

# Practica 1\_nba

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```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.2.1 --
```

```
## v ggplot2 3.2.1    v purrr  0.3.2
## v tibble  2.1.3    v dplyr  0.8.3
## v tidyr   1.0.0    v stringr 1.4.0
## v readr   1.3.1    v forcats 0.4.0
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(readr)
library(dplyr)
library(TeachingDemos)
library(nortest)
library(car)
```

```
## Loading required package: carData
```

```
##
```

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

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

```
##
```

```
##      recode
```

```
## The following object is masked from 'package:purrr':
```

```
##
```

```
##      some
```

```
library(MASS)
```

```
##
```

```
## Attaching package: 'MASS'
```

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

```
##
```

```
##      select
```

```
library(leaps)
library(gvlma)
library(ISLR)
library(tinytex)
```

#El objetivo del ejercicio es crear un modelo para estimar el salario de los jugadores de #la nba en funcion de las estadísticas y habilidades de cada jugador.

#En primer lugar importo el dataset correspondiente a los datos de “nba”.

```
nba<-read.csv("C:/Users/alvar/Desktop/CUNEF/PREDICCIÓN/nba.csv")
```

#El siguiente paso es suprimir los NAs pertenecientes al objeto nba.

```
nba <- unique(nba)
```

```
nba <- na.omit(nba)
```

#Cambio el nombre de las variables.

```
nba<-rename(nba,minutes.played = "MP")
nba<-rename(nba, team= "Tm")
nba<-rename(nba, games= "G")
nba<-rename(nba, efficiency= "PER")
nba<-rename(nba, success= "TS.")
nba<-rename(nba, triple.try= "X3PAr")
nba<-rename(nba, free.try= "FTr")
nba<-rename(nba, rebound.attack= "ORB.")
nba<-rename(nba, rebound.deffence= "DRB.")
nba<-rename(nba, total.rebounds= "TRB.")
nba<-rename(nba, assistance= "AST.")
nba<-rename(nba, stealing= "STL.")
nba<-rename(nba, blocking= "BLK.")
nba<-rename(nba, turnover.percentage= "TOV.")
nba<-rename(nba, usage.percentage= "USG.")
nba<-rename(nba, good.attack= "OWS")
nba<-rename(nba,good.defence= "DWS")
nba<-rename(nba, total.good= "WS")
nba<-rename(nba, win.shares.per.48= "WS.48")
nba<-rename(nba, puntos.ofensivosVSmedia= "OBPM")
nba<-rename(nba, puntos.defensivosVSmedia= "DBPM")
nba<-rename(nba, puntosVSmedia= "BPM")
nba<-rename(nba, contribution= "VORP")
```

#Creamos el objeto regres01 para estudiar el modelo de regresión y ver la relacion #que hay entre la variable dependiente “Salary” y las variables explicativas de la #base de datos.

#No se incluyen las variables equipo, nombre del jugador y pais.

```
vY=nba$Salary
View(vY)

mX=cbind(1,nba[,2:28])
```

View(mX)

head(vY)

```
## [1] 815615 3477600 12307692 3202217 3057240 1312611
```

head.matrix(mX)

```
## 1 Salary NBA_Country NBA_DraftNumber Age team games minutes.played
## 1 1 815615 China 43 22 HOU 16 87
## 2 1 3477600 Georgia 42 33 GSW 66 937
## 3 1 12307692 USA 19 36 SAC 59 1508
## 4 1 3202217 USA 13 22 CHI 24 656
## 5 1 3057240 USA 10 20 POR 62 979
## 6 1 1312611 USA 62 24 DAL 79 2238
## efficiency success triple.try free.try rebound.attack rebound.deffence
## 1 0.6 0.303 0.593 0.370 6.5 16.8
## 2 16.8 0.608 0.004 0.337 11.0 25.0
## 3 17.3 0.529 0.193 0.140 7.0 23.8
## 4 14.6 0.499 0.346 0.301 1.4 14.4
## 5 8.2 0.487 0.387 0.146 4.9 18.3
## 6 11.5 0.543 0.489 0.141 1.3 11.3
## total.rebounds assistance stealing blocking turnover.percentage
## 1 11.7 1.5 1.1 6.8 18.2
## 2 18.5 15.4 1.9 1.3 19.3
## 3 15.0 14.9 1.4 0.6 12.5
## 4 7.7 18.6 1.8 0.5 9.7
## 5 11.7 7.3 0.8 2.5 15.6
## 6 6.1 13.3 1.4 0.3 9.1
## usage.percentage good.attack good.defence total.good win.shares.per.48
## 1 19.5 -0.4 0.1 -0.2 -0.121
## 2 17.2 1.7 1.4 3.1 0.160
## 3 27.6 0.3 1.1 1.4 0.046
## 4 29.5 -0.1 0.5 0.4 0.027
## 5 15.5 -0.4 1.2 0.8 0.038
## 6 17.0 1.6 1.6 3.1 0.067
## puntos.ofensivosVSmedia puntos.defensivosVSmedia puntosVSmedia
## 1 -10.6 0.5 -10.1
## 2 -0.6 1.3 0.8
## 3 -0.6 -1.3 -1.9
## 4 -0.7 -2.0 -2.6
## 5 -3.7 0.9 -2.9
## 6 -0.4 -0.5 -0.9
## contribution
## 1 -0.2
## 2 0.7
## 3 0.0
## 4 -0.1
## 5 -0.2
## 6 0.6
```

```
regres01<-lm(Salary~NBA_DraftNumber + Age + games + minutes.played + efficiency + success
+ triple.try + free.try + rebound.attack + rebound.deffence + total.rebounds + assistance
+ turnover.percentage + usage.percentage + good.attack + good.defence +total.good + win.sh
+ puntos.ofensivosVSmedia + puntos.defensivosVSmedia + puntosVSmedia + contribution, data=)

summary(regres01)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + games + minutes.played +
##     efficiency + success + triple.try + free.try + rebound.attack +
##     rebound.deffence + total.rebounds + assistance + stealing +
##     blocking + turnover.percentage + usage.percentage + good.attack +
##     good.defence + total.good + win.shares.per.48 + puntos.ofensivosVSmedia +
##     puntos.defensivosVSmedia + puntosVSmedia + contribution,
##     data = nba)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-15344434	-2975002	-390994	2102049	21682686

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-2250339	4899140	-0.459	0.646
NBA_DraftNumber	-60481	12755	-4.742	2.83e-06 ***
Age	516821	56511	9.146	< 2e-16 ***
games	-154411	24983	-6.181	1.41e-09 ***
minutes.played	5657	1083	5.224	2.67e-07 ***
efficiency	-355059	280566	-1.266	0.206
success	-2162767	5149052	-0.420	0.675
triple.try	-3458209	2357905	-1.467	0.143
free.try	-158470	886960	-0.179	0.858
rebound.attack	-1055234	902929	-1.169	0.243
rebound.deffence	-855005	893502	-0.957	0.339
total.rebounds	2006676	1788697	1.122	0.263
assistance	-19606	47542	-0.412	0.680
stealing	-196551	421995	-0.466	0.642
blocking	110238	318223	0.346	0.729
turnover.percentage	4208	52692	0.080	0.936
usage.percentage	169431	104692	1.618	0.106
good.attack	-1271685	4493110	-0.283	0.777
good.defence	-1735775	4518854	-0.384	0.701
total.good	1827796	4498896	0.406	0.685
win.shares.per.48	1914677	11797052	0.162	0.871
puntos.ofensivosVSmedia	1878971	4744860	0.396	0.692
puntos.defensivosVSmedia	1438902	4660878	0.309	0.758
puntosVSmedia	-1295954	4678340	-0.277	0.782
contribution	629465	633799	0.993	0.321

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5103000 on 458 degrees of freedom
## Multiple R-squared:  0.5479, Adjusted R-squared:  0.5242
```

```
## F-statistic: 23.13 on 24 and 458 DF, p-value: < 2.2e-16
```

#Utilizo el Modelo Backward Stepwise y de manera iterativa se van eliminando una a una #las variables menos útiles.

```
stepAIC(regres01, direction="backward")
```

```
## Start: AIC=14944.57
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + free.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + turnover.percentage +
## usage.percentage + good.attack + good.defence + total.good +
## win.shares.per.48 + puntos.ofensivosVSmedia + puntos.defensivosVSmedia +
## puntosVSmedia + contribution
##
##           Df Sum of Sq      RSS   AIC
## - turnover.percentage    1 1.6611e+11 1.1928e+16 14943
## - win.shares.per.48       1 6.8601e+11 1.1928e+16 14943
## - free.try                1 8.3134e+11 1.1928e+16 14943
## - puntosVSmedia           1 1.9984e+12 1.1930e+16 14943
## - good.attack             1 2.0862e+12 1.1930e+16 14943
## - puntos.defensivosVSmedia 1 2.4821e+12 1.1930e+16 14943
## - blocking                1 3.1253e+12 1.1931e+16 14943
## - good.defence            1 3.8425e+12 1.1931e+16 14943
## - puntos.ofensivosVSmedia 1 4.0840e+12 1.1932e+16 14943
## - total.good              1 4.2986e+12 1.1932e+16 14943
## - assistance              1 4.4292e+12 1.1932e+16 14943
## - success                 1 4.5947e+12 1.1932e+16 14943
## - stealing                1 5.6497e+12 1.1933e+16 14943
## - rebound.deffence       1 2.3847e+13 1.1951e+16 14944
## - contribution            1 2.5688e+13 1.1953e+16 14944
## - total.rebounds          1 3.2777e+13 1.1960e+16 14944
## - rebound.attack          1 3.5570e+13 1.1963e+16 14944
## - efficiency              1 4.1708e+13 1.1969e+16 14944
## <none>                    1.1928e+16 14945
## - triple.try              1 5.6019e+13 1.1984e+16 14945
## - usage.percentage        1 6.8209e+13 1.1996e+16 14945
## - NBA_DraftNumber         1 5.8559e+14 1.2513e+16 14966
## - minutes.played          1 7.1068e+14 1.2638e+16 14970
## - games                   1 9.9486e+14 1.2922e+16 14981
## - Age                     1 2.1782e+15 1.4106e+16 15024
##
## Step: AIC=14942.58
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + free.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + usage.percentage +
## good.attack + good.defence + total.good + win.shares.per.48 +
## puntos.ofensivosVSmedia + puntos.defensivosVSmedia + puntosVSmedia +
## contribution
##
##           Df Sum of Sq      RSS   AIC
## - free.try                1 7.9833e+11 1.1929e+16 14941
## - win.shares.per.48       1 8.4655e+11 1.1929e+16 14941
```

```

## - puntosVSmedia      1 1.9026e+12 1.1930e+16 14941
## - good.attack        1 2.1455e+12 1.1930e+16 14941
## - puntos.defensivosVSmedia 1 2.3904e+12 1.1930e+16 14941
## - blocking           1 2.9601e+12 1.1931e+16 14941
## - puntos.ofensivosVSmedia 1 3.9353e+12 1.1932e+16 14941
## - good.defence       1 3.9548e+12 1.1932e+16 14941
## - total.good         1 4.3514e+12 1.1932e+16 14941
## - assistance         1 4.5028e+12 1.1932e+16 14941
## - success            1 4.9180e+12 1.1933e+16 14941
## - stealing           1 6.1216e+12 1.1934e+16 14941
## - rebound.deffence  1 2.4001e+13 1.1952e+16 14942
## - contribution       1 2.7419e+13 1.1955e+16 14942
## - total.rebounds     1 3.2857e+13 1.1961e+16 14942
## - rebound.attack    1 3.5528e+13 1.1963e+16 14942
## - efficiency         1 4.1563e+13 1.1969e+16 14942
## <none>                1.1928e+16 14943
## - triple.try         1 5.7496e+13 1.1985e+16 14943
## - usage.percentage   1 6.8726e+13 1.1997e+16 14943
## - NBA_DraftNumber    1 5.8642e+14 1.2514e+16 14964
## - minutes.played     1 7.4947e+14 1.2677e+16 14970
## - games              1 1.0114e+15 1.2939e+16 14980
## - Age                1 2.1979e+15 1.4126e+16 15022
##
## Step: AIC=14940.61
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + usage.percentage +
## good.attack + good.defence + total.good + win.shares.per.48 +
## puntos.ofensivosVSmedia + puntos.defensivosVSmedia + puntosVSmedia +
## contribution
##
##           Df Sum of Sq      RSS   AIC
## - win.shares.per.48  1 5.7347e+11 1.1929e+16 14939
## - puntosVSmedia      1 1.9275e+12 1.1931e+16 14939
## - good.attack        1 1.9626e+12 1.1931e+16 14939
## - puntos.defensivosVSmedia 1 2.4285e+12 1.1931e+16 14939
## - blocking           1 2.7993e+12 1.1931e+16 14939
## - good.defence       1 3.6782e+12 1.1932e+16 14939
## - puntos.ofensivosVSmedia 1 3.9842e+12 1.1933e+16 14939
## - total.good         1 4.0905e+12 1.1933e+16 14939
## - assistance         1 5.1064e+12 1.1934e+16 14939
## - success            1 6.3387e+12 1.1935e+16 14939
## - stealing           1 6.3954e+12 1.1935e+16 14939
## - rebound.deffence  1 2.4198e+13 1.1953e+16 14940
## - contribution       1 2.7213e+13 1.1956e+16 14940
## - total.rebounds     1 3.2956e+13 1.1962e+16 14940
## - rebound.attack    1 3.5633e+13 1.1964e+16 14940
## - efficiency         1 4.0808e+13 1.1969e+16 14940
## <none>                1.1929e+16 14941
## - triple.try         1 5.6710e+13 1.1985e+16 14941
## - usage.percentage   1 6.7933e+13 1.1997e+16 14941
## - NBA_DraftNumber    1 5.9452e+14 1.2523e+16 14962
## - minutes.played     1 7.5413e+14 1.2683e+16 14968
## - games              1 1.0231e+15 1.2952e+16 14978

```

```

## - Age                1 2.2119e+15 1.4140e+16 15021
##
## Step: AIC=14938.63
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + usage.percentage +
## good.attack + good.defence + total.good + puntos.ofensivosVSmedia +
## puntos.defensivosVSmedia + puntosVSmedia + contribution
##
##           Df Sum of Sq      RSS   AIC
## - puntosVSmedia      1 1.8942e+12 1.1931e+16 14937
## - good.attack         1 2.0437e+12 1.1931e+16 14937
## - puntos.defensivosVSmedia 1 2.3824e+12 1.1932e+16 14937
## - blocking            1 2.4986e+12 1.1932e+16 14937
## - good.defence        1 3.5778e+12 1.1933e+16 14937
## - puntos.ofensivosVSmedia 1 4.0291e+12 1.1933e+16 14937
## - total.good          1 4.1872e+12 1.1933e+16 14937
## - stealing            1 8.0410e+12 1.1937e+16 14937
## - success             1 9.3658e+12 1.1939e+16 14937
## - assistance          1 9.4428e+12 1.1939e+16 14937
## - rebound.deffence    1 2.5243e+13 1.1954e+16 14938
## - contribution        1 2.7016e+13 1.1956e+16 14938
## - total.rebounds      1 3.4128e+13 1.1963e+16 14938
## - rebound.attack      1 3.7538e+13 1.1967e+16 14938
## <none>                  1.1929e+16 14939
## - efficiency          1 5.7754e+13 1.1987e+16 14939
## - triple.try           1 6.0637e+13 1.1990e+16 14939
## - usage.percentage     1 1.1780e+14 1.2047e+16 14941
## - NBA_DraftNumber      1 6.0118e+14 1.2530e+16 14960
## - minutes.played       1 9.0827e+14 1.2837e+16 14972
## - games               1 1.0261e+15 1.2955e+16 14976
## - Age                 1 2.2128e+15 1.4142e+16 15019
##
## Step: AIC=14936.71
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + usage.percentage +
## good.attack + good.defence + total.good + puntos.ofensivosVSmedia +
## puntos.defensivosVSmedia + contribution
##
##           Df Sum of Sq      RSS   AIC
## - good.attack         1 2.2294e+12 1.1933e+16 14935
## - blocking            1 2.2870e+12 1.1933e+16 14935
## - good.defence        1 3.7974e+12 1.1935e+16 14935
## - total.good          1 4.4397e+12 1.1935e+16 14935
## - puntos.defensivosVSmedia 1 6.8285e+12 1.1938e+16 14935
## - stealing            1 8.3145e+12 1.1939e+16 14935
## - success             1 9.0223e+12 1.1940e+16 14935
## - assistance          1 9.2490e+12 1.1940e+16 14935
## - rebound.deffence    1 2.6121e+13 1.1957e+16 14936
## - contribution        1 2.7041e+13 1.1958e+16 14936
## - total.rebounds      1 3.5074e+13 1.1966e+16 14936
## - rebound.attack      1 3.8494e+13 1.1970e+16 14936
## <none>                  1.1931e+16 14937

```

```

## - efficiency          1 5.6419e+13 1.1987e+16 14937
## - triple.try          1 5.9173e+13 1.1990e+16 14937
## - puntos.ofensivosVSmedia 1 6.4825e+13 1.1996e+16 14937
## - usage.percentage    1 1.1818e+14 1.2049e+16 14940
## - NBA_DraftNumber     1 6.0585e+14 1.2537e+16 14959
## - minutes.played      1 9.0701e+14 1.2838e+16 14970
## - games               1 1.0244e+15 1.2955e+16 14974
## - Age                 1 2.2211e+15 1.4152e+16 15017
##
## Step: AIC=14934.8
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + blocking + usage.percentage +
## good.defence + total.good + puntos.ofensivosVSmedia + puntos.defensivosVSmedia +
## contribution
##
##           Df Sum of Sq      RSS   AIC
## - blocking          1 2.5147e+12 1.1936e+16 14933
## - puntos.defensivosVSmedia 1 6.5060e+12 1.1940e+16 14933
## - stealing          1 7.5744e+12 1.1941e+16 14933
## - success           1 8.6969e+12 1.1942e+16 14933
## - assistance        1 9.9197e+12 1.1943e+16 14933
## - good.defence      1 1.1664e+13 1.1945e+16 14933
## - rebound.deffence  1 2.6162e+13 1.1959e+16 14934
## - contribution      1 2.7873e+13 1.1961e+16 14934
## - total.rebounds    1 3.5184e+13 1.1968e+16 14934
## - rebound.attack    1 3.8644e+13 1.1972e+16 14934
## <none>              1.1933e+16 14935
## - efficiency        1 5.7350e+13 1.1991e+16 14935
## - triple.try         1 5.9448e+13 1.1993e+16 14935
## - puntos.ofensivosVSmedia 1 6.5481e+13 1.1999e+16 14935
## - total.good         1 6.7750e+13 1.2001e+16 14936
## - usage.percentage   1 1.1919e+14 1.2052e+16 14938
## - NBA_DraftNumber    1 6.1313e+14 1.2546e+16 14957
## - minutes.played     1 9.0492e+14 1.2838e+16 14968
## - games              1 1.0223e+15 1.2956e+16 14972
## - Age                1 2.2192e+15 1.4153e+16 15015
##
## Step: AIC=14932.9
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## success + triple.try + rebound.attack + rebound.deffence +
## total.rebounds + assistance + stealing + usage.percentage +
## good.defence + total.good + puntos.ofensivosVSmedia + puntos.defensivosVSmedia +
## contribution
##
##           Df Sum of Sq      RSS   AIC
## - success           1 7.5740e+12 1.1943e+16 14931
## - assistance         1 1.3347e+13 1.1949e+16 14931
## - good.defence       1 1.5885e+13 1.1952e+16 14932
## - stealing           1 1.9633e+13 1.1955e+16 14932
## - puntos.defensivosVSmedia 1 2.5586e+13 1.1961e+16 14932
## - rebound.deffence  1 2.8375e+13 1.1964e+16 14932
## - contribution       1 3.2257e+13 1.1968e+16 14932
## - total.rebounds    1 3.6606e+13 1.1972e+16 14932

```



```

## - rebound.attack          1 3.9541e+13 1.1975e+16 14932
## <none>                      1.1936e+16 14933
## - triple.try              1 6.4339e+13 1.2000e+16 14934
## - total.good              1 6.5921e+13 1.2002e+16 14934
## - efficiency              1 8.0494e+13 1.2016e+16 14934
## - puntos.ofensivosVSmedia 1 9.1834e+13 1.2028e+16 14935
## - usage.percentage        1 1.2190e+14 1.2058e+16 14936
## - NBA_DraftNumber         1 6.2550e+14 1.2561e+16 14956
## - minutes.played          1 9.1109e+14 1.2847e+16 14966
## - games                   1 1.0249e+15 1.2961e+16 14971
## - Age                     1 2.2300e+15 1.4166e+16 15014
##
## Step: AIC=14931.21
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
##      triple.try + rebound.attack + rebound.deffence + total.rebounds +
##      assistance + stealing + usage.percentage + good.defence +
##      total.good + puntos.ofensivosVSmedia + puntos.defensivosVSmedia +
##      contribution
##
##              Df Sum of Sq      RSS   AIC
## - assistance    1 1.0384e+13 1.1954e+16 14930
## - good.defence   1 1.5493e+13 1.1959e+16 14930
## - stealing       1 1.5653e+13 1.1959e+16 14930
## - puntos.defensivosVSmedia 1 2.4747e+13 1.1968e+16 14930
## - rebound.deffence 1 2.5711e+13 1.1969e+16 14930
## - contribution   1 3.2902e+13 1.1976e+16 14930
## - total.rebounds 1 3.3685e+13 1.1977e+16 14931
## - rebound.attack 1 3.6474e+13 1.1980e+16 14931
## <none>           1.1943e+16 14931
## - triple.try     1 5.7859e+13 1.2001e+16 14932
## - total.good     1 6.4533e+13 1.2008e+16 14932
## - efficiency     1 8.2947e+13 1.2026e+16 14932
## - puntos.ofensivosVSmedia 1 8.4435e+13 1.2028e+16 14933
## - usage.percentage 1 1.2686e+14 1.2070e+16 14934
## - NBA_DraftNumber 1 6.3399e+14 1.2577e+16 14954
## - minutes.played 1 9.2840e+14 1.2872e+16 14965
## - games          1 1.0565e+15 1.3000e+16 14970
## - Age            1 2.2224e+15 1.4166e+16 15012
##
## Step: AIC=14929.63
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
##      triple.try + rebound.attack + rebound.deffence + total.rebounds +
##      stealing + usage.percentage + good.defence + total.good +
##      puntos.ofensivosVSmedia + puntos.defensivosVSmedia + contribution
##
##              Df Sum of Sq      RSS   AIC
## - good.defence    1 1.3674e+13 1.1967e+16 14928
## - stealing         1 1.6114e+13 1.1970e+16 14928
## - puntos.defensivosVSmedia 1 2.1136e+13 1.1975e+16 14928
## - contribution     1 2.6050e+13 1.1980e+16 14929
## - rebound.deffence 1 2.6228e+13 1.1980e+16 14929
## - total.rebounds   1 3.4480e+13 1.1988e+16 14929
## - rebound.attack   1 3.6227e+13 1.1990e+16 14929
## - triple.try       1 4.8211e+13 1.2002e+16 14930

```

```

## <none> 1.1954e+16 14930
## - total.good 1 7.3809e+13 1.2028e+16 14931
## - efficiency 1 7.5833e+13 1.2030e+16 14931
## - puntos.ofensivosVSmedia 1 7.6710e+13 1.2030e+16 14931
## - usage.percentage 1 1.1702e+14 1.2071e+16 14932
## - NBA_DraftNumber 1 6.4110e+14 1.2595e+16 14953
## - minutes.played 1 9.2446e+14 1.2878e+16 14964
## - games 1 1.0656e+15 1.3019e+16 14969
## - Age 1 2.2142e+15 1.4168e+16 15010
##
## Step: AIC=14928.18
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## triple.try + rebound.attack + rebound.deffence + total.rebounds +
## stealing + usage.percentage + total.good + puntos.ofensivosVSmedia +
## puntos.defensivosVSmedia + contribution
##
## Df Sum of Sq RSS AIC
## - puntos.defensivosVSmedia 1 1.3054e+13 1.1980e+16 14927
## - stealing 1 1.3666e+13 1.1981e+16 14927
## - rebound.deffence 1 2.0975e+13 1.1988e+16 14927
## - contribution 1 2.1716e+13 1.1989e+16 14927
## - total.rebounds 1 2.8572e+13 1.1996e+16 14927
## - rebound.attack 1 3.0256e+13 1.1998e+16 14927
## <none> 1.1967e+16 14928
## - triple.try 1 6.1444e+13 1.2029e+16 14929
## - total.good 1 6.6537e+13 1.2034e+16 14929
## - efficiency 1 9.3372e+13 1.2061e+16 14930
## - puntos.ofensivosVSmedia 1 9.6756e+13 1.2064e+16 14930
## - usage.percentage 1 1.1213e+14 1.2080e+16 14931
## - NBA_DraftNumber 1 6.3970e+14 1.2607e+16 14951
## - minutes.played 1 9.8628e+14 1.2954e+16 14964
## - games 1 1.0824e+15 1.3050e+16 14968
## - Age 1 2.2279e+15 1.4195e+16 15009
##
## Step: AIC=14926.7
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
## triple.try + rebound.attack + rebound.deffence + total.rebounds +
## stealing + usage.percentage + total.good + puntos.ofensivosVSmedia +
## contribution
##
## Df Sum of Sq RSS AIC
## - stealing 1 3.5855e+12 1.1984e+16 14925
## - rebound.deffence 1 2.3335e+13 1.2004e+16 14926
## - total.rebounds 1 3.2795e+13 1.2013e+16 14926
## - rebound.attack 1 3.5036e+13 1.2016e+16 14926
## - contribution 1 4.1953e+13 1.2022e+16 14926
## <none> 1.1980e+16 14927
## - total.good 1 5.6198e+13 1.2037e+16 14927
## - triple.try 1 7.7957e+13 1.2058e+16 14928
## - efficiency 1 8.9034e+13 1.2070e+16 14928
## - puntos.ofensivosVSmedia 1 9.2588e+13 1.2073e+16 14928
## - usage.percentage 1 1.0474e+14 1.2085e+16 14929
## - NBA_DraftNumber 1 6.6016e+14 1.2641e+16 14951
## - minutes.played 1 9.9212e+14 1.2973e+16 14963

```

```

## - games          1 1.1006e+15 1.3081e+16 14967
## - Age            1 2.2220e+15 1.4202e+16 15007
##
## Step: AIC=14924.85
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
##      triple.try + rebound.attack + rebound.deffence + total.rebounds +
##      usage.percentage + total.good + puntos.ofensivosVSmedia +
##      contribution
##
##              Df Sum of Sq      RSS   AIC
## - rebound.deffence    1 2.3150e+13 1.2007e+16 14924
## - total.rebounds       1 3.2841e+13 1.2017e+16 14924
## - rebound.attack       1 3.4831e+13 1.2019e+16 14924
## - contribution         1 3.8380e+13 1.2022e+16 14924
## <none>                  1.1984e+16 14925
## - total.good           1 6.4134e+13 1.2048e+16 14925
## - triple.try           1 7.7528e+13 1.2062e+16 14926
## - efficiency           1 9.9444e+13 1.2084e+16 14927
## - puntos.ofensivosVSmedia 1 1.0195e+14 1.2086e+16 14927
## - usage.percentage     1 1.1594e+14 1.2100e+16 14928
## - NBA_DraftNumber      1 6.5690e+14 1.2641e+16 14949
## - minutes.played       1 9.8877e+14 1.2973e+16 14961
## - games                1 1.1071e+15 1.3091e+16 14966
## - Age                  1 2.2380e+15 1.4222e+16 15006
##
## Step: AIC=14923.78
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
##      triple.try + rebound.attack + total.rebounds + usage.percentage +
##      total.good + puntos.ofensivosVSmedia + contribution
##
##              Df Sum of Sq      RSS   AIC
## - contribution         1 3.5447e+13 1.2043e+16 14923
## <none>                  1.2007e+16 14924
## - triple.try           1 7.6377e+13 1.2084e+16 14925
## - total.good           1 8.1813e+13 1.2089e+16 14925
## - efficiency           1 1.0378e+14 1.2111e+16 14926
## - puntos.ofensivosVSmedia 1 1.0573e+14 1.2113e+16 14926
## - rebound.attack       1 1.0859e+14 1.2116e+16 14926
## - usage.percentage     1 1.1926e+14 1.2126e+16 14927
## - total.rebounds       1 2.3228e+14 1.2239e+16 14931
## - NBA_DraftNumber      1 6.6587e+14 1.2673e+16 14948
## - minutes.played       1 9.6566e+14 1.2973e+16 14959
## - games                1 1.0977e+15 1.3105e+16 14964
## - Age                  1 2.2414e+15 1.4249e+16 15004
##
## Step: AIC=14923.2
## Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
##      triple.try + rebound.attack + total.rebounds + usage.percentage +
##      total.good + puntos.ofensivosVSmedia
##
##              Df Sum of Sq      RSS   AIC
## <none>                  1.2043e+16 14923
## - triple.try           1 8.1804e+13 1.2124e+16 14924
## - rebound.attack       1 1.2227e+14 1.2165e+16 14926

```

```
## - efficiency          1 1.3025e+14 1.2173e+16 14926
## - puntos.ofensivosVSmedia 1 1.3257e+14 1.2175e+16 14926
## - usage.percentage    1 1.3897e+14 1.2182e+16 14927
## - total.rebounds      1 2.6687e+14 1.2310e+16 14932
## - NBA_DraftNumber     1 6.5783e+14 1.2700e+16 14947
## - total.good          1 6.9027e+14 1.2733e+16 14948
## - minutes.played      1 9.6191e+14 1.3005e+16 14958
## - games               1 1.3853e+15 1.3428e+16 14974
## - Age                 1 2.2154e+15 1.4258e+16 15003

##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + games + minutes.played +
##     efficiency + triple.try + rebound.attack + total.rebounds +
##     usage.percentage + total.good + puntos.ofensivosVSmedia,
##     data = nba)
##
## Coefficients:
##             (Intercept)          NBA_DraftNumber              Age
##             -3893333          -62168              511716
##             games          minutes.played          efficiency
##             -159255              5204          -320540
##             triple.try          rebound.attack          total.rebounds
##             -3013992          -201694              333889
##             usage.percentage          total.good puntos.ofensivosVSmedia
##             130873              845245              550301
```

#Procedo a crear el segundo modelo en función de las variables que menor AIC poseen.

#En este caso eligo el modelo AIC=14923.2.

```
regres02<-lm (Salary ~ NBA_DraftNumber + Age + games + minutes.played + efficiency +
  triple.try + rebound.attack + total.rebounds + usage.percentage +
  total.good + puntos.ofensivosVSmedia, data=nba)

summary(regres02)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + games + minutes.played +
##     efficiency + triple.try + rebound.attack + total.rebounds +
##     usage.percentage + total.good + puntos.ofensivosVSmedia,
##     data = nba)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -15532214 -3017494 -365840  2303991 21608017
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -3893332.9  2785355.5  -1.398  0.16284
## NBA_DraftNumber    -62168.3   12256.4  -5.072 5.66e-07 ***
## Age             511716.1    54973.9   9.308 < 2e-16 ***
## games          -159255.3    21635.8  -7.361 8.21e-13 ***
```

```
## minutes.played          5204.0          848.4    6.134 1.82e-09 ***
## efficiency              -320540.0    142016.8   -2.257  0.02446 *
## triple.try              -3013991.7    1685020.0   -1.789  0.07431 .
## rebound.attack         -201694.2     92231.2   -2.187  0.02925 *
## total.rebounds          333889.2    103347.8    3.231  0.00132 **
## usage.percentage        130872.7     56136.5    2.331  0.02016 *
## total.good              845245.2    162676.1    5.196 3.04e-07 ***
## puntos.ofensivosVSmedia  550301.3    241673.6    2.277  0.02323 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5057000 on 471 degrees of freedom
## Multiple R-squared:  0.5435, Adjusted R-squared:  0.5329
## F-statistic: 50.98 on 11 and 471 DF,  p-value: < 2.2e-16
```

#Para detectar la multicolinealidad utilizo el Factor de Inflación de Varianza (VIF).

```
vif(regres02)
```

```
##      NBA_DraftNumber      Age      games
##      1.262696      1.040844      5.394048
##      minutes.played      efficiency      triple.try
##      8.895257      29.081095      2.755505
##      rebound.attack      total.rebounds      usage.percentage
##      3.366046      4.884538      2.016768
##      total.good puntos.ofensivosVSmedia
##      3.568490      27.843542
```

```
sqrt(vif(regres02)) > 2
```

```
##      NBA_DraftNumber      Age      games
##      FALSE      FALSE      TRUE
##      minutes.played      efficiency      triple.try
##      TRUE      TRUE      FALSE
##      rebound.attack      total.rebounds      usage.percentage
##      FALSE      TRUE      FALSE
##      total.good puntos.ofensivosVSmedia
##      FALSE      TRUE
```

#Cuando la raiz cuadrada de VIF>2 se considera que hay problemas de multicolinealidad.

#Suprimo las variables que dan problemas y creo un nuevo modelo de regresion.

```
regres03<-lm (Salary ~ NBA_DraftNumber + Age + minutes.played + efficiency +
              triple.try + rebound.attack + total.rebounds + usage.percentage +
              total.good + puntos.ofensivosVSmedia, data=nba)
```

```
summary(regres03)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + minutes.played +
```

```
##      efficiency + triple.try + rebound.attack + total.rebounds +
##      usage.percentage + total.good + puntos.ofensivosVSmedia,
##      data = nba)
##
## Residuals:
##      Min        1Q      Median        3Q        Max
## -14111096  -3427694  -413634    2465195   24139185
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -9456814.3   2827838.5   -3.344  0.000891 ***
## NBA_DraftNumber    -64754.8    12923.2   -5.011  7.68e-07 ***
## Age              506234.7    57982.8    8.731  < 2e-16 ***
## minutes.played      255.1      545.9     0.467  0.640482
## efficiency       -256423.5   149521.8   -1.715  0.087009 .
## triple.try       -1713977.4  1767622.1   -0.970  0.332717
## rebound.attack   -175407.6    97215.4   -1.804  0.071819 .
## total.rebounds    232414.7   108040.3    2.151  0.031968 *
## usage.percentage   248633.4    56758.9    4.381  1.46e-05 ***
## total.good       1108406.1   167400.5    6.621  9.71e-11 ***
## puntos.ofensivosVSmedia 388790.8   253872.1    1.531  0.126330
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5334000 on 472 degrees of freedom
## Multiple R-squared:  0.491, Adjusted R-squared:  0.4802
## F-statistic: 45.53 on 10 and 472 DF, p-value: < 2.2e-16
```

```
vif(regres03)
```

```
##      NBA_DraftNumber      Age      minutes.played
##      1.261658      1.040653      3.309359
##      efficiency      triple.try      rebound.attack
##      28.971692      2.725233      3.361000
##      total.rebounds      usage.percentage      total.good
##      4.797623      1.852964      3.396129
## puntos.ofensivosVSmedia
##      27.614018
```

```
sqrt(vif(regres03)) > 2
```

```
##      NBA_DraftNumber      Age      minutes.played
##      FALSE      FALSE      FALSE
##      efficiency      triple.try      rebound.attack
##      TRUE      FALSE      FALSE
##      total.rebounds      usage.percentage      total.good
##      TRUE      FALSE      FALSE
## puntos.ofensivosVSmedia
##      TRUE
```

```
regres04<-lm (Salary ~ NBA_DraftNumber + Age + minutes.played +
              triple.try + rebound.attack + total.rebounds + usage.percentage +
```

```
total.good + puntos.ofensivosVSmedia, data=nba)

summary(regres04)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + minutes.played +
##     triple.try + rebound.attack + total.rebounds + usage.percentage +
##     total.good + puntos.ofensivosVSmedia, data = nba)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13902696 -3358110  -373446   2282127  23887665
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.252e+07  2.199e+06  -5.692 2.21e-08 ***
## NBA_DraftNumber -6.644e+04  1.291e+04  -5.145 3.92e-07 ***
## Age           5.134e+05  5.795e+04   8.860 < 2e-16 ***
## minutes.played  5.364e+02  5.217e+02   1.028  0.304
## triple.try      1.761e+05  1.385e+06   0.127  0.899
## rebound.attack -1.405e+05  9.526e+04  -1.475  0.141
## total.rebounds  1.268e+05  8.897e+04   1.426  0.155
## usage.percentage 1.896e+05  4.521e+04   4.193 3.28e-05 ***
## total.good      1.112e+06  1.677e+05   6.631 9.10e-11 ***
## puntos.ofensivosVSmedia -3.594e+04  5.591e+04  -0.643  0.521
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5345000 on 473 degrees of freedom
## Multiple R-squared:  0.4878, Adjusted R-squared:  0.4781
## F-statistic: 50.06 on 9 and 473 DF,  p-value: < 2.2e-16
```

```
vif(regres04)
```

```
##      NBA_DraftNumber      Age      minutes.played
##      1.254376      1.035220      3.010552
##      triple.try      rebound.attack      total.rebounds
##      1.665766      3.213682      3.240222
##      usage.percentage      total.good puntos.ofensivosVSmedia
##      1.170819      3.395508      1.333826
```

```
sqrt(vif(regres04)) > 2
```

```
##      NBA_DraftNumber      Age      minutes.played
##      FALSE      FALSE      FALSE
##      triple.try      rebound.attack      total.rebounds
##      FALSE      FALSE      FALSE
##      usage.percentage      total.good puntos.ofensivosVSmedia
##      FALSE      FALSE      FALSE
```

#De esta manera he conseguido eliminar la multicolinealidad.

#Comparo el modelo inicial “regres01” con el nuevo modelo obtenido “regres04”.

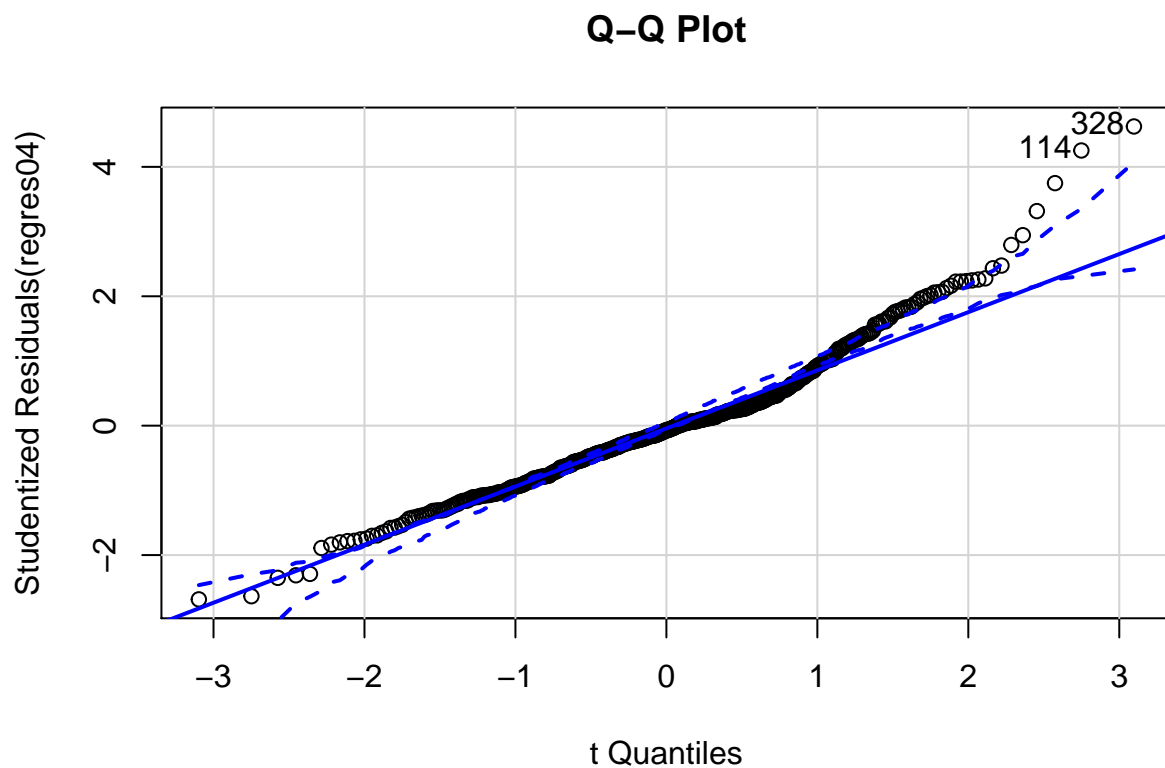
```
BIC(regres01,regres04)
```

```
##          df          BIC
## regres01 26 16425.94
## regres04 11 16393.47
```

#Observo que el modelo regres04 es mejor ya que tiene menor BIC.

#Compruebo la Normalidad para comprobar si se verifica la hipótesis de normalidad #necesaria para que el resultado de de esta prediccion sea fiable.

```
qqPlot(regres04, labels=row.names(nba), id.method="identify",
        simulate=TRUE, main="Q-Q Plot")
```



```
## 114 328
## 112 326
```

#Contrastamos todas las hipótesis del modelo mediante el test de Peña de Validacion Global.



```
validacion_global <- gvlma(regres04)
summary(validacion_global)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + minutes.played +
##      triple.try + rebound.attack + total.rebounds + usage.percentage +
##      total.good + puntos.ofensivosVSmedia, data = nba)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13902696 -3358110  -373446   2282127  23887665
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.252e+07  2.199e+06  -5.692 2.21e-08 ***
## NBA_DraftNumber -6.644e+04  1.291e+04  -5.145 3.92e-07 ***
## Age             5.134e+05  5.795e+04   8.860 < 2e-16 ***
## minutes.played  5.364e+02  5.217e+02   1.028  0.304
## triple.try      1.761e+05  1.385e+06   0.127  0.899
## rebound.attack -1.405e+05  9.526e+04  -1.475  0.141
## total.rebounds  1.268e+05  8.897e+04   1.426  0.155
## usage.percentage 1.896e+05  4.521e+04   4.193 3.28e-05 ***
## total.good      1.112e+06  1.677e+05   6.631 9.10e-11 ***
## puntos.ofensivosVSmedia -3.594e+04  5.591e+04  -0.643  0.521
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5345000 on 473 degrees of freedom
## Multiple R-squared:  0.4878, Adjusted R-squared:  0.4781
## F-statistic: 50.06 on 9 and 473 DF,  p-value: < 2.2e-16
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance = 0.05
##
## Call:
## gvlma(x = regres04)
##
##              Value    p-value              Decision
## Global Stat    121.9845 0.000e+00 Assumptions NOT satisfied!
## Skewness       48.0446 4.166e-12 Assumptions NOT satisfied!
## Kurtosis       53.8223 2.195e-13 Assumptions NOT satisfied!
## Link Function  19.9294 8.036e-06 Assumptions NOT satisfied!
## Heteroscedasticity 0.1881 6.645e-01 Assumptions acceptable.
```

```
gvlma(x = regres04)
```

```
##
## Call:
## lm(formula = Salary ~ NBA_DraftNumber + Age + minutes.played +
```

```
##      triple.try + rebound.attack + total.rebounds + usage.percentage +
##      total.good + puntos.ofensivosVSmedia, data = nba)
##
## Coefficients:
##              (Intercept)          NBA_DraftNumber              Age
##             -1.252e+07             -6.644e+04             5.134e+05
##      minutes.played          triple.try          rebound.attack
##             5.364e+02             1.761e+05             -1.405e+05
##      total.rebounds          usage.percentage          total.good
##             1.268e+05             1.896e+05             1.112e+06
## puntos.ofensivosVSmedia
##             -3.594e+04
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance = 0.05
##
## Call:
## gvlma(x = regres04)
##
##              Value    p-value              Decision
## Global Stat      121.9845 0.000e+00 Assumptions NOT satisfied!
## Skewness         48.0446 4.166e-12 Assumptions NOT satisfied!
## Kurtosis         53.8223 2.195e-13 Assumptions NOT satisfied!
## Link Function    19.9294 8.036e-06 Assumptions NOT satisfied!
## Heteroscedasticity 0.1881 6.645e-01 Assumptions acceptable.
```

#Realizamos la prediccion de un jugador al azar.

```
predict.lm(regres04, data.frame(NBA_DraftNumber=7, Age=29, minutes.played=1631,
                                triple.try=0.58, rebound.attack=2.7, total.rebounds=8.9,
                                usage.percentage=31,
                                total.good=9.2, puntos.ofensivosVSmedia=9.8 ))
```

```
##          1
## 19392638
```

#Realizamos un Cross Validation para dividir los datos del modelo predictivo en nuevos #conjuntos de datos, que son el conjunto de entrenamiento del modelo y el de #validacion el modelo

```
set.seed(6)
nbanum<-nrow(nba)
```

```
training<-sample(nbanum, nbanum/2)
```

```
regres.training <- lm(Salary~(NBA_DraftNumber + Age + minutes.played +
                              triple.try + rebound.attack + total.rebounds + usage.percentage +
                              total.good + puntos.ofensivosVSmedia), nba , subset =training )
```

```
attach(nba)
```

```
mean((Salary-predict(regres.training, Auto))[-training ]^2)
```

```
## Warning: 'newdata' had 392 rows but variables found have 483 rows
```

```
## [1] 3.191535e+13
```

```
sqrt(mean((Salary-predict(regres.training, Auto))[-training ]^2))
```

```
## Warning: 'newdata' had 392 rows but variables found have 483 rows
```

```
## [1] 5649367
```

#Por último, he obtenido el resultado de error que da mi modelo de prediccion.