

15FuncionesEstadisticasBasicas_cheatsheet.R

moka

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```
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# Objetivo:
# Referencia: Basado en R Programming Fundamentals, StanfordOnline XDFS112

require(ggplot2)
library(dplyr)
# Limpiar el espacio de trabajo
rm(list=ls())

# Configurar el directorio

midirectorio<-setwd("~/Dropbox/0.POST-PHD/GOALS/2.CODE/R/Ecomienza/15FuncionesEstadisticasBasicas")
midirectorio

## [1] "/Users/moka/Dropbox/0.POST-PHD/GOALS/2.CODE/R/Ecomienza/15FuncionesEstadisticasBasicas"

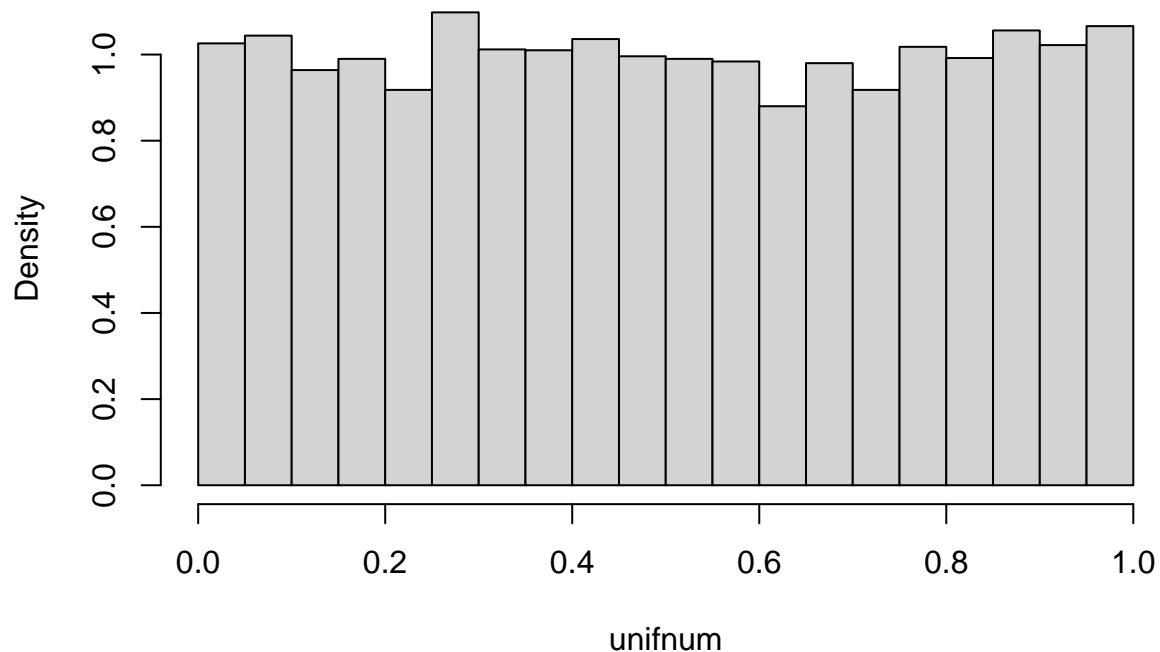
# Establecer la semilla
set.seed(1)

# Generar numeros aleatorios

? Distributions

# Distribucion uniforme
unifnum=runif(10000,0,1)
hist(unifnum,freq=F)
```

Histogram of unifnum



```
summary(unifnum)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.0001064 0.2525854 0.4957413 0.5001680 0.7568858 0.9999306
```

```
# Cambiar el intervalo a -10,10
```

```
# Para simulaciones es mejor utilizar replicate(), apply(), tambien es mejor usar funciones que ya han
```

```
reps=10000
```

```
system.time(x1<-colSums(matrix(runif(100*5),nrow=5)))
```

```
##      user  system elapsed
```

```
##       0       0       0
```

```
system.time(x1<-apply(matrix(runif(100*5),ncol=5),2,sum))
```

```
##      user  system elapsed
```

```
##    0.010   0.000   0.011
```

```
system.time(x1<-lapply(1:reps,function(i){sum(runif(n=5))}))
```

```
##      user  system elapsed
```

```
##    0.013   0.001   0.013
```

```
system.time(x1<-sapply(1:reps,function(i){sum(runif(n=5))}))
```

```
##      user  system elapsed
```

```
##    0.013   0.001   0.014
```

```
system.time(sum5<-replicate(reps,sum(runif(5))))
```

```
##      user  system elapsed
```

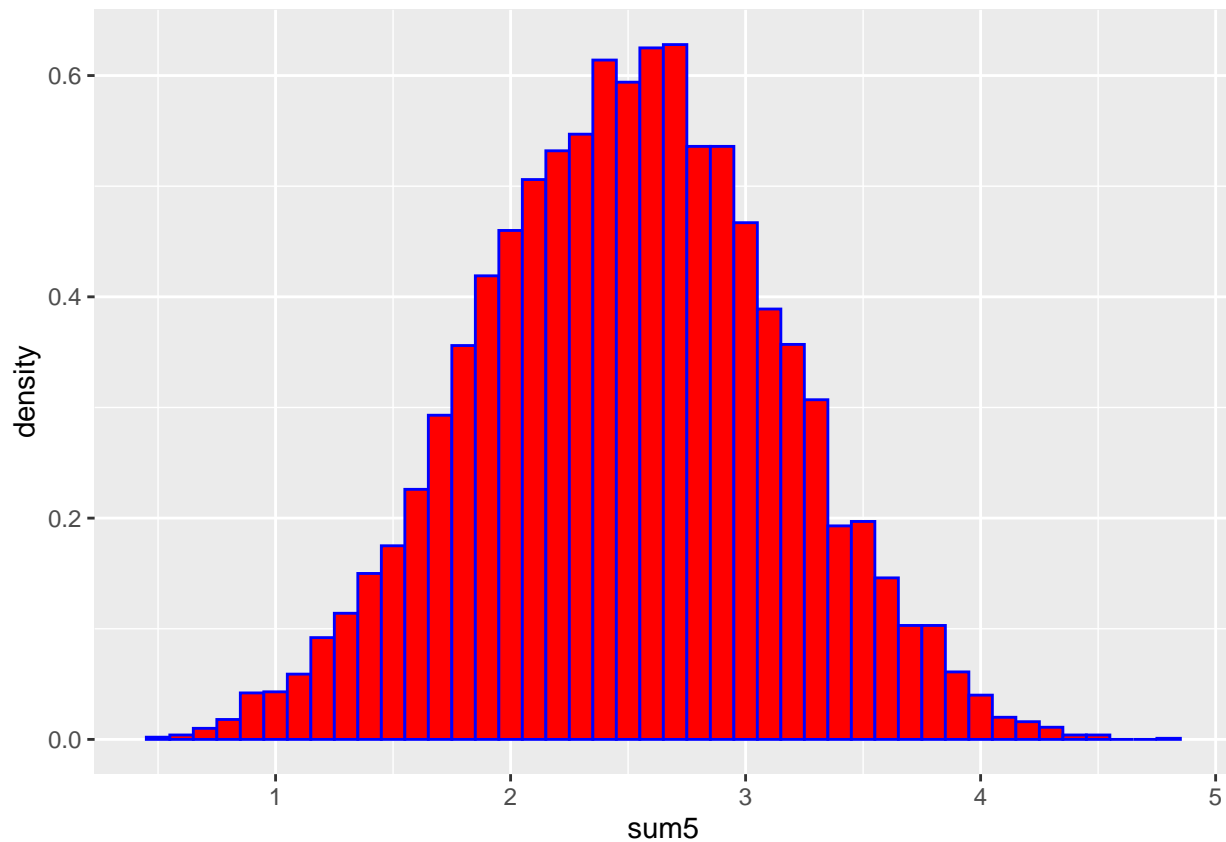
```
##    0.012   0.001   0.013
```

```
d5=data.frame(sum5)
summary(sum5) # Simetrica, media y mediana iguales.
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.4766  2.0565  2.5073  2.4985  2.9341  4.7685
```

```
# Para usar ggplot2, primero hay que declarar dataframe y la variable
#luego el tipo de grafico.
```

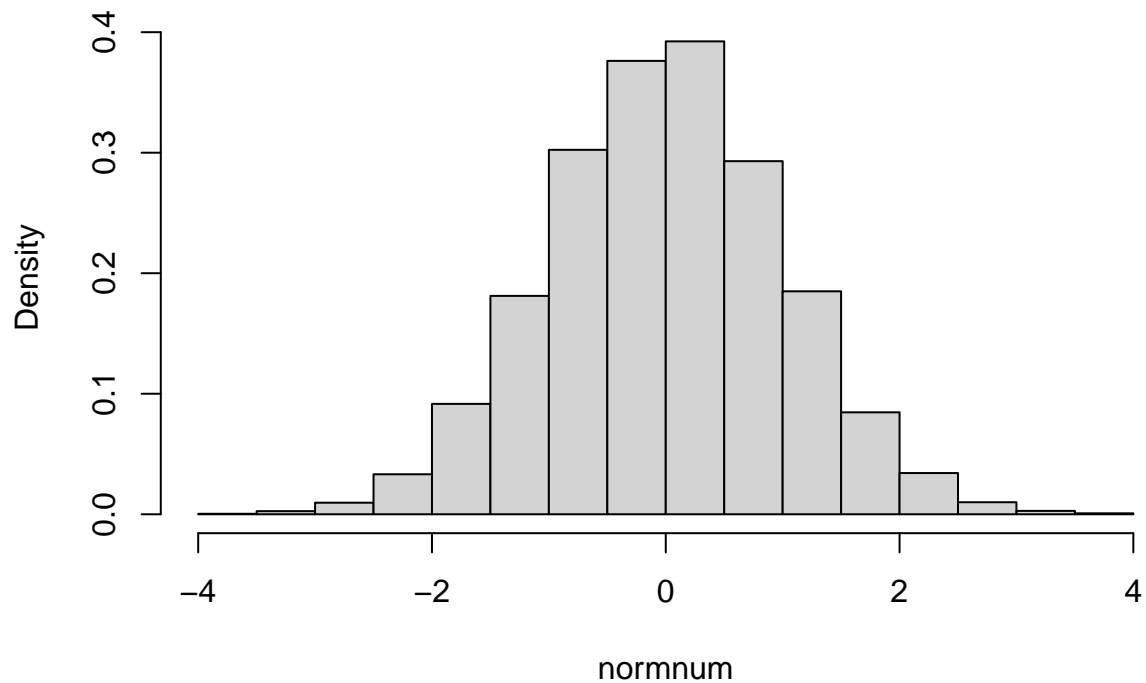
```
ggplot(d5,aes(sum5))+geom_histogram(aes(y=..density..),binwidth=0.1,color="blue",fill="red")
```



```
?geom_histogram
```

```
# Distribucion normal
normnum=rnorm(10000,0,1)
dn=data.frame(normnum)
hist(normnum,freq=F)
```

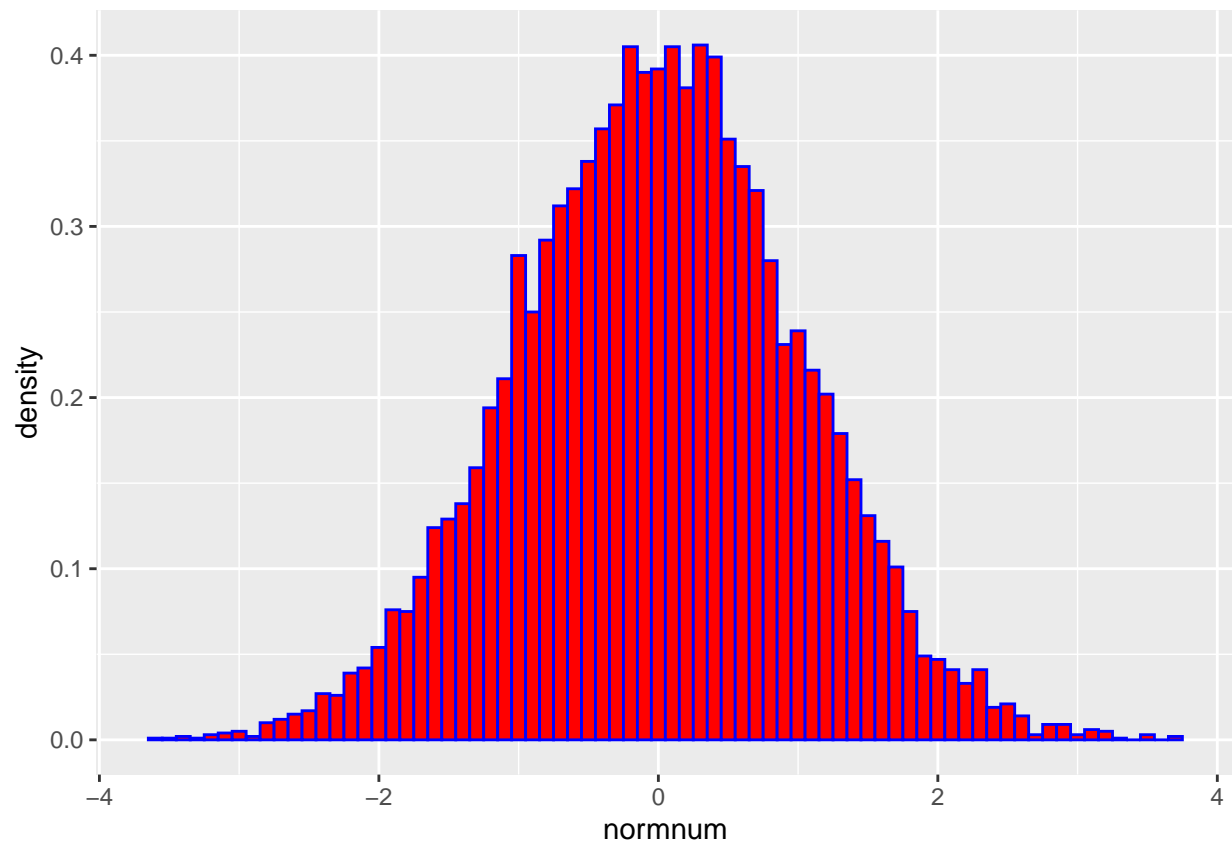
Histogram of normnum



```
summary(normnum)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## -3.604587 -0.680872  0.003936 -0.001864  0.663745  3.724271
```

```
ggplot(dn,aes(normnum))+geom_histogram(aes(y=..density..),binwidth=0.1,color="blue",fill="red")
```



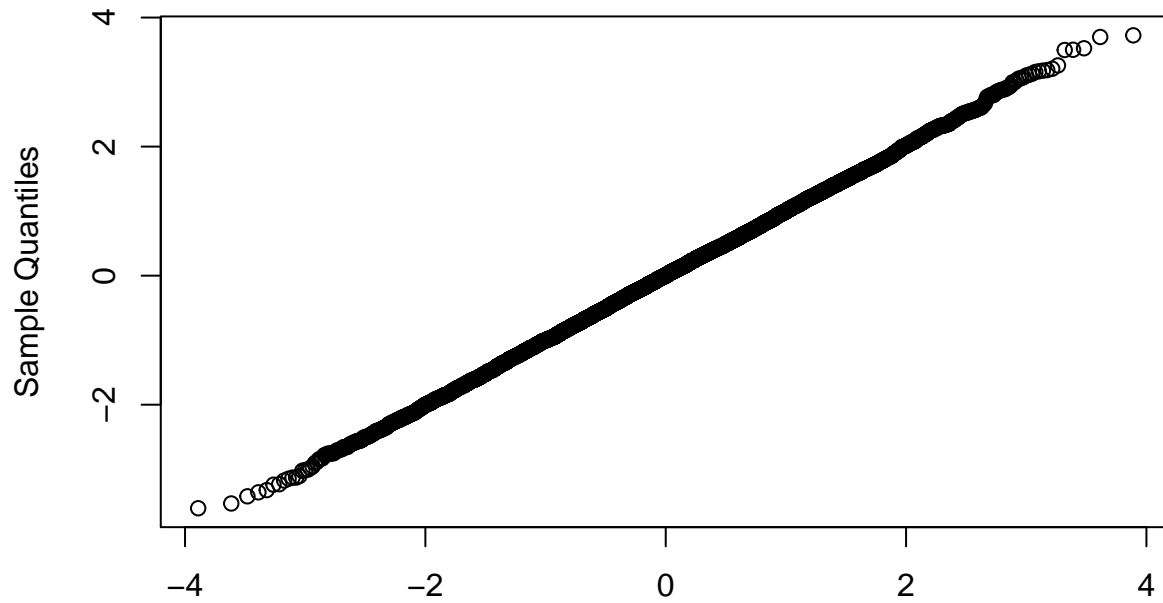
```
# Cambiar parametros a  $\mu=1$ ,  $\sigma=10$ 
```

```
# Comparar distribuciones: Cuantil Cuantil plot (qqplot)
```

```
?qqnorm
```

```
qqnorm(normnum)
```

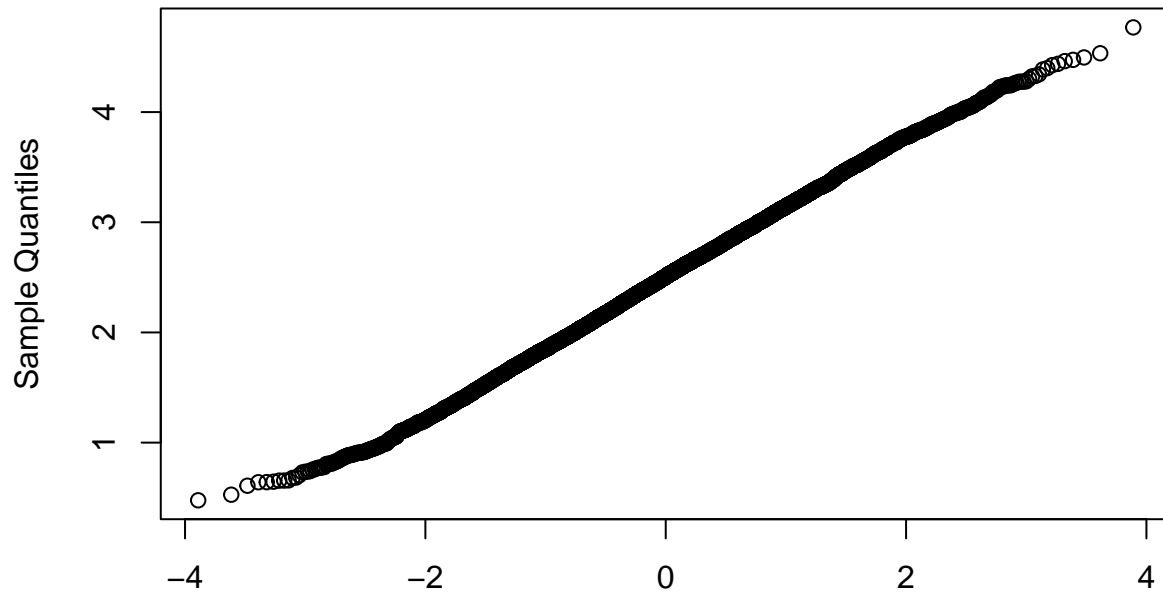
Normal Q-Q Plot



Theoretical Quantiles

```
# Podemos ver que la linea tiene un efecto de borde en las colas, esto es normal y causado por la generacion de datos aleatoria. Lo que buscamos son desviaciones en el medio: si este es el caso los datos no son distribucion normal.  
qqnorm(sum5)
```

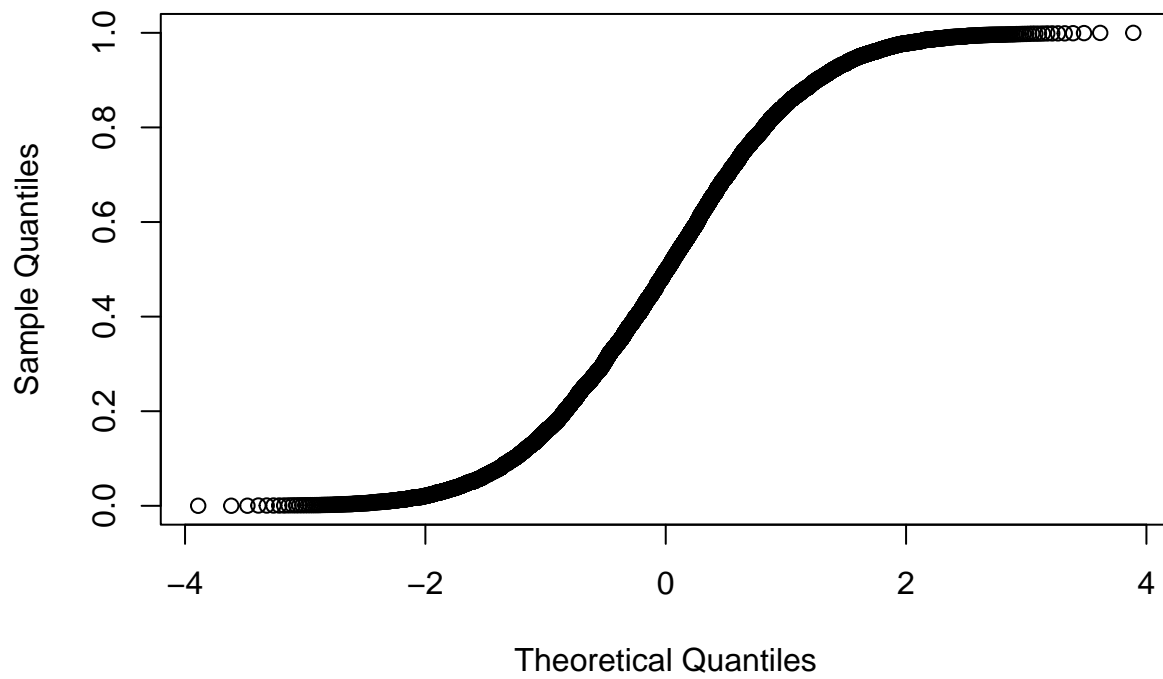
Normal Q-Q Plot



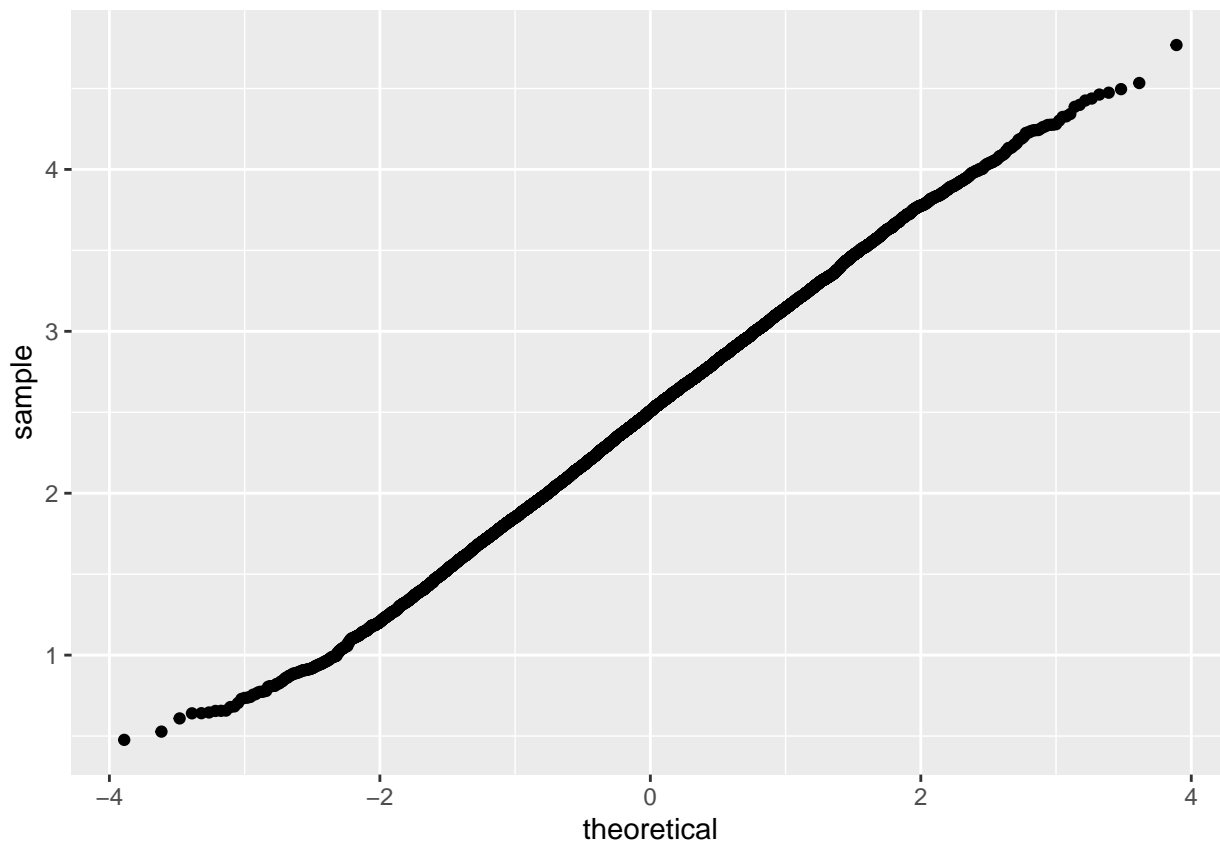
Theoretical Quantiles

```
qqnorm(unifnum)
```

Normal Q-Q Plot

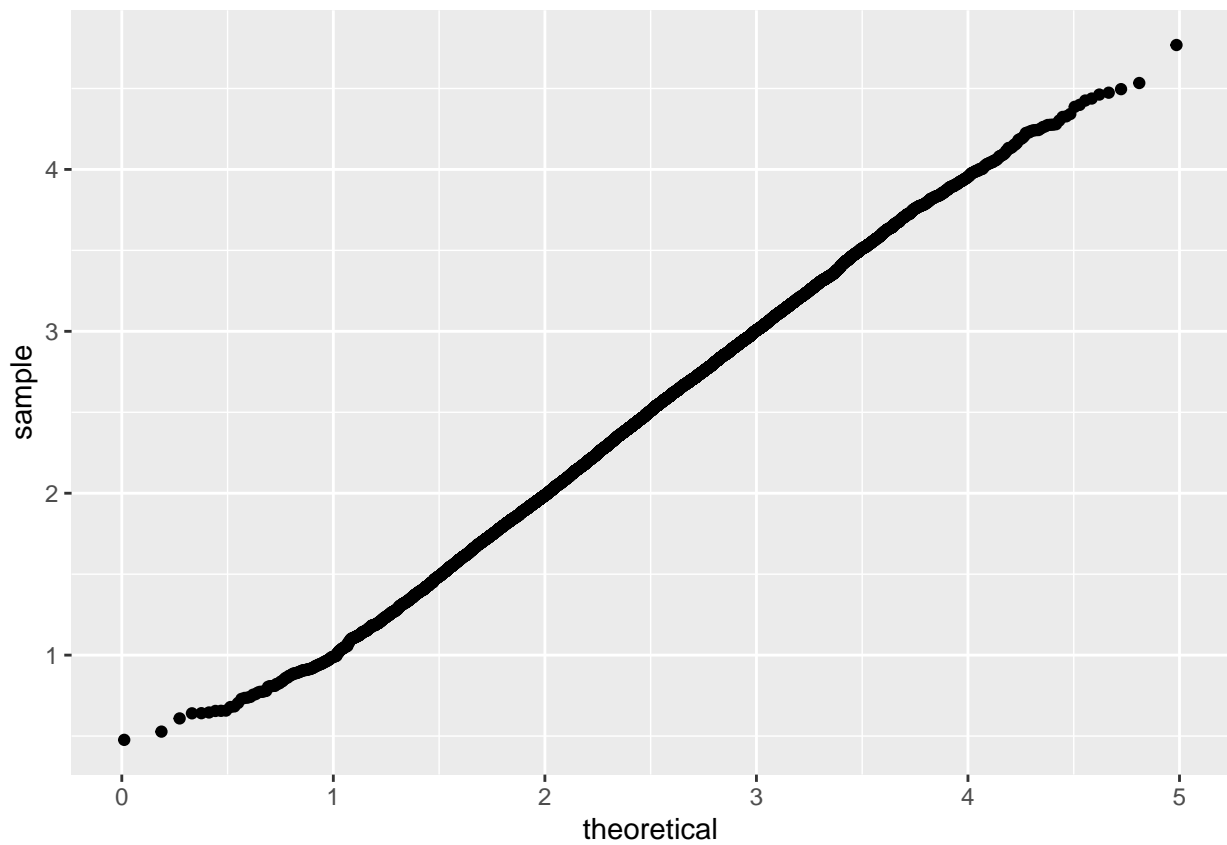


```
ggplot(data.frame(sum5), aes(sample=sum5))+geom_point(stat="qq")
```

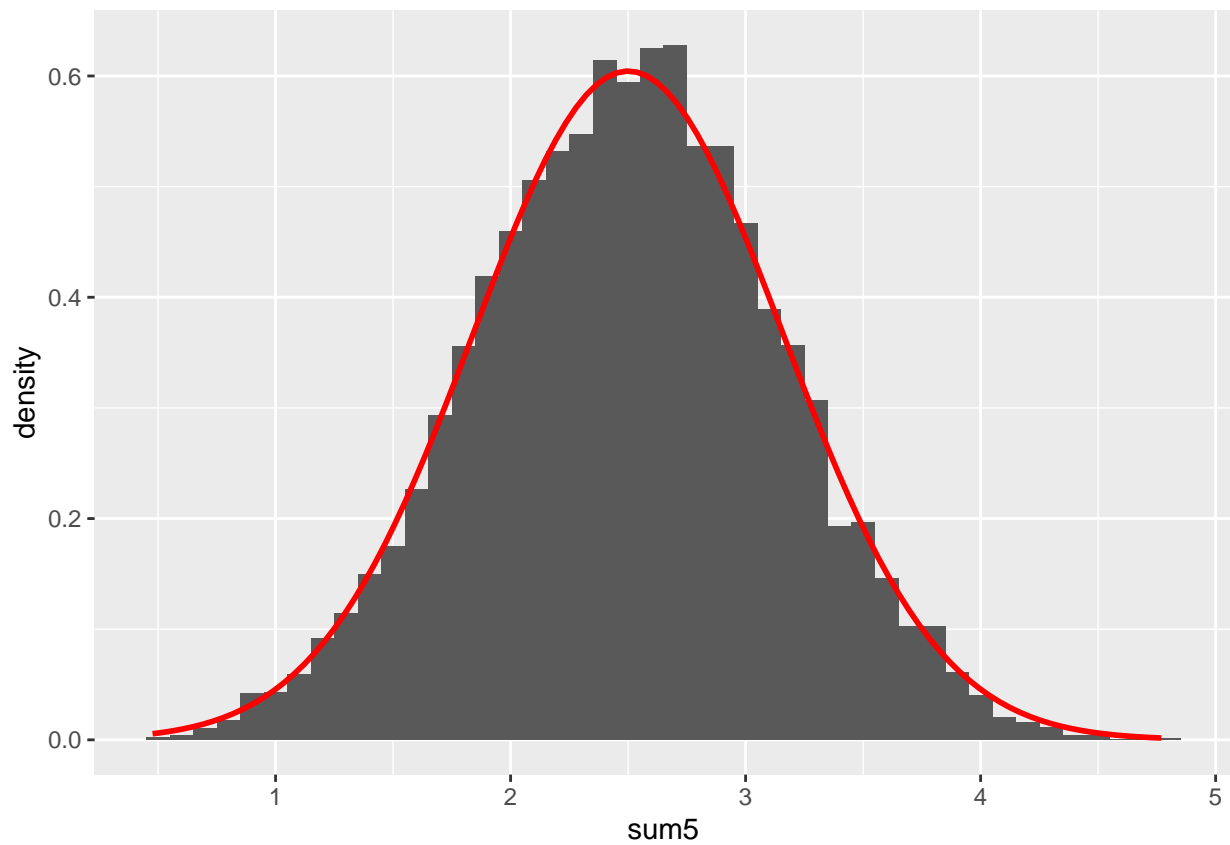


En este caso la distribucion teorica esta centrada en 0 pero la distribucion muestral en 2.5
Se puede corregir de la siguiente manera.

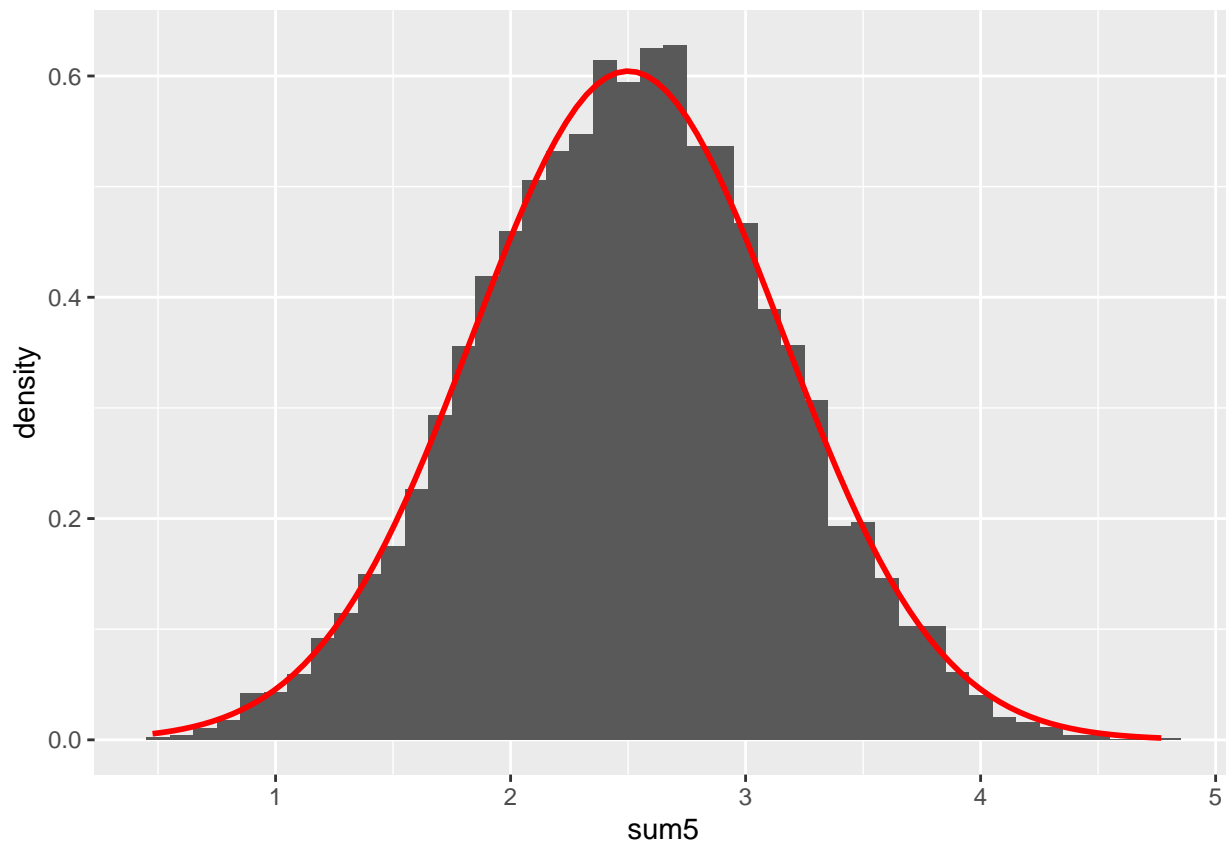
```
ggplot(data.frame(sum5), aes(sample=sum5))+stat_qq(distribution=qnorm, dparams=list(mean(sum5), sd(sum5)))
```

```
# Histograms con la funcion de densidad de probailidad de la normal  
ggplot(data.frame(sum5),aes(x=sum5))+geom_histogram(aes(y=..density..),binwidth=0.1)+stat_function(fun=
```



```
ggplot(data.frame(unifnum),aes(x=sum5))+geom_histogram(aes(y=..density..),binwidth=0.1)+stat_function(f
```



```
# Testeo
```

```
?htest
```

```
## No documentation for 'htest' in specified packages and libraries:
## you could try '??htest'
```

```
?t.test
```

```
# Ejemplo
```

```
?sleep
```

```
head(sleep)
```

```
##   extra group ID
## 1  0.7     1  1
## 2 -1.6     1  2
## 3 -0.2     1  3
## 4 -1.2     1  4
## 5 -0.1     1  5
## 6  3.4     1  6
```

```
attach(sleep)
```

```
## The following objects are masked from sleep (pos = 4):
```

```
##
```

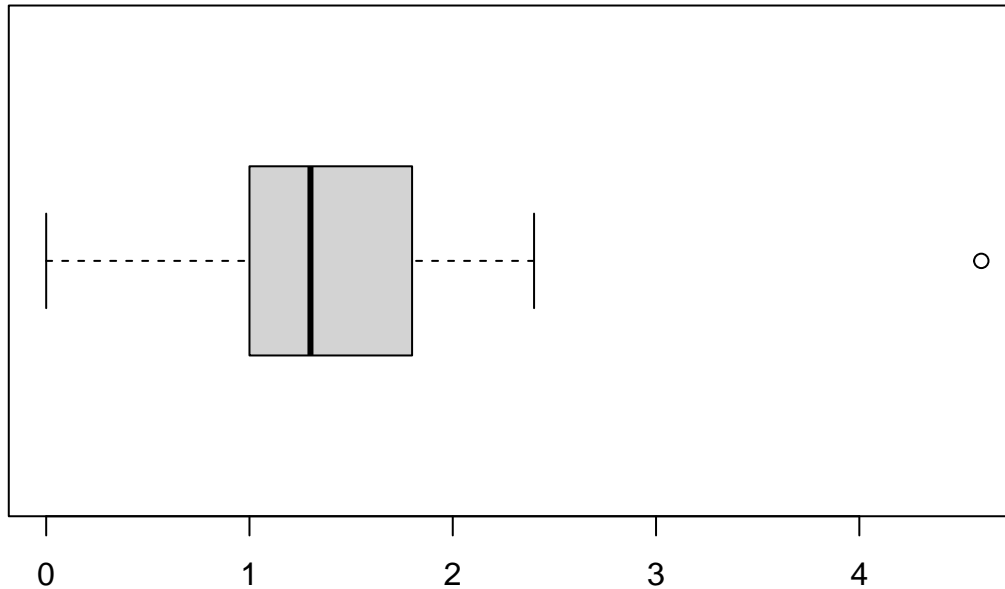
```
##   extra, group, ID
```

```
## The following objects are masked from sleep (pos = 7):
```

```
##
```

```
##   extra, group, ID
```

```
# Crear variables
difsleep=extra[group==2]-extra[group==1]
boxplot(difsleep,horizontal = T)
```



```
result.t=t.test(difsleep)
# Rechazamos la Ho, a NS of 5%, 10%, 1%.
# Problema: numero de observaciones bajo por lo que no podemos
# testear los supuestos del test.
# t-test (normalidad, homoscedasticidad)
class(t.test)
```

```
## [1] "function"
```

```
?t.test # ver argumentos y resultados de la funcion
result.t
```

```
##
## One Sample t-test
##
## data: difsleep
## t = 4.0621, df = 9, p-value = 0.002833
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 0.7001142 2.4598858
## sample estimates:
## mean of x
## 1.58
```

```
class(result.t) # Clase disenada para test de hipotesis.
```

```
## [1] "htest"
```

```
# Podemos ver que el resultados es una lista, entonces podemos acceder los valores que necesitamos usando
result.t$p.value
```

```
## [1] 0.00283289
```

```
str(result.t)
```

```
## List of 10
## $ statistic : Named num 4.06
##   ..- attr(*, "names")= chr "t"
## $ parameter : Named num 9
##   ..- attr(*, "names")= chr "df"
## $ p.value    : num 0.00283
## $ conf.int   : num [1:2] 0.7 2.46
##   ..- attr(*, "conf.level")= num 0.95
## $ estimate   : Named num 1.58
##   ..- attr(*, "names")= chr "mean of x"
## $ null.value : Named num 0
##   ..- attr(*, "names")= chr "mean"
## $ stderr     : num 0.389
## $ alternative: chr "two.sided"
## $ method     : chr "One Sample t-test"
## $ data.name  : chr "difsleep"
## - attr(*, "class")= chr "htest"
```

```
t.test(extra~group,paired=TRUE)
```

```
##
## Paired t-test
##
## data: extra by group
## t = -4.0621, df = 9, p-value = 0.002833
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -2.4598858 -0.7001142
## sample estimates:
## mean difference
## -1.58
```

```
#Obtenemos un estadístico t que es negativo porque se toma la diferencia entre G1 y G2 mientras que R
t.test(extra~group,paired=FALSE) # Aquí no tomamos en cuenta que son los mismos estudiantes y por lo ta
```

```
##
## Welch Two Sample t-test
##
## data: extra by group
## t = -1.8608, df = 17.776, p-value = 0.07939
## alternative hypothesis: true difference in means between group 1 and group 2 is not equal to 0
## 95 percent confidence interval:
## -3.3654832 0.2054832
## sample estimates:
## mean in group 1 mean in group 2
## 0.75 2.33
```

```
# Tener los mismos estudiantes aumenta el poder
# Hay que tener en cuenta el diseño de experimento
```

```
# Testear grupos multiples
```

```
load("~/Dropbox/0.POST-PHD/GOALS/2.CODE/R/Ecomienza/datos/procesados/trafico1.Rdata")
trafico[1:6,]
```

```
##   year month date_of_month day_of_week   autos
## 1 2000     1             1           6 9084.165
## 2 2000     1             2           7 8005.351
## 3 2000     1             3           1 11363.158
## 4 2000     1             4           2 13030.962
## 5 2000     1             5           3 12559.135
## 6 2000     1             6           4 12465.477
```

```
# Agrupar datos por dia
```

```
library(dplyr)
sumsperday<-trafico%>% group_by(day_of_week) %>% summarise(sum=sum(autos)) %>% arrange()
sumsperday
```

```
## # A tibble: 7 x 2
##   day_of_week      sum
##       <int>    <dbl>
## 1         1 9316003.
## 2         2 10274895.
## 3         3 10109101.
## 4         4 10045369.
## 5         5 9850212.
## 6         6 6704468.
## 7         7 5886926.
```

```
# Testear si hay la misma probailidad de nacimiento en cada dia.
```

```
# Supuestos: independencia entre dias y nacimientos
```

```
res=chisq.test(sumsperday[,2])
```

```
str(res)
```

```
## List of 9
```

```
## $ statistic: Named num 2210442
```

```
## ..- attr(*, "names")= chr "X-squared"
```

```
## $ parameter: Named num 6
```

```
## ..- attr(*, "names")= chr "df"
```

```
## $ p.value : num 0
```

```
## $ method : chr "Chi-squared test for given probabilities"
```

```
## $ data.name: chr "sumsperday[, 2]"
```

```
## $ observed : num [1:7] 9316003 10274895 10109101 10045369 9850212 ...
```

```
## $ expected : num [1:7] 8883853 8883853 8883853 8883853 8883853 ...
```

```
## $ residuals: num [1:7] 145 467 411 390 324 ...
```

```
## $ stdres : num [1:7] 157 504 444 421 350 ...
```

```
## - attr(*, "class")= chr "htest"
```

```
# Regresion
```

```
#install.packages("HistData")
```

```
library(HistData)
```

```
?Galton
```

```
head(Galton)
```

```
##   parent child
## 1   70.5  61.7
## 2   68.5  61.7
## 3   65.5  61.7
## 4   64.5  61.7
## 5   64.0  61.7
## 6   67.5  62.2
```

```
attach(Galton)
```

```
## The following objects are masked from Galton (pos = 4):
```

```
##
```

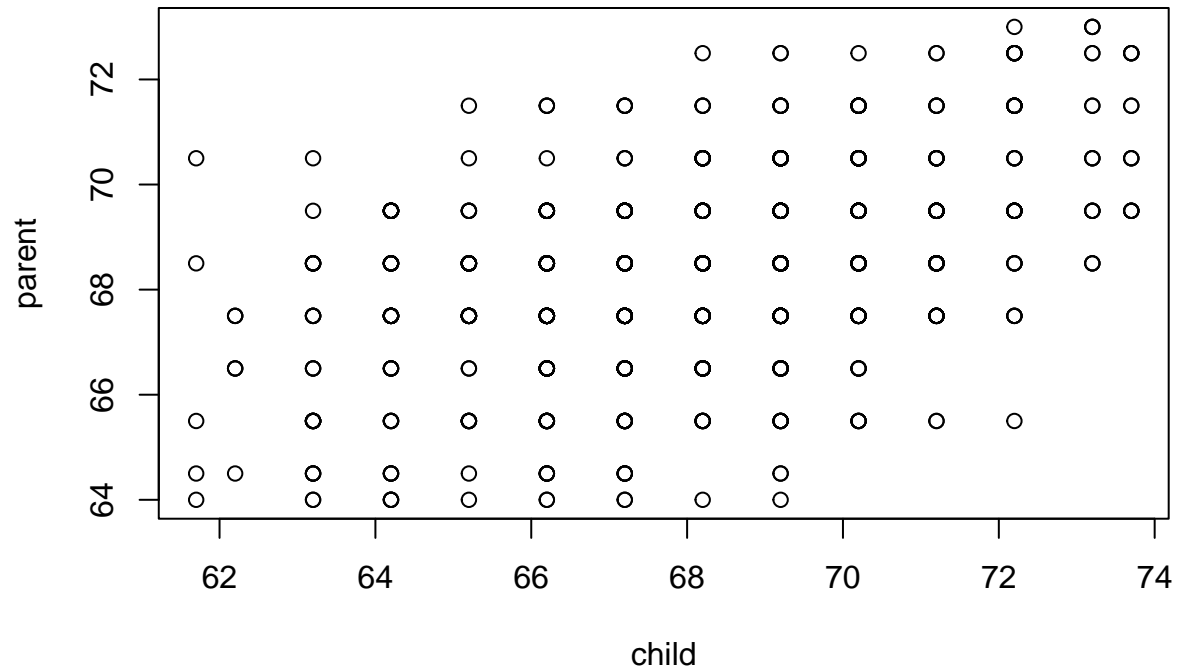
```
##   child, parent
```

```
## The following objects are masked from Galton (pos = 6):
```

```
##
```

```
##   child, parent
```

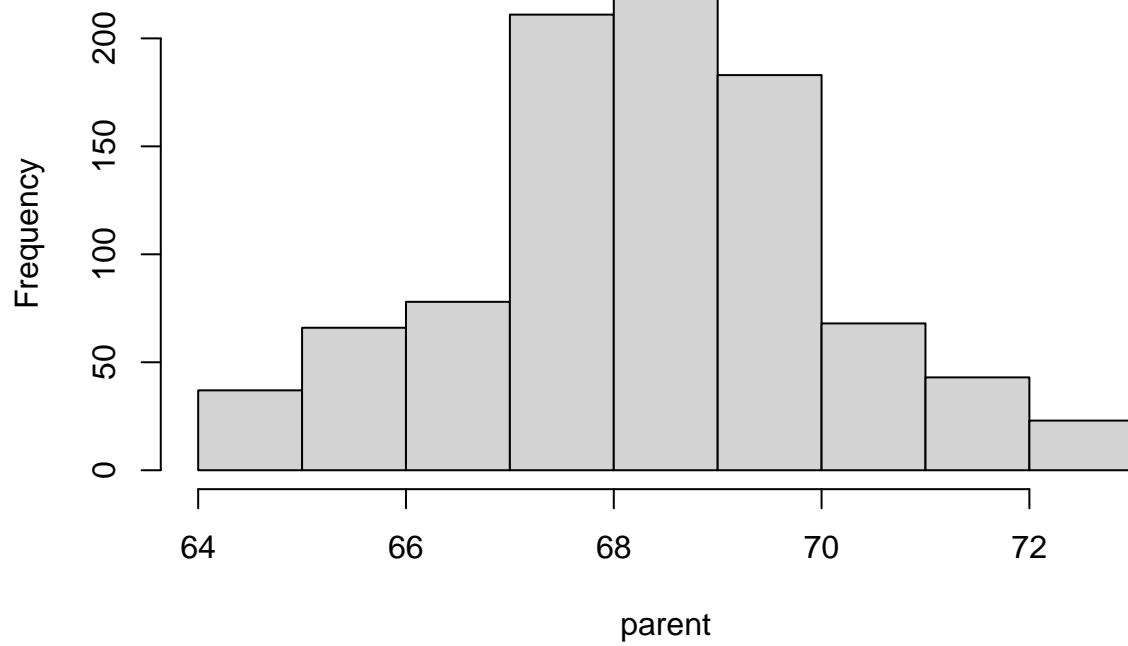
```
plot(child,parent)
```



Los datos no se han obtenido de manera continua, lo que conduce a una pérdida de precisión.

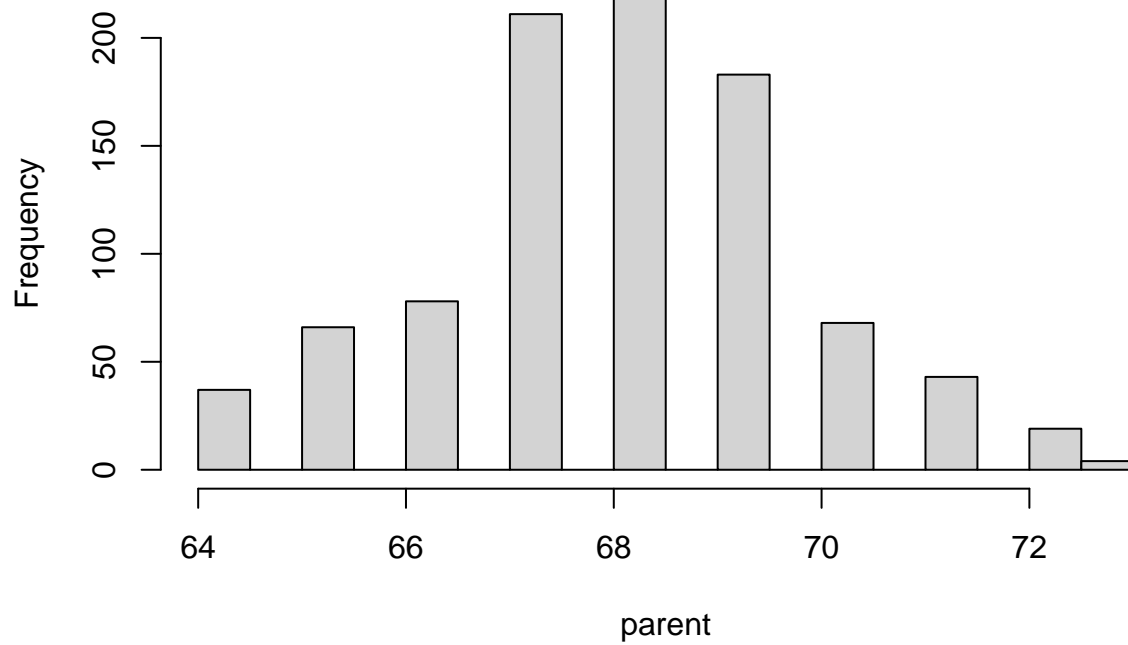
```
hist(parent)
```

Histogram of parent



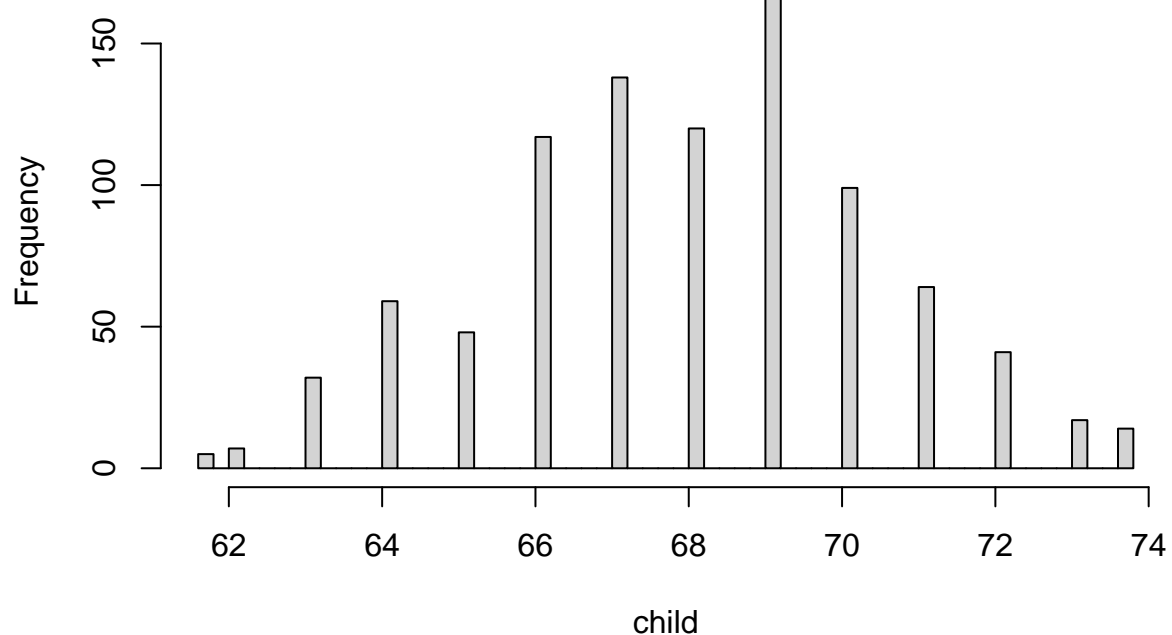
Al aumentar la anchura de clase, se ve el problema
`hist(parent,20)`

Histogram of parent

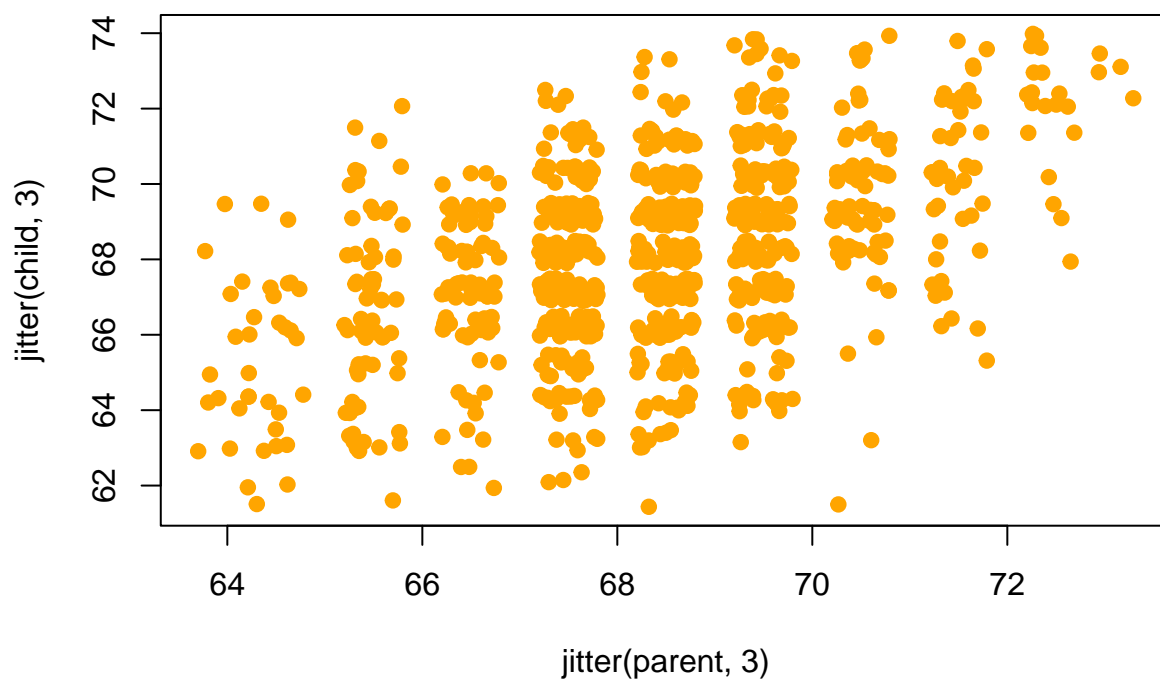


`hist(child,60)`

Histogram of child



```
# Vamos a corregir utilizando la funcion jitter
plot(jitter(parent,3),jitter(child,3),pch=19,col="orange")
```



```
# Ahora observamos elipsoides que se deben a la distribucion normal en ambas direcciones.
```

```
cor(parent,child)
```

```
## [1] 0.4587624
```

```
# Correlacion importante
# Hay un componente genetico
linearreg<-lm(child~parent)
summary(linearreg)
```

```
##
## Call:
## lm(formula = child ~ parent)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8050 -1.3661  0.0487  1.6339  5.9264
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 23.94153    2.81088   8.517  <2e-16 ***
## parent       0.64629    0.04114  15.711  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.239 on 926 degrees of freedom
## Multiple R-squared:  0.2105, Adjusted R-squared:  0.2096
## F-statistic: 246.8 on 1 and 926 DF,  p-value: < 2.2e-16
```

```
class(linearreg)
```

```
## [1] "lm"
```

```
?lm
```

```
names(linearreg)
```

```
## [1] "coefficients" "residuals"      "effects"        "rank"           "fitted.values" "assign"
## [9] "xlevels"      "call"          "terms"         "model"
```

```
linearreg$residuals
```

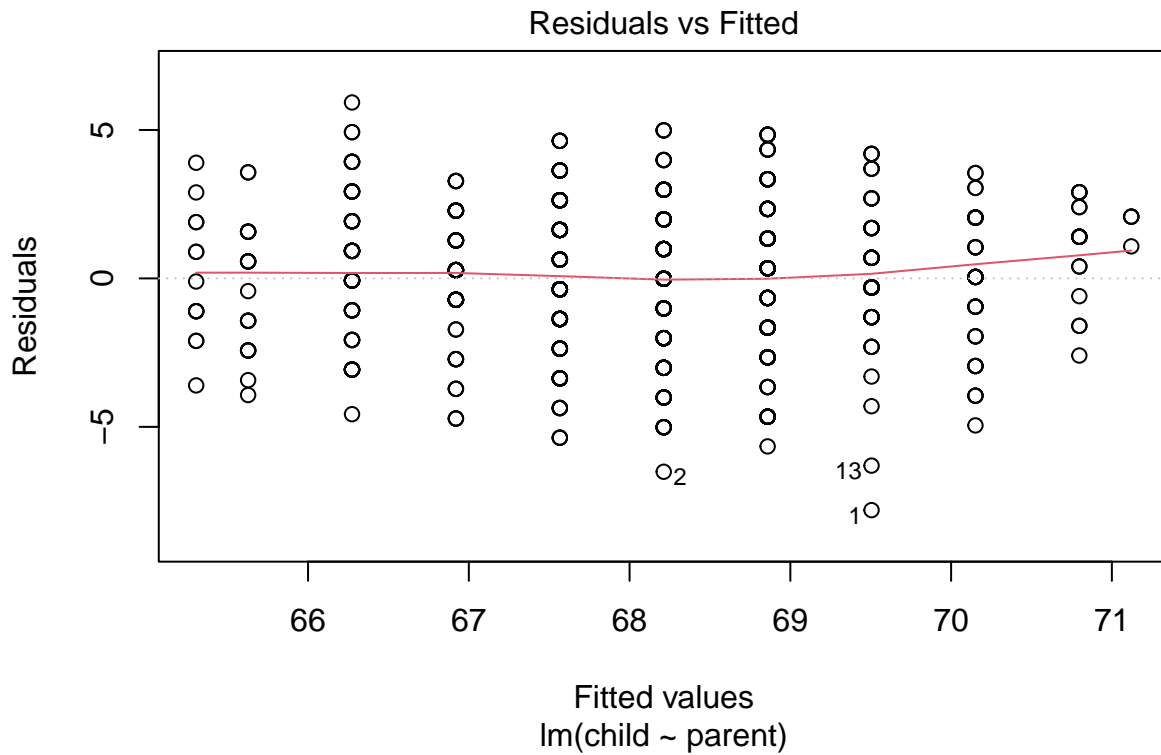
```
##      1      2      3      4      5      6      7      8
## -7.80501621 -6.51243505 -4.57356330 -3.92727272 -3.60412743 -5.36614446 -5.36614446 -5.36614446 -4.7
##      13     14     15     16     17     18     19     20
## -6.30501621 -5.65872563 -5.01243505 -5.01243505 -5.01243505 -5.01243505 -5.01243505 -5.01243505 -5.0
##      25     26     27     28     29     30     31     32
## -4.36614446 -4.36614446 -3.71985388 -3.71985388 -3.71985388 -3.07356330 -3.07356330 -3.07356330 -3.0
##      37     38     39     40     41     42     43     44
## -3.07356330 -3.07356330 -2.42727272 -2.42727272 -2.42727272 -2.42727272 -2.10412743 -2.10412743 -4.6
##      49     50     51     52     53     54     55     56
## -4.65872563 -4.65872563 -4.65872563 -4.65872563 -4.65872563 -4.65872563 -4.65872563 -4.65872563 -4.6
##      61     62     63     64     65     66     67     68
## -4.01243505 -4.01243505 -4.01243505 -4.01243505 -4.01243505 -4.01243505 -4.01243505 -4.01243505 -4.0
##      73     74     75     76     77     78     79     80
## -3.36614446 -3.36614446 -3.36614446 -3.36614446 -3.36614446 -3.36614446 -3.36614446 -3.36614446 -3.3
##      85     86     87     88     89     90     91     92
## -3.36614446 -2.71985388 -2.71985388 -2.71985388 -2.71985388 -2.71985388 -2.07356330 -2.07356330 -2.0
##      97     98     99     100    101    102    103    104
## -1.42727272 -1.42727272 -1.42727272 -1.10412743 -1.10412743 -1.10412743 -1.10412743 -4.95130679 -4.3
##      109    110    111    112    113    114    115    116
## -3.65872563 -3.01243505 -3.01243505 -3.01243505 -3.01243505 -3.01243505 -3.01243505 -3.01243505 -3.0
```

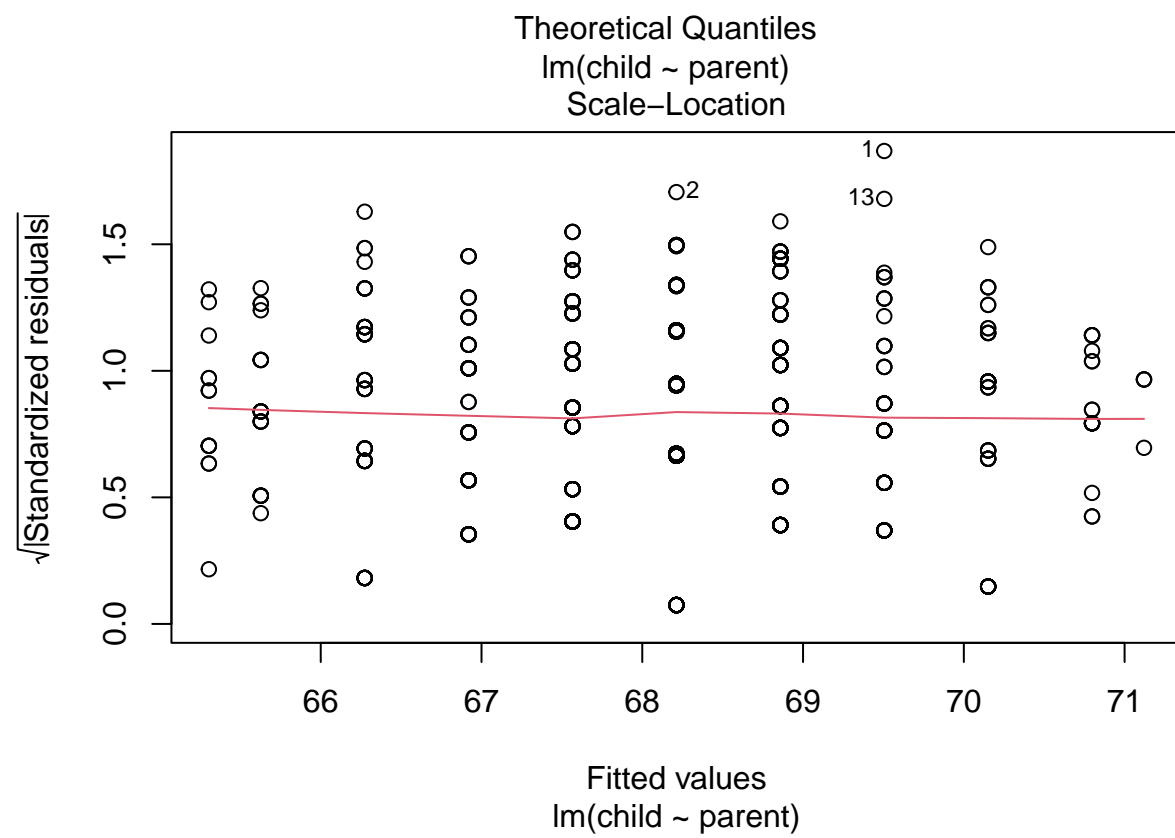
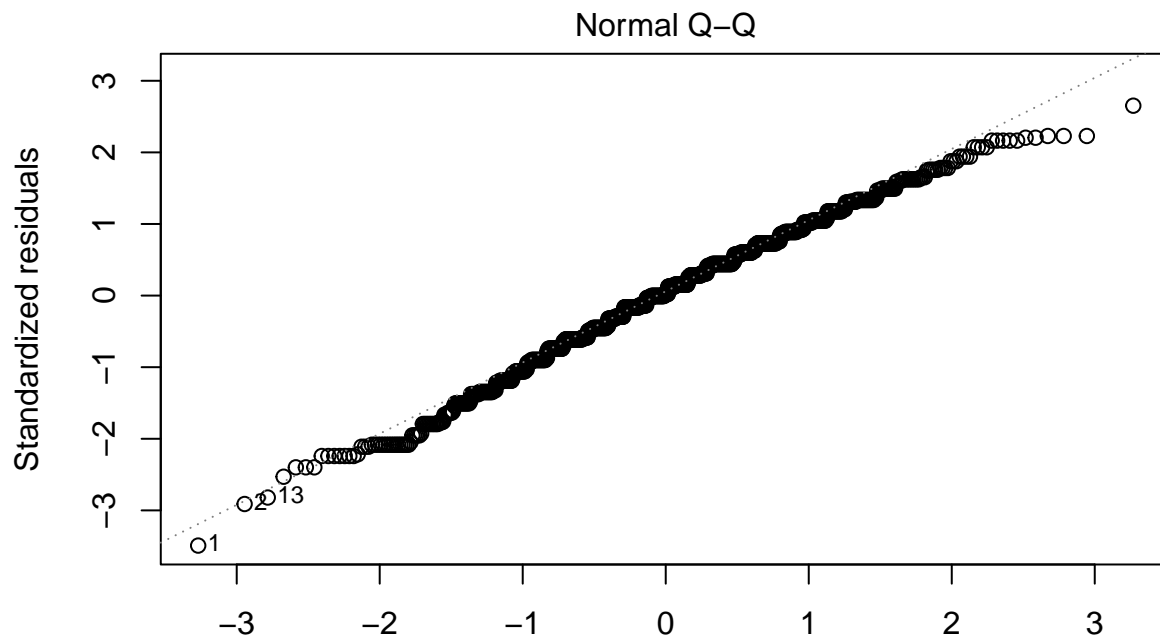
##	121	122	123	124	125	126	127	128
##	-3.01243505	-3.01243505	-3.01243505	-3.01243505	-3.01243505	-2.36614446	-2.36614446	-2.36614446
##	133	134	135	136	137	138	139	140
##	-2.36614446	-2.36614446	-2.36614446	-2.36614446	-2.36614446	-2.36614446	-2.36614446	-2.36614446
##	145	146	147	148	149	150	151	152
##	-1.07356330	-1.07356330	-1.07356330	-1.07356330	-1.07356330	-0.42727272	-0.10412743	-3.95130679
##	157	158	159	160	161	162	163	164
##	-2.65872563	-2.65872563	-2.65872563	-2.65872563	-2.65872563	-2.65872563	-2.65872563	-2.65872563
##	169	170	171	172	173	174	175	176
##	-2.65872563	-2.65872563	-2.65872563	-2.65872563	-2.01243505	-2.01243505	-2.01243505	-2.01243505
##	181	182	183	184	185	186	187	188
##	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-2.01243505
##	193	194	195	196	197	198	199	200
##	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-2.01243505	-1.36614446	-1.36614446	-1.36614446
##	205	206	207	208	209	210	211	212
##	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446
##	217	218	219	220	221	222	223	224
##	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446
##	229	230	231	232	233	234	235	236
##	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-1.36614446	-0.71985388	-0.71985388	-0.71985388
##	241	242	243	244	245	246	247	248
##	-0.71985388	-0.71985388	-0.71985388	-0.71985388	-0.71985388	-0.71985388	-0.71985388	-0.71985388
##	253	254	255	256	257	258	259	260
##	-0.07356330	-0.07356330	-0.07356330	-0.07356330	-0.07356330	-0.07356330	-0.07356330	-0.07356330
##	265	266	267	268	269	270	271	272
##	0.57272728	0.57272728	0.89587257	0.89587257	-2.95130679	-2.95130679	-2.95130679	-2.95130679
##	277	278	279	280	281	282	283	284
##	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563
##	289	290	291	292	293	294	295	296
##	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563	-1.65872563
##	301	302	303	304	305	306	307	308
##	-1.65872563	-1.65872563	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505
##	313	314	315	316	317	318	319	320
##	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505
##	325	326	327	328	329	330	331	332
##	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505	-1.01243505
##	337	338	339	340	341	342	343	344
##	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446
##	349	350	351	352	353	354	355	356
##	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446
##	361	362	363	364	365	366	367	368
##	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446	-0.36614446
##	373	374	375	376	377	378	379	380
##	0.28014612	0.28014612	0.28014612	0.28014612	0.28014612	0.28014612	0.28014612	0.28014612
##	385	386	387	388	389	390	391	392
##	0.28014612	0.28014612	0.28014612	0.28014612	0.92643670	0.92643670	0.92643670	0.92643670
##	397	398	399	400	401	402	403	404
##	0.92643670	0.92643670	0.92643670	1.57272728	1.57272728	1.57272728	1.57272728	1.57272728
##	409	410	411	412	413	414	415	416
##	-1.95130679	-1.95130679	-1.30501621	-1.30501621	-1.30501621	-1.30501621	-1.30501621	-1.30501621
##	421	422	423	424	425	426	427	428
##	-1.30501621	-1.30501621	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563
##	433	434	435	436	437	438	439	440
##	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563	-0.65872563

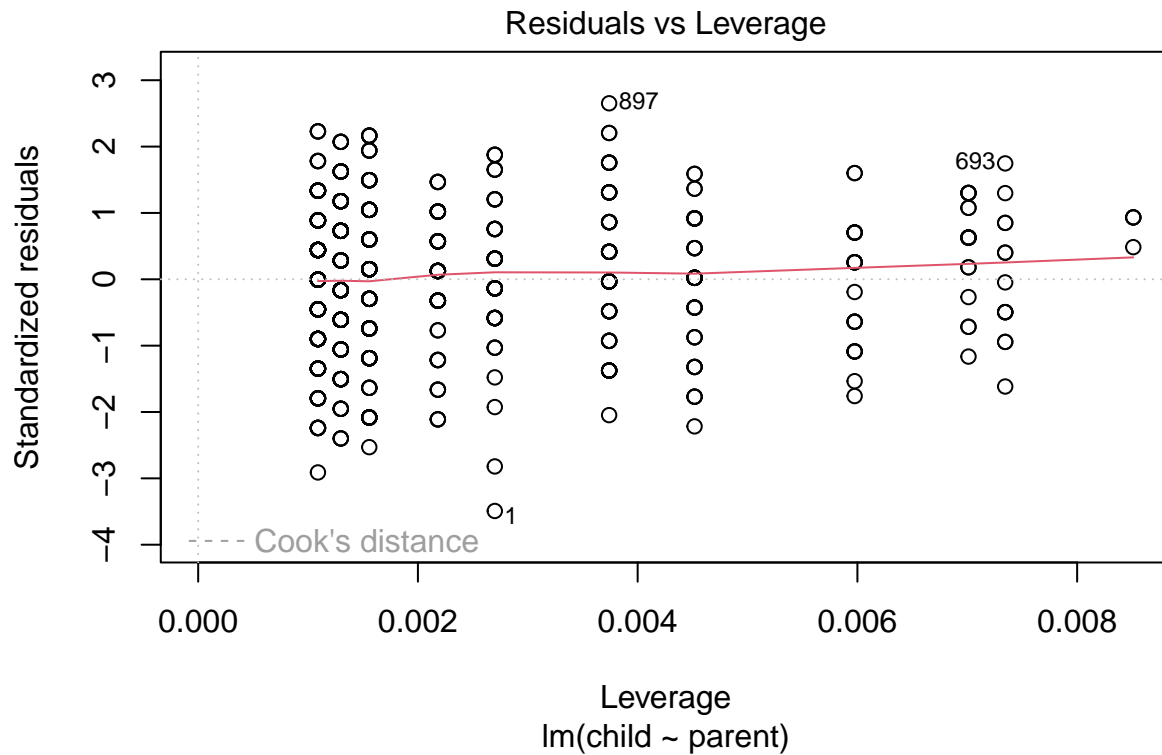
##	445	446	447	448	449	450	451	452	
##	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505
##	457	458	459	460	461	462	463	464	
##	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505
##	469	470	471	472	473	474	475	476	
##	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	-0.01243505	0.63385554
##	481	482	483	484	485	486	487	488	
##	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554
##	493	494	495	496	497	498	499	500	
##	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554	0.63385554
##	505	506	507	508	509	510	511	512	
##	1.28014612	1.28014612	1.28014612	1.28014612	1.28014612	1.28014612	1.28014612	1.28014612	1.28014612
##	517	518	519	520	521	522	523	524	
##	1.28014612	1.28014612	1.92643670	1.92643670	1.92643670	1.92643670	1.92643670	1.92643670	1.92643670
##	529	530	531	532	533	534	535	536	
##	-0.95130679	-0.95130679	-0.95130679	-0.95130679	-0.95130679	-0.30501621	-0.30501621	-0.30501621	-0.30501621
##	541	542	543	544	545	546	547	548	
##	-0.30501621	-0.30501621	-0.30501621	-0.30501621	-0.30501621	-0.30501621	-0.30501621	-0.30501621	-0.30501621
##	553	554	555	556	557	558	559	560	
##	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437
##	565	566	567	568	569	570	571	572	
##	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437
##	577	578	579	580	581	582	583	584	
##	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.34127437	0.98756495
##	589	590	591	592	593	594	595	596	
##	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495
##	601	602	603	604	605	606	607	608	
##	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495
##	613	614	615	616	617	618	619	620	
##	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495
##	625	626	627	628	629	630	631	632	
##	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	0.98756495	1.63385554
##	637	638	639	640	641	642	643	644	
##	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554
##	649	650	651	652	653	654	655	656	
##	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554
##	661	662	663	664	665	666	667	668	
##	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554	1.63385554
##	673	674	675	676	677	678	679	680	
##	2.28014612	2.28014612	2.28014612	2.28014612	2.28014612	2.28014612	2.28014612	2.28014612	2.28014612
##	685	686	687	688	689	690	691	692	
##	2.92643670	2.92643670	2.92643670	2.92643670	2.92643670	2.92643670	3.57272728	3.57272728	3.85416667
##	697	698	699	700	701	702	703	704	
##	0.04869321	0.04869321	0.04869321	0.04869321	0.04869321	0.04869321	0.04869321	0.04869321	0.63385554
##	709	710	711	712	713	714	715	716	
##	0.69498379	0.69498379	0.69498379	0.69498379	0.69498379	0.69498379	0.69498379	0.69498379	0.69498379
##	721	722	723	724	725	726	727	728	
##	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437
##	733	734	735	736	737	738	739	740	
##	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437	1.34127437
##	745	746	747	748	749	750	751	752	
##	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495
##	757	758	759	760	761	762	763	764	
##	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	1.98756495	2.63157895

##	769	770	771	772	773	774	775	776
##	2.63385554	2.63385554	2.63385554	2.63385554	2.63385554	2.63385554	2.63385554	2.63385554
##	781	782	783	784	785	786	787	788
##	2.63385554	2.63385554	2.63385554	3.28014612	3.28014612	3.28014612	3.28014612	3.92643670
##	793	794	795	796	797	798	799	800
##	0.40240263	0.40240263	1.04869321	1.04869321	1.04869321	1.04869321	1.69498379	1.69498379
##	805	806	807	808	809	810	811	812
##	1.69498379	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437
##	817	818	819	820	821	822	823	824
##	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437	2.34127437
##	829	830	831	832	833	834	835	836
##	2.98756495	2.98756495	2.98756495	2.98756495	2.98756495	2.98756495	2.98756495	2.98756495
##	841	842	843	844	845	846	847	848
##	2.98756495	2.98756495	2.98756495	3.63385554	3.63385554	3.63385554	3.63385554	3.63385554
##	853	854	855	856	857	858	859	860
##	3.63385554	3.63385554	4.92643670	4.92643670	1.07925733	1.40240263	1.40240263	1.40240263
##	865	866	867	868	869	870	871	872
##	2.04869321	2.04869321	2.04869321	2.04869321	2.04869321	2.04869321	2.04869321	2.04869321
##	877	878	879	880	881	882	883	884
##	2.69498379	3.34127437	3.34127437	3.34127437	3.34127437	3.34127437	3.34127437	3.34127437
##	889	890	891	892	893	894	895	896
##	3.98756495	3.98756495	3.98756495	3.98756495	4.63385554	4.63385554	4.63385554	4.63385554
##	901	902	903	904	905	906	907	908
##	2.40240263	2.40240263	3.04869321	3.04869321	3.69498379	3.69498379	3.69498379	4.34127437
##	913	914	915	916	917	918	919	920
##	4.98756495	4.98756495	2.90240263	2.90240263	2.90240263	2.90240263	3.54869321	3.54869321
##	925	926	927	928				
##	4.84127437	4.84127437	4.84127437	4.84127437				

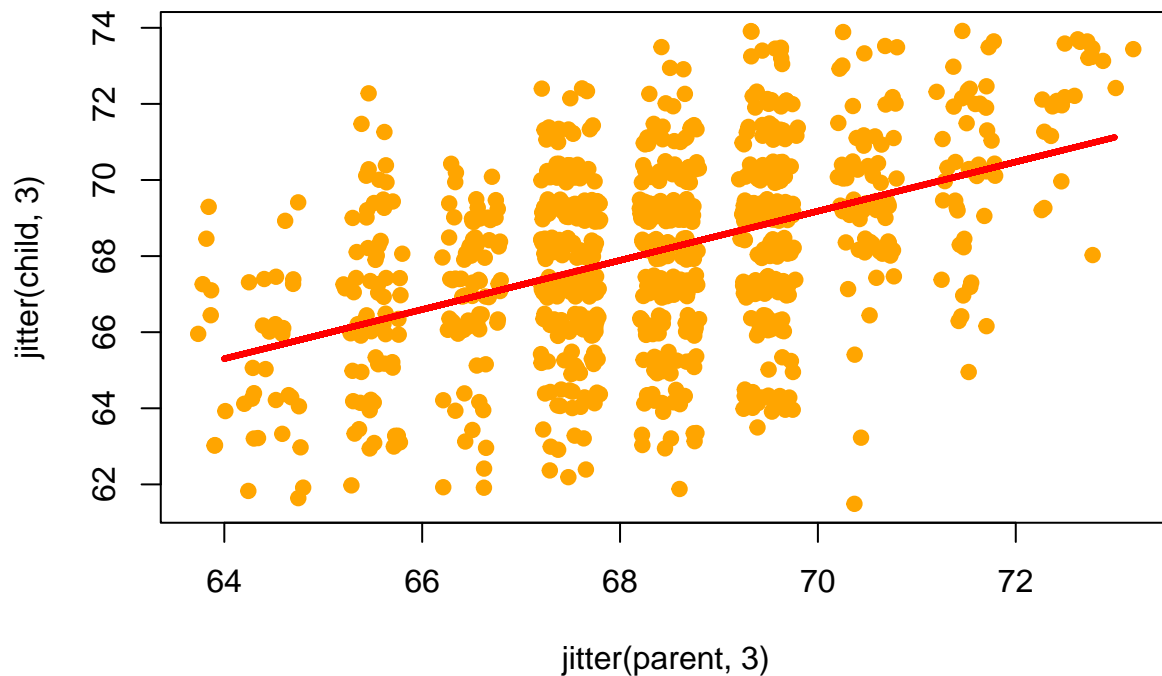
```
plot(linearreg)
```







```
plot(jitter(parent,3),jitter(child,3),pch=19,col="orange")
#lines(parent,linearreg$fitted.values,col="red",lwd=3)
lines(parent,linearreg$fitted,col="red",lwd=3)
```



```
# No quitar el comentario de la linea inferior. Solamente copiar en la consola para que ejecute
#rmarkdown::render("15FuncionesEstadisticasBasicas_cheatsheet.R",c("pdf_document","html_document"))
```