

**EQUIPE: 23**

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#

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```
# instalar o pacotes necessários
install.packages("mlbench", repos = "http://cran.us.r-project.org")
```

### 1.1 Carregue a base de dados Satellite

```
##
## The downloaded binary packages are in
## /var/folders/7c/43lddvnj7bsfph_4fmdrns140000gp/T//Rtmpphp1xa/downloaded_packages
install.packages("e1017", repos = "http://cran.us.r-project.org")
```

```
## Warning: package 'e1017' is not available for this version of R
##
## A version of this package for your version of R might be available elsewhere,
## see the ideas at
## https://cran.r-project.org/doc/manuals/r-patched/R-admin.html#Installing-packages
install.packages("randomForest", repos = "http://cran.us.r-project.org")
```

```
##
## The downloaded binary packages are in
## /var/folders/7c/43lddvnj7bsfph_4fmdrns140000gp/T//Rtmpphp1xa/downloaded_packages
install.packages("kernlab", repos = "http://cran.us.r-project.org")
```

```
##
## The downloaded binary packages are in
## /var/folders/7c/43lddvnj7bsfph_4fmdrns140000gp/T//Rtmpphp1xa/downloaded_packages
install.packages("caret", repos = "http://cran.us.r-project.org")
```

```
##
## The downloaded binary packages are in
## /var/folders/7c/43lddvnj7bsfph_4fmdrns140000gp/T//Rtmpphp1xa/downloaded_packages
```

```

# usar os pacotes necessários
library(mlbench)
library(caret)

## Loading required package: ggplot2

## Loading required package: lattice

# carregar os dados Satellite
data(Satellite)

# exibir estrutura dos dados Satellite
str(Satellite)

## 'data.frame': 6435 obs. of 37 variables:
## $ x.1 : num 92 84 84 80 84 80 76 76 76 ...
## $ x.2 : num 115 102 102 102 94 94 102 102 89 94 ...
## $ x.3 : num 120 106 102 102 102 98 106 106 98 98 ...
## $ x.4 : num 94 79 83 79 79 76 83 87 76 76 ...
## $ x.5 : num 84 84 80 84 80 80 76 80 76 76 ...
## $ x.6 : num 102 102 102 94 94 102 102 98 94 98 ...
## $ x.7 : num 106 102 102 102 98 102 106 106 98 102 ...
## $ x.8 : num 79 83 79 79 76 79 87 79 76 72 ...
## $ x.9 : num 84 80 84 80 80 76 80 76 76 76 ...
## $ x.10 : num 102 102 94 94 102 102 98 94 98 94 ...
## $ x.11 : num 102 102 102 98 102 102 106 102 102 90 ...
## $ x.12 : num 83 79 79 76 79 79 79 76 72 76 ...
## $ x.13 : num 101 92 84 84 84 76 80 80 80 76 ...
## $ x.14 : num 126 112 103 99 99 99 107 112 95 91 ...
## $ x.15 : num 133 118 104 104 104 104 118 118 104 104 ...
## $ x.16 : num 103 85 81 78 81 81 88 88 74 74 ...
## $ x.17 : num 92 84 84 84 76 76 80 80 76 76 ...
## $ x.18 : num 112 103 99 99 99 99 112 107 91 95 ...
## $ x.19 : num 118 104 104 104 104 108 118 113 104 100 ...
## $ x.20 : num 85 81 78 81 81 85 88 85 74 78 ...
## $ x.21 : num 84 84 84 76 76 76 80 80 76 76 ...
## $ x.22 : num 103 99 99 99 99 103 107 95 95 91 ...
## $ x.23 : num 104 104 104 104 108 118 113 100 100 100 ...
## $ x.24 : num 81 78 81 81 85 88 85 78 78 74 ...
## $ x.25 : num 102 88 84 84 84 84 79 79 75 75 ...
## $ x.26 : num 126 121 107 99 99 103 107 103 91 91 ...
## $ x.27 : num 134 128 113 104 104 104 113 104 96 96 ...
## $ x.28 : num 104 100 87 79 79 79 87 83 75 71 ...
## $ x.29 : num 88 84 84 84 84 79 79 79 75 79 ...
## $ x.30 : num 121 107 99 99 103 107 103 103 91 87 ...
## $ x.31 : num 128 113 104 104 104 109 104 104 96 93 ...
## $ x.32 : num 100 87 79 79 79 87 83 79 71 71 ...
## $ x.33 : num 84 84 84 84 79 79 79 79 79 79 ...
## $ x.34 : num 107 99 99 103 107 107 103 95 87 87 ...
## $ x.35 : num 113 104 104 104 109 109 104 100 93 93 ...
## $ x.36 : num 87 79 79 79 87 87 79 79 71 67 ...
## $ classes: Factor w/ 6 levels "red soil","cotton crop",...: 3 3 3 3 3 3 3 3 4 4 ...

# apresentar alguma medidas estatísticas do dados Satellite
summary(Satellite)

```

##	x.1	x.2	x.3	x.4
##	Min. : 39.0	Min. : 27.00	Min. : 53.00	Min. : 33.00
##	1st Qu.: 60.0	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.00
##	Median : 68.0	Median : 87.00	Median :101.00	Median : 81.00
##	Mean : 69.4	Mean : 83.59	Mean : 99.29	Mean : 82.59
##	3rd Qu.: 80.0	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.00
##	Max. :104.0	Max. :137.00	Max. :140.00	Max. :154.00
##	x.5	x.6	x.7	x.8
##	Min. : 39.00	Min. : 27.00	Min. : 50.00	Min. : 29.0
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.0
##	Median : 68.00	Median : 85.00	Median :101.00	Median : 81.0
##	Mean : 69.15	Mean : 83.24	Mean : 99.11	Mean : 82.5
##	3rd Qu.: 80.00	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.0
##	Max. :104.00	Max. :137.00	Max. :145.00	Max. :157.0
##	x.9	x.10	x.11	x.12
##	Min. : 40.00	Min. : 27.00	Min. : 50.00	Min. : 29.00
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 68.00
##	Median : 67.00	Median : 85.00	Median :100.00	Median : 81.00
##	Mean : 68.91	Mean : 82.89	Mean : 98.85	Mean : 82.39
##	3rd Qu.: 79.00	3rd Qu.:102.00	3rd Qu.:113.00	3rd Qu.: 92.00
##	Max. :104.00	Max. :130.00	Max. :145.00	Max. :157.00
##	x.13	x.14	x.15	x.16
##	Min. : 39.00	Min. : 27.00	Min. : 50.00	Min. : 29.00
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.00
##	Median : 68.00	Median : 85.00	Median :101.00	Median : 81.00
##	Mean : 69.29	Mean : 83.48	Mean : 99.31	Mean : 82.64
##	3rd Qu.: 80.00	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.00
##	Max. :104.00	Max. :137.00	Max. :145.00	Max. :154.00
##	x.17	x.18	x.19	x.20
##	Min. : 40.00	Min. : 27.00	Min. : 50.00	Min. : 29.0
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.0
##	Median : 68.00	Median : 85.00	Median :100.00	Median : 81.0
##	Mean : 69.05	Mean : 83.17	Mean : 99.15	Mean : 82.6
##	3rd Qu.: 79.00	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.0
##	Max. :104.00	Max. :130.00	Max. :145.00	Max. :157.0
##	x.21	x.22	x.23	x.24
##	Min. : 39.00	Min. : 27.00	Min. : 50.00	Min. : 29.00
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 68.00
##	Median : 67.00	Median : 84.00	Median :100.00	Median : 81.00
##	Mean : 68.84	Mean : 82.86	Mean : 98.95	Mean : 82.47
##	3rd Qu.: 79.00	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.00
##	Max. :104.00	Max. :130.00	Max. :145.00	Max. :157.00
##	x.25	x.26	x.27	x.28
##	Min. : 39.00	Min. : 27.00	Min. : 50.00	Min. : 29.00
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.00
##	Median : 68.00	Median : 85.00	Median :100.00	Median : 81.00
##	Mean : 69.16	Mean : 83.37	Mean : 99.21	Mean : 82.66
##	3rd Qu.: 79.00	3rd Qu.:103.00	3rd Qu.:113.00	3rd Qu.: 92.00
##	Max. :104.00	Max. :131.00	Max. :140.00	Max. :154.00
##	x.29	x.30	x.31	x.32
##	Min. : 39.00	Min. : 27.00	Min. : 50.00	Min. : 29.00
##	1st Qu.: 60.00	1st Qu.: 71.00	1st Qu.: 85.00	1st Qu.: 69.00
##	Median : 68.00	Median : 85.00	Median :100.00	Median : 81.00
##	Mean : 68.94	Mean : 83.15	Mean : 99.11	Mean : 82.62

```
## 3rd Qu.: 79.00 3rd Qu.:103.00 3rd Qu.:113.00 3rd Qu.: 92.00
## Max. :104.00 Max. :130.00 Max. :145.00 Max. :157.00
## x.33 x.34 x.35 x.36
## Min. : 39.00 Min. : 27.00 Min. : 50.00 Min. : 29.00
## 1st Qu.: 60.00 1st Qu.: 71.00 1st Qu.: 85.00 1st Qu.: 68.00
## Median : 67.00 Median : 84.00 Median :100.00 Median : 81.00
## Mean : 68.73 Mean : 82.86 Mean : 98.93 Mean : 82.51
## 3rd Qu.: 79.00 3rd Qu.:103.00 3rd Qu.:113.00 3rd Qu.: 92.00
## Max. :104.00 Max. :130.00 Max. :145.00 Max. :157.00
## classes
## red soil :1533
## cotton crop : 703
## grey soil :1358
## damp grey soil : 626
## vegetation stubble : 707
## very damp grey soil:1508
```

```
# exibir alguns dados do Satellite
head(Satellite, n = 6)
```

```
## x.1 x.2 x.3 x.4 x.5 x.6 x.7 x.8 x.9 x.10 x.11 x.12 x.13 x.14 x.15 x.16 x.17
## 1 92 115 120 94 84 102 106 79 84 102 102 83 101 126 133 103 92
## 2 84 102 106 79 84 102 102 83 80 102 102 79 92 112 118 85 84
## 3 84 102 102 83 80 102 102 79 84 94 102 79 84 103 104 81 84
## 4 80 102 102 79 84 94 102 79 80 94 98 76 84 99 104 78 84
## 5 84 94 102 79 80 94 98 76 80 102 102 79 84 99 104 81 76
## 6 80 94 98 76 80 102 102 79 76 102 102 79 76 99 104 81 76
## x.18 x.19 x.20 x.21 x.22 x.23 x.24 x.25 x.26 x.27 x.28 x.29 x.30 x.31 x.32
## 1 112 118 85 84 103 104 81 102 126 134 104 88 121 128 100
## 2 103 104 81 84 99 104 78 88 121 128 100 84 107 113 87
## 3 99 104 78 84 99 104 81 84 107 113 87 84 99 104 79
## 4 99 104 81 76 99 104 81 84 99 104 79 84 99 104 79
## 5 99 104 81 76 99 108 85 84 99 104 79 84 103 104 79
## 6 99 108 85 76 103 118 88 84 103 104 79 79 107 109 87
## x.33 x.34 x.35 x.36 classes
## 1 84 107 113 87 grey soil
## 2 84 99 104 79 grey soil
## 3 84 99 104 79 grey soil
## 4 84 103 104 79 grey soil
## 5 79 107 109 87 grey soil
## 6 79 107 109 87 grey soil
```

```
# Para reproductibilidade
set.seed(7)
```

```
# particionar em 80% para treino e 20% para teste
indices <- createDataPartition(Satellite$classes, p=0.8, list=F)
treino <- Satellite[indices, ]
teste <- Satellite[-indices, ]
```

1.2 Crie partições contendo 80% para treino e 20% para teste

```
# treinar modelos RandomForest, SVM e RNA
rf <- train(classes ~ ., data=treino, method="rf")
svm <- train(classes ~ ., data=treino, method="svmRadial")
rna <- train(classes ~ ., data=treino, method="nnet", trace=F)
```

1.3 Treine modelos RandomForest, SVM e RNA para predição destes dados.

```
# predições
predict.rf <- predict(rf, teste)
predict.svm <- predict(svm, teste)
predict.rna <- predict(rna, teste)

# matrizes de confusões de cada uma das predições

# matriz de confusão para o modelo RF
conf_matrix.rf <- confusionMatrix(predict.rf, teste$classes)

print(conf_matrix.rf)
```

1.4 Escolha o melhor modelo com base em suas matrizes de confusão.

```
## Confusion Matrix and Statistics
##
##               Reference
## Prediction      red soil cotton crop grey soil damp grey soil
## red soil           300         1         3         1
## cotton crop         0        137         1         0
## grey soil           3         0        263        25
## damp grey soil      0         0         2        78
## vegetation stubble  3         0         1         0
## very damp grey soil 0         2         1        21
##
##               Reference
## Prediction      vegetation stubble very damp grey soil
## red soil                4             0
## cotton crop              1             1
## grey soil                0             3
## damp grey soil           1            20
## vegetation stubble       128           4
## very damp grey soil       7            273
##
## Overall Statistics
##
##               Accuracy : 0.9182
##               95% CI : (0.9019, 0.9326)
##      No Information Rate : 0.2383
##      P-Value [Acc > NIR] : < 2.2e-16
##
##               Kappa : 0.8987
##
## McNemar's Test P-Value : NA
##
## Statistics by Class:
```

```
##
##          Class: red soil Class: cotton crop Class: grey soil
## Sensitivity          0.9804          0.9786          0.9705
## Specificity          0.9908          0.9974          0.9694
## Pos Pred Value       0.9709          0.9786          0.8946
## Neg Pred Value       0.9938          0.9974          0.9919
## Prevalence           0.2383          0.1090          0.2111
## Detection Rate       0.2336          0.1067          0.2048
## Detection Prevalence 0.2407          0.1090          0.2290
## Balanced Accuracy     0.9856          0.9880          0.9699
##
##          Class: damp grey soil Class: vegetation stubble
## Sensitivity          0.62400          0.90780
## Specificity          0.98016          0.99300
## Pos Pred Value       0.77228          0.94118
## Neg Pred Value       0.96027          0.98868
## Prevalence           0.09735          0.10981
## Detection Rate       0.06075          0.09969
## Detection Prevalence 0.07866          0.10592
## Balanced Accuracy     0.80208          0.95040
##
##          Class: very damp grey soil
## Sensitivity          0.9070
## Specificity          0.9685
## Pos Pred Value       0.8980
## Neg Pred Value       0.9714
## Prevalence           0.2344
## Detection Rate       0.2126
## Detection Prevalence 0.2368
## Balanced Accuracy     0.9377
```

```
cat('\n')
```

```
# matriz de confusão para o modelo SVM
```

```
conf_matrix.svm <- confusionMatrix(predict.svm, teste$classes)
```

```
print(conf_matrix.svm)
```

```
## Confusion Matrix and Statistics
```

```
##
##          Reference
## Prediction    red soil cotton crop grey soil damp grey soil
##   red soil      303          0          2          0
##   cotton crop    0         138          2          2
##   grey soil      2          0        261         27
##   damp grey soil 0          1          5         74
##   vegetation stubble 1          0          0          1
##   very damp grey soil 0          1          1         21
```

```
##          Reference
## Prediction    vegetation stubble very damp grey soil
##   red soil          5          0
##   cotton crop        2          2
##   grey soil          0          7
##   damp grey soil      1         21
##   vegetation stubble 126          3
##   very damp grey soil 7         268
```

```
##
```

```
## Overall Statistics
##
##           Accuracy : 0.9112
##           95% CI   : (0.8943, 0.9262)
##           No Information Rate : 0.2383
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.8901
##
## Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: red soil Class: cotton crop Class: grey soil
## Sensitivity           0.9902           0.9857           0.9631
## Specificity           0.9928           0.9930           0.9645
## Pos Pred Value        0.9774           0.9452           0.8788
## Neg Pred Value        0.9969           0.9982           0.9899
## Prevalence            0.2383           0.1090           0.2111
## Detection Rate        0.2360           0.1075           0.2033
## Detection Prevalence  0.2414           0.1137           0.2313
## Balanced Accuracy      0.9915           0.9894           0.9638
##
##           Class: damp grey soil Class: vegetation stubble
## Sensitivity           0.59200           0.89362
## Specificity           0.97584           0.99563
## Pos Pred Value        0.72549           0.96183
## Neg Pred Value        0.95685           0.98699
## Prevalence            0.09735           0.10981
## Detection Rate        0.05763           0.09813
## Detection Prevalence  0.07944           0.10202
## Balanced Accuracy      0.78392           0.94462
##
##           Class: very damp grey soil
## Sensitivity           0.8904
## Specificity           0.9695
## Pos Pred Value        0.8993
## Neg Pred Value        0.9665
## Prevalence            0.2344
## Detection Rate        0.2087
## Detection Prevalence  0.2321
## Balanced Accuracy      0.9299
```

```
cat('\n')
```

```
# matriz de confusão para o modelo RNA
conf_matrix.rna <- confusionMatrix(predict.rna, teste$classes)

print(conf_matrix.rna)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    red soil cotton crop grey soil damp grey soil
##   red soil         289         132           3           1
##   cotton crop         6           0           0           0
##   grey soil          3           5        244        104
```

```

##      damp grey soil           0           0           0           0
##      vegetation stubble       7           3          12          13
##      very damp grey soil      1           0          12          7
##
##      Reference
## Prediction      vegetation stubble very damp grey soil
##      red soil           31           1
##      cotton crop        1           0
##      grey soil          13          267
##      damp grey soil     0           0
##      vegetation stubble  91          15
##      very damp grey soil  5          18
##
## Overall Statistics
##
##      Accuracy : 0.5
##      95% CI : (0.4723, 0.5277)
##      No Information Rate : 0.2383
##      P-Value [Acc > NIR] : < 2.2e-16
##
##      Kappa : 0.3672
##
## McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##      Class: red soil Class: cotton crop Class: grey soil
## Sensitivity          0.9444          0.000000          0.9004
## Specificity          0.8282          0.993881          0.6130
## Pos Pred Value       0.6324          0.000000          0.3836
## Neg Pred Value       0.9794          0.890368          0.9583
## Prevalence           0.2383          0.109034          0.2111
## Detection Rate       0.2251          0.000000          0.1900
## Detection Prevalence 0.3559          0.005452          0.4953
## Balanced Accuracy    0.8863          0.496941          0.7567
##
##      Class: damp grey soil Class: vegetation stubble
## Sensitivity          0.00000          0.64539
## Specificity          1.00000          0.95626
## Pos Pred Value       NaN            0.64539
## Neg Pred Value       0.90265          0.95626
## Prevalence           0.09735          0.10981
## Detection Rate       0.00000          0.07087
## Detection Prevalence 0.00000          0.10981
## Balanced Accuracy    0.50000          0.80082
##
##      Class: very damp grey soil
## Sensitivity          0.05980
## Specificity          0.97457
## Pos Pred Value       0.41860
## Neg Pred Value       0.77196
## Prevalence           0.23442
## Detection Rate       0.01402
## Detection Prevalence 0.03349
## Balanced Accuracy    0.51718

```



```
cat('\n')
```

**1.5 Indique qual modelo dá o melhor o resultado e a métrica utilizada** O melhor modelo foi random forest com acurácia de 0.9182 e kappa de 0.8987. A métrica utilizada foram a acurácia e kappa.

```
dados <- read.csv("http://www.razer.net.br/datasets/Volumes.csv", sep=";", dec=",")
head(dados)
```

## 2.1 Carregar o arquivo Volumes.csv (<http://www.razer.net.br/datasets/Volumes.csv>)

```
##  NR  DAP  HT  HP  VOL
## 1   1 34.0 27.00 1.80 0.8971441
## 2   2 41.5 27.95 2.75 1.6204441
## 3   3 29.6 26.35 1.15 0.8008181
## 4   4 34.3 27.15 1.95 1.0791682
## 5   5 34.5 26.20 1.00 0.9801112
## 6   6 29.9 27.10 1.90 0.9067022
```

```
dados$NR <- NULL
```

## 2.2 Eliminar a coluna NR, que só apresenta um número sequencial

```
regression.indices <- caret::createDataPartition(dados$VOL, p=0.8, list=F)
regression.treino <- dados[regression.indices, ]
regression.teste <- dados[-regression.indices, ]
```

## 2.3 Criar partição de dados: treinamento 80%, teste 20%

```
# Para reproductibilidade
set.seed(7)
```

```
regression.rf <- caret::train(VOL ~ ., data=regression.treino, method="rf")
```

## 2.4 Usando o pacote “caret”, treinar os modelos: Random Forest (rf), SVM (svmRadial), Redes Neurais (neuralnet) e o modelo alométrico de SPURR

```
## note: only 2 unique complexity parameters in default grid. Truncating the grid to 2 .
```

```
regression.svm <- caret::train(VOL ~ ., data=regression.treino, method="svmRadial")
regression.rna <- caret::train(
  VOL ~ ., data=regression.treino,
  method="nnet",
  trControl=trainControl(method = "LOOCV"),
  trace=F
)
```

```
regression.spurr <- nls(
  VOL ~ b0 + b1*DAP*DAP*HT,
  data=regression.treino,
```

```
start=list(b0=0.5, b1=0.5)
)
```

treino do modelo Spurr

```
summary(regression.spurr)
```

visualizar os resultados de Spurr

```
##
## Formula: VOL ~ b0 + b1 * DAP * DAP * HT
##
## Parameters:
##      Estimate Std. Error t value Pr(>|t|)
## b0 2.490e-02  5.328e-02   0.467   0.642
## b1 3.860e-05  1.466e-06  26.325  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1377 on 78 degrees of freedom
##
## Number of iterations to convergence: 2
## Achieved convergence tolerance: 1.134e-09
```

2.5 Efetue as predições nos dados de teste    predições

```
# Para reprodutibilidade
set.seed(7)
```

```
predict.regression.rf <- predict(regression.rf, regression.teste)
predict.regression.svm <- predict(regression.svm, regression.teste)
predict.regression.rna <- predict(regression.rna, regression.teste)
predict.regression.suprr <- predict(regression.spurr, regression.teste)
```

2.6 Crie suas próprias funções (UDF) e calcule as seguintes métricas entre a predição e os dados observados

- Erro padrão de estimativa: Syx

```
Syx <- function(reals, predicted, n) {
  return (sqrt(sum((reals - predicted)^2)/(n - 2)))
}
```

- Erro padrão de estimativa em porcentagem: Syx%

```
SyxPercent <- function(reals, predicted, n) {
  return ((Syx(reals, predicted, n)/mean(reals))*100)
}
```

- Coeficiente de determinação: R2

```
R2 <- function (reals, predicted) {
  return (1 - sum((reals - predicted)^2)/sum((reals - mean(reals))^2))
}
```

2.7 Escolha o melhor modelo.

### métrica de estimativas para o modelo RandomForest - Regressão

- coeficiente de determinação

```
R2(regression.teste$VOL, predict.regression.rf)
```

```
## [1] 0.8223603
```

- Erro padrão estimativa

```
n <- nrow(regression.teste)
Syx(regression.teste$VOL, predict.regression.rf, n)
```

```
## [1] 0.1376052
```

- Erro padrão estimativa em porcentagem

```
n <- nrow(regression.teste)
SyxPercent(regression.teste$VOL, predict.regression.rf, n)
```

```
## [1] 10.42195
```

### métrica de estimativas para o modelo SVM - Regressão

- coeficiente de determinação

```
R2(regression.teste$VOL, predict.regression.svm)
```

```
## [1] 0.6254546
```

- Erro padrão estimativa

```
n <- nrow(regression.teste)
Syx(regression.teste$VOL, predict.regression.svm, n)
```

```
## [1] 0.19981
```

- Erro padrão estimativa em porcentagem

```
n <- nrow(regression.teste)
SyxPercent(regression.teste$VOL, predict.regression.svm, n)
```

```
## [1] 15.13322
```

### métricas de estimativas para o modelo nnet - Regressão

- coeficiente de determinação

```
R2(regression.teste$VOL, predict.regression.rna)
```

```
## [1] -1.069672
```

- Erro padrão estimativa

```
n <- nrow(regression.teste)
Syx(regression.teste$VOL, predict.regression.rna, n)
```

```
## [1] 0.4696948
```

- Erro padrão estimativa em porcentagem

```
n <- nrow(regression.teste)
SyxPercent(regression.teste$VOL, predict.regression.rna, n)
```

```
## [1] 35.57377
```

## métricas de estimativas para o modelo Spurr

- coeficiente de determinação

```
R2(regression.teste$VOL, predict.regression.suprr)
```

```
## [1] 0.7734018
```

- Erro padrão estimativa

```
n <- nrow(regression.teste)
```

```
Syx(regression.teste$VOL, predict.regression.suprr, n)
```

```
## [1] 0.1554151
```

- Erro padrão estimativa em porcentagem

```
n <- nrow(regression.teste)
```

```
SyxPercent(regression.teste$VOL, predict.regression.suprr, n)
```

```
## [1] 11.77084
```

## 2.7 escolha o melhor modelo

### Resumo dos resultados RF:

1. coeficiente de determinação: 0.8223603.
2. Erro padrão estimativa: 0.1376052.
3. Erro padrão estimativa em porcentagem: 10.42195

### SVM:

1. coeficiente de determinação: 0.6254546
2. Erro padrão estimativa: 0.19981
3. Erro padrão estimativa em porcentagem: 15.13322

### nnet:

1. coeficiente de determinação: -1.069672
2. Erro padrão estimativa: 0.4696948
3. Erro padrão estimativa em porcentagem: 35.57377

### spurr:

1. coeficiente de determinação: 0.7734018
2. Erro padrão estimativa: 0.1554151
3. Erro padrão estimativa em porcentagem: 11.77084

Com base nas métricas, o modelo que se saiu melhor foi o RandomForest, com R2 igual 0.8223603, Erro padrão estimativa de 0.1376052 e Erro padrão de estimativa em porcentagem de 10.42195