Practical 2: Model Evaluation

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Question 1

Load the data into R. Name the columns to better identify the board, as visited from left to right and from top to down.

Check for missing values.

```
any(is.na(data))
```

[1] FALSE

Question 2

Read the "data splitting" section at the web page of caret. Then split the data into 70% training and 30% test by keeping the original proportion of classes.

```
set.seed(825)
inTraining <- createDataPartition(data$Class, p=.7, list=FALSE)
data_training <- data[ inTraining,]
data_testing <- data[-inTraining,]
str(data_training)</pre>
```

```
## 'data.frame':
                    672 obs. of 10 variables:
## $ top.left.square
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 3 ...
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 3 ...
   $ top.middle.square
## $ top.right.square
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 3 ...
## $ middle.left.square : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 ...
## $ middle.middle.square: Factor w/ 3 levels "b", "o", "x": 2 2 2 2 2 2 2 1 1 ...
## $ middle.right.square : Factor w/ 3 levels "b", "o", "x": 2 2 2 2 2 1 1 2 2 ...
## $ bottom.left.square : Factor w/ 3 levels "b", "o", "x": 3 2 2 2 1 1 2 2 2 1 ...
## $ bottom.middle.square: Factor w/ 3 levels "b", "o", "x": 2 3 2 1 2 1 2 1 2 2 ...
## $ bottom.right.square : Factor w/ 3 levels "b", "o", "x": 2 2 3 1 1 2 1 2 1 2 ...
                          : Factor w/ 2 levels "negative", "positive": 2 2 2 2 2 2 2 2 2 ...
   $ Class
str(data testing)
## 'data.frame':
                    286 obs. of 10 variables:
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 ...
   $ top.left.square
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 ...
   $ top.middle.square
                          : Factor w/ 3 levels "b", "o", "x": 3 3 3 3 3 3 3 3 3 3 ...
## $ top.right.square
## $ middle.left.square : Factor w/ 3 levels "b", "o", "x": 3 3 2 2 2 2 2 2 2 ...
## $ middle.middle.square: Factor w/ 3 levels "b", "o", "x": 2 1 3 3 3 3 2 1 1 1 ...
## $ middle.right.square : Factor w/ 3 levels "b", "o", "x": 1 2 2 2 2 1 1 3 3 2 ...
## $ bottom.left.square : Factor w/ 3 levels "b", "o", "x": 1 2 3 2 1 2 2 2 1 3 ...
## $ bottom.middle.square: Factor w/ 3 levels "b", "o", "x": 2 1 2 1 2 1 2 2 2 ...
## $ bottom.right.square : Factor w/ 3 levels "b", "o", "x": 2 2 2 1 1 2 3 1 2 1 ...
## $ Class
                          : Factor w/ 2 levels "negative", "positive": 2 2 2 2 2 2 2 2 2 ...
```

Question 3

Specifive the type of resampling.

Apply the models: Naive Bayes, Decision Tree, Neural Networks, Nearest Neighbour and SVM (linear kernel) to the data training dataset using the same seed.

1. Model Naive Bayes

```
set.seed(825)
nb <- train(Class ~ .,</pre>
            data=data training,
            method="naive bayes",
            trControl=fitControl)
nb
## Naive Bayes
##
## 672 samples
##
     9 predictor
     2 classes: 'negative', 'positive'
##
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 1 times)
## Summary of sample sizes: 606, 605, 604, 605, 605, 605, ...
```

```
## Resampling results across tuning parameters:
##
##
     usekernel Accuracy
                           Kappa
##
    FALSE
               0.6756219 0.2666418
##
      TRUE
                0.6845565 0.1130575
##
## Tuning parameter 'laplace' was held constant at a value of 0
## parameter 'adjust' was held constant at a value of 1
## Accuracy was used to select the optimal model using the largest value.
## The final values used for the model were laplace = 0, usekernel = TRUE
## and adjust = 1.
  2. Model Decision Tree
set.seed(825)
dt <- train(Class ~ .,
            data=data_training,
            method="rpart2",
            trControl=fitControl)
dt
## CART
##
## 672 samples
    9 predictor
     2 classes: 'negative', 'positive'
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 1 times)
## Summary of sample sizes: 606, 605, 604, 605, 605, 605, ...
## Resampling results across tuning parameters:
##
##
     maxdepth Accuracy
                          Kappa
##
     1
               0.6889703 0.3190082
##
     5
               0.7530403 0.3667066
               0.9107511 0.7973708
     10
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was maxdepth = 10.
  3. Model Neural Network
set.seed(825)
nn <- train(Class ~ .,
            data=data_training,
           method="nnet",
            trControl=fitControl)
## # weights: 21
## initial value 428.135581
## iter 10 value 289.303629
## iter 20 value 108.534568
## iter 30 value 54.227076
## iter 40 value 39.682967
## iter 50 value 38.809099
## iter 60 value 38.771115
```

```
## iter 70 value 38.740181
## iter 80 value 38.713385
## iter 90 value 38.706327
## iter 100 value 38.705033
## final value 38.705033
## stopped after 100 iterations
## # weights: 61
## initial value 409.748868
## iter 10 value 225.361874
## iter 20 value 55.125209
## iter 30 value 17.572002
## iter 40 value 8.456200
## iter 50 value 3.459482
## iter 60 value 2.787821
## iter 70 value 2.773113
## iter 80 value 2.772654
## iter 90 value 2.772608
## iter 100 value 2.772594
## final value 2.772594
## stopped after 100 iterations
## # weights: 101
## initial value 402.977890
## iter 10 value 87.837037
## iter 20 value 23.313795
## iter 30 value 6.976151
## iter 40 value 1.086989
## iter 50 value 0.020694
## iter 60 value 0.001078
## final value 0.000073
## converged
## # weights: 21
## initial value 467.935687
## iter 10 value 362.131248
## iter 20 value 117.552842
## iter 30 value 76.182936
## iter 40 value 72.591056
## iter 50 value 72.506759
## final value 72.506758
## converged
## # weights: 61
## initial value 446.107384
## iter 10 value 185.289024
## iter 20 value 109.555309
## iter 30 value 80.846582
## iter 40 value 69.151292
## iter 50 value 65.635869
## iter 60 value 65.225771
## iter 70 value 64.936708
## iter 80 value 64.600876
## iter 90 value 64.091111
## iter 100 value 64.013221
## final value 64.013221
## stopped after 100 iterations
## # weights: 101
```

```
## initial value 716.824579
## iter 10 value 284.802404
## iter 20 value 131.251454
## iter 30 value 78.126861
## iter 40 value 68.321864
## iter 50 value 65.440310
## iter 60 value 64.305701
## iter 70 value 63.319378
## iter 80 value 62.485986
## iter 90 value 61.417437
## iter 100 value 61.120069
## final value 61.120069
## stopped after 100 iterations
## # weights: 21
## initial value 427.759482
## iter 10 value 187.921995
## iter 20 value 54.190690
## iter 30 value 43.627902
## iter 40 value 43.304724
## iter 50 value 43.168029
## iter 60 value 43.024684
## iter 70 value 42.984117
## iter 80 value 42.881267
## iter 90 value 42.806270
## iter 100 value 42.762894
## final value 42.762894
## stopped after 100 iterations
## # weights: 61
## initial value 396.497776
## iter 10 value 125.280610
## iter 20 value 38.251504
## iter 30 value 25.187885
## iter 40 value 22.758594
## iter 50 value 22.252956
## iter 60 value 21.929126
## iter 70 value 21.692130
## iter 80 value 21.580860
## iter 90 value 20.675959
## iter 100 value 20.633867
## final value 20.633867
## stopped after 100 iterations
## # weights: 101
## initial value 409.505062
## iter 10 value 59.342845
## iter 20 value 16.811621
## iter 30 value 2.675744
## iter 40 value 1.152053
## iter 50 value 1.076752
## iter 60 value 1.023898
## iter 70 value 0.973495
## iter 80 value 0.914359
## iter 90 value 0.813829
## iter 100 value 0.762220
## final value 0.762220
```

```
## stopped after 100 iterations
## # weights: 21
## initial value 411.657119
## iter 10 value 294.492879
## iter 20 value 174.418377
## iter 30 value 68.552092
## iter 40 value 56.618336
## iter 50 value 55.767514
## iter 60 value 28.938387
## iter 70 value 25.457037
## iter 80 value 25.380148
## iter 90 value 25.364489
## iter 100 value 25.364317
## final value 25.364317
## stopped after 100 iterations
## # weights: 61
## initial value 429.743200
## iter 10 value 184.192743
## iter 20 value 45.242237
## iter 30 value 43.082135
## iter 40 value 35.263298
## iter 50 value 35.059453
## iter 60 value 34.903249
## iter 70 value 34.819326
## iter 80 value 34.740756
## iter 90 value 34.684698
## iter 100 value 34.605328
## final value 34.605328
## stopped after 100 iterations
## # weights: 101
## initial value 510.747380
## iter 10 value 241.194625
## iter 20 value 35.412051
## iter 30 value 10.690401
## iter 40 value 2.506429
## iter 50 value 1.504326
## iter 60 value 1.416288
## iter 70 value 1.399686
## iter 80 value 1.392177
## iter 90 value 1.389628
## iter 100 value 1.387394
## final value 1.387394
## stopped after 100 iterations
## # weights: 21
## initial value 557.710442
## iter 10 value 381.206606
## iter 20 value 276.248517
## iter 30 value 88.199624
## iter 40 value 72.891539
## iter 50 value 71.683216
## final value 71.639099
## converged
## # weights: 61
## initial value 490.785438
```

```
## iter 10 value 314.660999
## iter 20 value 120.252350
## iter 30 value 80.755015
## iter 40 value 66.729903
## iter 50 value 66.028259
## iter 60 value 65.238305
## iter 70 value 63.893983
## iter 80 value 63.672146
## iter 90 value 63.302650
## iter 100 value 62.341260
## final value 62.341260
## stopped after 100 iterations
## # weights: 101
## initial value 492.750018
## iter 10 value 183.714106
## iter 20 value 91.463704
## iter 30 value 66.665071
## iter 40 value 63.830026
## iter 50 value 62.799715
## iter 60 value 62.022679
## iter 70 value 61.389989
## iter 80 value 61.097181
## iter 90 value 60.805134
## iter 100 value 59.979732
## final value 59.979732
## stopped after 100 iterations
## # weights: 21
## initial value 432.214242
## iter 10 value 292.709576
## iter 20 value 210.982843
## iter 30 value 171.389787
## iter 40 value 119.642329
## iter 50 value 69.397954
## iter 60 value 45.096628
## iter 70 value 27.733291
## iter 80 value 26.428389
## iter 90 value 26.371664
## iter 100 value 26.232388
## final value 26.232388
## stopped after 100 iterations
## # weights: 61
## initial value 455.720833
## iter 10 value 165.766896
## iter 20 value 39.483005
## iter 30 value 26.528899
## iter 40 value 24.597632
## iter 50 value 21.295962
## iter 60 value 21.156429
## iter 70 value 21.061029
## iter 80 value 20.978141
## iter 90 value 20.920029
## iter 100 value 20.865165
## final value 20.865165
## stopped after 100 iterations
```

```
## # weights: 101
## initial value 400.583560
## iter 10 value 237.142619
## iter 20 value 41.460643
## iter 30 value 32.575781
## iter 40 value 28.738984
## iter 50 value 25.166708
## iter 60 value 17.666664
## iter 70 value 11.240679
## iter 80 value 6.319110
## iter 90 value 5.393882
## iter 100 value 5.046587
## final value 5.046587
## stopped after 100 iterations
## # weights: 21
## initial value 444.006507
## iter 10 value 252.815120
## iter 20 value 59.955165
## iter 30 value 32.296492
## iter 40 value 27.841836
## iter 50 value 27.575944
## iter 60 value 27.518928
## iter 70 value 27.482167
## iter 80 value 27.480886
## iter 90 value 27.480422
## iter 100 value 27.480379
## final value 27.480379
## stopped after 100 iterations
## # weights: 61
## initial value 439.652464
## iter 10 value 389.552180
## final value 389.550783
## converged
## # weights: 101
## initial value 507.973993
## iter 10 value 264.773668
## iter 20 value 58.356380
## iter 30 value 24.380534
## iter 40 value 14.970900
## iter 50 value 9.366670
## iter 60 value 4.865087
## iter 70 value 1.417054
## iter 80 value 0.100625
## iter 90 value 0.021311
## iter 100 value 0.009847
## final value 0.009847
## stopped after 100 iterations
## # weights: 21
## initial value 456.065856
## iter 10 value 296.436107
## iter 20 value 128.150785
## iter 30 value 76.780912
## iter 40 value 74.783638
## iter 50 value 73.453054
```

```
## iter 60 value 73.308278
## iter 70 value 73.306938
## iter 70 value 73.306938
## iter 70 value 73.306938
## final value 73.306938
## converged
## # weights: 61
## initial value 415.642980
## iter 10 value 288.918999
## iter 20 value 168.202984
## iter 30 value 82.980811
## iter 40 value 71.395616
## iter 50 value 68.314445
## iter 60 value 66.573317
## iter 70 value 65.398388
## iter 80 value 64.610014
## iter 90 value 64.362300
## iter 100 value 64.204179
## final value 64.204179
## stopped after 100 iterations
## # weights: 101
## initial value 387.559440
## iter 10 value 182.741553
## iter 20 value 99.106874
## iter 30 value 76.808533
## iter 40 value 71.030357
## iter 50 value 67.744976
## iter 60 value 66.210018
## iter 70 value 64.799390
## iter 80 value 63.212129
## iter 90 value 61.418052
## iter 100 value 61.024856
## final value 61.024856
## stopped after 100 iterations
## # weights: 21
## initial value 501.433590
## iter 10 value 222.953962
## iter 20 value 70.432261
## iter 30 value 30.201784
## iter 40 value 28.007653
## iter 50 value 27.963714
## iter 60 value 27.916762
## iter 70 value 27.894582
## iter 80 value 27.851320
## iter 90 value 27.846413
## iter 100 value 27.842579
## final value 27.842579
## stopped after 100 iterations
## # weights: 61
## initial value 465.408885
## iter 10 value 263.004537
## iter 20 value 33.117590
## iter 30 value 15.054783
## iter 40 value 9.392535
```

```
## iter 50 value 6.080453
## iter 60 value 4.237885
## iter 70 value 2.784670
## iter 80 value 2.152231
## iter 90 value 1.863833
## iter 100 value 1.646038
## final value 1.646038
## stopped after 100 iterations
## # weights: 101
## initial value 455.559198
## iter 10 value 244.540507
## iter 20 value 35.168925
## iter 30 value 24.838494
## iter 40 value 19.505173
## iter 50 value 16.194087
## iter 60 value 15.099177
## iter 70 value 14.542706
## iter 80 value 12.856721
## iter 90 value 12.017949
## iter 100 value 11.350726
## final value 11.350726
## stopped after 100 iterations
## # weights: 21
## initial value 396.972864
## iter 10 value 257.042716
## iter 20 value 47.444349
## iter 30 value 43.681566
## iter 40 value 39.912898
## iter 50 value 37.799205
## iter 60 value 37.072056
## iter 70 value 36.839462
## iter 80 value 36.456780
## iter 90 value 36.181248
## iter 100 value 36.088202
## final value 36.088202
## stopped after 100 iterations
## # weights: 61
## initial value 612.916830
## final value 390.610752
## converged
## # weights: 101
## initial value 502.785999
## iter 10 value 246.134663
## iter 20 value 21.049221
## iter 30 value 5.677905
## iter 40 value 0.369477
## iter 50 value 0.008507
## iter 60 value 0.000595
## final value 0.000087
## converged
## # weights: 21
## initial value 408.152374
## iter 10 value 329.348715
## iter 20 value 281.222597
```

```
## iter 30 value 181.133350
## iter 40 value 93.213679
## iter 50 value 74.087562
## iter 60 value 73.024711
## final value 73.024506
## converged
## # weights: 61
## initial value 391.771545
## iter 10 value 268.617707
## iter 20 value 101.114113
## iter 30 value 76.194381
## iter 40 value 72.331361
## iter 50 value 69.042115
## iter 60 value 66.889249
## iter 70 value 65.563836
## iter 80 value 65.096876
## iter 90 value 65.017418
## iter 100 value 64.884365
## final value 64.884365
## stopped after 100 iterations
## # weights: 101
## initial value 466.329515
## iter 10 value 214.048515
## iter 20 value 89.935527
## iter 30 value 70.849756
## iter 40 value 62.285901
## iter 50 value 60.995317
## iter 60 value 60.515542
## iter 70 value 60.148261
## iter 80 value 59.923720
## iter 90 value 59.780831
## iter 100 value 59.749840
## final value 59.749840
## stopped after 100 iterations
## # weights: 21
## initial value 402.858423
## iter 10 value 234.841933
## iter 20 value 47.101072
## iter 30 value 41.697678
## iter 40 value 40.600228
## iter 50 value 39.989645
## iter 60 value 39.799882
## iter 70 value 39.643809
## iter 80 value 39.522201
## iter 90 value 39.403313
## iter 100 value 39.316832
## final value 39.316832
## stopped after 100 iterations
## # weights: 61
## initial value 508.375150
## iter 10 value 334.792633
## iter 20 value 60.326393
## iter 30 value 42.586598
## iter 40 value 41.954306
```

```
## iter 50 value 39.606902
## iter 60 value 33.061640
## iter 70 value 23.729555
## iter 80 value 20.933973
## iter 90 value 18.974371
## iter 100 value 17.383742
## final value 17.383742
## stopped after 100 iterations
## # weights: 101
## initial value 394.252639
## iter 10 value 106.238662
## iter 20 value 21.428034
## iter 30 value 9.169676
## iter 40 value 4.330583
## iter 50 value 4.190665
## iter 60 value 4.066497
## iter 70 value 3.972582
## iter 80 value 3.807971
## iter 90 value 3.710709
## iter 100 value 3.657040
## final value 3.657040
## stopped after 100 iterations
## # weights: 21
## initial value 438.052718
## iter 10 value 315.010099
## iter 20 value 66.938302
## iter 30 value 36.004817
## iter 40 value 33.410900
## iter 50 value 33.313485
## iter 60 value 33.312979
## iter 70 value 33.294215
## iter 80 value 33.286280
## iter 90 value 33.273693
## iter 100 value 33.272135
## final value 33.272135
## stopped after 100 iterations
## # weights: 61
## initial value 476.888907
## iter 10 value 299.704510
## iter 20 value 57.213615
## iter 30 value 21.606972
## iter 40 value 10.334798
## iter 50 value 8.775352
## iter 60 value 8.729450
## iter 70 value 8.728462
## iter 80 value 8.726730
## iter 90 value 8.725537
## iter 100 value 8.724163
## final value 8.724163
## stopped after 100 iterations
## # weights: 101
## initial value 452.030607
## iter 10 value 168.779988
## iter 20 value 19.683509
```

```
## iter 30 value 9.611372
## iter 40 value 4.978332
## iter 50 value 3.285333
## iter 60 value 2.833183
## iter 70 value 2.752811
## iter 80 value 2.746720
## iter 90 value 2.743843
## iter 100 value 2.738909
## final value 2.738909
## stopped after 100 iterations
## # weights: 21
## initial value 422.411570
## iter 10 value 303.667087
## iter 20 value 110.340182
## iter 30 value 75.949846
## iter 40 value 68.164933
## iter 50 value 68.111498
## iter 60 value 68.109888
## final value 68.109858
## converged
## # weights: 61
## initial value 434.509779
## iter 10 value 295.874390
## iter 20 value 121.617421
## iter 30 value 72.319881
## iter 40 value 61.204065
## iter 50 value 59.572369
## iter 60 value 58.808003
## iter 70 value 58.546371
## iter 80 value 58.489822
## iter 90 value 58.460880
## iter 100 value 58.426295
## final value 58.426295
## stopped after 100 iterations
## # weights: 101
## initial value 445.701218
## iter 10 value 143.235497
## iter 20 value 83.152291
## iter 30 value 67.679588
## iter 40 value 62.274566
## iter 50 value 60.552586
## iter 60 value 59.193632
## iter 70 value 58.628019
## iter 80 value 58.224022
## iter 90 value 57.767746
## iter 100 value 57.556517
## final value 57.556517
## stopped after 100 iterations
## # weights: 21
## initial value 390.917599
## iter 10 value 187.309494
## iter 20 value 78.845651
## iter 30 value 24.520086
## iter 40 value 21.897360
```

```
## iter 50 value 21.895519
## iter 60 value 21.891619
## iter 70 value 21.890243
## iter 80 value 21.890006
## iter 90 value 21.889757
## iter 90 value 21.889757
## iter 90 value 21.889757
## final value 21.889757
## converged
## # weights: 61
## initial value 424.706816
## iter 10 value 172.516194
## iter 20 value 36.387139
## iter 30 value 24.662978
## iter 40 value 17.912885
## iter 50 value 12.219917
## iter 60 value 11.794813
## iter 70 value 11.732984
## iter 80 value 11.660048
## iter 90 value 11.587682
## iter 100 value 11.554150
## final value 11.554150
## stopped after 100 iterations
## # weights: 101
## initial value 429.733007
## iter 10 value 83.303347
## iter 20 value 19.350366
## iter 30 value 3.822145
## iter 40 value 1.326291
## iter 50 value 1.140141
## iter 60 value 0.937164
## iter 70 value 0.827472
## iter 80 value 0.734473
## iter 90 value 0.652112
## iter 100 value 0.573117
## final value 0.573117
## stopped after 100 iterations
## # weights: 21
## initial value 516.954277
## iter 10 value 358.825723
## iter 20 value 107.037363
## iter 30 value 33.904006
## iter 40 value 29.833607
## iter 50 value 29.678090
## iter 60 value 29.639371
## iter 70 value 29.629549
## iter 80 value 29.627850
## iter 90 value 29.627537
## final value 29.627117
## converged
## # weights: 61
## initial value 471.612155
## iter 10 value 390.614733
## iter 20 value 390.610694
```

```
## final value 390.610687
## converged
## # weights: 101
## initial value 385.292934
## iter 10 value 148.711171
## iter 20 value 28.358579
## iter 30 value 11.192443
## iter 40 value 1.292816
## iter 50 value 0.086804
## iter 60 value 0.002238
## iter 70 value 0.000406
## iter 80 value 0.000134
## final value 0.000088
## converged
## # weights: 21
## initial value 461.281690
## iter 10 value 252.618686
## iter 20 value 132.816218
## iter 30 value 78.404647
## iter 40 value 68.878834
## iter 50 value 67.963775
## iter 60 value 67.878306
## iter 70 value 67.877180
## iter 70 value 67.877180
## iter 70 value 67.877180
## final value 67.877180
## converged
## # weights: 61
## initial value 420.492664
## iter 10 value 194.682143
## iter 20 value 116.140799
## iter 30 value 70.903027
## iter 40 value 67.795295
## iter 50 value 65.990334
## iter 60 value 64.821808
## iter 70 value 63.430792
## iter 80 value 62.439375
## iter 90 value 61.867082
## iter 100 value 61.199695
## final value 61.199695
## stopped after 100 iterations
## # weights: 101
## initial value 405.297607
## iter 10 value 220.591309
## iter 20 value 98.272079
## iter 30 value 73.355841
## iter 40 value 64.761076
## iter 50 value 61.948190
## iter 60 value 60.767626
## iter 70 value 60.039271
## iter 80 value 59.407898
## iter 90 value 58.876913
## iter 100 value 57.944621
## final value 57.944621
```

```
## stopped after 100 iterations
## # weights: 21
## initial value 448.141047
## iter 10 value 203.660792
## iter 20 value 41.083392
## iter 30 value 31.808840
## iter 40 value 31.420004
## iter 50 value 31.291732
## iter 60 value 31.127682
## iter 70 value 31.062023
## iter 80 value 30.586214
## iter 90 value 25.411297
## iter 100 value 23.024168
## final value 23.024168
## stopped after 100 iterations
## # weights: 61
## initial value 510.918830
## iter 10 value 378.139032
## iter 20 value 221.990209
## iter 30 value 83.963141
## iter 40 value 64.969787
## iter 50 value 42.147125
## iter 60 value 16.488592
## iter 70 value 10.541265
## iter 80 value 8.291379
## iter 90 value 8.102661
## iter 100 value 7.960948
## final value 7.960948
## stopped after 100 iterations
## # weights: 101
## initial value 407.916654
## iter 10 value 93.845660
## iter 20 value 26.647392
## iter 30 value 19.781555
## iter 40 value 10.234929
## iter 50 value 2.391242
## iter 60 value 1.391115
## iter 70 value 1.213528
## iter 80 value 1.049890
## iter 90 value 0.922240
## iter 100 value 0.853647
## final value 0.853647
## stopped after 100 iterations
## # weights: 21
## initial value 392.978324
## iter 10 value 119.795614
## iter 20 value 39.494050
## iter 30 value 21.385659
## iter 40 value 20.642584
## iter 50 value 20.609246
## iter 60 value 20.546000
## iter 70 value 20.478578
## iter 80 value 20.447796
## iter 90 value 20.423517
```

```
## iter 100 value 20.422856
## final value 20.422856
## stopped after 100 iterations
## # weights: 61
## initial value 404.109059
## iter 10 value 149.042022
## iter 20 value 33.012405
## iter 30 value 27.315754
## iter 40 value 17.528935
## iter 50 value 10.106429
## iter 60 value 5.230756
## iter 70 value 4.165854
## iter 80 value 3.913260
## iter 90 value 3.838856
## iter 100 value 3.829890
## final value 3.829890
## stopped after 100 iterations
## # weights: 101
## initial value 498.076913
## iter 10 value 172.112514
## iter 20 value 33.218582
## iter 30 value 28.298557
## iter 40 value 24.367890
## iter 50 value 19.480100
## iter 60 value 13.637468
## iter 70 value 12.616877
## iter 80 value 10.845451
## iter 90 value 8.568159
## iter 100 value 8.087527
## final value 8.087527
## stopped after 100 iterations
## # weights: 21
## initial value 460.928952
## iter 10 value 251.233917
## iter 20 value 186.632399
## iter 30 value 75.742340
## iter 40 value 68.064460
## iter 50 value 67.195691
## iter 60 value 67.005933
## iter 70 value 67.003866
## final value 67.003866
## converged
## # weights: 61
## initial value 503.792439
## iter 10 value 259.810387
## iter 20 value 108.029932
## iter 30 value 71.232788
## iter 40 value 63.844764
## iter 50 value 60.114082
## iter 60 value 59.168919
## iter 70 value 59.025492
## iter 80 value 58.959896
## iter 90 value 58.644063
## final value 58.641107
```

```
## converged
## # weights: 101
## initial value 437.573082
## iter 10 value 253.915618
## iter 20 value 125.486580
## iter 30 value 72.641756
## iter 40 value 61.922841
## iter 50 value 59.774944
## iter 60 value 57.909022
## iter 70 value 56.946510
## iter 80 value 56.418659
## iter 90 value 56.219353
## iter 100 value 56.177302
## final value 56.177302
## stopped after 100 iterations
## # weights: 21
## initial value 419.981746
## iter 10 value 283.603715
## iter 20 value 192.328834
## iter 30 value 174.843414
## iter 40 value 42.510365
## iter 50 value 36.444511
## iter 60 value 36.333147
## iter 70 value 36.285637
## iter 80 value 36.254856
## iter 90 value 36.229176
## iter 100 value 36.196157
## final value 36.196157
## stopped after 100 iterations
## # weights: 61
## initial value 549.469156
## iter 10 value 187.590726
## iter 20 value 40.507182
## iter 30 value 33.909470
## iter 40 value 28.392888
## iter 50 value 15.750207
## iter 60 value 9.382771
## iter 70 value 8.502960
## iter 80 value 5.426419
## iter 90 value 3.692690
## iter 100 value 3.166885
## final value 3.166885
## stopped after 100 iterations
## # weights: 101
## initial value 579.784317
## iter 10 value 228.832184
## iter 20 value 31.118032
## iter 30 value 14.039463
## iter 40 value 11.629221
## iter 50 value 10.522687
## iter 60 value 9.907288
## iter 70 value 9.543127
## iter 80 value 8.815738
## iter 90 value 8.420401
```

```
## iter 100 value 8.238160
## final value 8.238160
## stopped after 100 iterations
## # weights: 21
## initial value 437.509703
## iter 10 value 277.783520
## iter 20 value 64.534698
## iter 30 value 39.585531
## iter 40 value 27.492955
## iter 50 value 25.605752
## iter 60 value 25.407138
## iter 70 value 25.397713
## final value 25.397708
## converged
## # weights: 61
## initial value 485.302828
## iter 10 value 211.774613
## iter 20 value 83.672244
## iter 30 value 22.848393
## iter 40 value 18.109513
## iter 50 value 17.731317
## iter 60 value 17.314943
## iter 70 value 16.185220
## iter 80 value 16.009755
## iter 90 value 15.873639
## iter 100 value 15.825861
## final value 15.825861
## stopped after 100 iterations
## # weights: 101
## initial value 440.777586
## iter 10 value 156.755315
## iter 20 value 15.165217
## iter 30 value 8.626069
## iter 40 value 5.340378
## iter 50 value 3.118578
## iter 60 value 0.762540
## iter 70 value 0.121455
## iter 80 value 0.048836
## iter 90 value 0.023429
## iter 100 value 0.015485
## final value 0.015485
## stopped after 100 iterations
## # weights: 21
## initial value 421.786173
## iter 10 value 304.849212
## iter 20 value 147.587937
## iter 30 value 74.949416
## iter 40 value 71.821561
## iter 50 value 71.626992
## final value 71.626980
## converged
## # weights: 61
## initial value 408.670675
## iter 10 value 247.228868
```

```
## iter 20 value 122.451507
## iter 30 value 83.249990
## iter 40 value 74.729114
## iter 50 value 68.375074
## iter 60 value 66.611558
## iter 70 value 66.062641
## iter 80 value 65.264740
## iter 90 value 64.988902
## iter 100 value 64.599557
## final value 64.599557
## stopped after 100 iterations
## # weights: 101
## initial value 391.668643
## iter 10 value 222.200019
## iter 20 value 103.302269
## iter 30 value 72.844250
## iter 40 value 64.652312
## iter 50 value 63.126121
## iter 60 value 62.545917
## iter 70 value 61.856639
## iter 80 value 60.383705
## iter 90 value 59.983009
## iter 100 value 59.861455
## final value 59.861455
## stopped after 100 iterations
## # weights: 21
## initial value 406.960246
## iter 10 value 173.111090
## iter 20 value 39.751251
## iter 30 value 36.230818
## iter 40 value 35.859999
## iter 50 value 35.683754
## iter 60 value 35.516635
## iter 70 value 35.453819
## iter 80 value 35.337440
## iter 90 value 35.152194
## iter 100 value 35.069106
## final value 35.069106
## stopped after 100 iterations
## # weights: 61
## initial value 428.410028
## iter 10 value 188.953366
## iter 20 value 38.624474
## iter 30 value 25.888535
## iter 40 value 20.256338
## iter 50 value 19.078202
## iter 60 value 17.625551
## iter 70 value 14.572934
## iter 80 value 14.251166
## iter 90 value 14.116950
## iter 100 value 14.049671
## final value 14.049671
## stopped after 100 iterations
## # weights: 101
```

```
## initial value 514.198452
## iter 10 value 198.324012
## iter 20 value 69.665690
## iter 30 value 13.867371
## iter 40 value 8.249255
## iter 50 value 6.069297
## iter 60 value 5.630271
## iter 70 value 5.218450
## iter 80 value 4.839793
## iter 90 value 4.628133
## iter 100 value 2.757298
## final value 2.757298
## stopped after 100 iterations
## # weights: 21
## initial value 448.268486
## iter 10 value 226.968587
## iter 20 value 67.432583
## iter 30 value 27.253746
## iter 40 value 24.431604
## iter 50 value 24.250842
## iter 60 value 24.248179
## iter 70 value 24.247660
## final value 24.247656
## converged
## # weights: 61
## initial value 425.630529
## iter 10 value 88.449342
## iter 20 value 32.336614
## iter 30 value 21.245709
## iter 40 value 17.472783
## iter 50 value 14.978417
## iter 60 value 12.019135
## iter 70 value 10.366781
## iter 80 value 10.217504
## iter 90 value 10.200555
## iter 100 value 10.193970
## final value 10.193970
## stopped after 100 iterations
## # weights: 101
## initial value 454.479383
## iter 10 value 77.277857
## iter 20 value 16.526992
## iter 30 value 2.289291
## iter 40 value 0.046670
## iter 50 value 0.001642
## iter 60 value 0.000283
## final value 0.000096
## converged
## # weights: 21
## initial value 528.060178
## iter 10 value 333.898486
## iter 20 value 140.890916
## iter 30 value 78.294825
## iter 40 value 70.521899
```

```
## iter 50 value 70.011462
## iter 60 value 69.896301
## final value 69.895204
## converged
## # weights: 61
## initial value 394.935817
## iter 10 value 200.225509
## iter 20 value 115.858937
## iter 30 value 72.902786
## iter 40 value 67.859849
## iter 50 value 65.498674
## iter 60 value 65.131528
## iter 70 value 64.785067
## iter 80 value 63.717930
## iter 90 value 62.988262
## iter 100 value 62.603206
## final value 62.603206
## stopped after 100 iterations
## # weights: 101
## initial value 398.071017
## iter 10 value 165.071416
## iter 20 value 86.469069
## iter 30 value 73.692899
## iter 40 value 71.092563
## iter 50 value 68.522472
## iter 60 value 65.902246
## iter 70 value 63.855087
## iter 80 value 59.708117
## iter 90 value 57.889865
## iter 100 value 57.404710
## final value 57.404710
## stopped after 100 iterations
## # weights: 21
## initial value 456.213210
## iter 10 value 332.407962
## iter 20 value 113.514553
## iter 30 value 50.211239
## iter 40 value 28.857073
## iter 50 value 26.555070
## iter 60 value 26.234337
## iter 70 value 26.103867
## iter 80 value 25.774571
## iter 90 value 25.623769
## iter 100 value 25.582323
## final value 25.582323
## stopped after 100 iterations
## # weights: 61
## initial value 396.343083
## iter 10 value 153.620907
## iter 20 value 41.832998
## iter 30 value 24.251244
## iter 40 value 20.433163
## iter 50 value 19.127143
## iter 60 value 18.891564
```

```
## iter 70 value 18.799113
## iter 80 value 18.227767
## iter 90 value 17.308913
## iter 100 value 16.614500
## final value 16.614500
## stopped after 100 iterations
## # weights: 101
## initial value 434.981939
## iter 10 value 103.203698
## iter 20 value 28.780425
## iter 30 value 8.990703
## iter 40 value 3.520549
## iter 50 value 3.007868
## iter 60 value 2.849576
## iter 70 value 2.517832
## iter 80 value 2.129184
## iter 90 value 1.981326
## iter 100 value 1.820086
## final value 1.820086
## stopped after 100 iterations
## # weights: 21
## initial value 401.251377
## iter 10 value 291.277956
## iter 20 value 196.963314
## iter 30 value 194.396323
## iter 40 value 193.315774
## iter 50 value 193.041302
## iter 60 value 191.697756
## iter 70 value 191.496332
## iter 80 value 165.691905
## iter 90 value 43.917431
## iter 100 value 38.878414
## final value 38.878414
## stopped after 100 iterations
## # weights: 61
## initial value 449.141392
## iter 10 value 288.539594
## iter 20 value 62.221090
## iter 30 value 25.287415
## iter 40 value 17.866214
## iter 50 value 15.830695
## iter 60 value 15.333168
## iter 70 value 15.211025
## iter 80 value 15.132177
## iter 90 value 14.419467
## iter 100 value 14.410230
## final value 14.410230
## stopped after 100 iterations
## # weights: 101
## initial value 505.929114
## iter 10 value 68.504083
## iter 20 value 21.859211
## iter 30 value 11.412750
## iter 40 value 1.848571
```

```
## iter 50 value 0.052427
## iter 60 value 0.000813
## iter 70 value 0.000121
## final value 0.000090
## converged
## # weights: 21
## initial value 391.144625
## iter 10 value 185.780818
## iter 20 value 86.249546
## iter 30 value 73.714360
## iter 40 value 71.930337
## final value 71.927929
## converged
## # weights: 61
## initial value 442.209949
## iter 10 value 175.749230
## iter 20 value 113.870398
## iter 30 value 74.078843
## iter 40 value 65.860509
## iter 50 value 64.330625
## iter 60 value 63.685593
## iter 70 value 63.502369
## iter 80 value 63.417238
## iter 90 value 63.301681
## iter 100 value 63.301320
## final value 63.301320
## stopped after 100 iterations
## # weights: 101
## initial value 432.769872
## iter 10 value 284.659839
## iter 20 value 118.396346
## iter 30 value 76.893786
## iter 40 value 66.958897
## iter 50 value 63.843132
## iter 60 value 60.647629
## iter 70 value 59.426229
## iter 80 value 58.998312
## iter 90 value 58.881067
## iter 100 value 58.855462
## final value 58.855462
## stopped after 100 iterations
## # weights: 21
## initial value 403.944254
## iter 10 value 181.931891
## iter 20 value 54.683164
## iter 30 value 26.525177
## iter 40 value 26.140896
## iter 50 value 26.135110
## iter 60 value 26.114858
## iter 70 value 26.107409
## iter 80 value 26.104516
## iter 90 value 26.103555
## iter 100 value 26.103469
## final value 26.103469
```

```
## stopped after 100 iterations
## # weights: 61
## initial value 393.283009
## iter 10 value 93.040297
## iter 20 value 32.907280
## iter 30 value 17.185625
## iter 40 value 9.326391
## iter 50 value 7.126204
## iter 60 value 6.308960
## iter 70 value 6.113914
## iter 80 value 5.997302
## iter 90 value 5.943665
## iter 100 value 5.890769
## final value 5.890769
## stopped after 100 iterations
## # weights: 101
## initial value 402.657075
## iter 10 value 43.741724
## iter 20 value 21.033373
## iter 30 value 13.099429
## iter 40 value 5.140715
## iter 50 value 2.702058
## iter 60 value 2.449590
## iter 70 value 2.305706
## iter 80 value 2.132539
## iter 90 value 1.995674
## iter 100 value 1.889234
## final value 1.889234
## stopped after 100 iterations
## # weights: 61
## initial value 427.695589
## iter 10 value 190.238193
## iter 20 value 101.708924
## iter 30 value 81.980341
## iter 40 value 76.535008
## iter 50 value 74.740896
## iter 60 value 71.486805
## iter 70 value 68.441450
## iter 80 value 66.603724
## iter 90 value 65.044527
## iter 100 value 64.484083
## final value 64.484083
## stopped after 100 iterations
nn
## Neural Network
##
## 672 samples
##
    9 predictor
     2 classes: 'negative', 'positive'
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 1 times)
## Summary of sample sizes: 606, 605, 604, 605, 605, 605, ...
```

```
## Resampling results across tuning parameters:
##
##
     size decay Accuracy
                             Kappa
           0e+00 0.9732221 0.9402993
##
##
           1e-04 0.9717296 0.9369543
##
           1e-01 0.9776778 0.9498311
    1
          0e+00 0.8646593 0.6307553
##
    3
          1e-04 0.9612592 0.9140460
##
     3
##
     3
          1e-01 0.9776997 0.9499157
##
          0e+00 0.9598771 0.9103137
     5
          1e-04 0.9582954 0.9085208
           1e-01 0.9761846 0.9466124
##
     5
##
## Accuracy was used to select the optimal model using the largest value.
## The final values used for the model were size = 3 and decay = 0.1.
  4. Model Nearest Neighbour
set.seed(825)
knn <- train(Class ~ .,
             data=data_training,
             method="knn",
             trControl=fitControl)
knn
## k-Nearest Neighbors
##
## 672 samples
##
    9 predictor
     2 classes: 'negative', 'positive'
##
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 1 times)
## Summary of sample sizes: 606, 605, 604, 605, 605, 605, ...
## Resampling results across tuning parameters:
##
##
    k Accuracy
                   Kappa
##
    5 0.9420751 0.8667396
    7 0.8066433 0.5374263
##
##
    9 0.7709534 0.4451248
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was k = 5.
  5. Model SVM (linear kernel)
set.seed(825)
svm <- train(Class ~ .,</pre>
             data=data_training,
             method="svmLinear",
             trControl=fitControl)
svm
## Support Vector Machines with Linear Kernel
##
## 672 samples
   9 predictor
```

```
2 classes: 'negative', 'positive'
##
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 1 times)
## Summary of sample sizes: 606, 605, 604, 605, 605, 605, ...
## Resampling results:
##
##
     Accuracy
                Kappa
##
     0.9806629 0.9563793
##
## Tuning parameter 'C' was held constant at a value of 1
Collect the results for all the models.
resamps <- resamples(list("Naive Bayes"=nb,
                           "Decision Tree"=dt,
                           "Neural Network"=nn,
                           "Nearest Neighbour"=knn,
                           "SVM (linear kernel)"=svm))
summary(resamps)
##
## Call:
## summary.resamples(object = resamps)
## Models: Naive Bayes, Decision Tree, Neural Network, Nearest Neighbour, SVM (linear kernel)
## Number of resamples: 10
##
## Accuracy
##
                            Min.
                                    1st Qu.
                                               Median
                                                           Mean
                                                                   3rd Qu.
                       0.6567164 0.6642340 0.6865672 0.6845565 0.6943691 0.7205882
## Naive Bayes
## Decision Tree
                       0.8507463\ 0.9000668\ 0.9104478\ 0.9107511\ 0.9253731\ 0.9701493
                       0.9552239 0.9702590 0.9850746 0.9776997 0.9852392 1.0000000
## Neural Network
                       0.8955224 0.9188982 0.9402985 0.9420751 0.9700362 0.9850746
## Nearest Neighbour
## SVM (linear kernel) 0.9552239 0.9702590 0.9850746 0.9806629 0.9852392 1.0000000
##
                       NA's
## Naive Bayes
                           0
                           0
## Decision Tree
## Neural Network
                           0
## Nearest Neighbour
                           0
## SVM (linear kernel)
                           0
##
## Kappa
##
                                                Median
                            Min.
                                     1st Qu.
                                                            Mean
                       0.0000000 0.05403608 0.1111813 0.1130575 0.1504189
## Naive Bayes
                       0.6469968 0.77864355 0.7984862 0.7973708 0.8318554
## Decision Tree
## Neural Network
                       0.8975013 0.93288438 0.9665502 0.9499157 0.9672590
## Nearest Neighbour
                       0.7556019 0.81184946 0.8647830 0.8667396 0.9330473
## SVM (linear kernel) 0.8975013 0.93288438 0.9665502 0.9563793 0.9672590
                            Max. NA's
## Naive Bayes
                       0.2540416
                                     0
## Decision Tree
                       0.9337945
                                     0
## Neural Network
                       1.0000000
                                     0
## Nearest Neighbour
                                     0
                       0.9665502
## SVM (linear kernel) 1.0000000
                                     0
```

Complete the following table with the final values of accuracy and kappa for the training data:

Question 4

##

Apply the models: Naive Bayes, Decision Tree, Neural Networks, Nearest Neighbour and SVM (linear kernel) to the data testing dataset. Print the confusion matrix of each model.

1. Model Naive Bayes

```
nbPredict <- predict(nb, newdata=data_testing)</pre>
confusionMatrix(nbPredict, data_testing$Class)
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction negative positive
##
     negative
                     7
                               0
##
     positive
                    92
                             187
##
##
                  Accuracy : 0.6783
##
                    95% CI: (0.6208, 0.7321)
##
       No Information Rate: 0.6538
       P-Value [Acc > NIR] : 0.2102
##
##
##
                      Kappa: 0.0905
##
    Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 0.07071
##
##
               Specificity: 1.00000
##
            Pos Pred Value: 1.00000
##
            Neg Pred Value: 0.67025
##
                Prevalence: 0.34615
##
            Detection Rate: 0.02448
##
      Detection Prevalence: 0.02448
         Balanced Accuracy: 0.53535
##
##
##
          'Positive' Class : negative
##
  2. Model Decision Tree
dtPredict <- predict(dt, newdata=data_testing)</pre>
confusionMatrix(dtPredict, data_testing$Class)
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction negative positive
##
     negative
                    92
                              14
                      7
     positive
                             173
##
##
##
                  Accuracy : 0.9266
##
                    95% CI: (0.8899, 0.954)
##
       No Information Rate: 0.6538
       P-Value [Acc > NIR] : <2e-16
##
```

```
##
                     Kappa: 0.8404
##
##
    Mcnemar's Test P-Value: 0.1904
##
##
               Sensitivity: 0.9293
               Specificity: 0.9251
##
##
            Pos Pred Value: 0.8679
            Neg Pred Value: 0.9611
##
##
                Prevalence: 0.3462
            Detection Rate: 0.3217
##
##
      Detection Prevalence: 0.3706
##
         Balanced Accuracy: 0.9272
##
##
          'Positive' Class : negative
##
  3. Model Neural Network
nnPredict <- predict(nn, newdata=data_testing)</pre>
confusionMatrix(nnPredict, data_testing$Class)
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction negative positive
##
     negative
                    96
     positive
                     3
                             187
##
##
##
                  Accuracy: 0.9895
##
                    95% CI: (0.9697, 0.9978)
##
       No Information Rate: 0.6538
##
       P-Value [Acc > NIR] : <2e-16
##
                     Kappa: 0.9767
##
##
##
    Mcnemar's Test P-Value: 0.2482
##
##
               Sensitivity: 0.9697
               Specificity: 1.0000
##
##
            Pos Pred Value: 1.0000
##
            Neg Pred Value: 0.9842
##
                Prevalence: 0.3462
##
            Detection Rate: 0.3357
##
      Detection Prevalence: 0.3357
##
         Balanced Accuracy: 0.9848
##
##
          'Positive' Class : negative
##
  4. Model Nearest Neighbour
knnPredict <- predict(knn, newdata=data_testing)</pre>
confusionMatrix(knnPredict, data_testing$Class)
## Confusion Matrix and Statistics
```

##

```
##
             Reference
## Prediction negative positive
     negative
##
                    92
     positive
                             187
##
##
##
                  Accuracy: 0.9755
##
                    95% CI: (0.9502, 0.9901)
       No Information Rate: 0.6538
##
##
       P-Value [Acc > NIR] : < 2e-16
##
##
                     Kappa: 0.945
##
    Mcnemar's Test P-Value: 0.02334
##
##
##
               Sensitivity: 0.9293
##
               Specificity: 1.0000
##
            Pos Pred Value: 1.0000
##
            Neg Pred Value: 0.9639
##
                Prevalence: 0.3462
            Detection Rate: 0.3217
##
##
      Detection Prevalence: 0.3217
##
         Balanced Accuracy: 0.9646
##
##
          'Positive' Class : negative
##
  5. Model SVM (linear kernel)
svmPredict <- predict(svm, newdata=data_testing)</pre>
confusionMatrix(svmPredict, data_testing$Class)
## Confusion Matrix and Statistics
##
##
             Reference
  Prediction negative positive
##
##
     negative
                    96
##
     positive
                     3
                             187
##
##
                  Accuracy : 0.9895
                    95% CI: (0.9697, 0.9978)
##
##
       No Information Rate: 0.6538
##
       P-Value [Acc > NIR] : <2e-16
##
##
                     Kappa: 0.9767
##
##
    Mcnemar's Test P-Value: 0.2482
##
##
               Sensitivity: 0.9697
##
               Specificity: 1.0000
##
            Pos Pred Value: 1.0000
##
            Neg Pred Value: 0.9842
                Prevalence: 0.3462
##
##
            Detection Rate: 0.3357
##
      Detection Prevalence: 0.3357
##
         Balanced Accuracy: 0.9848
```

```
##
## 'Positive' Class : negative
##
Calculate the AUC value for all the models.
auc(roc(nbPredict, data_testing$Class))
## [1] 0.5353535
auc(roc(dtPredict, data_testing$Class))
## [1] 0.9272133
auc(roc(nnPredict, data_testing$Class))
## [1] 0.9848485
auc(roc(knnPredict, data_testing$Class))
## [1] 0.9646465
auc(roc(svmPredict, data_testing$Class))
## [1] 0.9848485
```

Complete the following table with the final values of accuracy, kappa and AUC for the testing data.

Question 5

Plot the ROC curves of the models.

- a) Calculate again the predictions on the test set but now setting the type parameter of the predict function to "prob".
- 1. Model Naive Bayes

```
nbPredictProb <- predict(nb, newdata=data_testing, type = "prob")
head(nbPredictProb)

## negative positive</pre>
```

```
## 1 0.105410725 0.8945893
## 2 0.048803363 0.9511966
## 3 0.002235508 0.9977645
## 4 0.004361079 0.9956389
## 5 0.001757359 0.9982426
## 6 0.012029302 0.9879707
```

2. Model Decision Tree

```
dtPredictProb <- predict(dt, newdata=data_testing, type = "prob")
head(dtPredictProb)</pre>
```

```
##
      negative positive
## 9
           0.00
                     1.00
## 11
           0.16
                     0.84
           0.00
                     1.00
## 13
## 16
           0.00
                     1.00
## 17
           0.00
                     1.00
## 20
           0.16
                     0.84
```

3. Model Neural Network

```
nnPredictProb <- predict(nn, newdata=data_testing, type = "prob")</pre>
head(nnPredictProb)
##
         negative positive
## 9 0.014915208 0.9850848
## 11 0.022068633 0.9779314
## 13 0.018293877 0.9817061
## 16 0.009860493 0.9901395
## 17 0.005332153 0.9946678
## 20 0.015123710 0.9848763
  4. Model Nearest Neighbour
knnPredictProb <- predict(knn, newdata=data_testing, type = "prob")</pre>
head(knnPredictProb)
##
     negative positive
## 1
            0
## 2
            0
            0
## 3
                      1
            0
            0
## 5
                      1
            0
## 6
  5. Model SVM (linear kernel)
svmPredictProb <- predict(svm, newdata=data_testing, type = "prob")</pre>
head(svmPredictProb)
##
       negative positive
## 1 0.03087548 0.9691245
## 2 0.03086058 0.9691394
## 3 0.03082860 0.9691714
## 4 0.03084267 0.9691573
## 5 0.03084003 0.9691600
## 6 0.03085089 0.9691491
  b) Construct a "prediction" object for each classifier using the vector of estimated probabilities for the
     positive class as the first parameter, and the vector of actual class labels as the second parameter.
  1. Model Naive Bayes
nbPred <- prediction(nbPredictProb$positive, data_testing$Class)
  2. Model Decision Tree
dtPred <- prediction(dtPredictProb$positive, data_testing$Class)</pre>
  3. Model Neural Network
nnPred <- prediction(nnPredictProb$positive, data_testing$Class)
  4. Model Nearest Neighbour
knnPred <- prediction(knnPredictProb$positive, data_testing$Class)
  5. Model SVM (linear kernel)
svmPred <- prediction(svmPredictProb$positive, data_testing$Class)</pre>
```

- c) Calculate the measures we want to plot on the y-axis (TPR) and on the x-axis (FPR) by using the performance function.
- 1. Model Naive Bayes

```
nbPerf <- performance(nbPred, "tpr", "fpr")</pre>
```

2. Model Decision Tree

```
dtPerf <- performance(dtPred, "tpr", "fpr")</pre>
```

3. Model Neural Network

```
nnPerf <- performance(nnPred, "tpr", "fpr")</pre>
```

4. Model Nearest Neighbour

```
knnPerf <- performance(knnPred, "tpr", "fpr")</pre>
```

5. Model SVM (linear kernel)

```
svmPerf <- performance(svmPred, "tpr", "fpr")</pre>
```

d) Draw all the curves in the same plot.

Curvas ROC

