

Progresión radiográfica en pacientes con Espondiloartritis Axial en tratamiento con anti-TNF

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Análisis de datos y técnicas de clustering

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FICHA DEL TRABAJO FINAL

Título del trabajo:	<i>Progresión radiográfica en pacientes con Espondiloartritis Axial en tratamiento con anti-TNF</i>
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Palabras clave	<i>Espondiloartritis axial (EspAx), anti-TNF, progresión radiográfica</i>
<p>Resumen del Trabajo (máximo 250 palabras): <i>Con la finalidad, contexto de aplicación, metodología, resultados i conclusiones del trabajo.</i></p>	
<p>Contexto: La eficacia clínica de los anti-TNF en la espondiloartritis axial (EspAax) ha sido ampliamente demostrada en ensayos clínicos aleatorizados. Algunos estudios observacionales sugieren que el tratamiento a largo plazo con anti-TNF (más de 4 años) podrían ralentizar la progresión radiográfica en EspAax. Existen datos que sugieren que el efecto del anti-TNF en la progresión radiográfica puede estar mediado por la reducción de la actividad de la enfermedad.</p> <p>Objetivo: Evaluar la relación de la actividad de la enfermedad con la progresión radiográfica en pacientes con EspAax en tratamiento con anti-TNF.</p> <p>Metodología: Un total de 101 paciente de la base de datos del registro español de pacientes con Espondiloartritis en terapia biológica (REGISPONSERBIO) han sido incluidos en el análisis. Para valorar la relación de la actividad de la enfermedad (medida por ASDAS) con la progresión radiográfica se ha realizado un modelo de regresión lineal múltiple. Tomando como variable independiente de progresión radiográfica, la diferencia entre el mSASSS de seguimiento con el inicial. Los predictores del modelo se ajustan según el método AIC.</p> <p>Resultados: El modelo de regresión lineal múltiple resultante es capaz de explicar el 61.8% de la variabilidad observada en la progresión radiográfica (R²: 0.6891, R²-Adjusted: 0.6182). El test F muestra que es significativo (p valor de 3.046e-10). La beta de la variable diferencia de ASDAS es de 0.893 (p= 0.011) con un IC al 95% de 0.225-1.561.</p>	

Conclusiones: Por cada punto de diferencia de ASDAS se puede esperar que la progresión radiográfica aumente un promedio de 0.89. Es decir, a menor actividad clínica medida por ASDAS menor progresión radiográfica medida por el mSASSS.

Abstract (in English, 250 words or less):

Background: Clinical efficacy of TNF inhibitors (TNFi) in axial spondyloarthritis (axSpA) has been widely probed in randomized control trials. In clinical practice, some studies suggested that long-term treatment with TNFi (more than 4 years) could slow down radiographic progression in axSpA. Furthermore, it is thought that the effect of TNFi on radiographic progression may be mediated by reducing disease activity.

Objective: To evaluate the relationship between disease activity and radiographic progression in axSpA patients treated with TNFi

Methods: 101 patients with axSpA, from the Spanish Register of Biological Therapy in Spondyloarthritides (REGISPONSERBIO) were included in the analysis. To evaluate the relationship between disease activity (measured by ASDAS) with radiographic progression we have performed a multiple linear regression. The independent variable of radiographic progression used was the difference between the follow-up mSASSS with the initial one. The model predictors are adjusted according to the AIC method.

Results: The multiple lineal regression model can explain the 61.8% of the observed variability in radiographic progression (R^2 : 0.6891, R^2 -Adjusted: 0.6182). The F test shows that it is significant (p value of $3.046e-10$). The beta of the ASDAS difference variable is 0.893 ($p= 0.011$) with a 95% CI of 0.225-1.561.

Conclusions: For each point of difference in ASDAS, the radiographic progression can be expected to increase by an average of 0.89. Meaning that the lower the clinical activity measured by ASDAS, the lower the radiographic progression measured by the mSASSS.

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1. Introducción

1.1 Contexto y justificación del Trabajo

La Espondiloartritis axial (EspAax) es una enfermedad inflamatoria crónica con síntomas predominantemente axiales, cursa con inflamación predominantemente en las articulaciones sacroilíacas y en la columna (1). La enfermedad se asocia en un 90% de los casos con el gen HLA-B27. La prevalencia estimada entre el 0.32-1.4% (2, 3), se correlaciona con el gen HLA-B27. Los pacientes con EspAax pueden presentar otros síntomas como artritis periférica (mono o oligoartritis de los miembros inferiores), entesitis y dactilitis, así como manifestaciones extra musculoesqueléticas como la psoriasis, la uveítis y la enfermedad inflamatoria intestinal. Característicamente la enfermedad se manifiesta con dolor y rigidez, y por el desarrollo de nueva formación ósea que conlleva la anquilosis del esqueleto axial (1).

La EspAax, puede ser clasificada como EspAax-radiológica (EspAax-r) o no EspAax-no radiológica (EspAax-nr), dependiendo de la presencia o ausencia de sacroiliitis radiográfica según los criterios modificados de Nueva York, Figura 1 (4). La EspAax-nr se considera como la forma inicial o más leve de la EspAax y los pacientes no siempre desarrollan daño estructural óseo en el esqueleto axial.

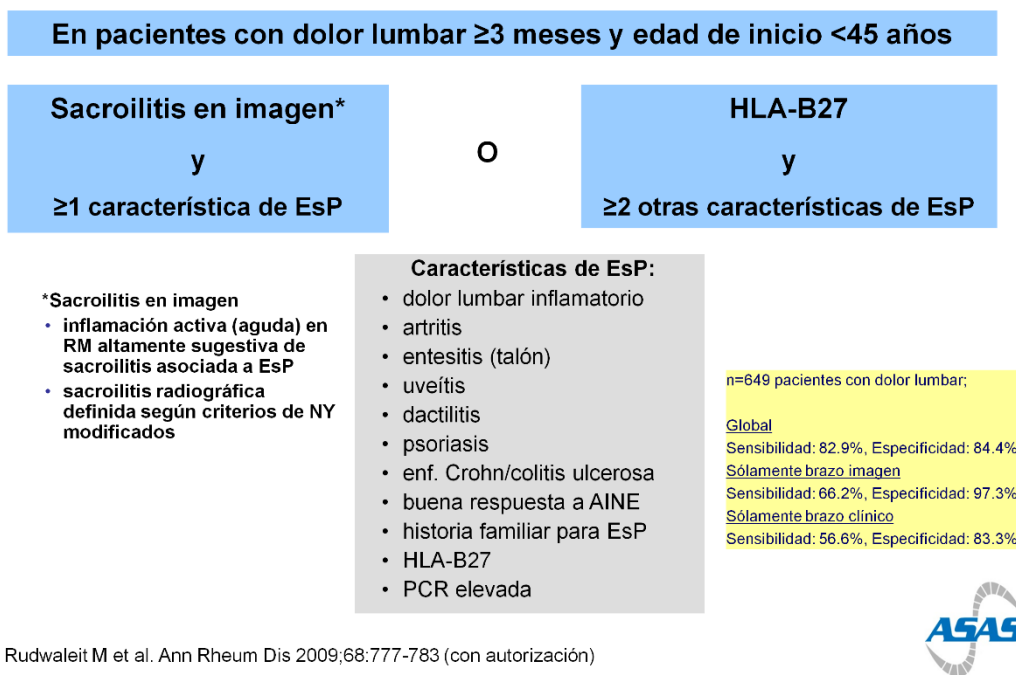
Figura 1. Criterios modificados de Nueva York para la valoración de sacroiliitis radiográfica.

Valoración Radiológica de Sacroiliitis (1966)

- **Grado 0** normal
- **Grado 1** cambios sospechosos
- **Grado 2** mínimos cambios – áreas pequeñas localizadas con erosión o esclerosis, sin alteración de la amplitud articular
- **Grado 3** alteraciones inequívocas – sacroiliitis moderada o avanzada con uno o más de los siguientes: erosiones, esclerosis, ensanchamiento, estrechamiento o anquilosis parcial
- **Grado 4** alteraciones graves – anquilosis total

Los criterios de clasificación ASAS (Assesment in Spondylo Arthriits international Society) (5) se utilizan en la práctica clínica habitual como a criterios diagnósticos de EspAax (Figura 2). Su diagnóstico requiere de la presencia de dolor crónico lumbar, de más de 3 meses de duración, antes de los 45 años. También, requiere de la presencia de sacroilitis, determinada por radiografía o por resonancia magnética aparte de como mínimo una característica clínica típica de la EspAax, o bien de la presencia de HLA-B27 positivo y de dos características clínicas típicas. La sensibilidad de los criterios en los diferentes estudios difiere entre el 67 y el 87% y la especificidad entre el 62 y el 95% (6-8).

Figura 2. Criterios de clasificación ASAS para Espondiloartritis axial.

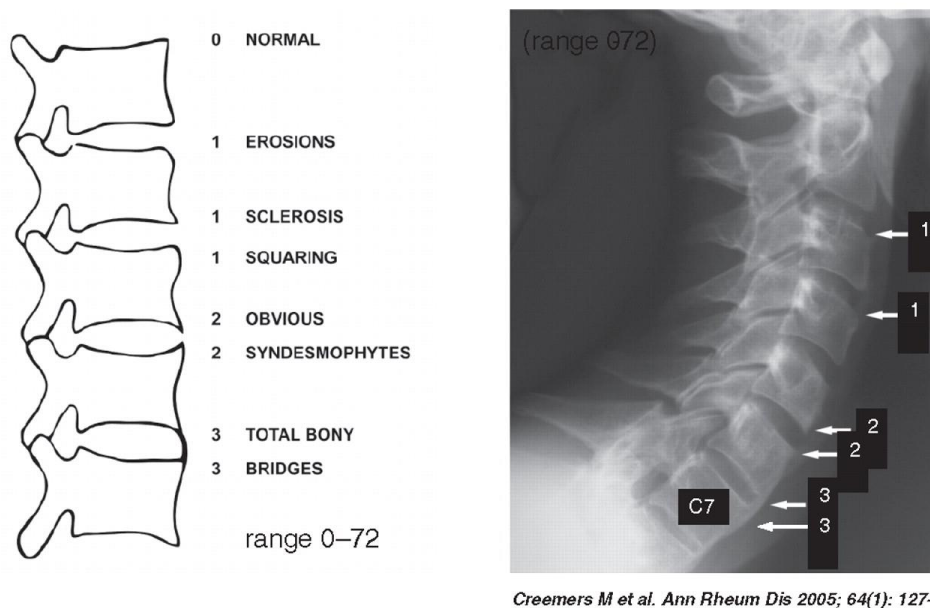


Una de las mayores preocupaciones en el campo de la espondiloartritis es el desarrollo de una nueva formación ósea (sindesmofitos) en la columna vertebral debido a su contribución en la gravedad de la enfermedad. Como también es relevante en determinar si los tratamientos son eficaces para inhibir esta progresión estructural. Es por ello, que resulta de vital importancia la evaluación del daño estructural en los pacientes con EspAax. El método de imagen de elección para la evaluación de la progresión estructural son las radiografías de columna (cervical y lumbar) y el índice numérico más utilizado es el mSASSS (modified Stoke Ankylosing Spondylitis Spine

Score). El mSASSS (Figura 3) captura las esquinas vertebrales anteriores de la columna cervical y lumbar en proyección lateral, evalúa la presencia de erosiones, esclerosis, cuadratura y sindesmofitos y puentes óseos con una puntuación total de 72 (9).

Figura 3. Escala mSASSS (modified Stoke Ankylosing Spondylitis Spine Score), escala para evaluar el daño estructural en pacientes con Espondiloartritis.

Modified Stoke Ankylosing Spondylitis Spinal Score (mSASSS)



Para medir la actividad de la enfermedad, se dispone de la escala Bath Ankylosing Spondylitis Disease Activity Index (BASDAI) que fue descrito inicialmente en 1994 (10). Es un índice clínico que consiste en 6 preguntas referentes a fatiga, afectación axial, afectación articular, entesopatía y rigidez matutina (2 preguntas). Por otro lado, el Ankylosing Spondylitis Disease Activity Score (ASDAS) (11) es un índice de actividad más reciente, y es considerado actualmente como el índice de elección para medir la actividad de la enfermedad. Combina datos clínicos (cuestionarios reportados por los pacientes) y datos analíticos (la proteína C reactiva (PCR)).

Estudios previos han demostrado que la progresión estructural está asociada con la presencia de daño estructural basal (sindemosfitos), PCR elevada, fumar, diagnóstico de EspAax-radiológica y presencia de alta actividad medida por ASDAS (12). La asociación entre la presencia de daño estructural a nivel axial y la reducción de la movilidad espinal (medida por BASMI) es bien conocida en pacientes con EspAax.

La eficacia clínica del tratamiento con inhibidores del factor de necrosis tumoral alfa (anti-TNF) en la espondiloartritis axial (EspAax) ha sido ampliamente probada en ensayos clínicos, sin embargo, su eficacia en inhibir la progresión radiográfica en este grupo de pacientes permanece incierto. En la práctica clínica habitual, algunos estudios sugieren que el tratamiento a largo plazo con anti-TNF (más de 4 años) podría ralentizar la progresión radiográfica (13). Además, se cree que el efecto del anti-TNF en la progresión radiográfica puede estar mediado por la reducción de la actividad de la enfermedad (13).

Por todo ello, resulta de vital interés conocer los factores que influyen en la progresión radiográfica en pacientes en tratamiento con anti-TNF. Y particularmente, conocer la influencia de la baja actividad en el tiempo con la progresión radiográfica en dichos pacientes. En nuestro trabajo, gracias a la cohorte longitudinal REGISPONSERBIO de pacientes con Espondiloartritis axial en tratamiento con anti-TNF, nos proponemos dar respuesta a ello.

1.2 Objetivos del Trabajo

- Evaluar la relación de la actividad de la enfermedad con la progresión radiográfica en pacientes con EspAax en tratamiento con anti-TNF.
- Evaluar el efecto de la baja actividad clínica medida por ASDAS y/o BASDAI en la progresión radiográfica en pacientes con EspAax en tratamiento con anti-TNF.

- Valorar si la baja actividad mantenida en el tiempo es más determinante que el tratamiento a largo plazo con anti-TNF en disminuir la progresión radiográfica.

1.3 Enfoque y método seguido

La evaluación de la progresión radiográfica se realizará mediante el índice numérico mSASSS. Dos lectores han evaluado las radiografías de forma independiente y con un orden cronológico conocido. Como definiciones de progresión radiográfica, disponemos:

- diferencia entre el mSASSS de seguimiento y el inicial (mSASSS Progress)
- cambio en la puntuación total del mSASSS en ≥ 2 puntos
- desarrollo de nuevos sindesmofitos
- desarrollo de nuevos sindesmofitos o crecimiento de los sindesmofitos existentes.

Para nuestro estudio, utilizaremos la variable de progresión numérica mSASSS Progress.

Como variables de actividad se utilizarán las variables numéricas: ASDAS y BASDAI m y como escala de funcionalidad la variable numérica BASFI.

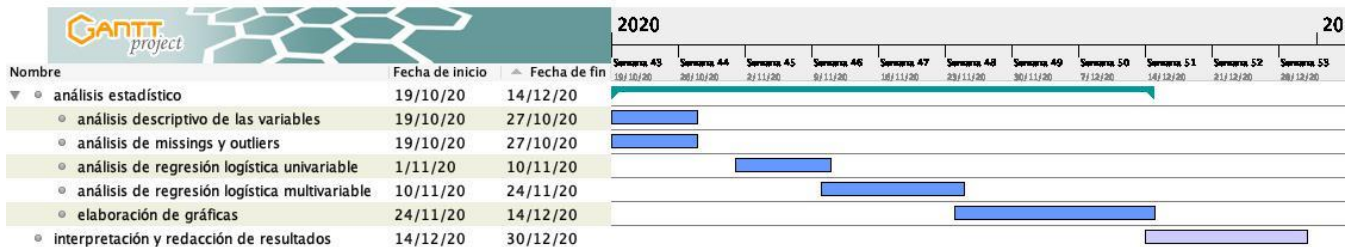
Primero se realizará un análisis descriptivo de la variable respuesta y de los factores/covariables, así como un análisis de missings y outliers. Para evaluar la relación de la actividad con la progresión radiográfica, una de las posibles estrategias a realizar es primero un análisis de regresión lineal multivariable.

1.4 Planificación del Trabajo

Para realizar el trabajo, dispongo de una base de datos clínicos obtenida en base a los datos de un registro multicéntrico español, REGISPONSERBIO, de pacientes con EspAax

en tratamiento con anti-TNF, concedida por el grupo de investigación en el campo de Espondiloartritis GRESSER.

El análisis estadístico se realizará con el programa estadístico R. A continuación, expongo una planificación temporal del proyecto con el diagrama de Gantt:



Análisis de riesgo, el factor tiempo es el principal limitante para la realización del proyecto, en el primer paso de análisis descriptivo de las variables resulta crucial para entender bien las cuales son las variables respuesta y las covariables y para entender bien el modelo estadístico a realizar.

1.5 Breve resumen de productos obtenidos

Con los resultados obtenidos en base a la interpretación de los resultados de las técnicas estadísticas realizadas y con la ayuda de tablas y gráficas, esperamos poder exponer de una forma clara la relación de la actividad de la enfermedad, medida por ASDAS, con la progresión radiográfica en pacientes con EspAax en tratamiento con anti-TNF.

1.6 Breve descripción de los otros capítulos de la memoria

Este Trabajo de Final de Máster (TFM) está estructurado de la siguiente forma. El Capítulo 2 y 3, explica extensamente el material y métodos del estudio. En el Capítulo 2, se realiza una descripción de la base de datos y de las variables utilizadas para llevar a

cabo el análisis, luego se realizará un análisis descriptivo de las variables incluidas. Por último, se realizará un análisis de los valores faltantes (missings). Posteriormente, en el Capítulo 3, se realiza primero un análisis de correlación de los predictores numéricos. Luego, se realizará un análisis de la variable respuesta, una validación de condiciones para la regresión lineal múltiple y luego un modelo de análisis de regresión múltiple lineal para cada variable de actividad (ASDAS y BASDAI). Por último, se realizará el ajuste de predictores del modelo de regresión lineal múltiple. En el Capítulo 4, se expondrán las conclusiones.

2. Material y métodos

2.1. Datos

Los datos con los que trabajaremos se han obtenido del registro español de pacientes con Espondiloartritis en terapia biológica (REGISPONSERBIO). Se trata de un registro observacional, longitudinal de pacientes con Espondiloartritis axial en tratamiento biológico con anti-TNF. Han participado un total de 20 centros de todo el territorio español. De los 200 pacientes incluidos en el registro, un total de 101 pacientes han sido incluidos en este estudio, en base a la disponibilidad de radiografías en el momento de inclusión del estudio y al seguimiento. La base de datos consta 66 variables, distribuidas entre variables clínicas y variables radiológicas. El registro, consta de paciente ya en tratamiento biológico previo al inicio del estudio (75%) y de pacientes que inician tratamiento biológico (25%). Al inicio del estudio, se recogieron datos demográficos, datos clínicos y radiografías. Las radiografías de seguimiento se obtuvieron en un intervalo medio de 3.45 años (± 0.98 sd). La lectura radiográfica, se realizó mediante la escala mSASSS, y como progresión radiográfica se utilizaron 4 definiciones: diferencia entre el MSASSS de seguimiento con el inicial, cambio en la puntuación total del mSASSS en ≥ 2 puntos, desarrollo de nuevos sindesmofitos y desarrollo de nuevos sindesmofitos o crecimiento de los sindesmofitos existentes. Los datos de actividad de la enfermedad (variables ASDAI, BASDAI y PCR) y de funcionalidad (variable BASFI) se han obtenido al inicio del tratamiento biológico, al inicio del estudio y luego cada 3 meses hasta un seguimiento total de 3 años.

He elegido la siguiente base de datos, porque es el único registro a nivel nacional dónde se disponen datos de actividad clínica y radiológicas de forma longitudinal con un seguimiento de 3 años. Por lo que resulta de vital interés estudiar la relación de la actividad clínica mantenida con la progresión radiográfica.

Tabla 1. Descripción de las variables al inicio del estudio, de la base de datos REGISPONSERBIO, registro español de pacientes con Espondiloartritis en terapia biológica.

Descripción de las variables	Tipo de variables	Cathegories	Median(IQR) N(%)
Edad:	Numérica		46.00 (16.00)
Sexo	Factor	Hombre	82 (81.2%)
		Mujer	19 (18.8%)
imc:	Numérica		26.02 (4.63)
imc.30: IMC > 30	Factor	No	84 (88.4%)
		Si	11 (11.6%)
tsint: Tiempo de síntomas de la enfermedad al inicio del estudio	Numérica		15.00 (18.00)
hlab27:	Factor	Negativo	13 (13.1%)
		Positivo	86 (86.9%)
ea: Espondiloartritis radiográfica	Factor	Negativo	15 (14.9%)
		Positivo	86 (85.1%)
tabaco.kat: tabaco	Factor	No	70 (69.3%)
		Si	31 (30.7%)
vsg: VSG al inicio del estudio	Numérica		19.00 (29.75)
basmi: BASMI al inicio del estudio	Numérica		2.87 (2.44)
uveitis: al inicio del estudio	Factor	No	79 (78.2%)
		Si	22 (21.8%)
psoriasis: PSORIASIS al inicio del estudio	Factor	No	92 (92.9%)
		Si	7 (7.1%)
enf.inf: ENF_INF al inicio del estudio	Factor	No	92 (92.9%)
		Si	7 (7.1%)
ttobio: Tipo de tto biológico al inicio del estudio	Factor	ETA	31 (30.7%)
		ADA	44 (43.6%)
		IFX	11 (10.9%)
		GOL	12 (11.9%)
		CTZ	3 (3.0%)
num.tb.prev: Número de ttos biológicos previos en el momento del inicio del estudio	Factor	No ttos previos	75 (74.3%)
		1 tto previo	20 (19.8%)
		2 ttos previos	6 (5.9%)
motivo: Motivo de cambio de tratamiento	Factor	Ineficacia	18 (72.0%)
		Efectos secundarios	7 (28.0%)
tiembio: Tiempo en tto biológico del estudio (meses)	Numérica		37.50 (70.00)
bio.inici: Tto biológico al inicio del estudio	Factor	No	25 (25.0%)
		Si	75 (75.0%)
biol.4.kat: Presencia de tto biológico durante más de 4 años al inicio del estudio	Factor	No	52 (51.5%)
		Si	49 (48.5%)

interval: Intervalo de tiempo entre RXs (entre el inicio y el seguimiento)	Factor	2 años	17 (17.0%)
		3 años	38 (38.0%)
		4 años	29 (29.0%)
		5 años	15 (15.0%)
		6 años	1 (1.0%)
aine: Tto con AINEs al inicio del estudio	Factor	No	42 (42.4%)
		Si	57 (57.6%)
fame: Tto con FAME al inicio del estudio	Factor	No	81 (81.8%)
		Si	18 (18.2%)
i.bas dai: BASDAI al inicio del tratamiento biológico	Numérica		5.40 (2.17)
bas dai.0: BASDAI al inicio del estudio	Numérica		3.20 (3.10)
bas dai.1: BASDAI a los 6 meses	Numérica		2.70 (3.00)
bas dai.2: BASDAI a los 12 meses	Numérica		2.50 (2.80)
bas dai.3: BASDAI a los 18 meses	Numérica		3.00 (3.60)
bas dai.4: BASDAI a los 24 meses	Numérica		2.80 (3.40)
bas dai.5: BASDAI a los 30 meses	Numérica		2.55 (2.92)
bas dai.6: BASDAI a los 36 meses	Numérica		2.80 (3.50)
as das.0: ASDAS al inicio del estudio	Numérica		2.06 (1.54)
as das.1: ASDAS a los 6 meses	Numérica		1.57 (1.26)
as das.2: ASDAS a los 12 meses	Numérica		1.83 (1.23)
as das.3: ASDAS a los 18 meses	Numérica		1.77 (1.31)
as das.4: ASDAS a los 24 meses	Numérica		1.67 (1.27)
as das.5: ASDAS a los 30 meses	Numérica		1.64 (1.13)
as das.6: ASDAS a los 36 meses	Numérica		1.68 (1.25)
pcr.0: PCR al inicio del estudio	Numérica		3.35 (8.10)
pcr.1: PCR a los 6 meses	Numérica		2.00 (3.68)
pcr.2: PCR a los 12 meses	Numérica		2.00 (4.70)
pcr.3: PCR a los 18 meses	Numérica		1.90 (4.85)
pcr.4: PCR a los 24 meses	Numérica		2.05 (4.68)
pcr.5: PCR a los 30 meses	Numérica		1.40 (5.38)
pcr.6: PCR a los 36 meses	Numérica		2.00 (5.35)
egp: Dolor global descrito por el paciente en el inicio del estudio	Numérica		3.00 (4.00)
dan: Dolor axial nocturno	Numérica		3.00 (5.00)
egm: Dolor global del paciente en la opinión del médico	Numérica		3.00 (3.50)
bas fi.0: BASFI al inicio del estudio	Numérica		3.85 (4.45)
bas fi.1: BASFI a los 6 meses	Numérica		2.70 (3.35)
bas fi.2: BASFI a los 12 meses	Numérica		2.30 (3.70)
bas fi.3: BASFI a los 18 meses	Numérica		2.80 (3.50)
bas fi.4: BASFI a los 24 meses	Numérica		3.50 (3.70)
bas fi.5: BASFI a los 30 meses	Numérica		3.20 (3.60)
bas fi.6: BASFI a los 36 meses	Numérica		3.10 (4.00)
ms ass.mean.bl: media de la puntuación Ms ass al inicio	Numérica		5.00 (23.00)
ms ass.mean.fl: media de la puntuación Ms ass en el seguimiento	Numérica		8.00 (30.00)
ms ass.progress: Diferencia entre el mSASSS del inicio con el del seguimiento	Numérica		0.00 (3.00)
new.syn.ms ass: Nuevos sindesmofitos	Factor	No	81 (80.2%)
		Nuevos sindesmofitos	20 (19.8%)

msasss.progress.kat: Msasss increment of 2 points or more during follow-up	Factor	No	71 (70.3%)
		Incremento de dos o más puntos del mSASSS	30 (29.7%)
new.prog.syn: Nuevos o progresión de sindesmofitos existentes	Factor	No	79 (78.2%)
		Nuevos o progresión de sindesmofitos	22 (21.8%)
mind.2.synd.fl: presencia de sindesmofitos en el seguimiento	Factor	No	46 (45.5%)
		Presencia de sindesmofitos	55 (54.5%)
mind.2.synd.bl: presencia de sindesmofitos al inicio	Factor	No	50 (49.5%)
		Presencia de sindesmofitos	51 (50.5%)
pcr.iniciob: PCR al inicio del tratamiento biológico	Numérica		10.50 (22.40)
i.asdas: ASDAS al inicio del tratamiento biológico	Numérica		3.40 (1.22)
hvgp: valoración global de la enfermedad al inicio	Numérica		6.00 (2.00)

2.2. Preparación de los datos para el análisis

La base de datos original consta de 66 variables, a continuación, preparamos la base de datos para el análisis. Definimos como variable respuesta la variable de progresión radiográfica, numérica Msasss Progress, que se define como la diferencia de la variable del mSASSS al inicio con la de seguimiento (Msasss_BL y Msasss_FL). Eliminamos la primera variable (el número de paciente), y las variables (Msasss_BL y Msasss_FL) puesto que forman parte de la variable y (respuesta).

Además, la base de datos original consta de diferentes mediciones de las variables BASDAI, ASDAS, PCR y BASFI medidas cada 6 meses en un seguimiento total de 3 años. Para poder realizar el análisis de regresión lineal, trataremos las covariables X de las diferentes mediciones a lo largo del tiempo del mismo modo que lo hemos hecho para la variable Y, es decir, como un cálculo de diferencia entre la variable inicial con la última

medición de seguimiento. Por lo tanto, en la base de datos resultante, transformamos las variables de seguimiento en una única variable. Las variables resultantes son: BASDAIDif, ASDASDif, PCRdif y BASFIDif.

2.3. Análisis descriptivo de las variables

Tras los ajustes realizados, obtenemos una base de datos con 39 variables y 101 observaciones (base_rx2). Una descripción gráfica de algunas de las variables de la base datos está representado en las Figuras 4 y 5.

Para el posterior análisis de regresión lineal, creamos también una base datos ajustada con los valores mencionados anteriormente, pero únicamente con variables numéricas (base_rx.num2). Los descriptivos de las variables se encuentran representados en la Tabla 2.

Figura 4. Boxplot y gráfico de densidad de las variables numéricas EDAD e IMC.

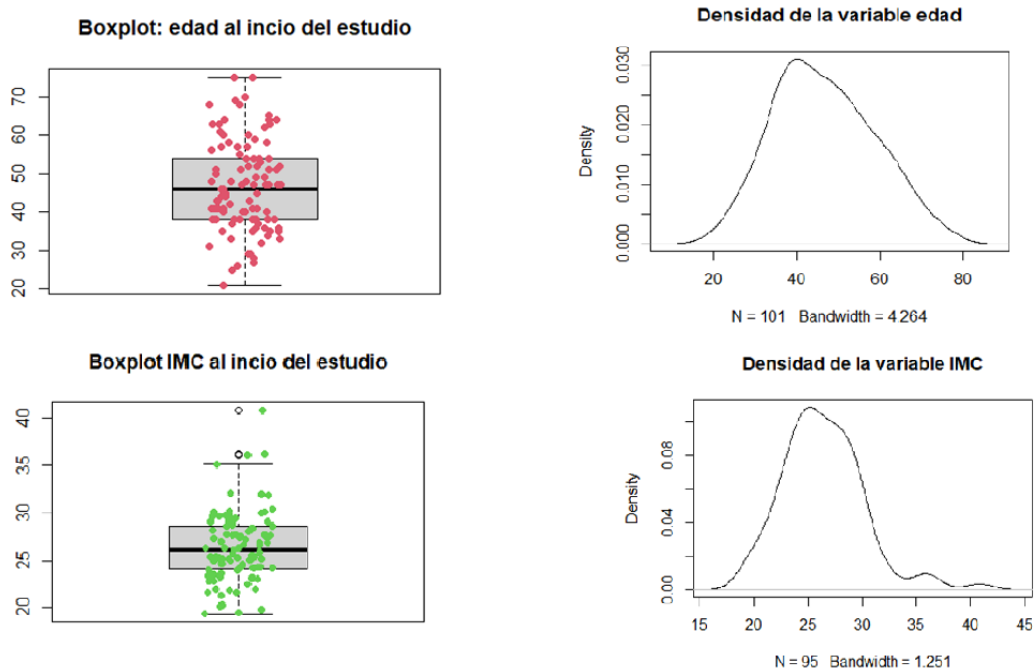


Figura 5. Histogramas de las variables tipo factor, HLA-B27, SEXO, IMC>30 y tipo de tratamiento biológico recibido.

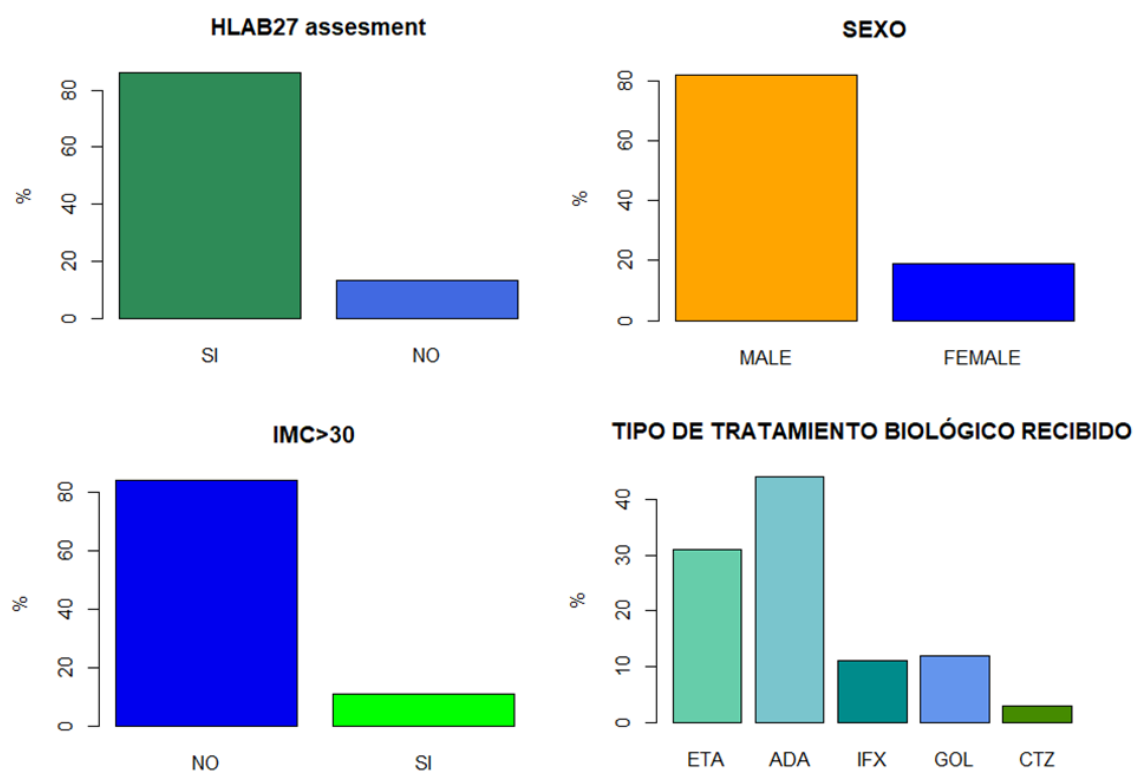


Tabla 2. Análisis descriptivo de las variables numéricas.

Variable	mean	Median	sd	min	max	var
ASDAS al inicio del tratamiento	3.4	3.5	0.8	1.5	5.3	0.7
BASDAI al inicio del tratamiento	5.3	5.4	1.8	0.7	8.8	3.3
BASMI	3.1	2.7	1.7	0.5	7.3	3
Diferencia entre ASDAS al inicio y final	-0.5	-0.3	1.2	-3.5	2	1.5
Diferencia entre el BASDAI al inicio final	-0.7	-0.1	2.1	-7	5.2	4.5
Diferencia entre el BASFI al inicio y final	-0.7	-0.2	2.6	-8.9	6.2	6.7
Diferencia entre el mSASSS al inicio y final	2	0	3.8	-2.5	21.5	14.1
Diferencia entre PCR al inicio y final	-3.6	-0.2	13.2	-88.4	28	173.5

Dolor global del paciente	4	3	2.8	0	10	7.7
Dolor global en la opinión del médico	3.4	3	2.4	0	10	5.7
Edad	46.7	46	11.9	21	75	142.2
IMC	26.4	26	3.7	19.4	40.8	13.9
Intervalo entre radiografías	3.5	3	1	2	6	1
PCR al inicio del tratamiento	18.5	11	19.2	0	98	370
Tiempo con tratamiento	45.1	42	40.1	0	132	1611.7
Tiempo síntomas	17.9	15	12.8	0	52	163.9
Valoración global del paciente al inicio del tratamiento	5.9	6	2	1	10	3.9
VSG	26.1	18	22.4	0.5	115	502.7

2.4. Análisis de missings

De la base datos ajustada hay 147 missings y en la base de datos de variables únicamente numéricas hay 128. Con el análisis de missings, vemos que los missings se concentran principalmente en 5 variables numéricas (BASMI, BASFIDif, BASDAIDif, ASDASDif, PCRDif).

Debido a que los missings se centran en las variables numéricas, sobre las variables numéricas de la base de datos original aplicamos un método de imputación de missings (Multivariate Imputation by Chained Equations). A este nuevo data.frame creado, lo unimos con las variables tipo factor, y creamos una nueva base de datos ((c_base_rx2).

3. Modelo de regresión lineal múltiple

3.1. Análisis de correlación de los predictores numéricos

Para controlar la multicolinealidad del modelo, estudiamos la correlación de los predictores numéricos. Para ellos creamos una matriz de correlaciones, las variables que están más correlacionadas (coeficiente de correlación >0.5), están representadas en la Tabla 3. Para construir el modelo de regresión, no incluiremos las variables numéricas que tienen una correlación superior a 0.7 para evitar la multicolinealidad.

Según apreciamos en la tabla, las variables EGM y EGP tienen una alta correlación 0.77. Hecho que es lógico, debido a que ambas variables expresan el grado del dolor del paciente. Por un lado, EGM, representa el grado del dolor del paciente en la opinión del médico y EGP, el grado de dolor reportado por el paciente. Por lo que, para construir el modelo, nos quedamos con la variable EGM, que en general refleja de forma más real el grado de dolor del paciente.

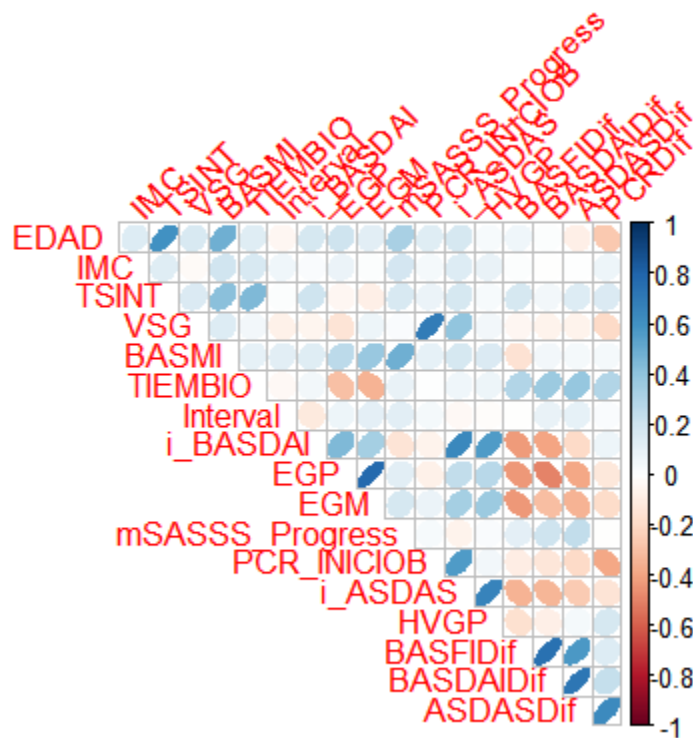
Otras variables que obtienen una correlación superior a 0,7, son las dos variables de actividad BASDAIDif y ASDASDif, correlación 0.728, y la variable de actividad BASDAIDif con la funcionalidad BASFIDif, correlación 0.747. Como la variable principal de interés para medir la actividad es el ASDAS, elegimos primero la variable ASDASDif para construir el modelo y la variable de funcionalidad, BASFIDif.

Tabla 3. Coeficientes de correlación superiores a 0.5 de las variables numéricas de la base de datos REGISPONSERBIO.

Table of correlation coefficients (0.5 or more)

Variable1	Variable2	Correlation Coefficient
EGM	EGP	0.7702752
BASDAIDif	BASFIDif	0.7468403
ASDASDif	BASDAIDif	0.7278965
PCR_INICIOB	VSG	0.6954562
HVGP	i_ASDas	0.6640062
i_ASDas	i_BASDAI	0.6418729
PCRDif	ASDASDif	0.6365525
TSINT	EDAD	0.6074403
ASDASDif	BASFIDif	0.5798891
i_ASDas	PCR_INICIOB	0.5682738
HVGP	i_BASDAI	0.5639104

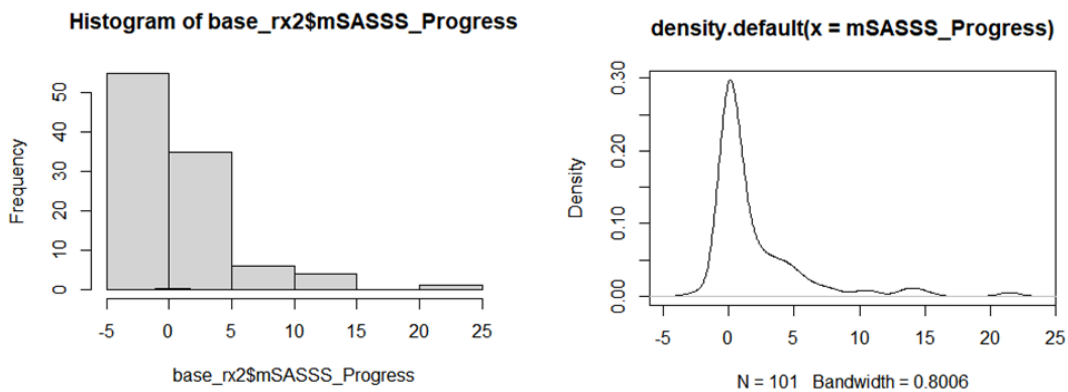
Figura 6. Correlation Plot de las variables numéricas de la base de datos REGISPONSERBIO.



3.2. Análisis de la variable respuesta

A continuación, estudiamos la variable respuesta numérica (mSASSS Progress) para ver si sigue una distribución normal. En la representación gráfica de la variable respuesta, se aprecia que la densidad es claramente asimétrica.

Figura 7. Histograma y diagrama de densidad de la variable respuesta, mSASSS Progress.

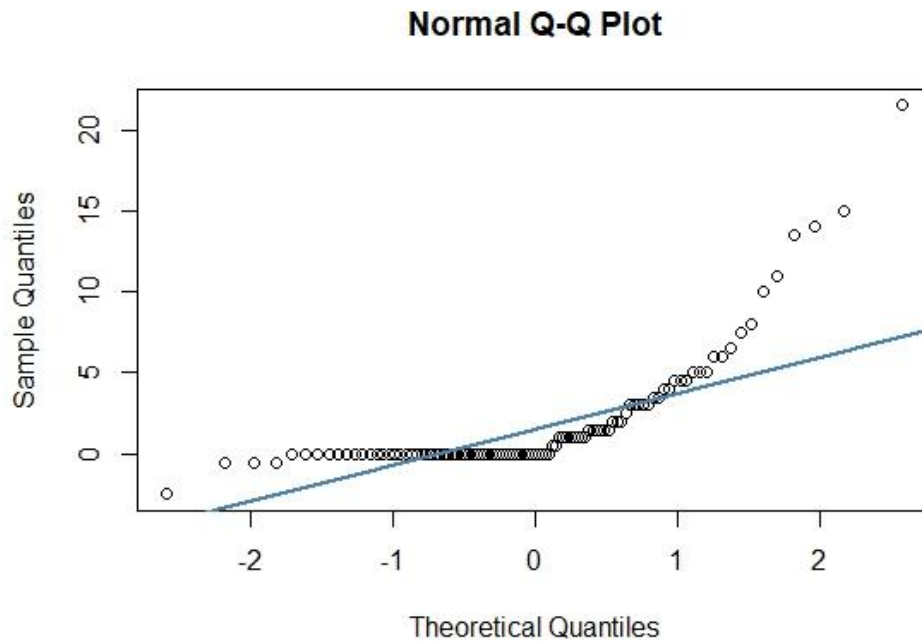


En el análisis descriptivo de la variable respuesta, la media y mediana no coincide (1.98 y 0, respectivamente), y que la kurtosis es superior a 3.

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	kurtosis	skew
-2.50	0.00	0.00	1.98	3.00	21.50	8.37	2.69

Con la gráfica qqplot, vemos que la variable respuesta no sigue una distribución normal (Figura 8). Hallazgos, que se confirman tanto en el Shapiro-Wilk's test como en el Kolmogorov-Smirnov (K-S) normality test. Pues al obtener un resultado significativo ($p = 2.092e-14$), debemos rechazar la hipótesis nula de igualdad. Como la variable respuesta no es normal, podemos intentar una transformación para mejorar esa situación, pero antes vamos a ver si los residuos siguen la suposición de normalidad, porque si es así no es necesario transformar la variable respuesta.

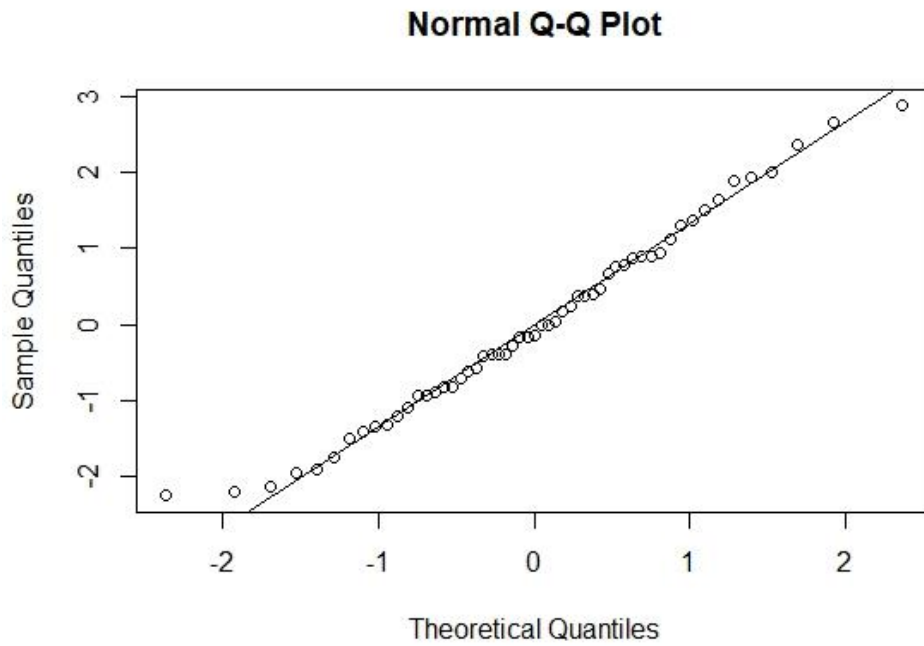
Figura 8. Q-Q Plot de la variable respuesta mSASSS Progress.



3.3. Validación de condiciones para la regresión múltiple lineal

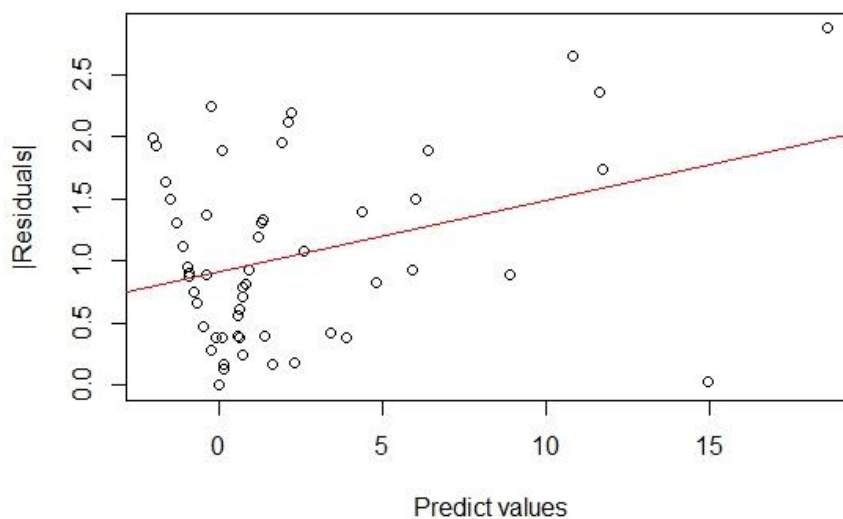
Primero realizamos un análisis de residuos para evaluar si cumplen la hipótesis de normalidad. Realizamos un gráfico Q-Q plot, donde se puede observar que los puntos se ajustan a la diagonal, lo que sugiere que los datos se ajustan a una distribución normal (Figura 9). Realizamos un test de normalidad (Shapiro-Wilks) para comprobar lo que hemos visto gráficamente y con un p-valor 0.58 permite rechazar la hipótesis nula, que nos dice que los residuos efectivamente siguen una distribución normal. Por lo tanto, al haber normalidad de los residuos, no hace falta realizar una transformación para la variable respuesta, mSASSS Progress.

Figura 9. Q-Q Plot de los residuos



Luego estudiamos la varianza de los residuos, que debe ser constante en todo el rango de observaciones. Para comprobarlo se representan los residuos gráficamente y se confirma los hallazgos con el Breusch-Pagan test. Con el test, obtenemos un resultado no significativo ($p= 0.3958$). Por lo que no hay evidencias de falta de homocedasticidad.

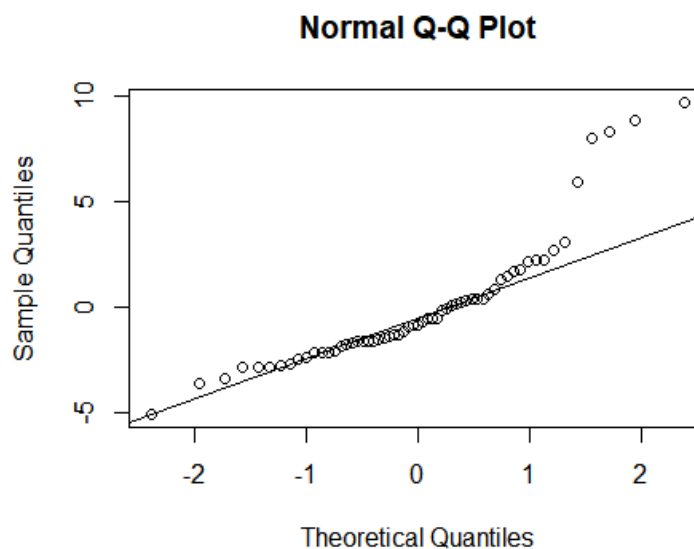
Figura 10. Representación gráfica de los residuos frente a los valores ajustados por el modelo.



Para estudiar la autocorrelación, realizamos el test D-W Statistic y al obtener también un resultado no significativo ($p=0.832$), no hay tampoco evidencia de autocorrelación.

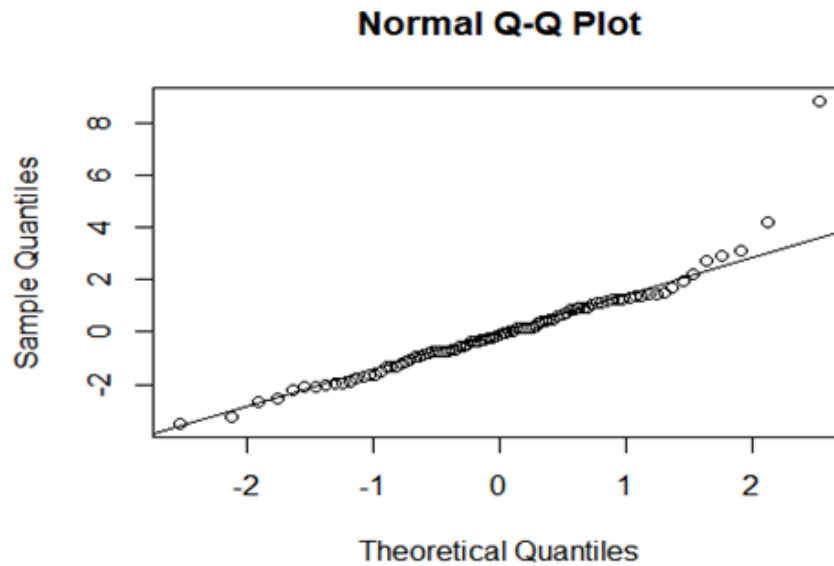
Por otro lado, cuando estudiamos las condiciones para el modelo con la base de datos únicamente numérica (base_rx.num2). En el análisis de residuos vemos que el Q-Q Plot los residuos se alejan de la normalidad, hallazgos que se confirman con el Saphiro test ($p= 1.724e-06$), por lo que los residuos no siguen la normalidad.

Figura 11. Q-Q Plot de los residuos de las variables numéricas.



Para la base de datos, con variables numéricas corregidas con imputación de missings (c_base_rx2). En el análisis de residuos vemos también que el Q-Q Plot los residuos se alejan de la normalidad, hallazgos que se confirman con el Saphiro test ($p= 5.607e-06$), por lo que los residuos no siguen la normalidad.

Figura 12. Q-Q Plot de los residuos con los missings corregidos.



3.4. Modelo de regresión múltiple lineal

Como hemos visto en el modelo, generado por la base de datos base_rx2, cumple la normalidad de los residuos, la homocedasticidad y no hay autocorrelación ni multicolinealidad. Para gestionar los missings, utilizamos la función `na.action = na.omit`, en el modelo resultante, 46 observaciones son eliminadas debido a los missings, quedando un total de 97 observaciones.

Con el modelo realizado, únicamente las variables EGM y la variable categórica mSASSS_Progress_Kat alcanzan la significación estadística.

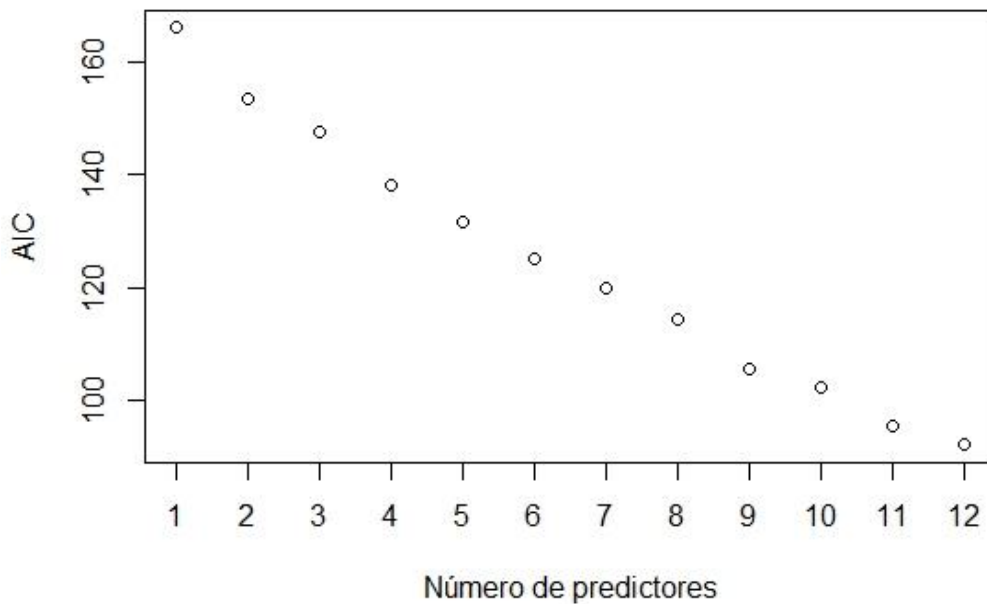
En conclusión, el modelo múltiple es capaz de explicar el 62.4% de la variabilidad observada en la progresión radiográfica (R^2 : 0.9197, R^2 -Adjusted: 0.6664). El test F muestra que es significativo ($p = 0.007572$). Se satisfacen todas las condiciones para este tipo de regresión múltiple.

3.5. Ajuste de predictores del modelo de regresión múltiple lineal

Por regla general, el número de observaciones debe ser como mínimo entre 10 y 20 veces el número de predictores del modelo. Como tenemos 97 observaciones, el número de predictores no debería ser mayor a 10. Por lo que debemos eliminar predictores, que no influyan en el modelo. Para ello, vamos a realizar a una regresión con variables seleccionadas por AIC.

Como apreciamos en la gráfica, el mínimo AIC se alcanza con 12 predictores, aunque con 10 también tenemos un buen resultado. Por lo que los predictores finales para el modelo son: EDAD, SEXO, HLAB27, EA, TTOBIO, EGM, New_Syn_mSASSS, mSASSS_Progress_Kat, New_Prog_Syn_Msasss y ASDASDif.

Figura 13. Selección de predictores por AIC.



Con este modelo las variables, presencia de nuevos sindesmofitos (New_Syn_mSASSS), la progresión radiográfica de > 2 puntos del mSASSS (mSASSS_Progress_Kat), y la diferencia entre el ASDAS del inicio con la variable final de seguimiento (ASDASDif), son las únicas variables significativas. La beta de la variable progresión radiográfica de > 2 puntos del mSASSS es de 4.820 ($p= 0.0003$) con un IC al 95% de 2.318-7.321. Y, la beta de la variable presencia de nuevos sindesmofitos es de 5.291 ($p= 0.012$) con un IC al 95% de 1.278-9.304. Por último, la beta de la variable diferencia de ASDAS es de 0.893 ($p= 0.011$) con un IC al 95% de 0.225-1.561

En conclusión, el modelo de regresión lineal múltiple resultante es capaz de explicar el 61.8% de la variabilidad observada en la progresión radiográfica ($R^2: 0.6891$, $R^2\text{-Adjusted}: 0.6182$). El test F muestra que es significativo (p valor de $3.046e-10$).

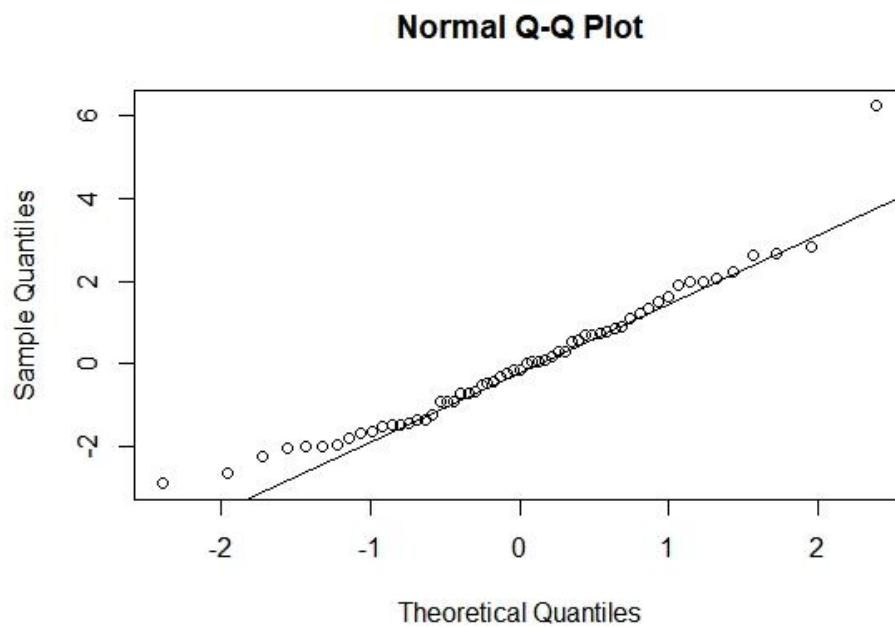
3.6. Modelo de regresión lineal múltiple con variable actividad BASDAI

A continuación, repetimos el mismo proceso anterior, pero tomando la variable BASDAIDif en vez de la variable ASDASDif como variable de actividad. Como hemos visto anteriormente ambas variables son utilizadas para medir la actividad clínica de la enfermedad. Debido a que presentan una alta correlación, para evitar la multicolinealidad del modelo, realizamos un modelo de regresión lineal múltiple para cada una de estas variables.

Como la variable respuesta es la misma (mSASSS_Progress) y ya hemos visto anteriormente que no sigue una distribución normal realizamos un análisis de residuos para ver si los residuos cumplen la hipótesis de normalidad. Realizamos primero un gráfico Q-Q plot, donde se puede observar que los puntos no se ajustan del todo a la diagonal, lo que sugiere que los datos no se ajustan a una distribución normal (Figura 14). Realizamos un test de normalidad (Shapiro-Wilks) para comprobar lo que hemos visto gráficamente y con un p -valor 0.0165 no se puede rechazar la hipótesis nula, que nos dice que los residuos efectivamente no siguen una distribución normal. Por lo tanto,

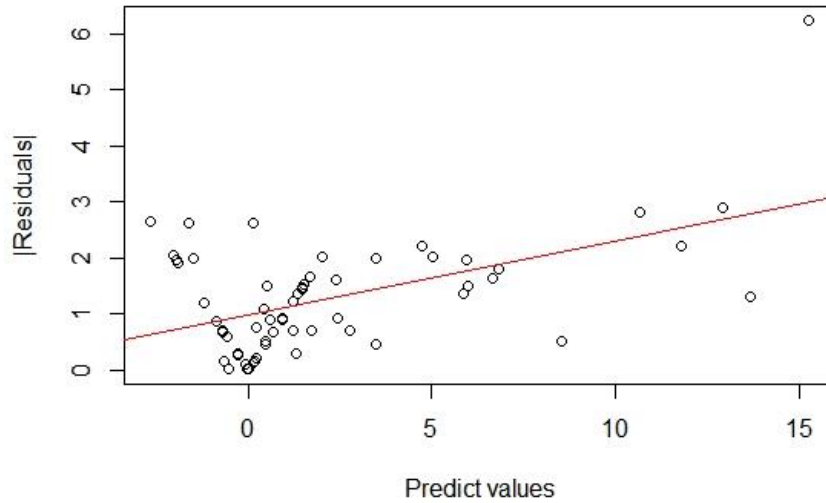
al no haber normalidad de los residuos, se debería realizar una transformación para la variable respuesta, mSASSS Progress.

Figura 14. Q-Q plot de los residuos con la variable BASDAI de actividad.



Estudiamos también la varianza de los residuos, que debe ser constante en todo el rango de observaciones. Para comprobarlo se representan los residuos gráficamente (Figura 15) y se confirma los hallazgos con el Breusch-Pagan test. Con el test, obtenemos un resultado no significativo ($p= 0.3916$). Por lo que no hay evidencias de falta de homocedasticidad.

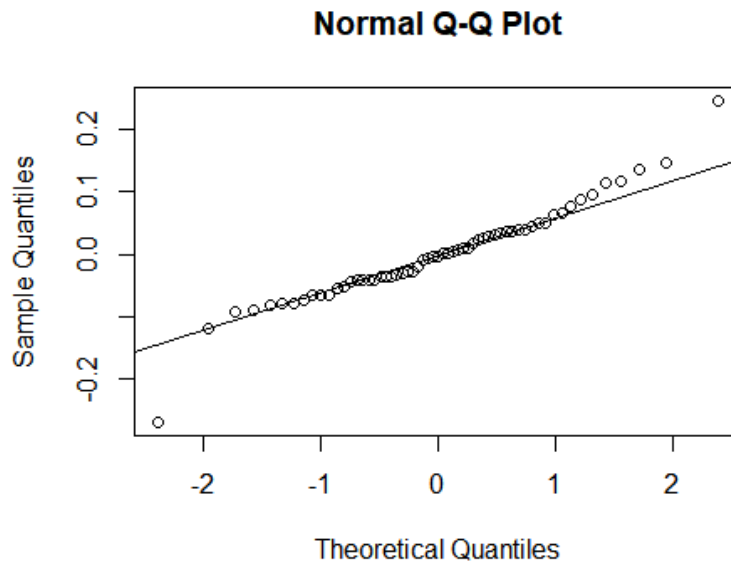
Figura 15. Representación gráfica de los residuos frente a los valores ajustados por el modelo.



Al tener valores negativos y 0 en la variable respuesta, se convierten en NaN cuando aplicamos una transformación logarítmica. Por lo tanto, debemos transformar los valores negativos y 0. Para ello, una técnica común para el manejo de valores negativos es agregar un valor constante a los datos antes de aplicar la transformación logarítmica. Como, por ejemplo: $LY = \log_{10}(Y + 1 - \min(Y))$.

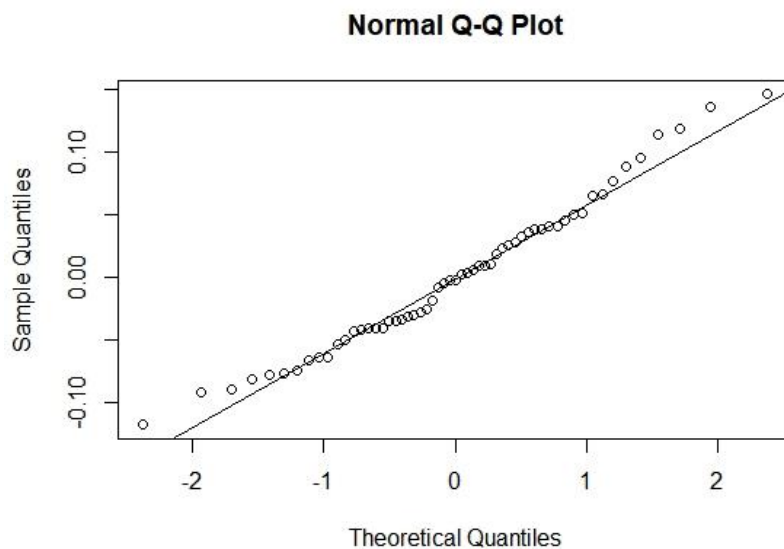
Con la variable respuesta transformada según lo indicado, realizamos ahora un análisis de residuos. Primero realizamos un gráfico Q-Q plot, por la disposición de los puntos impresiona que sigue sin satisfacerse la condición de normalidad posiblemente debido a dos datos atípicos (Figura 16). Realizamos un test de normalidad (Shapiro-Wilks) para comprobar lo que hemos visto gráficamente y con un p-valor 0.01936 no permite rechazar la hipótesis nula, lo que nos dice efectivamente que los residuos no siguen una distribución normal.

Figura 16. Q-plot de los residuos, con la variable respuesta transformada.



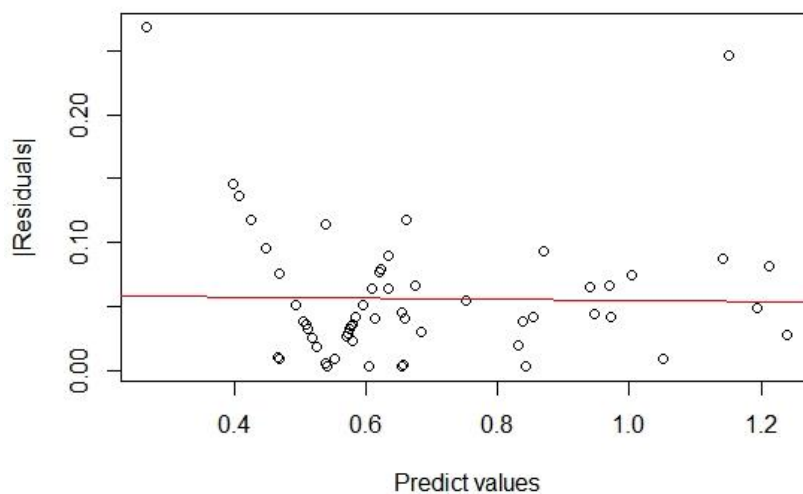
Con la función, `which.max(abc1$residuals)` y `which.min(abc1$residuals)`, vemos principalmente que hay dos datos atípicos en los residuos, uno por la parte superior de la gráfica (max) y otro en la parte inferior (min), el 24 y el 41. Realizamos de nuevo el gráfico q-q plot (figura 17) y el saphiro test ($p= 0.3819$) sin los dos valores extremos, y se confirma que los residuos sí se distribuyen de forma normal a excepción de dos datos extremos. Es necesario estudiar en detalle la influencia de esta observación para determinar si el modelo es más preciso sin ella.

Figura 17. Q-Q plot de los residus sin los dos datos extremos.



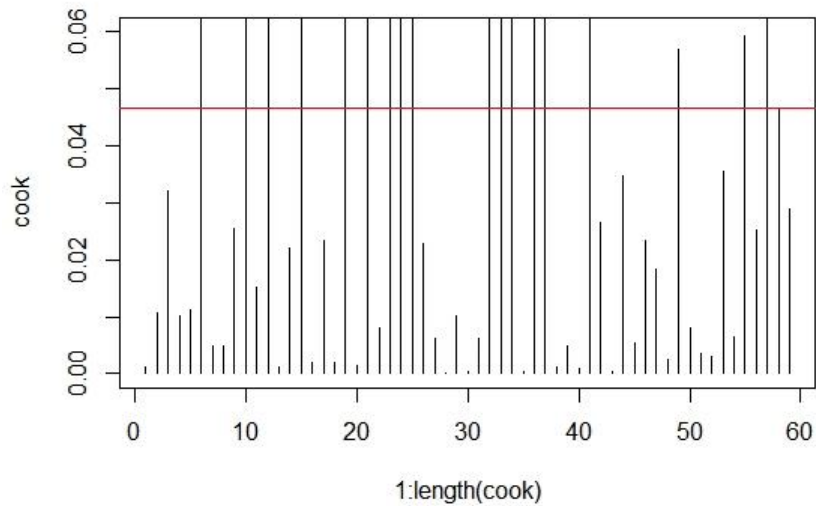
Estudiamos también la varianza de los residuos, que debe ser constante en todo el rango de observaciones. Para comprobarlo se representan los residuos gráficamente (Figura 18) y se confirma los hallazgos con el Breusch-Pagan test. Con el test, obtenemos un resultado no significativo ($p= 0.2969$). Por lo que no hay evidencias de falta de homocedasticidad. Ni tampoco de autocorrelación, $p= 0.688$, con el test D-W Statistic.

Figura 18. Representación gráfica de los residuos frente a los valores ajustados por el modelo, con la variable respuesta ajustada.



A continuación, estudiamos las observaciones influyentes. Calculamos la distancia de Cook como medida de la influencia de los puntos y la representamos contra los cuartiles de una distribución seminormal, figura 19, donde se aprecian varios valores influyentes (valores que sobrepasan al línea horizontal).

Figura 19. Gráfico de la distancia de Cook para detectar observaciones que influyen fuertemente en los valores ajustados del modelo.



Buscamos las observaciones con un mayor residuo studentizado externamente (*jackknife residual*).

42	66	7	51	61	36
4.813376	4.638805	2.399791	1.816236	1.745187	1.496918

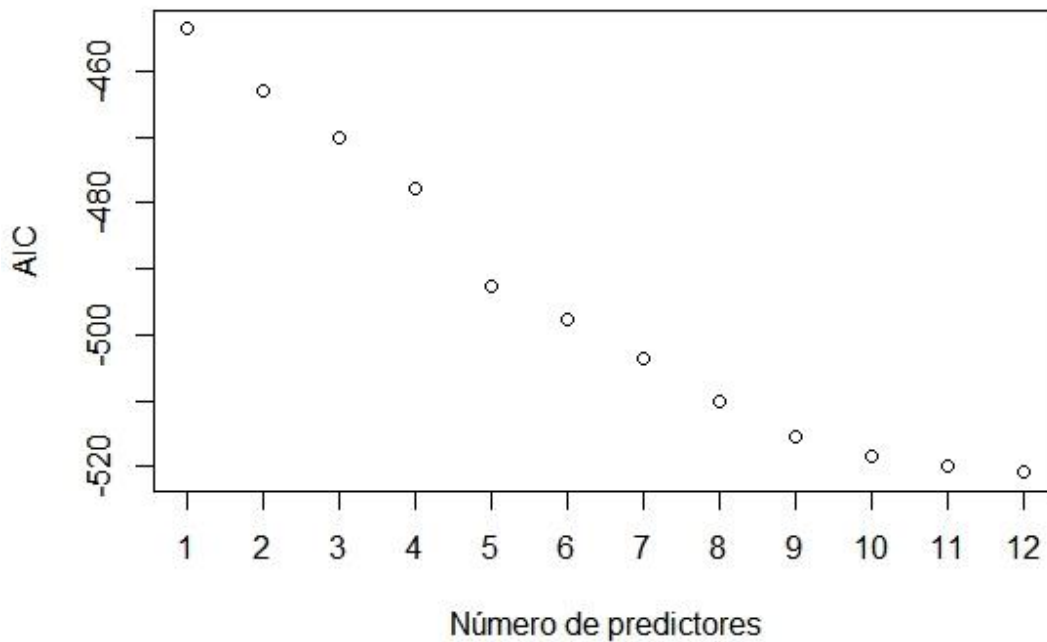
En cuanto al leverage:

1	15	63	57	51	14
0.9211843	0.9211843	0.9211843	0.8551536	0.8541275	0.8480441

En resumen, el punto 51 es un valor influyente con alto leverage y outlier. En cambio, el punto 1 es de alta influencia, tiene un alto leverage (está lejos del centro de los datos), pero no es outlier.

Por último, realizamos el modelo de regresión con variables seleccionadas por AIC. Como apreciamos en la gráfica (figura 20), el mínimo AIC se alcanza con 12 predictores, aunque con 10 también tenemos un buen resultado. Por lo que los predictores finales para el modelo son: EDAD, EA, TTOBIO, NUM_TB_PREV, MOTIVO, BIO_INICI, Mind_2_Synd_FL, mSASSS_Progress_Kat y BASDAIDif.

Figura 20. Selección de predictores por AIC.



Con este modelo las variables, EDAD, Espondilitis Anquilosante, Tratamiento biológico con Adalimumab y certolizumab (TTBIOADA y TTOBIOCTZ), presencia de tratamiento biológico al inicio, la progresión radiográfica de > 2 puntos del mSASSS (mSASSS_Progress_Kat), y la diferencia entre el BASDAI del inicio con la variable final de seguimiento (BASDAIDif), son variables significativas. La beta de la variable diferencia de BASDAI es de 0.000122 ($p= 0.011$) con un IC al 95% de 0.0145-0.0414. Recordemos que la variable respuesta esta transformada, por lo que tenemos que corregir el coeficiente de la siguiente forma: $(10\exp(0.000122)-1)*100= 0.028$. Por lo que por cada incremento en la diferencia de la progresión radiográfica, la diferencia del BASDAI aumenta en un 0.028%.

En conclusión, el modelo de regresión lineal múltiple resultante es capaz de explicar el 76.8% de la variabilidad observada en la progresión radiográfica ($R^2: 0.808$, $R^2\text{-Adjusted}: 0.7679$). El test F muestra que es significativo (p valor de $< 2.2e-16$).

4. Conclusiones

En el presente trabajo evaluamos la asociación de la actividad de la enfermedad con la progresión radiográfica en pacientes con Espondiloartritis axial tratados con anti-TNF de un registro Español (REGISPONSERBIO). Para poder evaluar dicha asociación, se ha realizado un modelo de regresión lineal múltiple, tomando como variable respuesta la progresión radiográfica. Se ha utilizado como definición de progresión radiográfica, la variable numérica, diferencia entre el mSASSS de seguimiento con el inicial. Para medir la actividad de la enfermedad, se ha utilizado el índice de elección para medir la actividad clínica de la enfermedad, el ASDAS. Disponemos de medidas de ASDAS y BASDAI (otra forma de medir la actividad de la enfermedad) al inicio del estudio y cada 6 meses hasta un seguimiento total de 3 años. Para incluir las variables de actividad en el modelo, se ha utilizado la diferencia entre el ASDAS y BASDAI del seguimiento (3er año) con la del inicio.

La duración del tratamiento biológico es un factor importante para evaluar su efecto en la progresión radiográfica en pacientes con EspAax. Según lo reportado en estudios previos, existe una asociación entre la duración del tratamiento con anti-TNF y el riesgo de progresión, sin que se observe ningún efecto del anti-TNF en un intervalo de seguimiento de 2 años. Además, se cree que el posible efecto a largo plazo del anti-TNF en la nueva formación ósea se debe a una supresión efectiva de la inflamación, ya que nuevas pruebas apoyan que los cambios inflamatorios van seguidos de la sustitución de la médula ósea subcondral por la reparación del tejido que luego estimula los osteoblastos, lo que da lugar a la nueva formación ósea.

Con los resultados obtenidos, podemos concluir que por cada punto de diferencia de ASDAS se puede esperar que la progresión radiográfica aumente un promedio de 0.89. Es decir, a menor actividad clínica medida por ASDAS menor progresión radiográfica medida por el mSASSS. Por otro lado, apreciamos sin embargo que la duración del

tratamiento biológico previo (tiembio), no resulta significativa en el modelo, lo que hace suponer que es más importante el control de la enfermedad (disminución del ASDAS) que el tiempo previo en tratamiento biológico.

Con los resultados obtenidos, apreciamos también que el índice de actividad ASDAS presenta una correlación muy alta con el BASDAI (coeficiente de correlación 0.727). Cuando evaluamos el modelo, utilizando la variable BASDAI en vez del ASDAS, como variable de actividad de la enfermedad, apreciamos que hay significación estadística y por lo tanto asociación con la progresión radiográfica, pero con menor beta y por tanto menor magnitud de impacto. Lo que confirma que el ASDAS es el mejor método para medir la actividad de la enfermedad y para medir los efectos de la progresión radiográfica en pacientes tratados con anti-TNF.

En conclusión, con los resultados descritos podemos confirmar nuestra hipótesis. Los pacientes con EspAax tratados con anti-TNF que presenten un descenso de la actividad clínica, medida por ASDAS principalmente y BASDAI, presentan una menor progresión radiográfica.

5. Glosario

EspAax: Espondiloartritis axial

EA: Espondilitis Anquilosante

ASDAS: Ankylosing Spondylitis Disease Activity Score

BASDAI: Bath Ankylosing Spondylitis Disease Activity Index

BASFI: Bath Ankylosing Spondylitis Functional Index

PCR: proteína C reactiva

VGP: valoración del dolor por el paciente

EGM: EVA (escala de dolor) Global en la opinión del Médico

EGP: EVA (escala de dolor Global en la opinión del Paciente

mSASSS: modified Stoke Ankylosing Spondylitis Spine Score

Anti-TNF: inhibidores del factor de necrosis tumoral alfa

6. Bibliografía

1. Sieper J, Poddubnyy D. Axial spondyloarthritis. *Lancet*. 2017;390(10089):73-84.
2. Reveille JD, Witter JP, Weisman MH. Prevalence of axial spondylarthritis in the United States: estimates from a cross-sectional survey. *Arthritis Care Res (Hoboken)*. 2012;64(6):905-10.
3. Costantino F, Talpin A, Said-Nahal R, Goldberg M, Henny J, Chiocchia G, et al. Prevalence of spondyloarthritis in reference to HLA-B27 in the French population: results of the GAZEL cohort. *Ann Rheum Dis*. 2015;74(4):689-93.
4. van der Linden S, Valkenburg HA, Cats A. Evaluation of diagnostic criteria for ankylosing spondylitis. A proposal for modification of the New York criteria. *Arthritis Rheum*. 1984;27(4):361-8.
5. Rudwaleit M, van der Heijde D, Landewé R, Listing J, Akkoc N, Brandt J, et al. The development of Assessment of SpondyloArthritis international Society classification criteria for axial spondyloarthritis (part II): validation and final selection. *Ann Rheum Dis*. 2009;68(6):777-83.
6. Tomero E, Mulero J, de Miguel E, Fernández-Espartero C, Gobbo M, Descalzo MA, et al. Performance of the Assessment of Spondyloarthritis International Society criteria for the classification of spondyloarthritis in early spondyloarthritis clinics participating in the ESPERANZA programme. *Rheumatology (Oxford)*. 2014;53(2):353-60.
7. Moltó A, Paternotte S, Comet D, Thibout E, Rudwaleit M, Claudepierre P, et al. Performances of the Assessment of SpondyloArthritis International Society axial spondyloarthritis criteria for diagnostic and classification purposes in patients visiting a rheumatologist because of chronic back pain: results from a multicenter, cross-sectional study. *Arthritis Care Res (Hoboken)*. 2013;65(9):1472-81.
8. Song IH, Hermann K, Haibel H, Althoff CE, Althoff C, Listing J, et al. Effects of etanercept versus sulfasalazine in early axial spondyloarthritis on active inflammatory lesions as detected by whole-body MRI (ESTHER): a 48-week randomised controlled trial. *Ann Rheum Dis*. 2011;70(4):590-6.

9. Creemers MC, Franssen MJ, van't Hof MA, Gribnau FW, van de Putte LB, van Riel PL. Assessment of outcome in ankylosing spondylitis: an extended radiographic scoring system. *Ann Rheum Dis.* 2005;64(1):127-9.
10. van der Heijde D, Lie E, Kvien TK, Sieper J, Van den Bosch F, Listing J, et al. ASDAS, a highly discriminatory ASAS-endorsed disease activity score in patients with ankylosing spondylitis. *Ann Rheum Dis.* 2009;68(12):1811-8.
11. Sieper J, Poddubnyy D. Inflammation, new bone formation and treatment options in axial spondyloarthritis. *Ann Rheum Dis.* 2014;73(8):1439-41.
12. Baraliakos X, Gensler LS, D'Angelo S, Iannone F, Favalli EG, de Peyrecave N, et al. Biologic therapy and spinal radiographic progression in patients with axial spondyloarthritis: A structured literature review. *Ther Adv Musculoskelet Dis.* 2020;12:1759720X20906040.
13. Molnar C, Scherer A, Baraliakos X, de Hooge M, Micheroli R, Exer P, et al. TNF blockers inhibit spinal radiographic progression in ankylosing spondylitis by reducing disease activity: results from the Swiss Clinical Quality Management cohort. *Ann Rheum Dis.* 2018;77(1):63-9.

7. Anexos

TFM

Maria Llop

28 de octubre de 2020

Descripción de los datos

```
install.packages("readxl", repos=c("http://rstudio.org/_packages",
"http://cran.rstudio.com"),dependencies=TRUE))

## Installing package into 'C:/Users/maria/OneDrive/Documentos/R/win-library/4.0'
## (as 'lib' is unspecified)

## Warning: unable to access index for repository TRUE/src/contrib:
## scheme not supported in URL 'TRUE/src/contrib/PACKAGES'

## Warning: unable to access index for repository http://rstudio.org/_packages/bin/windows/
contrib/4.0:
## no fue posible abrir la URL 'http://rstudio.org/_packages/bin/windows/contrib/4.0/PACKA
GES'

## Warning: unable to access index for repository TRUE/bin/windows/contrib/4.0:
## scheme not supported in URL 'TRUE/bin/windows/contrib/4.0/PACKAGES'

## package 'readxl' successfully unpacked and MD5 sums checked

## Warning: cannot remove prior installation of package 'readxl'

## Warning in file.copy(savedcopy, lib, recursive = TRUE):
## problema al copiar C:\Users\maria\OneDrive\Documentos\R\win-
## library\4.0\00LOCK\readxl\libs\x64\readxl.dll a C:
## \Users\maria\OneDrive\Documentos\R\win-library\4.0\readxl\libs\x64\readxl.dll:
## Permission denied

## Warning: restored 'readxl'

##
## The downloaded binary packages are in
## C:\Users\maria\AppData\Local\Temp\RtmpGGIHpg\downloaded_packages

library(readxl)

base_rx<-read_excel("C:/Users/maria/OneDrive/Documentos/TFM/variables_regisponserbio.x
lsx")
```

Realizamos una descripción de las variables, para ver como se han codificado y ver los missin

gs.

summary(base_rx)

```
## NUM_PAC      EDAD      SEXO      IMC
## Length:101   Min. :21.00 Min. :1.000 Min. :19.37
## Class:character 1st Qu.:38.00 1st Qu.:1.000 1st Qu.:24.05
## Mode :character Median :46.00 Median :1.000 Median :26.02
##           Mean :46.68 Mean :1.188 Mean :26.38
##           3rd Qu.:54.00 3rd Qu.:1.000 3rd Qu.:28.68
##           Max. :75.00 Max. :2.000 Max. :40.83
##           NA's :6
## IMC>30      TSINT      HLAB27      EA
## Min. :0.0000 Min. :0.00 Min. :1.000 Min. :1.000
## 1st Qu.:0.0000 1st Qu.: 8.00 1st Qu.:1.000 1st Qu.:1.000
## Median :0.0000 Median :15.00 Median :1.000 Median :1.000
## Mean :0.1158 Mean :17.76 Mean :1.131 Mean :1.149
## 3rd Qu.:0.0000 3rd Qu.:26.00 3rd Qu.:1.000 3rd Qu.:1.000
## Max. :1.0000 Max. :52.00 Max. :2.000 Max. :2.000
## NA's :6      NA's :4      NA's :2
## TABACO_Kat  VSG      BASMI      UVEITIS
## Min. :0.0000 Min. : 0.50 Min. :0.5378 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.: 9.00 1st Qu.:1.6458 1st Qu.:0.0000
## Median :0.0000 Median :19.00 Median :2.8728 Median :0.0000
## Mean :0.3069 Mean :26.65 Mean :3.0915 Mean :0.2178
## 3rd Qu.:1.0000 3rd Qu.:38.75 3rd Qu.:4.0864 3rd Qu.:0.0000
## Max. :1.0000 Max. :115.00 Max. :7.2729 Max. :1.0000
##           NA's :3      NA's :19
## PSORIASIS  ENF_INF      TTOBIO  NUM_TB_PREV
## Min. :0.00000 Min. :0.00000 Min. :1.000 Min. :0.0000
## 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:1.000 1st Qu.:0.0000
## Median :0.00000 Median :0.00000 Median :2.000 Median :0.0000
## Mean :0.07071 Mean :0.07071 Mean :2.129 Mean :0.3168
## 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:3.000 3rd Qu.:1.0000
## Max. :1.00000 Max. :1.00000 Max. :5.000 Max. :2.0000
## NA's :2      NA's :2
## MOTIVO      TIEMBIO      BIO_INICI  Biol_4_kat
## Min. :0.0000 Min. : 0.00 Min. :0.00 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.: 8.00 1st Qu.:0.75 1st Qu.:0.0000
## Median :0.0000 Median :42.50 Median :1.00 Median :0.0000
## Mean :0.3069 Mean :45.51 Mean :0.75 Mean :0.4554
## 3rd Qu.:0.0000 3rd Qu.:76.75 3rd Qu.:1.00 3rd Qu.:1.0000
## Max. :2.0000 Max. :132.00 Max. :1.00 Max. :1.0000
##           NA's :1      NA's :1
## Interval  AINE      FAME      i_BASDAI
## Min. :2.00 Min. :0.0000 Min. :0.0000 Min. :0.700
## 1st Qu.:3.00 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:4.225
## Median :3.00 Median :1.0000 Median :0.0000 Median :5.400
## Mean :3.45 Mean :0.5758 Mean :0.1818 Mean :5.326
## 3rd Qu.:4.00 3rd Qu.:1.0000 3rd Qu.:0.0000 3rd Qu.:6.400
## Max. :6.00 Max. :1.0000 Max. :1.0000 Max. :8.800
## NA's :1      NA's :2      NA's :2      NA's :3
## BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3
```

```

## Min. :0.20 Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:2.00 1st Qu.:1.400 1st Qu.:1.400 1st Qu.:1.200
## Median :3.20 Median :2.700 Median :2.500 Median :3.000
## Mean :3.67 Mean :2.902 Mean :3.419 Mean :3.108
## 3rd Qu.:5.10 3rd Qu.:4.400 3rd Qu.:4.200 3rd Qu.:4.800
## Max. :8.80 Max. :8.500 Max. :65.000 Max. :8.000
## NA's :4 NA's :4 NA's :4 NA's :8
## BASDAI_4 BASDAI_5 BASDAI_6 ASDAS_0
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.2257
## 1st Qu.:1.200 1st Qu.:1.275 1st Qu.:1.100 1st Qu.:1.3874
## Median :2.800 Median :2.550 Median :2.800 Median :2.0565
## Mean :3.075 Mean :2.915 Mean :2.924 Mean :2.2160
## 3rd Qu.:4.600 3rd Qu.:4.200 3rd Qu.:4.600 3rd Qu.:2.9248
## Max. :8.000 Max. :8.000 Max. :7.300 Max. :5.0749
## NA's :10 NA's :17 NA's :16 NA's :5
## ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4
## Min. :0.2471 Min. :0.001737 Min. :0.04892 Min. :0.00058
## 1st Qu.:1.1572 1st Qu.:1.088078 1st Qu.:1.16497 1st Qu.:1.17363
## Median :1.5738 Median :1.827288 Median :1.76656 Median :1.66660
## Mean :1.8249 Mean :1.773470 Mean :1.81523 Mean :1.81274
## 3rd Qu.:2.4204 3rd Qu.:2.316346 3rd Qu.:2.47797 3rd Qu.:2.44173
## Max. :7.1255 Max. :4.131246 Max. :3.98374 Max. :4.45150
## NA's :13 NA's :9 NA's :14 NA's :22
## ASDAS_5 ASDAS_6 PCR_0 PCR_1
## Min. :0.0338 Min. :0.096 Min. :0.000 Min. :0.000
## 1st Qu.:1.0522 1st Qu.:1.143 1st Qu.:1.000 1st Qu.:1.000
## Median :1.6400 Median :1.680 Median :3.350 Median :2.000
## Mean :1.7329 Mean :1.807 Mean :8.181 Mean :4.358
## 3rd Qu.:2.1781 3rd Qu.:2.397 3rd Qu.:9.100 3rd Qu.:4.675
## Max. :4.0579 Max. :3.988 Max. :88.700 Max. :61.300
## NA's :30 NA's :24 NA's :7 NA's :3
## PCR_2 PCR_3 PCR_4 PCR_5
## Min. :0.00 Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:0.90 1st Qu.:0.750 1st Qu.:0.700 1st Qu.:0.615
## Median :2.00 Median :1.900 Median :2.050 Median :1.400
## Mean :4.36 Mean :4.452 Mean :5.368 Mean :4.118
## 3rd Qu.:5.60 3rd Qu.:5.600 3rd Qu.:5.375 3rd Qu.:6.000
## Max. :33.90 Max. :37.000 Max. :68.500 Max. :65.900
## NA's :2 NA's :4 NA's :7 NA's :10
## PCR_6 EGP EGM BASFI_0
## Min. :0.00 Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:0.60 1st Qu.:2.000 1st Qu.:1.500 1st Qu.:1.525
## Median :2.00 Median :3.000 Median :3.000 Median :3.850
## Mean :4.42 Mean :4.031 Mean :3.455 Mean :3.931
## 3rd Qu.:5.95 3rd Qu.:6.000 3rd Qu.:5.000 3rd Qu.:5.975
## Max. :39.70 Max. :10.000 Max. :10.000 Max. :9.300
## NA's :14 NA's :3 NA's :2 NA's :3
## BASFI_1 BASFI_2 BASFI_3 BASFI_4 BASFI_5
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.00 Min. :0.000
## 1st Qu.:1.200 1st Qu.:0.550 1st Qu.:1.000 1st Qu.:1.30 1st Qu.:1.100
## Median :2.700 Median :2.330 Median :2.800 Median :3.50 Median :3.200
## Mean :3.059 Mean :2.667 Mean :2.882 Mean :3.34 Mean :3.118

```

```

## 3rd Qu.:4.550 3rd Qu.:4.340 3rd Qu.:4.500 3rd Qu.:5.00 3rd Qu.:4.700
## Max. :8.800 Max. :7.890 Max. :7.900 Max. :8.40 Max. :8.100
## NA's :2          NA's :8  NA's :12  NA's :22
## BASFI_6 mSASSS_Mean_BL mSASSS_Mean_FL mSASSS_Progress
## Min. :0.000 Min. :0.00 Min. :0.00 Min. :-2.50
## 1st Qu.:1.100 1st Qu.:0.00 1st Qu.:0.00 1st Qu.:0.00
## Median :3.100 Median :5.00 Median :8.00 Median :0.00
## Mean :3.313 Mean :16.46 Mean :18.44 Mean :1.98
## 3rd Qu.:5.100 3rd Qu.:23.00 3rd Qu.:30.00 3rd Qu.:3.00
## Max. :8.300 Max. :72.00 Max. :72.00 Max. :21.50
## NA's :16
## New_Syn_mSASSS Prog_Syn_mSASSS mSASSS_Progress_Kat New_Prog_Syn_Msasss
## Min. :0.000 Min. :0.0000 Min. :0.000 Min. :0.0000
## 1st Qu.:0.000 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:0.0000
## Median :0.000 Median :0.0000 Median :0.000 Median :0.0000
## Mean :0.198 Mean :0.1089 Mean :0.297 Mean :0.2178
## 3rd Qu.:0.000 3rd Qu.:0.0000 3rd Qu.:1.000 3rd Qu.:0.0000
## Max. :1.000 Max. :1.0000 Max. :1.000 Max. :1.0000
##
## Mind_2_Synd_FL Mind_2_Synd_BL PCR_INICIOB i_ASIDAS
## Min. :0.0000 Min. :0.000 Min. :0.00 Min. :1.500
## 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:4.60 1st Qu.:2.700
## Median :1.0000 Median :1.000 Median :11.10 Median :3.500
## Mean :0.5446 Mean :0.505 Mean :18.70 Mean :3.402
## 3rd Qu.:1.0000 3rd Qu.:1.000 3rd Qu.:28.05 3rd Qu.:4.000
## Max. :1.0000 Max. :1.000 Max. :98.00 Max. :5.300
##
##          NA's :2  NA's :2
## HVGP
## Min. :1.000
## 1st Qu.:5.000
## Median :6.000
## Mean :5.923
## 3rd Qu.:7.000
## Max. :10.000
##

```

`str(base_rx)`

```

## tibble [101 x 66] (S3: tbl_df/tbl/data.frame)
## $ NUM_PAC      : chr [1:101] "U01001" "U01007" "U01017" "U02001" ...
## $ EDAD        : num [1:101] 63 34 26 57 41 40 43 65 60 41 ...
## $ SEXO        : num [1:101] 1 1 1 2 1 2 1 1 1 1 ...
## $ IMC         : num [1:101] 26.3 26.2 24.6 28.7 21.6 ...
## $ IMC>30      : num [1:101] 0 0 0 0 0 0 0 0 0 ...
## $ TSINT       : num [1:101] 1 13 6 13 18 16 13 38 21 13 ...
## $ HLAB27      : num [1:101] 1 1 1 1 1 2 1 1 1 1 ...
## $ EA          : num [1:101] 1 1 1 2 1 1 2 1 1 1 ...
## $ TABACO_Kat  : num [1:101] 0 0 0 0 1 0 1 0 1 0 ...
## $ VSG         : num [1:101] 50 16 3 24 38 25 9 18 10 52 ...
## $ BASMI       : num [1:101] 1.083 NA 0.979 2.947 3.265 ...
## $ UVEITIS     : num [1:101] 0 0 0 0 1 0 0 0 0 0 ...
## $ PSORIASIS   : num [1:101] 0 0 0 1 0 0 1 0 0 0 ...

```

```

## $ ENF_INF      : num [1:101] 0 0 0 0 0 0 1 0 0 ...
## $ TTOBIO      : num [1:101] 5 3 4 2 3 4 2 3 1 2 ...
## $ NUM_TB_PREV  : num [1:101] 0 0 0 1 0 0 0 0 0 0 ...
## $ MOTIVO      : num [1:101] 0 0 0 1 0 0 0 0 0 0 ...
## $ TIEMBIO     : num [1:101] 0 72 0 84 24 0 0 132 1 58 ...
## $ BIO_INICI   : num [1:101] 0 1 0 1 1 0 0 1 0 1 ...
## $ Biol_4_kat  : num [1:101] 0 1 0 1 0 0 0 1 0 1 ...
## $ Interval    : num [1:101] 4 4 3 4 4 4 4 3 3 4 ...
## $ AINE        : num [1:101] 1 1 1 1 1 1 1 1 1 1 ...
## $ FAME        : num [1:101] 1 1 0 1 0 0 1 0 0 0 ...
## $ i_BASDAI    : num [1:101] 2.2 5.9 2.2 3.8 4.5 5 8 8 7 5 ...
## $ BASDAI_0    : num [1:101] 2.2 3.9 2.2 3 2 5 8 5.5 7 3 ...
## $ BASDAI_1    : num [1:101] 2.8 2.3 1.7 5.2 1 2.5 6 6 4 3 ...
## $ BASDAI_2    : num [1:101] 1.8 3.4 1.7 3 0.5 5.5 4 65 3 1.9 ...
## $ BASDAI_3    : num [1:101] 0.6 1.8 1.7 5 1 4 5 8 1 2.4 ...
## $ BASDAI_4    : num [1:101] 1.8 1.8 1.7 6 1 5 8 5 6.1 2.8 ...
## $ BASDAI_5    : num [1:101] 1 2.7 0.8 6 1 4 8 NA 2.1 3 ...
## $ BASDAI_6    : num [1:101] 0.8 2.9 1.1 3 1 5 1 NA 2.4 3 ...
## $ ASDAS_0     : num [1:101] 3.52 3 2.05 2.45 1.18 ...
## $ ASDAS_1     : num [1:101] 1.112 1.648 1.467 2.468 0.762 ...
## $ ASDAS_2     : num [1:101] 0.44 2.68 1.865 2.367 0.492 ...
## $ ASDAS_3     : num [1:101] 0.425 1.907 0.65 2.69 2.35 ...
## $ ASDAS_4     : num [1:101] 0.795 1.426 1.534 3.209 0.79 ...
## $ ASDAS_5     : num [1:101] 0.466 2.261 1.245 2.562 1.58 ...
## $ ASDAS_6     : num [1:101] 0.402 2.851 1.055 1.903 0.957 ...
## $ PCR_0       : num [1:101] 88.7 8.8 11.3 10 2 1.6 9.2 5.4 16 2 ...
## $ PCR_1       : num [1:101] 0.3 3.1 2 3.7 1 3 2.2 4 4 2 ...
## $ PCR_2       : num [1:101] 0 9.8 5.6 5.4 0.9 2 0.4 1.3 4 2 ...
## $ PCR_3       : num [1:101] 0.5 6.1 0 4.8 37 15 0.6 3.6 1 3 ...
## $ PCR_4       : num [1:101] 0.4 2.1 2.8 5.1 1.1 4.1 0.5 20 1 2 ...
## $ PCR_5       : num [1:101] 0.2 7.5 3.22 1 7.2 6 1 NA 1 9 ...
## $ PCR_6       : num [1:101] 0.3 11 1.79 4 1.8 18 1 NA 1.5 1.9 ...
## $ EGP         : num [1:101] 3 7 1 4 2 5 9 6 10 2 ...
## $ EGM         : num [1:101] 3 7 1 4 2 5 5 3 5 1 ...
## $ BASFI_0     : num [1:101] 1.3 5.4 0.9 5.1 1.8 4.3 6 7.2 8.1 2.2 ...
## $ BASFI_1     : num [1:101] 0.8 3.5 0.6 5.9 0.8 3.4 2 7.6 4.8 1.6 ...
## $ BASFI_2     : num [1:101] 0.11 2.67 0.44 4.89 0.55 4.33 5.21 6.23 2.99 1.88 ...
## $ BASFI_3     : num [1:101] 0.1 2.7 0.4 4.9 0.5 4.3 5.1 6.3 2.9 1.8 ...
## $ BASFI_4     : num [1:101] 1.5 2.9 6 3.9 0.6 3.5 4 6.7 5.5 1.8 ...
## $ BASFI_5     : num [1:101] 4 3 0.6 5 0 4 4 NA 5.2 2.4 ...
## $ BASFI_6     : num [1:101] 0.8 5.2 1.6 3 0 4 1 NA 5 2.4 ...
## $ mSASSS_Mean_BL : num [1:101] 2.5 2.5 0 1 34 0 0 60 4 3 ...
## $ mSASSS_Mean_FL : num [1:101] 6 4 0 2 34 1.5 0 60 5.5 3 ...
## $ mSASSS_Progress : num [1:101] 3.5 1.5 0 1 0 1.5 0 0 1.5 0 ...
## $ New_Syn_mSASSS : num [1:101] 1 0 0 0 0 0 0 0 0 0 ...
## $ Prog_Syn_mSASSS : num [1:101] 0 0 0 0 0 0 0 0 0 0 ...
## $ mSASSS_Progress_Kat: num [1:101] 1 0 0 0 0 0 0 0 0 ...
## $ New_Prog_Syn_Msasss: num [1:101] 1 0 0 0 0 0 0 0 0 ...
## $ Mind_2_Synd_FL : num [1:101] 1 0 0 0 1 0 0 1 1 0 ...
## $ Mind_2_Synd_BL : num [1:101] 0 0 0 0 1 0 0 1 1 0 ...
## $ PCR_INICIOB : num [1:101] 88.7 18 11.3 26 15 16 9.2 26 16 30 ...

```

```
## $ i_ASIDAS      : num [1:101] 3.4 3.8 2.2 3.6 3 3.4 3.8 4.3 3.9 4.1 ...
## $ HVGP          : num [1:101] 3 7 1 7 4 5 5 5 8 ...
```

las siguientes variables no están bien codificadas, están codificadas como variables numéricas cuando en realidad son variables categóricas, corrijo el problema:

```
base_rx$NUM_PAC <- factor( base_rx$NUM_PAC )
base_rx$EA <- factor(base_rx$EA, labels = c("SI", "NO"))
base_rx$IMC>30 <- factor(base_rx$IMC>30, labels = c("NO", "SI"))
base_rx$HLAB27 <- factor(base_rx$HLAB27, labels = c("SI", "NO"))
base_rx$SEXO <- factor(base_rx$SEXO, labels = c("MALE", "FEMALE"))
base_rx$TABACO_Kat <- factor(base_rx$TABACO_Kat, labels = c("SI", "NO"))
base_rx$VSG = as.numeric(base_rx$VSG)
base_rx$UVEITIS <- factor(base_rx$UVEITIS, labels = c("NO", "SI"))
base_rx$PSORIASIS <- factor(base_rx$PSORIASIS, labels = c("NO", "SI"))
base_rx$ENF_INF <- factor(base_rx$ENF_INF, labels = c("NO", "SI"))
base_rx$TTOBIO <- factor(base_rx$TTOBIO, labels = c("ETA", "ADA", "IFX", "GOL", "CTZ"))
base_rx$NUM_TB_PREV <- factor(base_rx$NUM_TB_PREV, labels = c("treatments.0", "treatments.1", "treatments.2"))
base_rx$MOTIVO <- factor(base_rx$MOTIVO, labels = c("non.change", "Inefficacy", "Side.effects"))
base_rx$BIO_INICI <- factor(base_rx$BIO_INICI, labels = c("NO", "SI"))
base_rx$Biol_4_kat <- factor(base_rx$Biol_4_kat, labels = c("NO", "SI"))
base_rx$AINE <- factor(base_rx$AINE, labels = c("NO", "SI"))
base_rx$FAME <- factor(base_rx$FAME, labels = c("NO", "SI"))
base_rx$New_Syn_mSASSS <- factor(base_rx$New_Syn_mSASSS, labels = c("NO", "SI"))
base_rx$Prog_Syn_mSASSS <- factor(base_rx$Prog_Syn_mSASSS, labels = c("NO", "SI"))
base_rx$mSASSS_Progress_Kat <- factor(base_rx$mSASSS_Progress_Kat, labels = c("NO", "SI"))
base_rx$New_Prog_Syn_Msasss <- factor(base_rx$New_Prog_Syn_Msasss, labels = c("NO", "SI"))
base_rx$Mind_2_Synd_FL <- factor(base_rx$Mind_2_Synd_FL, labels = c("NO", "SI"))
base_rx$Mind_2_Synd_BL <- factor(base_rx$Mind_2_Synd_BL, labels = c("NO", "SI"))
```

Ahora repetimos el proceso de descripción de las variables:

```
str(base_rx)

## tibble [101 x 66] (S3: tbl_df/tbl/data.frame)
## $ NUM_PAC      : Factor w/ 101 levels "U01001", "U01007",...: 1 2 3 4 5 6 7 8 9 10 ...
## $ EDAD        : num [1:101] 63 34 26 57 41 40 43 65 60 41 ...
## $ SEXO        : Factor w/ 2 levels "MALE", "FEMALE": 1 1 1 2 1 2 1 1 1 1 ...
## $ IMC         : num [1:101] 26.3 26.2 24.6 28.7 21.6 ...
## $ IMC>30      : Factor w/ 2 levels "NO", "SI": 1 1 1 1 1 1 1 1 1 1 ...
## $ TSINT       : num [1:101] 1 13 6 13 18 16 13 38 21 13 ...
## $ HLAB27      : Factor w/ 2 levels "SI", "NO": 1 1 1 1 1 2 1 1 1 1 ...
## $ EA          : Factor w/ 2 levels "SI", "NO": 1 1 1 2 1 1 2 1 1 1 ...
## $ TABACO_Kat  : Factor w/ 2 levels "SI", "NO": 1 1 1 1 2 1 2 1 2 1 ...
## $ VSG         : num [1:101] 50 16 3 24 38 25 9 18 10 52 ...
## $ BASMI       : num [1:101] 1.083 NA 0.979 2.947 3.265 ...
## $ UVEITIS     : Factor w/ 2 levels "NO", "SI": 1 1 1 1 2 1 1 1 1 1 ...
## $ PSORIASIS   : Factor w/ 2 levels "NO", "SI": 1 1 1 2 1 1 2 1 1 1 ...
## $ ENF_INF     : Factor w/ 2 levels "NO", "SI": 1 1 1 1 1 1 1 2 1 1 ...
## $ TTOBIO      : Factor w/ 5 levels "ETA", "ADA", "IFX",...: 5 3 4 2 3 4 2 3 1 2 ...
## $ NUM_TB_PREV : Factor w/ 3 levels "treatments.0",...: 1 1 1 2 1 1 1 1 1 1 ...
```

```

## $ MOTIVO      : Factor w/ 3 levels "non.change","Inefficacy",...: 1 1 1 2 1 1 1 1 1 1 ...
## $ TIEMBIO     : num [1:101] 0 72 0 84 24 0 0 132 1 58 ...
## $ BIO_INICI   : Factor w/ 2 levels "NO","SI": 1 2 1 2 2 1 1 2 1 2 ...
## $ Biol_4_kat  : Factor w/ 2 levels "NO","SI": 1 2 1 2 1 1 1 2 1 2 ...
## $ Interval    : num [1:101] 4 4 3 4 4 4 4 3 3 4 ...
## $ AINE        : Factor w/ 2 levels "NO","SI": 2 2 2 2 2 2 2 2 2 2 ...
## $ FAME        : Factor w/ 2 levels "NO","SI": 2 2 1 2 1 1 2 1 1 1 ...
## $ i_BASDAI   : num [1:101] 2.2 5.9 2.2 3.8 4.5 5 8 8 7 5 ...
## $ BASDAI_0    : num [1:101] 2.2 3.9 2.2 3 2 5 8 5.5 7 3 ...
## $ BASDAI_1    : num [1:101] 2.8 2.3 1.7 5.2 1 2.5 6 6 4 3 ...
## $ BASDAI_2    : num [1:101] 1.8 3.4 1.7 3 0.5 5.5 4 65 3 1.9 ...
## $ BASDAI_3    : num [1:101] 0.6 1.8 1.7 5 1 4 5 8 1 2.4 ...
## $ BASDAI_4    : num [1:101] 1.8 1.8 1.7 6 1 5 8 5 6.1 2.8 ...
## $ BASDAI_5    : num [1:101] 1 2.7 0.8 6 1 4 8 NA 2.1 3 ...
## $ BASDAI_6    : num [1:101] 0.8 2.9 1.1 3 1 5 1 NA 2.4 3 ...
## $ ASDAS_0     : num [1:101] 3.52 3 2.05 2.45 1.18 ...
## $ ASDAS_1     : num [1:101] 1.112 1.648 1.467 2.468 0.762 ...
## $ ASDAS_2     : num [1:101] 0.44 2.68 1.865 2.367 0.492 ...
## $ ASDAS_3     : num [1:101] 0.425 1.907 0.65 2.69 2.35 ...
## $ ASDAS_4     : num [1:101] 0.795 1.426 1.534 3.209 0.79 ...
## $ ASDAS_5     : num [1:101] 0.466 2.261 1.245 2.562 1.58 ...
## $ ASDAS_6     : num [1:101] 0.402 2.851 1.055 1.903 0.957 ...
## $ PCR_0       : num [1:101] 88.7 8.8 11.3 10 2 1.6 9.2 5.4 16 2 ...
## $ PCR_1       : num [1:101] 0.3 3.1 2 3.7 1 3 2.2 4 4 2 ...
## $ PCR_2       : num [1:101] 0 9.8 5.6 5.4 0.9 2 0.4 1.3 4 2 ...
## $ PCR_3       : num [1:101] 0.5 6.1 0 4.8 37 15 0.6 3.6 1 3 ...
## $ PCR_4       : num [1:101] 0.4 2.1 2.8 5.1 1.1 4.1 0.5 20 1 2 ...
## $ PCR_5       : num [1:101] 0.2 7.5 3.22 1 7.2 6 1 NA 1 9 ...
## $ PCR_6       : num [1:101] 0.3 11 1.79 4 1.8 18 1 NA 1.5 1.9 ...
## $ EGP         : num [1:101] 3 7 1 4 2 5 9 6 10 2 ...
## $ EGM         : num [1:101] 3 7 1 4 2 5 5 3 5 1 ...
## $ BASFI_0     : num [1:101] 1.3 5.4 0.9 5.1 1.8 4.3 6 7.2 8.1 2.2 ...
## $ BASFI_1     : num [1:101] 0.8 3.5 0.6 5.9 0.8 3.4 2 7.6 4.8 1.6 ...
## $ BASFI_2     : num [1:101] 0.11 2.67 0.44 4.89 0.55 4.33 5.21 6.23 2.99 1.88 ...
## $ BASFI_3     : num [1:101] 0.1 2.7 0.4 4.9 0.5 4.3 5.1 6.3 2.9 1.8 ...
## $ BASFI_4     : num [1:101] 1.5 2.9 6 3.9 0.6 3.5 4 6.7 5.5 1.8 ...
## $ BASFI_5     : num [1:101] 4 3 0.6 5 0 4 4 NA 5.2 2.4 ...
## $ BASFI_6     : num [1:101] 0.8 5.2 1.6 3 0 4 1 NA 5 2.4 ...
## $ mSASSS_Mean_BL : num [1:101] 2.5 2.5 0 1 34 0 0 60 4 3 ...
## $ mSASSS_Mean_FL : num [1:101] 6 4 0 2 34 1.5 0 60 5.5 3 ...
## $ mSASSS_Progress : num [1:101] 3.5 1.5 0 1 0 1.5 0 0 1.5 0 ...
## $ New_Syn_mSASSS : Factor w/ 2 levels "NO","SI": 2 1 1 1 1 1 1 1 1 1 ...
## $ Prog_Syn_mSASSS : Factor w/ 2 levels "NO","SI": 1 1 1 1 1 1 1 1 1 1 ...
## $ mSASSS_Progress_Kat: Factor w/ 2 levels "NO","SI": 2 1 1 1 1 1 1 1 1 1 ...
## $ New_Prog_Syn_Msasss: Factor w/ 2 levels "NO","SI": 2 1 1 1 1 1 1 1 1 1 ...
## $ Mind_2_Synd_FL : Factor w/ 2 levels "NO","SI": 2 1 1 1 2 1 1 2 2 1 ...
## $ Mind_2_Synd_BL : Factor w/ 2 levels "NO","SI": 1 1 1 1 2 1 1 2 2 1 ...
## $ PCR_INICIOB  : num [1:101] 88.7 18 11.3 26 15 16 9.2 26 16 30 ...
## $ i_ASIDAS    : num [1:101] 3.4 3.8 2.2 3.6 3 3.4 3.8 4.3 3.9 4.1 ...
## $ HVGP        : num [1:101] 3 7 1 7 4 5 5 5 5 8 ...

```

[summary\(base_rx\)](#)


```

## NUM_PAC EDAD SEXO IMC IMC>30
## U01001 :1 Min. :21.00 MALE :82 Min. :19.37 NO :84
## U01007 :1 1st Qu.:38.00 FEMALE:19 1st Qu.:24.05 SI :11
## U01017 :1 Median :46.00 Median :26.02 NA's: 6
## U02001 :1 Mean :46.68 Mean :26.38
## U02003 :1 3rd Qu.:54.00 3rd Qu.:28.68
## U02004 :1 Max. :75.00 Max. :40.83
## (Other):95 NA's :6
## TSINT HLAB27 EA TABACO_Kat VSG BASMI
## Min. :0.00 SI :86 SI:86 SI:70 Min. : 0.50 Min. :0.5378
## 1st Qu.:8.00 NO :13 NO:15 NO:31 1st Qu.: 9.00 1st Qu.:1.6458
## Median :15.00 NA's: 2 Median :19.00 Median :2.8728
## Mean :17.76 Mean :26.65 Mean :3.0915
## 3rd Qu.:26.00 3rd Qu.:38.75 3rd Qu.:4.0864
## Max. :52.00 Max. :115.00 Max. :7.2729
## NA's :4 NA's :3 NA's :19
## UVEITIS PSORIASIS ENF_INF TTBIO NUM_TB_PREV MOTIVO
## NO:79 NO :92 NO :92 ETA:31 treatments.0:75 non.change :77
## SI:22 SI :7 SI :7 ADA:44 treatments.1:20 Inefficacy :17
## NA's: 2 NA's: 2 IFX:11 treatments.2: 6 Side.effects: 7
## GOL:12
## CTZ: 3
##
##
## TIEMBIO BIO_INICI Biol_4_kat Interval AINE FAME
## Min. : 0.00 NO :25 NO:55 Min. :2.00 NO :42 NO :81
## 1st Qu.: 8.00 SI :75 SI:46 1st Qu.:3.00 SI :57 SI :18
## Median :42.50 NA's: 1 Median :3.00 NA's: 2 NA's: 2
## Mean :45.51 Mean :3.45
## 3rd Qu.:76.75 3rd Qu.:4.00
## Max. :132.00 Max. :6.00
## NA's :1 NA's :1
## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2
## Min. :0.700 Min. :0.20 Min. :0.000 Min. :0.000
## 1st Qu.:4.225 1st Qu.:2.00 1st Qu.:1.400 1st Qu.: 1.400
## Median :5.400 Median :3.20 Median :2.700 Median :2.500
## Mean :5.326 Mean :3.67 Mean :2.902 Mean :3.419
## 3rd Qu.:6.400 3rd Qu.:5.10 3rd Qu.:4.400 3rd Qu.: 4.200
## Max. :8.800 Max. :8.80 Max. :8.500 Max. :65.000
## NA's :3 NA's :4 NA's :4 NA's :4
## BASDAI_3 BASDAI_4 BASDAI_5 BASDAI_6
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:1.200 1st Qu.:1.200 1st Qu.:1.275 1st Qu.:1.100
## Median :3.000 Median :2.800 Median :2.550 Median :2.800
## Mean :3.108 Mean :3.075 Mean :2.915 Mean :2.924
## 3rd Qu.:4.800 3rd Qu.:4.600 3rd Qu.:4.200 3rd Qu.:4.600
## Max. :8.000 Max. :8.000 Max. :8.000 Max. :7.300
## NA's :8 NA's :10 NA's :17 NA's :16
## ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3
## Min. :0.2257 Min. :0.2471 Min. :0.001737 Min. :0.04892
## 1st Qu.:1.3874 1st Qu.:1.1572 1st Qu.:1.088078 1st Qu.:1.16497
## Median :2.0565 Median :1.5738 Median :1.827288 Median :1.76656

```

```

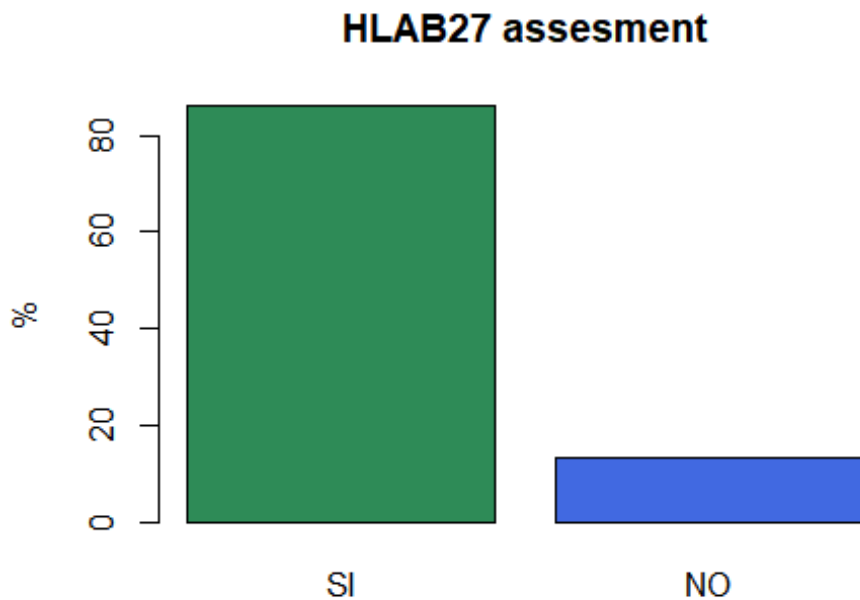
## Mean :2.2160 Mean :1.8249 Mean :1.773470 Mean :1.81523
## 3rd Qu.:2.9248 3rd Qu.:2.4204 3rd Qu.:2.316346 3rd Qu.:2.47797
## Max. :5.0749 Max. :7.1255 Max. :4.131246 Max. :3.98374
## NA's :5 NA's :13 NA's :9 NA's :14
## ASDAS_4 ASDAS_5 ASDAS_6 PCR_0
## Min. :0.00058 Min. :0.0338 Min. :0.096 Min. :0.000
## 1st Qu.:1.17363 1st Qu.:1.0522 1st Qu.:1.143 1st Qu.:1.000
## Median:1.66660 Median:1.6400 Median:1.680 Median:3.350
## Mean :1.81274 Mean :1.7329 Mean :1.807 Mean :8.181
## 3rd Qu.:2.44173 3rd Qu.:2.1781 3rd Qu.:2.397 3rd Qu.:9.100
## Max. :4.45150 Max. :4.0579 Max. :3.988 Max. :88.700
## NA's :22 NA's :30 NA's :24 NA's :7
## PCR_1 PCR_2 PCR_3 PCR_4
## Min. :0.000 Min. :0.00 Min. :0.000 Min. :0.000
## 1st Qu.:1.000 1st Qu.:0.90 1st Qu.:0.750 1st Qu.:0.700
## Median:2.000 Median:2.00 Median:1.900 Median:2.050
## Mean :4.358 Mean :4.36 Mean :4.452 Mean :5.368
## 3rd Qu.:4.675 3rd Qu.:5.60 3rd Qu.:5.600 3rd Qu.:5.375
## Max. :61.300 Max. :33.90 Max. :37.000 Max. :68.500
## NA's :3 NA's :2 NA's :4 NA's :7
## PCR_5 PCR_6 EGP EGM
## Min. :0.000 Min. :0.00 Min. :0.000 Min. :0.000
## 1st Qu.:0.615 1st Qu.:0.60 1st Qu.:2.000 1st Qu.:1.500
## Median:1.400 Median:2.00 Median:3.000 Median:3.000
## Mean :4.118 Mean :4.42 Mean :4.031 Mean :3.455
## 3rd Qu.:6.000 3rd Qu.:5.95 3rd Qu.:6.000 3rd Qu.:5.000
## Max. :65.900 Max. :39.70 Max. :10.000 Max. :10.000
## NA's :10 NA's :14 NA's :3 NA's :2
## BASFI_0 BASFI_1 BASFI_2 BASFI_3 BASFI_4
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.00
## 1st Qu.:1.525 1st Qu.:1.200 1st Qu.:0.550 1st Qu.:1.000 1st Qu.:1.30
## Median:3.850 Median:2.700 Median:2.330 Median:2.800 Median:3.50
## Mean :3.931 Mean :3.059 Mean :2.667 Mean :2.882 Mean :3.34
## 3rd Qu.:5.975 3rd Qu.:4.550 3rd Qu.:4.340 3rd Qu.:4.500 3rd Qu.:5.00
## Max. :9.300 Max. :8.800 Max. :7.890 Max. :7.900 Max. :8.40
## NA's :3 NA's :2 NA's :8 NA's :12
## BASFI_5 BASFI_6 mSASSS_Mean_BL mSASSS_Mean_FL
## Min. :0.000 Min. :0.000 Min. :0.00 Min. :0.00
## 1st Qu.:1.100 1st Qu.:1.100 1st Qu.:0.00 1st Qu.:0.00
## Median:3.200 Median:3.100 Median:5.00 Median:8.00
## Mean :3.118 Mean :3.313 Mean :16.46 Mean :18.44
## 3rd Qu.:4.700 3rd Qu.:5.100 3rd Qu.:23.00 3rd Qu.:30.00
## Max. :8.100 Max. :8.300 Max. :72.00 Max. :72.00
## NA's :22 NA's :16
## mSASSS_Progress New_Syn_mSASSS Prog_Syn_mSASSS mSASSS_Progress_Kat
## Min. :-2.50 NO:81 NO:90 NO:71
## 1st Qu.:0.00 SI:20 SI:11 SI:30
## Median:0.00
## Mean :1.98
## 3rd Qu.:3.00
## Max. :21.50
##

```

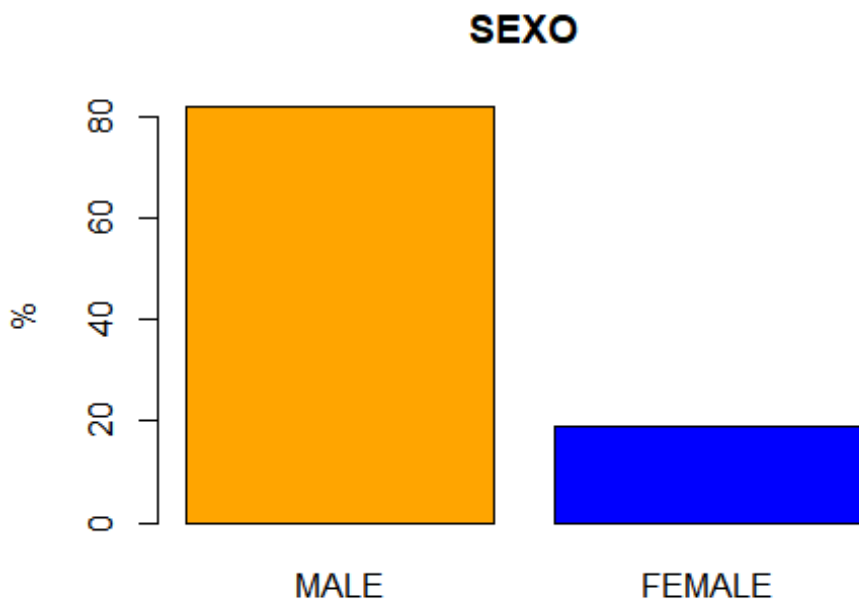
```
## New_Prog_Syn_Msassy Mind_2_Synd_FL Mind_2_Synd_BL PCR_INICIOB
## NO:79      NO:46      NO:50      Min. : 0.00
## SI:22      SI:55      SI:51      1st Qu.: 4.60
##                                     Median :11.10
##                                     Mean  :18.70
##                                     3rd Qu.:28.05
##                                     Max.  :98.00
##                                     NA's  :2
## i_ASIDAS   HVGP
## Min. :1.500 Min. : 1.000
## 1st Qu.:2.700 1st Qu.: 5.000
## Median :3.500 Median : 6.000
## Mean  :3.402 Mean  : 5.923
## 3rd Qu.:4.000 3rd Qu.: 7.000
## Max.  :5.300 Max.  :10.000
## NA's  :2
```

Análisis de outliers. Representación gráfica de las variables categóricas

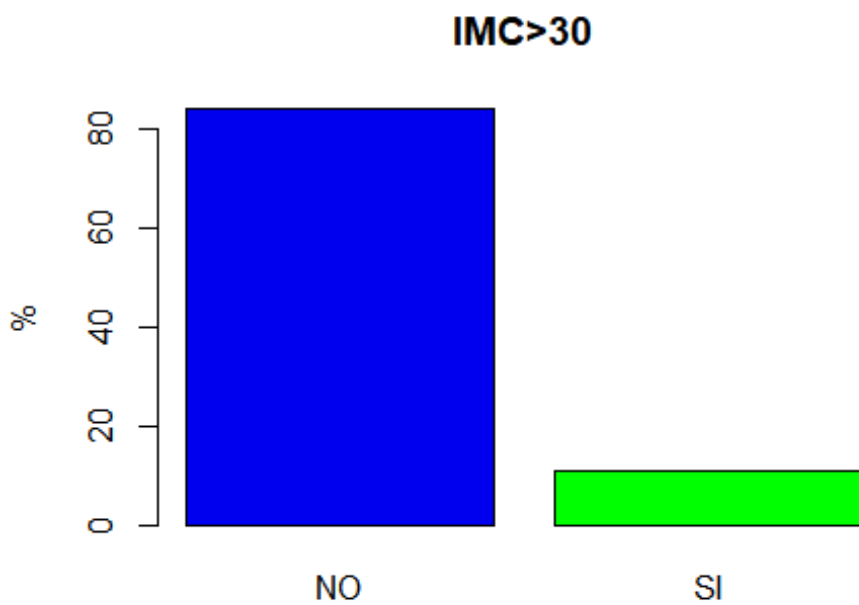
```
plot(x=base_rx$HLAB27, col=c("seagreen", "royalblue"), ylab="%", main="HLAB27 assesment")
```



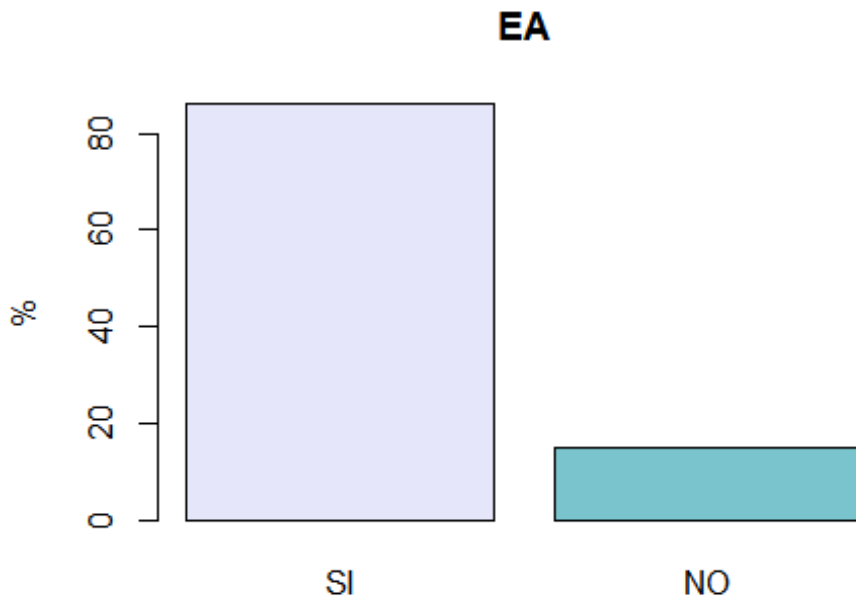
```
plot(x=base_rx$SEXO, col=c("orange", "blue"), ylab="%", main="SEXO")
```



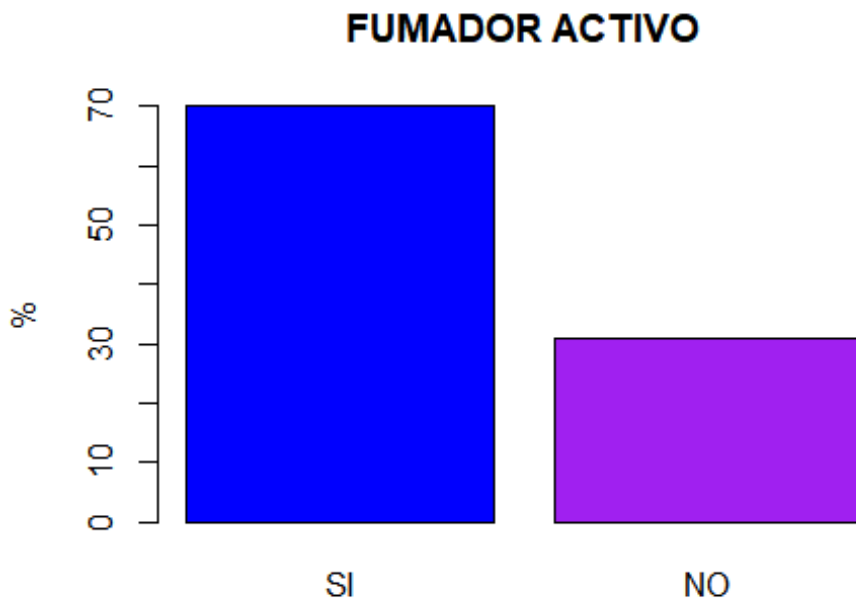
```
plot(x=base_rx$IMC>30, col=c("blue2","green"), ylab="%", main="IMC>30")
```



```
plot(x=base_rx$EA, col=c("lavender","cadetblue3"), ylab="%", main="EA")
```

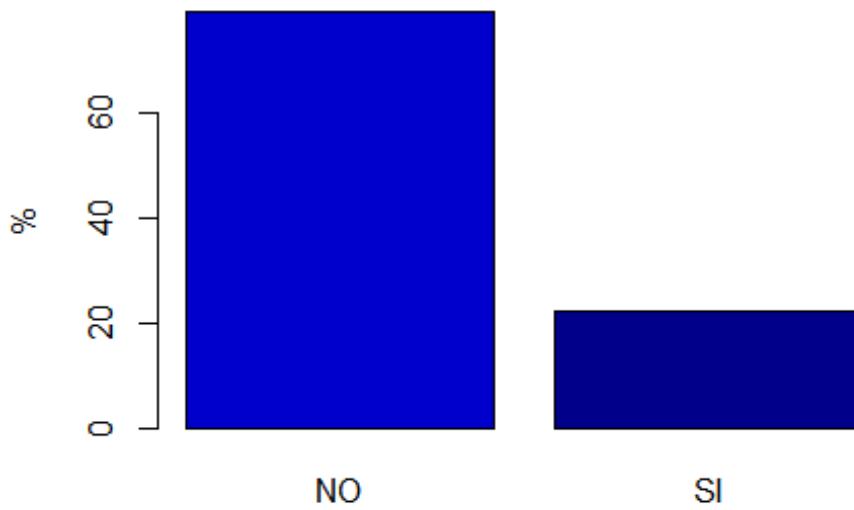


```
plot(x=base_rx$TABACO_Kat, col=c("blue", "purple"), ylab="%", main="FUMADOR ACTIVO")
```



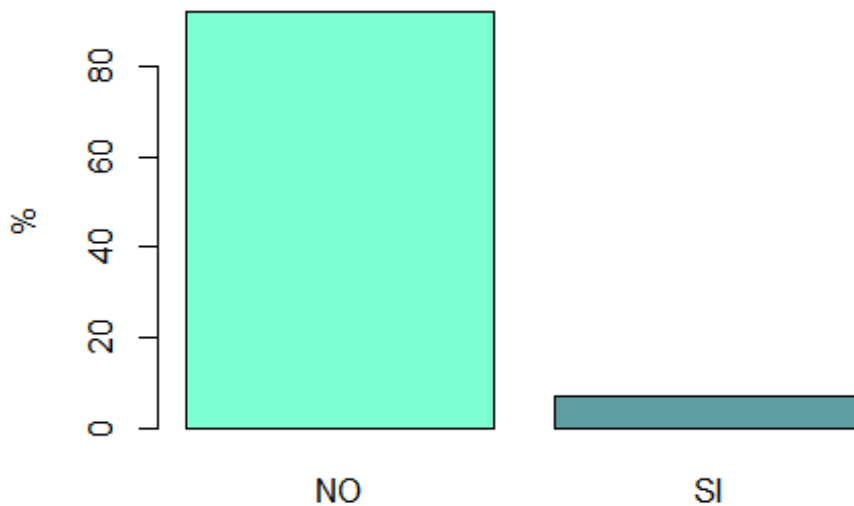
```
plot(x=base_rx$UVEITIS, col=c("blue3", "blue4"), ylab="%", main="PRESENCIA DE UVEÍTIS")
```

PRESENICA DE UVEĂTIS



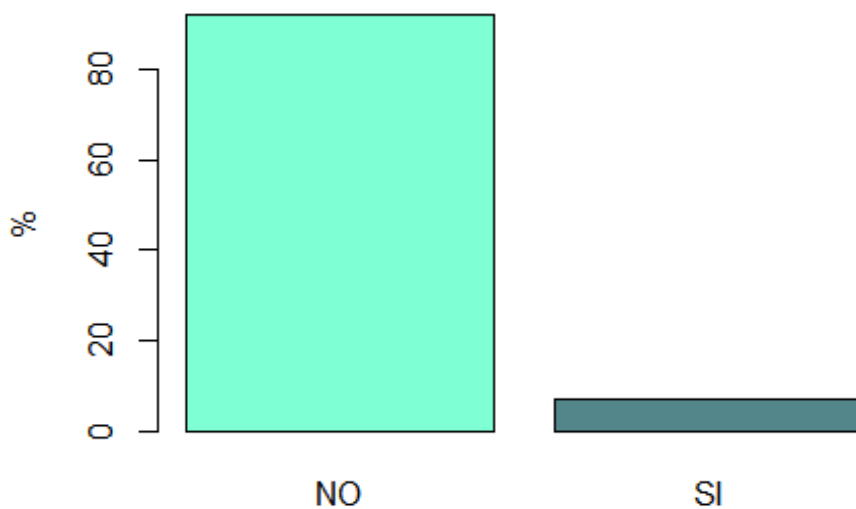
```
plot(x=base_rx$PSORIASIS, col=c("aquamarine", "cadetblue"), ylab="%", main="PRESENICA DE PSORIASIS")
```

PRESENICA DE PSORIASIS



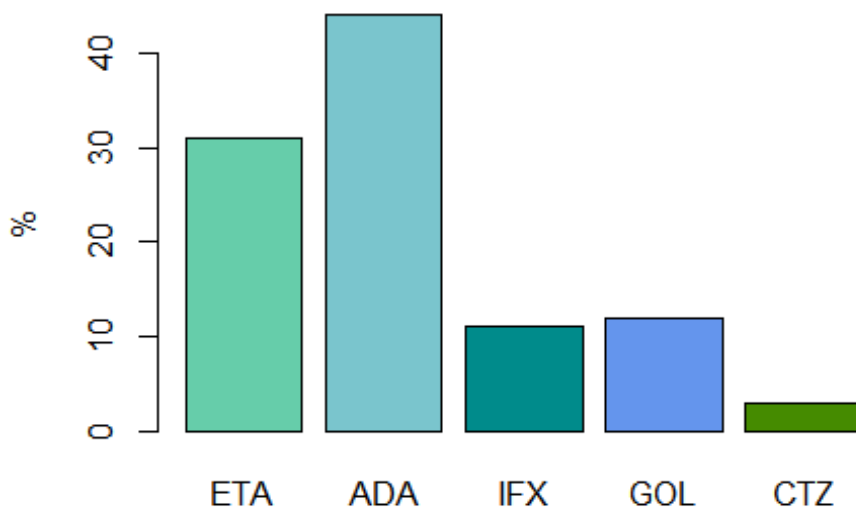
```
plot(x=base_rx$ENF_INF, col=c("aquamarine1", "cadetblue4"), ylab="%", main="PRESENICA DE ENFERMEDAD INFLAMATORIA INTESTINAL")
```

RESEÑA DE ENFERMEDAD INFLAMATORIA INTES'



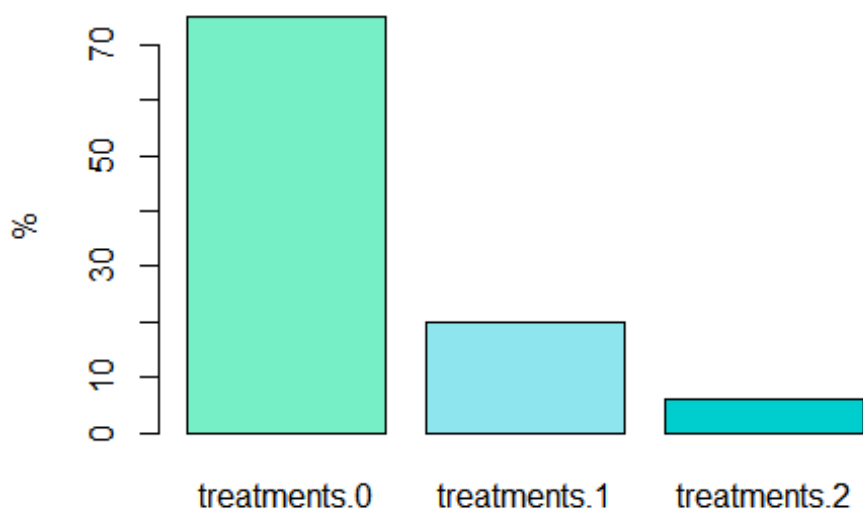
```
plot(x=base_rx$TTOBIO, col=c("aquamarine3", "cadetblue3", "cyan4", "cornflowerblue", "chartreuse4"), ylab="%", main="TIPO DE TRATAMIENTO BIOLÓGICO RECIBIDO")
```

TIPO DE TRATAMIENTO BIOLÓGICO RECIBIDO



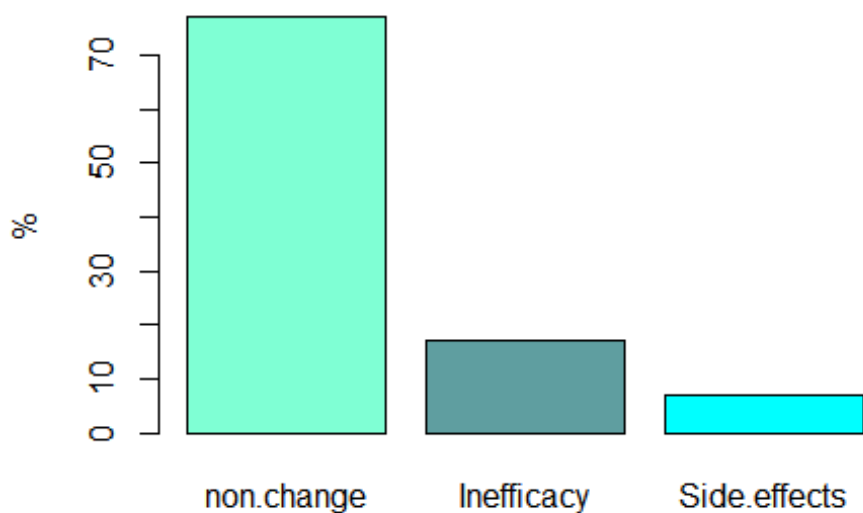
```
plot(x=base_rx$NUM_TB_PREV, col=c("aquamarine2", "cadetblue2", "cyan3"), ylab="%", main="NÚMERO DE TRATAMIENTOS BIOLÓGICOS PREVIOS")
```

09A>MERO DE TRATAMIENTOS BIOLĂ<U+0093>GIC



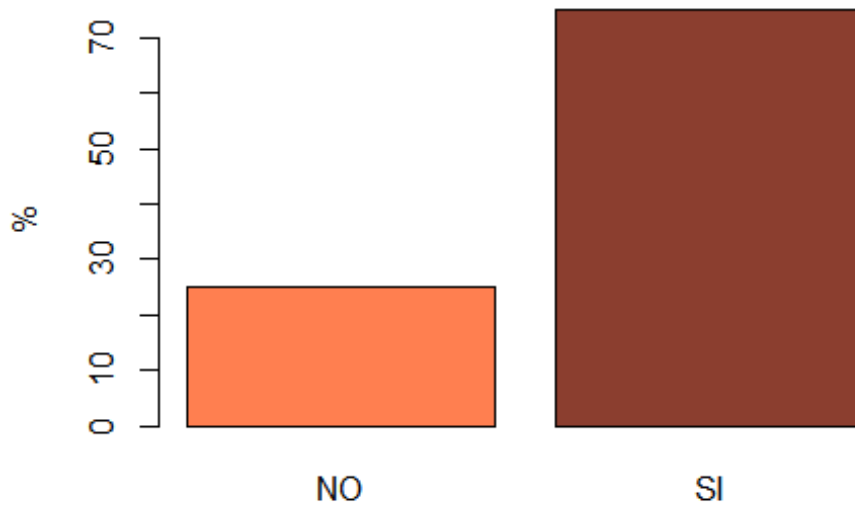
```
plot(x=base_rx$MOTIVO, col=c("aquamarine", "cadetblue", "cyan"), ylab="%", main="MOTIVO DE CAMBIO TRATAMIENTO BIOLĂ“GICO")
```

IOTIVO DE CAMBIO TRATAMIENTO BIOLĂ<U+0093>



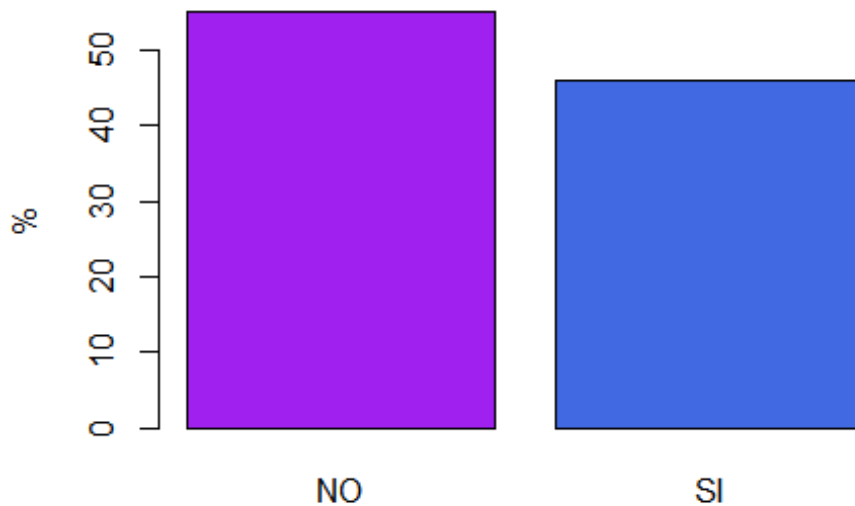
```
plot(x=base_rx$BIO_INICI, col=c("coral", "coral4"), ylab="%", main="TRATAMIENTO BIOLĂ“GICO AL INICIO DEL ESTUDIO")
```


TRATAMIENTO BIOLÓGICO AL INICIO DEL ESTUDIO



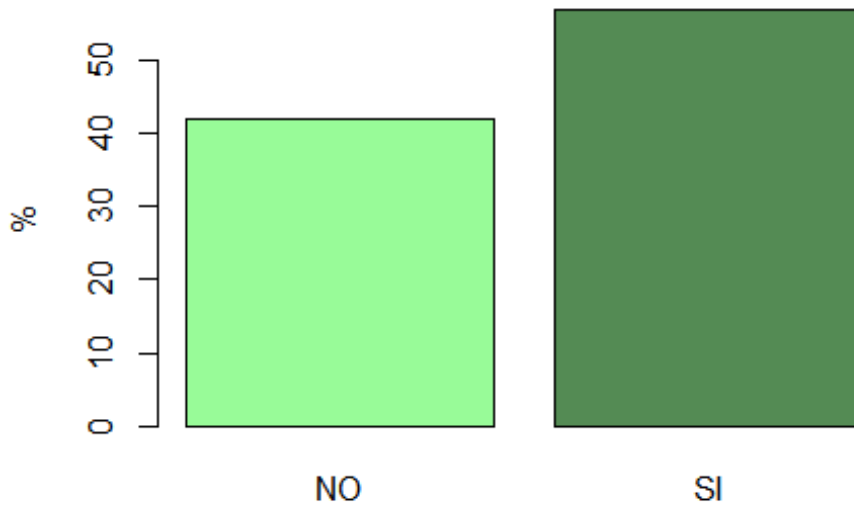
```
plot(x=base_rx$Biol_4_kat, col=c("purple", "royalblue"), ylab="%", main="TRATAMIENTO BIOLÓGICO DE MÁS DE 4 AÑOS AL INICIO DEL ESTUDIO")
```

TRATAMIENTO BIOLÓGICO DE MÁS DE 4 AÑOS AL



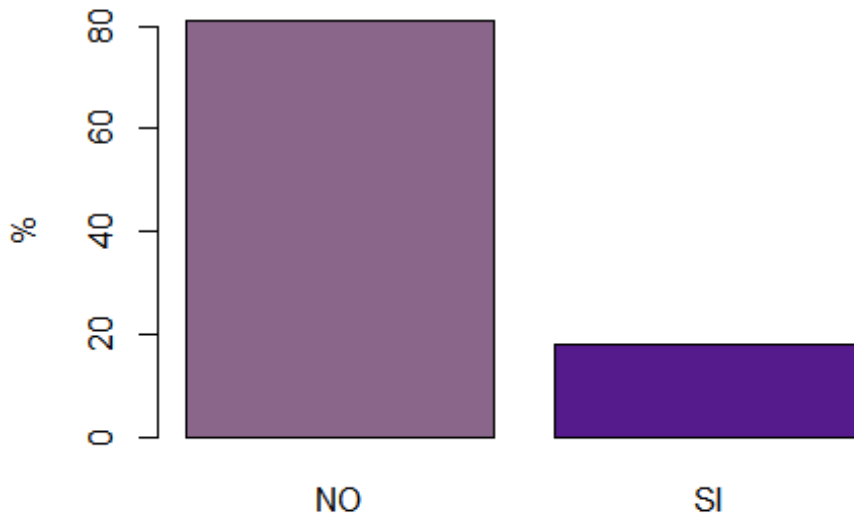
```
plot(x=base_rx$AINE, col=c("palegreen", "palegreen4"), ylab="%", main="TRATAMIENTO CON ANTIINFLAMATORIOS")
```

TRATAMIENTO CON ANTIINFLAMATORIOS



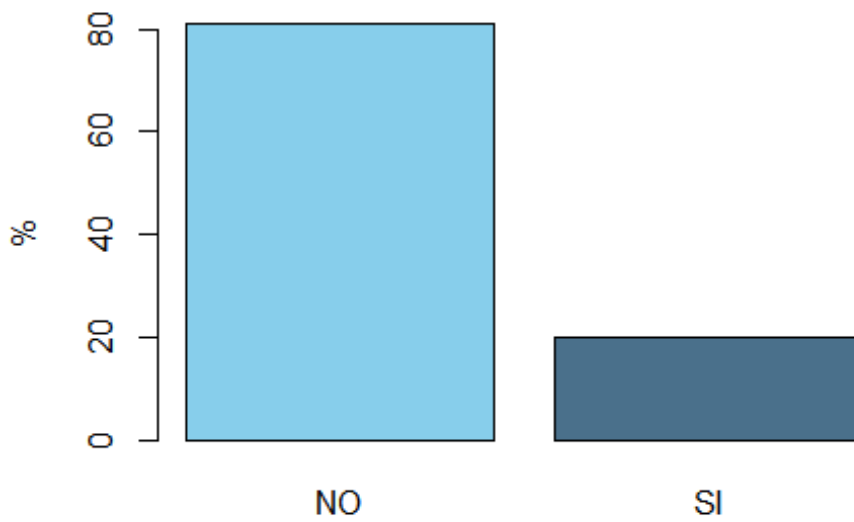
```
plot(x=base_rx$FAME, col=c("plum4","purple4"), ylab="%", main="TRATAMIENTO CON FÁRMACOS MODIFICADORES DE LA ENFERMEDAD")
```

TRATAMIENTO CON FÁRMACOS MODIFICADORES DE LA ENFERMEDAD



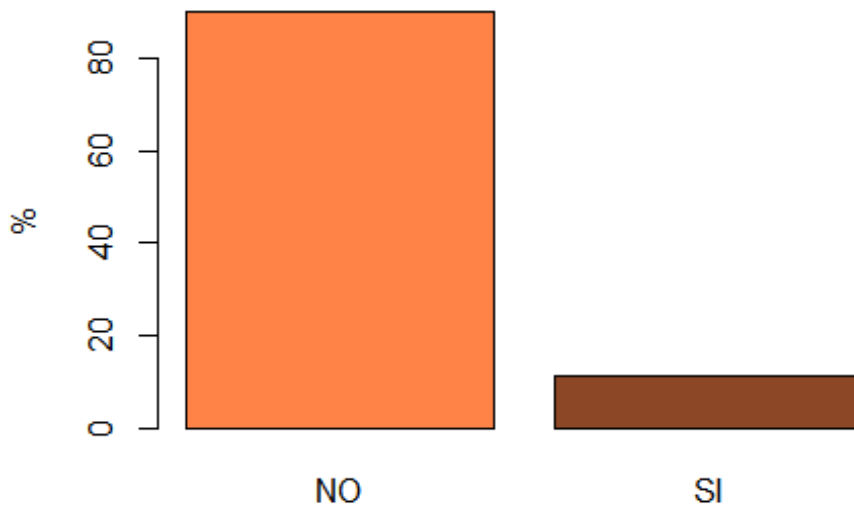
```
plot(x=base_rx$New_Syn_mSASSS, col=c("skyblue","skyblue4"), ylab="%", main="Nuevos síntomas")
```

Nuevos sindesmofitos



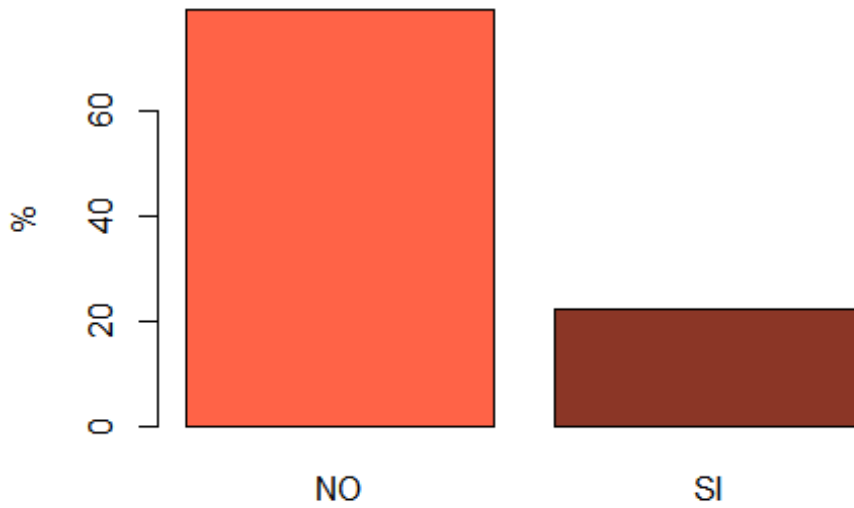
```
plot(x=base_rx$Prog_Syn_mSASSS, col=c("sienna1", "sienna4"), ylab="%", main="Progresi3n d e sindesmofitos")
```

Progresi3n de sindesmofitos



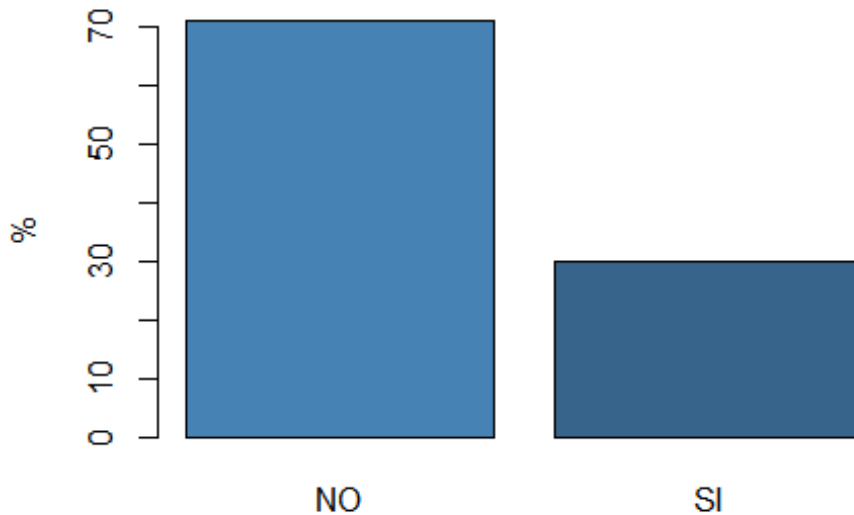
```
plot(x=base_rx$New_Prog_Syn_Msasss, col=c("tomato", "tomato4"), ylab="%", main="Nuevos o progresi3n de sindesmofitos")
```

Nuevos o progresión de sindesmoftos



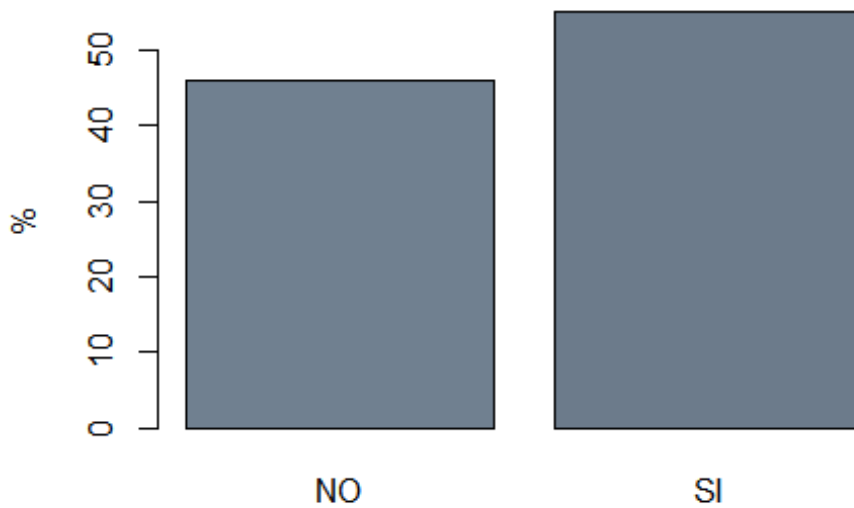
```
plot(x=base_rx$mSASSS_Progress_Kat, col=c("steelblue", "steelblue4"), ylab="%", main="Progresin de mÃ¡s de >2 puntos en la puntuaciÃ³n del Msasss")
```

Progresin de mÃ¡s de >2 puntos en la puntuaciÃ³n del I



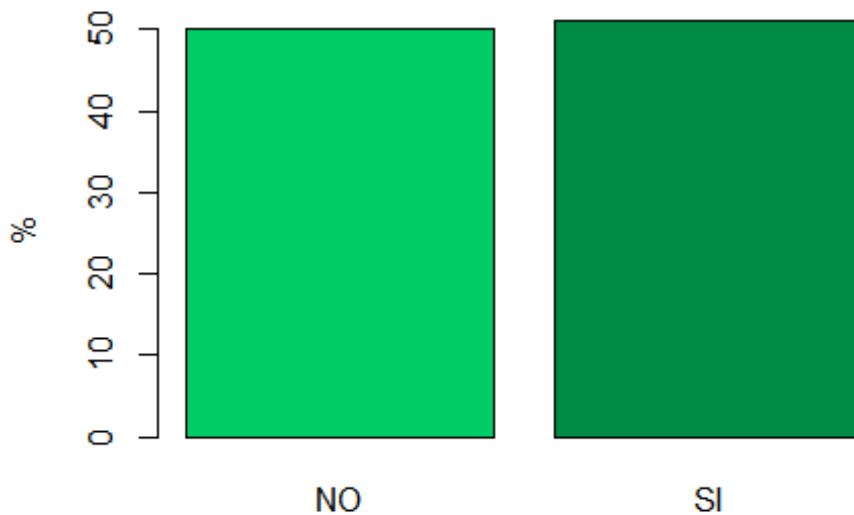
```
plot(x=base_rx$Mind_2_Synd_FL, col=c("slategray", "slategray4"), ylab="%", main="TRATAMIENTO BIOLÓGICO AL INICIO DEL ESTUDIO")
```

TRATAMIENTO BIOLÓGICO AL INICIO DEL ESTUDIO



```
plot(x=base_rx$Mind_2_Synd_BL, col=c("springgreen3", "springgreen4"), ylab="%", main="TRATAMIENTO BIOLÓGICO AL INICIO DEL ESTUDIO")
```

TRATAMIENTO BIOLÓGICO AL INICIO DEL ESTUDIO

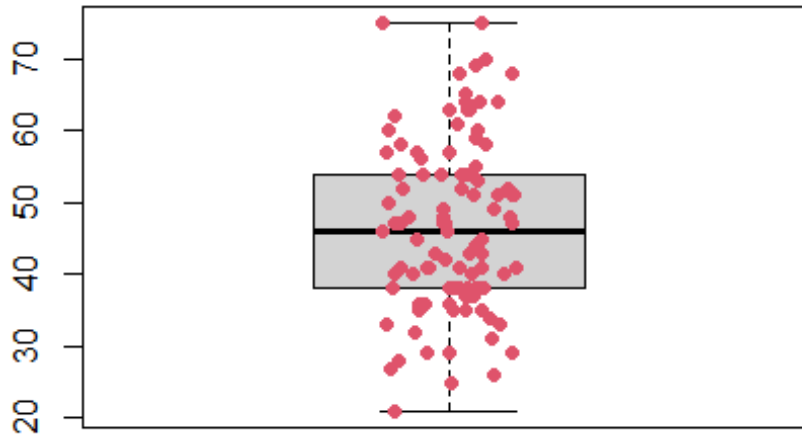


Análisis de outliers. Variables numéricas

Con la función stripchart observamos la dispersión de los datos

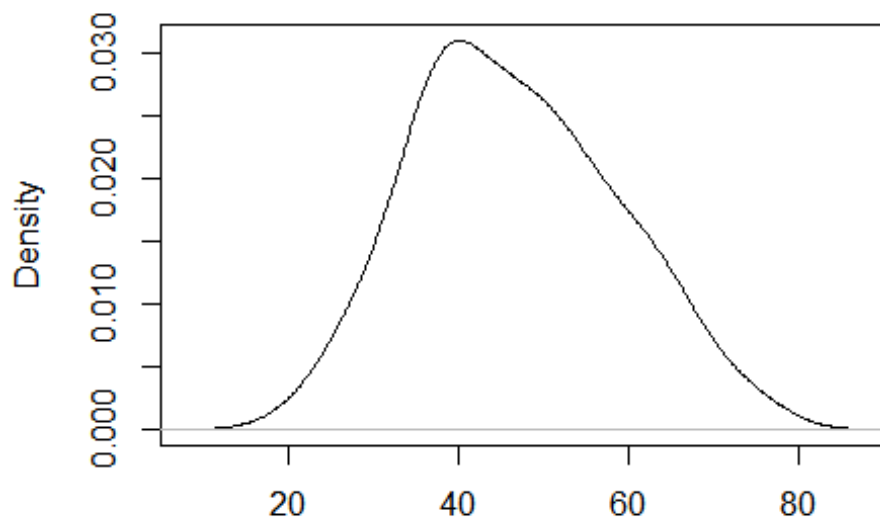
```
boxplot(x=base_rx$EDAD, main = "Boxplot: edad al inicio del estudio")
stripchart(x=base_rx$EDAD, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 2)
```

Boxplot: edad al inicio del estudio



```
plot(density(base_rx$EDAD), main = "Densidad de la variable edad")
```

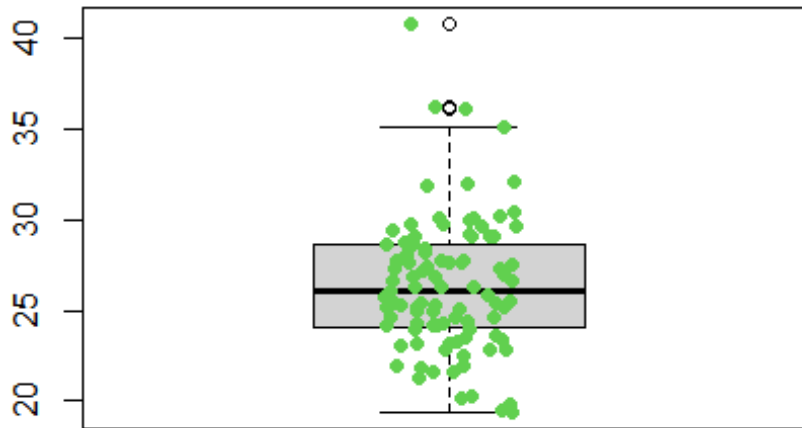
Densidad de la variable edad



N = 101 Bandwidth = 4.264

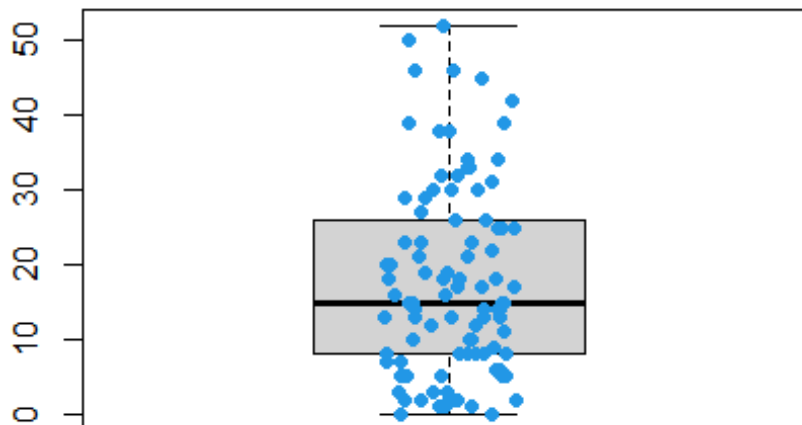
```
boxplot(x=base_rx$IMC, main = "Boxplot IMC al inicio del estudio")
stripchart(x=base_rx$IMC, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 3)
```

Boxplot IMC al inicio del estudio



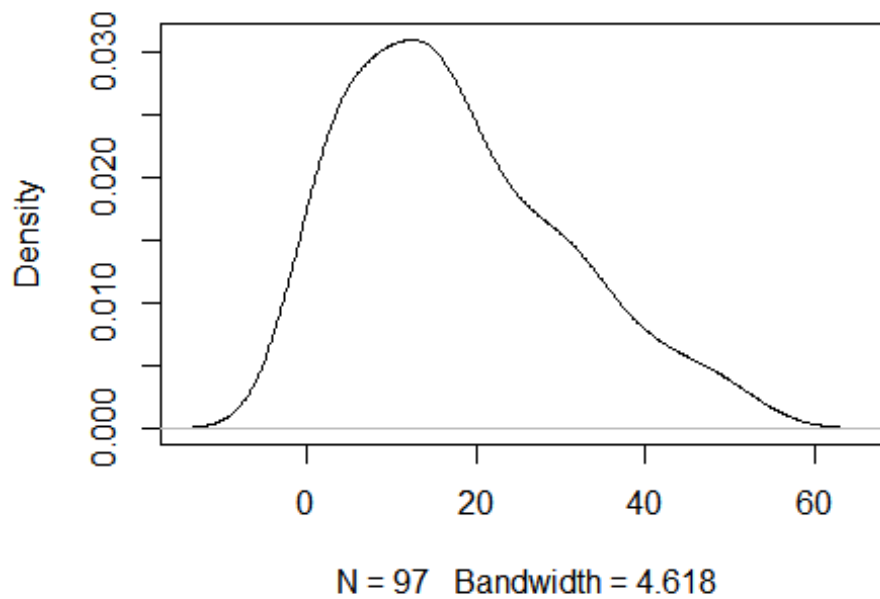
```
boxplot(x=base_rx$TSINT, main = "Boxplot: tiempo con tto biolÃ³gico al inicio del estudio")  
stripchart(x=base_rx$TSINT, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 4)
```

Boxplot: tiempo con tto biolÃ³gico al inicio del estudio



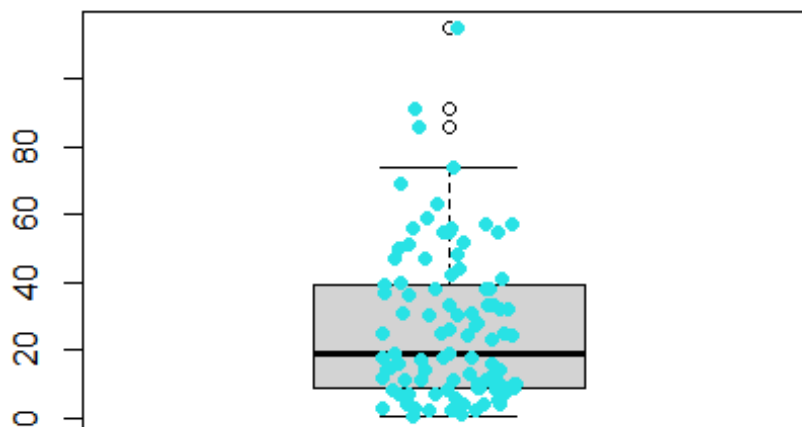
```
plot(density(base_rx$TSINT, na.rm = T), main = "Densidad de la variable tiempo con tto biolÃ³gico al inicio del estudio")
```

dad de la variable tiempo con tto biol³gico al inicio d



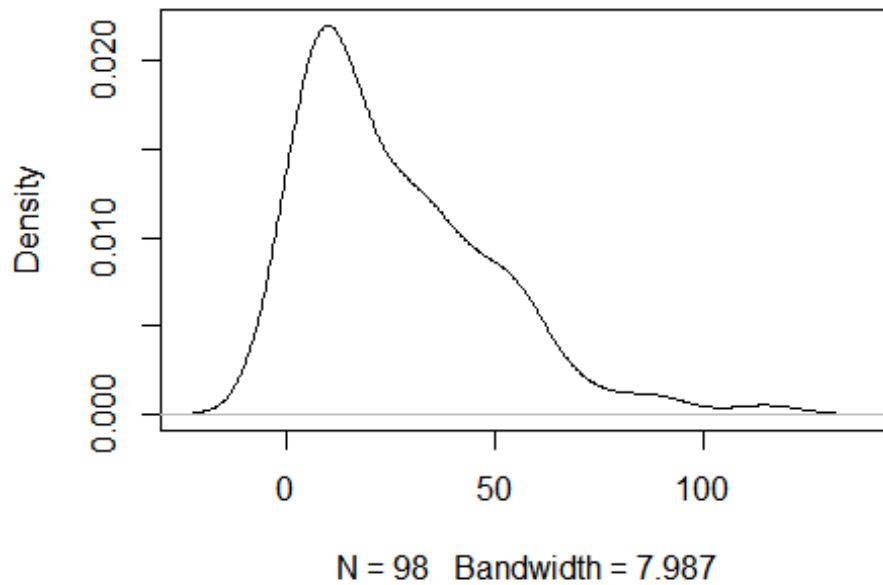
```
boxplot(x=base_rx$VSG, main = "Boxplot VSG al inicio del estudio")
stripchart(x=base_rx$VSG, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 5)
```

Boxplot VSG al inicio del estudio



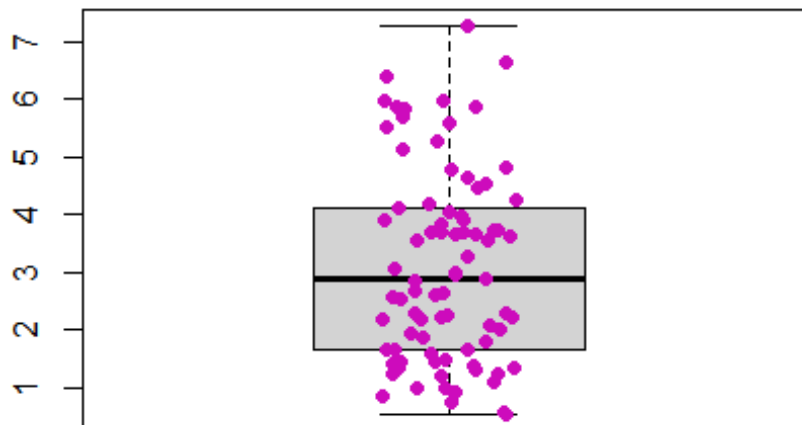
```
plot(density(base_rx$VSG, na.rm = T), main = "Densidad de la variable VSG al inicio del estudio")
```


Densidad de la variable VSG al inicio del estudio



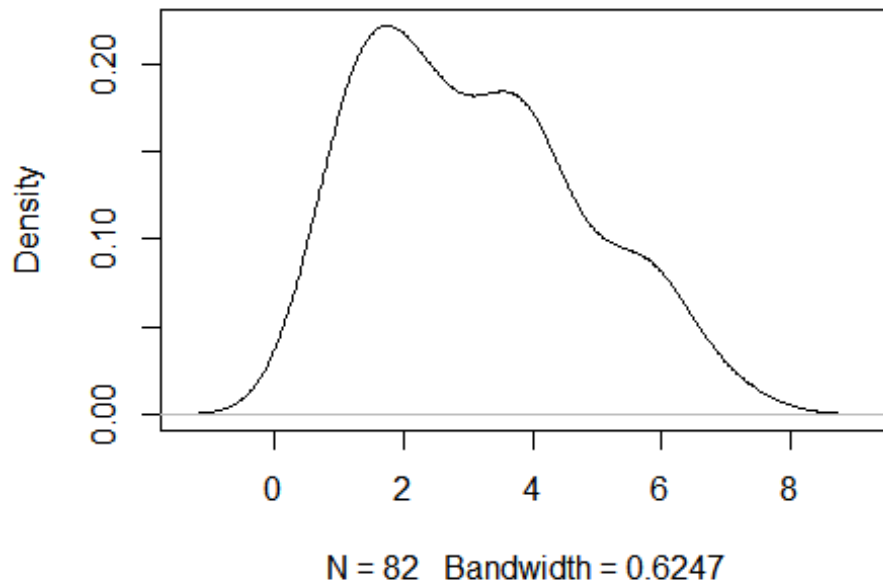
```
boxplot(x=base_rx$BASMI, main = "Boxplot BASMI al inicio del estudio")  
stripchart(x=base_rx$BASMI, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 6)
```

Boxplot BASMI al inicio del estudio



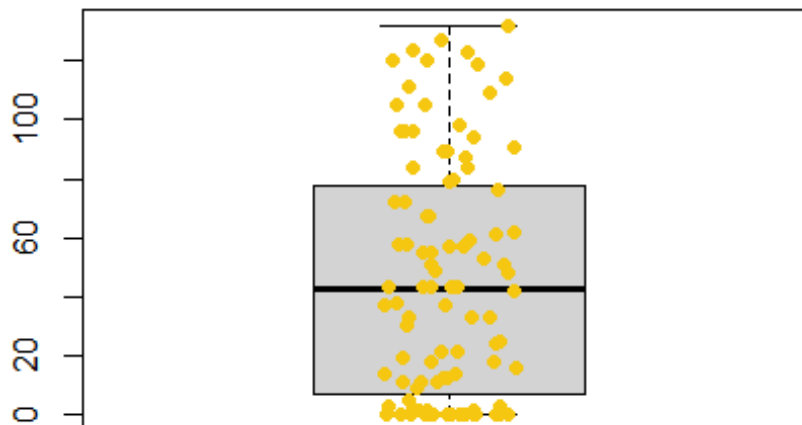
```
plot(density(base_rx$BASMI, na.rm = T), main = "Densidad de la variable BASMI al inicio del estudio")
```

Densidad de la variable BASMI al inicio del estudio



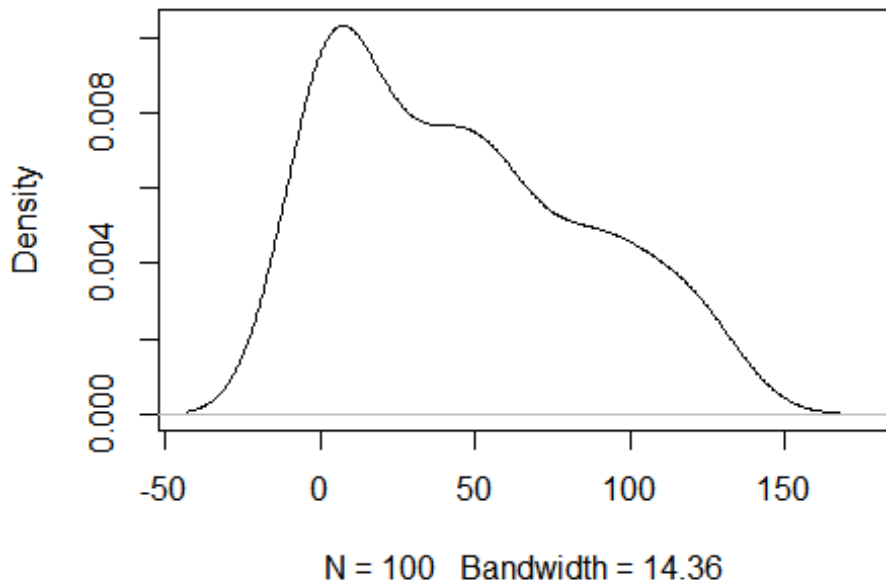
```
boxplot(x=base_rx$TIEMBIO, main = "Boxplot tiempo en tto biolÃ³gico al inicio del estudio")  
stripchart(x=base_rx$TIEMBIO, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 7)
```

Boxplot tiempo en tto biolÃ³gico al inicio del estudio



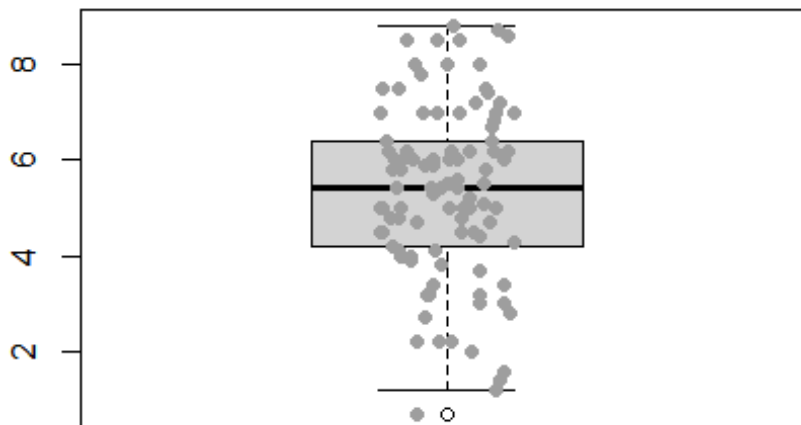
```
plot(density(base_rx$TIEMBIO, na.rm = T), main = "Densidad de la variable tiempo en tto biolÃ³gico al inicio del estudio")
```

idad de la variable tiempo en tto biol³gico al inicio de



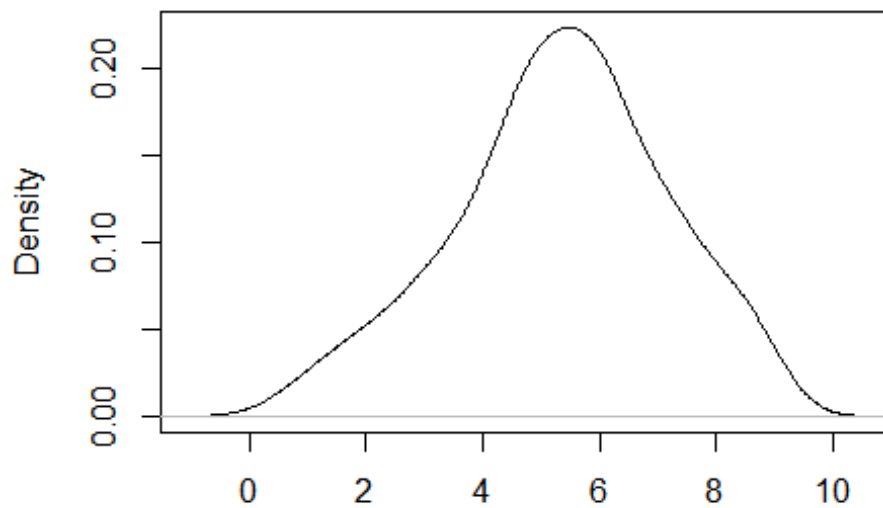
```
boxplot(x=base_rx$i_BASDAI, main = "Boxplot BASDAI al empezar tto biol3gico")  
stripchart(x=base_rx$i_BASDAI, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI al empezar tto biol³gico



```
plot(density(base_rx$i_BASDAI, na.rm = T), main = "Densidad de la variable BASDAI al empezar  
tto biol3gico")
```

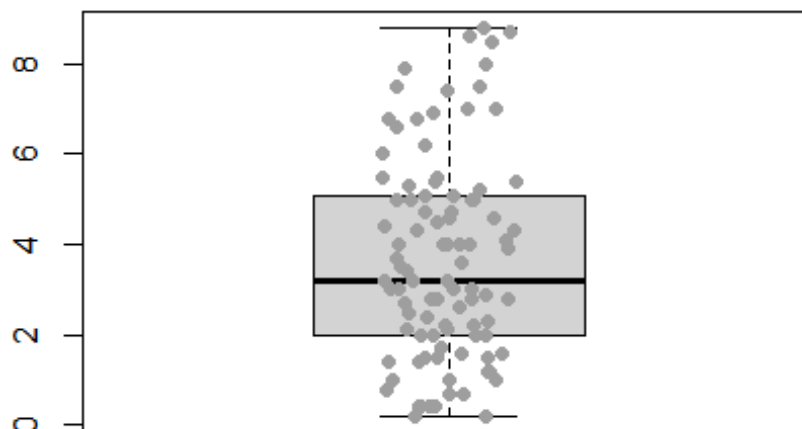
Densidad de la variable BASDAI al empezar tto bioIÃ³



N = 98 Bandwidth = 0.5839

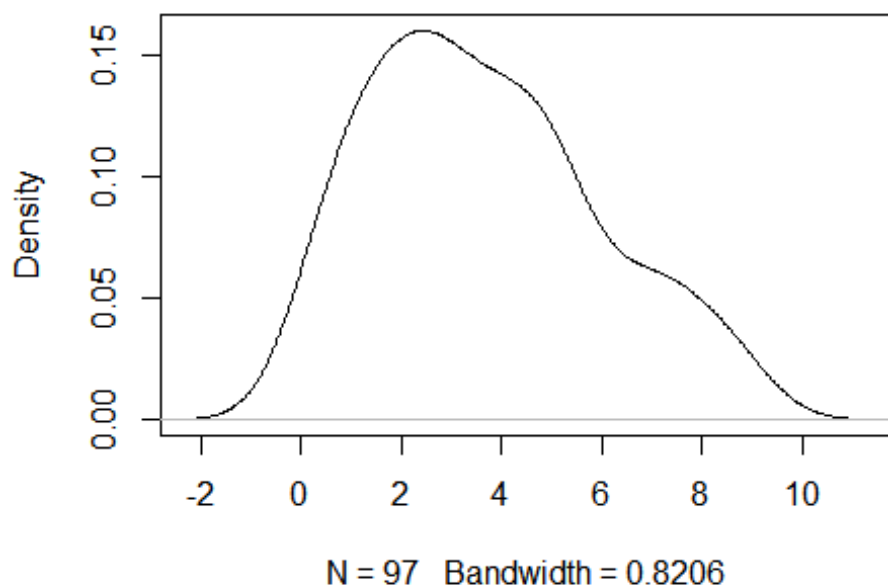
```
boxplot(x=base_rx$BASDAI_0, main = "Boxplot: BASDAI al empezar el estudio")  
stripchart(x=base_rx$BASDAI_0, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot: BASDAI al empezar el estudio



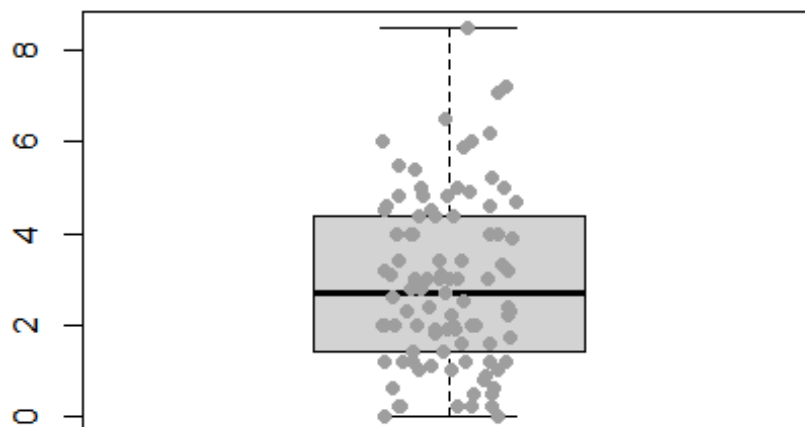
```
plot(density(base_rx$BASDAI_0, na.rm = T), main = "Densidad de la variable BASDAI al empezar el estudio")
```

Densidad de la variable BASDAI al empezar el estudio



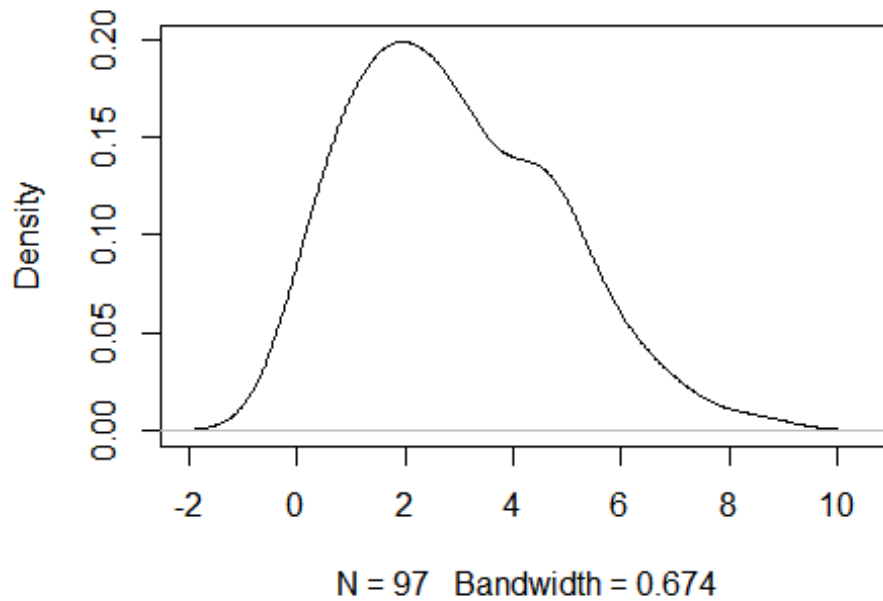
```
boxplot(x=base_rx$BASDAI_1, main = "Boxplot: BASDAI a los 6 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_1, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot: BASDAI a los 6 meses de empezar el estudio



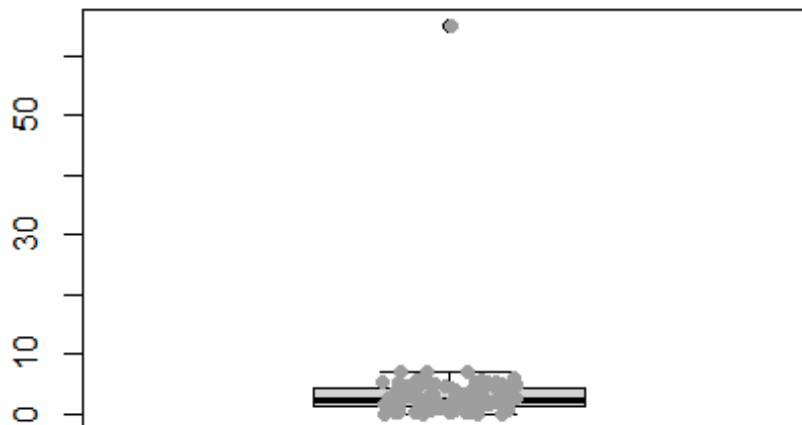
```
plot(density(base_rx$BASDAI_1, na.rm = T), main = "Densidad de la variable BASDAI a los 6 me  
ses de empezar el estudio")
```

dad de la variable BASDAI a los 6 meses de empezar



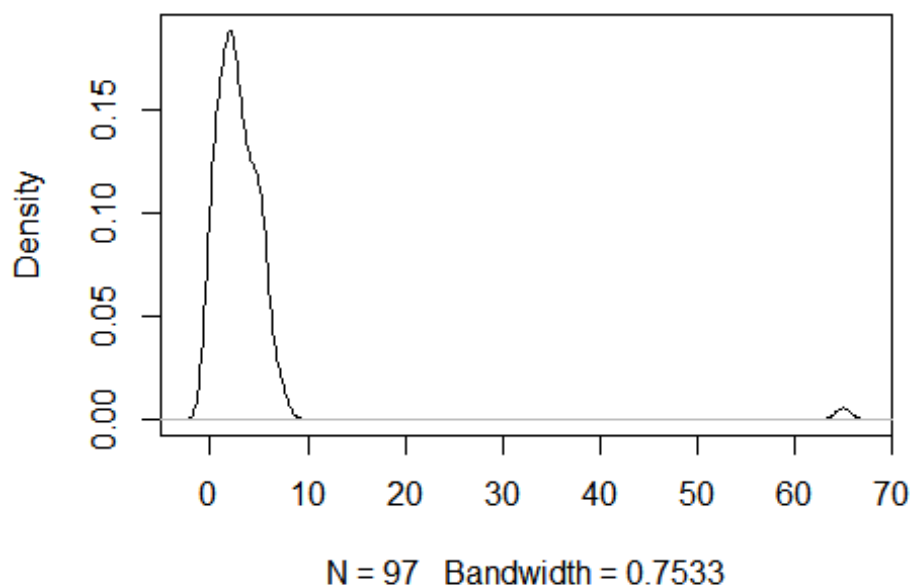
```
boxplot(x=base_rx$BASDAI_2, main = "Boxplot BASDAI a los 12 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_2, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI a los 12 meses de empezar el estu



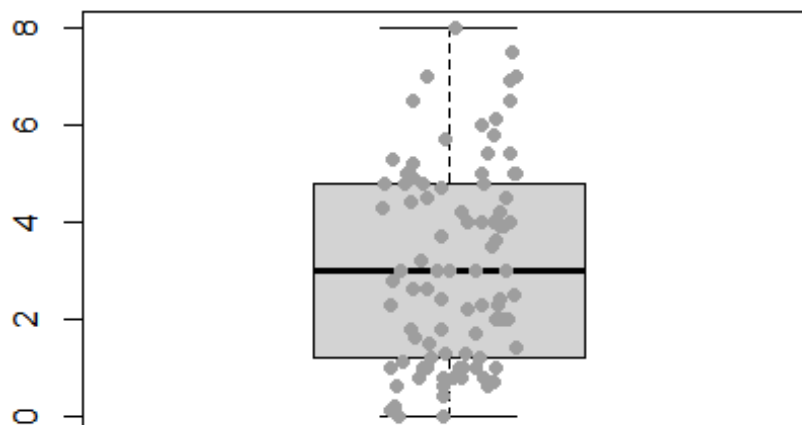
```
plot(density(base_rx$BASDAI_2, na.rm = T), main = "Densidad BASDAI a los 12 meses de empezar el estudio")
```

Densidad BASDAI a los 12 meses de empezar el estu



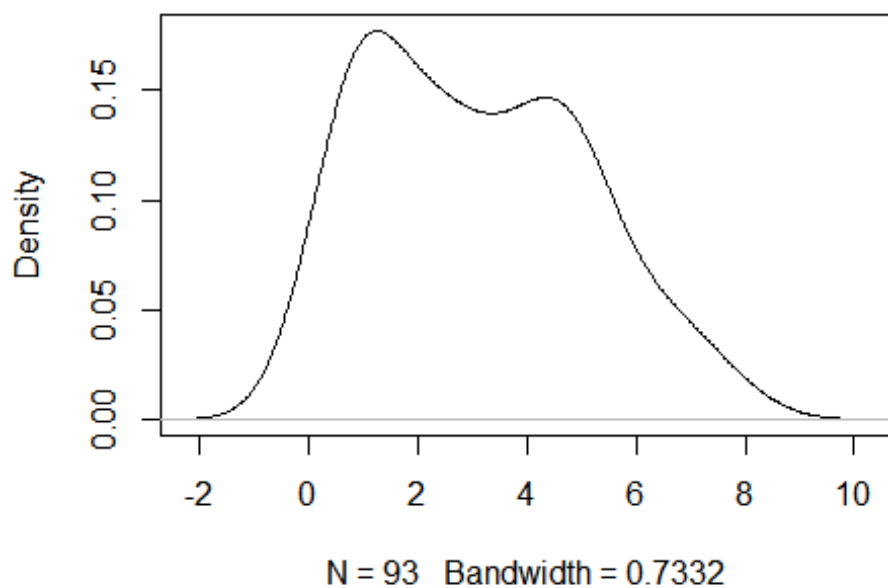
```
boxplot(x=base_rx$BASDAI_3, main = "Boxplot BASDAI a los 18 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_3, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI a los 18 meses de empezar el estu



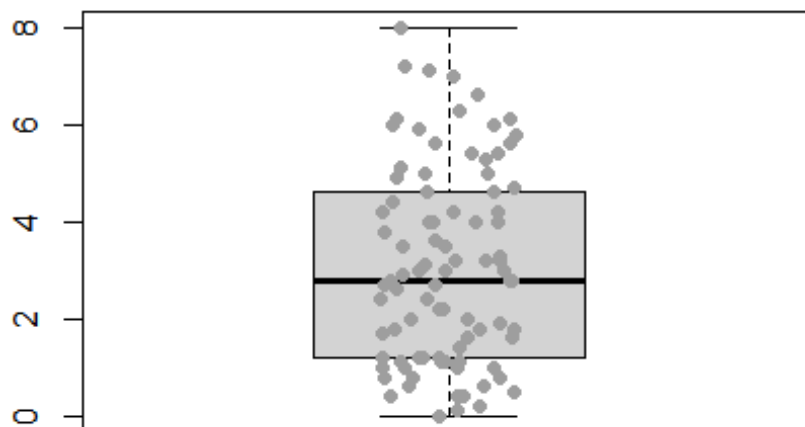
```
plot(density(base_rx$BASDAI_3, na.rm = T), main = "Densidad de la variable BASDAI a los 18 m  
eses de empezar el estudio")
```

Densidad de la variable BASDAI a los 18 meses de empezar el estudio



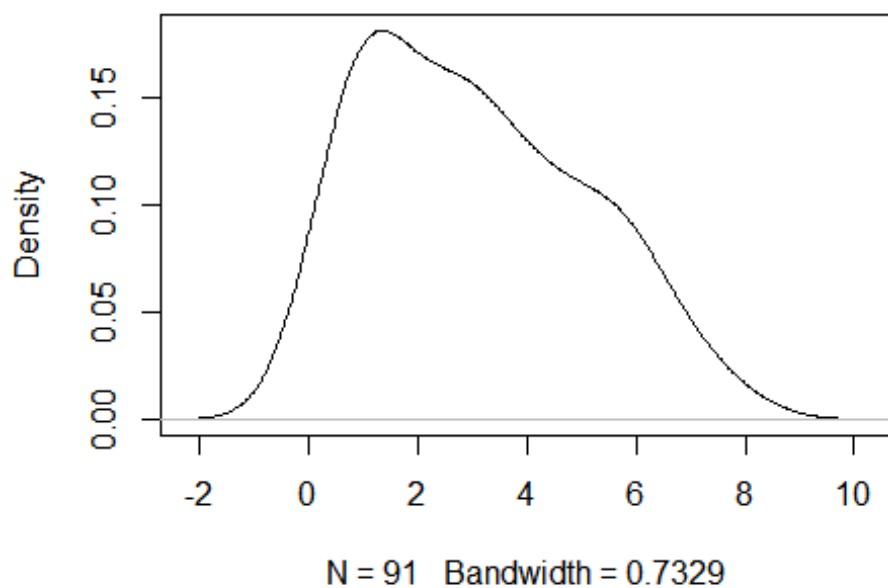
```
boxplot(x=base_rx$BASDAI_4, main = "Boxplot BASDAI a los 24 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_4, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI a los 24 meses de empezar el estudio



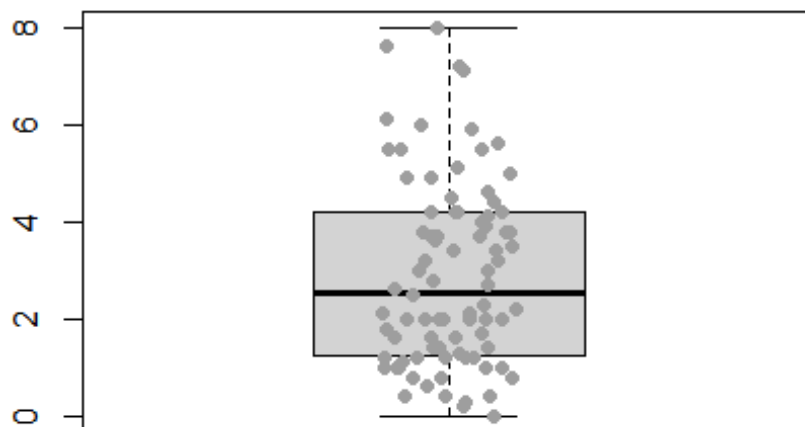
```
plot(density(base_rx$BASDAI_4, na.rm = T), main = "Densidad de la variable BASDAI a los 24 meses de empezar el estudio")
```


Densidad de la variable BASDAI a los 24 meses de empezar el estudio



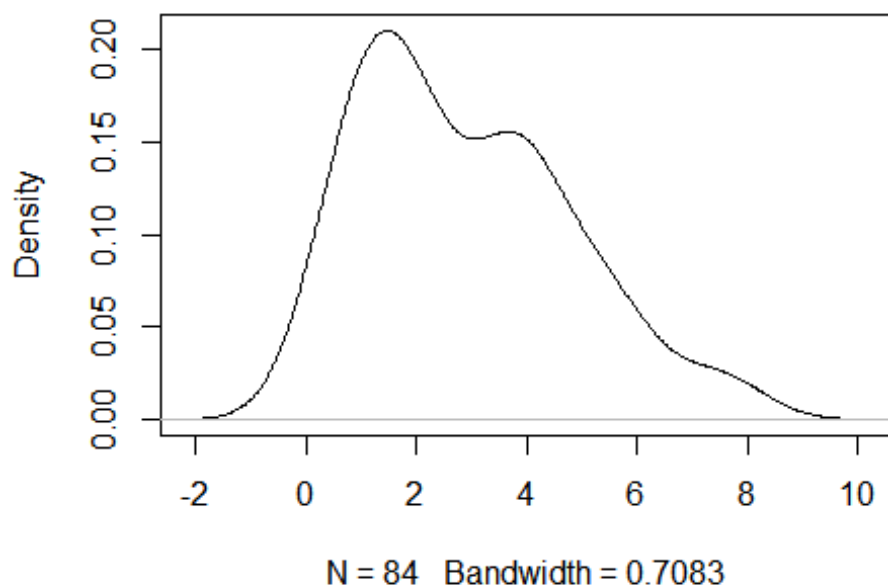
```
boxplot(x=base_rx$BASDAI_5, main = "Boxplot BASDAI a los 30 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_5, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI a los 30 meses de empezar el estudio



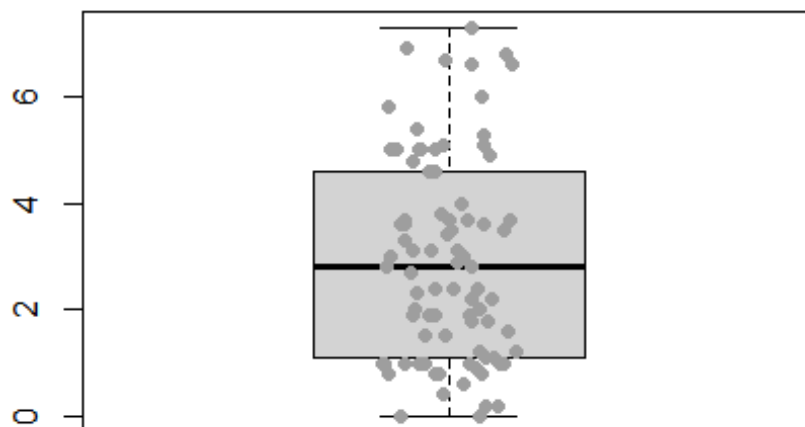
```
plot(density(base_rx$BASDAI_5, na.rm = T), main = "Densidad de la variable BASDAI a los 30 meses de empezar el estudio")
```

Densidad de la variable BASDAI a los 30 meses de empezar el estudio



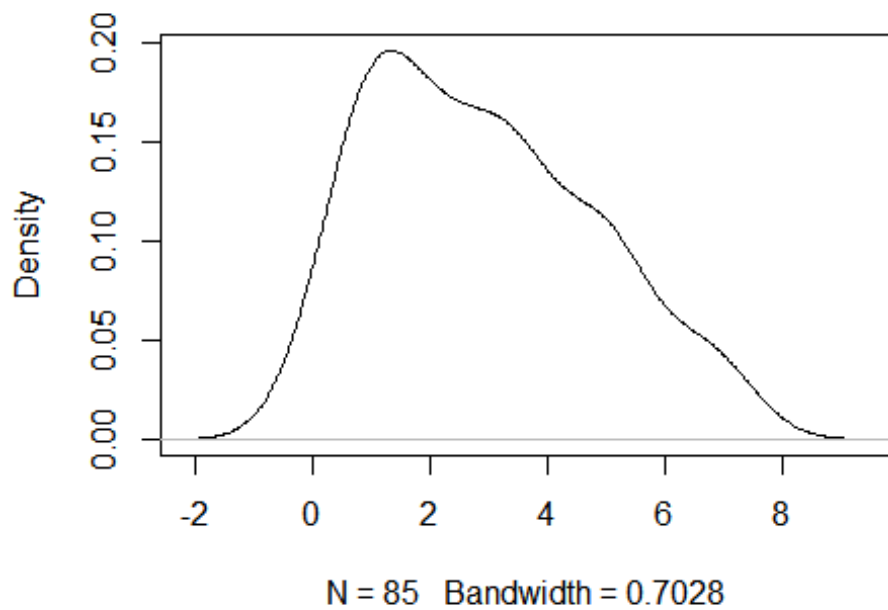
```
boxplot(x=base_rx$BASDAI_6, main = "Boxplot BASDAI a los 36 meses de empezar el estudio")  
stripchart(x=base_rx$BASDAI_6, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 8)
```

Boxplot BASDAI a los 36 meses de empezar el estudio



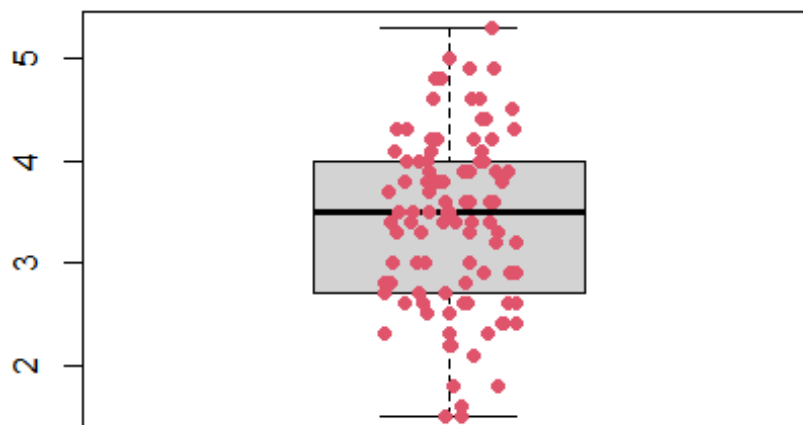
```
plot(density(base_rx$BASDAI_6, na.rm = T), main = "Densidad de la variable BASDAI a los 36 meses de empezar el estudio")
```

Densidad de la variable BASDAI a los 36 meses de empezar



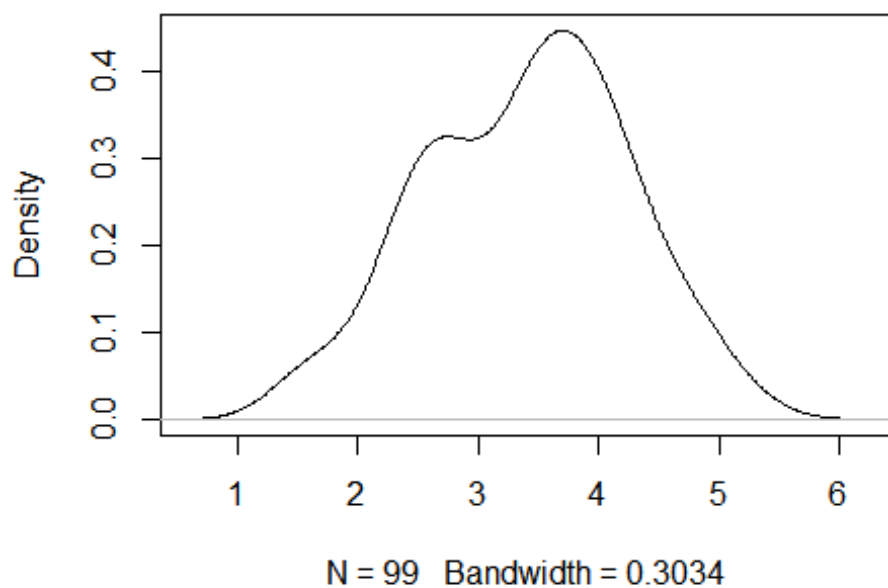
```
boxplot(x=base_rx$i_ASDAS, main = "Boxplot: ASDAS al empezar el tto biológico")  
stripchart(x=base_rx$i_ASDAS, method = "jitter", pch=16, add=TRUE, vertical=TRUE, col = 10)
```

Boxplot: ASDAS al empezar el tto biológico



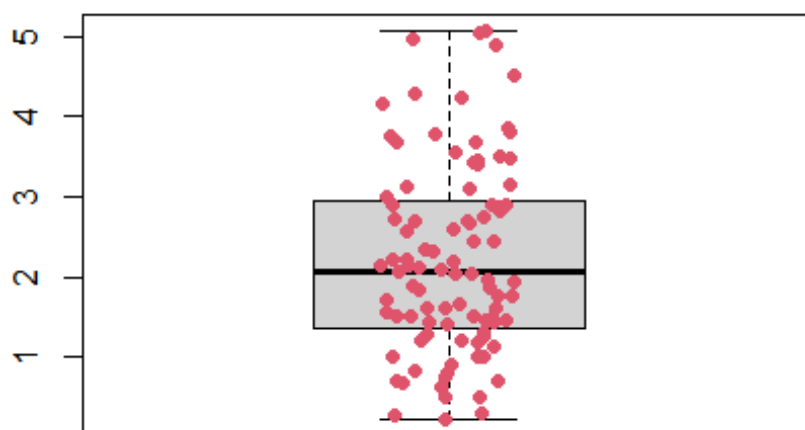
```
plot(density(base_rx$i_ASDAS, na.rm = T), main = "Densidad de la variable ASDAS al empezar  
el tto biológico")
```

Densidad de la variable ASDAS al empezar el tto bioIÃ



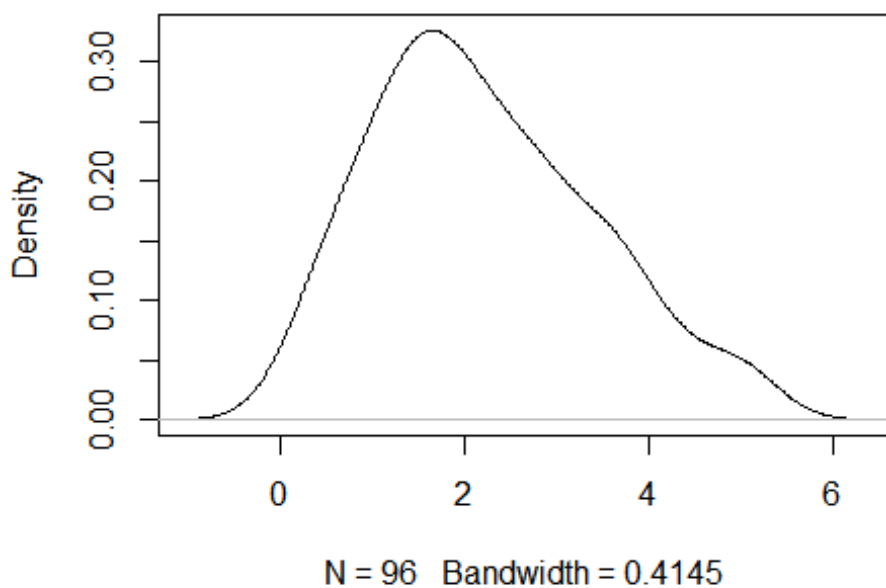
```
boxplot(x=base_rx$ASDAS_0, main = "Boxplot: ASDAS al empezar el estudio")  
stripchart(x=base_rx$ASDAS_0, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10  
)
```

Boxplot: ASDAS al empezar el estudio



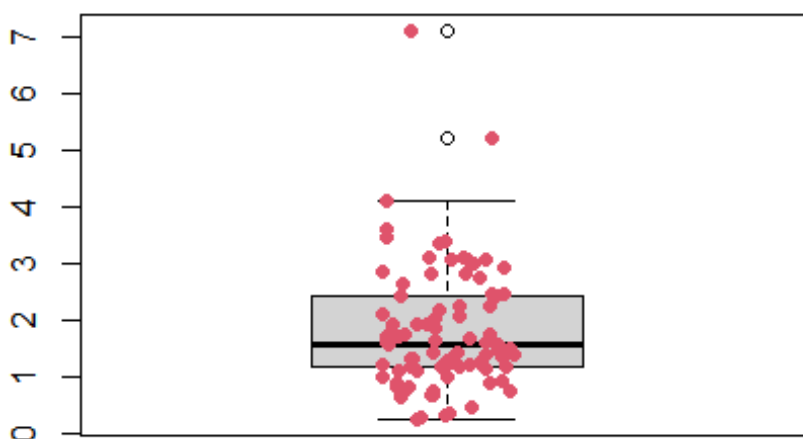
```
plot(density(base_rx$ASDAS_0, na.rm = T), main = "Densidad de la variable ASDAS al empezar  
el estudio")
```

Densidad de la variable ASDAS al empezar el estudio



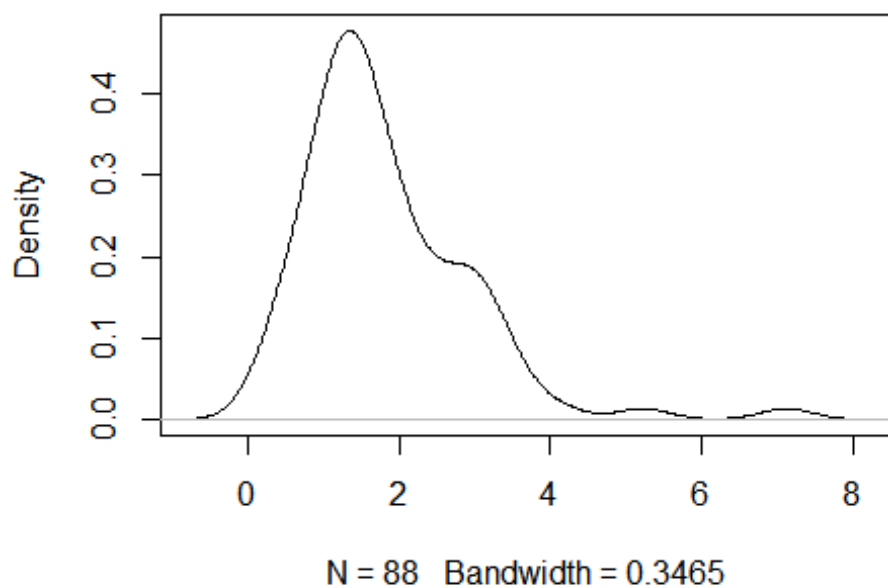
```
boxplot(x=base_rx$ASDAS_1, main = "Boxplot: ASDAS a los 6 meses de empezar estudio")
stripchart(x=base_rx$ASDAS_1, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10
)
```

Boxplot: ASDAS a los 6 meses de empezar estudio



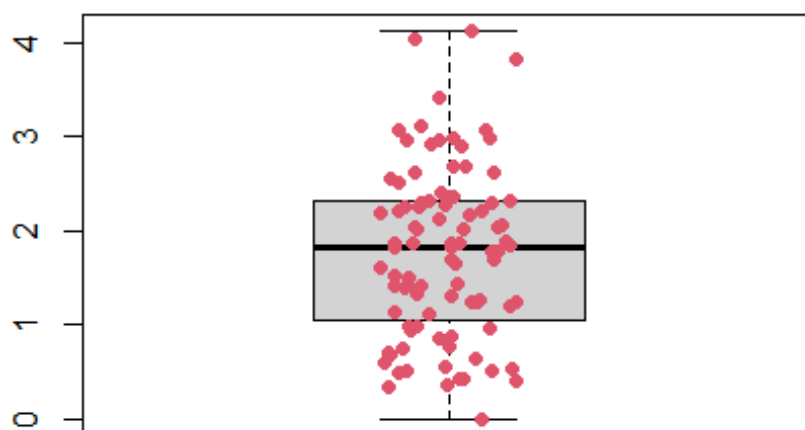
```
plot(density(base_rx$ASDAS_1, na.rm = T), main = "Densidad de la variable ASDAS a los 6 mes  
es de empezar el estudio")
```

Densidad de la variable ASDAS a los 6 meses de empezar



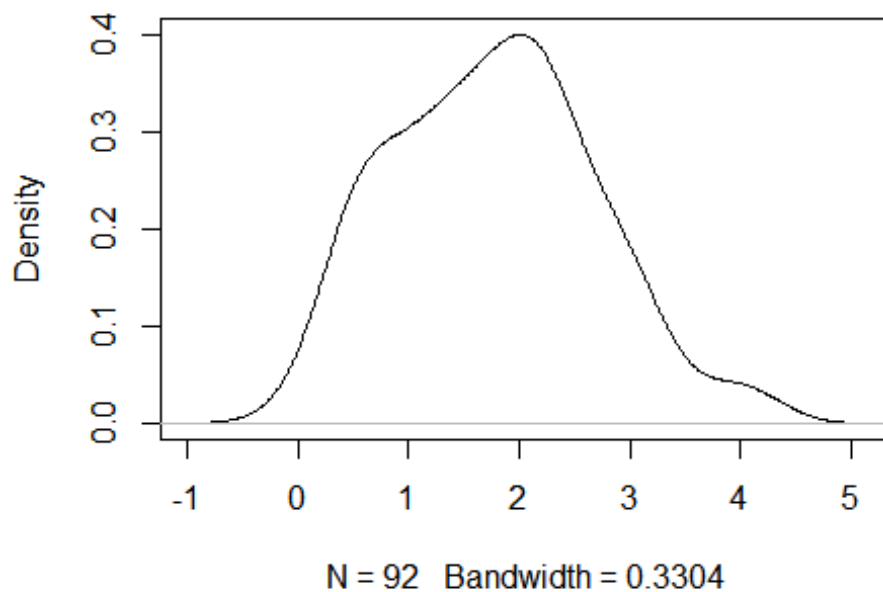
```
boxplot(x=base_rx$ASDAS_2, main = "Boxplot: ASDAS a los 12 meses de empezar estudio")  
stripchart(x=base_rx$ASDAS_2, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10  
)
```

Boxplot: ASDAS a los 12 meses de empezar estud



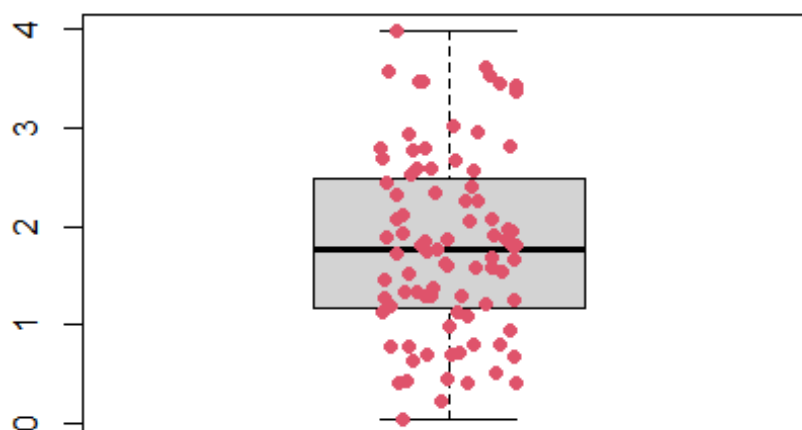
```
plot(density(base_rx$ASDAS_2, na.rm = T), main = "Densidad de la variable ASDAS a los 12 me  
ses de empezar el estudio")
```

Densidad de la variable ASDAS a los 12 meses de empezar



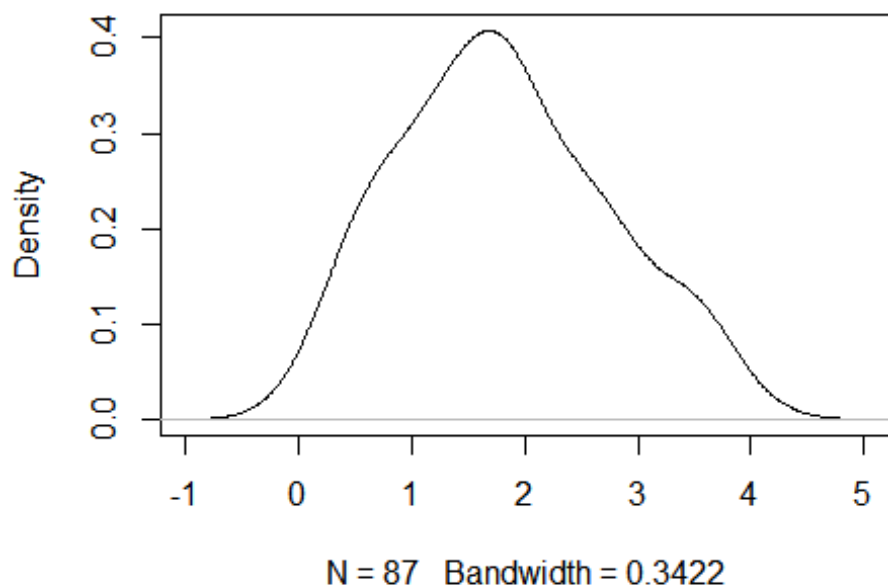
```
boxplot(x=base_rx$ASDAS_3, main = "Boxplot: ASDAS a los 18 meses de empezar estudio")  
stripchart(x=base_rx$ASDAS_3, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10  
)
```

Boxplot: ASDAS a los 18 meses de empezar estudio



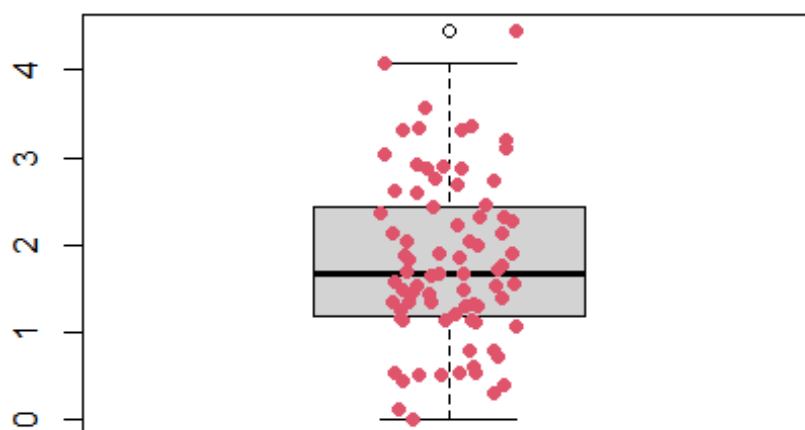
```
plot(density(base_rx$ASDAS_3, na.rm = T), main = "Densidad de la variable ASDAS a los 18 me  
ses de empezar el estudio")
```

Densidad de la variable ASDAS a los 18 meses de empezar



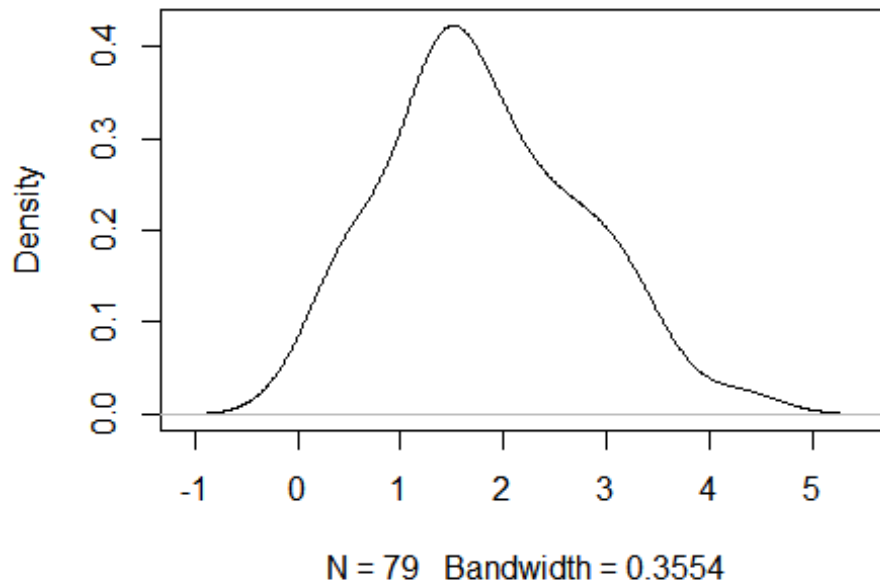
```
boxplot(x=base_rx$ASDAS_4, main = "Boxplot: ASDAS a los 24 meses de empezar estudio")
stripchart(x=base_rx$ASDAS_4, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10
)
```

Boxplot: ASDAS a los 24 meses de empezar estudio



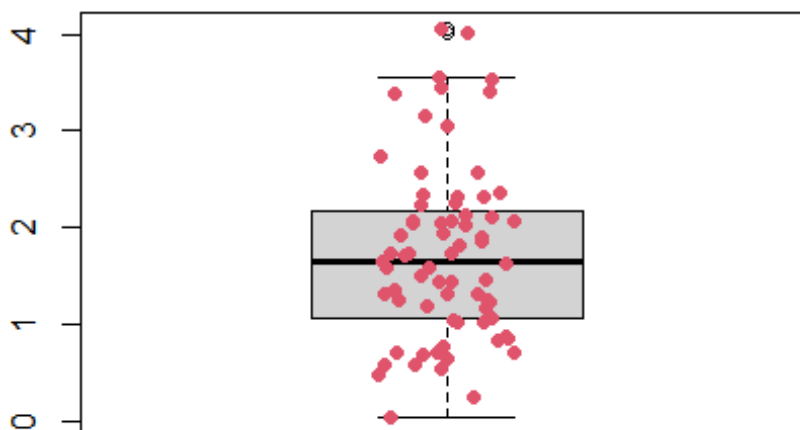
```
plot(density(base_rx$ASDAS_4, na.rm = T), main = "Densidad de la variable ASDAS a los 24 me  
ses de empezar el estudio")
```


Densidad de la variable ASDAS a los 24 meses de empezar



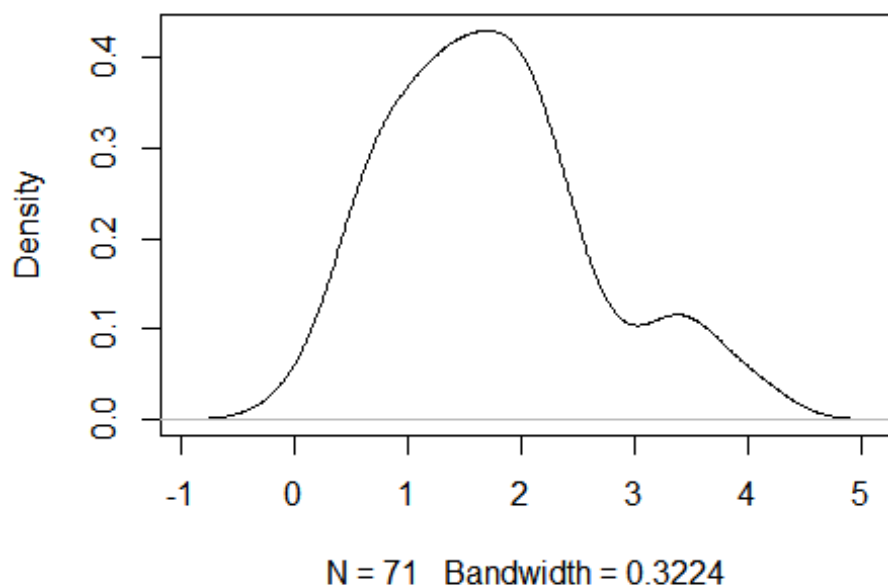
```
boxplot(x=base_rx$ASDAS_5, main = "Boxplot: ASDAS a los 30 meses de empezar estudio")
stripchart(x=base_rx$ASDAS_5, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10
)
```

Boxplot: ASDAS a los 30 meses de empezar estudio



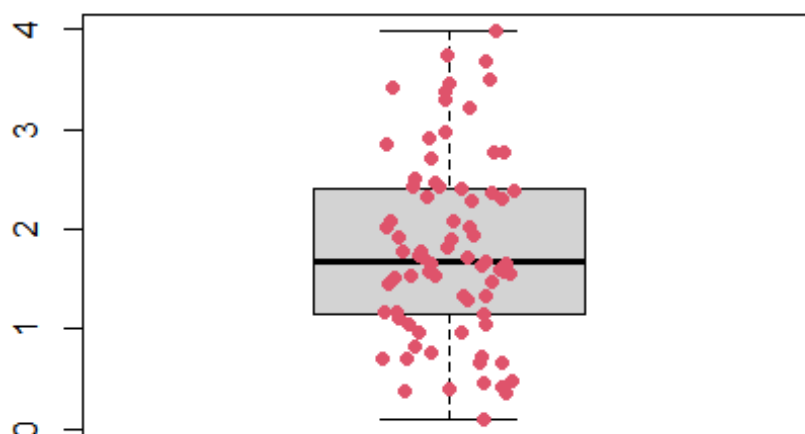
```
plot(density(base_rx$ASDAS_5, na.rm = T), main = "Densidad de la variable ASDAS a los 30 me  
ses de empezar el estudio")
```

Densidad de la variable ASDAS a los 30 meses de empezar



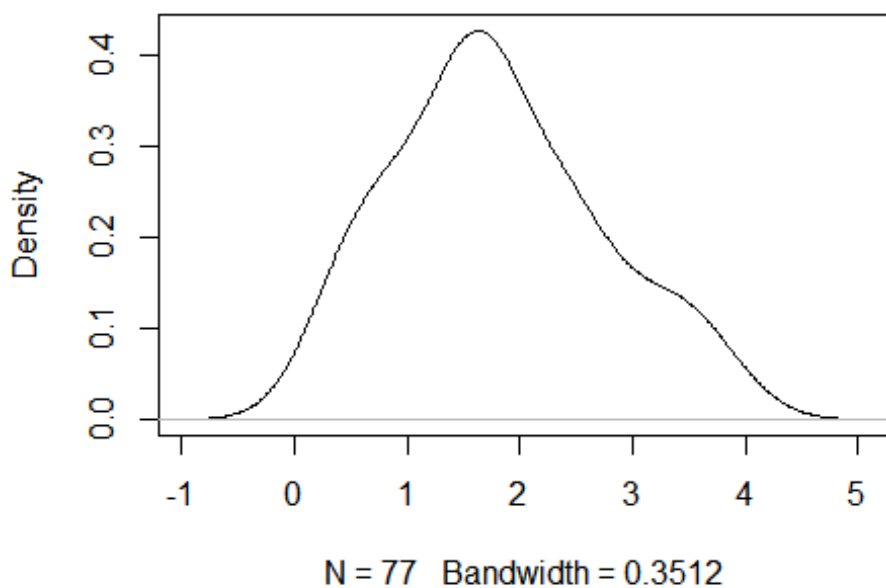
```
boxplot(x=base_rx$ASDAS_6, main = "Boxplot: ASDAS a los 36 meses de empezar estudio")  
stripchart(x=base_rx$ASDAS_6, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 10  
)
```

Boxplot: ASDAS a los 36 meses de empezar estudio



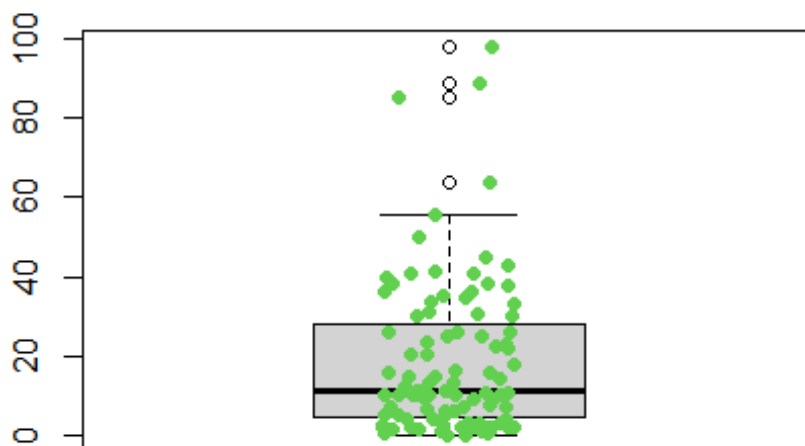
```
plot(density(base_rx$ASDAS_6, na.rm = T), main = "Densidad de la variable ASDAS a los 36 me  
ses de empezar el estudio")
```

Densidad de la variable ASDAS a los 36 meses de empezar



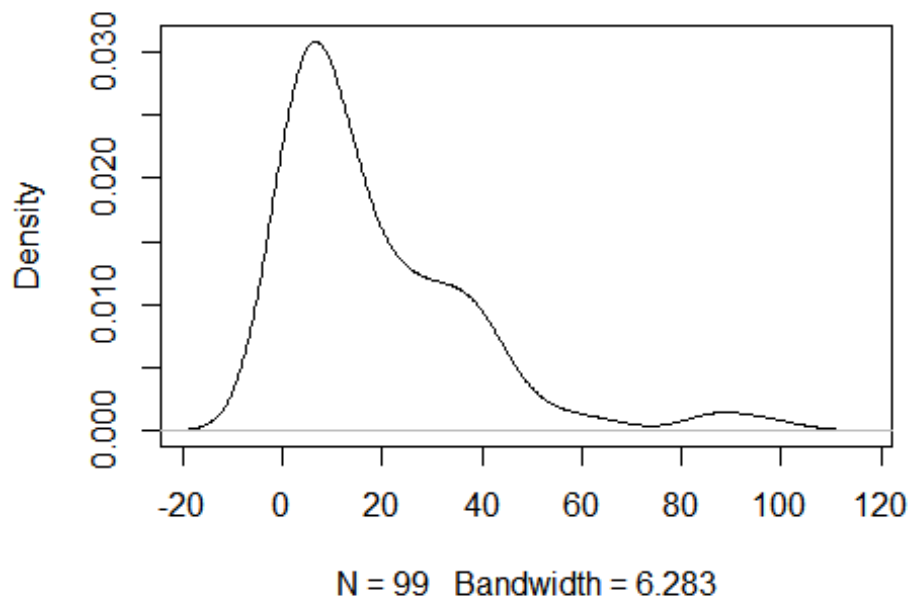
```
boxplot(x=base_rx$PCR_INICIOB, main = "Boxplot: PCR al empezar el tto biolÃ³gico")  
stripchart(x=base_rx$PCR_INICIOB, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col  
= 11)
```

Boxplot: PCR al empezar el tto biolÃ³gico



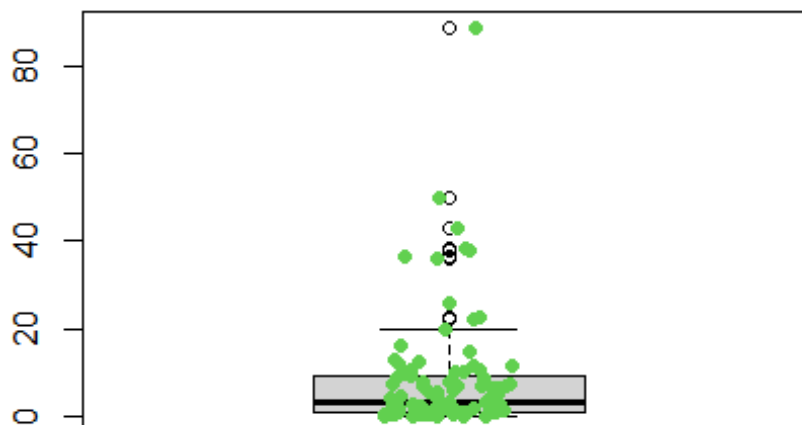
```
plot(density(base_rx$PCR_INICIOB, na.rm = T), main = "Densidad de la variable PCR al empeza  
r el tto biolÃ³gico")
```

Densidad de la variable PCR al empezar el tto bioI³



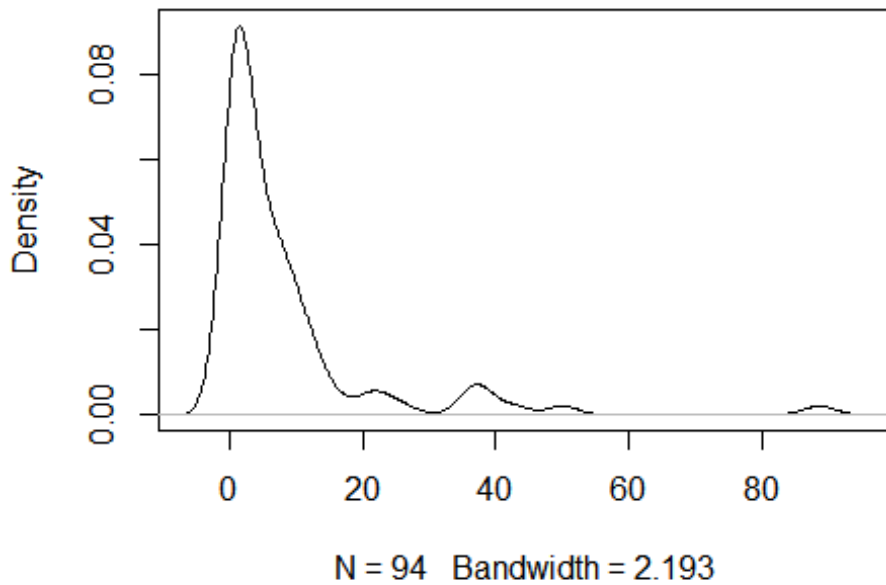
```
boxplot(x=base_rx$PCR_0, main = "Boxplot: PCR al empezar el estudio")  
stripchart(x=base_rx$PCR_0, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR al empezar el estudio



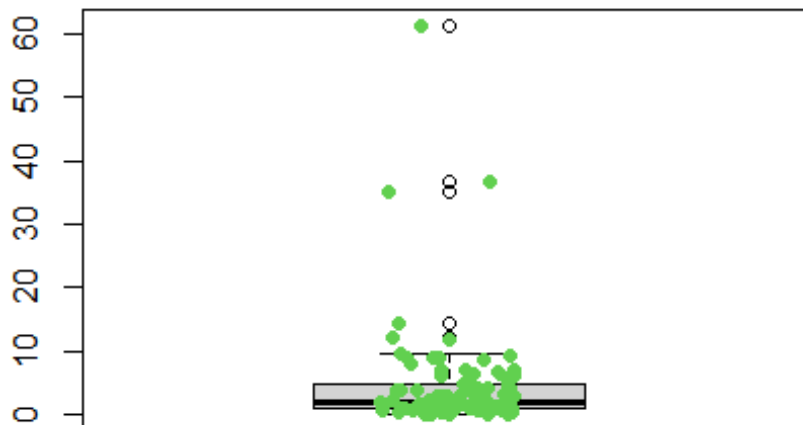
```
plot(density(base_rx$PCR_0, na.rm = T), main = "Densidad de la variable PCR al empezar el estudio")
```

Densidad de la variable PCR al empezar el estudio



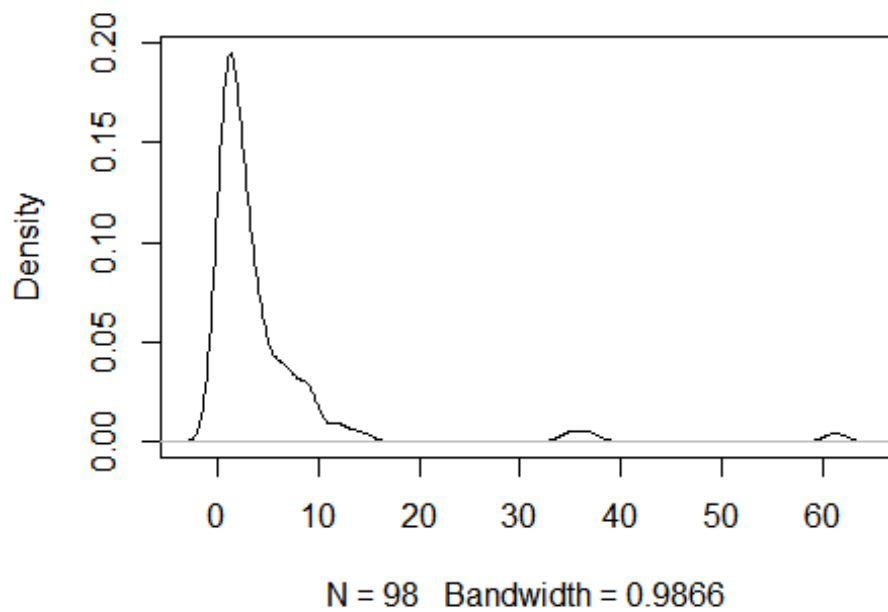
```
boxplot(x=base_rx$PCR_1, main = "Boxplot: PCR a los 6 meses de empezar estudio")  
stripchart(x=base_rx$PCR_1, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 6 meses de empezar estudio



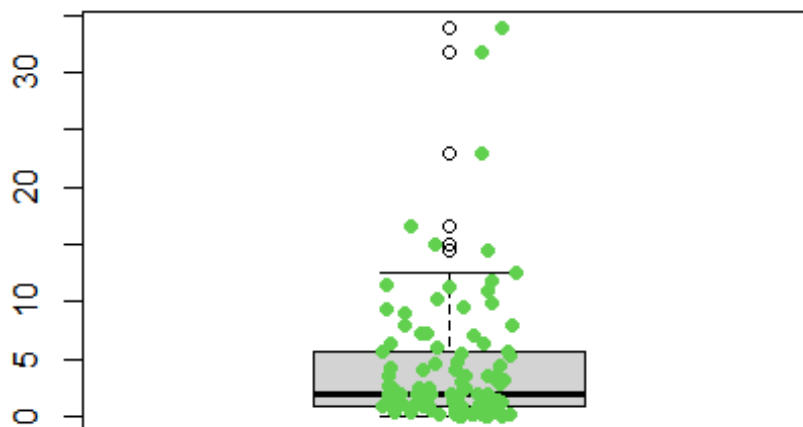
```
plot(density(base_rx$PCR_1, na.rm = T), main = "Densidad de la variable PCR a los 6 meses de  
empezar estudio")
```

Densidad de la variable PCR a los 6 meses de empezar



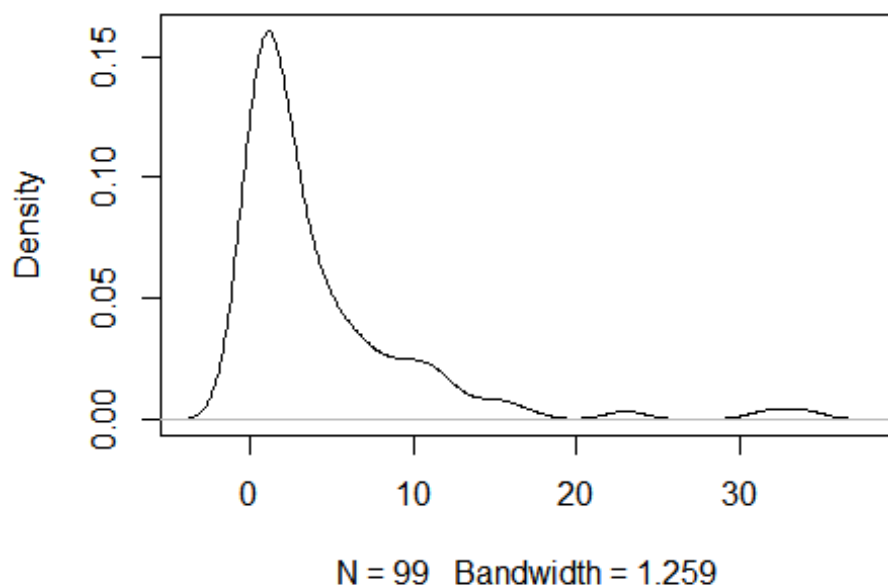
```
boxplot(x=base_rx$PCR_2, main = "Boxplot: PCR a los 12 meses de empezar estudio")  
stripchart(x=base_rx$PCR_2, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 12 meses de empezar estudio



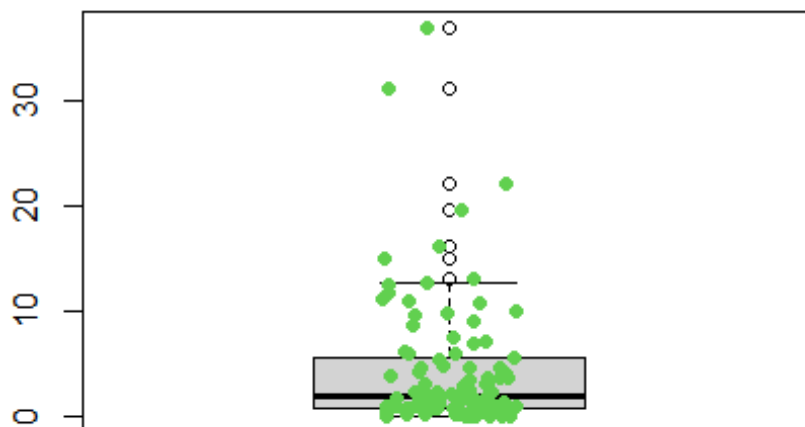
```
plot(density(base_rx$PCR_2, na.rm = T), main = "Densidad de la variable PCR a los 12 meses de empezar estudio")
```

Densidad de la variable PCR a los 12 meses de empezar



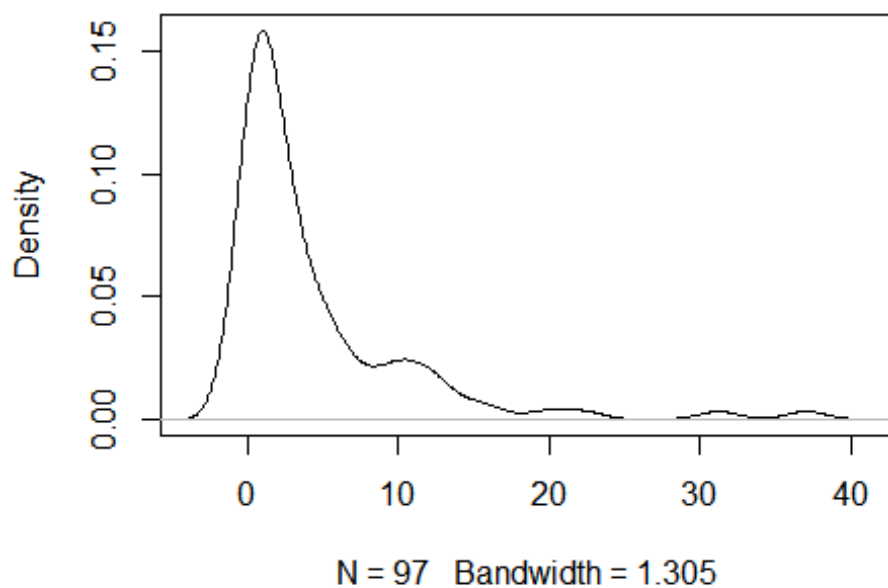
```
boxplot(x=base_rx$PCR_3, main = "Boxplot: PCR a los 18 meses de empezar estudio")  
stripchart(x=base_rx$PCR_3, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 18 meses de empezar estudio



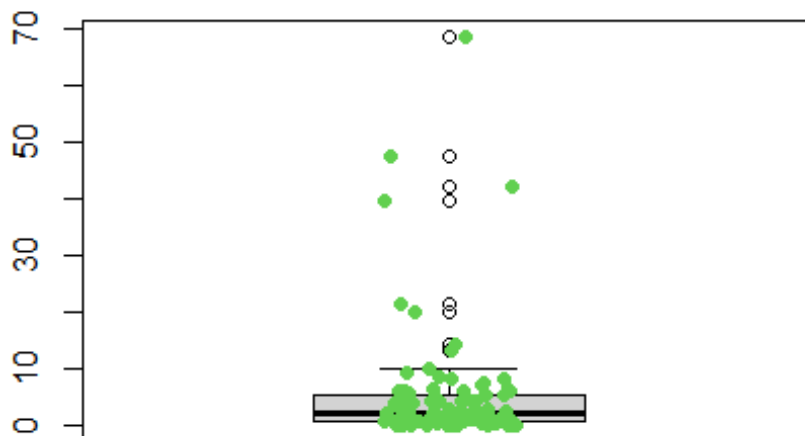
```
plot(density(base_rx$PCR_3, na.rm = T), main = "Densidad de la variable PCR a los 18 meses de empezar estudio")
```

densidad de la variable PCR a los 18 meses de empezar



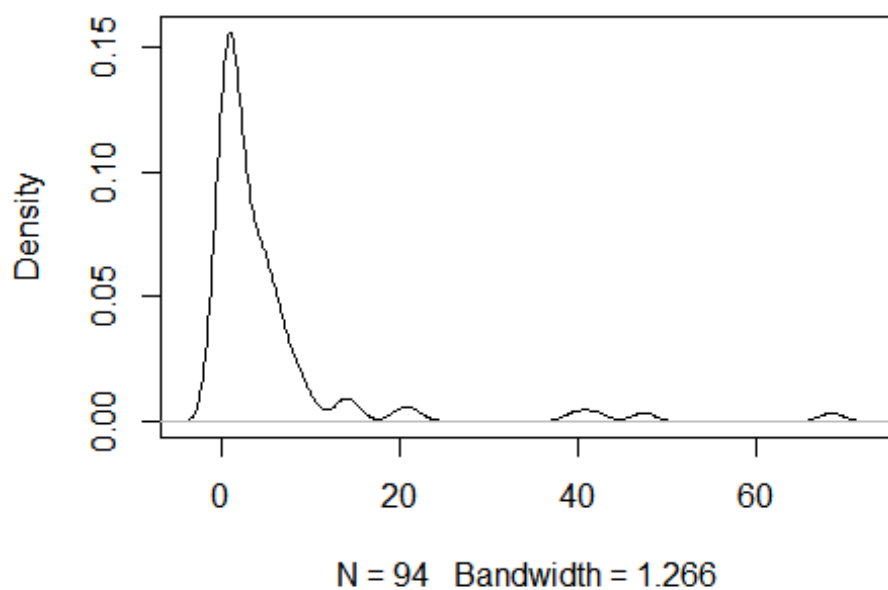
```
boxplot(x=base_rx$PCR_4, main = "Boxplot: PCR a los 24 meses de empezar estudio")
stripchart(x=base_rx$PCR_4, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 24 meses de empezar estudio



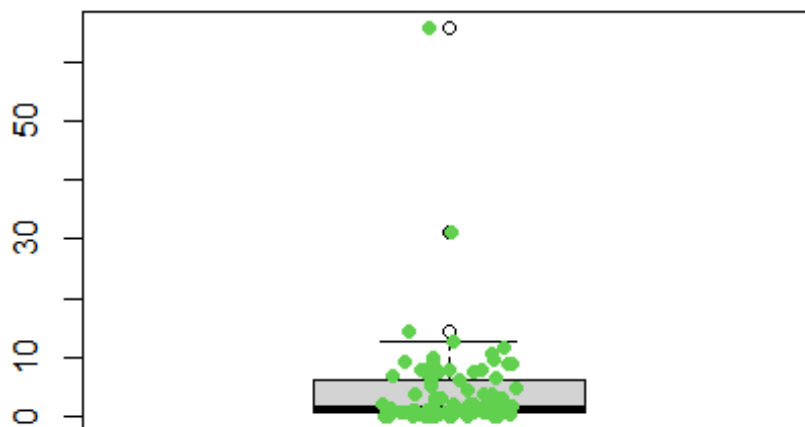
```
plot(density(base_rx$PCR_4, na.rm = T), main = "Densidad de la variable PCR a los 24 meses de empezar estudio")
```


densidad de la variable PCR a los 24 meses de empezar



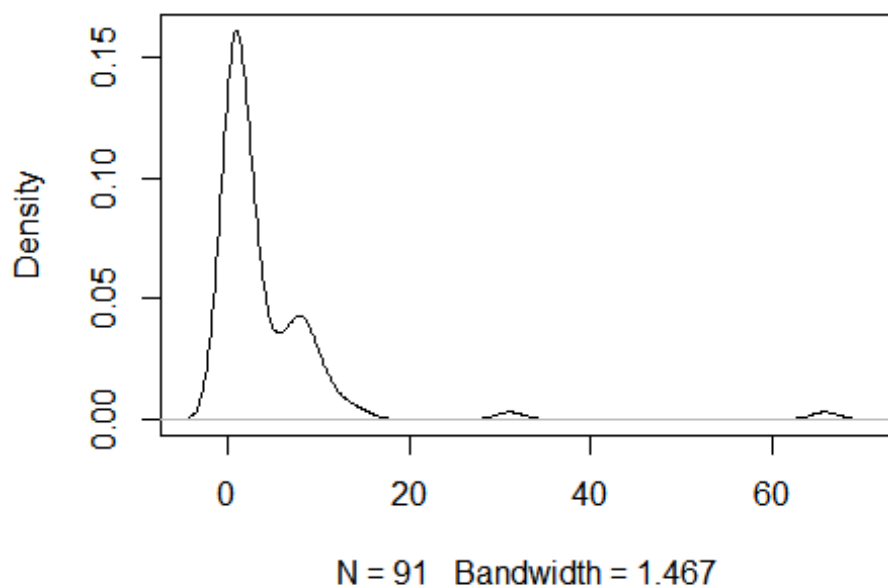
```
boxplot(x=base_rx$PCR_5, main = "Boxplot: PCR a los 30 meses de empezar estudio")  
stripchart(x=base_rx$PCR_5, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 30 meses de empezar estudio



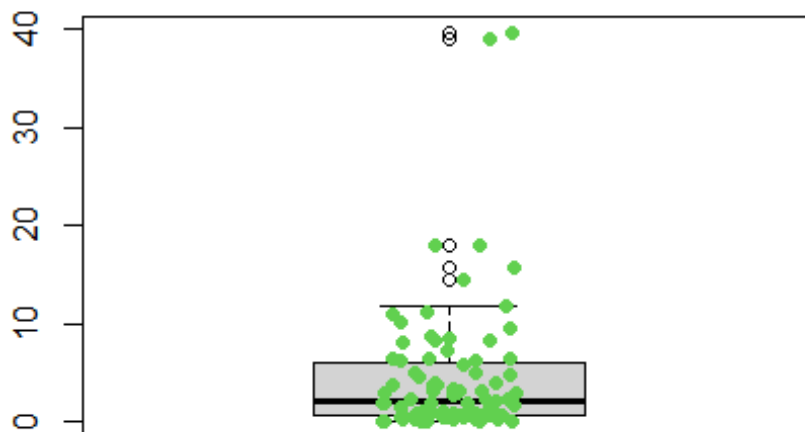
```
plot(density(base_rx$PCR_5, na.rm = T), main = "Densidad de la variable PCR a los 30 meses de empezar estudio")
```

densidad de la variable PCR a los 30 meses de empezar



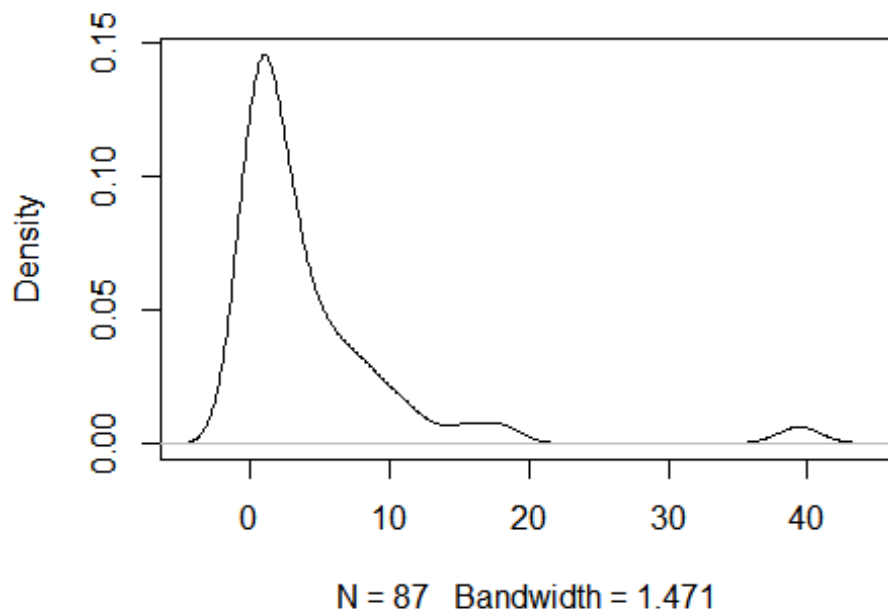
```
boxplot(x=base_rx$PCR_6, main = "Boxplot: PCR a los 36 meses de empezar estudio")  
stripchart(x=base_rx$PCR_6, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 11)
```

Boxplot: PCR a los 36 meses de empezar estudio



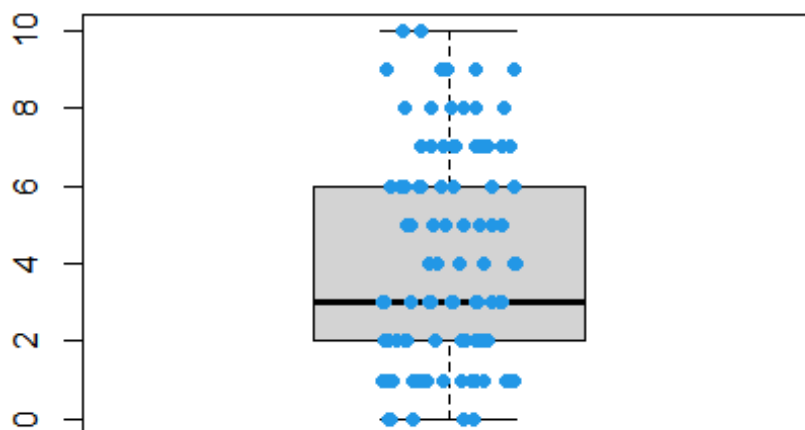
```
plot(density(base_rx$PCR_6, na.rm = T), main = "Densidad de la variable PCR a los 36 meses de empezar estudio")
```

idad de la variable PCR a los 36 meses de empezar



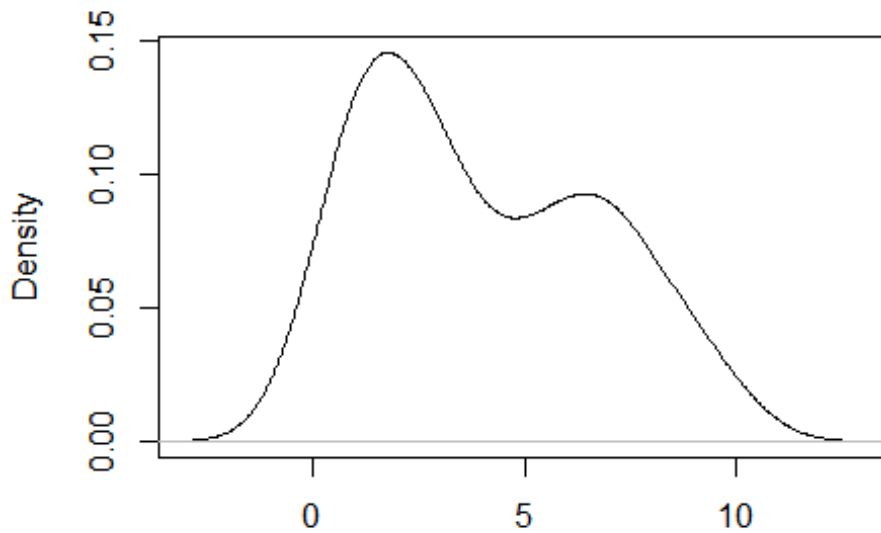
```
boxplot(x=base_rx$EGP, main = "Boxplot: EVA GLOBAL DEL PACIENTE")  
stripchart(x=base_rx$EGP, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 12)
```

Boxplot: EVA GLOBAL DEL PACIENTE



```
plot(density(base_rx$EGP, na.rm = T), main = "Densidad de la variable EVA GLOBAL DEL PACIENTE")
```

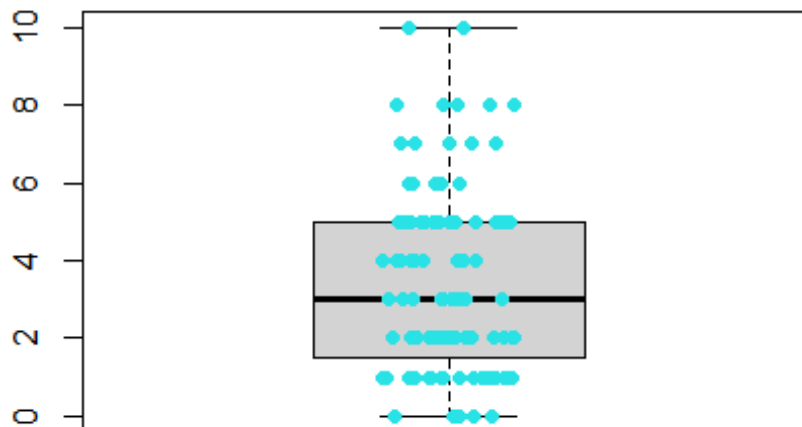
Densidad de la variable EVA GLOBAL DEL PACIEN



N = 98 Bandwidth = 1.001

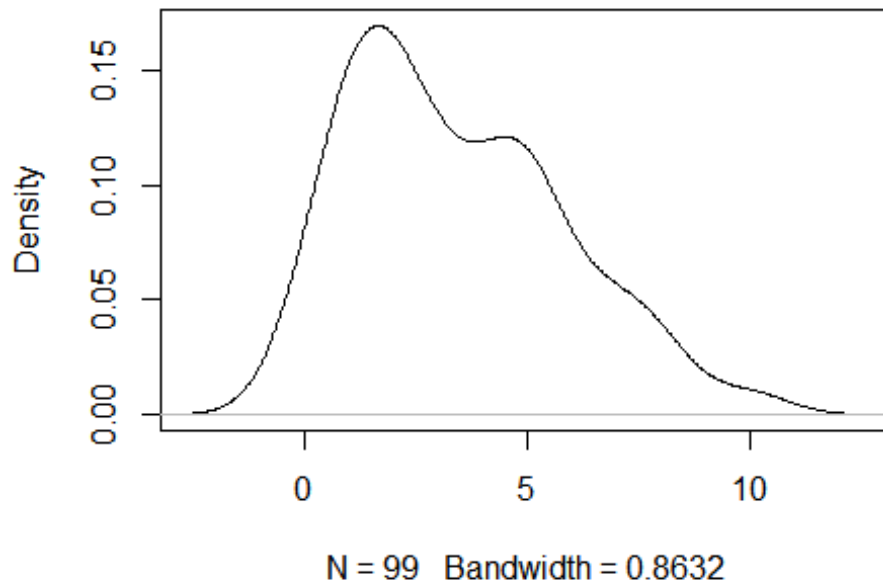
```
boxplot(x=base_rx$EGM, main = "Boxplot: EVA GLOBAL DEL MÃ%0089DICO")  
stripchart(x=base_rx$EGM, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 13)
```

Boxplot: EVA GLOBAL DEL MÃ<U+0089>DICO



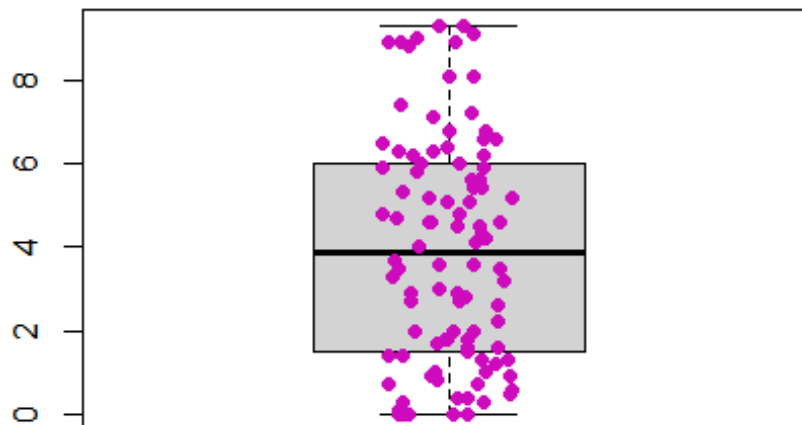
```
plot(density(base_rx$EGM, na.rm = T), main = "Densidad de la variable EVA GLOBAL DEL MÃ  
%0089DICO")
```

Densidad de la variable EVA GLOBAL DEL MÃ<U+0089



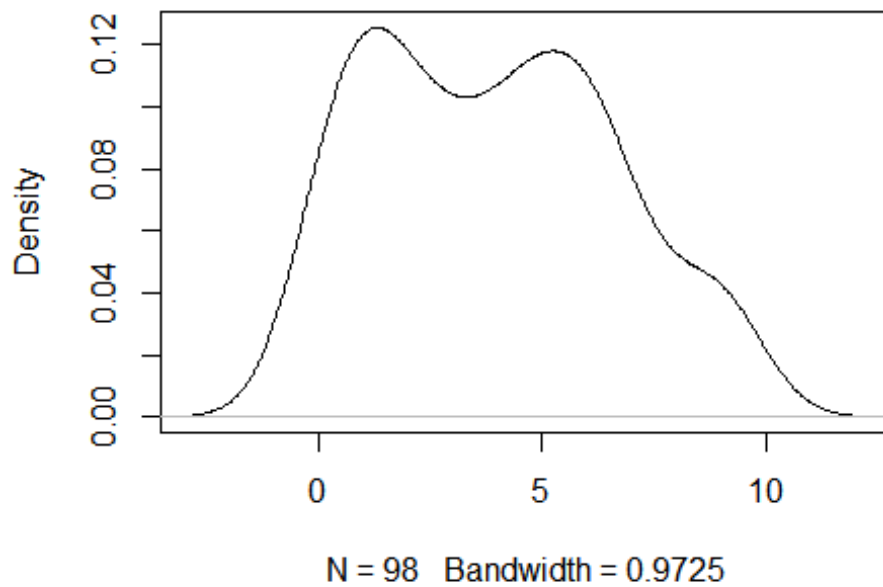
```
boxplot(x=base_rx$BASFI_0, main = "Boxplot: BASFI al empezar el estudio")  
stripchart(x=base_rx$BASFI_0, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI al empezar el estudio



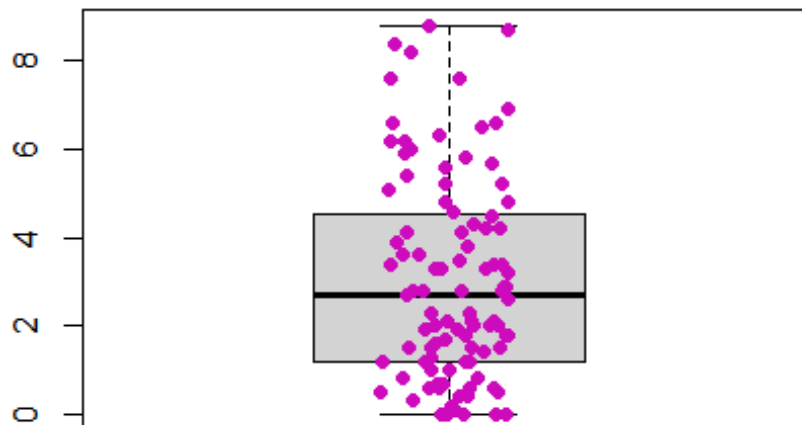
```
plot(density(base_rx$BASFI_0, na.rm = T), main = "Densidad de la variable BASFI al empezar el estudio")
```

Densidad de la variable BASFI al empezar el estud



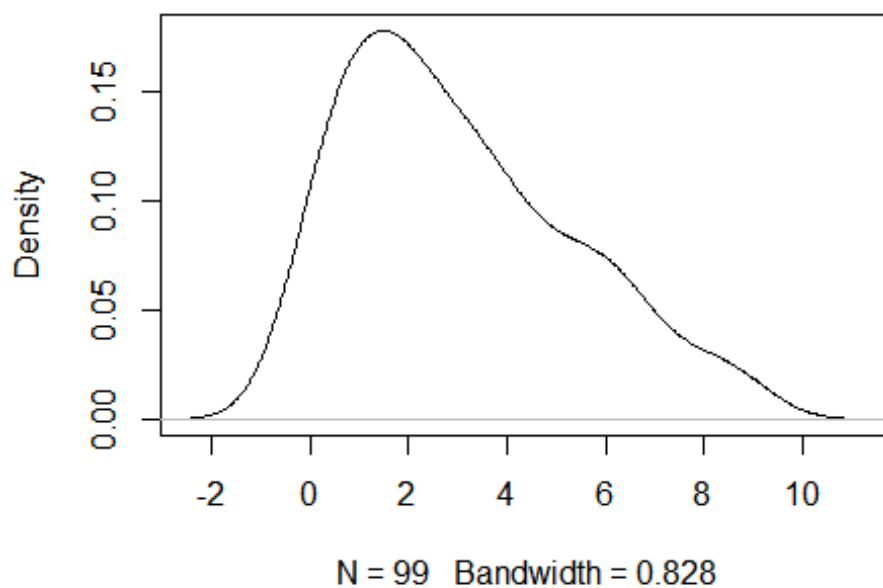
```
boxplot(x=base_rx$BASFI_1, main = "Boxplot: BASFI a los 6 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_1, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 6 meses de empezar estudio



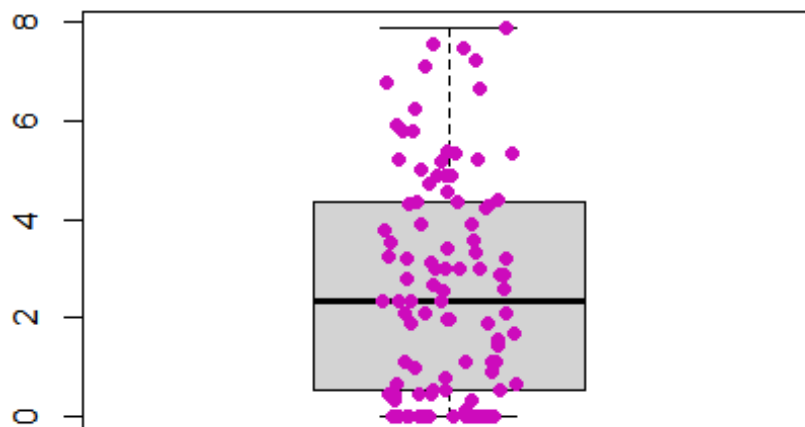
```
plot(density(base_rx$BASFI_1, na.rm = T), main = "Densidad de la variable BASFI a los 6 meses  
de empezar estudio")
```

Densidad de la variable BASFI a los 6 meses de empezar



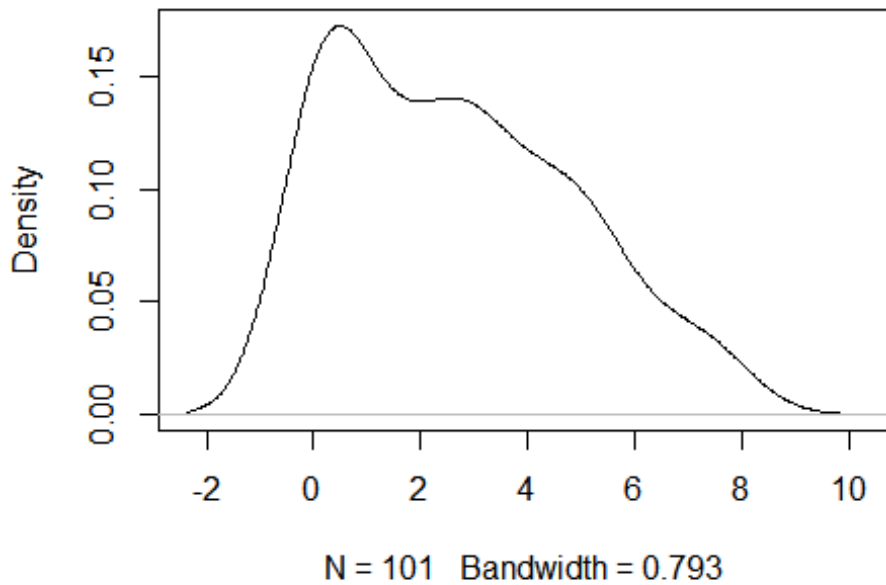
```
boxplot(x=base_rx$BASFI_2, main = "Boxplot: BASFI a los 12 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_2, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 12 meses de empezar estudi



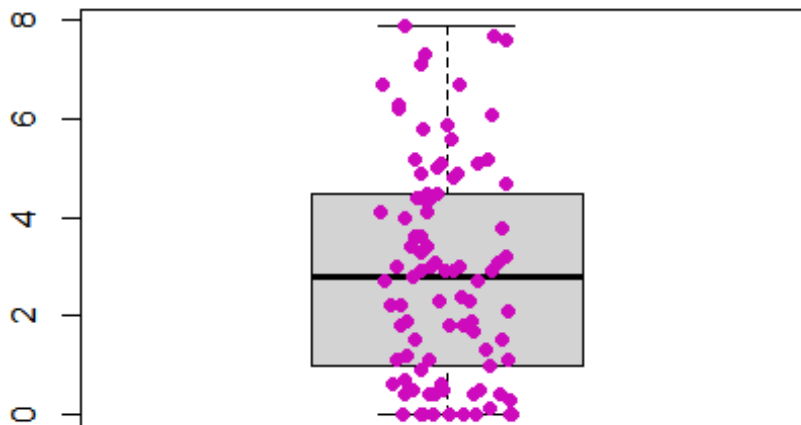
```
plot(density(base_rx$BASFI_2, na.rm = T), main = "Densidad de la variable BASFI a los 12 mese  
s de empezar estudio")
```

Densidad de la variable BASFI a los 12 meses de empeza



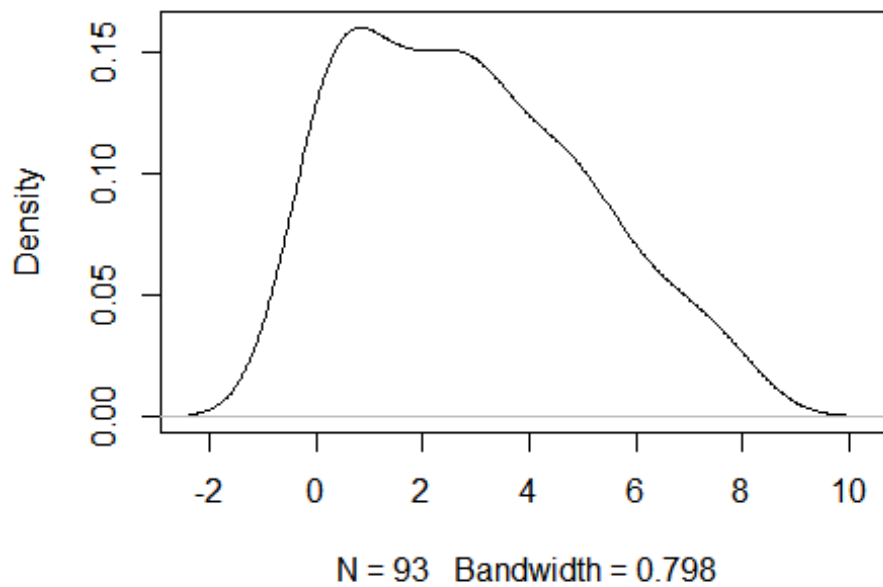
```
boxplot(x=base_rx$BASFI_3, main = "Boxplot: BASFI a los 18 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_3, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 18 meses de empezar estudi



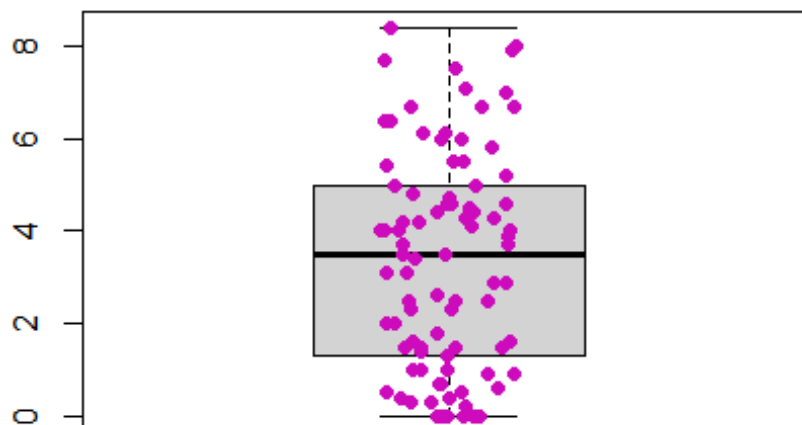
```
plot(density(base_rx$BASFI_3, na.rm = T), main = "Densidad de la variable BASFI a los 18 mese  
s de empezar estudio")
```


Densidad de la variable BASFI a los 18 meses de empeza



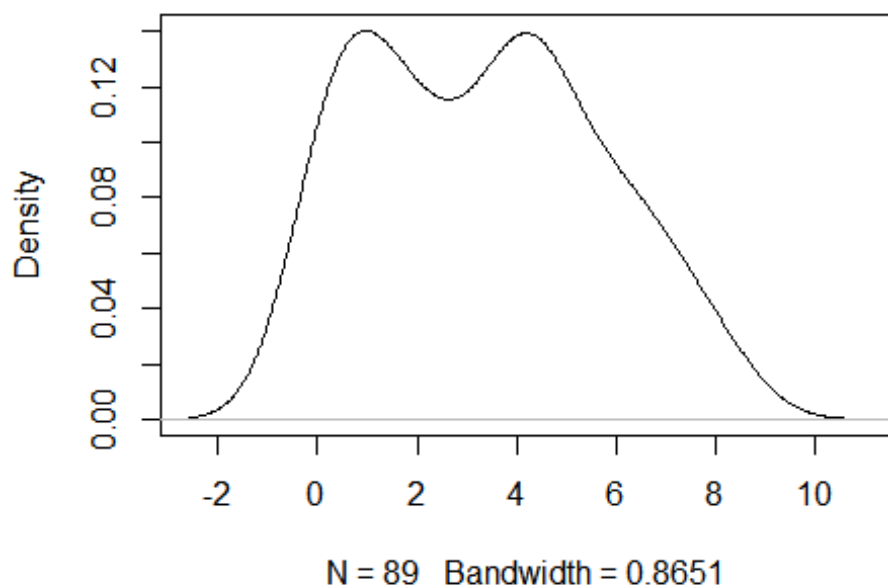
```
boxplot(x=base_rx$BASFI_4, main = "Boxplot: BASFI a los 24 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_4, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 24 meses de empezar estudi



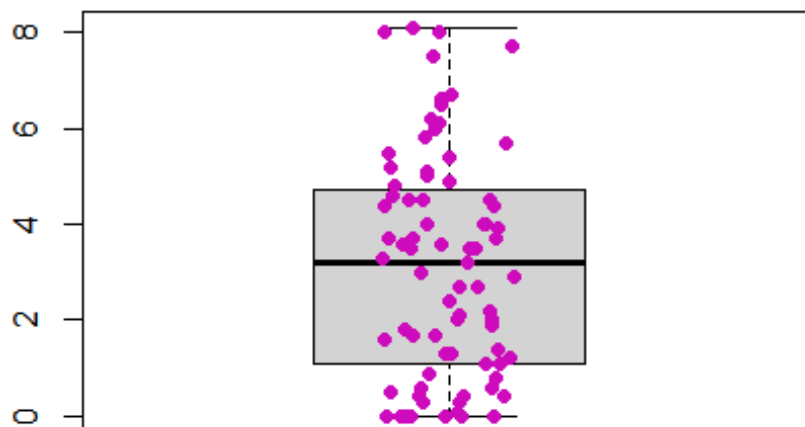
```
plot(density(base_rx$BASFI_4, na.rm = T), main = "Densidad de la variable BASFI a los 24 mese  
s de empezar estudio")
```

Densidad de la variable BASFI a los 24 meses de empeza



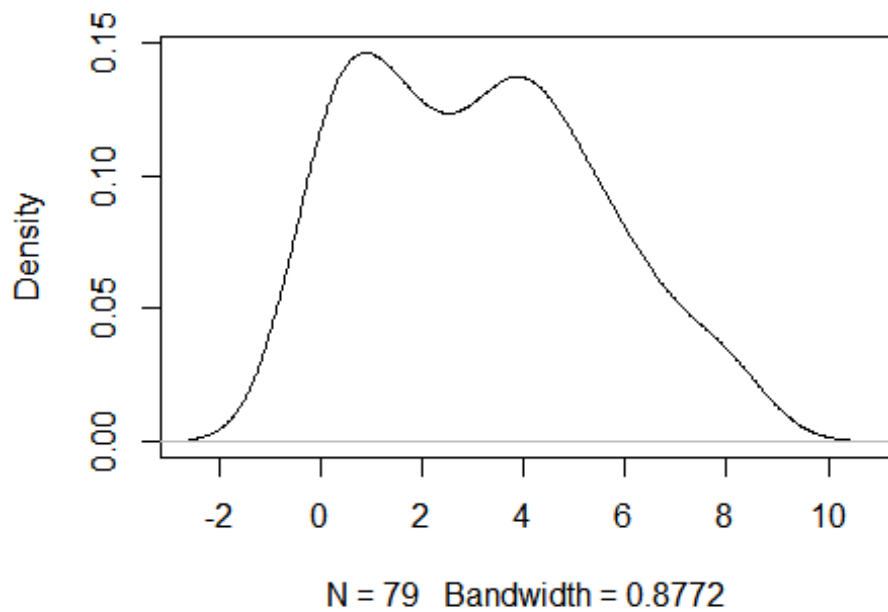
```
boxplot(x=base_rx$BASFI_5, main = "Boxplot: BASFI a los 30 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_5, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 30 meses de empezar estudi



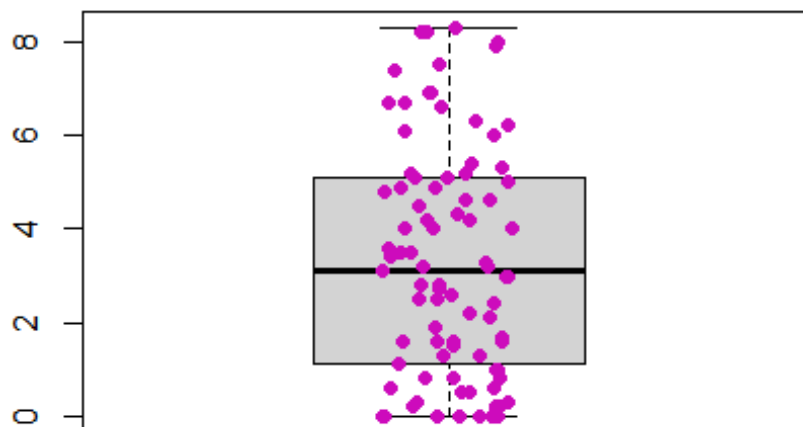
```
plot(density(base_rx$BASFI_5, na.rm = T), main = "Densidad de la variable BASFI a los 30 mese  
s de empezar estudio")
```

Densidad de la variable BASFI a los 30 meses de empeza



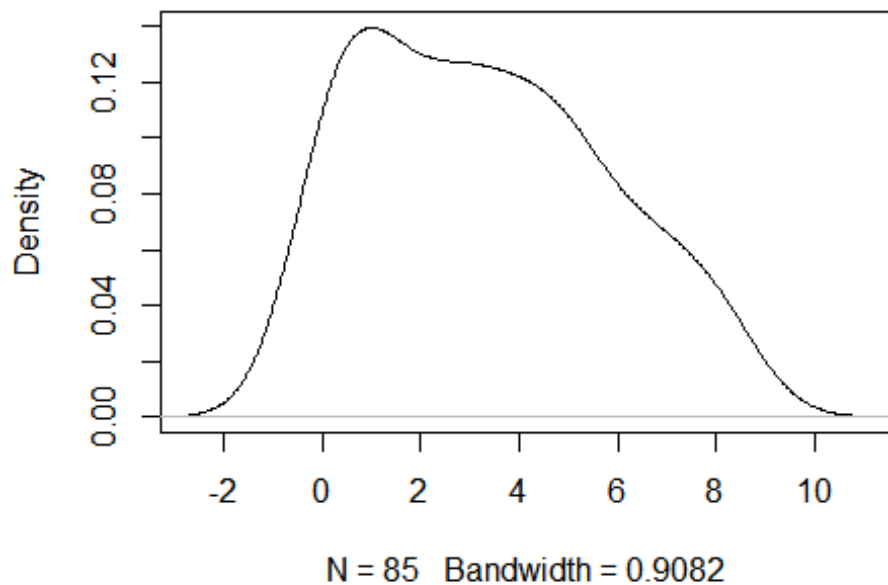
```
boxplot(x=base_rx$BASFI_6, main = "Boxplot: BASFI a los 36 meses de empezar estudio")  
stripchart(x=base_rx$BASFI_6, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 14)
```

Boxplot: BASFI a los 36 meses de empezar estudi



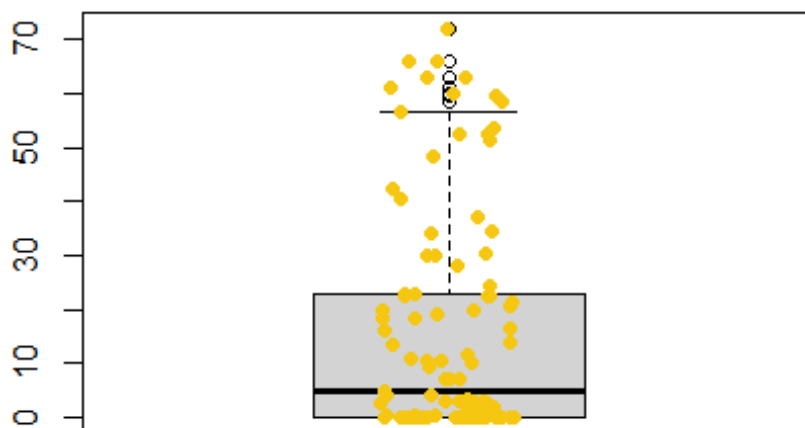
```
plot(density(base_rx$BASFI_6, na.rm = T), main = "Densidad de la variable BASFI a los 36 mese  
s de empezar estudio")
```

Densidad de la variable BASFI a los 36 meses de empeza



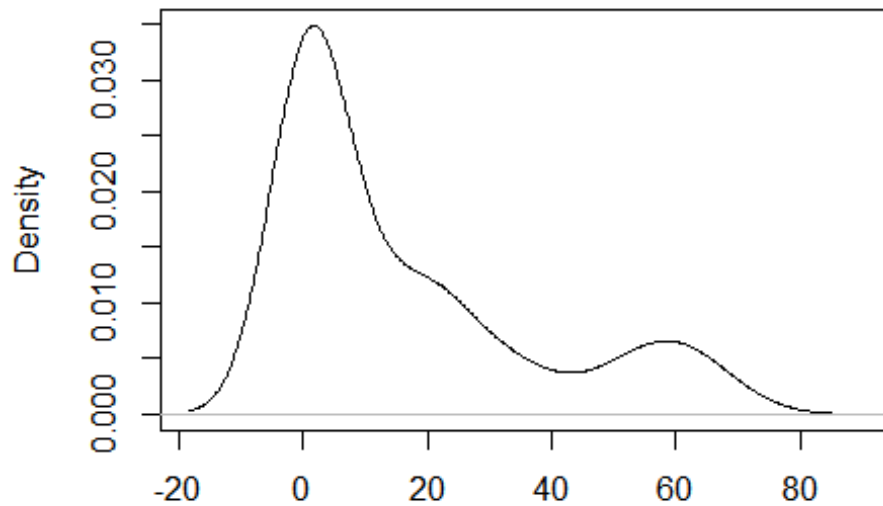
```
boxplot(x=base_rx$mSASSS_Mean_BL, main = "Boxplot: edad al inicio del estudio")  
stripchart(x=base_rx$mSASSS_Mean_BL, method = "jitter", pch=16, add= TRUE, vertical= TRUE  
, col = 15)
```

Boxplot: edad al inicio del estudio



```
plot(density(base_rx$mSASSS_Mean_BL, na.rm = T), main = "Densidad de la variable edad al in  
cio del estudio")
```

Densidad de la variable edad al inicio del estudio

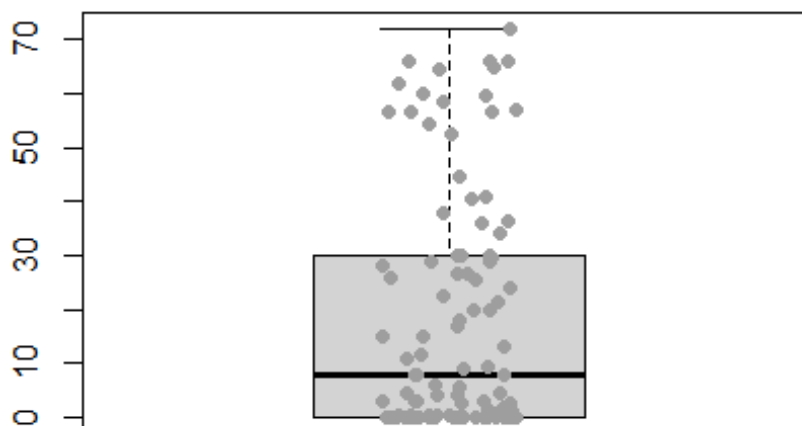


N = 101 Bandwidth = 6.138

```
boxplot(x=base_rx$mSASSS_Mean_FL, main = "Boxplot: edad al inicio del estudio")
stripchart(x=base_rx$mSASSS_Mean_FL, na.rm = T, method = "jitter", pch=16, add= TRUE, vertical= TRUE, col = 16)
```

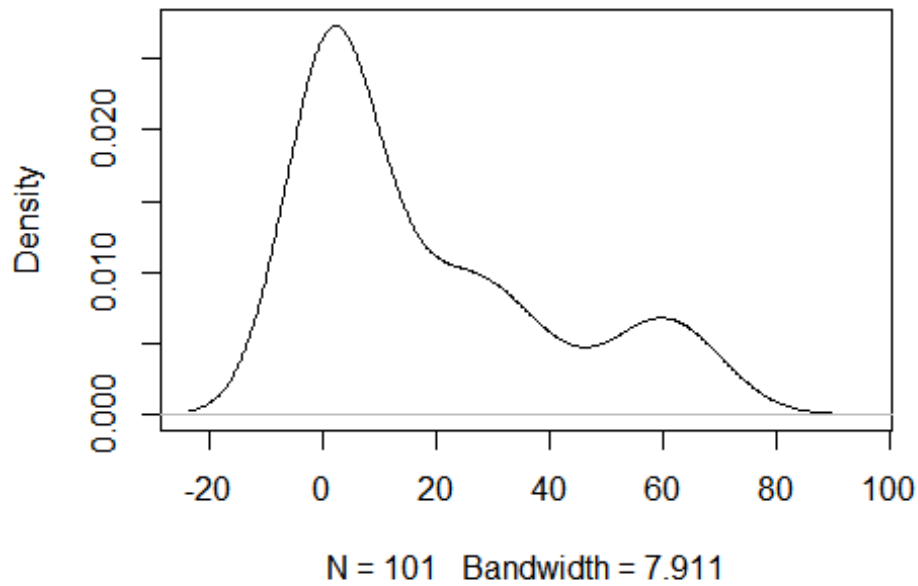
```
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "na.rm" is not a
## graphical parameter
```

Boxplot: edad al inicio del estudio



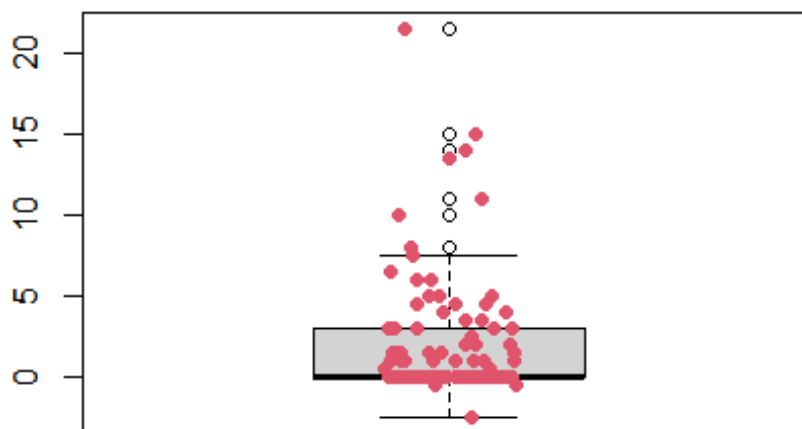
```
plot(density(base_rx$mSASSS_Mean_FL, na.rm = T), main = "Densidad de la variable edad al inicio del estudio")
```

Densidad de la variable edad al inicio del estudio



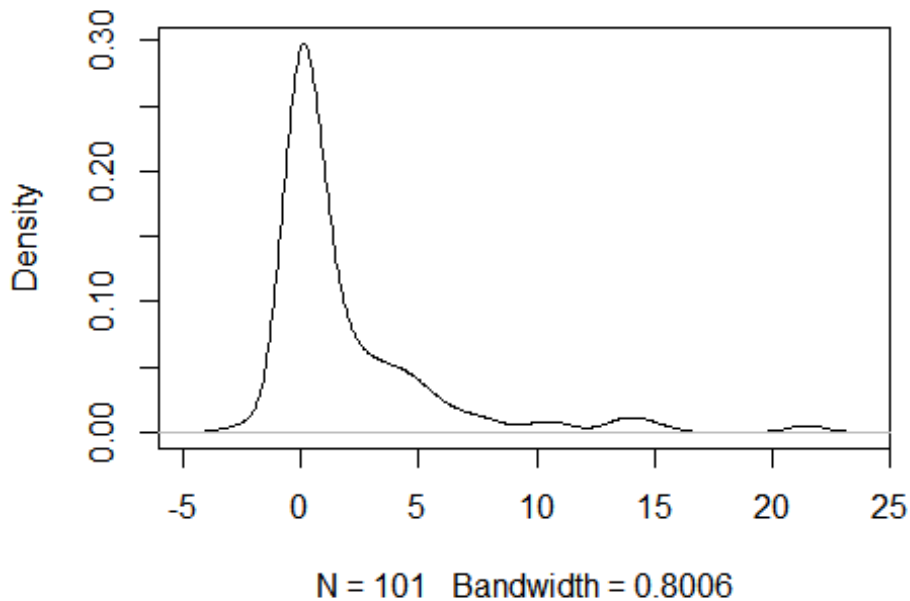
```
boxplot(x=base_rx$mSASSS_Progress, main = "Boxplot: edad al inicio del estudio")  
stripchart(x=base_rx$mSASSS_Progress, method = "jitter", pch=16, add= TRUE, vertical= TRUE,  
col = 18)
```

Boxplot: edad al inicio del estudio



```
plot(density(base_rx$mSASSS_Progress, na.rm = T), main = "Densidad de la variable edad al inicio del estudio")
```

Densidad de la variable edad al inicio del estudio



Ajustamos las covariables, de medidas repetidos en el tiempo, de la misma forma que ajustamos la variable respuesta. Del mismo modo elimino las variables mSASSS_Mean_BL y mSASSS_Mean_FL porque la diferencia de ellas, es la variable respuesta, la variable y (mSASSS_Progress)

```
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

base_rx1 <-
  base_rx %>%
  mutate( BASFIDif = (BASFI_6-BASFI_0), BASDAIDif = (BASDAI_6-BASDAI_0
), ASDASDif = (ASDAS_6-ASDAS_0), PCRDif = (PCR_6-PCR_0))
summary(base_rx1)

##      NUM_PAC      EDAD      SEXO      IMC      IMC>30
## U01001 : 1  Min.   :21.00  MALE  :82  Min.   :19.37  NO   :84
## U01007 : 1  1st Qu.:38.00  FEMALE:19  1st Qu.:24.05  SI   :11
## U01017 : 1  Median :46.00                Median :26.02  NA's: 6
## U02001 : 1  Mean    :46.68                Mean    :26.38
```

```

## U02003 : 1 3rd Qu.:54.00 3rd Qu.:28.68
## U02004 : 1 Max. :75.00 Max. :40.83
## (Other):95 NA's :6
## TSINT HLAB27 EA TABACO_Kat VSG
BASMI
## Min. : 0.00 SI :86 SI:86 SI:70 Min. : 0.50 Min.
:0.5378
## 1st Qu.: 8.00 NO :13 NO:15 NO:31 1st Qu.: 9.00 1st
Qu.:1.6458
## Median :15.00 NA's: 2 Median : 19.00 Medi
an :2.8728
## Mean :17.76 Mean : 26.65 Mean
:3.0915
## 3rd Qu.:26.00 3rd Qu.: 38.75 3rd
Qu.:4.0864
## Max. :52.00 Max. :115.00 Max.
:7.2729
## NA's :4 NA's :3 NA's
:19
## UVEITIS PSORIASIS ENF_INF TTOBIO NUM_TB_PREV MO
TIVO
## NO:79 NO :92 NO :92 ETA:31 treatments.0:75 non.change
:77
## SI:22 SI : 7 SI : 7 ADA:44 treatments.1:20 Inefficacy
:17
## NA's: 2 NA's: 2 IFX:11 treatments.2: 6 Side.effect
s: 7
## GOL:12
## CTZ: 3
##
##
## TIEMBIO BIO_INICI Biol_4_kat Interval AINE F
AME
## Min. : 0.00 NO :25 NO:55 Min. :2.00 NO :42 NO
:81
## 1st Qu.: 8.00 SI :75 SI:46 1st Qu.:3.00 SI :57 SI
:18
## Median : 42.50 NA's: 1 Median :3.00 NA's: 2 NA'
s: 2
## Mean : 45.51 Mean :3.45
## 3rd Qu.: 76.75 3rd Qu.:4.00
## Max. :132.00 Max. :6.00
## NA's :1 NA's :1
## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2
## Min. :0.700 Min. :0.20 Min. :0.000 Min. : 0.000
## 1st Qu.:4.225 1st Qu.:2.00 1st Qu.:1.400 1st Qu.: 1.400
## Median :5.400 Median :3.20 Median :2.700 Median : 2.500
## Mean :5.326 Mean :3.67 Mean :2.902 Mean : 3.419
## 3rd Qu.:6.400 3rd Qu.:5.10 3rd Qu.:4.400 3rd Qu.: 4.200
## Max. :8.800 Max. :8.80 Max. :8.500 Max. :65.000
## NA's :3 NA's :4 NA's :4 NA's :4
## BASDAI_3 BASDAI_4 BASDAI_5 BASDAI_6
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:1.200 1st Qu.:1.200 1st Qu.:1.275 1st Qu.:1.100

```


##	Median :3.000	Median :2.800	Median :2.550	Median :2.800
##	Mean :3.108	Mean :3.075	Mean :2.915	Mean :2.924
##	3rd Qu.:4.800	3rd Qu.:4.600	3rd Qu.:4.200	3rd Qu.:4.600
##	Max. :8.000	Max. :8.000	Max. :8.000	Max. :7.300
##	NA's :8	NA's :10	NA's :17	NA's :16
##	ASDAS_0	ASDAS_1	ASDAS_2	ASDAS_3
##	Min. :0.2257	Min. :0.2471	Min. :0.001737	Min. :0.048
92				
##	1st Qu.:1.3874	1st Qu.:1.1572	1st Qu.:1.088078	1st Qu.:1.164
97				
##	Median :2.0565	Median :1.5738	Median :1.827288	Median :1.766
56				
##	Mean :2.2160	Mean :1.8249	Mean :1.773470	Mean :1.815
23				
##	3rd Qu.:2.9248	3rd Qu.:2.4204	3rd Qu.:2.316346	3rd Qu.:2.477
97				
##	Max. :5.0749	Max. :7.1255	Max. :4.131246	Max. :3.983
74				
##	NA's :5	NA's :13	NA's :9	NA's :14
##	ASDAS_4	ASDAS_5	ASDAS_6	PCR_0
##	Min. :0.00058	Min. :0.0338	Min. :0.096	Min. :0.000
##	1st Qu.:1.17363	1st Qu.:1.0522	1st Qu.:1.143	1st Qu.:1.000
##	Median :1.66660	Median :1.6400	Median :1.680	Median :3.350
##	Mean :1.81274	Mean :1.7329	Mean :1.807	Mean :8.181
##	3rd Qu.:2.44173	3rd Qu.:2.1781	3rd Qu.:2.397	3rd Qu.:9.100
##	Max. :4.45150	Max. :4.0579	Max. :3.988	Max. :88.700
##	NA's :22	NA's :30	NA's :24	NA's :7
##	PCR_1	PCR_2	PCR_3	PCR_4
##	Min. :0.000	Min. :0.00	Min. :0.000	Min. :0.000
##	1st Qu.:1.000	1st Qu.:0.90	1st Qu.:0.750	1st Qu.:0.700
##	Median :2.000	Median :2.00	Median :1.900	Median :2.050
##	Mean :4.358	Mean :4.36	Mean :4.452	Mean :5.368
##	3rd Qu.:4.675	3rd Qu.:5.60	3rd Qu.:5.600	3rd Qu.:5.375
##	Max. :61.300	Max. :33.90	Max. :37.000	Max. :68.500
##	NA's :3	NA's :2	NA's :4	NA's :7
##	PCR_5	PCR_6	EGP	EGM
##	Min. :0.000	Min. :0.00	Min. :0.000	Min. :0.000
##	1st Qu.:0.615	1st Qu.:0.60	1st Qu.:2.000	1st Qu.:1.500
##	Median :1.400	Median :2.00	Median :3.000	Median :3.000
##	Mean :4.118	Mean :4.42	Mean :4.031	Mean :3.455
##	3rd Qu.:6.000	3rd Qu.:5.95	3rd Qu.:6.000	3rd Qu.:5.000
##	Max. :65.900	Max. :39.70	Max. :10.000	Max. :10.000
##	NA's :10	NA's :14	NA's :3	NA's :2
##	BASFI_0	BASFI_1	BASFI_2	BASFI_3
BASFI_4				
##	Min. :0.000	Min. :0.000	Min. :0.000	Min. :0.000
Min. :0.00				
##	1st Qu.:1.525	1st Qu.:1.200	1st Qu.:0.550	1st Qu.:1.000
1st Qu.:1.30				
##	Median :3.850	Median :2.700	Median :2.330	Median :2.800
Median :3.50				
##	Mean :3.931	Mean :3.059	Mean :2.667	Mean :2.882
Mean :3.34				
##	3rd Qu.:5.975	3rd Qu.:4.550	3rd Qu.:4.340	3rd Qu.:4.500
3rd				

```

d Qu.:5.00
## Max. :9.300 Max. :8.800 Max. :7.890 Max. :7.900 Ma
x. :8.40
## NA's :3 NA's :2 NA's :8 NA
's :12
## BASFI_5 BASFI_6 mSASSS_Mean_BL mSASSS_Mean_FL
## Min. :0.000 Min. :0.000 Min. : 0.00 Min. : 0.00
## 1st Qu.:1.100 1st Qu.:1.100 1st Qu.: 0.00 1st Qu.: 0.00
## Median :3.200 Median :3.100 Median : 5.00 Median : 8.00
## Mean :3.118 Mean :3.313 Mean :16.46 Mean :18.44
## 3rd Qu.:4.700 3rd Qu.:5.100 3rd Qu.:23.00 3rd Qu.:30.00
## Max. :8.100 Max. :8.300 Max. :72.00 Max. :72.00
## NA's :22 NA's :16
## mSASSS_Progress New_Syn_mSASSS Prog_Syn_mSASSS mSASSS_Progress_Kat
## Min. :-2.50 NO:81 NO:90 NO:71
## 1st Qu.: 0.00 SI:20 SI:11 SI:30
## Median : 0.00
## Mean : 1.98
## 3rd Qu.: 3.00
## Max. :21.50
##
## New_Prog_Syn_Msasss Mind_2_Synd_FL Mind_2_Synd_BL PCR_INICIOB
## NO:79 NO:46 NO:50 Min. : 0.00
## SI:22 SI:55 SI:51 1st Qu.: 4.60
## Median :11.10
## Mean :18.70
## 3rd Qu.:28.05
## Max. :98.00
## NA's :2
## i_ASIDAS HVGP BASFIDif BASDAIDif
## Min. :1.500 Min. : 1.000 Min. :-8.9000 Min. :-7.0000
## 1st Qu.:2.700 1st Qu.: 5.000 1st Qu.:-1.8000 1st Qu.:-1.1000
## Median :3.500 Median : 6.000 Median :-0.2000 Median :-0.1000
## Mean :3.402 Mean : 5.923 Mean :-0.5667 Mean :-0.5354
## 3rd Qu.:4.000 3rd Qu.: 7.000 3rd Qu.: 0.6250 3rd Qu.: 0.4750
## Max. :5.300 Max. :10.000 Max. : 6.2000 Max. : 5.2000
## NA's :2 NA's :17 NA's :19
## ASDASDif PCRDif
## Min. :-3.4650 Min. :-88.40
## 1st Qu.:-1.0300 1st Qu.: -5.40
## Median :-0.2247 Median : -0.20
## Mean :-0.3908 Mean : -4.14
## 3rd Qu.: 0.2567 3rd Qu.: 1.00
## Max. : 1.9947 Max. : 28.00
## NA's :28 NA's :20

base_rx2 <- base_rx1[,c(-1,-25:-45,-48:-56)]
summary(base_rx2)

## EDAD SEXO IMC IMC>30 TSINT
## Min. :21.00 MALE :82 Min. :19.37 NO :84 Min. : 0.0
0
## 1st Qu.:38.00 FEMALE:19 1st Qu.:24.05 SI :11 1st Qu.: 8.0
0

```

```

## Median :46.00          Median :26.02  NA's: 6  Median :15.0
0
## Mean   :46.68          Mean    :26.38          Mean    :17.7
6
## 3rd Qu.:54.00          3rd Qu.:28.68          3rd Qu.:26.0
0
## Max.   :75.00          Max.    :40.83          Max.    :52.0
0
##                                     NA's   :6          NA's   :4
## HLAB27   EA   TABACO_Kat   VSG          BASMI   UVE
ITIS
## SI :86   SI:86   SI:70   Min.   : 0.50   Min.   :0.5378   NO:
79
## NO :13   NO:15   NO:31   1st Qu.: 9.00   1st Qu.:1.6458   SI:
22
## NA's: 2          Median : 19.00   Median :2.8728
##                                     Mean   : 26.65   Mean   :3.0915
##                                     3rd Qu.: 38.75   3rd Qu.:4.0864
##                                     Max.   :115.00   Max.   :7.2729
##                                     NA's   :3          NA's   :19
## PSORIASIS ENF_INF   TTOBIO          NUM_TB_PREV          MOTIVO
## NO :92   NO :92   ETA:31   treatments.0:75   non.change :77
## SI : 7   SI : 7   ADA:44   treatments.1:20   Inefficacy :17
## NA's: 2   NA's: 2   IFX:11   treatments.2: 6   Side.effects: 7
##                                     GOL:12
##                                     CTZ: 3
##
##
##
## TIEMBIO          BIO_INICI Biol_4_kat   Interval   AINE   F
AME
## Min.   : 0.00   NO :25   NO:55   Min.   :2.00   NO :42   NO
:81
## 1st Qu.: 8.00   SI :75   SI:46   1st Qu.:3.00   SI :57   SI
:18
## Median : 42.50   NA's: 1          Median :3.00   NA's: 2   NA'
s: 2
## Mean   : 45.51          Mean   :3.45
## 3rd Qu.: 76.75          3rd Qu.:4.00
## Max.   :132.00          Max.   :6.00
## NA's   :1          NA's   :1
## i_BASDAI          EGP          EGM          mSASSS_Progress
## Min.   :0.700   Min.   : 0.000   Min.   : 0.000   Min.   :-2.50
## 1st Qu.:4.225   1st Qu.: 2.000   1st Qu.: 1.500   1st Qu.: 0.00
## Median :5.400   Median : 3.000   Median : 3.000   Median : 0.00
## Mean   :5.326   Mean   : 4.031   Mean   : 3.455   Mean   : 1.98
## 3rd Qu.:6.400   3rd Qu.: 6.000   3rd Qu.: 5.000   3rd Qu.: 3.00
## Max.   :8.800   Max.   :10.000   Max.   :10.000   Max.   :21.50
## NA's   :3          NA's   :3          NA's   :2
## New_Syn_mSASSS Prog_Syn_mSASSS mSASSS_Progress_Kat New_Prog_Syn_Ms
asss
## NO:81          NO:90          NO:71          NO:79
## SI:20          SI:11          SI:30          SI:22
##
##

```

```

##
##
##
## Mind_2_Synd_FL Mind_2_Synd_BL PCR_INICIOB i_ASDas
HVGP
## NO:46 NO:50 Min. : 0.00 Min. :1.500 Min.
: 1.000
## SI:55 SI:51 1st Qu.: 4.60 1st Qu.:2.700 1st
Qu.: 5.000
## Median :11.10 Median :3.500 Medi
an : 6.000
## Mean :18.70 Mean :3.402 Mean
: 5.923
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd
Qu.: 7.000
## Max. :98.00 Max. :5.300 Max.
:10.000
## NA's :2 NA's :2
## BASFIDif BASDAIDif ASDASDif PCRDif
## Min. :-8.9000 Min. :-7.0000 Min. :-3.4650 Min. :-88.
40
## 1st Qu.:-1.8000 1st Qu.:-1.1000 1st Qu.:-1.0300 1st Qu.: -5.
40
## Median :-0.2000 Median :-0.1000 Median :-0.2247 Median : -0.
20
## Mean :-0.5667 Mean :-0.5354 Mean :-0.3908 Mean : -4.
14
## 3rd Qu.: 0.6250 3rd Qu.: 0.4750 3rd Qu.: 0.2567 3rd Qu.: 1.
00
## Max. : 6.2000 Max. : 5.2000 Max. : 1.9947 Max. : 28.
00
## NA's :17 NA's :19 NA's :28 NA's :20

base_rx21 <- base_rx1[,c(-1,-25:-46,-48:-56,-68)]
summary(base_rx21)

## EDAD SEXO IMC IMC>30 TSINT
## Min. :21.00 MALE :82 Min. :19.37 NO :84 Min. : 0.0
0
## 1st Qu.:38.00 FEMALE:19 1st Qu.:24.05 SI :11 1st Qu.: 8.0
0
## Median :46.00 Median :26.02 NA's: 6 Median :15.0
0
## Mean :46.68 Mean :26.38 Mean :17.7
6
## 3rd Qu.:54.00 3rd Qu.:28.68 3rd Qu.:26.0
0
## Max. :75.00 Max. :40.83 Max. :52.0
0
## NA's :6 NA's :4
## HLAB27 EA TABACO_Kat VSG BASMI UVE
ITIS
## SI :86 SI:86 SI:70 Min. : 0.50 Min. :0.5378 NO:
79

```

```

## NO :13 NO:15 NO:31 1st Qu.: 9.00 1st Qu.:1.6458 SI:
22
## NA's: 2 Median : 19.00 Median :2.8728
## Mean : 26.65 Mean :3.0915
## 3rd Qu.: 38.75 3rd Qu.:4.0864
## Max. :115.00 Max. :7.2729
## NA's :3 NA's :19
## PSORIASIS ENF_INF TTOBIO NUM_TB_PREV MOTIVO
## NO :92 NO :92 ETA:31 treatments.0:75 non.change :77
## SI : 7 SI : 7 ADA:44 treatments.1:20 Inefficacy :17
## NA's: 2 NA's: 2 IFX:11 treatments.2: 6 Side.effects: 7
## GOL:12
## CTZ: 3
##
##
## TIEMBIO BIO_INICI Biol_4_kat Interval AINE F
AME
## Min. : 0.00 NO :25 NO:55 Min. :2.00 NO :42 NO
:81
## 1st Qu.: 8.00 SI :75 SI:46 1st Qu.:3.00 SI :57 SI
:18
## Median : 42.50 NA's: 1 Median :3.00 NA's: 2 NA'
s: 2
## Mean : 45.51 Mean :3.45
## 3rd Qu.: 76.75 3rd Qu.:4.00
## Max. :132.00 Max. :6.00
## NA's :1 NA's :1
## i_BASDAI EGM mSASSS_Progress New_Syn_mSASSS
## Min. :0.700 Min. : 0.000 Min. :-2.50 NO:81
## 1st Qu.:4.225 1st Qu.: 1.500 1st Qu.: 0.00 SI:20
## Median :5.400 Median : 3.000 Median : 0.00
## Mean :5.326 Mean : 3.455 Mean : 1.98
## 3rd Qu.:6.400 3rd Qu.: 5.000 3rd Qu.: 3.00
## Max. :8.800 Max. :10.000 Max. :21.50
## NA's :3 NA's :2
## Prog_Syn_mSASSS mSASSS_Progress_Kat New_Prog_Syn_Msasss Mind_2_Syn
d_FL
## NO:90 NO:71 NO:79 NO:46
## SI:11 SI:30 SI:22 SI:55
##
##
##
## Mind_2_Synd_BL PCR_INICIOB i_ASIDAS HVGP
## NO:50 Min. : 0.00 Min. :1.500 Min. : 1.000
## SI:51 1st Qu.: 4.60 1st Qu.:2.700 1st Qu.: 5.000
## Median :11.10 Median :3.500 Median : 6.000
## Mean :18.70 Mean :3.402 Mean : 5.923
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd Qu.: 7.000
## Max. :98.00 Max. :5.300 Max. :10.000
## NA's :2 NA's :2
## BASFIDif ASDASdif PCRDif
## Min. :-8.9000 Min. :-3.4650 Min. :-88.40

```

```
## 1st Qu.: -1.8000 1st Qu.: -1.0300 1st Qu.: -5.40
## Median : -0.2000 Median : -0.2247 Median : -0.20
## Mean : -0.5667 Mean : -0.3908 Mean : -4.14
## 3rd Qu.: 0.6250 3rd Qu.: 0.2567 3rd Qu.: 1.00
## Max. : 6.2000 Max. : 1.9947 Max. : 28.00
## NA's : 17 NA's : 28 NA's : 20
```

```
base_rx22 <- base_rx1[,c(-1, -25:-46, -48:-56, -67, -69)]
summary(base_rx22)
```

```
##          EDAD          SEXO          IMC          IMC>30          TSINT
## Min.   :21.00  MALE :82  Min.   :19.37  NO :84  Min.   : 0.0
0
## 1st Qu.:38.00  FEMALE:19  1st Qu.:24.05  SI :11  1st Qu.: 8.0
0
## Median :46.00          Median :26.02  NA's: 6  Median :15.0
0
## Mean   :46.68          Mean   :26.38          Mean   :17.7
6
## 3rd Qu.:54.00          3rd Qu.:28.68          3rd Qu.:26.0
0
## Max.   :75.00          Max.   :40.83          Max.   :52.0
0
##          NA's :6          NA's :4
##  HLAB27  EA  TABACO_Kat  VSG  BASMI  UVE
ITIS
## SI :86  SI:86  SI:70  Min.   : 0.50  Min.   :0.5378  NO:
79
## NO :13  NO:15  NO:31  1st Qu.: 9.00  1st Qu.:1.6458  SI:
22
## NA's: 2          Median : 19.00  Median :2.8728
##          Mean   : 26.65  Mean   :3.0915
##          3rd Qu.: 38.75  3rd Qu.:4.0864
##          Max.   :115.00  Max.   :7.2729
##          NA's   :3      NA's   :19
##  PSORIASIS ENF_INF  TTOBIO  NUM_TB_PREV  MOTIVO
## NO :92  NO :92  ETA:31  treatments.0:75  non.change :77
## SI : 7  SI : 7  ADA:44  treatments.1:20  Inefficacy :17
## NA's: 2  NA's: 2  IFX:11  treatments.2: 6  Side.effects: 7
##          GOL:12
##          CTZ: 3
##
##
##          TIEMBIO          BIO_INICI Biol_4_kat  Interval  AINE  F
AME
## Min.   : 0.00  NO :25  NO:55  Min.   :2.00  NO :42  NO
:81
## 1st Qu.: 8.00  SI :75  SI:46  1st Qu.:3.00  SI :57  SI
:18
## Median : 42.50  NA's: 1          Median :3.00  NA's: 2  NA'
s: 2
## Mean   : 45.51          Mean   :3.45
## 3rd Qu.: 76.75          3rd Qu.:4.00
## Max.   :132.00          Max.   :6.00
```

```

## NA's :1 NA's :1
## i_BASDAI EGM mSASSS_Progress New_Syn_mSASSS
## Min. :0.700 Min. : 0.000 Min. :-2.50 NO:81
## 1st Qu.:4.225 1st Qu.: 1.500 1st Qu.: 0.00 SI:20
## Median :5.400 Median : 3.000 Median : 0.00
## Mean :5.326 Mean : 3.455 Mean : 1.98
## 3rd Qu.:6.400 3rd Qu.: 5.000 3rd Qu.: 3.00
## Max. :8.800 Max. :10.000 Max. :21.50
## NA's :3 NA's :2
## Prog_Syn_mSASSS mSASSS_Progress_Kat New_Prog_Syn_Msasss Mind_2_Syn
d_FL
## NO:90 NO:71 NO:79 NO:46
## SI:11 SI:30 SI:22 SI:55
##
##
##
##
## Mind_2_Synd_BL PCR_INICIOB i_ASIDAS HVGP
## NO:50 Min. : 0.00 Min. :1.500 Min. : 1.000
## SI:51 1st Qu.: 4.60 1st Qu.:2.700 1st Qu.: 5.000
## Median :11.10 Median :3.500 Median : 6.000
## Mean :18.70 Mean :3.402 Mean : 5.923
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd Qu.: 7.000
## Max. :98.00 Max. :5.300 Max. :10.000
## NA's :2 NA's :2
## BASDAIDif PCRDif
## Min. :-7.0000 Min. :-88.40
## 1st Qu.:-1.1000 1st Qu.: -5.40
## Median :-0.1000 Median : -0.20
## Mean :-0.5354 Mean : -4.14
## 3rd Qu.: 0.4750 3rd Qu.: 1.00
## Max. : 5.2000 Max. : 28.00
## NA's :19 NA's :20

```

```

base_rx3 <- base_rx1[,c(-1,-25:-45,-48:-56,-58:-61)]
summary(base_rx3)

```

```

## EDAD SEXO IMC IMC>30 TSINT
## Min. :21.00 MALE :82 Min. :19.37 NO :84 Min. : 0.0
0
## 1st Qu.:38.00 FEMALE:19 1st Qu.:24.05 SI :11 1st Qu.: 8.0
0
## Median :46.00 Median :26.02 NA's: 6 Median :15.0
0
## Mean :46.68 Mean :26.38 Mean :17.7
6
## 3rd Qu.:54.00 3rd Qu.:28.68 3rd Qu.:26.0
0
## Max. :75.00 Max. :40.83 Max. :52.0
0
## HLAB27 EA TABACO_Kat VSG BASMI UVE
ITIS

```

```

## SI :86 SI:86 SI:70 Min. : 0.50 Min. :0.5378 NO:
79
## NO :13 NO:15 NO:31 1st Qu.: 9.00 1st Qu.:1.6458 SI:
22
## NA's: 2 Median : 19.00 Median :2.8728
## Mean : 26.65 Mean :3.0915
## 3rd Qu.: 38.75 3rd Qu.:4.0864
## Max. :115.00 Max. :7.2729
## NA's :3 NA's :19
## PSORIASIS ENF_INF TTOBIO NUM_TB_PREV MOTIVO
## NO :92 NO :92 ETA:31 treatments.0:75 non.change :77
## SI : 7 SI : 7 ADA:44 treatments.1:20 Inefficacy :17
## NA's: 2 NA's: 2 IFX:11 treatments.2: 6 Side.effects: 7
## GOL:12
## CTZ: 3
##
##
## TIEMBIO BIO_INICI Biol_4_kat Interval AINE F
AME
## Min. : 0.00 NO :25 NO:55 Min. :2.00 NO :42 NO
:81
## 1st Qu.: 8.00 SI :75 SI:46 1st Qu.:3.00 SI :57 SI
:18
## Median : 42.50 NA's: 1 Median :3.00 NA's: 2 NA'
s: 2
## Mean : 45.51 Mean :3.45
## 3rd Qu.: 76.75 3rd Qu.:4.00
## Max. :132.00 Max. :6.00
## NA's :1 NA's :1
## i_BASDAI EGP EGM mSASSS_Progress
## Min. :0.700 Min. : 0.000 Min. : 0.000 Min. : -2.50
## 1st Qu.:4.225 1st Qu.: 2.000 1st Qu.: 1.500 1st Qu.: 0.00
## Median :5.400 Median : 3.000 Median : 3.000 Median : 0.00
## Mean :5.326 Mean : 4.031 Mean : 3.455 Mean : 1.98
## 3rd Qu.:6.400 3rd Qu.: 6.000 3rd Qu.: 5.000 3rd Qu.: 3.00
## Max. :8.800 Max. :10.000 Max. :10.000 Max. :21.50
## NA's :3 NA's :3 NA's :2
## Mind_2_Synd_FL Mind_2_Synd_BL PCR_INICIOB i_ASIDAS
HVGP
## NO:46 NO:50 Min. : 0.00 Min. :1.500 Min.
: 1.000
## SI:55 SI:51 1st Qu.: 4.60 1st Qu.:2.700 1st
Qu.: 5.000
## Median :11.10 Median :3.500 Medi
an : 6.000
## Mean :18.70 Mean :3.402 Mean
: 5.923
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd
Qu.: 7.000
## Max. :98.00 Max. :5.300 Max.
:10.000
## NA's :2 NA's :2
## BASFIDif BASDAIDif ASDASDif PCRDif
## Min. :-8.9000 Min. :-7.0000 Min. :-3.4650 Min. :-88.

```



```

40
## 1st Qu.: -1.8000 1st Qu.: -1.1000 1st Qu.: -1.0300 1st Qu.: -5.
40
## Median : -0.2000 Median : -0.1000 Median : -0.2247 Median : -0.
20
## Mean : -0.5667 Mean : -0.5354 Mean : -0.3908 Mean : -4.
14
## 3rd Qu.: 0.6250 3rd Qu.: 0.4750 3rd Qu.: 0.2567 3rd Qu.: 1.
00
## Max. : 6.2000 Max. : 5.2000 Max. : 1.9947 Max. : 28.
00
## NA's :17 NA's :19 NA's :28 NA's :20

```

```

library(psych)
describe(base_rx2)

```

```

##          vars  n mean   sd median trimmed  mad  m
in      max
## EDAD          1 101 46.68 11.92 46.00  46.37 11.86 21.
00 75.00
## SEXO*         2 101  1.19  0.39  1.00  1.11  0.00  1.
00  2.00
## IMC           3  95 26.38  3.74 26.02  26.18  3.49 19.
37 40.83
## IMC>30*      4  95  1.12  0.32  1.00  1.03  0.00  1.
00  2.00
## TSINT        5  97 17.76 12.81 15.00  16.65 13.34  0.
00 52.00
## HLAB27*      6  99  1.13  0.34  1.00  1.05  0.00  1.
00  2.00
## EA*          7 101  1.15  0.36  1.00  1.06  0.00  1.
00  2.00
## TABACO_Kat*  8 101  1.31  0.46  1.00  1.26  0.00  1.
00  2.00
## VSG          9  98 26.65 22.53 19.00  24.02 19.27  0.
50 115.00
## BASMI       10  82  3.09  1.68  2.87  2.98  1.82  0.
54  7.27
## UVEITIS*    11 101  1.22  0.41  1.00  1.15  0.00  1.
00  2.00
## PSORIASIS*  12  99  1.07  0.26  1.00  1.00  0.00  1.
00  2.00
## ENF_INF*    13  99  1.07  0.26  1.00  1.00  0.00  1.
00  2.00
## TTOBIO*    14 101  2.13  1.07  2.00  2.00  1.48  1.
00  5.00
## NUM_TB_PREV* 15 101  1.32  0.58  1.00  1.20  0.00  1.
00  3.00
## MOTIVO*    16 101  1.31  0.60  1.00  1.17  0.00  1.
00  3.00
## TIEMBIO    17 100 45.51 40.09 42.50  41.90 51.89  0.
00 132.00
## BIO_INICI*  18 100  1.75  0.44  2.00  1.81  0.00  1.
00  2.00

```

## Biol_4_kat*	19	101	1.46	0.50	1.00	1.44	0.00	1.
00 2.00								
## Interval	20	100	3.45	0.98	3.00	3.42	1.48	2.
00 6.00								
## AINE*	21	99	1.58	0.50	2.00	1.59	0.00	1.
00 2.00								
## FAME*	22	99	1.18	0.39	1.00	1.11	0.00	1.
00 2.00								
## i_BASDAI	23	98	5.33	1.83	5.40	5.38	1.70	0.
70 8.80								
## EGP	24	98	4.03	2.78	3.00	3.88	2.97	0.
00 10.00								
## EGM	25	99	3.45	2.40	3.00	3.27	2.97	0.
00 10.00								
## mSASSS_Progress	26	101	1.98	3.75	0.00	1.12	0.00	-2.
50 21.50								
## New_Syn_mSASSS*	27	101	1.20	0.40	1.00	1.12	0.00	1.
00 2.00								
## Prog_Syn_mSASSS*	28	101	1.11	0.31	1.00	1.01	0.00	1.
00 2.00								
## mSASSS_Progress_Kat*	29	101	1.30	0.46	1.00	1.25	0.00	1.
00 2.00								
## New_Prog_Syn_Msasss*	30	101	1.22	0.41	1.00	1.15	0.00	1.
00 2.00								
## Mind_2_Synd_FL*	31	101	1.54	0.50	2.00	1.56	0.00	1.
00 2.00								
## Mind_2_Synd_BL*	32	101	1.50	0.50	2.00	1.51	0.00	1.
00 2.00								
## PCR_INICIOB	33	99	18.70	19.39	11.10	15.72	13.49	0.
00 98.00								
## i_ASIDAS	34	99	3.40	0.85	3.50	3.41	0.89	1.
50 5.30								
## HVGP	35	101	5.92	1.99	6.00	6.02	1.48	1.
00 10.00								
## BASFIDif	36	84	-0.57	2.42	-0.20	-0.47	1.63	-8.
90 6.20								
## BASDAIDif	37	82	-0.54	1.99	-0.10	-0.37	1.33	-7.
00 5.20								
## ASDASDif	38	73	-0.39	1.15	-0.22	-0.32	0.77	-3.
46 1.99								
## PCRDif	39	81	-4.14	13.70	-0.20	-1.86	3.41	-88.
40 28.00								
##			range	skew	kurtosis	se		
## EDAD	54.00		0.27	-0.59	1.19			
## SEXO*	1.00		1.57	0.48	0.04			
## IMC	21.46		0.84	1.71	0.38			
## IMC>30*	1.00		2.36	3.63	0.03			
## TSINT	52.00		0.69	-0.29	1.30			
## HLAB27*	1.00		2.15	2.65	0.03			
## EA*	1.00		1.95	1.81	0.04			
## TABACO_Kat*	1.00		0.82	-1.33	0.05			
## VSG	114.50		1.22	1.58	2.28			
## BASMI	6.74		0.47	-0.76	0.19			
## UVEITIS*	1.00		1.35	-0.19	0.04			

```

## PSORIASIS*          1.00  3.30    8.97  0.03
## ENF_INF*           1.00  3.30    8.97  0.03
## TTOBIO*            4.00  0.90    0.04  0.11
## NUM_TB_PREV*       2.00  1.64    1.60  0.06
## MOTIVO*            2.00  1.75    1.88  0.06
## TIEMBIO           132.00  0.50   -0.99  4.01
## BIO_INICI*         1.00 -1.14   -0.71  0.04
## Biol_4_kat*        1.00  0.18   -1.99  0.05
## Interval           4.00  0.23   -0.73  0.10
## AINE*              1.00 -0.30   -1.93  0.05
## FAME*              1.00  1.62    0.65  0.04
## i_BASDAI           8.10 -0.26   -0.35  0.18
## EGP                10.00  0.38   -1.10  0.28
## EGM                10.00  0.62   -0.37  0.24
## mSASSS_Progress    24.00  2.69    8.37  0.37
## New_Syn_mSASSS*    1.00  1.49    0.23  0.04
## Prog_Syn_mSASSS*   1.00  2.47    4.16  0.03
## mSASSS_Progress_Kat* 1.00  0.88   -1.25  0.05
## New_Prog_Syn_Msasss* 1.00  1.35   -0.19  0.04
## Mind_2_Synd_FL*    1.00 -0.18   -1.99  0.05
## Mind_2_Synd_BL*    1.00 -0.02   -2.02  0.05
## PCR_INICIOB       98.00  1.79    3.82  1.95
## i_ASIDAS           3.80 -0.15   -0.58  0.08
## HVGP               9.00 -0.41   -0.29  0.20
## BASFIDif          15.10 -0.45    1.67  0.26
## BASDAIDif         12.20 -0.73    1.71  0.22
## ASDASDif           5.46 -0.58    0.28  0.13
## PCRDif            116.40 -3.28   16.60  1.52

```

`describe(base_rx3)`

```

##          vars   n  mean   sd median trimmed  mad   min
max range
## EDAD          1 101 46.68 11.92  46.00  46.37 11.86 21.00  7
5.00 54.00
## SEXO*         2 101  1.19  0.39   1.00   1.11  0.00  1.00
2.00  1.00
## IMC           3  95 26.38  3.74  26.02  26.18  3.49 19.37  4
0.83 21.46
## IMC>30*       4  95  1.12  0.32   1.00   1.03  0.00  1.00
2.00  1.00
## TSINT         5  97 17.76 12.81  15.00  16.65 13.34  0.00  5
2.00 52.00
## HLAB27*       6  99  1.13  0.34   1.00   1.05  0.00  1.00
2.00  1.00
## EA*           7 101  1.15  0.36   1.00   1.06  0.00  1.00
2.00  1.00
## TABACO_Kat*   8 101  1.31  0.46   1.00   1.26  0.00  1.00
2.00  1.00
## VSG           9  98 26.65 22.53  19.00  24.02 19.27  0.50 11
5.00 114.50
## BASMI        10  82  3.09  1.68   2.87   2.98  1.82  0.54
7.27  6.74
## UVEITIS*     11 101  1.22  0.41   1.00   1.15  0.00  1.00

```

2.00	1.00									
## PSORIASIS*		12	99	1.07	0.26	1.00	1.00	0.00	1.00	
2.00	1.00									
## ENF_INF*		13	99	1.07	0.26	1.00	1.00	0.00	1.00	
2.00	1.00									
## TTOBIO*		14	101	2.13	1.07	2.00	2.00	1.48	1.00	
5.00	4.00									
## NUM_TB_PREV*		15	101	1.32	0.58	1.00	1.20	0.00	1.00	
3.00	2.00									
## MOTIVO*		16	101	1.31	0.60	1.00	1.17	0.00	1.00	
3.00	2.00									
## TIEMBIO		17	100	45.51	40.09	42.50	41.90	51.89	0.00	13
2.00	132.00									
## BIO_INICI*		18	100	1.75	0.44	2.00	1.81	0.00	1.00	
2.00	1.00									
## Biol_4_kat*		19	101	1.46	0.50	1.00	1.44	0.00	1.00	
2.00	1.00									
## Interval		20	100	3.45	0.98	3.00	3.42	1.48	2.00	
6.00	4.00									
## AINE*		21	99	1.58	0.50	2.00	1.59	0.00	1.00	
2.00	1.00									
## FAME*		22	99	1.18	0.39	1.00	1.11	0.00	1.00	
2.00	1.00									
## i_BASDAI		23	98	5.33	1.83	5.40	5.38	1.70	0.70	
8.80	8.10									
## EGP		24	98	4.03	2.78	3.00	3.88	2.97	0.00	1
0.00	10.00									
## EGM		25	99	3.45	2.40	3.00	3.27	2.97	0.00	1
0.00	10.00									
## mSASSS_Progress		26	101	1.98	3.75	0.00	1.12	0.00	-2.50	2
1.50	24.00									
## Mind_2_Synd_FL*		27	101	1.54	0.50	2.00	1.56	0.00	1.00	
2.00	1.00									
## Mind_2_Synd_BL*		28	101	1.50	0.50	2.00	1.51	0.00	1.00	
2.00	1.00									
## PCR_INICIOB		29	99	18.70	19.39	11.10	15.72	13.49	0.00	9
8.00	98.00									
## i_ASIDAS		30	99	3.40	0.85	3.50	3.41	0.89	1.50	
5.30	3.80									
## HVGP		31	101	5.92	1.99	6.00	6.02	1.48	1.00	1
0.00	9.00									
## BASFIDif		32	84	-0.57	2.42	-0.20	-0.47	1.63	-8.90	
6.20	15.10									
## BASDAIDif		33	82	-0.54	1.99	-0.10	-0.37	1.33	-7.00	
5.20	12.20									
## ASDASdif		34	73	-0.39	1.15	-0.22	-0.32	0.77	-3.46	
1.99	5.46									
## PCRDif		35	81	-4.14	13.70	-0.20	-1.86	3.41	-88.40	2
8.00	116.40									
##				skew	kurtosis	se				
## EDAD				0.27	-0.59	1.19				
## SEXO*				1.57	0.48	0.04				
## IMC				0.84	1.71	0.38				
## IMC>30*				2.36	3.63	0.03				

```

## TSINT          0.69    -0.29  1.30
## HLAB27*       2.15     2.65  0.03
## EA*           1.95     1.81  0.04
## TABACO_Kat*  0.82    -1.33  0.05
## VSG           1.22     1.58  2.28
## BASMI         0.47    -0.76  0.19
## UVEITIS*     1.35    -0.19  0.04
## PSORIASIS*   3.30     8.97  0.03
## ENF_INF*     3.30     8.97  0.03
## TTOBIO*      0.90     0.04  0.11
## NUM_TB_PREV* 1.64     1.60  0.06
## MOTIVO*      1.75     1.88  0.06
## TIEMBIO      0.50    -0.99  4.01
## BIO_INICI*   -1.14    -0.71  0.04
## Biol_4_kat*  0.18    -1.99  0.05
## Interval     0.23    -0.73  0.10
## AINE*       -0.30    -1.93  0.05
## FAME*       1.62     0.65  0.04
## i_BASDAI    -0.26    -0.35  0.18
## EGP         0.38    -1.10  0.28
## EGM         0.62    -0.37  0.24
## mSASSS_Progress 2.69     8.37  0.37
## Mind_2_Synd_FL* -0.18    -1.99  0.05
## Mind_2_Synd_BL* -0.02    -2.02  0.05
## PCR_INICIOB  1.79     3.82  1.95
## i_ASIDAS    -0.15    -0.58  0.08
## HVGP       -0.41    -0.29  0.20
## BASFIDif    -0.45     1.67  0.26
## BASDAIDif   -0.73     1.71  0.22
## ASDASDif   -0.58     0.28  0.13
## PCRDif     -3.28    16.60  1.52

```

```

base_rx.num<-base_rx[,c(-1,-3,-5,-7,-8,-9,-12,-13,-14,-15,-16,-17,-19,
-20,-22,-23,-58,-59,-60,-61,-62,-63)] # data frame con únicamente variables numéricas

```

```

library(psych)
describe(base_rx.num)

```

```

##          vars  n  mean   sd median trimmed  mad  min
max range
## EDAD          1 101 46.68 11.92  46.00   46.37 11.86 21.00  75
.00  54.00
## IMC           2  95 26.38  3.74  26.02   26.18  3.49 19.37  40
.83  21.46
## TSINT         3  97 17.76 12.81  15.00   16.65 13.34  0.00  52
.00  52.00
## VSG           4  98 26.65 22.53  19.00   24.02 19.27  0.50 115
.00 114.50
## BASMI         5  82  3.09  1.68  2.87    2.98  1.82  0.54  7
.27  6.74
## TIEMBIO       6 100 45.51 40.09  42.50   41.90 51.89  0.00 132
.00 132.00
## Interval     7 100  3.45  0.98  3.00    3.42  1.48  2.00  6

```

.00	4.00									
## i_BASDAI		8	98	5.33	1.83	5.40	5.38	1.70	0.70	8
.80	8.10									
## BASDAI_0		9	97	3.67	2.28	3.20	3.54	2.52	0.20	8
.80	8.60									
## BASDAI_1		10	97	2.90	1.87	2.70	2.79	2.22	0.00	8
.50	8.50									
## BASDAI_2		11	97	3.42	6.59	2.50	2.73	2.22	0.00	65
.00	65.00									
## BASDAI_3		12	93	3.11	2.02	3.00	2.99	2.67	0.00	8
.00	8.00									
## BASDAI_4		13	91	3.07	2.01	2.80	2.96	2.37	0.00	8
.00	8.00									
## BASDAI_5		14	84	2.92	1.91	2.55	2.77	2.00	0.00	8
.00	8.00									
## BASDAI_6		15	85	2.92	1.90	2.80	2.79	2.52	0.00	7
.30	7.30									
## ASDAS_0		16	96	2.22	1.20	2.06	2.15	1.23	0.23	5
.07	4.85									
## ASDAS_1		17	88	1.82	1.10	1.57	1.72	0.80	0.25	7
.13	6.88									
## ASDAS_2		18	92	1.77	0.91	1.83	1.74	0.87	0.00	4
.13	4.13									
## ASDAS_3		19	87	1.82	0.93	1.77	1.78	0.95	0.05	3
.98	3.93									
## ASDAS_4		20	79	1.81	0.96	1.67	1.78	0.88	0.00	4
.45	4.45									
## ASDAS_5		21	71	1.73	0.92	1.64	1.66	0.88	0.03	4
.06	4.02									
## ASDAS_6		22	77	1.81	0.93	1.68	1.77	0.97	0.10	3
.99	3.89									
## PCR_0		23	94	8.18	13.25	3.35	5.12	4.23	0.00	88
.70	88.70									
## PCR_1		24	98	4.36	8.03	2.00	2.83	1.93	0.00	61
.30	61.30									
## PCR_2		25	99	4.36	5.95	2.00	3.22	2.37	0.00	33
.90	33.90									
## PCR_3		26	97	4.45	6.33	1.90	3.19	2.52	0.00	37
.00	37.00									
## PCR_4		27	94	5.37	10.46	2.05	2.94	2.89	0.00	68
.50	68.50									
## PCR_5		28	91	4.12	8.00	1.40	2.73	1.82	0.00	65
.90	65.90									
## PCR_6		29	87	4.42	6.81	2.00	3.05	2.52	0.00	39
.70	39.70									
## EGP		30	98	4.03	2.78	3.00	3.88	2.97	0.00	10
.00	10.00									
## EGM		31	99	3.45	2.40	3.00	3.27	2.97	0.00	10
.00	10.00									
## BASFI_0		32	98	3.93	2.70	3.85	3.80	3.34	0.00	9
.30	9.30									
## BASFI_1		33	99	3.06	2.31	2.70	2.87	2.22	0.00	8
.80	8.80									
## BASFI_2		34	101	2.67	2.22	2.33	2.48	2.80	0.00	7

.89	7.89									
## BASFI_3		35	93	2.88	2.20	2.80	2.73	2.52	0.00	7
.90	7.90									
## BASFI_4		36	89	3.34	2.36	3.50	3.24	2.97	0.00	8
.40	8.40									
## BASFI_5		37	79	3.12	2.34	3.20	2.98	2.82	0.00	8
.10	8.10									
## BASFI_6		38	85	3.31	2.45	3.10	3.17	2.97	0.00	8
.30	8.30									
## mSASSS_Mean_BL		39	101	16.46	20.92	5.00	12.80	7.41	0.00	72
.00	72.00									
## mSASSS_Mean_FL		40	101	18.44	22.12	8.00	15.10	11.86	0.00	72
.00	72.00									
## mSASSS_Progress		41	101	1.98	3.75	0.00	1.12	0.00	-2.50	21
.50	24.00									
## PCR_INICIOB		42	99	18.70	19.39	11.10	15.72	13.49	0.00	98
.00	98.00									
## i_ASDas		43	99	3.40	0.85	3.50	3.41	0.89	1.50	5
.30	3.80									
## HVGP		44	101	5.92	1.99	6.00	6.02	1.48	1.00	10
.00	9.00									
##				skew	kurtosis	se				
## EDAD				0.27	-0.59	1.19				
## IMC				0.84	1.71	0.38				
## TSINT				0.69	-0.29	1.30				
## VSG				1.22	1.58	2.28				
## BASMI				0.47	-0.76	0.19				
## TIEMBIO				0.50	-0.99	4.01				
## Interval				0.23	-0.73	0.10				
## i_BASDAI				-0.26	-0.35	0.18				
## BASDAI_0				0.47	-0.66	0.23				
## BASDAI_1				0.53	-0.32	0.19				
## BASDAI_2				8.41	75.78	0.67				
## BASDAI_3				0.36	-0.88	0.21				
## BASDAI_4				0.42	-0.87	0.21				
## BASDAI_5				0.62	-0.39	0.21				
## BASDAI_6				0.46	-0.79	0.21				
## ASDAS_0				0.48	-0.49	0.12				
## ASDAS_1				1.71	5.06	0.12				
## ASDAS_2				0.28	-0.44	0.09				
## ASDAS_3				0.28	-0.70	0.10				
## ASDAS_4				0.38	-0.35	0.11				
## ASDAS_5				0.61	-0.13	0.11				
## ASDAS_6				0.36	-0.60	0.11				
## PCR_0				3.34	14.15	1.37				
## PCR_1				4.88	27.71	0.81				
## PCR_2				2.75	9.26	0.60				
## PCR_3				2.70	8.90	0.64				
## PCR_4				3.90	16.88	1.08				
## PCR_5				5.52	37.64	0.84				
## PCR_6				3.35	13.58	0.73				
## EGP				0.38	-1.10	0.28				
## EGM				0.62	-0.37	0.24				
## BASFI_0				0.27	-1.02	0.27				

```
## BASFI_1      0.67    -0.47  0.23
## BASFI_2      0.49    -0.81  0.22
## BASFI_3      0.46    -0.79  0.23
## BASFI_4      0.24    -1.05  0.25
## BASFI_5      0.35    -0.92  0.26
## BASFI_6      0.35    -0.99  0.27
## mSASSS_Mean_BL 1.18     0.09  2.08
## mSASSS_Mean_FL 0.98    -0.42  2.20
## mSASSS_Progress 2.69     8.37  0.37
## PCR_INICIOB  1.79     3.82  1.95
## i_ASDas      -0.15    -0.58  0.08
## HVGP         -0.41    -0.29  0.20
```

```
base_rx.num1 <-
  base_rx.num %>%
  mutate( BASFI_Dif = (BASFI_6-BASFI_0), BASDAI_Dif = (BASDAI_6-BASDAI_0
), ASDAS_Dif = (ASDAS_6-ASDAS_0), PCRDif = (PCR_6-PCR_0))
summary(base_rx.num1)
```

```
##          EDAD          IMC          TSINT          VSG
## Min.   :21.00   Min.   :19.37   Min.    : 0.00   Min.    : 0.50
## 1st Qu.:38.00   1st Qu.:24.05   1st Qu.: 8.00   1st Qu.: 9.00
## Median :46.00   Median :26.02   Median :15.00   Median :19.00
## Mean   :46.68   Mean   :26.38   Mean   :17.76   Mean   :26.65
## 3rd Qu.:54.00   3rd Qu.:28.68   3rd Qu.:26.00   3rd Qu.:38.75
## Max.   :75.00   Max.   :40.83   Max.   :52.00   Max.   :115.00
##
##          NA's      :6      NA's      :4      NA's      :3
##          BASMI          TIEMBIO          Interval          i_BASDAI
## Min.   :0.5378   Min.    : 0.00   Min.    :2.00   Min.    :0.700
## 1st Qu.:1.6458   1st Qu.: 8.00   1st Qu.:3.00   1st Qu.:4.225
## Median :2.8728   Median :42.50   Median :3.00   Median :5.400
## Mean   :3.0915   Mean   :45.51   Mean   :3.45   Mean   :5.326
## 3rd Qu.:4.0864   3rd Qu.:76.75   3rd Qu.:4.00   3rd Qu.:6.400
## Max.   :7.2729   Max.   :132.00   Max.   :6.00   Max.   :8.800
## NA's   :19      NA's    :1      NA's    :1      NA's    :3
##          BASDAI_0          BASDAI_1          BASDAI_2          BASDAI_3
## Min.   :0.20   Min.   :0.000   Min.    : 0.000   Min.    :0.000
## 1st Qu.:2.00   1st Qu.:1.400   1st Qu.: 1.400   1st Qu.:1.200
## Median :3.20   Median :2.700   Median : 2.500   Median :3.000
## Mean   :3.67   Mean   :2.902   Mean   : 3.419   Mean   :3.108
## 3rd Qu.:5.10   3rd Qu.:4.400   3rd Qu.: 4.200   3rd Qu.:4.800
## Max.   :8.80   Max.   :8.500   Max.   :65.000   Max.   :8.000
## NA's   :4      NA's    :4      NA's    :4      NA's    :8
##          BASDAI_4          BASDAI_5          BASDAI_6          ASDAS_0
## Min.   :0.000   Min.   :0.000   Min.    :0.000   Min.    :0.2257
## 1st Qu.:1.200   1st Qu.:1.275   1st Qu.:1.100   1st Qu.:1.3874
## Median :2.800   Median :2.550   Median :2.800   Median :2.0565
## Mean   :3.075   Mean   :2.915   Mean   :2.924   Mean   :2.2160
## 3rd Qu.:4.600   3rd Qu.:4.200   3rd Qu.:4.600   3rd Qu.:2.9248
## Max.   :8.000   Max.   :8.000   Max.   :7.300   Max.   :5.0749
## NA's   :10     NA's    :17     NA's    :16     NA's    :5
##          ASDAS_1          ASDAS_2          ASDAS_3          ASDAS_4
## Min.   :0.2471   Min.   :0.001737   Min.    :0.04892   Min.    :0.00
058
```


##	1st Qu.:1.1572	1st Qu.:1.088078	1st Qu.:1.16497	1st Qu.:1.17363
##	Median :1.5738	Median :1.827288	Median :1.76656	Median :1.66660
##	Mean :1.8249	Mean :1.773470	Mean :1.81523	Mean :1.81274
##	3rd Qu.:2.4204	3rd Qu.:2.316346	3rd Qu.:2.47797	3rd Qu.:2.44173
##	Max. :7.1255	Max. :4.131246	Max. :3.98374	Max. :4.45150
##	NA's :13	NA's :9	NA's :14	NA's :22
##	ASDAS_5	ASDAS_6	PCR_0	PCR_1
##	Min. :0.0338	Min. :0.096	Min. : 0.000	Min. : 0.000
##	1st Qu.:1.0522	1st Qu.:1.143	1st Qu.: 1.000	1st Qu.: 1.000
##	Median :1.6400	Median :1.680	Median : 3.350	Median : 2.000
##	Mean :1.7329	Mean :1.807	Mean : 8.181	Mean : 4.358
##	3rd Qu.:2.1781	3rd Qu.:2.397	3rd Qu.: 9.100	3rd Qu.: 4.675
##	Max. :4.0579	Max. :3.988	Max. :88.700	Max. :61.300
##	NA's :30	NA's :24	NA's :7	NA's :3
##	PCR_2	PCR_3	PCR_4	PCR_5
##	Min. : 0.00	Min. : 0.000	Min. : 0.000	Min. : 0.000
##	1st Qu.: 0.90	1st Qu.: 0.750	1st Qu.: 0.700	1st Qu.: 0.615
##	Median : 2.00	Median : 1.900	Median : 2.050	Median : 1.400
##	Mean : 4.36	Mean : 4.452	Mean : 5.368	Mean : 4.118
##	3rd Qu.: 5.60	3rd Qu.: 5.600	3rd Qu.: 5.375	3rd Qu.: 6.000
##	Max. :33.90	Max. :37.000	Max. :68.500	Max. :65.900
##	NA's :2	NA's :4	NA's :7	NA's :10
##	PCR_6	EGP	EGM	BASFI_0
##	Min. : 0.00	Min. : 0.000	Min. : 0.000	Min. :0.000
##	1st Qu.: 0.60	1st Qu.: 2.000	1st Qu.: 1.500	1st Qu.:1.525
##	Median : 2.00	Median : 3.000	Median : 3.000	Median :3.850
##	Mean : 4.42	Mean : 4.031	Mean : 3.455	Mean :3.931
##	3rd Qu.: 5.95	3rd Qu.: 6.000	3rd Qu.: 5.000	3rd Qu.:5.975
##	Max. :39.70	Max. :10.000	Max. :10.000	Max. :9.300
##	NA's :14	NA's :3	NA's :2	NA's :3
##	BASFI_1	BASFI_2	BASFI_3	BASFI_4
##	BASFI_5			
##	Min. :0.000	Min. :0.000	Min. :0.000	Min. :0.00
##	1st Qu.:1.200	1st Qu.:0.550	1st Qu.:1.000	1st Qu.:1.30
##	Median :2.700	Median :2.330	Median :2.800	Median :3.50
##	Mean :3.059	Mean :2.667	Mean :2.882	Mean :3.34
##	3rd Qu.:4.550	3rd Qu.:4.340	3rd Qu.:4.500	3rd Qu.:5.00
##	Max. :8.800	Max. :7.890	Max. :7.900	Max. :8.40
##	NA's :2		NA's :8	NA's :12
##	BASFI_6	mSASSS_Mean_BL	mSASSS_Mean_FL	mSASSS_Progress
##	Min. :0.000	Min. : 0.00	Min. : 0.00	Min. :-2.50
##	1st Qu.:1.100	1st Qu.: 0.00	1st Qu.: 0.00	1st Qu.: 0.00

```

## Median :3.100 Median : 5.00 Median : 8.00 Median : 0.00
## Mean :3.313 Mean :16.46 Mean :18.44 Mean : 1.98
## 3rd Qu.:5.100 3rd Qu.:23.00 3rd Qu.:30.00 3rd Qu.: 3.00
## Max. :8.300 Max. :72.00 Max. :72.00 Max. :21.50
## NA's :16
## PCR_INICIOB i_ASIDAS HVGP BASFIDif
## Min. : 0.00 Min. :1.500 Min. : 1.000 Min. : -8.9000
## 1st Qu.: 4.60 1st Qu.:2.700 1st Qu.: 5.000 1st Qu.: -1.8000
## Median :11.10 Median :3.500 Median : 6.000 Median : -0.2000
## Mean :18.70 Mean :3.402 Mean : 5.923 Mean : -0.5667
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd Qu.: 7.000 3rd Qu.: 0.6250
## Max. :98.00 Max. :5.300 Max. :10.000 Max. : 6.2000
## NA's :2 NA's :2 NA's :17
## BASDAIDif ASDASdif PCRdif
## Min. : -7.0000 Min. : -3.4650 Min. : -88.40
## 1st Qu.: -1.1000 1st Qu.: -1.0300 1st Qu.: -5.40
## Median : -0.1000 Median : -0.2247 Median : -0.20
## Mean : -0.5354 Mean : -0.3908 Mean : -4.14
## 3rd Qu.: 0.4750 3rd Qu.: 0.2567 3rd Qu.: 1.00
## Max. : 5.2000 Max. : 1.9947 Max. : 28.00
## NA's :19 NA's :28 NA's :20

```

```

base_rx.num2<-base_rx.num1[,c(-9:-29,-32:-40)]
summary(base_rx.num2)

```

```

## EDAD IMC TSINT VSG
## Min. :21.00 Min. :19.37 Min. : 0.00 Min. : 0.50
## 1st Qu.:38.00 1st Qu.:24.05 1st Qu.: 8.00 1st Qu.: 9.00
## Median :46.00 Median :26.02 Median :15.00 Median : 19.00
## Mean :46.68 Mean :26.38 Mean :17.76 Mean : 26.65
## 3rd Qu.:54.00 3rd Qu.:28.68 3rd Qu.:26.00 3rd Qu.: 38.75
## Max. :75.00 Max. :40.83 Max. :52.00 Max. :115.00
## NA's :6 NA's :4 NA's :3
## BASMI TIEMBIO Interval i_BASDAI
## Min. :0.5378 Min. : 0.00 Min. :2.00 Min. :0.700
## 1st Qu.:1.6458 1st Qu.: 8.00 1st Qu.:3.00 1st Qu.:4.225
## Median :2.8728 Median : 42.50 Median :3.00 Median :5.400
## Mean :3.0915 Mean : 45.51 Mean :3.45 Mean :5.326
## 3rd Qu.:4.0864 3rd Qu.: 76.75 3rd Qu.:4.00 3rd Qu.:6.400
## Max. :7.2729 Max. :132.00 Max. :6.00 Max. :8.800
## NA's :19 NA's :1 NA's :1 NA's :3
## EGP EGM mSASSS_Progress PCR_INICIOB
## Min. : 0.000 Min. : 0.000 Min. : -2.50 Min. : 0.00
## 1st Qu.: 2.000 1st Qu.: 1.500 1st Qu.: 0.00 1st Qu.: 4.60
## Median : 3.000 Median : 3.000 Median : 0.00 Median :11.10
## Mean : 4.031 Mean : 3.455 Mean : 1.98 Mean :18.70
## 3rd Qu.: 6.000 3rd Qu.: 5.000 3rd Qu.: 3.00 3rd Qu.:28.05
## Max. :10.000 Max. :10.000 Max. :21.50 Max. :98.00
## NA's :3 NA's :2 NA's :2
## i_ASIDAS HVGP BASFIDif BASDAIDif
## Min. :1.500 Min. : 1.000 Min. : -8.9000 Min. : -7.0000
## 1st Qu.:2.700 1st Qu.: 5.000 1st Qu.: -1.8000 1st Qu.: -1.1000
## Median :3.500 Median : 6.000 Median : -0.2000 Median : -0.1000
## Mean :3.402 Mean : 5.923 Mean : -0.5667 Mean : -0.5354

```

```
## 3rd Qu.:4.000 3rd Qu.: 7.000 3rd Qu.: 0.6250 3rd Qu.: 0.4750
## Max. :5.300 Max. :10.000 Max. : 6.2000 Max. : 5.2000
## NA's :2 NA's :17 NA's :19
## ASDASDif PCRdif
## Min. :-3.4650 Min. :-88.40
## 1st Qu.:-1.0300 1st Qu.: -5.40
## Median :-0.2247 Median : -0.20
## Mean :-0.3908 Mean : -4.14
## 3rd Qu.: 0.2567 3rd Qu.: 1.00
## Max. : 1.9947 Max. : 28.00
## NA's :28 NA's :20
```

```
library(dplyr)
library(tidyr)
library(broom)
```

```
base_rx.num2 <-base_rx.num2[complete.cases(base_rx.num2), ]
sumstat <- base_rx.num2 %>%
  # Select and rename five variables
  select(
    `Edad` = EDAD,
    `IMC` = IMC,
    `Tiempo sintomas` = TSINT,
    `VSG` = VSG,
    `BASMI` = BASMI,
    `Tiempo con tratamiento` = TIEMBIO,
    `Intervalo entre radiografías` = Interval,
    `BASDAI al inicio del tratamiento` = i_BASDAI,
    `Dolor global del paciente` = EGP,
    `Dolor global en la opinión del médico` = EGM,
    `Diferencia entre el mSASSS al inicio y final` = mSASSS_Progre
ss,
    `PCR al inicio del tratamiento` = PCR_INICIOB,
    `ASDAS al inicio del tratamiento` = i_ASIDAS,
    `Valoración global del paciente al inicio del tratamiento` = H
VGP,
    `Diferencia entre el BASFI al inicio y final` = BASFIDif,
    `Diferencia entre el BASDAI al inicio final` = BASDAIDif,
    `Diferencia entre ASDAS al inicio y final` = ASDASDif,
    `Diferencia entre PCR al inicio y final` = PCRDif
  ) %>%
  # Find the mean, st. dev., min, and max for each variable
  summarise_each(funs(mean, median, sd, min, max, var)) %>%

  # Move summary stats to columns
  gather(key, value, everything()) %>%
  separate(key, into = c("variable", "stat"), sep = "_") %>%
  spread(stat, value) %>%

  # Set order of summary statistics
  select(variable, mean, median, sd, min, max, var) %>%

  # Round all numeric variables to one decimal point
  mutate_each(funs(round(., 1)), -variable)
```

```
## Warning: `summarise_each()` is deprecated as of dplyr 0.7.0.
## Please use `across()` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_warnings()` to see where this warning was generated.
```

```
## Warning: `funs()` is deprecated as of dplyr 0.8.0.
## Please use a list of either functions or lambdas:
##
## # Simple named list:
## list(mean = mean, median = median)
##
## # Auto named with `tibble::lst()`:
## tibble::lst(mean, median)
##
## # Using lambdas
## list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_warnings()` to see where this warning was generated.
```

```
## Warning: `mutate_each()` is deprecated as of dplyr 0.7.0.
## Please use `across()` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_warnings()` to see where this warning was generated.
```

sumstat

```
## # A tibble: 18 x 7
##   variable          mean median    sd  min
max   var
##   <chr>             <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
## 1 ASDAS al inicio del tratamiento      3.4   3.4  0.7  2.1
5.3  0.5
## 2 BASDAI al inicio del tratamiento      5     5    1.9  0.7
8.5  3.4
## 3 BASMI      3.2   2.9  1.7  0.5
6.6  2.7
## 4 Diferencia entre ASDAS al inicio y fin~ -0.4  -0.3  1.1 -3.5
1.8  1.3
## 5 Diferencia entre el BASDAI al inicio f~ -0.4   0    2    -7
5.2  4
## 6 Diferencia entre el BASFI al inicio y ~ -0.3  -0.1  2.1 -5.5
6.2  4.2
## 7 Diferencia entre el mSASSS al inicio y~  2.2   0    4.4 -2.5
21.5 19.4
## 8 Diferencia entre PCR al inicio y final  -4.5  -0.4  15  -88.4
28  224.
## 9 Dolor global del paciente      3.7   3    2.6  0
10  6.7
## 10 Dolor global en la opinión del médico  3.3   3    2.3  0
8    5.1
## 11 Edad      48.2  47   11.5  25
```

```

75    133.
## 12 IMC                26.9  26.7  4    19.8
40.8  16.1
## 13 Intervalo entre radiografías
5     0.8
## 14 PCR al inicio del tratamiento
88.7  318.
## 15 Tiempo con tratamiento
124   1425.
## 16 Tiempo síntomas
52    154.
## 17 Valoración global del paciente al inic~
10    3.8
## 18 VSG                28.5  24.8  24.2  0.5
115   588.

```

```

write.table(sumstat, file = "sumstats.txt", sep = ",", quote = FALSE,
row.names = F)

```

```

library(dlookr)

```

```

## Loading required package: mice

```

```

##

```

```

## Attaching package: 'mice'

```

```

## The following object is masked from 'package:stats':

```

```

##

```

```

##   filter

```

```

## The following objects are masked from 'package:base':

```

```

##

```

```

##   cbind, rbind

```

```

## Registered S3 method overwritten by 'quantmod':

```

```

##   method          from

```

```

##   as.zoo.data.frame zoo

```

```

##

```

```

## Attaching package: 'dlookr'

```

```

## The following object is masked from 'package:psych':

```

```

##

```

```

##   describe

```

```

## The following object is masked from 'package:base':

```

```

##

```

```

##   transform

```

```

library(ISLR)

```

```

library(SmartEDA)

```

```

## Registered S3 method overwritten by 'GGally':

```

```

##   method from

```

```

##   +.gg  ggplot2

```

```

library(dplyr)
eda_report(base_rx.num2, mSASSS_Progress, output_format = "html")

## Warning: Unquoting language objects with `!!!` is deprecated as of
rlang 0.4.0.
## Please use `!!` instead.
##
## # Bad:
## dplyr::select(data, !!!enquo(x))
##
## # Good:
## dplyr::select(data, !!enquo(x)) # Unquote single quosure
## dplyr::select(data, !!!enquos(x)) # Splice list of quosures
##
## This warning is displayed once per session.

##
##
## processing file: EDA_Report.Rmd

##
|
|
| 0%
|
| ...
| 4%
## inline R code fragments
##
##
|
| .....
| 7%
## label: setup (with options)
## List of 1
## $ include: logi FALSE
##
##
|
| .....
| 11%
## ordinary text without R code
##
##
|
| .....
| 15%
## label: environment (with options)
## List of 3
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE

##
## Attaching package: 'magrittr'

```

```

## The following object is masked from 'package:tidyr':
##
##   extract

##
## Attaching package: 'ggplot2'

## The following objects are masked from 'package:psych':
##
##   %+%, alpha

##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':
##
##   combine

##
## Attaching package: 'moments'

## The following objects are masked from 'package:dlookr':
##
##   kurtosis, skewness

##
## Attaching package: 'kableExtra'

## The following object is masked from 'package:dplyr':
##
##   group_rows

##
## |
## | .....
## | 19%
## ordinary text without R code
##
##
## |
## | .....
## | 22%
## label: udf (with options)
## List of 3
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
##
##
## |
## | .....
## | 26%
## ordinary text without R code
##
##
## |

```

```

| .....
| 30%
## label: check_variables (with options)
## List of 4
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
## $ comment: chr ""
##
##
| .....
| 33%
## inline R code fragments
##
##
| .....
| 37%
## label: info_variables (with options)
## List of 5
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
## $ comment: chr ""
## $ results: chr "asis"
##
##
| .....
| 41%
## inline R code fragments
##
##
| .....
| 44%
## label: describe_univariate (with options)
## List of 4
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
## $ comment: chr ""

## Warning in png(file, width = 1 + k * w, height = h): 'width=13, height=13' are
## unlikely values in pixels

##
| .....
| 48%
## ordinary text without R code
##

```



```

## Warning in sqrt(x): Se han producido NaNs

##
|
| .....
| 56%
## ordinary text without R code
##
##
|
| .....
| 59%
## label: correlations (with options)
## List of 4
## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
## $ comment: chr ""
##
##
|
| .....
| 63%
## ordinary text without R code
##
##
|
| .....
| 67%
## label: plot_correlations (with options)
## List of 6
## $ echo : logi FALSE
## $ warning : logi FALSE
## $ message : logi FALSE
## $ comment : chr ""
## $ fig.height: num 4
## $ fig.width : num 6
##
##
|
| .....
| 70%
## ordinary text without R code
##
##
|
| .....
| 74%
## label: numeric_variables (with options)
## List of 7
## $ echo : logi FALSE
## $ warning : logi FALSE
## $ message : logi FALSE
## $ comment : chr ""

```

```

## $ fig.height: num 4
## $ fig.width : num 8
## $ results   : chr "asis"

## `geom_smooth()` using formula 'y ~ x'

## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'
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## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'

##
|
| .....
| 78%
## ordinary text without R code
##
##
|
| .....
| 81%
## label: category_variables (with options)
## List of 7
## $ echo      : logi FALSE
## $ warning   : logi FALSE
## $ message   : logi FALSE
## $ comment   : chr ""
## $ fig.height: num 4
## $ fig.width : num 7
## $ results   : chr "asis"
##
##
|
| .....
| 85%
## ordinary text without R code
##
##
|
| .....
| 89%
## label: group_correlations (with options)
## List of 4

```

```

## $ echo : logi FALSE
## $ warning: logi FALSE
## $ message: logi FALSE
## $ comment: chr ""
##
##
|
| .....
| 93%
## ordinary text without R code
##
##
|
| .....
| 96%
## label: plot_group_correlations (with options)
## List of 6
## $ echo : logi FALSE
## $ warning : logi FALSE
## $ message : logi FALSE
## $ comment : chr ""
## $ fig.height: num 4
## $ fig.width : num 6
##
##
|
| .....
...| 100%
## ordinary text without R code

## output file: EDA_Report.knit.md

## "C:/Program Files/RStudio/bin/pandoc/pandoc" +RTS -K512m -RTS EDA_Report.utf8.md --to html4 --from markdown+autolink_bare_uris+tex_math_single_backslash --output pandoc3c586e5316fc.html --email-obfuscation none --self-contained --standalone --section-divs --table-of-contents --toc-depth 3 --template "C:/Users/maria/OneDrive/Documents/R/win-library/4.0/prettydoc/resources/templates/cayman.html" --highlight-style pygments --number-sections --include-in-header "C:\Users\maria\AppData\Local\Temp\Rtmp4GADut\rmarkdown-str3c585d172210.html" --mathjax --variable "mathjax-url:https://mathjax.rstudio.com/latest/MathJax.js?config=TeX-AMS-MML_HTMLorMML" --css "C:\Users\maria\AppData\Local\Temp\Rtmp4GADut\EDA_Report_files/style.css"

##
## Output created: C:\Users\maria\AppData\Local\Temp\Rtmp4GADut\EDA_Report.html

str(base_rx.num2)

## tibble [59 x 18] (S3: tbl_df/tbl/data.frame)
## $ EDAD : num [1:59] 63 26 57 41 40 43 60 41 51 38 ...
## $ IMC : num [1:59] 26.3 24.6 28.7 21.6 27.7 ...
## $ TSINT : num [1:59] 1 6 13 18 16 13 21 13 33 2 ...
## $ VSG : num [1:59] 50 3 24 38 25 9 10 52 48 33 ...

```

```

## $ BASMI : num [1:59] 1.083 0.979 2.947 3.265 2.202 ...
## $ TIEMBIO : num [1:59] 0 0 84 24 0 0 1 58 59 14 ...
## $ Interval : num [1:59] 4 3 4 4 4 4 3 4 3 4 ...
## $ i_BASDAI : num [1:59] 2.2 2.2 3.8 4.5 5 8 7 5 7 6 ...
## $ EGP : num [1:59] 3 1 4 2 5 9 10 2 1 6 ...
## $ EGM : num [1:59] 3 1 4 2 5 5 5 1 1 4 ...
## $ mSASSS_Progress : num [1:59] 3.5 0 1 0 1.5 0 1.5 0 0 0 ...
## $ PCR_INICIOB : num [1:59] 88.7 11.3 26 15 16 9.2 16 30 31 5 ..
.
## $ i_ASIDAS : num [1:59] 3.4 2.2 3.6 3 3.4 3.8 3.9 4.1 4.5 3.
3 ...
## $ HVGP : num [1:59] 3 1 7 4 5 5 5 8 7 7 ...
## $ BASFIDif : num [1:59] -0.5 0.7 -2.1 -1.8 -0.3 ...
## $ BASDAIDif : num [1:59] -1.4 -1.1 0 -1 0 -7 -4.6 0 1.8 -2.5
...
## $ ASDASDif : num [1:59] -3.116 -0.99 -0.547 -0.22 1.794 ...
## $ PCRDif : num [1:59] -88.4 -9.51 -6 -0.2 16.4 -8.2 -14.5
-0.1 -0.2 1 ...

```

Análisis de missings

#De La base de datos original:

```
sum(is.na(base_rx))
```

```
## [1] 353
```

```
sum(is.na(base_rx.num))
```

```
## [1] 336
```

Hay 353 valores missing, y en la base de datos de variables únicamente e numéricas hay 336.

#De La base de datos ajustada:

```
sum(is.na(base_rx2))
```

```
## [1] 147
```

```
sum(is.na(base_rx.num2))
```

```
## [1] 0
```

Hay 147 valores missing, y en la base de datos de variables únicamente e numéricas hay 128.

```
colSums(is.na(base_rx))
```

##	NUM_PAC	EDAD	SEXO
IMC			
##	0	0	0
6			
##	IMC>30	TSINT	HLAB27
EA			
##	6	4	2
0			
##	TABACO_Kat	VSG	BASMI
UVEITIS			

##	0	3	19	
0				
##	PSORIASIS	ENF_INF	TTOBIO	
NUM_TB_PREV				
##	2	2	0	
0				
##	MOTIVO	TIEMBIO	BIO_INICI	
Biol_4_kat				
##	0	1	1	
0				
##	Interval	AINE	FAME	
i_BASDAI				
##	1	2	2	
3				
##	BASDAI_0	BASDAI_1	BASDAI_2	
BASDAI_3				
##	4	4	4	
8				
##	BASDAI_4	BASDAI_5	BASDAI_6	
ASDAS_0				
##	10	17	16	
5				
##	ASDAS_1	ASDAS_2	ASDAS_3	
ASDAS_4				
##	13	9	14	
22				
##	ASDAS_5	ASDAS_6	PCR_0	
PCR_1				
##	30	24	7	
3				
##	PCR_2	PCR_3	PCR_4	
PCR_5				
##	2	4	7	
10				
##	PCR_6	EGP	EGM	
BASFI_0				
##	14	3	2	
3				
##	BASFI_1	BASFI_2	BASFI_3	
BASFI_4				
##	2	0	8	
12				
##	BASFI_5	BASFI_6	mSASSS_Mean_BL	mS
ASSS_Mean_FL				
##	22	16	0	
0				
##	mSASSS_Progress	New_Syn_mSASSS	Prog_Syn_mSASSS	mSASSS_
Progress_Kat				
##	0	0	0	
0				
##	New_Prog_Syn_Msasss	Mind_2_Synd_FL	Mind_2_Synd_BL	
PCR_INICIOB				
##	0	0	0	
2				

```

##          i_ASDas          HVGP
##          2                0

#vemos que Los missings se concentran principalmente en 5 variables nu
méricas (BASMI, BASFIDif, BASDAIDif, ASDASDif, PCRDif)

# Multivariate Imputation by Chained Equations de Las variables numéri
cas
library(mice)
miceMod <- mice(base_rx.num[, !names(base_rx.num) %in% "medv"], method
="rf")

##
## iter imp variable
##  1  1  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

```

```

do
## regression?

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do
## regression?

##   i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##   1  2  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

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## regression?

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: The

```



```

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##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      1      3  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

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## regression?

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```



```
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```

```

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##   i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##   1   5  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

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```

```

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## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      2      1  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do

```

```

## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      2      2  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to

```

```

do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDAI_5
BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 ASDAS_6
PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BASFI_0 BASFI_1
BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 2 3 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The

```

```
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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do
## regression?

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: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?
```



```

## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDA
I_5 BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 A
SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 2 4 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

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## regression?

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```

```

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do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  ASDA
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      2      5  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

```

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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      3      1  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      3      2  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do

```

```
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

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## regression?

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: The
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do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
```

```

do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDA
I_5 BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 A
SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 3 3 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The

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## response has five or fewer unique values. Are you sure you want to
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## regression?

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## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      3      4  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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: The
## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

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: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDA
I_5 BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 A
SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 3 5 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?
```



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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDA
I_5 BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 A
SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 4 1 IMC TSINT VSG BASMI TIEM BIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

```

```
## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?
```

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## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDAI_5
BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 ASDAS_6
PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BASFI_0 BASFI_1
BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 4 2 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

```

```

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##   i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##   4   3  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do

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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      4      4  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to

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do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDAI_5
BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 ASDAS_6
PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BASFI_0 BASFI_1
BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 4 5 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The

```



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SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S
## 5 1 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## response has five or fewer unique values. Are you sure you want to
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: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?
```



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: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##   i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##   5   2  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

```

```

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## regression?

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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      5      3  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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```

```
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDA
I_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  A
SDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BAS
FI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
S
##      5      4  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
```

```
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

##      i_BASDAI  BASDAI_0  BASDAI_1  BASDAI_2  BASDAI_3  BASDAI_4  BASDAI_5  BASDAI_6  ASDAS_0  ASDAS_1  ASDAS_2  ASDAS_3  ASDAS_4  ASDAS_5  ASDAS_6  PCR_0  PCR_1  PCR_2  PCR_3  PCR_4  PCR_5  PCR_6  EGP  EGM  BASFI_0  BASFI_1  BASFI_3  BASFI_4  BASFI_5  BASFI_6  PCR_INICIOB  i_ASDA
```

```
S
## 5 5 IMC TSINT VSG BASMI TIEMBIO Interval

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
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## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
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## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?
```

```

do
## regression?

## Warning in randomForest.default(x = xobs, y = yobs, ntree = 1, ...)
: The
## response has five or fewer unique values. Are you sure you want to
do
## regression?

## i_BASDAI BASDAI_0 BASDAI_1 BASDAI_2 BASDAI_3 BASDAI_4 BASDA
I_5 BASDAI_6 ASDAS_0 ASDAS_1 ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5 A
SDAS_6 PCR_0 PCR_1 PCR_2 PCR_3 PCR_4 PCR_5 PCR_6 EGP EGM BAS
FI_0 BASFI_1 BASFI_3 BASFI_4 BASFI_5 BASFI_6 PCR_INICIOB i_ASDA
S

## Warning: Number of logged events: 950

c_base_rx.num<- complete(miceMod)
anyNA(c_base_rx.num)

## [1] FALSE

summary(c_base_rx.num)

##          EDAD          IMC          TSINT          VSG
## Min.   :21.00   Min.   :19.37   Min.    : 0.00   Min.    : 0.5
## 1st Qu.:38.00   1st Qu.:23.94   1st Qu.: 8.00   1st Qu.: 9.0
## Median :46.00   Median :26.02   Median :15.00   Median : 19.0
## Mean   :46.68   Mean    :26.38   Mean    :17.84   Mean    : 26.3
## 3rd Qu.:54.00   3rd Qu.:28.69   3rd Qu.:26.00   3rd Qu.: 38.0
## Max.   :75.00   Max.    :40.83   Max.    :52.00   Max.    :115.0
##          BASMI          TIEMBIO          Interval          i_BASDAI
## Min.   :0.5378   Min.    : 0.0   Min.    :2.000   Min.    :0.700
## 1st Qu.:1.5881   1st Qu.: 9.0   1st Qu.:3.000   1st Qu.:4.200
## Median :2.8480   Median : 42.0   Median :3.000   Median :5.400
## Mean   :3.1584   Mean    : 45.2   Mean    :3.446   Mean    :5.315
## 3rd Qu.:4.1775   3rd Qu.: 76.0   3rd Qu.:4.000   3rd Qu.:6.400
## Max.   :7.2729   Max.    :132.0   Max.    :6.000   Max.    :8.800
##          BASDAI_0          BASDAI_1          BASDAI_2          BASDAI_3
BASDAI_4
## Min.   :0.200   Min.   :0.00   Min.    : 0.00   Min.    :0.000   Min
.   :0.000
## 1st Qu.:2.000   1st Qu.:1.40   1st Qu.: 1.40   1st Qu.:1.000   1st
Qu.:1.200
## Median :3.200   Median :2.80   Median : 2.50   Median :2.800   Med
ian :2.800
## Mean   :3.611   Mean    :2.94   Mean    : 3.36   Mean    :3.011   Mea
n    :3.015
## 3rd Qu.:5.000   3rd Qu.:4.40   3rd Qu.: 4.10   3rd Qu.:4.700   3rd
Qu.:4.400
## Max.   :8.800   Max.    :8.50   Max.    :65.00   Max.    :8.000   Max
.   :8.000
##          BASDAI_5          BASDAI_6          ASDAS_0          ASDAS_1
## Min.   :0.00   Min.   :0.000   Min.    :0.2257   Min.    :0.2471
## 1st Qu.:1.40   1st Qu.:1.000   1st Qu.:1.4166   1st Qu.:1.1572

```

```

## Median :2.70 Median :2.400 Median :2.0966 Median :1.5609
## Mean :2.96 Mean :2.832 Mean :2.2336 Mean :1.7996
## 3rd Qu.:4.20 3rd Qu.:4.000 3rd Qu.:3.0000 3rd Qu.:2.4395
## Max. :8.00 Max. :7.300 Max. :5.0749 Max. :7.1255
## ASDAS_2 ASDAS_3 ASDAS_4 ASDAS_5
## Min. :0.001737 Min. :0.04892 Min. :0.00058 Min. :0.0
338
## 1st Qu.:1.118372 1st Qu.:1.13776 1st Qu.:1.29202 1st Qu.:1.1
620
## Median :1.837270 Median :1.76656 Median :1.68322 Median :1.6
142
## Mean :1.769730 Mean :1.81207 Mean :1.85161 Mean :1.7
300
## 3rd Qu.:2.307195 3rd Qu.:2.43783 3rd Qu.:2.45260 3rd Qu.:2.1
218
## Max. :4.131246 Max. :3.98374 Max. :4.45150 Max. :4.0
579
## ASDAS_6 PCR_0 PCR_1 PCR_2
## Min. :0.096 Min. : 0.000 Min. : 0.000 Min. : 0.000
## 1st Qu.:1.108 1st Qu.: 1.000 1st Qu.: 1.000 1st Qu.: 0.900
## Median :1.641 Median : 3.600 Median : 2.000 Median : 2.000
## Mean :1.753 Mean : 8.228 Mean : 4.324 Mean : 4.377
## 3rd Qu.:2.359 3rd Qu.: 9.200 3rd Qu.: 4.700 3rd Qu.: 5.600
## Max. :3.988 Max. :88.700 Max. :61.300 Max. :33.900
## PCR_3 PCR_4 PCR_5 PCR_6
## Min. : 0.00 Min. : 0.000 Min. : 0.00 Min. : 0.000
## 1st Qu.: 0.75 1st Qu.: 0.700 1st Qu.: 0.60 1st Qu.: 0.600
## Median : 2.00 Median : 2.000 Median : 1.30 Median : 1.800
## Mean : 4.52 Mean : 5.149 Mean : 3.95 Mean : 4.291
## 3rd Qu.: 5.90 3rd Qu.: 5.300 3rd Qu.: 6.00 3rd Qu.: 5.800
## Max. :37.00 Max. :68.500 Max. :65.90 Max. :39.700
## EGM EGM BASFI_0 BASFI_1
## Min. : 0.00 Min. : 0.000 Min. :0.000 Min. :0.000
## 1st Qu.: 2.00 1st Qu.: 1.000 1st Qu.:1.600 1st Qu.:1.200
## Median : 3.00 Median : 3.000 Median :3.700 Median :2.800
## Mean : 4.02 Mean : 3.416 Mean :3.912 Mean :3.086
## 3rd Qu.: 6.00 3rd Qu.: 5.000 3rd Qu.:5.900 3rd Qu.:4.600
## Max. :10.00 Max. :10.000 Max. :9.300 Max. :8.800
## BASFI_2 BASFI_3 BASFI_4 BASFI_5
BASFI_6
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.000 Mi
n. :0.00
## 1st Qu.:0.550 1st Qu.:1.100 1st Qu.:1.400 1st Qu.:0.900 1s
t Qu.:1.00
## Median :2.330 Median :2.700 Median :3.500 Median :3.000 Me
dian :2.80
## Mean :2.667 Mean :2.847 Mean :3.364 Mean :3.062 Me
an :3.24
## 3rd Qu.:4.340 3rd Qu.:4.400 3rd Qu.:5.000 3rd Qu.:4.800 3r
d Qu.:5.00
## Max. :7.890 Max. :7.900 Max. :8.400 Max. :8.100 Ma
x. :8.30
## mSASSS_Mean_BL mSASSS_Mean_FL mSASSS_Progress PCR_INICIOB
## Min. : 0.00 Min. : 0.00 Min. : -2.50 Min. : 0.00

```

```
## 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 4.20
## Median : 5.00 Median : 8.00 Median : 0.00 Median :11.10
## Mean :16.46 Mean :18.44 Mean : 1.98 Mean :18.64
## 3rd Qu.:23.00 3rd Qu.:30.00 3rd Qu.: 3.00 3rd Qu.:30.00
## Max. :72.00 Max. :72.00 Max. :21.50 Max. :98.00
## i_ASDas HVGP
## Min. :1.500 Min. : 1.000
## 1st Qu.:2.700 1st Qu.: 5.000
## Median :3.500 Median : 6.000
## Mean :3.416 Mean : 5.923
## 3rd Qu.:4.000 3rd Qu.: 7.000
## Max. :5.300 Max. :10.000
```

Ajustamos el data frame con los missings corregidos

```
c_base_rx.num1 <-
  c_base_rx.num %>%
  mutate( BASFIDif = (BASFI_6-BASFI_0), BASDAIDif = (BASDAI_6-BASDAI_0
), ASDASdif = (ASDAS_6-ASDAS_0), PCRDif = (PCR_6-PCR_0))
summary(base_rx.num1)
```

```
##      EDAD      IMC      TSINT      VSG
## Min.   :21.00  Min.   :19.37  Min.    : 0.00  Min.    : 0.50
## 1st Qu.:38.00  1st Qu.:24.05  1st Qu.: 8.00  1st Qu.: 9.00
## Median :46.00  Median :26.02  Median :15.00  Median : 19.00
## Mean   :46.68  Mean   :26.38  Mean    :17.76  Mean    : 26.65
## 3rd Qu.:54.00  3rd Qu.:28.68  3rd Qu.:26.00  3rd Qu.: 38.75
## Max.   :75.00  Max.   :40.83  Max.    :52.00  Max.    :115.00
##      NA's      NA's      NA's      NA's
##      :6        :4        :3
##      BASMI      TIEMBIO      Interval      i_BASDAI
## Min.   :0.5378  Min.    : 0.00  Min.    :2.00  Min.    :0.700
## 1st Qu.:1.6458  1st Qu.: 8.00  1st Qu.:3.00  1st Qu.:4.225
## Median :2.8728  Median :42.50  Median :3.00  Median :5.400
## Mean   :3.0915  Mean   :45.51  Mean    :3.45  Mean    :5.326
## 3rd Qu.:4.0864  3rd Qu.:76.75  3rd Qu.:4.00  3rd Qu.:6.400
## Max.   :7.2729  Max.   :132.00  Max.    :6.00  Max.    :8.800
## NA's   :19      NA's    :1      NA's    :1      NA's    :3
##      BASDAI_0      BASDAI_1      BASDAI_2      BASDAI_3
## Min.   :0.20  Min.   :0.000  Min.    : 0.000  Min.    :0.000
## 1st Qu.:2.00  1st Qu.:1.400  1st Qu.: 1.400  1st Qu.:1.200
## Median :3.20  Median :2.700  Median : 2.500  Median :3.000
## Mean   :3.67  Mean   :2.902  Mean    : 3.419  Mean    :3.108
## 3rd Qu.:5.10  3rd Qu.:4.400  3rd Qu.: 4.200  3rd Qu.:4.800
## Max.   :8.80  Max.   :8.500  Max.    :65.000  Max.    :8.000
## NA's   :4      NA's    :4      NA's    :4      NA's    :8
##      BASDAI_4      BASDAI_5      BASDAI_6      ASDAS_0
## Min.   :0.000  Min.   :0.000  Min.    :0.000  Min.    :0.2257
## 1st Qu.:1.200  1st Qu.:1.275  1st Qu.:1.100  1st Qu.:1.3874
## Median :2.800  Median :2.550  Median :2.800  Median :2.0565
## Mean   :3.075  Mean   :2.915  Mean    :2.924  Mean    :2.2160
## 3rd Qu.:4.600  3rd Qu.:4.200  3rd Qu.:4.600  3rd Qu.:2.9248
## Max.   :8.000  Max.   :8.000  Max.    :7.300  Max.    :5.0749
## NA's   :10     NA's    :17     NA's    :16     NA's    :5
##      ASDAS_1      ASDAS_2      ASDAS_3      ASDAS_4
```


##	Min.	:0.2471	Min.	:0.001737	Min.	:0.04892	Min.	:0.00
058								
##	1st Qu.:	1.1572	1st Qu.:	1.088078	1st Qu.:	1.16497	1st Qu.:	1.17
363								
##	Median	:1.5738	Median	:1.827288	Median	:1.76656	Median	:1.66
660								
##	Mean	:1.8249	Mean	:1.773470	Mean	:1.81523	Mean	:1.81
274								
##	3rd Qu.:	2.4204	3rd Qu.:	2.316346	3rd Qu.:	2.47797	3rd Qu.:	2.44
173								
##	Max.	:7.1255	Max.	:4.131246	Max.	:3.98374	Max.	:4.45
150								
##	NA's	:13	NA's	:9	NA's	:14	NA's	:22
##	ASDAS_5		ASDAS_6		PCR_0		PCR_1	
##	Min.	:0.0338	Min.	:0.096	Min.	:0.000	Min.	:0.000
##	1st Qu.:	1.0522	1st Qu.:	1.143	1st Qu.:	1.000	1st Qu.:	1.000
##	Median	:1.6400	Median	:1.680	Median	:3.350	Median	:2.000
##	Mean	:1.7329	Mean	:1.807	Mean	:8.181	Mean	:4.358
##	3rd Qu.:	2.1781	3rd Qu.:	2.397	3rd Qu.:	9.100	3rd Qu.:	4.675
##	Max.	:4.0579	Max.	:3.988	Max.	:88.700	Max.	:61.300
##	NA's	:30	NA's	:24	NA's	:7	NA's	:3
##	PCR_2		PCR_3		PCR_4		PCR_5	
##	Min.	:0.00	Min.	:0.000	Min.	:0.000	Min.	:0.000
##	1st Qu.:	0.90	1st Qu.:	0.750	1st Qu.:	0.700	1st Qu.:	0.615
##	Median	:2.00	Median	:1.900	Median	:2.050	Median	:1.400
##	Mean	:4.36	Mean	:4.452	Mean	:5.368	Mean	:4.118
##	3rd Qu.:	5.60	3rd Qu.:	5.600	3rd Qu.:	5.375	3rd Qu.:	6.000
##	Max.	:33.90	Max.	:37.000	Max.	:68.500	Max.	:65.900
##	NA's	:2	NA's	:4	NA's	:7	NA's	:10
##	PCR_6		EGP		EGM		BASFI_0	
##	Min.	:0.00	Min.	:0.000	Min.	:0.000	Min.	:0.000
##	1st Qu.:	0.60	1st Qu.:	2.000	1st Qu.:	1.500	1st Qu.:	1.525
##	Median	:2.00	Median	:3.000	Median	:3.000	Median	:3.850
##	Mean	:4.42	Mean	:4.031	Mean	:3.455	Mean	:3.931
##	3rd Qu.:	5.95	3rd Qu.:	6.000	3rd Qu.:	5.000	3rd Qu.:	5.975
##	Max.	:39.70	Max.	:10.000	Max.	:10.000	Max.	:9.300
##	NA's	:14	NA's	:3	NA's	:2	NA's	:3
##	BASFI_1		BASFI_2		BASFI_3		BASFI_4	
BASFI_5								
##	Min.	:0.000	Min.	:0.000	Min.	:0.000	Min.	:0.00
.	:0.000							Min
##	1st Qu.:	1.200	1st Qu.:	0.550	1st Qu.:	1.000	1st Qu.:	1.30
Qu.:	1.100							1st
##	Median	:2.700	Median	:2.330	Median	:2.800	Median	:3.50
ian	:3.200							Med
##	Mean	:3.059	Mean	:2.667	Mean	:2.882	Mean	:3.34
n	:3.118							Mea
##	3rd Qu.:	4.550	3rd Qu.:	4.340	3rd Qu.:	4.500	3rd Qu.:	5.00
Qu.:	4.700							3rd
##	Max.	:8.800	Max.	:7.890	Max.	:7.900	Max.	:8.40
.	:8.100							Max
##	NA's	:2			NA's	:8	NA's	:12
s	:22							NA'
##	BASFI_6		mSASSS_Mean_BL		mSASSS_Mean_FL		mSASSS_Progress	

```

## Min. :0.000 Min. : 0.00 Min. : 0.00 Min. : -2.50
## 1st Qu.:1.100 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.00
## Median :3.100 Median : 5.00 Median : 8.00 Median : 0.00
## Mean :3.313 Mean :16.46 Mean :18.44 Mean : 1.98
## 3rd Qu.:5.100 3rd Qu.:23.00 3rd Qu.:30.00 3rd Qu.: 3.00
## Max. :8.300 Max. :72.00 Max. :72.00 Max. :21.50
## NA's :16
## PCR_INICIOB i_ASIDAS HVGP BASFIDif
## Min. : 0.00 Min. :1.500 Min. : 1.000 Min. : -8.9000
## 1st Qu.: 4.60 1st Qu.:2.700 1st Qu.: 5.000 1st Qu.:-1.8000
## Median :11.10 Median :3.500 Median : 6.000 Median : -0.2000
## Mean :18.70 Mean :3.402 Mean : 5.923 Mean : -0.5667
## 3rd Qu.:28.05 3rd Qu.:4.000 3rd Qu.: 7.000 3rd Qu.: 0.6250
## Max. :98.00 Max. :5.300 Max. :10.000 Max. : 6.2000
## NA's :2 NA's :2 NA's :17
## BASDAIDif ASDASDif PCRDif
## Min. : -7.0000 Min. : -3.4650 Min. : -88.40
## 1st Qu.: -1.1000 1st Qu.: -1.0300 1st Qu.: -5.40
## Median : -0.1000 Median : -0.2247 Median : -0.20
## Mean : -0.5354 Mean : -0.3908 Mean : -4.14
## 3rd Qu.: 0.4750 3rd Qu.: 0.2567 3rd Qu.: 1.00
## Max. : 5.2000 Max. : 1.9947 Max. : 28.00
## NA's :19 NA's :28 NA's :20

```

```
c_base_rx.num2<-c_base_rx.num1[,c(-9:-29,-32:-40)]
```

```
c_base_rx2= cbind(c_base_rx.num2,base_rx [,c(3,5,7,8,9,12,13,14,15,16,
17,19,20,22,23,58,59,60,61,62,63)])
summary(c_base_rx2)
```

```

## EDAD IMC TSINT VSG
## Min. :21.00 Min. :19.37 Min. : 0.00 Min. : 0.5
## 1st Qu.:38.00 1st Qu.:23.94 1st Qu.: 8.00 1st Qu.: 9.0
## Median :46.00 Median :26.02 Median :15.00 Median : 19.0
## Mean :46.68 Mean :26.38 Mean :17.84 Mean : 26.3
## 3rd Qu.:54.00 3rd Qu.:28.69 3rd Qu.:26.00 3rd Qu.: 38.0
## Max. :75.00 Max. :40.83 Max. :52.00 Max. :115.0
## BASMI TIEMBIO Interval i_BASDAI
## Min. :0.5378 Min. : 0.0 Min. :2.000 Min. :0.700
## 1st Qu.:1.5881 1st Qu.: 9.0 1st Qu.:3.000 1st Qu.:4.200
## Median :2.8480 Median : 42.0 Median :3.000 Median :5.400
## Mean :3.1584 Mean : 45.2 Mean :3.446 Mean :5.315
## 3rd Qu.:4.1775 3rd Qu.: 76.0 3rd Qu.:4.000 3rd Qu.:6.400
## Max. :7.2729 Max. :132.0 Max. :6.000 Max. :8.800
## EGP EGM mSASSS_Progress PCR_INICIOB
## Min. : 0.00 Min. : 0.000 Min. : -2.50 Min. : 0.00
## 1st Qu.: 2.00 1st Qu.: 1.000 1st Qu.: 0.00 1st Qu.: 4.20
## Median : 3.00 Median : 3.000 Median : 0.00 Median :11.10
## Mean : 4.02 Mean : 3.416 Mean : 1.98 Mean :18.64
## 3rd Qu.: 6.00 3rd Qu.: 5.000 3rd Qu.: 3.00 3rd Qu.:30.00
## Max. :10.00 Max. :10.000 Max. :21.50 Max. :98.00
## i_ASIDAS HVGP BASFIDif BASDAIDif
## Min. :1.500 Min. : 1.000 Min. : -8.9000 Min. : -7.8000
## 1st Qu.:2.700 1st Qu.: 5.000 1st Qu.:-1.8000 1st Qu.:-1.7000

```

```

## Median :3.500   Median : 6.000   Median :-0.2000   Median :-0.1000
## Mean    :3.416   Mean    : 5.923   Mean    :-0.6723   Mean    :-0.7792
## 3rd Qu.:4.000   3rd Qu.: 7.000   3rd Qu.: 0.7000   3rd Qu.: 0.5000
## Max.    :5.300   Max.    :10.000   Max.    : 6.2000   Max.    : 5.2000
##      ASDASDif      PCRDif      SEX0      IMC>30      HLAB27
EA
## Min.    :-3.5373   Min.    :-88.400   MALE   :82   NO    :84   SI    :86
SI:86
## 1st Qu.: -1.0443   1st Qu.: -5.400   FEMALE:19   SI    :11   NO    :13
NO:15
## Median  :-0.2886   Median   : -0.200   NA's: 6   NA's: 2
## Mean    :-0.4804   Mean     : -3.937
## 3rd Qu.: 0.2043   3rd Qu.: 1.000
## Max.    : 1.9947   Max.     : 28.000
## TABACO_Kat UVEITIS PSORIASIS ENF_INF   TTOBIO      NUM_TB_PREV
## SI:70      NO:79   NO   :92   NO   :92   ETA:31   treatments.0:75
## NO:31      SI:22   SI   : 7   SI   : 7   ADA:44   treatments.1:20
##                      NA's: 2   NA's: 2   IFX:11   treatments.2: 6
##                      GOL:12
##                      CTZ: 3
##
##          MOTIVO      BIO_INICI Biol_4_kat   AINE      FAME      New_Syn
_mSASSS
## non.change :77   NO   :25   NO:55      NO   :42   NO   :81   NO:81
## Inefficacy :17   SI   :75   SI:46      SI   :57   SI   :18   SI:20
## Side.effects: 7   NA's: 1      NA's: 2   NA's: 2   NA's: 2
##
##
## Prog_Syn_mSASSS mSASSS_Progress_Kat New_Prog_Syn_Msasss Mind_2_Syn
d_FL
## NO:90      NO:71      NO:79      NO:46
## SI:11      SI:30      SI:22      SI:55
##
##
## Mind_2_Synd_BL
## NO:50
## SI:51
##
##
##

```

Analizamos la relación entre variables numéricas

```

round(cor(x = c_base_rx.num2, method = "pearson"), 3)

##          EDAD   IMC   TSINT   VSG   BASMI   TIEMBIO   Interval
i_BASDAI
## EDAD      1.000 0.129 0.602 0.159 0.479 0.118 -0.056
0.164
## IMC       0.129 1.000 0.142 0.032 0.175 0.153 0.061

```

0.018							
## TSINT	0.602	0.142	1.000	0.133	0.416	0.408	0.058
0.193							
## VSG	0.159	0.032	0.133	1.000	0.079	0.072	-0.066
-0.045							
## BASMI	0.479	0.175	0.416	0.079	1.000	0.195	0.124
0.132							
## TIEMBIO	0.118	0.153	0.408	0.072	0.195	1.000	-0.033
0.051							
## Interval	-0.056	0.061	0.058	-0.066	0.124	-0.033	1.000
-0.140							
## i_BASDAI	0.164	0.018	0.193	-0.045	0.132	0.051	-0.140
1.000							
## EGP	0.199	0.054	0.002	-0.137	0.169	-0.297	0.071
0.434							
## EGM	0.122	0.013	-0.068	0.084	0.289	-0.347	0.108
0.330							
## mSASSS_Progress	0.327	0.181	0.163	0.033	0.432	0.091	0.124
-0.140							
## PCR_INICIOB	0.150	0.058	0.064	0.689	0.052	0.010	0.038
-0.054							
## i_ASIDAS	0.172	0.140	0.142	0.379	0.097	0.047	-0.072
0.622							
## HVGP	0.033	0.061	-0.006	0.069	0.103	0.060	-0.022
0.553							
## BASFIDif	0.163	0.076	0.155	-0.006	-0.063	0.297	0.114
-0.372							
## BASDAIDif	-0.015	0.030	0.088	-0.029	0.069	0.344	0.174
-0.432							
## ASDASdif	-0.018	0.027	0.164	-0.051	0.012	0.380	0.102
-0.240							
## PCRdif	-0.169	0.030	0.151	-0.179	-0.012	0.283	-0.043
0.048							
##	EGP	EGM	mSASSS_Progress	PCR_INICIOB	i_ASIDAS		
HVGP							
## EDAD	0.199	0.122	0.327	0.150	0.172		
0.033							
## IMC	0.054	0.013	0.181	0.058	0.140		
0.061							
## TSINT	0.002	-0.068	0.163	0.064	0.142	-	
0.006							
## VSG	-0.137	0.084	0.033	0.689	0.379		
0.069							
## BASMI	0.169	0.289	0.432	0.052	0.097		
0.103							
## TIEMBIO	-0.297	-0.347	0.091	0.010	0.047		
0.060							
## Interval	0.071	0.108	0.124	0.038	-0.072	-	
0.022							
## i_BASDAI	0.434	0.330	-0.140	-0.054	0.622		
0.553							
## EGP	1.000	0.749	0.123	-0.085	0.200		
0.276							
## EGM	0.749	1.000	0.180	0.062	0.290		

```

0.372
## mSASSS_Progress  0.123  0.180                1.000    0.041 -0.075
0.023
## PCR_INICIOB      -0.085  0.062                0.041    1.000  0.569
0.038
## i_ASIDAS         0.200  0.290               -0.075    0.569  1.000
0.634
## HVGP             0.276  0.372                0.023    0.038  0.634
1.000
## BASFIDif        -0.478 -0.456                0.099   -0.037 -0.198 -
0.179
## BASDAIDif       -0.541 -0.386                0.209   -0.052 -0.237 -
0.099
## ASDASDif        -0.465 -0.396                0.217   -0.108 -0.162
0.046
## PCRDif          -0.114 -0.187                0.026   -0.361 -0.198
0.079
##
##                BASFIDif BASDAIDif ASDASDif PCRDif
## EDAD             0.163   -0.015  -0.018 -0.169
## IMC              0.076    0.030   0.027  0.030
## TSINT            0.155    0.088   0.164  0.151
## VSG              -0.006   -0.029  -0.051 -0.179
## BASMI            -0.063    0.069   0.012 -0.012
## TIEMBIO          0.297    0.344   0.380  0.283
## Interval         0.114    0.174   0.102 -0.043
## i_BASDAI        -0.372   -0.432  -0.240  0.048
## EGP              -0.478   -0.541  -0.465 -0.114
## EGM              -0.456   -0.386  -0.396 -0.187
## mSASSS_Progress  0.099    0.209   0.217  0.026
## PCR_INICIOB     -0.037   -0.052  -0.108 -0.361
## i_ASIDAS        -0.198   -0.237  -0.162 -0.198
## HVGP            -0.179   -0.099   0.046  0.079
## BASFIDif         1.000    0.747   0.590  0.164
## BASDAIDif        0.747    1.000   0.697  0.165
## ASDASDif         0.590    0.697   1.000  0.556
## PCRDif           0.164    0.165   0.556  1.000

```

```
library(corrplot)
```

```
## corrplot 0.84 loaded
```

```
m<-cor(c_base_rx.num2)
```

```
head(round(m,2))
```

```

##          EDAD  IMC  TSINT  VSG  BASMI  TIEMBIO  Interval  i_BASDAI  EGP
EGM
## EDAD      1.00 0.13  0.60  0.16  0.48    0.12    -0.06    0.16  0.20
0.12
## IMC       0.13 1.00  0.14  0.03  0.17    0.15     0.06     0.02  0.05
0.01
## TSINT     0.60 0.14  1.00  0.13  0.42    0.41     0.06     0.19  0.00
-0.07
## VSG       0.16 0.03  0.13  1.00  0.08    0.07    -0.07    -0.04 -0.14
0.08
## BASMI     0.48 0.17  0.42  0.08  1.00    0.20     0.12     0.13  0.17

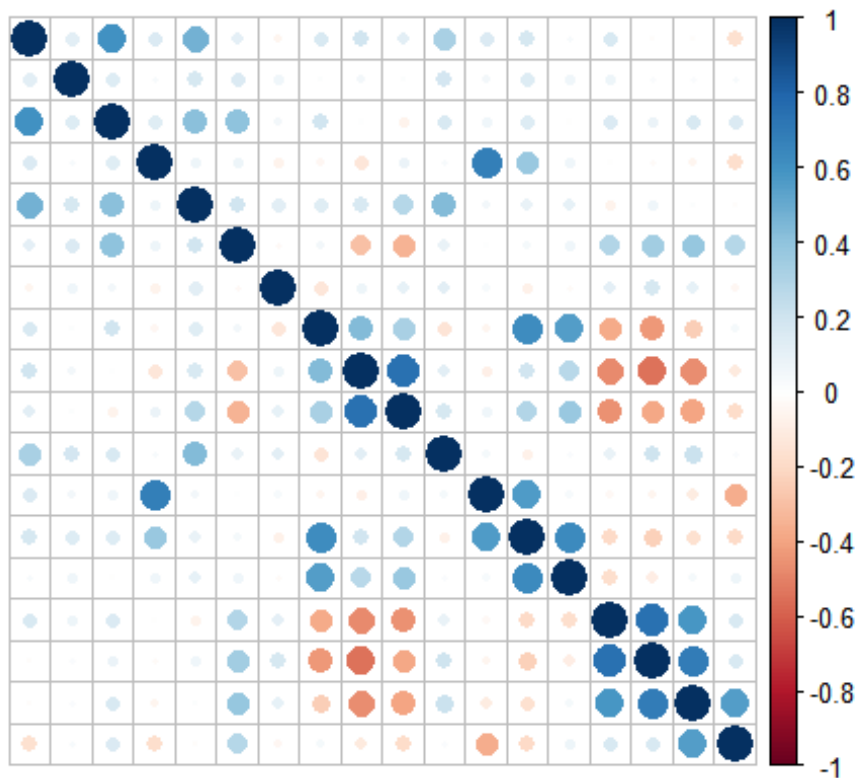
```

```

0.29
## TIEMBIO 0.12 0.15 0.41 0.07 0.20 1.00 -0.03 0.05 -0.30
-0.35
##          mSASSS_Progress PCR_INICIOB i_ASIDAS HVGP BASFIDif BASDAIDi
f ASDASDif
## EDAD          0.33      0.15      0.17      0.03      0.16      -0.0
2   -0.02
## IMC           0.18      0.06      0.14      0.06      0.08      0.0
3   0.03
## TSINT         0.16      0.06      0.14 -0.01      0.15      0.0
9   0.16
## VSG           0.03      0.69      0.38      0.07      -0.01     -0.0
3   -0.05
## BASMI         0.43      0.05      0.10      0.10      -0.06      0.0
7   0.01
## TIEMBIO       0.09      0.01      0.05      0.06      0.30      0.3
4   0.38
##          PCRdif
## EDAD        -0.17
## IMC          0.03
## TSINT        0.15
## VSG         -0.18
## BASMI       -0.01
## TIEMBIO      0.28

```

```
corrplot(m, method = "circle", tl.pos="n")
```



La correlación positiva está representada en azul, y la correlación negativa en rojo. La intensidad del del color y el tamaño del círculo son proporcionales al coeficiente de correlación.

Análisis de regresión lineal univariable y análisis de correlación de las variables

```
attach(c_base_rx.num)
b<-lm(mSASSS_Progress~EDAD,data=c_base_rx.num)
summary(b)

##
## Call:
## lm(formula = mSASSS_Progress ~ EDAD, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2771 -2.0128 -0.9391  0.2077 17.3258
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.82464    1.44047  -1.961 0.052698 .
## EDAD         0.10292    0.02991   3.442 0.000848 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.566 on 99 degrees of freedom
## Multiple R-squared:  0.1069, Adjusted R-squared:  0.09784
## F-statistic: 11.85 on 1 and 99 DF,  p-value: 0.0008483

cor.test(mSASSS_Progress,EDAD, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, EDAD, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and EDAD
## S = 122594, p-value = 0.00374
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.285997

c<-lm(mSASSS_Progress~IMC,data=c_base_rx.num)
summary(c)

##
## Call:
## lm(formula = mSASSS_Progress ~ IMC, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.6821 -1.8668 -1.2592  0.6264 17.7018
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.8955    2.6867  -1.078  0.2838
```

```

## IMC          0.1848    0.1009    1.832    0.0699 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.711 on 99 degrees of freedom
## Multiple R-squared:  0.03279,    Adjusted R-squared:  0.02302
## F-statistic: 3.357 on 1 and 99 DF,  p-value: 0.06994

cor.test(mSASSS_Progress,IMC, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, IMC, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and IMC
## S = 136602, p-value = 0.04032
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.204416

d<-lm(mSASSS_Progress~TSINT,data=c_base_rx.num)
summary(d)

##
## Call:
## lm(formula = mSASSS_Progress ~ TSINT, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5824 -1.8456 -1.2771  0.4649 18.2332
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.13498    0.63490   1.788  0.0769 .
## TSINT        0.04737    0.02890   1.639  0.1043
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.723 on 99 degrees of freedom
## Multiple R-squared:  0.02642,    Adjusted R-squared:  0.01659
## F-statistic: 2.687 on 1 and 99 DF,  p-value: 0.1043

cor.test(mSASSS_Progress,TSINT, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, TSINT, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and TSINT

```



```

## S = 146835, p-value = 0.1485
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.1448178

e<-lm(mSASSS_Progress~VSG,data=c_base_rx.num)
summary(e)

##
## Call:
## lm(formula = mSASSS_Progress ~ VSG, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.3474 -1.9840 -1.8583  0.8192 19.5378
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.836482   0.580375   3.164  0.00207 **
## VSG          0.005465   0.016836   0.325  0.74617
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.771 on 99 degrees of freedom
## Multiple R-squared:  0.001063, Adjusted R-squared: -0.009027
## F-statistic: 0.1054 on 1 and 99 DF, p-value: 0.7462

cor.test(mSASSS_Progress,VSG, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, VSG, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and VSG
## S = 169531, p-value = 0.9002
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.01263013

f<-lm(mSASSS_Progress~BASMI,data=c_base_rx.num)
summary(f)

##
## Call:
## lm(formula = mSASSS_Progress ~ BASMI, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.4813 -1.5879 -0.5962  0.4510 16.2866
##

```

```

## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.9500    0.7013  -1.355    0.179
## BASMI       0.9277    0.1945   4.771 6.34e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.402 on 99 degrees of freedom
## Multiple R-squared:  0.1869, Adjusted R-squared:  0.1787
## F-statistic: 22.76 on 1 and 99 DF,  p-value: 6.335e-06

cor.test(mSASSS_Progress,BASMI, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, BASMI, method = "spear
man"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and BASMI
## S = 95796, p-value = 3.688e-06
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.4420714

g<-lm(mSASSS_Progress~TIEMBIO,data=c_base_rx.num)
summary(g)

##
## Call:
## lm(formula = mSASSS_Progress ~ TIEMBIO, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.0954 -2.0296 -1.5954  0.7513 18.9596
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.595437    0.565656   2.821  0.00579 **
## TIEMBIO     0.008513    0.009391   0.906  0.36689
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.758 on 99 degrees of freedom
## Multiple R-squared:  0.008232, Adjusted R-squared:  -0.001786
## F-statistic: 0.8217 on 1 and 99 DF,  p-value: 0.3669

cor.test(mSASSS_Progress,TIEMBIO, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, TIEMBIO, method = "spe
arman"):
## Cannot compute exact p-value with ties

```

```

##
## Spearman's rank correlation rho
##
## data: mSASSS_Progress and TIEMBIO
## S = 164738, p-value = 0.6873
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.04054722

h<-lm(mSASSS_Progress~Interval,data=c_base_rx.num)
summary(h)

##
## Call:
## lm(formula = mSASSS_Progress ~ Interval, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7453 -1.7672 -1.2892  0.7547 19.2547
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.3331     1.3753   0.242   0.809
## Interval      0.4781     0.3842   1.244   0.216
##
## Residual standard error: 3.744 on 99 degrees of freedom
## Multiple R-squared:  0.0154, Adjusted R-squared:  0.00545
## F-statistic: 1.548 on 1 and 99 DF,  p-value: 0.2164

cor.test(mSASSS_Progress,Interval, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, Interval, method = "sp
earman"):
## Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data: mSASSS_Progress and Interval
## S = 153995, p-value = 0.3048
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.1031182

i<-lm(mSASSS_Progress~i_BASDAI,data=c_base_rx.num)
summary(i)

##
## Call:
## lm(formula = mSASSS_Progress ~ i_BASDAI, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max

```

```

## -3.8754 -1.9845 -1.4914 0.6163 18.4423
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  3.5218     1.1539   3.052 0.00292 **
## i_BASDAI    -0.2901     0.2055  -1.411 0.16132
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.736 on 99 degrees of freedom
## Multiple R-squared:  0.01972,    Adjusted R-squared:  0.009817
## F-statistic: 1.991 on 1 and 99 DF,  p-value: 0.1613

cor.test(mSASSS_Progress,i_BASDAI, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, i_BASDAI, method = "sp
earman"):
## Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and i_BASDAI
## S = 183897, p-value = 0.4802
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## -0.07103595

j<-lm(mSASSS_Progress~EGP,data=c_base_rx.num)
summary(j)

##
## Call:
## lm(formula = mSASSS_Progress ~ EGP, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1482 -1.8090 -1.4734  0.5196 19.3553
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.3055     0.6628   1.970 0.0517 .
## EGP          0.1678     0.1364   1.231 0.2214
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.745 on 99 degrees of freedom
## Multiple R-squared:  0.01507,    Adjusted R-squared:  0.005119
## F-statistic: 1.515 on 1 and 99 DF,  p-value: 0.2214

cor.test(mSASSS_Progress,EGP, method = "spearman")

```

```

## Warning in cor.test.default(mSASSS_Progress, EGP, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data: mSASSS_Progress and EGP
## S = 141747, p-value = 0.08102
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.1744509

l<-lm(mSASSS_Progress~EGM,data=c_base_rx.num)
summary(l)

##
## Call:
## lm(formula = mSASSS_Progress ~ EGM, data = c_base_rx.num)
##
## Residuals:
## Min 1Q Median 3Q Max
## -5.4898 -1.8631 -1.2997 0.2286 19.0736
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.0180 0.6451 1.578 0.1177
## EGM 0.2817 0.1549 1.819 0.0719 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.712 on 99 degrees of freedom
## Multiple R-squared: 0.03234, Adjusted R-squared: 0.02257
## F-statistic: 3.309 on 1 and 99 DF, p-value: 0.07193

cor.test(mSASSS_Progress,EGM, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, EGM, method = "spearman"): Cannot
## compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data: mSASSS_Progress and EGM
## S = 140236, p-value = 0.06661
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.1832498

r<-lm(mSASSS_Progress~PCR_INICIOB,data=c_base_rx.num)
summary(r)

```

```

##
## Call:
## lm(formula = mSASSS_Progress ~ PCR_INICIOB, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.3659 -1.9451 -1.8422  0.7716 19.4299
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.832707   0.522521   3.507 0.000682 ***
## PCR_INICIOB  0.007914   0.019516   0.405 0.685990
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.77 on 99 degrees of freedom
## Multiple R-squared:  0.001658, Adjusted R-squared:  -0.008426
## F-statistic: 0.1644 on 1 and 99 DF, p-value: 0.686

cor.test(mSASSS_Progress,PCR_INICIOB, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, PCR_INICIOB, method =
"spearmen"):
## Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  mSASSS_Progress and PCR_INICIOB
## S = 167716, p-value = 0.8178
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.02320423

s<-lm(mSASSS_Progress~i_ASDAS,data=c_base_rx.num)
summary(s)

##
## Call:
## lm(formula = mSASSS_Progress ~ i_ASDAS, data = c_base_rx.num)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4198 -2.0182 -1.6246  0.4898 19.2850
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.1006     1.5419   2.011  0.047 *
## i_ASDAS       -0.3280     0.4379  -0.749  0.456
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.763 on 99 degrees of freedom

```

```
## Multiple R-squared: 0.005636, Adjusted R-squared: -0.004408
## F-statistic: 0.5611 on 1 and 99 DF, p-value: 0.4556

cor.test(mSASSS_Progress,i_ASDAS, method = "spearman")

## Warning in cor.test.default(mSASSS_Progress, i_ASDAS, method = "spe
arman"):
## Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data: mSASSS_Progress and i_ASDAS
## S = 183572, p-value = 0.492
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## -0.06914314
```

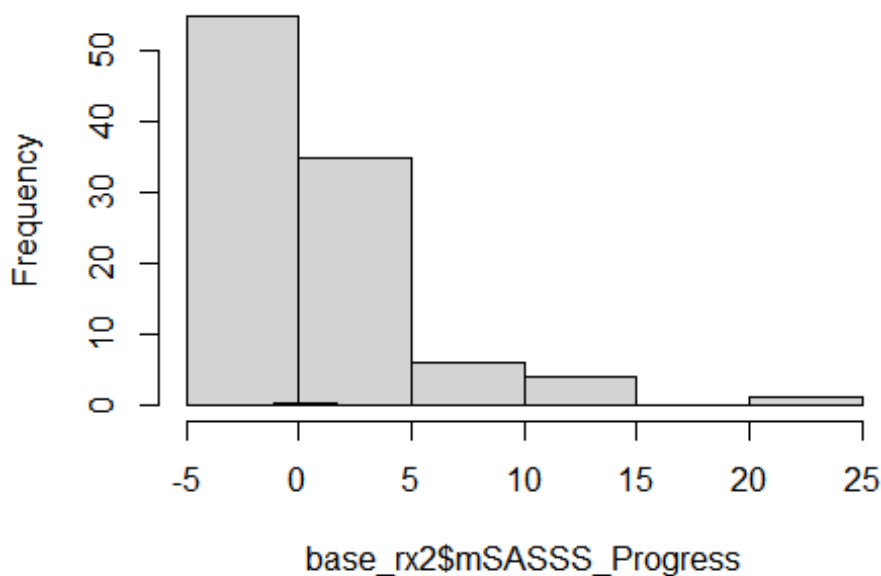
Análisis de la variable respuesta (mSASSS_Progress), numérica

```
attach(base_rx2)

## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, EGP, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

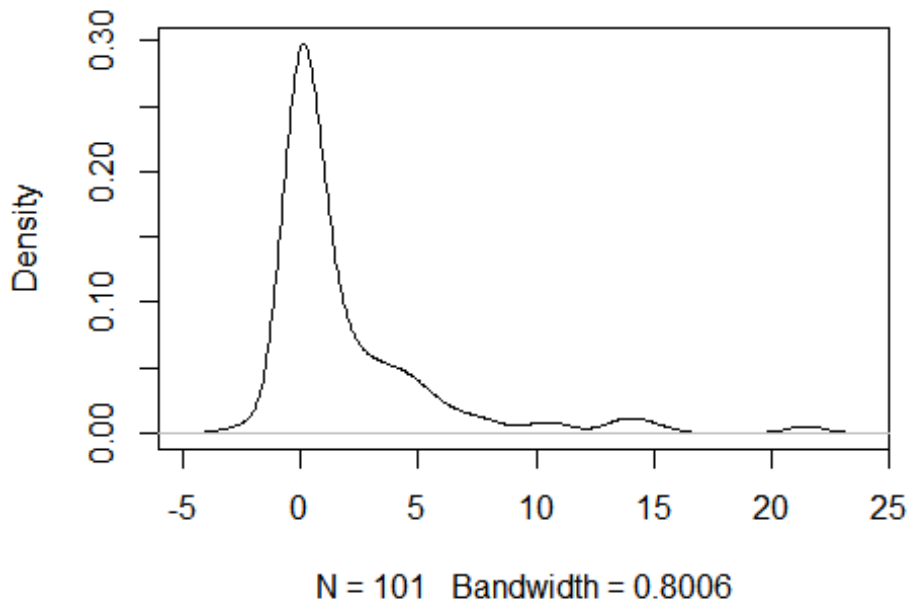
hist(base_rx2$mSASSS_Progress)
lines(density(mSASSS_Progress))
```

Histogram of base_rx2\$mSASSS_Progress



```
plot(density(mSASSS_Progress))
```

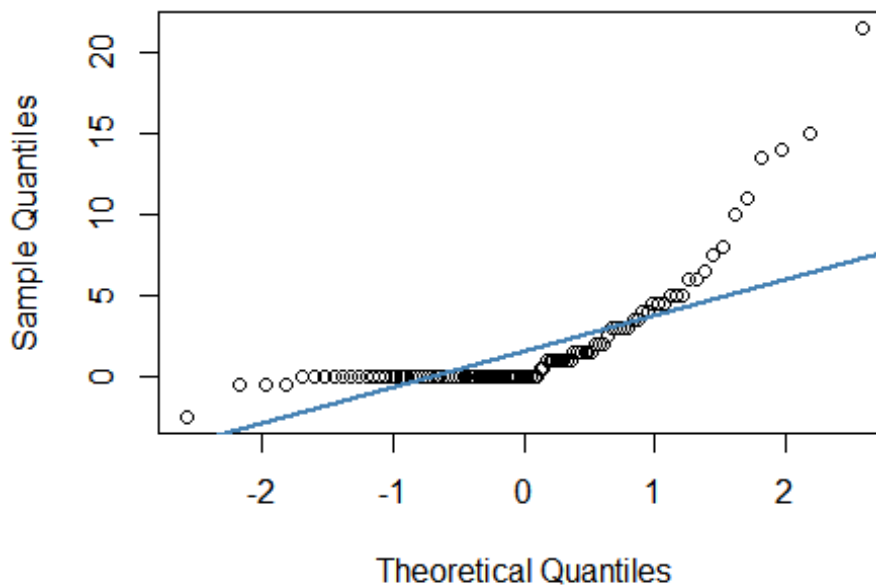
```
density.default(x = mSASSS_Progress)
```



Parece que la densidad es claramente asimétrica.

```
qqnorm(base_rx2$mSASSS_Progress)  
qqline(base_rx2$mSASSS_Progress, col = "steelblue", lwd = 2)
```

Normal Q-Q Plot



```
summary(base_rx2$mSASSS_Progress)
```



```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    -2.50   0.00   0.00   1.98   3.00   21.50
```

```
shapiro.test(base_rx2$mSASSS_Progress)
```

```
##
## Shapiro-Wilk normality test
##
## data:  base_rx2$mSASSS_Progress
## W = 0.63869, p-value = 2.092e-14
```

En el análisis descriptivos de la variable respuesta, vemos que la media y mediana no coincide, y que la kurtosis es superior a 3. Tanto en el Shapiro-Wilk's test como en el Kolmogorov-Smirnov (K-S) normality test vemos que la variable respuesta, no sigue una distribución normal. Pues al obtener un resultado significativo ($p < 0,05$), debemos rechazar la hipótesis nula de igualdad. Como la variable respuesta no es normal, podemos intentar una transformación para mejorar esa situación, pero antes vamos a ver si los residuos siguen la suposición de normalidad y homocedasticidad.

```
attach(base_rx2)
```

```
## The following objects are masked from base_rx2 (pos = 3):
##
##      AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
at,
##      EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BAS
DAI,
##      IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
##      mSASSS_Progress, mSASSS_Progress_Kat, New_Prog_Syn_Msasss,
##      New_Syn_mSASSS, NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSAS
SS,
##      PSORIASIS, SEXO, TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, V
SG

## The following objects are masked from c_base_rx.num:
##
##      BASMI, EDAD, EGM, EGP, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
##      mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG
```

```
a<-lm(mSASSS_Progress~ ., data=base_rx2, na.action = na.omit)
```

```
summary(a)
```

```
##
## Call:
## lm(formula = mSASSS_Progress ~ ., data = base_rx2, na.action = na.o
mit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0835 -0.8788 -0.1220  0.8867  2.7176
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6.030490  13.395121   0.450  0.66131
## EDAD           0.076896   0.095442   0.806  0.43749
```

```

## SEXOFEMALE          -1.681685    2.615599   -0.643    0.53343
## IMC                  -0.033612    0.253970   -0.132    0.89710
## `IMC>30`SI          -0.037253    3.009490   -0.012    0.99035
## TSINT                0.081354    0.100625    0.808    0.43595
## HLAB27NO            -5.040693    2.797895   -1.802    0.09905
## EANO                 -0.063666    5.218556   -0.012    0.99048
## TABACO_KatNO        0.725666    2.349126    0.309    0.76316
## VSG                  0.065548    0.060919    1.076    0.30497
## BASMI               -0.845203    0.716373   -1.180    0.26295
## UVEITISSI          -0.026928    1.973453   -0.014    0.98936
## PSORIASISSI         3.106432    3.351841    0.927    0.37393
## ENF_INFESI          4.710392    4.354606    1.082    0.30253
## TTOBIOADA           -3.037611    2.053408   -1.479    0.16712
## TTOBIOIFX           0.410255    2.700656    0.152    0.88201
## TTOBIOGOL           -3.172484    4.399943   -0.721    0.48593
## TTOBIOCTZ           -5.425373    4.128656   -1.314    0.21556
## NUM_TB_PREVtreatments.1 -4.314352    6.012165   -0.718    0.48796
## NUM_TB_PREVtreatments.2 -3.024290    7.419758   -0.408    0.69139
## MOTIVOinefficacy     5.409123    5.805011    0.932    0.37144
## MOTIVOSide.effects    1.273170    5.501541    0.231    0.82124
## TIEMBIO             -0.014120    0.044600   -0.317    0.75747
## BIO_INICISI         -2.875136    2.600103   -1.106    0.29242
## Biol_4_katSI        1.508230    2.541654    0.593    0.56491
## Interval            0.012851    0.994240    0.013    0.98992
## AINESI              0.330936    1.847061    0.179    0.86106
## FAMESI             -1.886349    2.837520   -0.665    0.51988
## i_BASDAI           -0.339930    0.839083   -0.405    0.69316
## EGP                 0.374793    0.690786    0.543    0.59825
## EGM                 0.641881    0.622347    1.031    0.32450
## New_Syn_mSASSSSI     7.897362    4.922208    1.604    0.13692
## Prog_Syn_mSASSSSI    0.804746    3.195528    0.252    0.80581
## mSASSS_Progress_KatSI 7.508523    2.288661    3.281    0.00732 **
## New_Prog_Syn_MsasssSI -6.756748    5.282319   -1.279    0.22718
## Mind_2_Synd_FLSI    -3.497963    3.594456   -0.973    0.35139
## Mind_2_Synd_BLSI    2.459721    3.650972    0.674    0.51440
## PCR_INICIOB         0.007389    0.134302    0.055    0.95711
## i_ASIDAS           -1.629755    3.200426   -0.509    0.62065
## HVGP               -0.050671    0.757349   -0.067    0.94786
## BASFIDif            0.072305    0.810919    0.089    0.93055
## BASDAIDif          -0.120637    0.674872   -0.179    0.86138
## ASDASDif            1.934157    1.712446    1.129    0.28273
## PCRDif             -0.022665    0.134179   -0.169    0.86893
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.772 on 11 degrees of freedom
## (46 observations deleted due to missingness)
## Multiple R-squared:  0.9235, Adjusted R-squared:  0.6244
## F-statistic: 3.087 on 43 and 11 DF, p-value: 0.02385

modelo<-update(a, .~. -BASDAIDif -EGP)
summary(modelo)

```

```

##
## Call:
## lm(formula = mSASSS_Progress ~ EDAD + SEXO + IMC + `IMC>30` +
##     TSINT + HLAB27 + EA + TABACO_Kat + VSG + BASMI + UVEITIS +
##     PSORIASIS + ENF_INF + TTOBIO + NUM_TB_PREV + MOTIVO + TIEMBIO +
##     BIO_INICI + Biol_4_kat + Interval + AINE + FAME + i_BASDAI +
##     EGM + New_Syn_mSASSS + Prog_Syn_mSASSS + mSASSS_Progress_Kat +
##     New_Prog_Syn_Msasss + Mind_2_Synd_FL + Mind_2_Synd_BL + PCR_INI
CIOB +
##     i_AS DAS + HVGP + BASFIDif + ASDASDif + PCRDif, data = base_rx2,
##     na.action = na.omit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2495 -0.9135 -0.1335  0.8903  2.8754
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      3.279166   11.544312    0.284  0.78085
## EDAD              0.105811    0.080815    1.309  0.21310
## SEXOFEMALE      -1.366577    2.418556   -0.565  0.58167
## IMC              -0.030907    0.236522   -0.131  0.89803
## `IMC>30`SI       0.159458    2.652687    0.060  0.95298
## TSINT            0.086091    0.094628    0.910  0.37949
## HLAB27NO        -4.835809    2.439601   -1.982  0.06900 .
## EANO             0.898020    4.438945    0.202  0.84281
## TABACO_KatNO    1.242937    2.061128    0.603  0.55686
## VSG              0.049249    0.047083    1.046  0.31462
## BASMI           -0.978005    0.652077   -1.500  0.15755
## UVEITISSI       0.234858    1.760230    0.133  0.89590
## PSORIASISSI     2.652616    3.103625    0.855  0.40820
## ENF_INFESI      4.797201    4.092863    1.172  0.26219
## TTOBIOADA       -3.291948    1.720120   -1.914  0.07792 .
## TTOBIOIFX       0.519086    2.529191    0.205  0.84057
## TTOBIOGOL       -2.952448    4.111204   -0.718  0.48537
## TTOBIOCTZ       -5.536354    3.859320   -1.435  0.17503
## NUM_TB_PREVtreatments.1 -4.310685    5.525908   -0.780  0.44931
## NUM_TB_PREVtreatments.2 -3.793466    6.723062   -0.564  0.58219
## MOTIVOInefficacy  5.415986    5.425952    0.998  0.33642
## MOTIVOSide.effects  1.497152    5.029598    0.298  0.77066
## TIEMBIO         -0.022553    0.040463   -0.557  0.58675
## BIO_INICISI     -2.984814    2.384523   -1.252  0.23272
## Biol_4_katSI    2.139469    2.161978    0.990  0.34044
## Interval        0.221040    0.851297    0.260  0.79920
## AINESI          0.849680    1.548010    0.549  0.59239
## FAMESI         -2.517002    2.358806   -1.067  0.30536
## i_BASDAI        -0.084631    0.700072   -0.121  0.90563
## EGM             0.891908    0.398355    2.239  0.04328 *
## New_Syn_mSASSSSI  7.005377    4.388975    1.596  0.13447
## Prog_Syn_mSASSSSI  0.708655    2.765453    0.256  0.80177
## mSASSS_Progress_KatSI  7.709396    2.140348    3.602  0.00322 **
## New_Prog_Syn_MsasssSI -5.686655    4.730863   -1.202  0.25079
## Mind_2_Synd_FL SI -2.835653    3.276390   -0.865  0.40246
## Mind_2_Synd_BLSI  1.767413    3.320724    0.532  0.60354

```

```

## PCR_INICIOB          0.023364    0.122191    0.191    0.85131
## i_ASIDAS            -1.838684    2.949624   -0.623    0.54383
## HVGP                0.066306    0.613889    0.108    0.91564
## BASFIDif           0.221046    0.483612    0.457    0.65516
## ASDASDif           1.371442    1.444511    0.949    0.35973
## PCRDif              0.008344    0.119027    0.070    0.94518
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.612 on 13 degrees of freedom
## (46 observations deleted due to missingness)
## Multiple R-squared:  0.9197, Adjusted R-squared:  0.6664
## F-statistic: 3.631 on 41 and 13 DF,  p-value: 0.007572

fitted(modelo)

##           1           3           4           5
6
##  3.883109e+00 -7.540416e-01  7.511298e-01  2.116825e+00  7.109927e-
01
##           7           9           10          11
12
## -1.505862e+00  1.664959e+00 -1.310363e+00  5.649057e-01 -1.996754e+
00
##           13          14          15          16
17
## -1.638241e+00  6.398036e+00  6.168910e-01  1.334632e-01  2.582696e+
00
##           19          20          21          34
35
## -3.959460e-01  9.341081e-01 -2.172219e-01 -1.068590e-15  6.121443e-
01
##           36          42          44          45
46
## -3.704619e-01 -2.504826e-01  4.398667e+00  7.091628e-01 -1.929421e+
00
##           47          49          50          51
56
##  6.061890e-01 -8.847289e-01 -2.353672e-03  1.306331e+00  5.930905e+
00
##           57          59          61          62
63
## -6.646363e-01  1.396165e+00 -1.119954e+00  1.019641e-01  1.168910e-
01
##           64          65          66          67
68
##  1.496427e+01 -9.531105e-01  1.862463e+01  4.826957e+00 -1.115725e-
01
##           69          70          71          72
73
##  1.174049e+01  1.953745e+00  1.711572e-01  2.198770e+00  8.896140e+
00
##           75          76          85          86
87

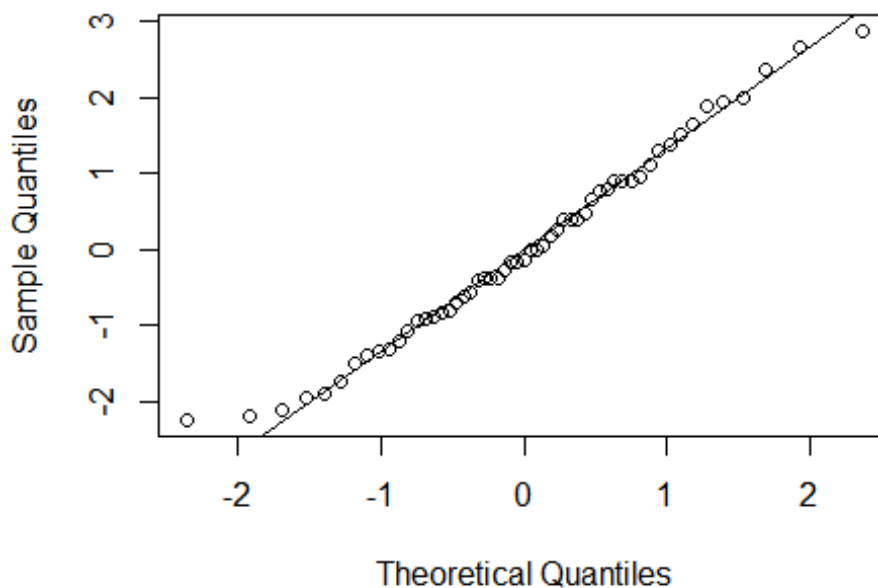
```

```
## 2.320867e+00 1.163738e+01 -4.786596e-01 -9.051875e-01 6.002216e+
00
##          88          89          94          96
97
## 3.421602e+00 8.114250e-01 1.197917e+00 1.085277e+01 1.333130e+
00
```

46 observaciones eliminadas debido a los missing, quedando un total de 97 observaciones.

```
qqnorm(modelo$residuals)
qqline(modelo$residuals)
```

Normal Q-Q Plot

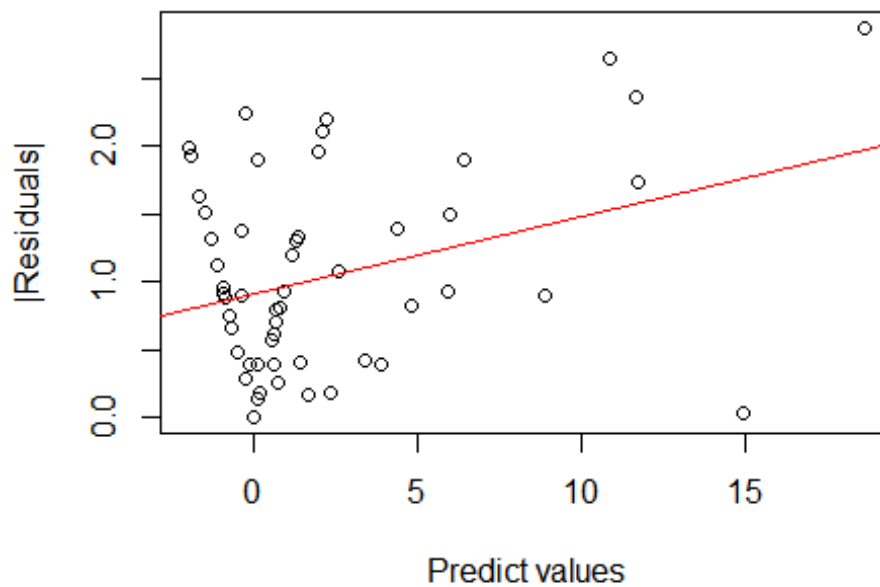


```
shapiro.test(modelo$residuals)
```

```
##
## Shapiro-Wilk normality test
##
## data:  modelo$residuals
## W = 0.98209, p-value = 0.5818
```

Se rechaza la hipótesis nula, por lo que hay normalidad de los residuos

```
plot(fitted(modelo),abs(residuals(modelo)),xlab="Predict values", ylab="
|Residuals|")
abline(lm(abs(residuals(modelo)) ~ fitted(modelo)), col="red", lwd=1.5
)
```



```
library(lmtest)

## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

bptest(modelo)

##
## studentized Breusch-Pagan test
##
## data: modelo
## BP = 42.753, df = 41, p-value = 0.3958

# No hay evidencias de falta de homocedasticidad.
```

Validación de condiciones para la regresión múltiple lineal, como ya hemos analizado los residuos y la homocedasticidad. Valoramos ahora la correlación.

```
library(car)

## Loading required package: carData

##
## Attaching package: 'car'
```

```
## The following object is masked from 'package:psych':
##
##   logit

## The following object is masked from 'package:dplyr':
##
##   recode

dwt(modelo, alternative = "two.sided")

##   lag Autocorrelation D-W Statistic p-value
##   1      0.005012348      1.968281  0.824
## Alternative hypothesis: rho != 0
```

No hay evidencia de autocorrelación

Identificación de posibles valores atípicos o influyentes

```
library(car)
outlierTest(modelo)

## No Studentized residuals with Bonferroni p < 0.05
## Largest |rstudent|:
##   rstudent unadjusted p-value Bonferroni p
## 66  3.51269      0.0042807      0.23116
```

No hay datos atípicos

Conclusión

El modelo múltiple es capaz de explicar el 62.4% de la variabilidad observada en la progresión radiográfica (R2: 0.9235, R2-Adjusted: 0.6244). El test F muestra que es significativo (0.02385). Se satisfacen todas las condiciones para este tipo de regresión múltiple.

```
attach(base_rx3)

## The following objects are masked from base_rx2 (pos = 7):
##
##   AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
##   at,
##   EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BAS
##   DAI,
##   IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
##   mSASSS_Progress, NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, S
##   EXO,
##   TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 8):
##
##   AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
##   at,
##   EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BAS
##   DAI,
##   IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
##   mSASSS_Progress, NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, S
```

```

EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx.num:
##
##      BASMI, EDAD, EGM, EGP, HVGP, i_ASIDAS, i_BASDAI, IMC, Interval,
##      mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

a3<-lm(mSASSS_Progress~ ., na.action= na.omit, data=base_rx3)
summary(a3)

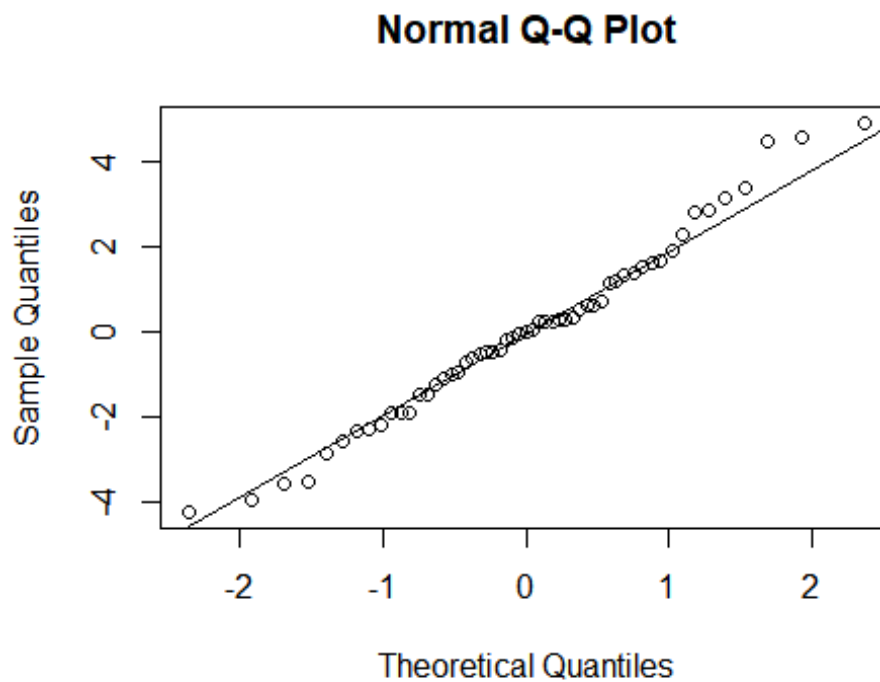
##
## Call:
## lm(formula = mSASSS_Progress ~ ., data = base_rx3, na.action = na.o
mit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.236 -1.350  0.000  1.249  4.897
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      8.103692   16.992383    0.477    0.640
## EDAD             -0.003886    0.125016   -0.031    0.976
## SEXOFEMALE      -2.243199    3.656405   -0.613    0.549
## IMC              -0.108577    0.358271   -0.303    0.766
## `IMC>30`SI      -1.097715    4.171210   -0.263    0.796
## TSINT            0.045964    0.142208    0.323    0.751
## HLAB27NO        -1.896123    3.860162   -0.491    0.630
## EANO             -2.841619    7.070704   -0.402    0.693
## TABACO_KatNO    -0.216501    2.868730   -0.075    0.941
## VSG              -0.019563    0.065021   -0.301    0.768
## BASMI            0.416655    0.851084    0.490    0.632
## UVEITISSI       -1.159512    2.654264   -0.437    0.668
## PSORIASISSI      3.137099    4.752488    0.660    0.519
## ENF_INFESI       3.609225    6.030499    0.598    0.558
## TTOBIOADA        -4.247040    2.686738   -1.581    0.135
## TTOBIOIFX        -0.546824    3.833681   -0.143    0.888
## TTOBIOGOL        -1.636413    5.776941   -0.283    0.781
## TTOBIOCTZ        -6.491962    5.877194   -1.105    0.287
## NUM_TB_PREVtreatments.1 -8.992542    8.412075   -1.069    0.302
## NUM_TB_PREVtreatments.2 -11.086607    9.511212   -1.166    0.262
## MOTIVOInefficacy  9.432894    8.162722    1.156    0.266
## MOTIVOSide.effects  5.104385    7.522292    0.679    0.508
## TIEMBIO          0.044958    0.050599    0.889    0.388
## BIO_INICISI      -2.551808    3.305274   -0.772    0.452
## Biol_4_katSI     -1.221309    3.298691   -0.370    0.716
## Interval         0.401322    1.348680    0.298    0.770
## AINESI           -2.294849    2.447854   -0.937    0.363
## FAMESI           2.819417    3.758669    0.750    0.465
## i_BASDAI         -1.057945    1.173140   -0.902    0.381
## EGP              0.719047    0.834054    0.862    0.402
## EGM              0.135346    0.806780    0.168    0.869
## Mind_2_Synd_FLSI -1.583067    5.027610   -0.315    0.757
## Mind_2_Synd_BLSI  3.469850    5.045243    0.688    0.502

```



```
## PCR_INICIOB          0.137779    0.174083    0.791    0.441
## i_ASIDAS            -0.997510    4.468997   -0.223    0.826
## HVGP                0.334195    1.033858    0.323    0.751
## BASFIDif            0.109019    0.940209    0.116    0.909
## BASDAIDif          -0.110290    0.857289   -0.129    0.899
## ASDASDif           0.977816    2.290934    0.427    0.676
## PCRDif              0.022501    0.171042    0.132    0.897
##
## Residual standard error: 3.984 on 15 degrees of freedom
## (46 observations deleted due to missingness)
## Multiple R-squared:  0.7844, Adjusted R-squared:  0.2239
## F-statistic: 1.4 on 39 and 15 DF, p-value: 0.245
```

```
#Análisis de residuos
qqnorm(a3$residuals)
qqline(a3$residuals)
```



```
shapiro.test(a3$residuals)
```

```
##
## Shapiro-Wilk normality test
##
## data:  a3$residuals
## W = 0.98334, p-value = 0.6412
```

```
# Los residuos siguen la normalidad
```

```
library(lmtest)
bptest(a3)
```

```
##
## studentized Breusch-Pagan test
```

```
##
## data: a3
## BP = 45.736, df = 39, p-value = 0.2126

# No hay evidencias de falta de homocedasticidad.

library(car)
dwt(a3,alternative = "two.sided")

## lag Autocorrelation D-W Statistic p-value
## 1 -0.1210956 2.221821 0.636
## Alternative hypothesis: rho != 0

# No hay evidencia de autocorrelación
```

Conclusión

El modelo múltiple es capaz de explicar el 22.4% de la variabilidad observada en la progresión radiográfica (R2: 0.7844, R2-Adjusted: 0.2239). El test F muestra que NO es significativo (0.245).

```
attach(base_rx.num2)

## The following objects are masked from base_rx3:
##
## ASDASdif, BASDAIDif, BASFIDif, BASMI, EDAD, EGM, EGP, HVGP,
## i_ASIDAS, i_BASDAI, IMC, Interval, mSASSS_Progress, PCR_INICIOB,
## PCRDif, TIEMBIO, TSINT, VSG

## The following objects are masked from base_rx2 (pos = 8):
##
## ASDASdif, BASDAIDif, BASFIDif, BASMI, EDAD, EGM, EGP, HVGP,
## i_ASIDAS, i_BASDAI, IMC, Interval, mSASSS_Progress, PCR_INICIOB,
## PCRDif, TIEMBIO, TSINT, VSG

## The following objects are masked from base_rx2 (pos = 9):
##
## ASDASdif, BASDAIDif, BASFIDif, BASMI, EDAD, EGM, EGP, HVGP,
## i_ASIDAS, i_BASDAI, IMC, Interval, mSASSS_Progress, PCR_INICIOB,
## PCRDif, TIEMBIO, TSINT, VSG

## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, EGP, HVGP, i_ASIDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

a2<-lm(mSASSS_Progress~ ., data=base_rx.num2)
summary(a2)

##
## Call:
## lm(formula = mSASSS_Progress ~ ., data = base_rx.num2)
##
## Residuals:
## Min 1Q Median 3Q Max
## -4.6650 -1.4206 -0.5989 0.5910 10.9876
##
```

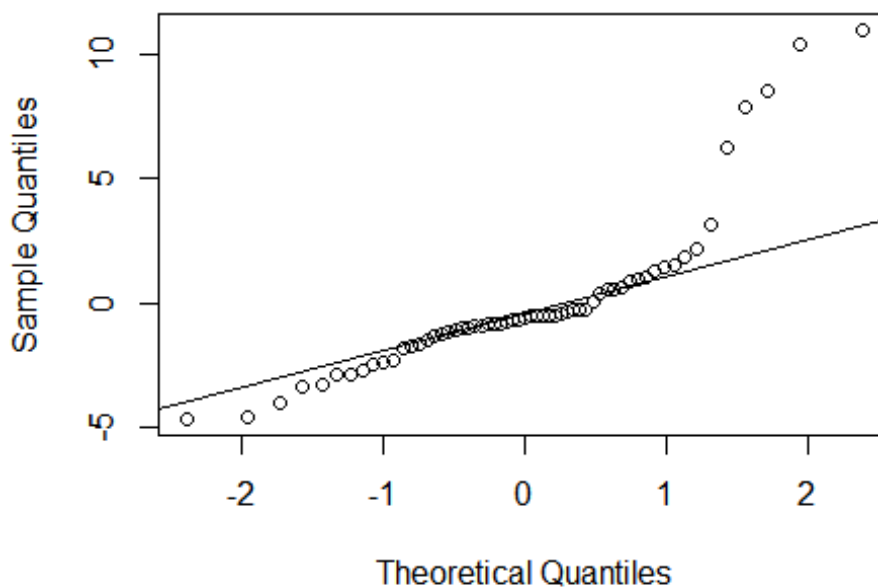
```

## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.30755    6.27543  -1.005   0.321
## EDAD         0.07762    0.08274   0.938   0.354
## IMC          0.02270    0.14503   0.157   0.876
## TSINT       -0.05750    0.06973  -0.825   0.414
## VSG         -0.03316    0.03730  -0.889   0.379
## BASMI        0.80028    0.54226   1.476   0.148
## TIEMBIO      0.01685    0.01915   0.880   0.384
## Interval     0.50775    0.66034   0.769   0.446
## i_BASDAI    -0.87809    0.58810  -1.493   0.143
## EGP          0.45201    0.39417   1.147   0.258
## EGM          0.26407    0.52132   0.507   0.615
## PCR_INICIOB  0.03217    0.08768   0.367   0.716
## i_ASIDAS     0.76942    2.24528   0.343   0.734
## HVGP         0.02271    0.51083   0.044   0.965
## BASFIDif    -0.03336    0.43809  -0.076   0.940
## BASDAIDif   -0.19473    0.52683  -0.370   0.714
## ASDASdif     1.66904    1.07943   1.546   0.130
## PCRDif      -0.03104    0.06995  -0.444   0.660
##
## Residual standard error: 3.763 on 41 degrees of freedom
## Multiple R-squared:  0.4834, Adjusted R-squared:  0.2692
## F-statistic: 2.257 on 17 and 41 DF,  p-value: 0.01685

#Análisis de residuos
qqnorm(a2$residuals)
qqline(a2$residuals)

```

Normal Q-Q Plot



```
shapiro.test(a2$residuals)
```

```

##
## Shapiro-Wilk normality test
##
## data: a2$residuals
## W = 0.7963, p-value = 1.311e-07

# Los residuos no siguen la normalidad

library(lmtest)
bptest(a2)

##
## studentized Breusch-Pagan test
##
## data: a2
## BP = 8.9985, df = 17, p-value = 0.9403

# No hay evidencias de falta de homocedasticidad.

library(car)
dwt(a2,alternative = "two.sided")

## lag Autocorrelation D-W Statistic p-value
## 1 -0.2147223 2.416146 0.114
## Alternative hypothesis: rho != 0

# No hay evidencia de autocorrelación

```

Conclusión, este modelo múltiple es capaz de explicar el 32.6% de la variabilidad observada en la progresión radiográfica (R2: 0.5235, R2-Adjusted: 0.326). El test F muestra que es significativo (0.005458). Sin embargo, el análisis de los residuos muestra que no siguen la normalidad.

```

attach(c_base_rx2)

## The following objects are masked from base_rx.num2:
##
## ASDASdif, BASDAIDif, BASFIDif, BASMI, EDAD, EGM, EGP, HVGP,
## i_ASIDAS, i_BASDAI, IMC, Interval, mSASSS_Progress, PCR_INICIOB,
## PCRDif, TIEMBIO, TSINT, VSG

## The following objects are masked from base_rx3:
##
## AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
at,
## EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASIDAS, i_BAS
DAI,
## IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
## mSASSS_Progress, NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 9):
##
## AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
at,

```

```

## EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BAS
DAI,
## IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
## mSASSS_Progress, mSASSS_Progress_Kat, New_Prog_Syn_Msasss,
## New_Syn_mSASSS, NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSAS
SS,
## PSORIASIS, SEXO, TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, V
SG

## The following objects are masked from base_rx2 (pos = 10):
##
## AINE, ASDASdif, BASDAIDif, BASFIDif, BASMI, BIO_INICI, Biol_4_k
at,
## EA, EDAD, EGM, EGP, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BAS
DAI,
## IMC, IMC>30, Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO,
## mSASSS_Progress, mSASSS_Progress_Kat, New_Prog_Syn_Msasss,
## New_Syn_mSASSS, NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSAS
SS,
## PSORIASIS, SEXO, TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, V
SG

## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, EGP, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

a4<-lm(mSASSS_Progress~ ., data=c_base_rx2)
summary(a4)

##
## Call:
## lm(formula = mSASSS_Progress ~ ., data = c_base_rx2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.1847 -0.9599 -0.1242  0.9775  7.9780
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.432440    3.693996  -0.117  0.9073
## EDAD           0.081921    0.042793   1.914  0.0619 .
## IMC          -0.003432    0.132654  -0.026  0.9795
## TSINT        -0.003237    0.041008  -0.079  0.9374
## VSG           0.029065    0.020344   1.429  0.1600
## BASMI        -0.116478    0.249814  -0.466  0.6433
## TIEMBIO       0.016939    0.019102   0.887  0.3799
## Interval     0.043426    0.375553   0.116  0.9085
## i_BASDAI     -0.408141    0.367417  -1.111  0.2725
## EGP           0.298590    0.249372   1.197  0.2374
## EGM           0.085729    0.236838   0.362  0.7191
## PCR_INICIOB  -0.044345    0.033021  -1.343  0.1860
## i_ASDAS      0.691693    1.015520   0.681  0.4993
## HVGP        -0.053475    0.285021  -0.188  0.8520
## BASFIDif    -0.234562    0.201987  -1.161  0.2517

```

```

## BASDAIDif          0.138215    0.285341    0.484    0.6305
## ASDASDif          0.965751    0.503800    1.917    0.0616 .
## PCRDif           -0.006256    0.037187   -0.168    0.8672
## SEXOFEMALE       -1.672282    0.983481   -1.700    0.0960 .
## `IMC>30`SI        1.205225    1.497020    0.805    0.4250
## HLAB27NO         -1.697130    1.013916   -1.674    0.1011
## EANO              1.203685    1.339031    0.899    0.3735
## TABACO_KatNO     -0.110267    0.726659   -0.152    0.8801
## UVEITISSI         0.214601    0.819317    0.262    0.7946
## PSORIASISSI      -1.262357    1.393738   -0.906    0.3699
## ENF_INFESI       -0.091538    1.583879   -0.058    0.9542
## TTOBIOADA        -1.863147    0.801557   -2.324    0.0247 *
## TTOBIOIFX         0.054218    1.235601    0.044    0.9652
## TTOBIOGOL        -0.526924    1.619044   -0.325    0.7463
## TTOBIOCTZ        -2.201403    2.048448   -1.075    0.2883
## NUM_TB_PREVtreatments.1  1.808198    2.026637    0.892    0.3770
## NUM_TB_PREVtreatments.2 -0.634034    2.708435   -0.234    0.8160
## MOTIVOInefficacy -0.880702    2.236344   -0.394    0.6956
## MOTIVOSide.effects -2.001092    2.290223   -0.874    0.3869
## BIO_INICISI      -1.925112    1.174174   -1.640    0.1081
## Biol_4_katSI     -1.121438    1.266724   -0.885    0.3807
## AINESI           -1.442735    0.735905   -1.960    0.0561 .
## FAMESI           1.283016    0.981828    1.307    0.1979
## New_Syn_mSASSSSI  6.049405    2.353675    2.570    0.0135 *
## Prog_Syn_mSASSSSI 0.902780    1.478001    0.611    0.5444
## mSASSS_Progress_KatSI 5.487751    1.221006    4.494 4.85e-05 ***
## New_Prog_Syn_MsasssSI -5.513072    2.802110   -1.967    0.0553 .
## Mind_2_Synd_FLSI -2.652211    2.074171   -1.279    0.2076
## Mind_2_Synd_BLSI  2.064728    2.088418    0.989    0.3281
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.293 on 45 degrees of freedom
## (12 observations deleted due to missingness)
## Multiple R-squared:  0.8261, Adjusted R-squared:  0.66
## F-statistic: 4.972 on 43 and 45 DF, p-value: 1.989e-07

```

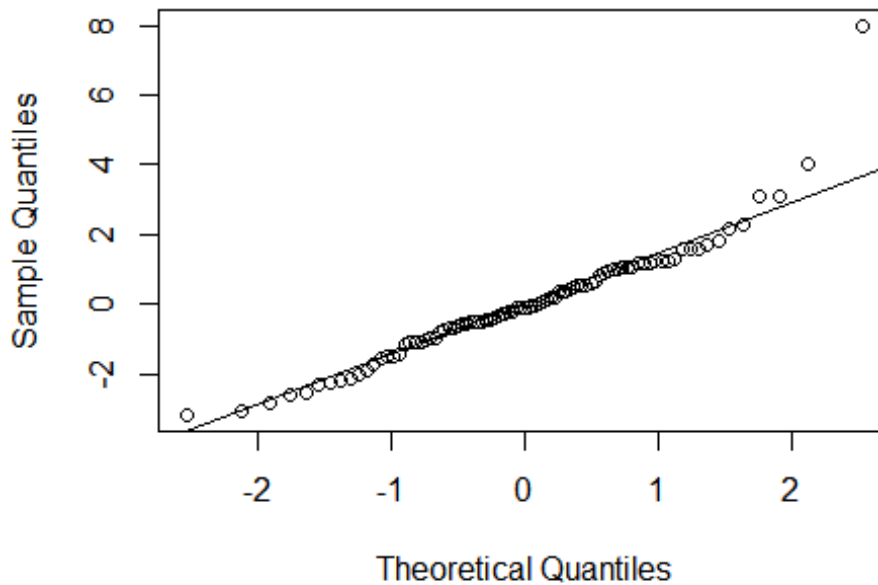
#Análisis de residuis

```

qqnorm(a4$residuals)
qqline(a4$residuals)

```

Normal Q-Q Plot



```
shapiro.test(a4$residuals)

##
## Shapiro-Wilk normality test
##
## data:  a4$residuals
## W = 0.92064, p-value = 4.247e-05

# pero Los residuos no siguen La normalidad

library(lmtest)
bptest(a4)

##
## studentized Breusch-Pagan test
##
## data:  a4
## BP = 55.722, df = 43, p-value = 0.09237

# No hay evidencias de falta de homocedasticidad.

library(car)
dwt(a4,alternative = "two.sided")

## lag Autocorrelation D-W Statistic p-value
## 1 -0.06955701 2.137721 0.672
## Alternative hypothesis: rho != 0

# No hay evidencia de autocorrelación
```

Conclusión, este modelo múltiple es capaz de explicar el 66.6% de la variabilidad observada en la progresión radiográfica (R2: 0.8296, R2-Adjusted: 0.6668).El test F

muestra que es significativo (1.365e-07). Sin embargo, el análisis de los residuos muestra que no siguen la normalidad.

Por lo que elegimos el primer modelo.

Por regla general, el número de observaciones debe ser como mínimo entre 10 y 20 veces el número de predictores del modelo. Como tenemos 97 observaciones, el número de predictores no debería ser mayor a 10. Por lo que debemos eliminar predictores, que no influyan en el modelo. Para ello, vamos a realizar una regresión con variables seleccionadas por AIC

```
library(leaps)
attach(base_rx21)

## The following objects are masked from c_base_rx2:
##
## AINE, ASDASdif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx.num2:
##
## ASDASdif, BASFIDif, BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI,
IMC,
## Interval, mSASSS_Progress, PCR_INICIOB, PCRDif, TIEMBIO, TSINT,
VSG

## The following objects are masked from base_rx3:
##
## AINE, ASDASdif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, SEXO, TABACO_Kat,
## TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 11):
##
## AINE, ASDASdif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
```



```

EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 12):
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

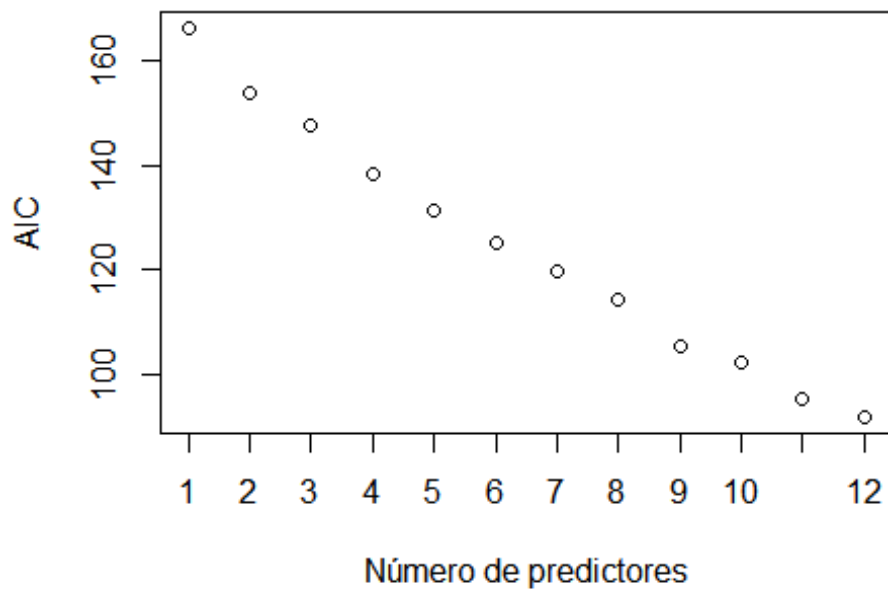
## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

b <- regsubsets( mSASSS_Progress~ ., data=base_rx21, na.action = na.om
it, nvmax=12)
rs <- summary(b)
n <- nrow(base_rx21)
k <- length(rs$rss) # Número de variables predictoras
p <- k + 1 # Número de parámetros (incluye la intercepción)
(AIC <- n*log(rs$rss/n) + (2:p)*2)

## [1] 166.12733 153.54661 147.48753 138.12852 131.52924 125.24759 11
9.94576
## [8] 114.31560 105.63619 102.32931 95.34395 92.00777

plot(1:k, AIC, ylab="AIC", xlab="Número de predictores", axes=F)
box(); axis(1,at=1:k); axis(2)

```



El mínimo AIC se alcanza con 12 predictores, aunque con 10 también tenemos un buen resultado.

```
rs$outmat[10,]
##          EDAD          SEXOFEMALE
IMC
##          "*"          " "
" "
##          `IMC>30`SI          TSINT          HLAB
27NO
##          " "          " "
" "
##          EANO          TABACO_KatNO
VSG
##          "*"          " "
" "
##          BASMI          UVEITISSI          PSORIAS
ISSI
##          " "          " "
" "
##          ENF_INFISI          TTOBIOADA          TTOBI
OIFX
##          " "          "*"
" "
##          TTOBIOGOL          TTOBIOCTZ NUM_TB_PREVtreatmen
ts.1
##          "*"          "*"
" "
## NUM_TB_PREVtreatments.2          MOTIVOInefficacy          MOTIVOSide.effects
##          "*"          " "
```

```

" "
##          TIEMBIO          BIO_INICISI          Biol_4_k
atSI
##          " "          "*"
" "
##          Interval          AINESI          FA
MESI
##          " "          " "
" "
##          i_BASDAI          EGM          New_Syn_mSAS
SSSI
##          " "          " "
"*"
##          Prog_Syn_mSASSSI  mSASSS_Progress_KatSI  New_Prog_Syn_Msas
ssSI
##          " "          "*"
" "
##          Mind_2_Synd_FLSI  Mind_2_Synd_BLSI          PCR_INI
CIOB
##          " "          " "
" "
##          i_ASIDAS          HVGP          BASF
IDif
##          " "          " "
" "
##          ASDASDif          PCRDif
##          "*"          " "

```

attach(base_rx21)

```

## The following objects are masked from base_rx21 (pos = 3):
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASIDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

```

```

## The following objects are masked from c_base_rx2:
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASIDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

```

```

## The following objects are masked from base_rx.num2:
##
## ASDASDif, BASFIDif, BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI,
IMC,
## Interval, mSASSS_Progress, PCR_INICIOB, PCRDif, TIEMBIO, TSINT,
VSG

## The following objects are masked from base_rx3:
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, SEXO, TABACO_Kat,
## TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 12):
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 13):
##
## AINE, ASDASDif, BASFIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDA
D,
## EGM, ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>3
0,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

s.aic<-lm(mSASSS_Progress ~ EDAD + SEXO + HLAB27 + EA + TTOBIO + EGM
+ New_Syn_mSASSS +
  mSASSS_Progress_Kat + New_Prog_Syn_Msasss + ASDASDif, na.action =
na.omit)
summary(s.aic)

```

```
##
## Call:
## lm(formula = mSASSS_Progress ~ EDAD + SEXO + HLAB27 + EA + TTOBIO +
##      EGM + New_Syn_mSASSS + mSASSS_Progress_Kat + New_Prog_Syn_Msass
s +
##      ASDASDif, na.action = na.omit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.5558 -1.0077 -0.3158  0.6955  9.8685
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.43128    1.51491  -0.945  0.348751
## EDAD           0.04491    0.03157   1.423  0.160307
## SEXOFEMALE    -0.19701    0.90764  -0.217  0.828939
## HLAB27NO      -0.60798    1.01252  -0.600  0.550577
## EANO           1.19672    1.03091   1.161  0.250547
## TTOBIOADA     -1.17795    0.77237  -1.525  0.132761
## TTOBIOIFX      0.15510    1.07691   0.144  0.885989
## TTOBIOGOL     -1.57378    1.38653  -1.135  0.261105
## TTOBIOCTZ     -1.89162    2.04345  -0.926  0.358505
## EGM            0.24254    0.16317   1.486  0.142691
## New_Syn_mSASSSI  5.29174    2.04740   2.585  0.012334 *
## mSASSS_Progress_KatSI 4.82019    1.27623   3.777  0.000381 ***
## New_Prog_Syn_MsassSI -3.05319    2.29198  -1.332  0.188123
## ASDASDif       0.89338    0.34092   2.621  0.011234 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.626 on 57 degrees of freedom
## (30 observations deleted due to missingness)
## Multiple R-squared:  0.6891, Adjusted R-squared:  0.6182
## F-statistic:  9.72 on 13 and 57 DF,  p-value: 3.046e-10
```

```
coef(s.aic)
```

```
##              (Intercept)              EDAD              SEXOFEMALE
##              -1.43127676              0.04491406              -0.19701044
##              HLAB27NO              EANO              TTOBIOADA
##              -0.60798117              1.19671506              -1.17794570
##              TTOBIOIFX              TTOBIOGOL              TTOBIOCTZ
##              0.15510074              -1.57377609              -1.89161660
##              EGM              New_Syn_mSASSSI mSASSS_Progress_KatSI
##              0.24253902              5.29173534              4.82019415
## New_Prog_Syn_MsassSI              ASDASDif
##              -3.05319384              0.89338432
```

Este valor indica que por cada diferencia entre el ASDAS inicial con el final, la odds estimada de progresión se multiplica por ese número. El intervalo de confianza es

```
(summary(s.aic)$coefficients["New_Syn_mSASSSI",1] +
qnorm(c(0.025,0.975)) * summary(s.aic)$coefficients["New_Syn_mSASSSI",2])
```

```

## [1] 1.278914 9.304557

(summary(s.aic)$coefficients["mSASSS_Progress_KatSI",1] +
qnorm(c(0.025,0.975)) * summary(s.aic)$coefficients["mSASSS_Progress_K
atSI",2])

## [1] 2.318822 7.321566

(summary(s.aic)$coefficients["ASDASDif",1] +
qnorm(c(0.025,0.975)) * summary(s.aic)$coefficients["ASDASDif",2])

## [1] 0.2251996 1.5615691

attach(base_rx22)

## The following objects are masked from base_rx21 (pos = 3):
##
## AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM
E,
## HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,
## Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx21 (pos = 4):
##
## AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM
E,
## HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,
## Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx2:
##
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx.num2:
##
## BASDAIDif, BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC,
## Interval, mSASSS_Progress, PCR_INICIOB, PCRDif, TIEMBIO, TSINT,
VSG

## The following objects are masked from base_rx3:
##

```

```

##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, SEXO, TABACO_Kat,
##      TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 13):
##
##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 14):
##
##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx.num:
##
##      BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
##      mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

abc<-lm(mSASSS_Progress~ ., data=base_rx22, na.action = na.omit)
summary(abc)

##
## Call:
## lm(formula = mSASSS_Progress ~ ., data = base_rx22, na.action = na.
omit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9083 -1.3655 -0.1359  0.8913  6.2356
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -5.378305   9.852182  -0.546  0.59184
## EDAD          0.157102   0.083216   1.888  0.07527 .
## SEXOFEMALE   -1.000685   2.306875  -0.434  0.66960
## IMC           0.089652   0.224743   0.399  0.69465
## `IMC>30`SI    0.542955   2.667651   0.204  0.84100
## TSINT         0.001852   0.091952   0.020  0.98415
## HLAB27NO     -4.500475   2.458216  -1.831  0.08374 .

```

```

## EANO          1.760100    3.519349    0.500    0.62305
## TABACO_KatNO  1.215674    1.951695    0.623    0.54117
## VSG           0.068615    0.039662    1.730    0.10074
## BASMI        -0.771863    0.561888   -1.374    0.18640
## UVEITISSI    0.719321    1.616122    0.445    0.66156
## PSORIASISSI -1.742260    2.573706   -0.677    0.50705
## ENF_INFESI   2.161013    4.266984    0.506    0.61869
## TTOBIOADA    -2.491257    1.437645   -1.733    0.10022
## TTOBIOIFX    0.262128    2.283534    0.115    0.90988
## TTOBIOGOL   -0.673301    4.048884   -0.166    0.86978
## TTOBIOCTZ   -3.458348    4.090163   -0.846    0.40892
## NUM_TB_PREVtreatments.1  3.014191    3.692527    0.816    0.42500
## NUM_TB_PREVtreatments.2  3.854029    5.161996    0.747    0.46494
## MOTIVOInefficacy -1.589475    3.762311   -0.422    0.67768
## MOTIVOSide.effects -4.524821    4.255796   -1.063    0.30174
## TIEMBIO      0.009117    0.039738    0.229    0.82113
## BIO_INICISI  -3.761606    2.478690   -1.518    0.14649
## Biol_4_katSI -0.028936    2.101266   -0.014    0.98916
## Interval     0.408044    0.752932    0.542    0.59451
## AINESI       0.402600    1.563484    0.258    0.79971
## FAMESI      -1.297081    2.480959   -0.523    0.60748
## i_BASDAI    -0.239530    0.737635   -0.325    0.74913
## EGM         0.452444    0.341935    1.323    0.20234
## New_Syn_mSASSSSI  7.996090    4.394555    1.820    0.08550 .
## Prog_Syn_mSASSSSI -1.202146    2.894825   -0.415    0.68285
## mSASSS_Progress_KatSI  7.911277    2.231201    3.546    0.00231 **
## New_Prog_Syn_MsasssSI -6.335959    4.855007   -1.305    0.20832
## Mind_2_Synd_FLISI -2.170821    3.431615   -0.633    0.53495
## Mind_2_Synd_BLSI  0.904186    3.500986    0.258    0.79913
## PCR_INICIOB  -0.071352    0.065928   -1.082    0.29342
## i_ASIDAS    -0.016251    2.571380   -0.006    0.99503
## HVGP        0.055180    0.592719    0.093    0.92686
## BASDAIDif   0.537715    0.462403    1.163    0.26007
## PCRDif      0.059146    0.070315    0.841    0.41129
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.975 on 18 degrees of freedom
## (42 observations deleted due to missingness)
## Multiple R-squared:  0.8578, Adjusted R-squared:  0.5418
## F-statistic: 2.715 on 40 and 18 DF, p-value: 0.0127

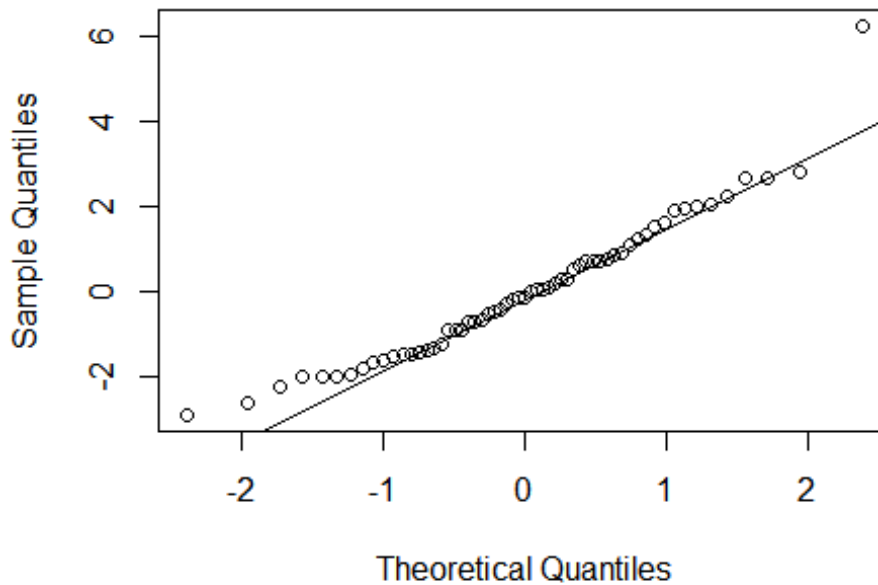
```

```

qqnorm(abc$residuals)
qqline(abc$residuals)

```


Normal Q-Q Plot

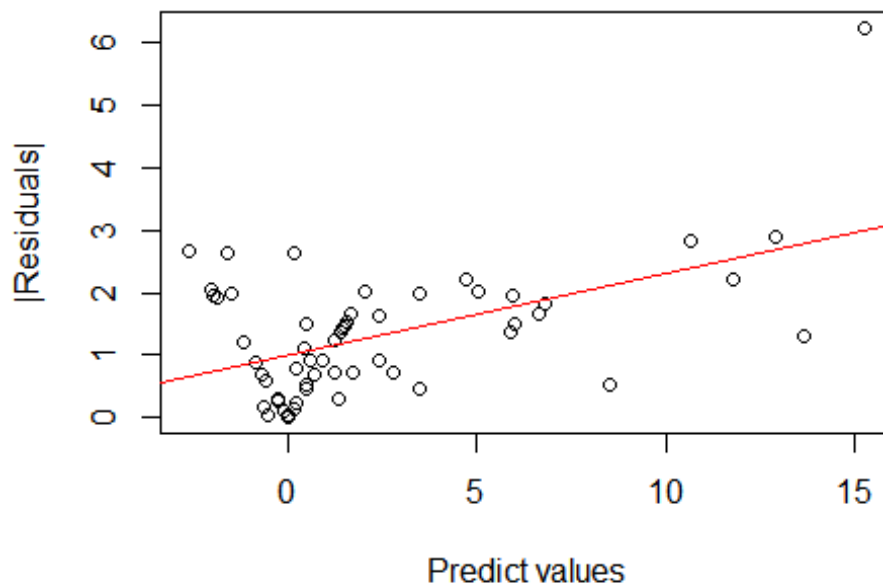


```
shapiro.test(abc$residuals)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: abc$residuals  
## W = 0.94982, p-value = 0.01651
```

No se rechaza la hipótesis nula, por lo que no hay normalidad de los residuos

```
plot(fitted(abc), abs(residuals(abc)), xlab="Predict values", ylab="|Residuals|")  
abline(lm(abs(residuals(abc)) ~ fitted(abc)), col="red", lwd=1.5)
```



```

library(lmtest)
bptest(abc)

##
## studentized Breusch-Pagan test
##
## data: abc
## BP = 41.823, df = 40, p-value = 0.3916

library(car)
dwt(abc,alternative = "two.sided")

## lag Autocorrelation D-W Statistic p-value
## 1 -0.04275237 2.056578 0.97
## Alternative hypothesis: rho != 0

Y = c(3.5,1.5,0,1,0,1.5,0,0.,1.5,0,0,0,0,4.5,1,0,1.5,0,0.5,0,-0.5,0,0
,0,0,0,0,3,0,6,0,0,3,0,0,1,0,0,0,4,0,0,0,1.5,0,0,-2.5,5.0,3.0,0,0,0,0,
1,0,5.0,0,0,0,0,0,0,4.5,0,0,0,0,6.5,5.0,0,0,3,0,1,0,0,0,0,0,2,0,0.5,15
.0,0,0,21.5,4,0,-0.5,10,0,0,0,0,0,0,8,0,2,0,2.5,14,0,3.5,1,0,0,0,0,0
,0,0,0,0,-0.5,1.5,0,0,0,0,4.5,3,0,0,0,1,0,0,0,11,0,7.5,0,0,0,13.5,0,0,
1,0,0,0,6,0,2,0)
LY = log10(Y + 1 - min(Y))
abc1<-lm(LY~.-mSASSS_Progress, data=base_rx22, na.action = na.omit)
summary(abc1)

##
## Call:
## lm(formula = LY ~ . - mSASSS_Progress, data = base_rx22, na.action
= na.omit)
##
## Residuals:

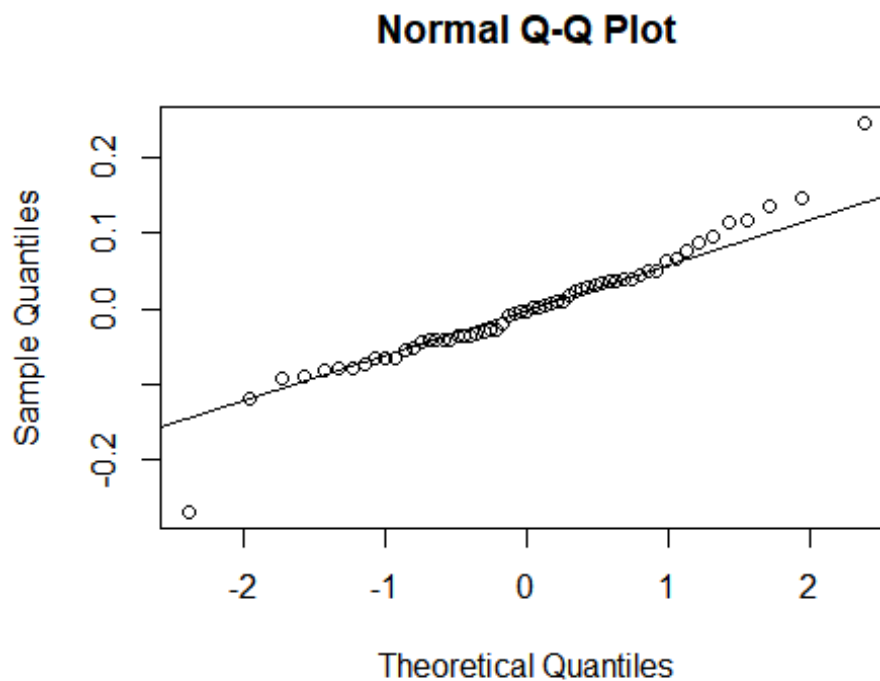
```

```

##          Min          1Q      Median          3Q          Max
## -0.268261 -0.041747 -0.002465  0.039164  0.246168
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    2.313e-01  4.533e-01   0.510  0.61612
## EDAD           8.508e-03  3.829e-03   2.222  0.03936 *
## SEXOFEMALE    -1.145e-02  1.061e-01  -0.108  0.91526
## IMC            9.242e-05  1.034e-02   0.009  0.99297
## `IMC>30`SI    2.183e-02  1.228e-01   0.178  0.86083
## TSINT         -3.076e-03  4.231e-03  -0.727  0.47663
## HLAB27NO      -6.550e-02  1.131e-01  -0.579  0.56973
## EANO          1.070e-01  1.619e-01   0.661  0.51699
## TABACO_KatNO -1.558e-02  8.981e-02  -0.174  0.86417
## VSG           1.979e-03  1.825e-03   1.085  0.29244
## BASMI        -1.222e-02  2.585e-02  -0.473  0.64207
## UVEITISSI    -2.197e-02  7.436e-02  -0.295  0.77102
## PSORIASISSI  -3.071e-02  1.184e-01  -0.259  0.79834
## ENF_INFESI    5.423e-03  1.963e-01   0.028  0.97827
## TTOBIOADA    -1.309e-01  6.615e-02  -1.979  0.06332 .
## TTOBIOIFX     2.622e-02  1.051e-01   0.250  0.80578
## TTOBIOGOL     3.872e-02  1.863e-01   0.208  0.83771
## TTOBIOCTZ    -1.417e-01  1.882e-01  -0.753  0.46123
## NUM_TB_PREVtreatments.1 -2.300e-02  1.699e-01  -0.135  0.89382
## NUM_TB_PREVtreatments.2 -1.280e-01  2.375e-01  -0.539  0.59647
## MOTIVOInefficacy 9.166e-02  1.731e-01   0.529  0.60296
## MOTIVOSide.effects -5.195e-02  1.958e-01  -0.265  0.79382
## TIEMBIO       1.142e-03  1.829e-03   0.625  0.54009
## BIO_INICISI   -1.296e-01  1.141e-01  -1.137  0.27064
## Biol_4_katSI  -2.922e-02  9.669e-02  -0.302  0.76598
## Interval      1.659e-02  3.465e-02   0.479  0.63780
## AINESI        -1.846e-02  7.194e-02  -0.257  0.80038
## FAMESI        -9.641e-03  1.142e-01  -0.084  0.93363
## i_BASDAI      1.658e-03  3.394e-02   0.049  0.96159
## EGM           1.295e-03  1.573e-02   0.082  0.93531
## New_Syn_mSASSSSI 2.344e-01  2.022e-01   1.159  0.26152
## Prog_Syn_mSASSSSI 1.257e-03  1.332e-01   0.009  0.99258
## mSASSS_Progress_KatSI 3.524e-01  1.027e-01   3.433  0.00297 **
## New_Prog_Syn_MsasssSI -2.139e-01  2.234e-01  -0.957  0.35110
## Mind_2_Synd_FLSI 3.428e-02  1.579e-01   0.217  0.83060
## Mind_2_Synd_BLSI 2.868e-02  1.611e-01   0.178  0.86067
## PCR_INICIOB  -2.462e-03  3.034e-03  -0.812  0.42759
## i_ASIDAS     4.040e-02  1.183e-01   0.341  0.73674
## HVGP         -9.490e-03  2.727e-02  -0.348  0.73190
## BASDAIDif    2.968e-02  2.128e-02   1.395  0.18000
## PCRDif       1.923e-03  3.235e-03   0.594  0.55972
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1369 on 18 degrees of freedom
## (42 observations deleted due to missingness)
## Multiple R-squared:  0.8972, Adjusted R-squared:  0.6688
## F-statistic: 3.929 on 40 and 18 DF, p-value: 0.001425

```

```
qqnorm(abc1$residuals)
qqline(abc1$residuals)
```



```
shapiro.test(abc1$residuals)
```

```
##
## Shapiro-Wilk normality test
##
## data: abc1$residuals
## W = 0.9513, p-value = 0.01936
```

```
which.max(abc1$residuals)
```

```
## 66
## 41
```

```
which.min(abc1$residuals)
```

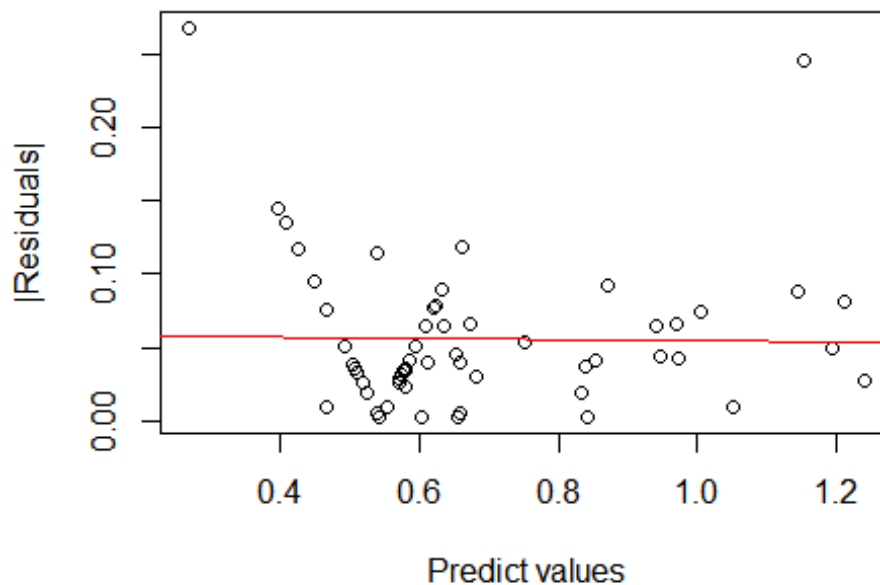
```
## 42
## 24
```

```
shapiro.test(abc1$residuals[c(-24, -41)])
```

```
##
## Shapiro-Wilk normality test
##
## data: abc1$residuals[c(-24, -41)]
## W = 0.97796, p-value = 0.3819
```

```
plot(fitted(abc1), abs(residuals(abc1)), xlab="Predict values", ylab="|Residuals|")
```

```
abline(lm(abs(residuals(abc1)) ~ fitted(abc1)), col="red", lwd=1.5)
```



```

library(lmtest)
bptest(abc1)

##
## studentized Breusch-Pagan test
##
## data: abc1
## BP = 44.25, df = 40, p-value = 0.2969

library(car)
dwt(abc1,alternative = "two.sided")

## lag Autocorrelation D-W Statistic p-value
## 1 0.01571134 1.963515 0.62
## Alternative hypothesis: rho != 0

library(car)
outlierTest(abc1)

## rstudent unadjusted p-value Bonferroni p
## 42 -4.813376 0.00016221 0.0095705
## 66 4.638805 0.00023484 0.0138550

summary(influence.measures(abc1))

## Potentially influential observations of
## lm(formula = LY ~ . - mSASSS_Progress, data = base_rx22, na.actio
n = na.omit) :
##
## dfb.1_ dfb.EDAD dfb.SEXO dfb.IMC dfb.`IMC dfb.TSIN dfb.HLAB dfb
.EANO
## 1 0.00 0.00 0.02 0.00 0.01 -0.01 0.01 -0

```

.01								
## 3	0.29	-0.21	-0.06	-0.06	-0.07	0.08	-0.06	-0
.06								
## 4	0.26	-0.23	-0.22	-0.10	0.34	0.13	-0.02	-0
.02								
## 5	0.11	-0.13	0.05	0.21	-0.16	0.13	0.06	-0
.13								
## 6	0.02	-0.02	0.23	0.10	-0.05	-0.15	0.26	-0
.12								
## 7	-0.33	-0.10	-1.04_*	-0.06	0.62	0.55	0.23	0
.77								
## 9	-0.07	0.10	0.12	0.07	-0.05	-0.04	-0.05	-0
.08								
## 11	-0.04	-0.01	0.16	0.08	0.03	-0.24	-0.16	0
.00								
## 13	-0.17	-0.05	0.12	0.23	0.04	-0.10	0.06	0
.06								
## 14	-0.20	0.45	0.01	-0.19	0.01	-0.40	-0.06	-0
.04								
## 15	0.00	0.00	0.02	0.00	0.01	-0.01	0.01	-0
.01								
## 19	0.01	-0.08	0.01	0.02	-0.07	0.05	0.00	-0
.02								
## 21	0.00	0.00	-0.02	0.01	-0.02	-0.06	-0.05	-0
.09								
## 35	0.08	-0.10	0.07	0.03	-0.05	-0.02	0.07	-0
.06								
## 36	0.94	-0.16	-0.49	-1.19_*	0.66	0.16	0.65	0
.29								
## 38	0.01	-0.06	0.21	0.10	-0.04	-0.16	0.00	-0
.12								
## 40	-0.48	0.02	-0.05	-0.11	0.32	-0.09	-0.14	0
.54								
## 42	-0.67	1.34_*	-0.50	-0.13	-0.14	-1.01_*	1.23_*	0
.62								
## 43	-0.38	0.27	-0.08	0.26	-0.02	-0.15	-0.17	0
.15								
## 44	0.13	-0.01	0.26	-0.10	0.17	-0.16	0.04	-0
.10								
## 45	-0.05	0.03	-0.03	0.02	-0.02	-0.10	-0.08	-0
.04								
## 46	0.00	0.00	0.00	0.00	0.00	0.01	0.00	0
.00								
## 47	-0.10	-0.02	0.01	0.08	-0.02	-0.02	-0.16	-0
.01								
## 49	0.03	-0.03	-0.01	0.00	-0.02	0.01	0.01	0
.01								
## 50	-0.04	-0.03	-0.01	0.01	-0.02	-0.01	-0.05	0
.05								
## 51	-0.39	-0.93	0.30	1.02_*	-1.07_*	-0.15	-0.07	-2
.70_*								
## 57	0.03	-0.22	-0.30	0.41	-0.19	0.30	-0.13	0
.18								
## 59	-0.01	0.00	0.00	0.01	0.00	0.00	-0.02	-0

.01								
## 63	0.00	0.00	0.02	0.00	0.01	-0.01	0.01	-0
.01								
## 64	-0.02	0.04	-0.04	0.04	-0.06	0.00	-0.12	0
.06								
## 65	-0.02	0.00	-0.05	0.00	0.00	0.02	-0.04	0
.04								
## 66	-0.43	0.72	-0.20	0.84	0.19	0.48	-0.96	-0
.24								
## 68	-0.03	0.01	-0.03	0.00	0.02	0.00	0.00	0
.01								
## 70	-0.04	-0.01	-0.13	-0.12	0.08	-0.07	0.14	0
.03								
## 71	0.05	-0.16	-0.21	0.28	-0.32	0.12	0.01	0
.11								
## 73	0.00	-0.02	-0.04	-0.02	0.07	0.07	0.07	0
.06								
## 76	-0.08	0.18	-0.01	0.07	-0.01	-0.15	-0.08	-0
.06								
## 85	0.00	-0.01	0.03	0.04	-0.09	0.00	-0.05	-0
.08								
## 86	0.06	-0.03	0.07	0.00	-0.03	-0.07	0.04	-0
.02								
## 87	0.08	0.13	-0.24	-0.07	0.10	0.14	-0.05	0
.15								
## 88	-0.04	0.01	0.07	0.14	-0.14	-0.02	-0.01	-0
.01								
## 94	-0.02	0.04	0.11	-0.25	0.30	-0.01	-0.30	-0
.05								
## 97	-0.04	0.22	-0.20	-0.19	-0.04	0.13	0.19	-0
.02								
## 99	-0.04	0.09	-0.12	0.11	-0.13	0.07	0.01	0
.12								
##	dfb.TABA	dfb.VSG	dfb.BASM	dfb.UVEI	dfb.PSOR	dfb.ENF_	dfb.TTOBIO	
A								
## 1	0.00	-0.02	-0.01	-0.01	-0.01	0.00	0.01	
## 3	-0.10	0.03	-0.03	-0.05	0.02	0.02	-0.01	
## 4	-0.13	0.24	-0.09	-0.10	-0.37	0.08	0.13	
## 5	-0.20	-0.12	-0.14	-0.18	-0.08	0.06	0.03	
## 6	-0.20	0.01	0.10	-0.15	-0.08	-0.07	0.19	
## 7	-0.02	-0.15	-0.06	-0.79	0.20	-0.60	0.90	
## 9	0.09	-0.08	0.03	-0.08	-0.02	0.01	-0.04	
## 11	0.06	-0.10	0.18	0.13	0.08	0.07	0.04	
## 13	-0.06	0.10	0.11	0.12	-0.06	-0.10	-0.04	
## 14	0.24	-0.54	0.22	0.34	0.12	0.00	0.28	
## 15	0.00	-0.02	-0.01	-0.01	-0.01	0.00	0.01	
## 19	-0.03	-0.04	-0.07	-0.01	0.04	0.00	-0.07	
## 21	-0.01	0.07	0.05	-0.04	0.01	0.04	0.02	
## 35	-0.05	0.04	0.04	-0.01	0.02	0.00	0.03	
## 36	-0.45	-0.55	0.69	0.26	0.39	-0.48	-0.38	
## 38	-0.01	-0.06	0.19	-0.12	-0.07	0.01	0.16	
## 40	0.21	-0.09	0.18	0.31	-0.95	-0.25	-0.12	
## 42	-1.49_*	0.35	0.65	-1.63_*	0.86	-0.81	-1.18_*	
## 43	0.15	0.32	-0.41	0.00	-0.24	-0.19	0.36	

## 44	-0.10	0.19	0.11	-0.06	0.01	-0.05	0.24
## 45	0.18	-0.03	0.00	-0.05	0.03	0.07	0.01
## 46	0.00	0.01	0.00	-0.01	0.00	0.00	0.01
## 47	0.02	-0.07	0.01	-0.01	-0.02	0.22	0.05
## 49	0.00	-0.04	0.01	0.01	-0.02	-0.01	-0.02
## 50	0.10	-0.04	0.04	0.00	-0.05	-0.20	0.01
## 51	0.41	-1.77_*	0.99	-0.12	1.09_*	2.48_*	-0.49
## 57	0.05	0.19	-0.64	0.22	-0.06	-0.31	0.24
## 59	-0.02	0.00	-0.01	0.00	0.01	-0.01	0.01
## 63	0.00	-0.02	-0.01	-0.01	-0.01	0.00	0.01
## 64	-0.01	0.14	-0.15	0.12	0.07	0.01	-0.11
## 65	0.07	-0.03	0.01	0.01	0.00	0.00	0.00
## 66	0.80	0.41	-0.71	-0.17	-0.86	0.46	-0.04
## 68	-0.01	0.01	0.03	0.04	0.02	0.00	-0.02
## 70	-0.19	0.05	0.14	0.05	0.04	-0.07	0.07
## 71	0.03	0.01	-0.30	0.13	-0.07	-0.16	-0.10
## 73	0.03	-0.11	0.01	0.00	-0.03	-0.06	0.03
## 76	-0.03	0.08	0.05	0.02	0.02	0.10	-0.11
## 85	0.05	-0.05	-0.09	0.09	0.01	0.06	-0.02
## 86	-0.04	0.05	0.03	-0.01	-0.01	-0.03	0.01
## 87	-0.06	-0.15	0.06	0.32	0.28	0.08	-0.61
## 88	0.00	-0.06	0.05	-0.11	-0.25	-0.02	0.04
## 94	0.29	-0.04	-0.21	0.37	0.06	0.31	-0.19
## 97	0.04	-0.12	-0.14	-0.01	-0.05	-0.10	0.07
## 99	0.02	0.11	0.09	0.20	0.06	-0.05	-0.19
##	dfb.TTOBIOI	dfb.TTOBIOG	dfb.TTOBIOC	dfb.NUM_TB_PREV.1	dfb.NUM_TB		
	_PREV.2						
## 1	0.01	0.00	0.04	0.01		0.01	
## 3	-0.01	0.12	-0.19	-0.02		0.05	
## 4	0.19	0.03	-0.03	0.26		0.26	
## 5	-0.26	-0.22	-0.06	0.06		-0.07	
## 6	0.14	0.02	0.04	-0.08		-0.08	
## 7	0.63	0.32	0.30	0.09		-0.70	
## 9	-0.04	-0.14	-0.04	-0.07		-0.04	
## 11	0.01	0.11	0.09	0.10		-0.14	
## 13	0.02	0.05	-0.01	0.04		0.05	
## 14	0.18	0.25	0.10	0.13		0.57	
## 15	0.01	0.00	0.04	0.01		0.01	
## 19	-0.12	-0.04	-0.05	-0.07		-0.05	
## 21	-0.03	0.00	0.01	0.00		0.04	
## 35	0.00	-0.13	-0.06	-0.02		-0.02	
## 36	0.17	0.03	-0.33	-0.35		-0.57	
## 38	0.07	-0.14	0.05	-0.01		-0.02	
## 40	-0.39	0.40	0.11	1.31_*		1.04_*	
## 42	0.97	1.02_*	-0.02	-1.69_*		-1.85_*	
## 43	0.27	0.21	0.14	0.45		0.34	
## 44	0.10	-0.15	-0.08	0.01		-0.01	
## 45	-0.12	0.05	0.12	0.08		0.09	
## 46	0.00	0.00	0.00	0.00		0.01	
## 47	-0.02	0.04	0.09	0.00		0.00	
## 49	0.00	-0.02	-0.01	0.00		-0.02	
## 50	0.05	0.10	-0.03	0.00		0.02	
## 51	-0.28	-0.23	0.28	-0.62		-0.87	
## 57	-0.01	-0.11	-0.34	0.18		-0.09	

## 59	-0.01	-0.01	0.02	0.00	0.00	0.00	
## 63	0.01	0.00	-0.13	0.01	0.01	0.01	
## 64	-0.10	0.04	-0.09	0.03	0.03	0.09	
## 65	0.01	0.05	0.00	0.02	0.02	-0.01	
## 66	0.94	-0.56	0.33	0.43	0.43	0.92	
## 68	0.00	0.03	-0.01	-0.01	-0.01	-0.02	
## 70	0.08	0.16	0.03	-0.18	-0.18	-0.08	
## 71	-0.07	0.05	-0.07	0.04	0.04	0.04	
## 73	0.01	0.00	-0.01	0.04	0.04	-0.02	
## 76	0.07	0.01	0.01	0.00	0.00	-0.04	
## 85	-0.16	-0.07	-0.04	-0.01	-0.01	0.00	
## 86	0.01	-0.02	0.00	-0.01	-0.01	0.03	
## 87	-0.27	-0.01	-0.36	-0.09	-0.09	-0.16	
## 88	0.04	-0.05	0.06	0.11	0.11	0.07	
## 94	-0.23	-0.14	0.01	0.22	0.22	0.14	
## 97	0.07	-0.03	-0.07	-0.07	-0.07	-0.42	
## 99	-0.19	0.02	-0.10	0.01	0.01	-0.03	
##	dfb.MOTIVOI	dfb.MOTIVOS	dfb.TIEM	dfb.BIO_	dfb.B_4_	dfb.Intr	dfb.
AINE							
## 100	-0.02	-0.01	0.00	0.00	0.00	0.00	0.
## 303	0.04	-0.05	0.10	-0.21	0.00	-0.06	-0.
## 424	-0.36	-0.16	-0.22	-0.01	0.14	-0.14	-0.
## 508	-0.12	-0.07	0.14	-0.12	-0.07	-0.24	-0.
## 602	0.06	0.03	0.18	-0.20	-0.07	0.00	0.
## 794	-0.12	0.86	-0.61	-0.04	-0.20	0.48	-0.
## 914	0.05	0.04	0.00	-0.10	-0.02	-0.06	0.
## 1107	-0.19	-0.13	-0.10	0.23	0.06	-0.02	-0.
## 1304	0.00	-0.04	-0.08	0.14	0.12	0.05	0.
## 1400	-0.16	0.00	-0.40	0.49	0.37	-0.19	0.
## 1500	-0.02	-0.01	0.00	0.00	0.00	0.00	0.
## 1903	0.04	0.05	0.04	0.04	-0.06	0.07	0.
## 2102	-0.03	-0.02	0.05	0.06	-0.04	-0.01	-0.
## 3502	0.00	-0.03	0.03	-0.09	0.02	-0.05	0.
## 3621	0.82	0.29	0.04	-0.18	-0.19	-0.37	-0.
## 3810	-0.05	-0.07	0.00	-0.01	-0.07	-0.20	0.
## 4018	-1.28_*	-0.97	0.30	0.04	-0.40	0.69	-0.
## 42	1.40_*	1.47_*	0.48	1.39_*	-0.36	-0.07	-2.

02_*							
## 43	-0.31	-0.14	-0.17	0.10	-0.19	0.17	-0.
33							
## 44	0.01	-0.25	0.13	-0.17	-0.21	-0.09	-0.
11							
## 45	-0.07	-0.09	-0.12	0.10	0.13	0.05	0.
04							
## 46	0.00	0.00	0.00	0.00	-0.01	0.00	0.
01							
## 47	0.02	0.05	-0.03	0.13	-0.03	-0.01	0.
06							
## 49	-0.01	-0.01	-0.01	0.01	0.01	-0.03	-0.
01							
## 50	0.06	0.02	-0.04	0.00	0.01	0.05	-0.
07							
## 51	0.29	0.72	0.17	0.46	-0.67	0.19	-0.
63							
## 57	0.03	0.41	0.03	-0.22	-0.71	-0.20	-0.
24							
## 59	-0.01	0.01	0.00	0.02	-0.01	-0.01	0.
02							
## 63	-0.02	-0.01	0.00	0.00	0.00	0.00	0.
00							
## 64	-0.06	-0.05	-0.01	0.01	0.04	0.10	0.
05							
## 65	-0.02	0.00	-0.04	0.01	0.01	0.02	-0.
03							
## 66	-0.21	-0.39	0.99	-0.68	-0.69	0.01	0.
56							
## 68	0.01	0.01	-0.03	0.02	0.02	0.03	-0.
04							
## 70	0.18	0.21	0.14	-0.02	0.04	0.12	-0.
04							
## 71	-0.02	-0.21	0.17	-0.01	-0.16	-0.19	0.
02							
## 73	-0.02	0.02	-0.05	0.00	-0.03	-0.02	-0.
04							
## 76	-0.05	-0.02	-0.08	0.00	0.04	-0.08	-0.
10							
## 85	-0.01	-0.02	0.17	-0.07	-0.14	0.03	0.
02							
## 86	-0.01	0.05	0.11	-0.04	-0.03	-0.08	0.
00							
## 87	-0.07	0.01	-0.25	-0.04	0.20	-0.09	0.
03							
## 88	-0.10	-0.08	-0.01	0.02	-0.04	0.00	0.
00							
## 94	-0.12	-0.16	-0.15	-0.09	-0.09	0.11	0.
27							
## 97	0.17	0.16	-0.05	-0.07	0.08	-0.06	0.
08							
## 99	-0.01	0.04	0.07	-0.15	-0.09	-0.17	0.
10							
##	dfb.FAME	dfb.i_BA	dfb.EGM	dfb.N_S_	dfb.P_S_	dfb.mSAS	dfb.N_P_ d

fb.M_2_S_F							
## 1	0.01	0.02	-0.02	-0.01	-0.01	0.01	0.02
0.03							
## 3	0.01	0.00	0.01	-0.01	-0.11	0.00	0.01
-0.05							
## 4	0.05	0.43	-0.02	0.05	-0.03	0.28	-0.03
0.17							
## 5	0.22	-0.04	0.19	-0.08	-0.12	0.03	0.13
0.01							
## 6	0.05	0.11	-0.08	-0.04	-0.06	-0.07	0.03
0.07							
## 7	1.08_*	0.97	-0.45	-0.65	0.28	-0.30	0.77
0.49							
## 9	0.00	-0.10	-0.02	-0.06	-0.02	-0.06	0.04
-0.05							
## 11	-0.02	0.03	0.01	-0.15	0.15	-0.02	0.10
0.09							
## 13	-0.20	0.19	-0.05	0.23	0.06	-0.01	-0.18
-0.17							
## 14	-0.33	0.07	0.05	1.29_*	0.43	0.06	-1.30_*
0.04							
## 15	0.01	0.02	-0.02	-0.01	-0.01	0.01	0.02
-0.16							
## 19	0.03	-0.01	0.05	-0.01	-0.03	-0.12	0.06
0.04							
## 21	0.08	-0.06	0.10	0.03	0.03	-0.01	-0.05
-0.04							
## 35	0.03	0.05	-0.01	-0.01	-0.09	-0.02	0.04
0.01							
## 36	-0.41	0.41	-0.45	-0.17	-0.29	-0.01	0.02
-0.10							
## 38	0.08	0.03	-0.08	-0.07	-0.04	0.18	-0.04
0.05							
## 40	0.14	-0.11	-0.27	0.16	0.10	-0.18	-0.21
0.05							
## 42	0.51	0.74	-1.41_*	-0.41	0.63	-1.37_*	0.20
0.91							
## 43	-0.08	-0.10	0.15	-0.08	-0.26	0.08	0.14
0.14							
## 44	0.00	0.22	-0.16	-0.02	-0.13	-0.25	0.16
0.05							
## 45	-0.03	-0.14	0.05	-0.07	-0.03	0.03	0.06
-0.05							
## 46	0.00	-0.01	0.00	0.00	0.00	0.00	0.00
0.00							
## 47	0.02	0.03	0.12	0.09	0.04	-0.06	-0.05
0.02							
## 49	0.04	-0.03	0.00	-0.03	0.01	0.00	0.02
0.01							
## 50	-0.02	-0.05	-0.02	-0.01	0.00	-0.06	-0.02
-0.03							
## 51	2.53_*	-1.59_*	-0.56	-0.97	0.70	-1.54_*	0.87
0.54							
## 57	0.06	0.28	0.17	-0.23	-0.44	-0.19	0.37

0.15							
## 59	0.01	0.02	0.02	0.02	0.01	0.00	0.00
0.01							
## 63	0.01	0.02	-0.02	-0.01	-0.01	0.01	0.02
0.03							
## 64	-0.10	0.08	0.13	0.17	0.10	0.02	-0.12
0.00							
## 65	0.01	0.02	-0.03	-0.04	0.00	-0.03	0.03
0.02							
## 66	-0.54	-1.35_*	1.40_*	0.75	-1.90_*	0.91	0.10
-1.01_*							
## 68	0.00	0.01	-0.02	0.00	0.02	-0.01	-0.02
0.02							
## 70	0.01	0.08	0.04	0.16	0.02	-0.08	-0.15
0.05							
## 71	-0.03	0.00	0.31	0.07	-0.12	0.10	-0.07
-0.09							
## 73	0.00	-0.05	0.02	-0.01	0.08	-0.01	-0.02
0.07							
## 76	0.07	-0.01	-0.16	0.03	0.22	0.07	-0.08
0.00							
## 85	0.04	-0.10	0.06	0.00	-0.04	-0.06	0.04
-0.03							
## 86	-0.02	0.04	0.03	0.06	-0.02	-0.03	-0.03
-0.05							
## 87	-0.05	0.07	0.04	-0.42	-0.25	0.07	0.21
0.19							
## 88	0.13	-0.06	0.01	-0.05	0.03	-0.15	0.08
0.01							
## 94	-0.09	-0.29	-0.10	-0.08	-0.05	0.34	0.10
-0.12							
## 97	-0.12	-0.01	0.01	-0.22	0.00	0.05	0.17
-0.10							
## 99	0.01	-0.08	-0.39	0.09	0.05	-0.04	-0.05
0.01							
##	dfb.M_2_S_B	dfb.PCR_	dfb.i_AS	dfb.HVGP	dfb.BASD	dfb.PCRD	dffit
cov.r							
## 1	-0.03	0.04	-0.02	0.01	-0.01	-0.03	0.21
130.94_*							
## 3	0.08	0.15	-0.20	0.09	-0.09	-0.07	0.65
18.07_*							
## 4	-0.22	0.06	-0.20	-0.13	-0.10	0.12	-1.12
31.72_*							
## 5	-0.06	0.11	-0.06	-0.05	-0.06	0.18	-0.63
15.88_*							
## 6	-0.02	0.05	-0.06	-0.12	-0.12	0.15	0.66
14.24_*							
## 7	-0.62	-0.16	0.28	-0.49	-0.38	0.50	4.18
0.00							
## 9	0.07	-0.08	0.16	-0.15	0.05	0.02	0.44
14.55_*							
## 11	-0.10	0.10	-0.07	0.03	-0.09	0.05	-1.00
24.58_*							
## 13	0.16	-0.08	-0.05	-0.08	-0.03	-0.34	0.77

24.83_*							
## 14	-0.10	0.08	0.23	-0.22	0.11	-0.21	-2.98
1.75							
## 15	0.16	0.04	-0.02	0.01	-0.01	-0.03	-0.21
130.94_*							
## 19	0.03	0.00	0.00	-0.02	0.03	-0.01	0.27
20.23_*							
## 21	0.03	-0.07	0.04	-0.04	-0.04	-0.02	0.28
47.04_*							
## 35	0.02	0.04	-0.07	-0.04	0.00	0.01	-0.24
42.59_*							
## 36	-0.01	0.90	-0.44	0.45	-0.14	-0.22	2.13
0.20							
## 38	-0.06	0.02	0.08	-0.18	0.07	0.14	0.56
17.51_*							
## 40	0.00	-0.28	0.37	0.20	0.03	-0.29	2.48
2.98							
## 42	0.03	-0.28	0.36	-0.78	0.71	-0.60	-6.28_
* 0.00							
## 43	-0.19	-0.76	0.35	-0.18	0.27	-0.09	-1.74
16.74_*							
## 44	-0.06	0.00	0.03	-0.32	0.08	0.06	-0.95
17.89_*							
## 45	0.04	0.00	0.08	-0.04	0.02	0.12	-0.49
25.42_*							
## 46	0.00	0.00	0.00	0.00	0.00	0.00	0.03
21.87_*							
## 47	0.00	-0.05	0.07	-0.09	0.07	0.02	0.63
16.69_*							
## 49	-0.01	0.01	0.01	-0.01	0.00	0.01	0.14
41.46_*							
## 50	0.09	0.04	0.02	0.07	0.00	-0.11	-0.49
17.77_*							
## 51	-0.39	0.17	0.79	0.28	-0.65	0.90	-4.39
0.05							
## 57	-0.06	0.58	-0.26	0.16	0.47	0.31	2.36
7.87_*							
## 59	-0.02	-0.02	0.01	-0.03	0.01	0.03	-0.12
41.73_*							
## 63	-0.03	0.04	-0.02	0.01	-0.01	-0.03	-0.21
130.94_*							
## 64	0.02	-0.08	-0.10	0.04	0.11	-0.09	0.43
22.05_*							
## 65	-0.03	0.03	-0.01	0.06	0.09	-0.02	0.18
19.27_*							
## 66	0.38	-0.47	0.18	-0.16	-0.36	1.07_*	6.74_
* 0.00							
## 68	-0.03	0.00	0.01	0.01	0.00	-0.02	0.10
21.21_*							
## 70	0.03	-0.08	0.02	-0.03	-0.12	-0.11	-0.46
22.30_*							
## 71	0.14	0.24	-0.33	0.42	-0.03	-0.04	-0.96
23.79_*							
## 73	-0.08	0.03	0.04	0.04	0.08	-0.01	0.31

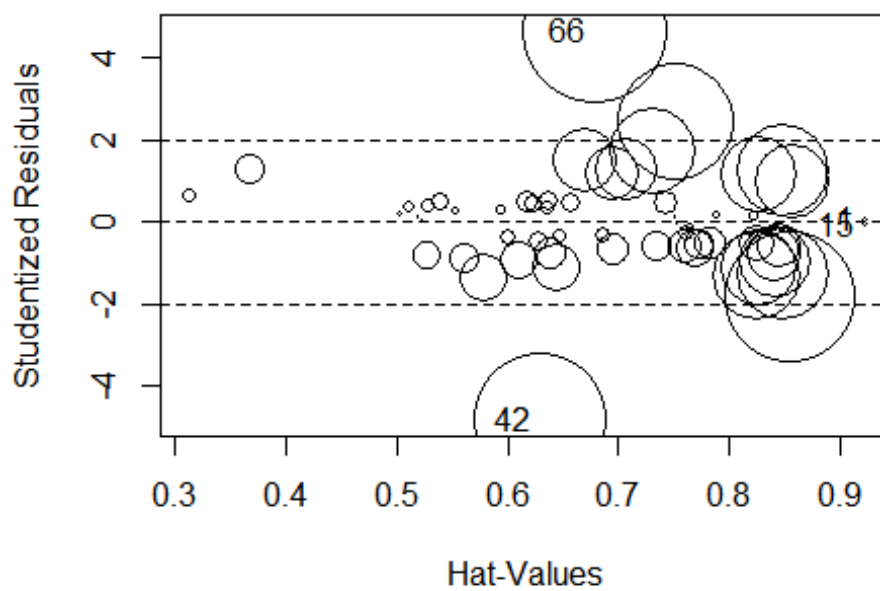
```

55.61_*
## 76 -0.06      -0.07      0.08      -0.02      -0.05      0.10      0.56
11.87_*
## 85  0.01      -0.05      0.06      -0.05      0.05      0.03      0.37
15.63_*
## 86  0.05      0.00      -0.08      0.01      -0.06      -0.05      0.34
21.10_*
## 87 -0.24      0.34      -0.19      0.34      0.28      -0.03      -1.19
16.28_*
## 88 -0.01      0.05      -0.02      0.01      -0.07      0.13      -0.50
35.90_*
## 94 -0.08      -0.17      0.42      -0.12      0.24      0.14      -1.00
12.05_*
## 97  0.08      -0.09      0.19      -0.14      0.16      -0.16      -1.36
30.19_*
## 99 -0.01      -0.22      0.02      0.15      0.04      -0.15      -1.07
18.16_*
##      cook.d hat
## 1      0.00      0.92
## 3      0.01      0.66
## 4      0.03      0.83
## 5      0.01      0.63
## 6      0.01      0.62
## 7      0.34      0.75
## 9      0.00      0.53
## 11     0.03      0.78
## 13     0.02      0.74
## 14     0.21      0.85
## 15     0.00      0.92
## 19     0.00      0.55
## 21     0.00      0.79
## 35     0.00      0.77
## 36     0.10      0.67
## 38     0.01      0.62
## 40     0.15      0.83
## 42     0.43      0.63
## 43     0.08      0.84
## 44     0.02      0.73
## 45     0.01      0.69
## 46     0.00      0.52
## 47     0.01      0.64
## 49     0.00      0.75
## 50     0.01      0.60
## 51     0.42      0.85
## 57     0.14      0.86
## 59     0.00      0.75
## 63     0.00      0.92
## 64     0.00      0.64
## 65     0.00      0.50
## 66     0.52      0.68
## 68     0.00      0.52
## 70     0.01      0.65
## 71     0.02      0.77
## 73     0.00      0.82

```

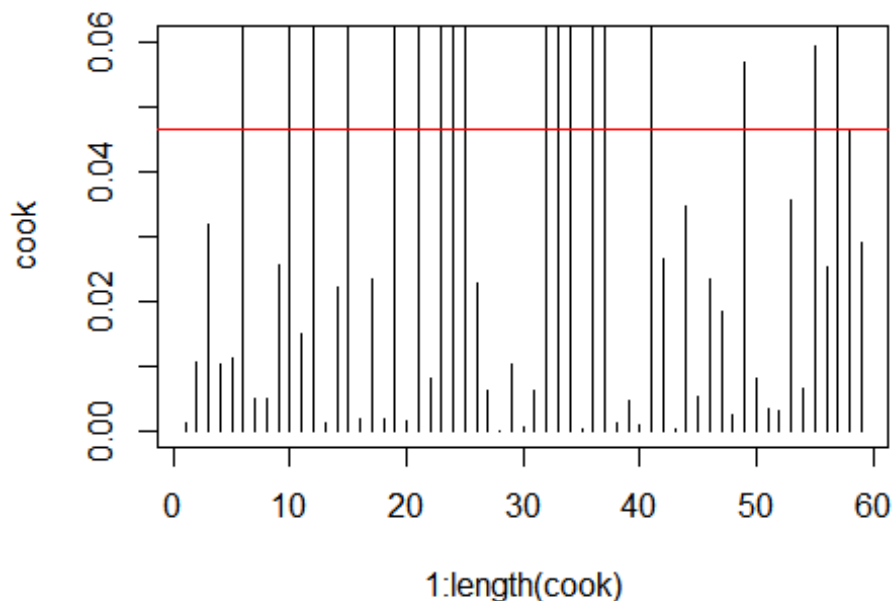
```
## 76  0.01  0.54
## 85  0.00  0.51
## 86  0.00  0.59
## 87  0.04  0.77
## 88  0.01  0.76
## 94  0.03  0.70
## 97  0.05  0.85
## 99  0.03  0.76
```

```
influencePlot(abc1)
```



```
##      StudRes      Hat      CookD
## 1  0.0623331 0.9211843 0.001172496
## 15 -0.0623331 0.9211843 0.001172496
## 42 -4.8133759 0.6302185 0.431566423
## 66  4.6388049 0.6785118 0.517635056
```

```
cook <- cooks.distance(abc1)
plot(1:length(cook), cook, ylim=c(0,0.06), type = "h")
abline(h=4/((n-p-2)), col="red")
```



```

stud <- rstudent(abc1)
head(sort(abs(stud),decreasing=TRUE))

##      42      66      7      51      61      36
## 4.813376 4.638805 2.399791 1.816236 1.745187 1.496918

hatv <- hatvalues(abc1)
head(sort(hatv,decreasing=T))

##      1      15      63      57      51      14
## 0.9211843 0.9211843 0.9211843 0.8551536 0.8541275 0.8480441

library(leaps)
attach(base_rx22)

## The following objects are masked from base_rx22 (pos = 3):
##
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
## ss,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
## EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx21 (pos = 4):
##
## AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM
## E,
## HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,
## Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,

```



```

##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx21 (pos = 5):
##
##      AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM
E,
##      HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,
##      Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,
##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx2:
##
##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx.num2:
##
##      BASDAIDif, BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC,
##      Interval, mSASSS_Progress, PCR_INICIOB, PCRDif, TIEMBIO, TSINT,
VSG

## The following objects are masked from base_rx3:
##
##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, SEXO, TABACO_Kat,
##      TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 14):
##
##      AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
##      ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
##      Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
##      mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
##      NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
##      TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 15):
##

```

```

## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_AS DAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
## SS,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
## EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

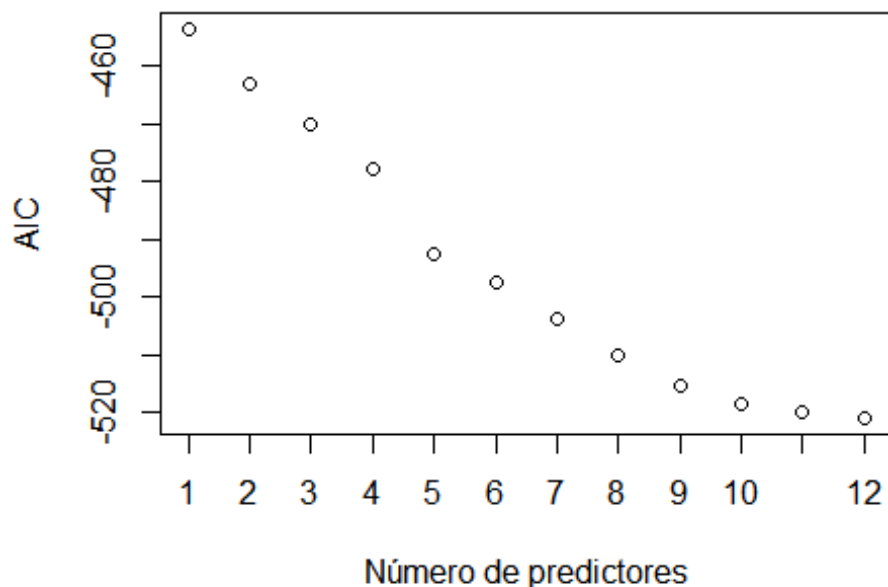
## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, HVGP, i_AS DAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

b1 <- regsubsets(LY~ .-mSASSS_Progress, data=base_rx22, na.action = na
.omit, nvmax=12)
rs1 <- summary(b1)
n1 <- nrow(base_rx22)
k1 <- length(rs1$rss) # Número de variables predictoras
p1 <- k1 + 1 # Número de parámetros (incluye la intercepción)
(AIC <- n*log(rs1$rss/n1) + (2:p1)*2)

## [1] -453.6466 -463.1659 -470.0006 -477.9064 -492.6002 -497.5672 -5
## [8] -510.2430 -515.3117 -518.4770 -519.9174 -520.9258

plot(1:k1, AIC, ylab="AIC", xlab="Número de predictores", axes=F)
box(); axis(1,at=1:k1); axis(2)

```



```
rs1$outmat[10,]
```

```

##          EDAD          SEXOFEMALE
IMC
##          "*"          " "
" "
##          `IMC>30`SI          TSINT          HLAB
27NO
##          " "          " "
" "
##          EANO          TABACO_KatNO
VSG
##          "*"          " "
" "
##          BASMI          UVEITISSI          PSORIAS
ISSI
##          " "          " "
" "
##          ENF_INFSI          TTOBIOADA          TTOBI
OIFX
##          " "          "*"
" "
##          TTOBIOGOL          TTOBIOCTZ NUM_TB_PREVtreatmen
ts.1
##          " "          "*"
" "
## NUM_TB_PREVtreatments.2          MOTIVOInefficacy          MOTIVOSide.eff
ects
##          "*"          "*"
" "
##          TIEMBIO          BIO_INICISI          Biol_4_k
atSI
##          " "          "*"
" "
##          Interval          AINESI          FA
MESI
##          " "          " "
" "
##          i_BASDAI          EGM          New_Syn_mSAS
SSSI
##          " "          " "
" "
##          Prog_Syn_mSASSSI          mSASSS_Progress_KatSI          New_Prog_Syn_Msas
ssSI
##          " "          "*"
" "
##          Mind_2_Synd_FLSI          Mind_2_Synd_BLSI          PCR_INI
CIOB
##          "*"          " "
" "
##          i_ASIDAS          HVGP          BASDA
IDif
##          " "          " "
"*"
##          PCRDif
##          " "

```

`attach(base_rx22)`

The following objects are masked from base_rx22 (pos = 3):

```
##  
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,  
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,  
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre  
ss,  
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,  
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S  
EXO,  
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG
```

The following objects are masked from base_rx22 (pos = 4):

```
##  
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,  
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,  
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre  
ss,  
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,  
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S  
EXO,  
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG
```

The following objects are masked from base_rx21 (pos = 5):

```
##  
## AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM  
E,  
## HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,  
## Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,  
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,  
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S  
EXO,  
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG
```

The following objects are masked from base_rx21 (pos = 6):

```
##  
## AINE, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM, ENF_INF, FAM  
E,  
## HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30, Interval,  
## Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progress,  
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,  
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S  
EXO,  
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG
```

The following objects are masked from c_base_rx2:

```
##  
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,  
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,  
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre  
ss,  
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,  
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
```

```

EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx.num2:
##
## BASDAIDif, BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC,
## Interval, mSASSS_Progress, PCR_INICIOB, PCRDif, TIEMBIO, TSINT,
VSG

## The following objects are masked from base_rx3:
##
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, PSORIASIS, SEXO, TABACO_Kat,
## TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 15):
##
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from base_rx2 (pos = 16):
##
## AINE, BASDAIDif, BASMI, BIO_INICI, Biol_4_kat, EA, EDAD, EGM,
## ENF_INF, FAME, HLAB27, HVGP, i_ASDAS, i_BASDAI, IMC, IMC>30,
## Interval, Mind_2_Synd_BL, Mind_2_Synd_FL, MOTIVO, mSASSS_Progre
SS,
## mSASSS_Progress_Kat, New_Prog_Syn_Msasss, New_Syn_mSASSS,
## NUM_TB_PREV, PCR_INICIOB, PCRDif, Prog_Syn_mSASSS, PSORIASIS, S
EXO,
## TABACO_Kat, TIEMBIO, TSINT, TTOBIO, UVEITIS, VSG

## The following objects are masked from c_base_rx.num:
##
## BASMI, EDAD, EGM, HVGP, i_ASDAS, i_BASDAI, IMC, Interval,
## mSASSS_Progress, PCR_INICIOB, TIEMBIO, TSINT, VSG

s.aic1<-lm(LY ~ EDAD + EA + TTOBIO + NUM_TB_PREV + MOTIVO + BIO_INICI
+ Mind_2_Synd_FL+ mSASSS_Progress_Kat + BASDAIDif, na.action = na.omit
)
summary(s.aic1)

##
## Call:
## lm(formula = LY ~ EDAD + EA + TTOBIO + NUM_TB_PREV + MOTIVO +
## BIO_INICI + Mind_2_Synd_FL + mSASSS_Progress_Kat + BASDAIDif,
## na.action = na.omit)

```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.44770 -0.05367 -0.01084  0.05426  0.32929
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.552830   0.059747   9.253 1.35e-13 ***
## EDAD           0.003388   0.001228   2.758 0.007484 **
## EANO           0.083079   0.038317   2.168 0.033699 *
## TTOBIOADA      -0.086619   0.029872  -2.900 0.005046 **
## TTOBIOIFX       0.038237   0.044039   0.868 0.388352
## TTOBIOGOL      -0.081640   0.048821  -1.672 0.099142 .
## TTOBIOCTZ      -0.209352   0.074456  -2.812 0.006457 **
## NUM_TB_PREVtreatments.1  0.024798   0.083548   0.297 0.767530
## NUM_TB_PREVtreatments.2 -0.181983   0.104824  -1.736 0.087146 .
## MOTIVOIinefficacy    0.053790   0.088756   0.606 0.546532
## MOTIVOSide.effects  -0.087947   0.092211  -0.954 0.343636
## BIO_INICISI        -0.130944   0.037371  -3.504 0.000822 ***
## Mind_2_Synd_FLSI     0.030141   0.033644   0.896 0.373520
## mSASSS_Progress_KatSI  0.342374   0.034759   9.850 1.18e-14 ***
## BASDAIDif          0.028008   0.006865   4.080 0.000122 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1073 on 67 degrees of freedom
## (19 observations deleted due to missingness)
## Multiple R-squared:  0.808, Adjusted R-squared:  0.7679
## F-statistic: 20.14 on 14 and 67 DF, p-value: < 2.2e-16
```

```
coef(s.aic1)
```

```
##              (Intercept)              EDAD
EANO
##              0.55282970              0.00338837              0.0830
7935
##              TTOBIOADA              TTOBIOIFX              TTOBI
OGOL
##              -0.08661902              0.03823681              -0.0816
3998
##              TTOBIOCTZ NUM_TB_PREVtreatments.1 NUM_TB_PREVtreatmen
ts.2
##              -0.20935230              0.02479779              -0.1819
8295
##              MOTIVOIinefficacy              MOTIVOSide.effects              BIO_INI
CISI
##              0.05379000              -0.08794693              -0.1309
4378
##              Mind_2_Synd_FLSI              mSASSS_Progress_KatSI              BASDA
IDif
##              0.03014132              0.34237371              0.0280
0761
```

```
(summary(s.aic1)$coefficients["BASDAIDif",1] +
qnorm(c(0.025,0.975)) * summary(s.aic1)$coefficients["BASDAIDif",2])
```

```
## [1] 0.01455265 0.04146257
```