Statistics in R: Task 22

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```
library(mlbench)
library(dplyr)
library(randomForest)
```

Random Forest

```
data(Vehicle)
df <- Vehicle
str(df)
```

```
## 'data.frame':    846 obs. of  19 variables:
##  $ Comp        : num  95 91 104 93 85 107 97 90 86 93 ...
##  $ Circ        : num  48 41 50 41 44 57 43 43 34 44 ...
##  $ D.Circ      : num  83 84 106 82 70 106 73 66 62 98 ...
##  $ Rad.Ra      : num  178 141 209 159 205 172 173 157 140 197 ...
##  $ Pr.Axis.Ra  : num  72 57 66 63 103 50 65 65 61 62 ...
##  $ Max.L.Ra    : num  10 9 10 9 52 6 6 9 7 11 ...
##  $ Scat.Ra     : num  162 149 207 144 149 255 153 137 122 183 ...
##  $ Elong       : num  42 45 32 46 45 26 42 48 54 36 ...
##  $ Pr.Axis.Rect: num  20 19 23 19 19 28 19 18 17 22 ...
##  $ Max.L.Rect  : num  159 143 158 143 144 169 143 146 127 146 ...
##  $ Sc.Var.Maxis: num  176 170 223 160 241 280 176 162 141 202 ...
##  $ Sc.Var.maxis: num  379 330 635 309 325 957 361 281 223 505 ...
##  $ Ra.Gyr      : num  184 158 220 127 188 264 172 164 112 152 ...
##  $ Skew.Maxis  : num  70 72 73 63 127 85 66 67 64 64 ...
##  $ Skew.maxis  : num  6 9 14 6 9 5 13 3 2 4 ...
##  $ Kurt.maxis  : num  16 14 9 10 11 9 1 3 14 14 ...
##  $ Kurt.Maxis  : num  187 189 188 199 180 181 200 193 200 195 ...
##  $ Holl.Ra     : num  197 199 196 207 183 183 204 202 208 204 ...
##  $ Class       : Factor w/ 4 levels "bus","opel","saab",..: 4 4 3 4 1 1 1 4 4 3 ...
```

```
summary(df$Class)
```

```
##  bus opel saab  van
##  218  212  217  199
```

```
df <- df %>%
    filter(Class != 'opel')
df$Class <- factor(df$Class)
str(df)
```

```
## 'data.frame':    634 obs. of  19 variables:
##  $ Comp        : num  95 91 104 93 85 107 97 90 86 93 ...
##  $ Circ        : num  48 41 50 41 44 57 43 43 34 44 ...
##  $ D.Circ      : num  83 84 106 82 70 106 73 66 62 98 ...
##  $ Rad.Ra      : num  178 141 209 159 205 172 173 157 140 197 ...
##  $ Pr.Axis.Ra  : num  72 57 66 63 103 50 65 65 61 62 ...
##  $ Max.L.Ra    : num  10 9 10 9 52 6 6 9 7 11 ...
##  $ Scat.Ra     : num  162 149 207 144 149 255 153 137 122 183 ...
##  $ Elong       : num  42 45 32 46 45 26 42 48 54 36 ...
##  $ Pr.Axis.Rect: num  20 19 23 19 19 28 19 18 17 22 ...
##  $ Max.L.Rect  : num  159 143 158 143 144 169 143 146 127 146 ...
##  $ Sc.Var.Maxis: num  176 170 223 160 241 280 176 162 141 202 ...
##  $ Sc.Var.maxis: num  379 330 635 309 325 957 361 281 223 505 ...
##  $ Ra.Gyr      : num  184 158 220 127 188 264 172 164 112 152 ...
##  $ Skew.Maxis  : num  70 72 73 63 127 85 66 67 64 64 ...
##  $ Skew.maxis  : num  6 9 14 6 9 5 13 3 2 4 ...
##  $ Kurt.maxis  : num  16 14 9 10 11 9 1 3 14 14 ...
##  $ Kurt.Maxis  : num  187 189 188 199 180 181 200 193 200 195 ...
##  $ Holl.Ra     : num  197 199 196 207 183 183 204 202 208 204 ...
##  $ Class       : Factor w/ 3 levels "bus","saab","van": 3 3 2 3 1 1 1 3 3 2 ...
```

```
summary(df$Class)
```

```
##  bus saab  van
##  218  217  199
```

Bagging Trees

```
set.seed(1)

train <- sample(1:nrow(df), nrow(df)/2)
bag <- randomForest(Class ~ ., data = df,
                    subset = train,
                    mtry = 18,
                    importance = T)
bag
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, mtry = 18, importance = T,      subset = train)
##                Type of random forest: classification
##                      Number of trees: 500
## No. of variables tried at each split: 18
##
##         OOB estimate of  error rate: 6.62%
## Confusion matrix:
##      bus saab van class.error
## bus  107    2   1  0.02727273
## saab   7   95  10  0.15178571
## van    0    1  94  0.01052632
```

```
set.seed(1)

bag <- randomForest(Class ~ ., data = df,
                    subset = train,
                    ntree = 25,
                    mtry = 18,
                    importance = T)
bag
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, ntree = 25, mtry = 18,      importance = T, subset = t
##                Type of random forest: classification
##                      Number of trees: 25
## No. of variables tried at each split: 18
##
##         OOB estimate of  error rate: 8.2%
## Confusion matrix:
##      bus saab van class.error
## bus  106    2   2  0.03636364
## saab   9   94   9  0.16071429
## van    2    2  91  0.04210526
```

Random Forest

```
set.seed(1)

rf <- randomForest(Class ~ .,
                   data = df,
                   subset = train,
                   mtry = 9,
                   importance = T)
rf
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, mtry = 9, importance = T,      subset = train)
##                Type of random forest: classification
##                      Number of trees: 500
## No. of variables tried at each split: 9
##
##         OOB estimate of  error rate: 6.62%
## Confusion matrix:
##      bus saab van class.error
## bus  108    1   1  0.01818182
## saab   7   95  10  0.15178571
## van    1    1  93  0.02105263
```

```
sample_size <- ceiling(.632*nrow(df[-train,]))
vars <- floor(sqrt(ncol(df)))
sample_size
```

```
## [1] 201
```

```
vars
```

```
## [1] 4
```

```
set.seed(1)
rf <- randomForest(Class ~ .,
                   data = df,
                   subset = train,
                   mtry = vars,
                   sampsize = sample_size,
                   importance = T)
rf
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, mtry = vars, sampsize = sample_size,      importance =
##                Type of random forest: classification
##                      Number of trees: 500
## No. of variables tried at each split: 4
##
##         OOB estimate of  error rate: 6.31%
## Confusion matrix:
##      bus saab van class.error
## bus  108    0   2  0.01818182
## saab   8   95   9  0.15178571
## van    1    0  94  0.01052632
```

```
# the smallest error rate for now
```

```
# test
est_class <- predict(rf, newdata = df[-train,])
mean(est_class != df$Class[-train])
```

```
## [1] 0.1041009
```

```
# error rate for the test data is larger maybe because of overfitting,
#                                        maybe because of unforfunate choice of test/train data
```

```
set.seed(1)

ntrees <- 500
rf <- randomForest(Class ~ .,
                   data = df,
                   subset = train,
                   ntree = ntrees,
                   mtry = vars,
                   sampsize = sample_size,
                   importance = T,
                   do.trace = ntrees/10)
```

```
## ntree      OOB      1      2      3
##    50:   7.57%  2.73% 17.86%  1.05%
##   100:   6.94%  1.82% 16.07%  2.11%
##   150:   6.31%  1.82% 15.18%  1.05%
##   200:   6.62%  1.82% 16.07%  1.05%
##   250:   5.99%  1.82% 14.29%  1.05%
##   300:   6.31%  1.82% 15.18%  1.05%
##   350:   6.62%  1.82% 16.07%  1.05%
##   400:   6.31%  1.82% 15.18%  1.05%
##   450:   6.31%  1.82% 15.18%  1.05%
##   500:   6.31%  1.82% 15.18%  1.05%
```

```
rf
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, ntree = ntrees,      mtry = vars, sampsize = sample_si
##                Type of random forest: classification
##                      Number of trees: 500
## No. of variables tried at each split: 4
##
##         OOB estimate of  error rate: 6.31%
## Confusion matrix:
##      bus saab van class.error
## bus  108    0   2  0.01818182
## saab   8   95   9  0.15178571
## van    1    0  94  0.01052632
```

```
set.seed(1)

ntrees <- 300
rf <- randomForest(Class ~ .,
                   data = df,
                   subset = train,
                   ntree = ntrees,
                   mtry = vars,
                   sampsize = sample_size,
                   importance = T,
                   do.trace = ntrees/30)
```

```
## ntree      OOB      1      2      3
##    10:   9.78%  6.36% 19.64%  2.11%
##    20:   7.57%  4.55% 15.18%  2.11%
##    30:   7.57%  3.64% 16.96%  1.05%
##    40:   6.62%  2.73% 15.18%  1.05%
##    50:   7.57%  2.73% 17.86%  1.05%
##    60:   6.31%  1.82% 15.18%  1.05%
##    70:   6.62%  1.82% 16.07%  1.05%
##    80:   6.94%  1.82% 16.07%  2.11%
##    90:   6.94%  1.82% 16.07%  2.11%
##   100:   6.94%  1.82% 16.07%  2.11%
##   110:   6.62%  1.82% 16.07%  1.05%
##   120:   6.62%  1.82% 16.07%  1.05%
```

```
##   130:   6.62%  1.82% 16.07%  1.05%
##   140:   6.62%  1.82% 16.07%  1.05%
##   150:   6.31%  1.82% 15.18%  1.05%
##   160:   6.31%  1.82% 15.18%  1.05%
##   170:   6.31%  1.82% 15.18%  1.05%
##   180:   6.31%  1.82% 15.18%  1.05%
##   190:   6.62%  1.82% 16.07%  1.05%
##   200:   6.62%  1.82% 16.07%  1.05%
##   210:   6.31%  1.82% 15.18%  1.05%
##   220:   6.62%  1.82% 15.18%  2.11%
##   230:   5.99%  1.82% 14.29%  1.05%
##   240:   5.99%  1.82% 14.29%  1.05%
##   250:   5.99%  1.82% 14.29%  1.05%
##   260:   6.31%  1.82% 15.18%  1.05%
##   270:   6.31%  1.82% 15.18%  1.05%
##   280:   6.31%  1.82% 15.18%  1.05%
##   290:   6.31%  1.82% 15.18%  1.05%
##   300:   6.31%  1.82% 15.18%  1.05%
```

```
rf
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, ntree = ntrees,      mtry = vars, sampsize = sample_si
##                Type of random forest: classification
##                      Number of trees: 300
## No. of variables tried at each split: 4
##
##         OOB estimate of  error rate: 6.31%
## Confusion matrix:
##      bus saab van class.error
## bus  108    0   2  0.01818182
## saab   8   95   9  0.15178571
## van    1    0  94  0.01052632
```

```
set.seed(1)

ntrees <- 230
rf <- randomForest(Class ~ .,
                   data = df,
                   subset = train,
                   ntree = ntrees,
                   mtry = vars,
                   sampsize = sample_size,
                   importance = T)
rf
```

```
##
## Call:
##  randomForest(formula = Class ~ ., data = df, ntree = ntrees,      mtry = vars, sampsize = sample_si
##                Type of random forest: classification
##                      Number of trees: 230
## No. of variables tried at each split: 4
```

```
##
##         OOB estimate of  error rate: 5.99%
## Confusion matrix:
##      bus saab van class.error
## bus  108    0   2  0.01818182
## saab   7   96   9  0.14285714
## van    1    0  94  0.01052632
```

```
est_class <- predict(rf, newdata = df[-train,])
mean(est_class != df$Class[-train])
```

```
## [1] 0.09463722
```

```
importance(rf)
```

```
##                    bus       saab         van MeanDecreaseAccuracy
## Comp          6.059377  4.7009247  4.85796831              7.442709
## Circ          6.034482  5.6443849  2.27603397              7.600410
## D.Circ       12.476425  8.8535369  8.96415111             14.680438
## Rad.Ra        6.169519  5.8866048  6.68848695              7.947228
## Pr.Axis.Ra   10.621176  7.1856596  5.67786171             12.040802
## Max.L.Ra     21.878787 17.7790224 18.23877766             24.998139
## Scat.Ra      10.756709  7.1122217 13.16633136             12.964160
## Elong        10.916012  7.9774177 14.64730755             13.531771
## Pr.Axis.Rect  6.538649  3.9292392  8.81201743              8.740163
## Max.L.Rect    9.088398  9.9761215  5.57174737             11.528983
## Sc.Var.Maxis  9.979156  6.6408687  8.67784453             10.251777
## Sc.Var.maxis  9.847864  7.7914692 12.58708627             12.508193
## Ra.Gyr        3.728664  3.7132889  2.62058044              5.727017
## Skew.Maxis    9.573700 10.5193031  5.10122319             12.660068
## Skew.maxis    5.870879 -0.6134069  0.02370082              4.918691
## Kurt.maxis    3.675142  5.0230821  3.14545219              6.431313
## Kurt.Maxis    8.566836  4.8730573  4.60439629              9.189796
## Holl.Ra       8.172886  8.7092794  6.33963654             11.388951
##              MeanDecreaseGini
## Comp                 4.047486
## Circ                 2.760725
## D.Circ              11.529290
## Rad.Ra               5.428790
## Pr.Axis.Ra           4.322950
## Max.L.Ra            22.718260
## Scat.Ra             12.094844
## Elong               11.622053
## Pr.Axis.Rect         6.534919
## Max.L.Rect           6.633753
## Sc.Var.Maxis         9.015377
## Sc.Var.maxis        12.547583
## Ra.Gyr               2.111020
## Skew.Maxis           7.435177
## Skew.maxis           2.152614
## Kurt.maxis           2.687833
## Kurt.Maxis           3.296486
## Holl.Ra              5.984613
```

```
# Max.L.Ra - max length rectangularity
```

```
varImpPlot(rf)
```