

Regressão Múltipla

Load data

```
boston = readxl::read_excel("boston.xlsx")
b = boston#[,-1]
#b = b[-c(365),]
```

Manipulacao inicial dos dados

Adicionando labels

```
b$chas=as.factor(b$chas)
levels(b$chas)=c("otherwise", "bounds river")
```

Sumario dos dados

```
summary(b)
```

```
##          id          crim          zn          indus
## Min.      : 1.0    Min.      : 0.00632  Min.      : 0.00    Min.      : 0.46
## 1st Qu.:127.2    1st Qu.: 0.08204  1st Qu.: 0.00    1st Qu.: 5.19
## Median :253.5    Median : 0.25651  Median : 0.00    Median : 9.69
## Mean      :253.5    Mean      : 3.61352  Mean      : 11.36   Mean      :11.14
## 3rd Qu.:379.8    3rd Qu.: 3.67708  3rd Qu.: 12.50   3rd Qu.:18.10
## Max.      :506.0    Max.      :88.97620  Max.      :100.00  Max.      :27.74
##          chas          nox          rm          age
## otherwise   :471    Min.      :0.3850  Min.      :3.561  Min.      : 2.90
## bounds river: 35    1st Qu.:0.4490  1st Qu.:5.886  1st Qu.: 45.02
##              Median :0.5380  Median :6.208  Median : 77.50
##              Mean      :0.5547  Mean      :6.285  Mean      : 68.57
##              3rd Qu.:0.6240  3rd Qu.:6.623  3rd Qu.: 94.08
##              Max.      :0.8710  Max.      :8.780  Max.      :100.00
##          dis          rad          tax          ptratio
## Min.      : 1.130    Min.      : 1.000  Min.      :187.0  Min.      :12.60
## 1st Qu.: 2.100    1st Qu.: 4.000  1st Qu.:279.0  1st Qu.:17.40
## Median : 3.207    Median : 5.000  Median :330.0  Median :19.05
## Mean      : 3.795    Mean      : 9.549  Mean      :408.2  Mean      :18.46
## 3rd Qu.: 5.188    3rd Qu.:24.000  3rd Qu.:666.0  3rd Qu.:20.20
## Max.      :12.127    Max.      :24.000  Max.      :711.0  Max.      :22.00
##          lstat          medv
## Min.      : 1.73    Min.      : 5.00
## 1st Qu.: 6.95    1st Qu.:17.02
## Median :11.36    Median :21.20
## Mean      :12.65    Mean      :22.53
## 3rd Qu.:16.95    3rd Qu.:25.00
## Max.      :37.97    Max.      :50.00
```

Filtrando dados

```
b = filter(b, medv < 50)

#b$dis = ifelse(b$dis >= 3, 3, b$dis)
b$rad = ifelse(b$rad >= 9, 9, b$rad)
b$tax = ifelse(b$tax >= 500, 500, b$tax)
#b$nox = ifelse(b$nox >= 0.8, 0.75, b$nox)
#b$rm = ifelse(b$rm >= 7.5, 7.5, b$rm)

#b = filter(b, id != 366)
```

Ajustando valores

```
b = filter(b, medv < 50)
```

Transformando log

```
#b$crim = log(b$crim)
b$lstat = log(b$lstat)
b$dis = log(b$dis)
#b$medv = log(b$medv)
```

Regressao inicial

Fazendo Regressão com todas as variaveis

```
reg.mlt=lm(data=b, medv ~ crim + zn + indus + chas + nox + rm + age + dis +
           rad + tax + ptratio + lstat)

summary(reg.mlt)
```

```
##
## Call:
## lm(formula = medv ~ crim + zn + indus + chas + nox + rm + age +
##      dis + rad + tax + ptratio + lstat, data = b)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.764  -2.142  -0.445   1.824  11.331
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   52.44904    3.898524  13.454 < 2e-16 ***
## crim         -0.140924    0.023190  -6.077 2.51e-09 ***
## zn            0.011309    0.009855   1.148  0.252
## indus        -0.044448    0.045653  -0.974  0.331
## chasbounds river  0.688448    0.690888   0.996  0.320
## nox          -14.966018    2.939639  -5.091 5.13e-07 ***
```

```
## rm          2.761520    0.344175    8.024 8.04e-15 ***
## age         -0.003908    0.010400   -0.376  0.707
## dis         -4.975750    0.671793   -7.407 5.92e-13 ***
## rad          0.516139    0.104161    4.955 1.01e-06 ***
## tax         -0.015118    0.003071   -4.923 1.18e-06 ***
## ptratio     -0.730224    0.094220   -7.750 5.58e-14 ***
## lstat       -6.895288    0.531185  -12.981 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.494 on 477 degrees of freedom
## Multiple R-squared:  0.8075, Adjusted R-squared:  0.8026
## F-statistic: 166.7 on 12 and 477 DF,  p-value: < 2.2e-16
```

Testando multicolinearidade

VIF > 5 indica alta chance de multicolinearidade.

```
round(vif(reg.mlt),1)
```

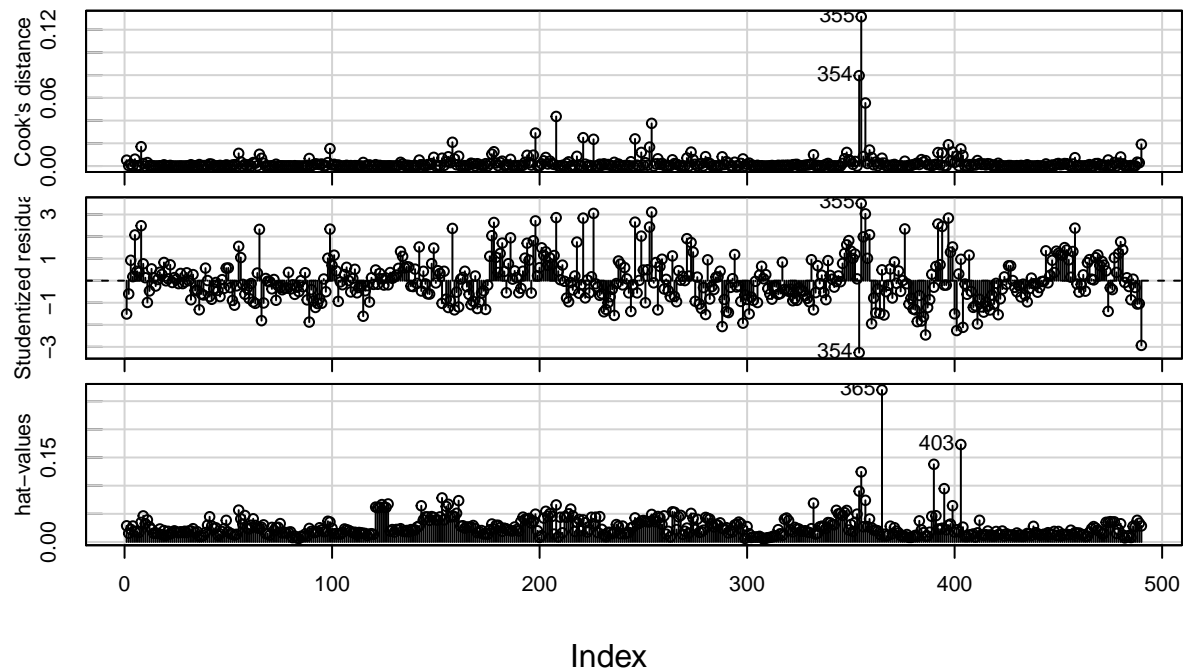
```
##          crim          zn          indus chasbounds river
##          1.6          2.0          3.9          1.1
##          nox          rm          age          dis
##          4.7          2.0          3.4          5.2
##          rad          tax          ptratio          lstat
##          2.6          3.9          1.6          3.7
```

Detecção de anomalias

Cooks Distances -> Pontos Influentes Studentized residuals -> Outliers em Y hat-values -> Outliers em X

```
influenceIndexPlot(reg.mlt , vars=c("Cook","Studentized","hat"))
```

Diagnostic Plots



Regressao com seleção de variáveis

```
reg.mlt2=step(reg.mlt)
```

```
## Start:  AIC=1238.98
## medv ~ crim + zn + indus + chas + nox + rm + age + dis + rad +
##      tax + ptratio + lstat
##
##           Df Sum of Sq  RSS   AIC
## - age      1      1.72 5826.4 1237.1
## - indus    1     11.57 5836.3 1238.0
## - chas     1     12.13 5836.8 1238.0
## - zn       1     16.08 5840.8 1238.3
## <none>             5824.7 1239.0
## - tax      1    295.90 6120.6 1261.3
## - rad      1    299.83 6124.6 1261.6
## - nox      1    316.51 6141.2 1262.9
## - crim     1    450.95 6275.7 1273.5
## - dis      1    669.89 6494.6 1290.3
## - ptratio  1    733.47 6558.2 1295.1
## - rm       1    786.13 6610.8 1299.0
## - lstat    1   2057.64 7882.4 1385.2
##
## Step:  AIC=1237.12
## medv ~ crim + zn + indus + chas + nox + rm + dis + rad + tax +
##      ptratio + lstat
##
##           Df Sum of Sq  RSS   AIC
## - indus    1     11.07 5837.5 1236.0
```

```

## - chas      1      11.84 5838.3 1236.1
## - zn        1      17.39 5843.8 1236.6
## <none>                5826.4 1237.1
## - tax       1      294.37 6120.8 1259.3
## - rad       1      301.96 6128.4 1259.9
## - nox       1      336.84 6163.3 1262.7
## - crim      1      449.24 6275.7 1271.5
## - ptratio   1      740.55 6567.0 1293.8
## - dis       1      740.68 6567.1 1293.8
## - rm        1      828.74 6655.2 1300.3
## - lstat     1     2603.44 8429.9 1416.1
##
## Step:  AIC=1236.05
## medv ~ crim + zn + chas + nox + rm + dis + rad + tax + ptratio +
##      lstat
##
##           Df Sum of Sq   RSS   AIC
## - chas     1      10.31 5847.8 1234.9
## - zn       1      20.63 5858.1 1235.8
## <none>                5837.5 1236.0
## - rad      1     330.18 6167.7 1261.0
## - nox      1     383.80 6221.3 1265.2
## - tax      1     388.43 6225.9 1265.6
## - crim     1     442.40 6279.9 1269.8
## - dis      1     744.80 6582.3 1292.9
## - ptratio  1     795.62 6633.1 1296.7
## - rm       1     848.48 6686.0 1300.5
## - lstat    1    2669.94 8507.5 1418.6
##
## Step:  AIC=1234.92
## medv ~ crim + zn + nox + rm + dis + rad + tax + ptratio + lstat
##
##           Df Sum of Sq   RSS   AIC
## - zn       1      20.14 5868.0 1234.6
## <none>                5847.8 1234.9
## - rad      1     344.49 6192.3 1261.0
## - nox      1     375.18 6223.0 1263.4
## - tax      1     413.69 6261.5 1266.4
## - crim     1     452.99 6300.8 1269.5
## - dis      1     744.58 6592.4 1291.6
## - ptratio  1     815.92 6663.7 1296.9
## - rm       1     855.49 6703.3 1299.8
## - lstat    1    2663.57 8511.4 1416.8
##
## Step:  AIC=1234.6
## medv ~ crim + nox + rm + dis + rad + tax + ptratio + lstat
##
##           Df Sum of Sq   RSS   AIC
## <none>                5868.0 1234.6
## - rad      1     326.38 6194.3 1259.1
## - nox      1     392.34 6260.3 1264.3
## - tax      1     395.85 6263.8 1264.6
## - crim     1     433.55 6301.5 1267.5
## - dis      1     757.11 6625.1 1292.1

```

```
## - rm      1      870.67 6738.6 1300.4
## - ptratio 1      1001.80 6869.8 1309.8
## - lstat   1      2770.00 8638.0 1422.1
```

Novo sumario da regressao

```
summary(reg.mlt2)
```

```
##
## Call:
## lm(formula = medv ~ crim + nox + rm + dis + rad + tax + ptratio +
##      lstat, data = b)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.2445  -2.2169  -0.3962   1.8000  11.3857
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  53.285115   3.851023   13.837 < 2e-16 ***
## crim        -0.135371   0.022708   -5.961 4.84e-09 ***
## nox         -15.828758   2.791156   -5.671 2.45e-08 ***
## rm           2.780929   0.329181    8.448 3.54e-16 ***
## dis          -4.449171   0.564771   -7.878 2.24e-14 ***
## rad           0.521068   0.100740    5.172 3.40e-07 ***
## tax          -0.015417   0.002707   -5.696 2.13e-08 ***
## ptratio      -0.792582   0.087463   -9.062 < 2e-16 ***
## lstat        -7.099162   0.471127  -15.068 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.493 on 481 degrees of freedom
## Multiple R-squared:  0.806, Adjusted R-squared:  0.8028
## F-statistic: 249.8 on 8 and 481 DF, p-value: < 2.2e-16
```

Nova deteccao de multicolinearidade

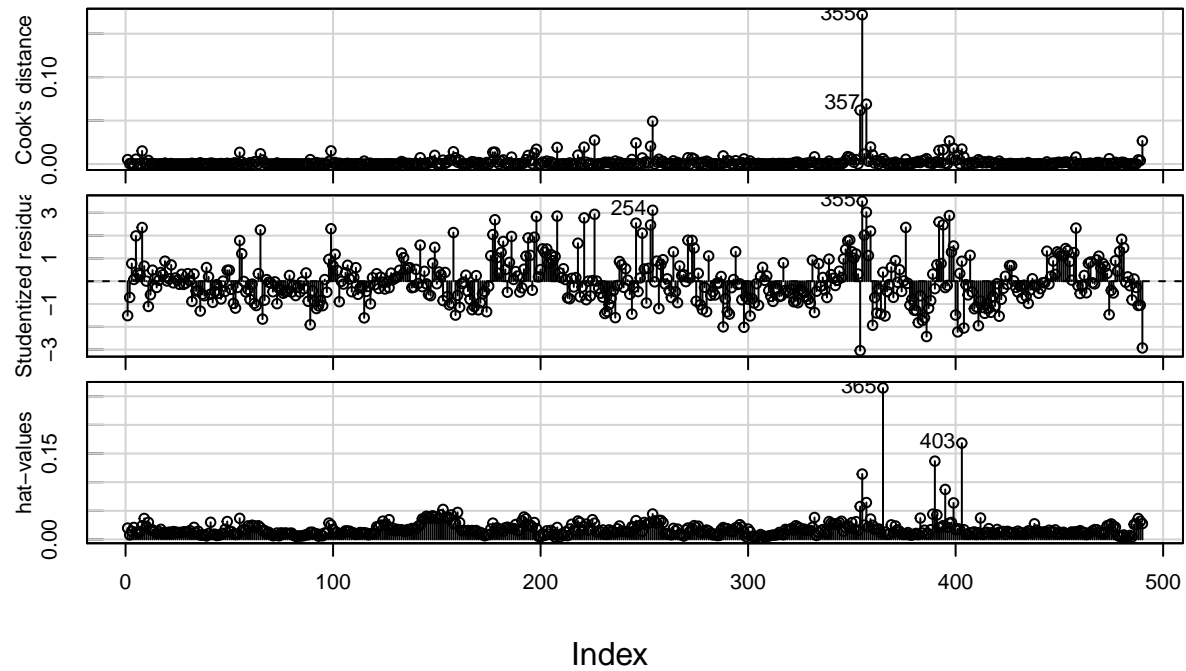
```
round(vif(reg.mlt2),1)
```

```
##      crim      nox      rm      dis      rad      tax ptratio  lstat
##      1.6      4.3      1.9      3.7      2.4      3.0      1.4      2.9
```

Novas anomalias

```
influenceIndexPlot(reg.mlt2 , vars=c("Cook","Studentized","hat"))
```

Diagnostic Plots

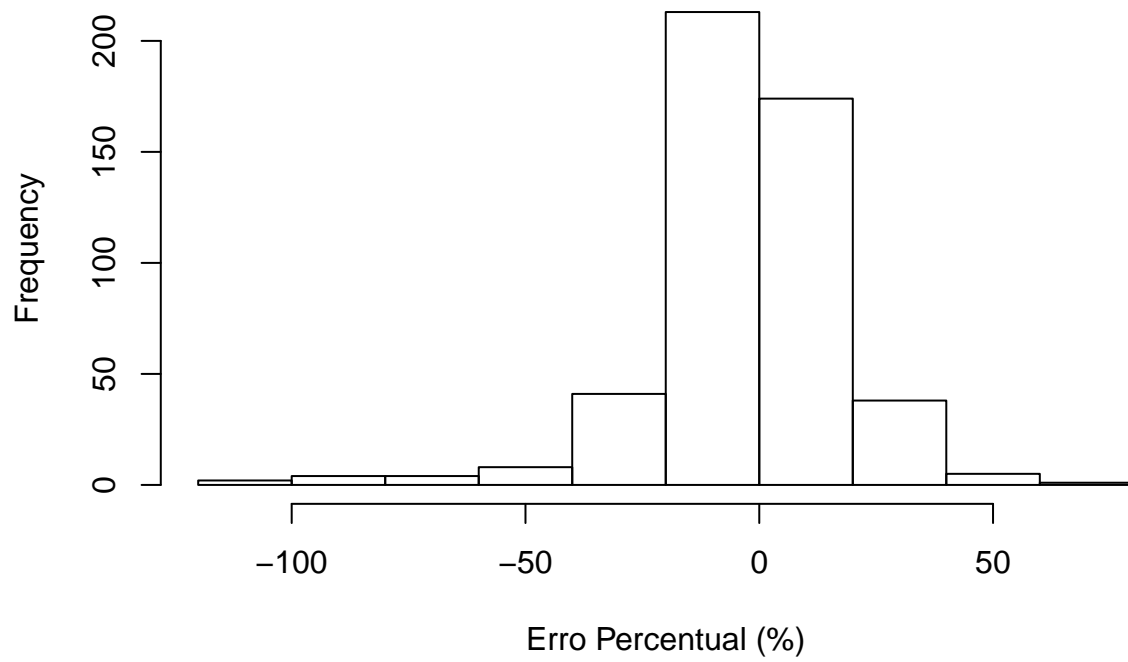


Criar previsoes

```
b$medv_HAT=fitted.values(reg.mlt2) #Previsoes
b$RES=residuals(reg.mlt2) #Residuais das previsoes
b$EP=b$RES/b$medv*100 #Erro percentual das previsoes
```

Erro percentual

```
hist(b$EP, xlab = 'Erro Percentual (%)', main = '')
```



Previsao e real

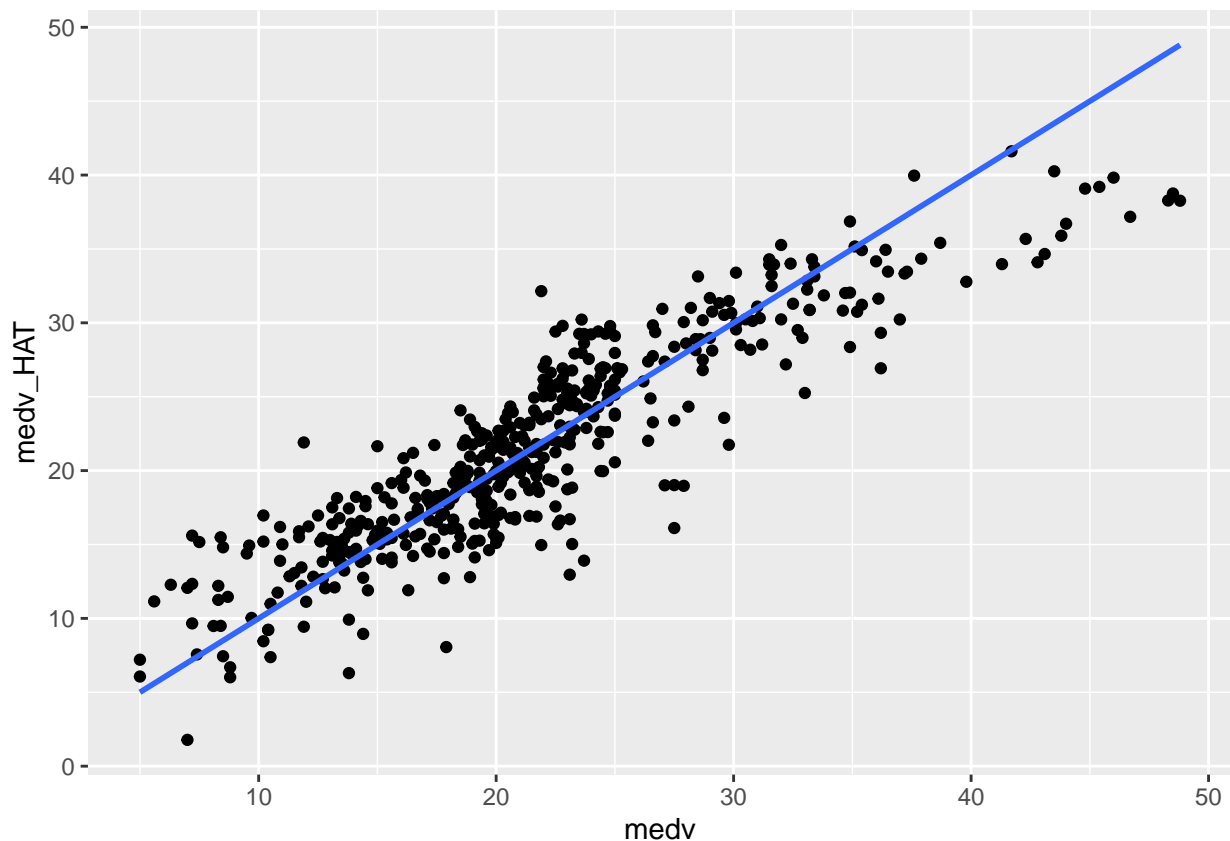
```
#plot(x = b$medv, y = b$medv_HAT, xlab = 'Preco Real', ylab = 'Previsao')
```

```
p = ggplot(b) +
  geom_point(aes(x = medv, y = medv_HAT,
                 name = id #rownames(b)
                )) +
  geom_smooth(method='lm', aes(x = medv, y = medv) )
```

```
## Warning: Ignoring unknown aesthetics: name
```

```
#ggplotly(p)
```

```
p
```

Teste anova

```
anova(reg.mlt2)
```

```
## Analysis of Variance Table
##
## Response: medv
##          Df Sum Sq Mean Sq  F value    Pr(>F)
## crim      1  6129.0   6129.0  502.3943 < 2.2e-16 ***
## nox       1  4128.7   4128.7  338.4309 < 2.2e-16 ***
## rm        1 8322.5   8322.5  682.2021 < 2.2e-16 ***
## dis       1   171.9    171.9   14.0905 0.0001955 ***
## rad       1    54.8     54.8    4.4911 0.0345843 *
## tax       1  1038.3   1038.3   85.1134 < 2.2e-16 ***
## ptratio   1  1767.8   1767.8  144.9093 < 2.2e-16 ***
## lstat     1  2770.0   2770.0  227.0586 < 2.2e-16 ***
## Residuals 481 5868.0    12.2
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Root mean squared error

```
mean((b$medv - b$medv_HAT) ** 2) **0.5
```

```
## [1] 3.460554
```