

Classification models

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- Load Boston database
- Generate binary variable out of criminality

```
library(MASS)
library(dplyr)
## load database
df <- Boston

## transform crim to binary
df$crim <- ifelse(df$crim >= mean(df$crim), 1, 0)
print(head(df))
```

	crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	black	lstat	medv
1	0	18	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	396.90	4.98	24.0
2	0	0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.90	9.14	21.6
3	0	0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83	4.03	34.7
4	0	0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63	2.94	33.4
5	0	0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.90	5.33	36.2
6	0	0	2.18	0	0.458	6.430	58.7	6.0622	3	222	18.7	394.12	5.21	28.7

- Generate train and test partition (0.7, 0.3)

```
## option 1 self-made
df_partition <- function(data, train_size)
{
  set.seed(101)
  # get indices
  sample <- sample.int(n = nrow(data),
                      size = floor(train_size*nrow(data)),
                      replace = F)
  # generate train data
```

```

train <- data[sample, ]
# and test data
test <- data[-sample, ]
# list of train and test
return(list(train, test))

# return it
}

## example
fun_list <- df_partition(df, 0.7)
df_train <- fun_list[[1]]
df_test <- fun_list[[2]]

## option 2 using caret
library(caret)
## LGOCV <- leave group out
## p <- train percentage
## number <- iterations
## same as 1 train/test split
train_control <- trainControl(method = "LGOCV",
                               p = 0.7,
                               number = 1,
                               savePredictions = TRUE)

## the first function will be used since it's easier to deal with
## native R objects



- Train following models
- KNN
- SVM
- Decision tree
- Neural net



#####
## K-nearest-neighbours ##
#####

```

```

## knn parameter is number neighbours 'k'
library(class)
knn_md1 <- knn(as.factor(crim) ~ .,
               data = df,
               k = 3)

## confusion matrix
tbl <- table(knn_md1, df_test$crim)
accuracy <- function(x){sum(diag(x)/(sum(rowSums(x)))) * 100}
accuracy(tbl)

## optimize with grid search
grid <- seq(from = 1, to = 200, by = 1)
k_opt <- sapply(grid, function(x) accuracy(table(knn(df_train,
                                                  df_test,
                                                  cl = as.factor(df_train$crim),
                                                  k = x), df_test$crim)))

## using caret
knn_caret <- train(as.factor(crim) ~ .,
                  data = df,
                  method = "knn",
                  trControl = train_control,
                  tuneGrid = expand.grid(k = 1:200))

## we can get the optimal parameter back a train using that
knn_md1_opt <- knn(df_train,
                  df_test,
                  cl = as.factor(df_train$crim),
                  k = knn_caret$bestTune$k)

## confusion matrix optimal model
tbl_2 <- table(knn_md1, df_test$crim)
accuracy(tbl_2)

#####
## Support vector machine ##
#####
## here on out I'll be using caret because is way much faster
## assumming a linear kernel there's only one parameter to optimize
## the cost o C

```

```

## parallel so it goes faster
library(doParallel)
registerDoParallel(cores=8)
svm_caret <- train(as.factor(crim) ~ .,
                  data = df, method = "svmLinear",
                  tuneGrid = expand.grid(C = 1:100),
                  preProcess = c("scale"),
                  metric = "Accuracy")

## we do the same as before
library(e1017)
svm_mdl_opt <- svm(as.factor(crim) ~ .,
                  data = df_train,
                  kernel.type="linear",
                  cost = svm_caret$bestTune$C)
svm_pred <- predict(svm_mdl_opt, newdata = df_test)
## confusion matrix optimal model
tbl_3 <- table(svm_pred, df_test$crim)
accuracy(tbl_3)

#####
## Decision tree ##
#####
## trained in all data
library(tree)
tree_boston <- tree(as.factor(crim) ~ .,
                  data = df)
par(mfrow = c(1,3))
plot(tree_boston)
text(tree_boston, pretty = 0)
## using train test
tree_boston_cv = tree(as.factor(crim) ~ .,
                  data = df_train)
plot(tree_boston_cv)
text(tree_boston_cv, pretty=0)
## check accuracy
tree_pred = predict(tree_boston_cv, newdata = df_train, type="class")
tbl_4 <- with(df_train, table(tree_pred, crim))

```

```

accuracy(tbl_4)

## prune the tree
tree_boston_prune <- cv.tree(tree_boston_cv, FUN = prune.misclass)
plot(tree_boston_prune)

#####
## neural network ##
#####
library(nnet)
nnet_boston <- train(as.factor(crim) ~ .,
                     data = df,
                     method = "nnet",
                     trControl = train_control,
                     tuneGrid = expand.grid(size = 1:20, decay = 0.1))

Error in knn(as.factor(crim) ~ ., data = df, k = 3) :
  unused argument (data = df)
[1] 98.02632
[1] 98.02632
Error in library(e1017) : there is no package called 'e1017'
[1] 98.02632
[1] 99.15254
# weights: 16
initial value 226.196132
iter 10 value 201.052318
iter 20 value 72.846098
iter 30 value 43.211325
# weights: 31
initial value 292.368582
iter 40 value 37.945742
iter 10 value 201.029879
iter 20 value 54.747725
iter 50 value 29.668717
iter 60 value 28.102580
iter 30 value 43.887130
iter 70 value 21.512648
# weights: 46

```

```

initial value 309.287302
iter 80 value 19.677944
iter 90 value 19.619891
iter 40 value 29.438416
iter 100 value 19.608291
final value 19.608291
stopped after 100 iterations
iter 10 value 173.198654
# weights: 61
initial value 214.048752
iter 50 value 20.226776
iter 20 value 157.993503
# weights: iter 10 value 39.557842
76
initial value 255.688362
iter 60 value 12.007125
iter 30 value 69.971068
iter 20 value 33.132042
iter 40 value 18.964819
iter 70 value 9.268370
iter 10 value 182.757707
# weights: 136
initial value 214.028284
iter 50 value 17.174742
iter 80 value 7.809193
iter 30 value 31.198290
iter 90 value 7.671440
iter 20 value 40.145347
iter 100 value 7.653837
final value 7.653837
stopped after 100 iterations
# weights: 91 iter 60 value 14.524319

initial value 340.440439
iter 40 value 30.921707
iter 10 value 172.297840
iter 30 value 32.763593
iter 70 value 13.387735

```

```

iter 10 value 40.655366
iter 50 value 28.857423
iter 80 value 11.049919
iter 40 value 31.875453
iter 20 value 21.516621
iter 20 value 135.271112
# weights: 151
iter 90 value 9.801624
initial value 559.517776
iter 60 value 20.806553
iter 50 value 28.671111
# weights: 121
initial value 229.795568
iter 70 value 9.624488
iter 100 value 9.455691
final value 9.455691
stopped after 100 iterations
iter 30 value 12.653412
iter 60 value 19.803904
iter 10 value 188.049895
iter 10 value 55.884273
iter 30 value 38.661605
iter 80 value 8.587769
iter 40 value 10.842930
iter 70 value 17.658193
iter 20 value 52.969410
# weights: iter 20 value 41.172773
iter 90 value 8.103390
iter 40 value 31.245365
iter 80 value 12.201893
iter 30 value 37.512875
# weights: 166
initial value 298.937291
iter 50 value 10.159459
iter 100 value 6.892904
final value 6.892904
stopped after 100 iterations
106

```

```
iter 90 value 8.984671
initial value 275.413411
iter 30 value 34.438604
iter 50 value 29.245895
iter 10 value 72.501052
iter 60 value 9.264950
iter 100 value 8.273510
final value 8.273510
stopped after 100 iterations
iter 40 value 35.790249
iter 10 value 50.579123
# weights: 181
initial value 272.266464
iter 60 value 26.492098
iter 20 value 40.808950
iter 70 value 7.648837
iter 40 value 32.098814
iter 50 value 30.341618
# weights: 196
initial value 241.582623
iter 70 value 15.221633
iter 80 value 7.430490
iter 10 value 72.235171
iter 30 value 30.459756
iter 50 value 26.898181
iter 60 value 27.346294
iter 80 value 12.539122
iter 90 value 7.113488
iter 10 value 117.076454
iter 20 value 35.800538
iter 20 value 47.142967
iter 40 value 28.080017
iter 100 value 6.136970
final value 6.136970
stopped after 100 iterations
iter 60 value 20.460130
iter 90 value 9.236030
iter 70 value 18.871383
```



```
iter 20 value 48.273706
iter 50 value 27.767165
iter 30 value 32.966870
iter 70 value 13.744073
iter 100 value 7.819880
final value 7.819880
stopped after 100 iterations
iter 80 value 16.419615
# weights: 211
initial value 368.251083
iter 80 value 11.023044
iter 90 value 8.512111
iter 60 value 23.608656
iter 30 value 45.136249
# weights: 256
initial value 262.384382
iter 40 value 29.517882
iter 10 value 42.773477
iter 30 value 43.857493
iter 100 value 7.495080
final value 7.495080
stopped after 100 iterations
iter 90 value 9.437385
iter 70 value 12.118256
iter 50 value 22.475089
iter 40 value 39.667434
iter 20 value 37.181665
iter 10 value 93.785797
# weights: 241
initial value 448.753982
iter 100 value 7.292606
final value 7.292606
stopped after 100 iterations
iter 80 value 9.026718
iter 60 value 17.068397
iter 40 value 32.806673
iter 50 value 14.929911
iter 10 value 46.681667
```

```
# weights: 271
iter 20 value 36.340653
initial value 436.949875
iter 70 value 15.173936
iter 90 value 7.278977
iter 30 value 31.227938
iter 50 value 25.997305
iter 60 value 12.625639
iter 20 value 36.741133
iter 80 value 14.120632
iter 100 value 6.715546
final value 6.715546
stopped after 100 iterations
iter 10 value 61.990227
iter 40 value 29.345213
iter 30 value 23.504202
iter 70 value 8.570259
iter 60 value 23.335348
iter 90 value 12.027490
iter 30 value 33.055696
# weights: 286
initial value 406.574475
iter 20 value 41.257323
iter 50 value 17.985714
iter 80 value 7.571126
iter 100 value 6.969828
final value 6.969828
stopped after 100 iterations
iter 40 value 20.083305
iter 70 value 17.229861
iter 40 value 29.815509
iter 10 value 114.777632
iter 60 value 15.383874
# weights: 301
iter 90 value 7.110509
initial value 260.956503
iter 80 value 14.952952
iter 30 value 32.329012
```

```
iter 50 value 16.132398
iter 90 value 13.747189
iter 100 value 5.512198
final value 5.512198
stopped after 100 iterations
iter 20 value 91.334228
iter 50 value 28.989031
iter 70 value 14.691812
iter 10 value 42.799434
iter 40 value 19.973625
iter 100 value 13.374171
iter 60 value 11.478912
final value 13.374171
stopped after 100 iterations
iter 30 value 68.231096
iter 60 value 23.404093
iter 80 value 14.243021
iter 20 value 38.966559
iter 50 value 11.724646
# weights: 226
initial value 218.478341
iter 70 value 9.548567
iter 40 value 52.879250
iter 70 value 12.594034
iter 90 value 12.789243
iter 30 value 36.965256
iter 60 value 7.561389
iter 10 value 45.918771
iter 100 value 10.899136
final value 10.899136
stopped after 100 iterations
iter 80 value 11.116552
iter 50 value 30.944902
iter 80 value 8.834259
iter 20 value 34.546945
iter 70 value 7.128295
iter 40 value 32.822016
iter 60 value 27.615291
```

```
iter 90 value 9.976340
iter 90 value 8.300747
iter 30 value 30.854265
iter 80 value 6.386393
iter 100 value 6.169060
final value 6.169060
stopped after 100 iterations
iter 70 value 17.429110
iter 50 value 31.872403
iter 100 value 9.360403
final value 9.360403
stopped after 100 iterations
iter 40 value 27.200476
iter 90 value 5.314960
iter 80 value 14.887858
iter 50 value 14.883211
iter 60 value 26.635680
iter 60 value 11.916040
iter 90 value 11.444747
iter 100 value 5.232587
final value 5.232587
stopped after 100 iterations
iter 70 value 9.975334
iter 70 value 23.239447
iter 80 value 8.028598
iter 100 value 8.502123
final value 8.502123
stopped after 100 iterations
iter 90 value 6.773246
iter 80 value 20.511832
iter 100 value 6.634517
final value 6.634517
stopped after 100 iterations
iter 90 value 16.150784
iter 100 value 12.085513
final value 12.085513
stopped after 100 iterations
# weights: 121
```

```
initial value 344.534530
iter 10 value 71.092929
iter 20 value 63.838360
iter 30 value 56.438977
iter 40 value 40.868469
iter 50 value 35.911126
iter 60 value 31.406140
iter 70 value 26.444076
iter 80 value 21.994367
iter 90 value 20.213973
iter 100 value 19.302098
final value 19.302098
stopped after 100 iterations
```