

Tarea 7. Regresión Lineal

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```
library(corrplot)

## corrplot 0.92 loaded

library(stats)
library(ggplot2)
library(lmtest)

## Loading required package: zoo

##
## Attaching package: 'zoo'

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

M = read.csv("Estatura-peso_HyM.csv")
MM = subset(M, M$Sexo == "M")
MH = subset(M, M$Sexo == "H")
M1 = data.frame(MH$Estatura, MH$Peso, MM$Estatura, MM$Peso)
M1
```

	MH.Estatura	MH.Peso	MM.Estatura	MM.Peso
## 1	1.61	72.21	1.53	50.07
## 2	1.61	65.71	1.60	59.78
## 3	1.70	75.08	1.54	50.66
## 4	1.65	68.55	1.58	56.96
## 5	1.72	70.77	1.61	51.03
## 6	1.63	77.18	1.57	64.27
## 7	1.76	81.21	1.61	68.62
## 8	1.67	75.71	1.52	54.53
## 9	1.67	76.57	1.62	66.96
## 10	1.65	68.78	1.63	66.94
## 11	1.63	65.13	1.55	59.84
## 12	1.70	77.53	1.60	55.46
## 13	1.69	70.91	1.51	57.54
## 14	1.59	71.77	1.59	50.05
## 15	1.71	80.98	1.53	50.25
## 16	1.66	74.11	1.67	64.36
## 17	1.65	72.45	1.56	53.79
## 18	1.59	64.60	1.65	59.07
## 19	1.59	62.08	1.52	45.19

## 20	1.67	66.01	1.61	61.36
## 21	1.71	83.67	1.65	62.32
## 22	1.68	76.17	1.61	44.74
## 23	1.59	69.66	1.57	54.06
## 24	1.70	73.40	1.63	64.00
## 25	1.68	72.91	1.69	74.50
## 26	1.61	67.22	1.54	55.31
## 27	1.70	74.79	1.59	49.31
## 28	1.70	79.19	1.53	49.86
## 29	1.63	69.44	1.54	51.47
## 30	1.72	83.29	1.57	69.89
## 31	1.62	71.23	1.55	55.81
## 32	1.69	75.60	1.52	59.31
## 33	1.58	66.97	1.61	43.31
## 34	1.68	77.18	1.56	47.79
## 35	1.62	70.26	1.58	54.92
## 36	1.65	79.13	1.61	55.84
## 37	1.58	65.61	1.56	44.44
## 38	1.68	78.34	1.59	52.35
## 39	1.64	74.04	1.55	51.77
## 40	1.74	79.77	1.56	51.36
## 41	1.63	81.80	1.57	44.07
## 42	1.60	65.65	1.62	55.77
## 43	1.61	72.97	1.63	60.38
## 44	1.65	72.51	1.57	55.42
## 45	1.69	79.56	1.54	59.78
## 46	1.67	73.36	1.57	56.32
## 47	1.61	65.14	1.64	49.37
## 48	1.56	61.87	1.56	47.73
## 49	1.63	73.37	1.55	58.44
## 50	1.59	68.26	1.57	44.90
## 51	1.58	73.23	1.48	45.47
## 52	1.66	70.13	1.62	69.63
## 53	1.55	56.43	1.53	62.16
## 54	1.71	72.46	1.56	54.30
## 55	1.65	69.52	1.57	53.92
## 56	1.65	68.04	1.64	57.27
## 57	1.71	73.39	1.55	47.50
## 58	1.80	90.05	1.55	47.54
## 59	1.59	73.83	1.66	62.52
## 60	1.56	60.66	1.53	60.01
## 61	1.56	64.74	1.68	67.30
## 62	1.71	86.37	1.45	47.39
## 63	1.77	84.91	1.61	61.55
## 64	1.72	81.56	1.61	52.00
## 65	1.68	77.36	1.62	56.90
## 66	1.64	78.54	1.55	59.40
## 67	1.71	77.18	1.54	53.67
## 68	1.71	84.70	1.58	57.70
## 69	1.55	57.51	1.48	52.28

## 70	1.71	82.13	1.74	70.63
## 71	1.63	65.95	1.62	63.08
## 72	1.66	71.20	1.60	55.28
## 73	1.63	68.25	1.64	58.22
## 74	1.74	75.72	1.62	59.86
## 75	1.54	67.57	1.61	54.48
## 76	1.66	74.14	1.47	49.03
## 77	1.68	72.67	1.63	62.14
## 78	1.71	79.57	1.60	64.37
## 79	1.70	73.05	1.52	58.38
## 80	1.59	70.52	1.53	44.87
## 81	1.56	61.16	1.65	61.80
## 82	1.64	71.88	1.57	60.08
## 83	1.75	81.26	1.53	49.15
## 84	1.77	82.24	1.57	53.70
## 85	1.70	76.33	1.52	51.33
## 86	1.60	74.07	1.64	57.98
## 87	1.54	57.36	1.63	53.79
## 88	1.68	75.33	1.54	48.45
## 89	1.59	69.12	1.54	59.78
## 90	1.65	69.00	1.58	43.67
## 91	1.70	71.26	1.47	51.63
## 92	1.63	73.02	1.59	50.59
## 93	1.71	77.38	1.58	57.65
## 94	1.71	73.24	1.63	58.09
## 95	1.61	70.22	1.62	61.73
## 96	1.69	80.33	1.60	58.12
## 97	1.72	75.93	1.55	44.47
## 98	1.62	67.40	1.60	55.09
## 99	1.73	81.90	1.59	47.43
## 100	1.64	72.76	1.49	45.65
## 101	1.67	77.14	1.58	55.63
## 102	1.71	78.51	1.54	54.25
## 103	1.65	69.92	1.59	61.71
## 104	1.67	73.90	1.56	52.57
## 105	1.62	65.22	1.52	59.21
## 106	1.64	75.17	1.56	57.24
## 107	1.74	83.95	1.66	77.07
## 108	1.59	62.42	1.51	45.62
## 109	1.77	82.05	1.48	60.04
## 110	1.57	64.17	1.61	67.96
## 111	1.65	72.02	1.63	51.18
## 112	1.62	74.86	1.49	37.39
## 113	1.68	77.62	1.50	44.09
## 114	1.66	69.83	1.62	63.59
## 115	1.61	67.76	1.55	44.76
## 116	1.68	71.05	1.51	53.22
## 117	1.70	77.52	1.58	55.18
## 118	1.71	84.47	1.55	53.75
## 119	1.60	63.87	1.55	52.40

## 120	1.69	73.77	1.57	52.12
## 121	1.65	70.20	1.58	68.31
## 122	1.62	76.16	1.51	50.06
## 123	1.61	70.00	1.55	49.08
## 124	1.66	72.28	1.47	50.69
## 125	1.71	74.50	1.54	58.85
## 126	1.62	71.11	1.58	53.36
## 127	1.62	77.41	1.49	50.16
## 128	1.68	77.41	1.61	68.73
## 129	1.56	65.61	1.56	57.84
## 130	1.55	65.78	1.52	52.01
## 131	1.62	65.66	1.64	64.62
## 132	1.67	78.72	1.64	64.74
## 133	1.58	65.59	1.56	54.49
## 134	1.70	75.49	1.57	58.34
## 135	1.55	66.33	1.58	68.31
## 136	1.57	56.69	1.53	48.57
## 137	1.63	69.50	1.56	48.29
## 138	1.66	73.21	1.55	57.06
## 139	1.75	79.06	1.59	62.60
## 140	1.73	81.14	1.44	48.79
## 141	1.52	59.07	1.53	45.25
## 142	1.78	82.66	1.60	64.35
## 143	1.71	77.84	1.62	56.02
## 144	1.74	80.47	1.58	49.08
## 145	1.70	84.18	1.61	66.38
## 146	1.78	86.74	1.53	47.90
## 147	1.64	72.82	1.55	50.33
## 148	1.69	73.53	1.55	54.06
## 149	1.69	69.95	1.56	54.46
## 150	1.64	63.81	1.58	54.32
## 151	1.67	74.32	1.55	42.95
## 152	1.54	60.85	1.49	51.95
## 153	1.67	82.10	1.64	73.85
## 154	1.57	69.74	1.60	46.85
## 155	1.62	63.58	1.60	52.14
## 156	1.59	62.15	1.59	60.57
## 157	1.69	74.48	1.48	41.82
## 158	1.68	78.51	1.63	63.98
## 159	1.65	74.34	1.54	57.28
## 160	1.70	78.27	1.55	43.92
## 161	1.71	79.17	1.62	46.59
## 162	1.65	71.21	1.65	40.01
## 163	1.62	70.16	1.60	64.88
## 164	1.63	72.02	1.48	46.36
## 165	1.64	74.41	1.62	64.46
## 166	1.63	73.11	1.51	48.44
## 167	1.66	67.01	1.48	39.73
## 168	1.62	70.38	1.54	46.27
## 169	1.68	79.32	1.49	53.41

## 170	1.68	72.02	1.58	42.95
## 171	1.66	77.57	1.57	54.75
## 172	1.61	70.35	1.57	55.41
## 173	1.48	60.35	1.58	57.38
## 174	1.65	72.98	1.58	43.60
## 175	1.77	81.69	1.62	63.67
## 176	1.66	70.08	1.61	57.63
## 177	1.60	67.94	1.63	49.99
## 178	1.67	75.72	1.54	53.56
## 179	1.61	64.22	1.59	49.10
## 180	1.66	71.06	1.50	52.11
## 181	1.60	68.15	1.59	65.62
## 182	1.67	75.62	1.48	55.53
## 183	1.74	80.75	1.55	44.04
## 184	1.67	69.56	1.56	44.07
## 185	1.65	79.16	1.56	57.69
## 186	1.54	58.36	1.54	53.21
## 187	1.63	73.29	1.52	50.56
## 188	1.63	79.85	1.60	48.68
## 189	1.65	67.79	1.56	58.85
## 190	1.61	71.75	1.51	52.42
## 191	1.64	76.40	1.55	51.13
## 192	1.63	70.42	1.60	69.30
## 193	1.67	73.55	1.55	52.48
## 194	1.73	78.27	1.54	39.54
## 195	1.80	83.60	1.63	64.34
## 196	1.80	90.49	1.55	46.38
## 197	1.74	81.06	1.55	48.45
## 198	1.61	67.56	1.60	47.98
## 199	1.67	78.69	1.53	47.10
## 200	1.51	61.90	1.66	61.30
## 201	1.57	59.58	1.57	51.59
## 202	1.63	71.16	1.66	49.41
## 203	1.66	72.77	1.68	75.52
## 204	1.72	74.07	1.51	59.77
## 205	1.69	74.43	1.64	57.19
## 206	1.58	61.79	1.54	59.13
## 207	1.52	61.38	1.55	51.13
## 208	1.78	87.55	1.57	44.37
## 209	1.75	87.66	1.59	51.87
## 210	1.56	66.29	1.58	40.15
## 211	1.64	72.55	1.69	57.37
## 212	1.66	70.59	1.57	57.14
## 213	1.61	66.86	1.59	61.06
## 214	1.59	66.13	1.57	59.44
## 215	1.79	90.02	1.64	63.81
## 216	1.54	59.06	1.58	66.39
## 217	1.75	82.11	1.57	65.89
## 218	1.64	73.79	1.56	56.48

```
## 219      1.58   64.66      1.61   59.16
## 220      1.65   70.50      1.67   80.87
```

```
A = lm(M$Peso~M$Estatura+M$Sexo)
```

```
A
```

```
##
```

```
## Call:
```

```
## lm(formula = M$Peso ~ M$Estatura + M$Sexo)
```

```
##
```

```
## Coefficients:
```

```
## (Intercept)    M$Estatura    M$SexoM
```

```
##      -74.75         89.26        -10.56
```

```
summary(A)
```

```
##
```

```
## Call:
```

```
## lm(formula = M$Peso ~ M$Estatura + M$Sexo)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -21.9505  -3.2491   0.0489   3.2880  17.1243
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) -74.7546     7.5555  -9.894  <2e-16 ***
```

```
## M$Estatura   89.2604     4.5635  19.560  <2e-16 ***
```

```
## M$SexoM     -10.5645     0.6317 -16.724  <2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 5.381 on 437 degrees of freedom
```

```
## Multiple R-squared:  0.7837, Adjusted R-squared:  0.7827
```

```
## F-statistic: 791.5 on 2 and 437 DF,  p-value: < 2.2e-16
```

El estadístico F pertenece a la significancia global y, como se puede observar, se encuentra muy lejano al valor 1, lo cual indica que la regresión lineal sí se puede usar en el modelo. El valor de t pertenece a la significancia individual, esta muy apartado del valor hipotético (19 y 16) El coeficiente de determinación corresponde al valor de r cuadrado ajustado porque tenemos más de una variable

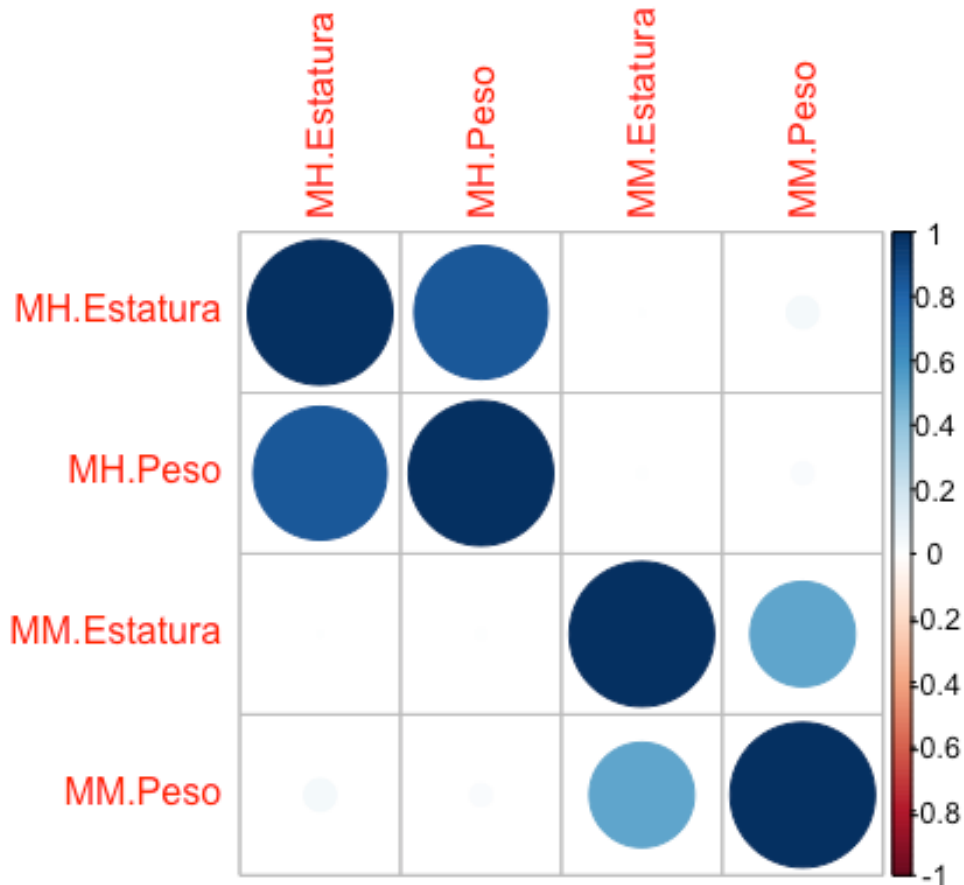
Por medio de la R^2 podemos interpretar que el modelo explica el 78.37% de la variabilidad del peso por medio de las variables estatura y sexo. La ecuación del modelo es: $y = -74.75 + 89.2604 * E - 10.564 * S$

La recta de mejor ajuste (Primera entrega)

```
# Calcular la matriz de correlación
```

```
matrizCorrelacion = cor(M1)
```

```
corrplot(matrizCorrelacion, method = "circle")
```



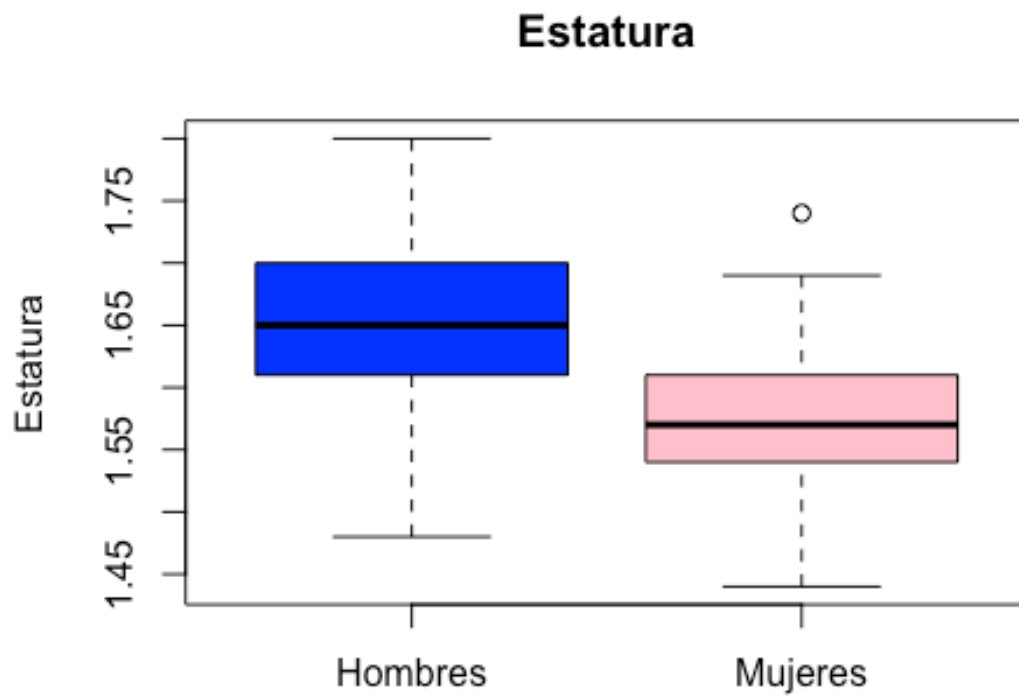
Interpretación:

```
n=4 #número de variables
d=matrix(NA,ncol=7,nrow=n)
for(i in 1:n){
  d[i,]<-c(as.numeric(summary(M1[,i])),sd(M1[,i]))
}
m=as.data.frame(d)

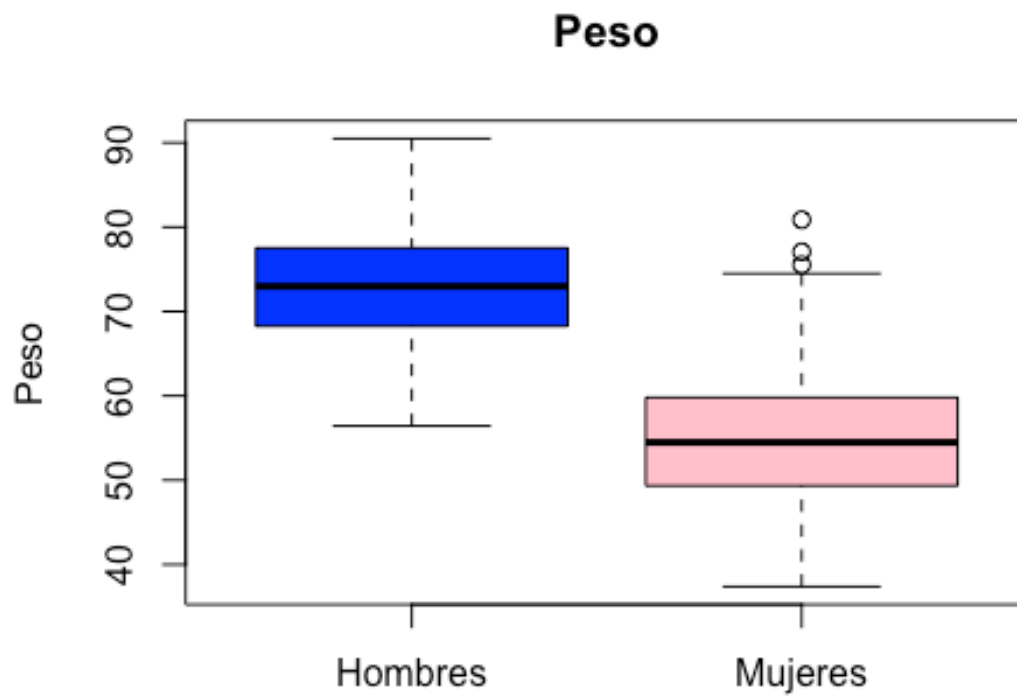
row.names(m)=c("H-Estatura","H-Peso","M-Estatura","M-Peso")
names(m)=c("Minimo","Q1","Mediana","Media","Q3","Máximo","Desv Est")
m

##           Minimo      Q1 Mediana      Media      Q3 Máximo  Desv Est
## H-Estatura   1.48  1.6100   1.650   1.653727  1.7000   1.80  0.06173088
## H-Peso       56.43 68.2575  72.975  72.857682  77.5225  90.49  6.90035408
## M-Estatura   1.44  1.5400   1.570   1.572955  1.6100   1.74  0.05036758
## M-Peso       37.39 49.3550  54.485  55.083409  59.7950  80.87  7.79278074

boxplot(M$Estatura~M$Sexo, ylab="Estatura", xlab="", col=c("blue","pink"),
names=c("Hombres", "Mujeres"), main="Estatura")
```



```
boxplot(M$Peso~M$Sexo, ylab="Peso", xlab="", names=c("Hombres", "Mujeres"),  
col=c("blue", "pink"), main="Peso")
```

```

b0 = A$coefficients[1]
b1 = A$coefficients[2]
b2 = A$coefficients[3]

cat("Peso =", b0, "+", b1, "Estatura", b2, "SexoM")

## Peso = -74.7546 + 89.26035 Estatura -10.56447 SexoM

# Para Mujeres (SexoM=1)
cat("Para mujeres", "\n")

## Para mujeres

cat("Peso =", b0+b2, "+", b1, "Estatura", "\n")

## Peso = -85.31907 + 89.26035 Estatura

# Para Mujeres (SexoM=0)
cat("Para hombres", "\n")

## Para hombres

cat("Peso =", b0, "+", b1, "Estatura")

## Peso = -74.7546 + 89.26035 Estatura

```

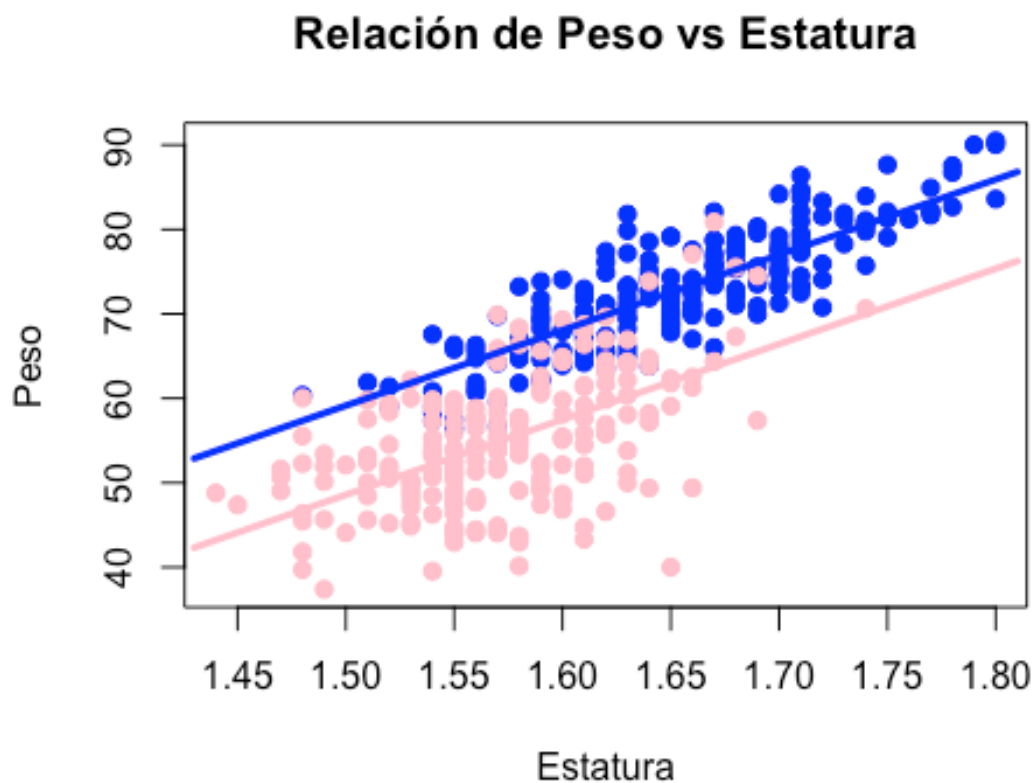
```

Ym = function(x) {b0 + b2 + b1 * x}
Yh = function(x) {b0 + b1 * x}

colores = c("blue", "pink")
plot(M$Estatura, M$Peso, col=colores[factor(M$Sexo)], pch=19, ylab="Peso",
xlab = "Estatura", main="Relación de Peso vs Estatura")

# 1.43 = min(M$Estatura)
# 1.81 = max(M$Estatura)
x = seq(1.43, 1.81, 0.01)
lines(x, Ym(x), col="pink", lwd=3)
lines(x, Yh(x), col="blue", lwd=3)

```



5. Interpreta en el contexto del problema cada uno de los análisis que hiciste.
6. Interpreta en el contexto del problema:
 - ¿Qué información proporciona $\hat{\beta}_0$ sobre la relación entre la estatura y el peso de hombres y mujeres?
 - ¿Cómo interpretas $\hat{\beta}_1$ en la relación entre la estatura y el peso de hombres y mujeres?

Validación del Modelo (segunda entrega)

Normalidad

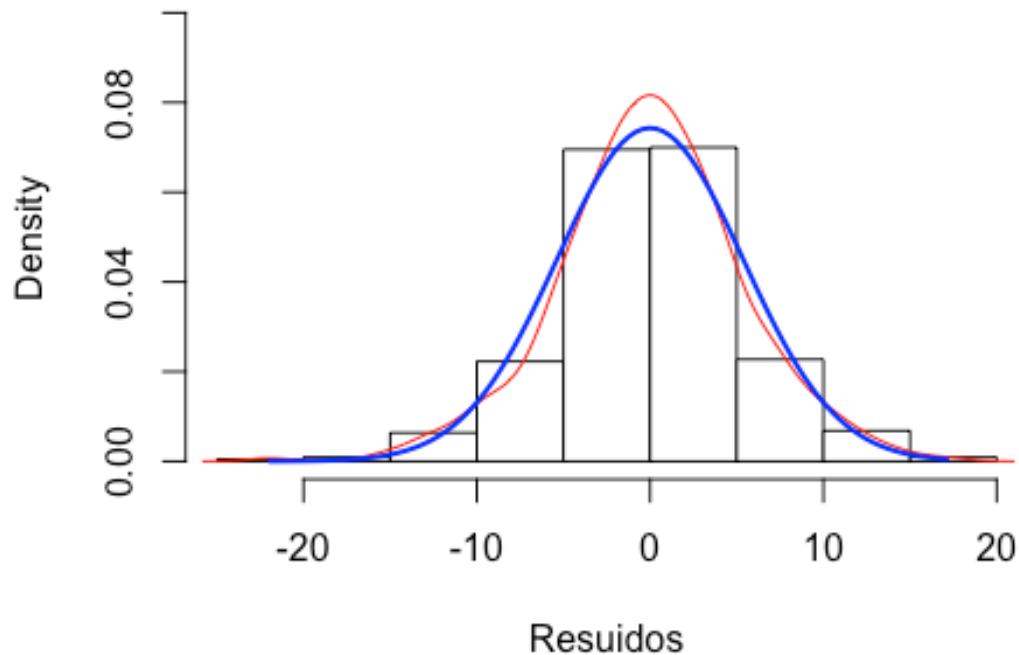
- Histograma
- Prueba de Hipótesis
- QQplot

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

```
##  
## Shapiro-Wilk normality test  
##  
## data: A$residuals  
## W = 0.99337, p-value = 0.0501
```

```
hist(A$residuals, freq = FALSE, ylim = c(0, 0.1), xlab = "Resuidos", col = 0,  
main = "Histograma de Residuos")  
lines(density(A$residuals), col = "red", ylim = c(0, 0.1))  
curve(dnorm(x, mean = mean(A$residuals), sd = sd(A$residuals)), from =  
min(A$residuals), to = max(A$residuals), add = TRUE, col = "blue", lwd = 2)
```

Histograma de Residuos



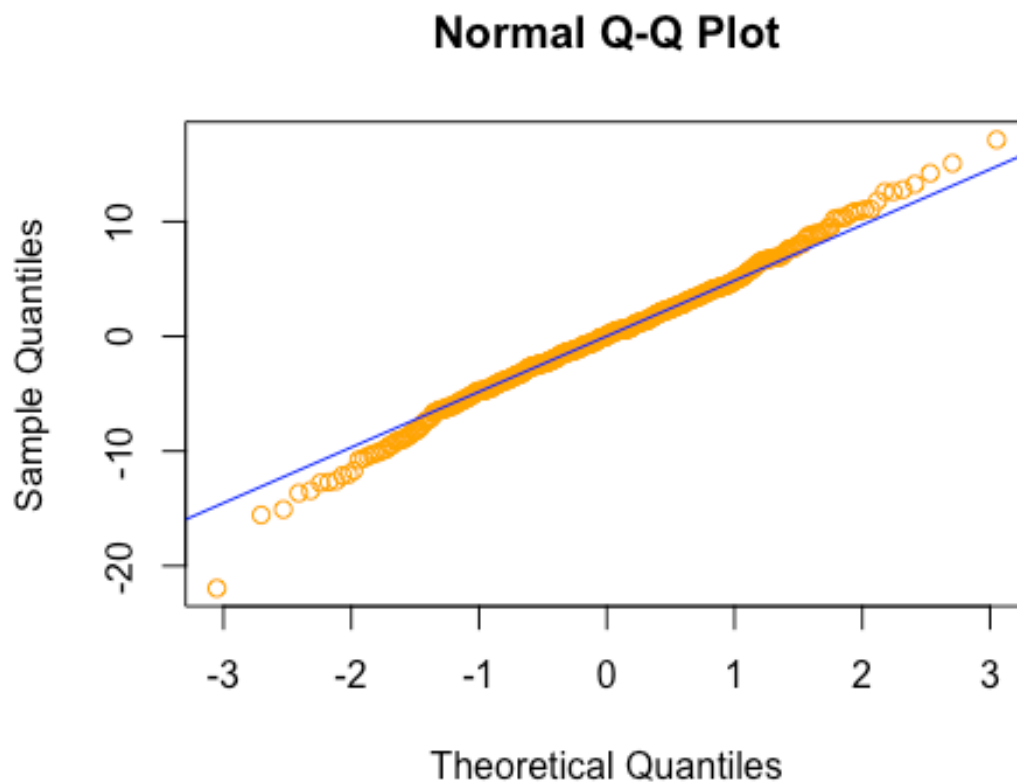
Se realiza la prueba de Breusch-Pagan para identificar si existe homocedasticidad en los datos. Por lo tanto, se establecen las siguientes hipótesis: H_0 : Homocedasticidad H_1 : Heterocedasticidad

```
bptest(A)

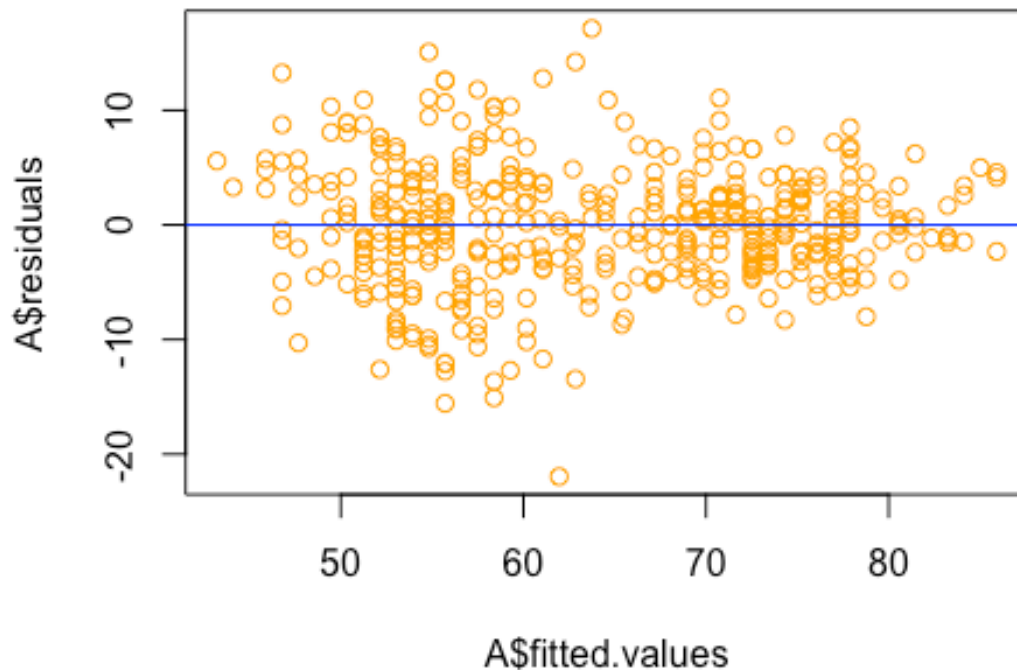
##
## studentized Breusch-Pagan test
##
## data: A
## BP = 48.202, df = 2, p-value = 3.413e-11
```

Como $\alpha > \text{Valor } p$ (3.413×10^{-11}), se rechaza H_0 a favor de H_a por lo que hay heterocedasticidad

```
qqnorm(A$residuals, col="orange")
qqline(A$residuals, col="blue")
```



```
plot(A$fitted.values, A$residuals, col=c('orange'))
abline(h=0, col="blue")
```



De la regresión que se realizó, primero se revisó que se cumplieran los supuestos de los Mínimos Cuadrados Ordinarios verificando, por su parte la normalidad, el supuesto de homocedasticidad y la correlación de los datos. Con este modelo se puede identificar el sexo de una persona con base en su estatura y peso. Durante el análisis se pudieron observar algunas tendencias en ambas variables independientes siendo, por ejemplo, que los hombres tienden a tener un mayor peso que las mujeres.

Al haber realizado pruebas de los supuestos (tanto gráficamente como numéricamente), graficado los residuos, y examinado gráficamente que se esté cumpliendo la normalidad, se puede determinar que el modelo tiene buena capacidad de predecir el sexo de una persona con base en las variables que se tienen disponibles.

Intervalos de confianza (última entrega)

```
resultado = t.test(M1, conf.level = 0.95)
resultado

##
## One Sample t-test
##
## data: M1
## t = 30.168, df = 879, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
```

```
## 30.65856 34.92532
## sample estimates:
## mean of x
## 32.79194

a = 0.03
confint(A, level= 1-a)

##              1.5 %      98.5 %
## (Intercept) -91.20451 -58.304689
## M$Estatura   79.32465  99.196052
## M$SexoM      -11.93983  -9.189113

lp = predict(object=A, interval="prediction", level = 0.97)

## Warning in predict.lm(object = A, interval = "prediction", level = 0.97):
## predictions on current data refer to _future_ responses

lp

##           fit      lwr      upr
## 1  68.95457 57.20540 80.70374
## 2  68.95457 57.20540 80.70374
## 3  76.98800 65.23787 88.73813
## 4  72.52498 60.78380 84.26617
## 5  78.77321 67.01363 90.53279
## 6  70.73978 58.99628 82.48327
## 7  82.34362 70.55511 94.13214
## 8  74.31019 62.56795 86.05243
## 9  74.31019 62.56795 86.05243
## 10 72.52498 60.78380 84.26617
## 11 70.73978 58.99628 82.48327
## 12 76.98800 65.23787 88.73813
## 13 76.09540 64.34874 87.84206
## 14 67.16936 55.41117 78.92755
## 15 77.88061 66.12617 89.63504
## 16 73.41759 61.67629 85.15888
## 17 72.52498 60.78380 84.26617
## 18 67.16936 55.41117 78.92755
## 19 67.16936 55.41117 78.92755
## 20 74.31019 62.56795 86.05243
## 21 77.88061 66.12617 89.63504
## 22 75.20279 63.45876 86.94683
## 23 67.16936 55.41117 78.92755
## 24 76.98800 65.23787 88.73813
## 25 75.20279 63.45876 86.94683
## 26 68.95457 57.20540 80.70374
## 27 76.98800 65.23787 88.73813
## 28 76.98800 65.23787 88.73813
## 29 70.73978 58.99628 82.48327
## 30 78.77321 67.01363 90.53279
```

##	31	69.84717	58.10126	81.59308
##	32	76.09540	64.34874	87.84206
##	33	66.27676	54.51280	78.04072
##	34	75.20279	63.45876	86.94683
##	35	69.84717	58.10126	81.59308
##	36	72.52498	60.78380	84.26617
##	37	66.27676	54.51280	78.04072
##	38	75.20279	63.45876	86.94683
##	39	71.63238	59.89046	83.37430
##	40	80.55842	68.78604	92.33079
##	41	70.73978	58.99628	82.48327
##	42	68.06197	56.30871	79.81523
##	43	68.95457	57.20540	80.70374
##	44	72.52498	60.78380	84.26617
##	45	76.09540	64.34874	87.84206
##	46	74.31019	62.56795	86.05243
##	47	68.95457	57.20540	80.70374
##	48	64.49155	52.71355	76.26955
##	49	70.73978	58.99628	82.48327
##	50	67.16936	55.41117	78.92755
##	51	66.27676	54.51280	78.04072
##	52	73.41759	61.67629	85.15888
##	53	63.59895	51.81267	75.38522
##	54	77.88061	66.12617	89.63504
##	55	72.52498	60.78380	84.26617
##	56	72.52498	60.78380	84.26617
##	57	77.88061	66.12617	89.63504
##	58	85.91404	74.08330	97.74477
##	59	67.16936	55.41117	78.92755
##	60	64.49155	52.71355	76.26955
##	61	64.49155	52.71355	76.26955
##	62	77.88061	66.12617	89.63504
##	63	83.23623	71.43840	95.03405
##	64	78.77321	67.01363	90.53279
##	65	75.20279	63.45876	86.94683
##	66	71.63238	59.89046	83.37430
##	67	77.88061	66.12617	89.63504
##	68	77.88061	66.12617	89.63504
##	69	63.59895	51.81267	75.38522
##	70	77.88061	66.12617	89.63504
##	71	70.73978	58.99628	82.48327
##	72	73.41759	61.67629	85.15888
##	73	70.73978	58.99628	82.48327
##	74	80.55842	68.78604	92.33079
##	75	62.70635	50.91097	74.50172
##	76	73.41759	61.67629	85.15888
##	77	75.20279	63.45876	86.94683
##	78	77.88061	66.12617	89.63504
##	79	76.98800	65.23787	88.73813
##	80	67.16936	55.41117	78.92755

## 81	64.49155	52.71355	76.26955
## 82	71.63238	59.89046	83.37430
## 83	81.45102	69.67099	93.23105
## 84	83.23623	71.43840	95.03405
## 85	76.98800	65.23787	88.73813
## 86	68.06197	56.30871	79.81523
## 87	62.70635	50.91097	74.50172
## 88	75.20279	63.45876	86.94683
## 89	67.16936	55.41117	78.92755
## 90	72.52498	60.78380	84.26617
## 91	76.98800	65.23787	88.73813
## 92	70.73978	58.99628	82.48327
## 93	77.88061	66.12617	89.63504
## 94	77.88061	66.12617	89.63504
## 95	68.95457	57.20540	80.70374
## 96	76.09540	64.34874	87.84206
## 97	78.77321	67.01363	90.53279
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## 102	77.88061	66.12617	89.63504
## 103	72.52498	60.78380	84.26617
## 104	74.31019	62.56795	86.05243
## 105	69.84717	58.10126	81.59308
## 106	71.63238	59.89046	83.37430
## 107	80.55842	68.78604	92.33079
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## 112	69.84717	58.10126	81.59308
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## 114	73.41759	61.67629	85.15888
## 115	68.95457	57.20540	80.70374
## 116	75.20279	63.45876	86.94683
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## 118	77.88061	66.12617	89.63504
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## 122	69.84717	58.10126	81.59308
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## 125	77.88061	66.12617	89.63504
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## 128	75.20279	63.45876	86.94683
## 129	64.49155	52.71355	76.26955
## 130	63.59895	51.81267	75.38522

131 69.84717 58.10126 81.59308
132 74.31019 62.56795 86.05243
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241 61.96051 50.19446 73.72657
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262 59.28270 47.53227 71.03313
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264 54.81969 43.07852 66.56085
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270 54.81969 43.07852 66.56085
271 46.78625 35.00886 58.56365
272 59.28270 47.53227 71.03313
273 51.24927 39.50039 62.99815
274 53.92708 42.18525 65.66892
275 54.81969 43.07852 66.56085
276 61.06791 49.30790 72.82792
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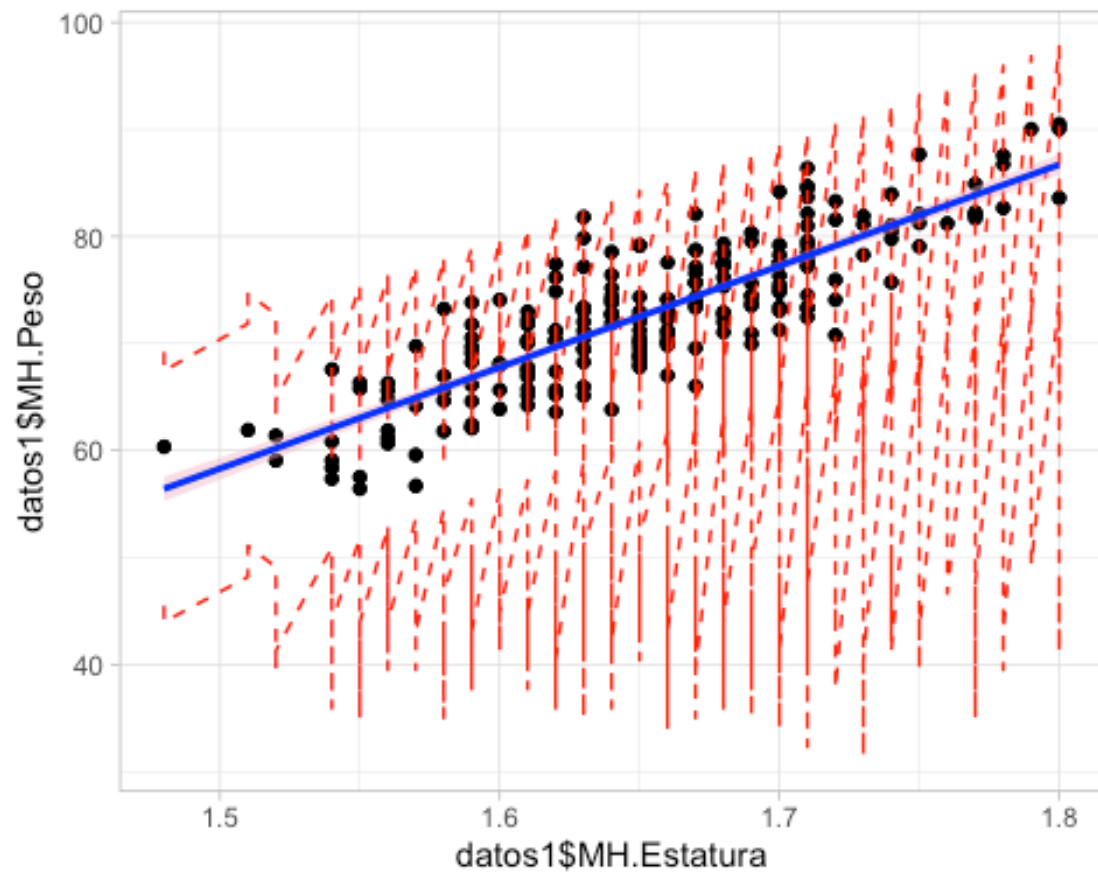
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303 51.24927 39.50039 62.99815
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306 61.06791 49.30790 72.82792
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309 52.14187 40.39618 63.88757
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364 55.71229 43.97095 67.45363
365 58.39010 46.64320 70.13700
366 51.24927 39.50039 62.99815
367 53.03448 41.29113 64.77782
368 53.03448 41.29113 64.77782
369 53.92708 42.18525 65.66892
370 55.71229 43.97095 67.45363
371 53.03448 41.29113 64.77782
372 47.67886 35.90883 59.44888
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374 57.49750 45.75329 69.24170
375 57.49750 45.75329 69.24170
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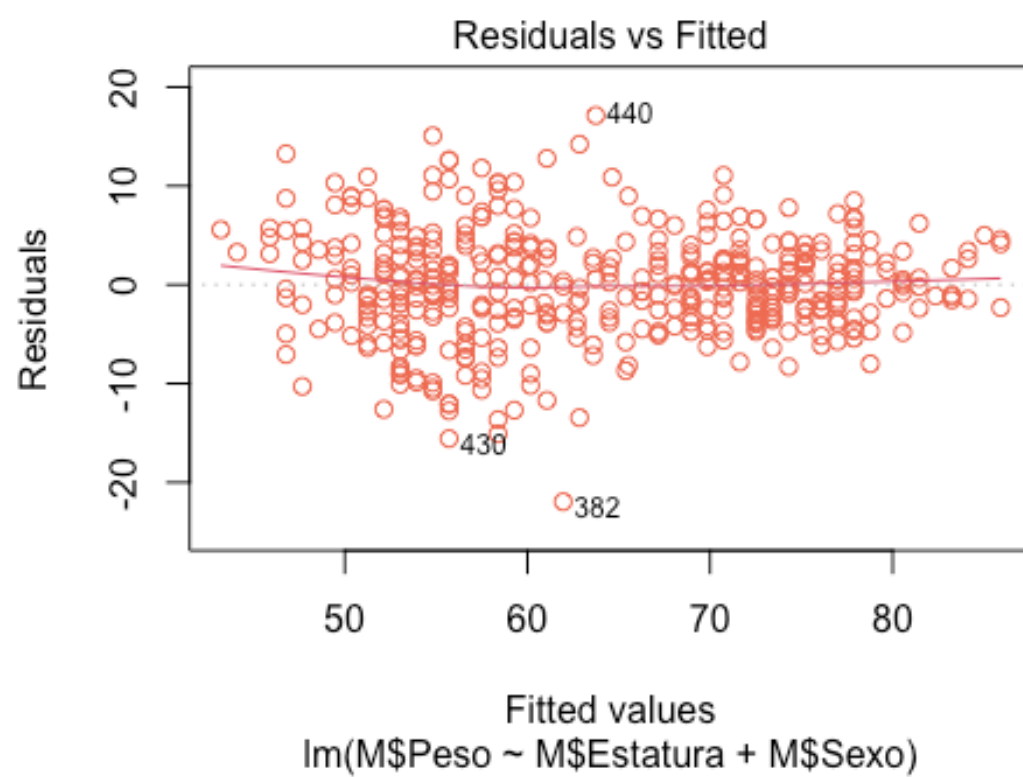
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410 49.46406 37.70628 61.22184
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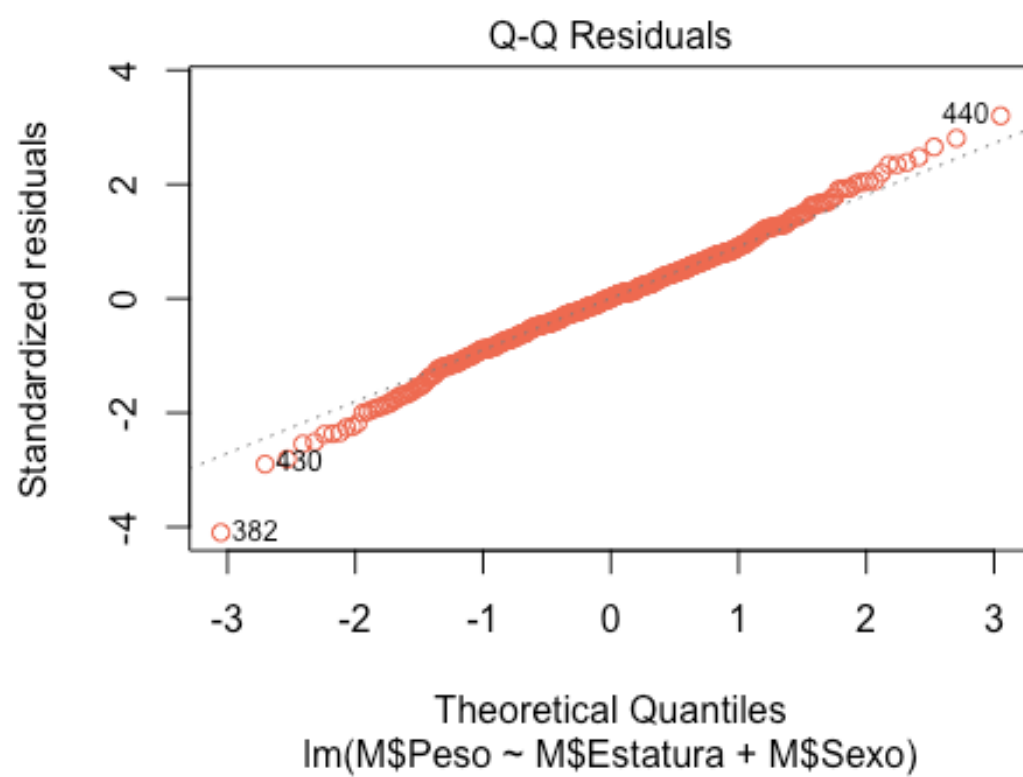
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## 435 61.06791 49.30790 72.82792
## 436 55.71229 43.97095 67.45363
## 437 54.81969 43.07852 66.56085
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```

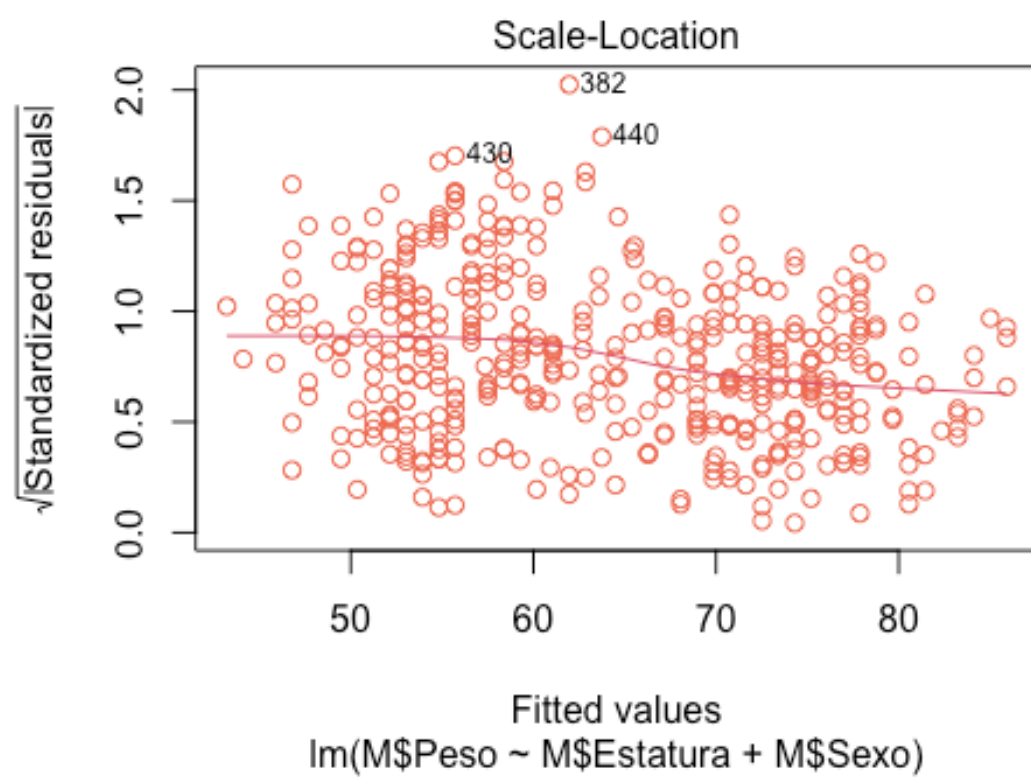
```
datos1 = cbind(M1, lp)
ggplot(datos1, aes(x = datos1$MH.Estatura, y = datos1$MH.Peso)) +
  geom_point() +
  geom_line(aes(y = lwr ), color = "red", linetype = "dashed") +
  geom_line(aes(y = upr), color = "red", linetype = "dashed") +
  geom_smooth(
    method = lm,
    formula = y ~ x,
    se = TRUE,
    level = 0.97,
    col = "blue",
    fill = "pink2"
  ) +
  theme_light()
```

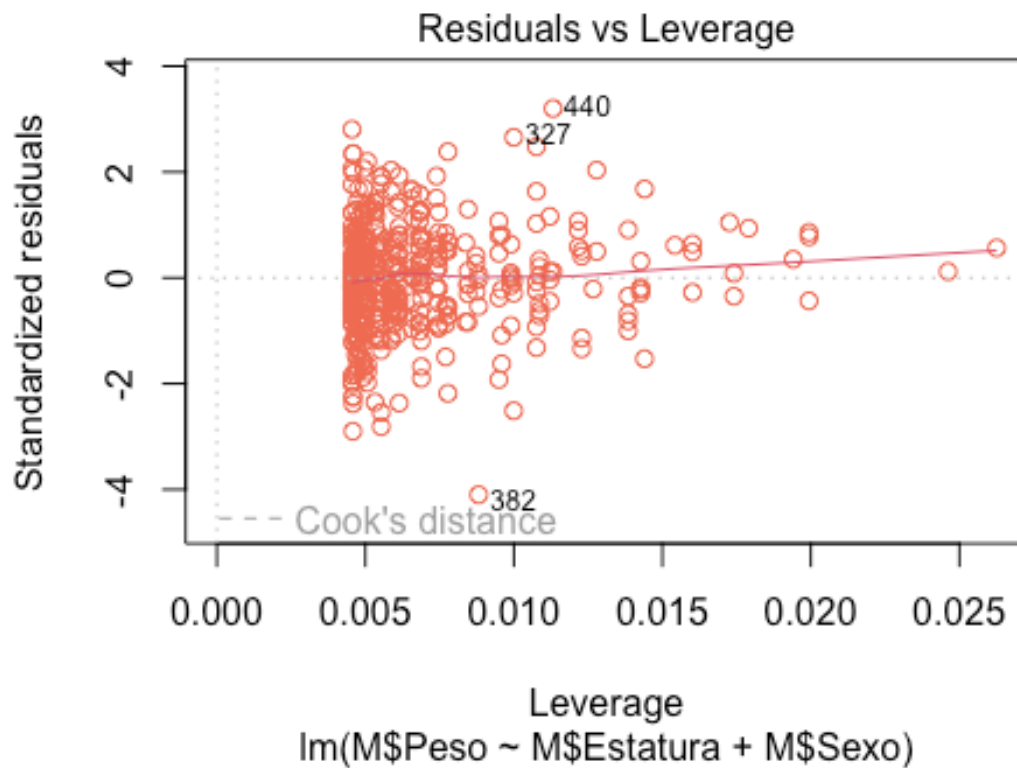


```
plot(A, col='coral2')
```







Cuáles

son las diferencias y similitudes de estos gráficos con respecto a los que ya habías analizado? Estos gráficos, ¿cambian en algo las conclusiones que ya habías obtenido?

Los gráficos son muy similares a los que ya había realizado. Particularmente los gráficos de Q-Q y el gráfico de residuales se observa que se asemejan a los que se hicieron con código en el ejercicio. No obstante, se pueden ver pequeñas diferencias. Por ejemplo, la línea de los residuales está un poco más curveada que la realizada sin la función "plot()". Además, la función arroja la gráfica de residuales con Leverage la cual no hicimos en esta tarea y aporta herramientas para el análisis del modelo. No obstante, los resultados no afectan las conclusiones anteriores.