Preventing Hospital Expenses

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For this analysis, we will use a data set simulating hypothetical medical expenses for a set of patients spread across 4 regions of Brazil. This dataset has 1,338 observations and 7 variables.

Step 1 - Data gathering

```
# Data gathering
df <- read.csv('C:\\Users\\Oracy\\Desktop\\DSA_Projetos\\DSA_Projetos\\Big Data Analytics com R e Micro head(df)

## idade sexo bmi filhos fumante regiao gastos
## 1 19 mulher 27.9 0 sim sudeste 16884.92</pre>
```

```
## 1
       19 mulher 27.9
                       0 sim
                                     sudeste 16884.92
## 2
       18 homem 33.8
                          1
                                         sul 1725.55
                                nao
## 3
       28 homem 33.0
                          3
                                             4449.46
                                nao
                                         sul
                           0
## 4
       33 homem 22.7
                                nao nordeste 21984.47
## 5
       32 homem 28.9
                           0
                                nao nordeste 3866.86
## 6
       31 mulher 25.7
                           0
                                         sul 3756.62
                                nao
```

Etapa 2 - Explorando os Dados

```
# Viewing variables
str(df)

## 'data.frame': 1338 obs. of 7 variables:
## $ idade : int 19 18 28 33 32 31 46 37 37 60 ...
## $ sexo : Factor w/ 2 levels "homem", "mulher": 2 1 1 1 1 2 2 2 1 2 ...
## $ bmi : num 27.9 33.8 33 22.7 28.9 25.7 33.4 27.7 29.8 25.8 ...
## $ filhos : int 0 1 3 0 0 0 1 3 2 0 ...
## $ fumante: Factor w/ 2 levels "nao", "sim": 2 1 1 1 1 1 1 1 1 1 ...
## $ regiao : Factor w/ 4 levels "nordeste", "norte", ...: 3 4 4 1 1 4 4 1 2 1 ...
## $ gastos : num 16885 1726 4449 21984 3867 ...
## Central Trend Averages of the variable spending
summary(df[c("gastos")])
```

```
## gastos

## Min. : 1122

## 1st Qu.: 4740

## Median : 9382

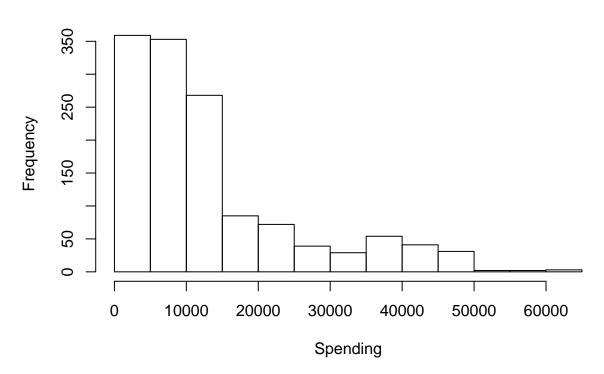
## Mean :13270

## 3rd Qu.:16640

## Max. :63770
```

```
# Building a Histogram
hist(df$gastos, main = 'Histogram', xlab = 'Spending')
```

Histogram



```
# Regions contingency table
table(df$regiao)
##
## nordeste
                       sudeste
                                     sul
               norte
        325
                           325
                                     364
                  324
{\it \# Exploring relation ships among variables: Correlation \ {\it Matrix}}
cor(df[c("idade","bmi", "filhos", "gastos")])
##
              idade
                                     filhos
                            bmi
                                                gastos
## idade 1.0000000 0.10934101 0.04246900 0.29900819
          0.1093410\ 1.00000000\ 0.01264471\ 0.19857626
## filhos 0.0424690 0.01264471 1.00000000 0.06799823
## gastos 0.2990082 0.19857626 0.06799823 1.00000000
# None of the correlations in the matrix are considered strong, but there are some interesting associat
```

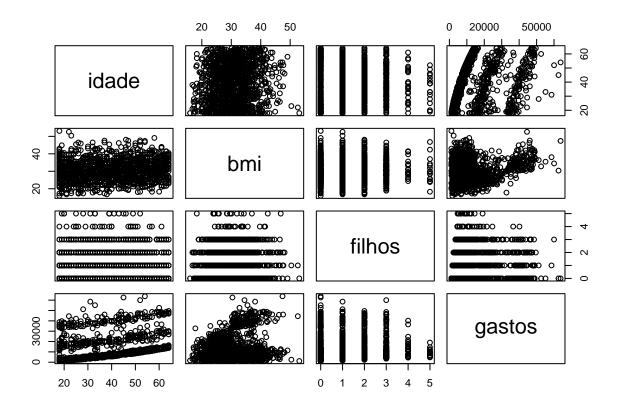
For example, age and bmi (BMI) appear to have a weak positive correlation, which means that

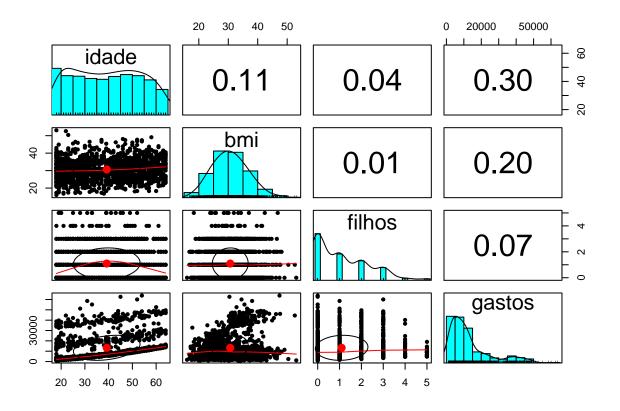
Moderate between age and expenditure, in addition to the number of children and expenses. These assoc # that as the average age, body mass and number of children increases, the expected cost of health insu

As age increases, body mass tends to increase. There is also a positive correlation

```
# Viewing relationship between variables: Scatterplot
# Note that there is no clear relationship between the variables
pairs(df[c("idade", "bmi", "filhos", "gastos")])

# Scatterplot Matrix
# Font: http://www.sthda.com/english/wiki/scatter-plot-matrices-r-base-graphs#use-the-r-package-psych
#install.packages ("psych")
library(psych)
```





This graphic provides more information about the relationship between variables

Step 3: Training the Model

```
# Font: https://www.rdocumentation.org/packages/stats/versions/3.5.1/topics/lm
str(df)
## 'data.frame':
                    1338 obs. of 7 variables:
## $ idade : int 19 18 28 33 32 31 46 37 37 60 ...
           : Factor w/ 2 levels "homem", "mulher": 2 1 1 1 1 2 2 2 1 2 ...
            : num 27.9 33.8 33 22.7 28.9 25.7 33.4 27.7 29.8 25.8 ...
## $ filhos : int 0 1 3 0 0 0 1 3 2 0 ...
## $ fumante: Factor w/ 2 levels "nao", "sim": 2 1 1 1 1 1 1 1 1 1 ...
## $ regiao : Factor w/ 4 levels "nordeste", "norte",..: 3 4 4 1 1 4 4 1 2 1 ...
    $ gastos : num 16885 1726 4449 21984 3867 ...
model <- lm(gastos ~ idade + filhos + bmi + sexo + fumante + regiao, df)
# Similar to the previous item
model_2 <- lm(gastos ~ ., df) # "." is the same as type all variables</pre>
# Viewing the coefficients
# Font: https://stackoverflow.com/questions/6577058/extract-regression-coefficient-values
```

```
model_summary <- summary(model)</pre>
model_summary
##
## Call:
## lm(formula = gastos ~ idade + filhos + bmi + sexo + fumante +
      regiao, data = df)
##
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
                                           Max
## -11302.7 -2850.9
                     -979.6
                              1383.9 29981.7
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
                         1000.7 -12.418 < 2e-16 ***
## (Intercept)
                -12425.7
                              11.9 21.586 < 2e-16 ***
## idade
                   256.8
## filhos
                   475.7
                             137.8 3.452 0.000574 ***
                              28.6 11.864 < 2e-16 ***
## bmi
                   339.3
## sexomulher
                   131.3
                              332.9 0.395 0.693255
## fumantesim
                 23847.5
                            413.1 57.723 < 2e-16 ***
## regiaonorte
                  352.8
                            476.3 0.741 0.458976
                            477.2 -1.271 0.203940
## regiaosudeste
                  -606.5
## regiaosul
                  -682.8
                              478.9 -1.426 0.154211
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 6062 on 1329 degrees of freedom
## Multiple R-squared: 0.7509, Adjusted R-squared: 0.7494
## F-statistic: 500.9 on 8 and 1329 DF, p-value: < 2.2e-16
model
##
## Call:
## lm(formula = gastos ~ idade + filhos + bmi + sexo + fumante +
      regiao, data = df)
##
##
## Coefficients:
##
     (Intercept)
                         idade
                                       filhos
                                                         bmi
                                                                 sexomulher
##
       -12425.7
                         256.8
                                        475.7
                                                       339.3
                                                                      131.4
##
     fumantesim regiaonorte regiaosudeste
                                                   regiaosul
##
        23847.5
                         352.8
                                       -606.5
                                                      -682.8
# Preventing medical expenses
# Font: https://stat.ethz.ch/R-manual/R-devel/library/stats/html/predict.lm.html
predicting <- predict(model)</pre>
class(predicting)
```

[1] "numeric"

```
head(predicting)
```

```
## 1 2 3 4 5 6
## 25292.740 3458.281 6706.619 3751.868 5598.626 3704.606
```

Step 4: Evaluating Model Performance

```
# More details about the model
# Font: https://stat.ethz.ch/R-manual/R-devel/library/stats/html/lm.html
summary(model)
##
## Call:
## lm(formula = gastos ~ idade + filhos + bmi + sexo + fumante +
      regiao, data = df)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -11302.7 -2850.9
                      -979.6
                               1383.9
                                       29981.7
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                -12425.7
                             1000.7 -12.418 < 2e-16 ***
## idade
                   256.8
                               11.9 21.586 < 2e-16 ***
## filhos
                   475.7
                              137.8 3.452 0.000574 ***
## bmi
                   339.3
                               28.6 11.864 < 2e-16 ***
## sexomulher
                              332.9 0.395 0.693255
                   131.3
## fumantesim
                 23847.5
                              413.1 57.723 < 2e-16 ***
                  352.8
                              476.3
                                     0.741 0.458976
## regiaonorte
## regiaosudeste
                  -606.5
                              477.2 -1.271 0.203940
## regiaosul
                  -682.8
                              478.9 -1.426 0.154211
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 6062 on 1329 degrees of freedom
## Multiple R-squared: 0.7509, Adjusted R-squared: 0.7494
## F-statistic: 500.9 on 8 and 1329 DF, p-value: < 2.2e-16
```

Step 5: Optimizing Model Performance

```
# Adding a variable with twice the age value
df$idade2 <- df$idade * 2
#df$idade2
#df$idade

# Adding a Bookmark to BMI> = 30
# Font: https://www.datamentor.io/r-programming/ifelse-function/
df$bmi30 <- ifelse(df$bmi >= 30, 1, 0)
#df$bmi30
```

```
# Creating the final template
str(df)
## 'data.frame':
                   1338 obs. of 9 variables:
## $ idade : int 19 18 28 33 32 31 46 37 37 60 ...
## $ sexo : Factor w/ 2 levels "homem", "mulher": 2 1 1 1 1 2 2 2 1 2 ...
           : num 27.9 33.8 33 22.7 28.9 25.7 33.4 27.7 29.8 25.8 ...
## $ bmi
## $ filhos : int 0 1 3 0 0 0 1 3 2 0 ...
## $ fumante: Factor w/ 2 levels "nao", "sim": 2 1 1 1 1 1 1 1 1 1 ...
## $ regiao : Factor w/ 4 levels "nordeste", "norte", ...: 3 4 4 1 1 4 4 1 2 1 ...
## $ gastos : num 16885 1726 4449 21984 3867 ...
## $ idade2 : num 38 36 56 66 64 62 92 74 74 120 ...
## $ bmi30 : num 0 1 1 0 0 0 1 0 0 0 ...
model_3 <- lm(gastos ~ idade + idade2 + sexo + filhos + bmi30 * fumante + regiao, df)
summary(model_3)
##
## Call:
## lm(formula = gastos ~ idade + idade2 + sexo + filhos + bmi30 *
      fumante + regiao, data = df)
##
##
## Residuals:
       Min
                 1Q Median
                                   30
## -18829.3 -1872.4 -1306.7 -582.2 24710.6
## Coefficients: (1 not defined because of singularities)
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   -2713.616
                                470.681 -5.765 1.01e-08 ***
## idade
                     265.486
                                  8.812 30.126 < 2e-16 ***
## idade2
                          NA
                                     NA
                                            NA
## sexomulher
                     479.924
                                247.474
                                         1.939
                                                  0.0527 .
## filhos
                     524.041
                                102.338
                                          5.121 3.49e-07 ***
## bmi30
                     201.748
                                281.354
                                         0.717
                                                 0.4735
## fumantesim
                   13360.059
                                445.408 29.995 < 2e-16 ***
## regiaonorte
                     278.942
                                353.726
                                         0.789
                                                 0.4305
## regiaosudeste
                    -881.758
                                353.916 -2.491
                                                 0.0128 *
## regiaosul
                    -308.196
                                348.633 -0.884
                                                 0.3768
## bmi30:fumantesim 19856.483
                                612.121 32.439 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 4502 on 1328 degrees of freedom
## Multiple R-squared: 0.8627, Adjusted R-squared: 0.8618
## F-statistic: 927.4 on 9 and 1328 DF, p-value: < 2.2e-16
```

Fim