

TRABALHO DE IAA003 – Linguagem R

EQUIPE: 23

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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//RtmpeUIvww/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//RtmpeUIvww/downloaded_packages
install.packages("kernlab", repos = "http://cran.us.r-project.org")
```

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

```
##
## The downloaded binary packages are in
## /var/folders/7c/43lddvnj7bsfph_4fmdrns140000gp/T//RtmpeUIvww/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 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, predicteds, n) {
  return (sqrt(sum((reals - predicteds)^2)/(n - 2)))
}
```

- Erro padrão de estimativa em porcentagem: Syx%

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

- Coeficiente de determinação: R2

```
R2 <- function (reals, predicteds) {
  return (1 - sum((reals - predicteds)^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