ISL Assignment -2-Q3

Question 3:

a) First, do the entire steps discussed in https://rpubs.com/pparacch/237109 to do naive Bayes classification on a dataset consisting of SMS messages. The data set on SMS messages is discussed at http://www.dt.fee.unicamp.br/~tiago/smsspamcollection/ and can be downloaded from http://www.dt.fee.unicamp.br/~tiago/smsspamcollection/smsspamcollection.zip

Solution:

> tmp2 <- read.csv("~/R/tmp2.txt")</pre>

```
View(tmp2)
> rawData<-tmp2</pre>
  str(rawData)
                     5573 obs. of 2 variables:
 'data.frame':
 $ ham : Factor w/ 2 levels "ham", "spam": 1 2 1 1 2 1 1 2 2 1 ...
$ Go.until.jurong.point: Factor w/ 4967 levels "", "'AH POOR BABY!HOPE URFEELING BETTERSN LUV! PRO
BTHAT OVERDOSE OF WORK HEY GO CAREFUL SPK 2 U SN LOTS OF LOVEJEN XXX.'",..: 3137 1033 4112 2797 10 58 967 437 4591 1250 1761 ... > colnames(rawData) <- c("type", "text")
> rawData$text <- iconv(rawData$text, to = "utf-8")
> rawData$type <- factor(rawData$type)</pre>
> summary(rawData)
                  text
Length:5573
    type
 ham :4826
 spam: 747
                   Class:character
                   Mode :character
> table(rawData$type)
 ham spam
1826 747
> prop.table(table(rawData$type)) * 100
       ham
86.59609 13.40391
> set.seed(0695)
> trainIndex <- createDataPartition(rawData$type, p = .80,</pre>
                                                    list = FALSE,
                                                    times = 1)
> trainData <- rawData[trainIndex,]</pre>
> testData <- rawData[-trainIndex,]</pre>
> prop.table(table(trainData$type)) * 100
86.58892 13.41108
> prop.table(table(testData$type)) * 100
ham spam
86.62478 13.37522
> trainData_ham <- trainData[trainData$type == "ham",]</pre>
> trainbata_nam <- trainbata_trainbata_type -- nam ,,
> head(trainData_ham$text)
[1] "Ok lar... Joking wif u oni..."
[2] "U dun say so early hor... U c already then say..."
[3] "Nah I don't think he goes to usf"
[4] "Even my brother is not like to speak with me. They treat me like aids patent."
[5] "Time come have soon and i don't want to talk about this stuff anymore tonig
 [5] "I'm gonna be home soon and i don't want to talk about this stuff anymore tonight"
[6] "I've been searching for the right words to thank you for this breather. I promise i wont take your help for granted and will fulfil my promise. You have been wonderful and a blessing at all ti
> tail(trainData_ham$text)
[1] "Ok lor... Sony ericsson salesman... I ask shuhui then she say quite gd 2 use so i considering
     "Ard 6 like dat lor."
"Huh y lei..."
     "Will AfA% b going to esplanade fr home?"
     "Pity"
[6] "The guy did some bitching but I acted like i'd be interested in buying something else next we
ek and he gave it to us for free"
> trainData_spam <- trainData[trainData$type == "spam",]</pre>
   head(trainData_spam$text)
[1] "Free entry in 2 a wkly comp to win FA Cup final thts 21st May 2005. Text FA to 87121 to recei
ve_entry question(std txt rate)T&C's apply 08452810075over18's"
                                                              week's now and no word back! I'd like some fun you up f
                          here darl
or it still? Tb ok! XxX std chgs to send"
[3] "WINNER!! As a valued network customer you have been selected to receivea Ã,£900 prize reward! To claim call 09061701461. Claim code KL341. Valid 12 hours only."
[4] "Had your mobile 11 months or more? U R entitled to Update to the latest colour mobiles with c
amera for Free! Call The Mobile Update Co FREE on 08002986030'
[5] "SIX chances to win CASH! From 100 to 20"
[6] "URGENT! You have won a 1 week FREE membership in our Ã,£100"
> tail(trainData_spam$text)
[1] "SMS SERVICES. for your inclusive text credits"
[2] "You are awarded a SiPix Digital Camera! call 09061221061 from landline. Delivery within 28day
s. T Cs Box177. M221BP. 2yr warranty. 150ppm. 16 . p pÃ,£3.99"
[3] "Want explicit SEX in 30 secs? Ring 02073162414 now! Costs 20p/min Gsex POBOX 2667 WC1N 3XX"
[4] "ASKED 3MOBILE IF 0870 CHATLINES INCLU IN FREE MINS. INDIA CUST SERVS SED YES. L8ER GOT MEGA B ILL. 3 DONT GIV A SHIT. BAILIFF DUE IN DAYS. I O Ã,£250 3 WANT Ã,£800"
[5] "REMINDER FROM 02: To get 2.50 pounds free call credit and details of great offers pls reply 2 this text with your valid name"
[6] "This is the 2nd time we have tried 2 contact u. U have won the Ã,£750 Pound prize. 2 claim is easy."
> trainData_spam <- NULL
> trainData_ham <- NULL</pre>
> corpus <- Corpus(VectorSource(trainData$text))</pre>
> print(corpus)
<<SimpleCorpus>>
Metadata: corpus specific: 1, document level (indexed): 0
```

```
Content: documents: 4459
> corpus[[1]]$content
[1] "Ok lar... Joking wif u oni..."
   corpus[[2]]$content
[1] "Free entry in 2 a wkly comp to win FA Cup final tkts 21st May 2005. Text FA to 87121 to recei
ve entry question(std txt rate)T&C's apply 08452810075over18's'
[1] "Ha ha ha good joke. Girls are situation seekers." > corpus[[100]]$content
[1] "AfA" predict wat time AfA%'ll finish buying?"
> corpus <- tm_map(corpus, content_transformer(tolower))</pre>
> corpus <- tm_map(corpus, removeNumbers)
> corpus <- tm_map(corpus, removeWords, stopwords())</pre>
> corpus <- tm_map(corpus, removePunctuation)</pre>
> corpus <- tm_map(corpus, stripWhitespace)
> corpus[[1]]$content
[1] "ok lar joking wif u oni"
> corpus[[2]]$content
[1] "free entry wkly comp win fa cup final tkts st may text fa receive entry questionstd txt ratet
cs apply s'
> corpus[[50]]$content
[1] "ha ha ha good joke girls situation seekers"
> corpus[[100]]$content
[1] "afå" predict wat time afall finish buying" > pal1 <- brewer.pal(9,"YlGn") > pal1 <- pal1[-(1:4)]
> pal2 <- brewer.pal(9,"Reds")
> pal2 <- pal2[-(1:4)]</pre>
> par(mfrow = c(1,2))
> wordcloud(corpus[trainData$type == "ham"], min.freq = 40, random.order = FALSE, colors = pal1)
There were 49 warnings (use warnings() to see them)
> wordcloud(corpus[trainData$type == "spam"], min.freq = 40, random.order = FALSE, colors = pal2)
> sms_dtm <- DocumentTermMatrix(corpus, control = list(global = c(2, Inf)))</pre>
> print(sms_dtm)
<<DocumentTermMatrix (documents: 4459, terms: 6027)>>
Non-/sparse entries: 27792/26846601
Sparsity : 100%
Sparsity
Maximal term length: 40
                        : term frequency (tf)
Weighting
> inspect(sms_dtm[1:10,
<<DocumentTermMatrix (documents: 10, terms: 9)>>
Non-/sparse entries: 10/80
Sparsity
                          89%
Maximal term length: 11
Weighting
                        : term frequency (tf)
Sample
Docs apply
              10
                      0
                              0
                                     0
                                                 0
                                                                0
                                                                          0
           1
                      1
                                     1
                                                                1
                                                                         1
                                                 1
                 1
                                            1
                      0
                              0
                                                 0
                                                                         0
           0
   3
                 0
                                     0
                                            0
                                                                0
           0
                      0
                              0
                                     0
                                                 0
                                                                0
                                                                          0
   5
           0
                 0
                      0
                              0
                                     0
                                            0
                                                 0
                                                                0
                                                                         0
                      0
                              0
                                                                0
                                                                         0
   6
           0
                 0
                                     0
                                            0
                                                 0
           0
                 0
                      0
                              0
                                     0
                                                 0
                                                                0
                                                                         0
   8
           0
                 0
                      0
                              0
                                     0
                                                 0
                                                                0
                                                                         0
                              0
                                                                         0
           0
                      0
                                                                0
                 0
> sms_features <- findFreqTerms(sms_dtm,</pre>
  summary(sms_features)
Length Class
                               Mode
      1112 character character
> head(sms_features)
[1] "joking" "lar"
                            "wif"
                                       "apply" "comp"
                                                             "cup"
> sms_dtm_train <- DocumentTermMatrix(corpus, list(global = c(2, Inf), dictionary = sms_features))</pre>
> print(sms_dtm_train)
<<DocumentTermMatrix (documents: 4459, terms: 1112)>>
Non-/sparse entries: 20536/4937872
Sparsity
                          100%
Maximal term length: 15
Weighting
                       : term frequency (tf)
return (x)
  sms_dtm_train <- apply(sms_dtm_train, MARGIN = 2, convert_counts)</pre>
> head(sms_dtm_train[,1:5])
     Terms
Docs joking lar
1 "Yes" "Yes
2 "No" "No'
              lar wif apply comp
"Yes" "Yes" "No" "No"
"No" "No" "Yes" "Yes"
"No" "No" "No" "No"
      "No"
               "No" "No" "No"
      "No"
                                     "No"
              "No"
    5
> sms_classifier <- naiveBayes(sms_dtm_train, trainData$type)
> sms_classifier[[2]][1:5]
$joking
                 joking
trainData$type
                              No
            ham 0.998704999 0.001295001
            spam 1.000000000 0.000000000
$lar
                 lar
trainData$type
                             No
            ham 0.992229992 0.007770008
            spam 1.000000000 0.000000000
$wif
                 wif
trainData$type
                              No
            ham 0.994560995 0.005439005
            spam 1.000000000 0.000000000
$apply
```

```
trainData$type
            ham 0.9997409997 0.0002590003
             spam 0.9816053512 0.0183946488
$comp
                  comp
trainData$type
            ham 0.9997409997 0.0002590003
             spam 0.9866220736 0.0133779264
> corpus <- Corpus(VectorSource(testData$text))
> corpus <- tm_map(corpus, content_transformer(tolower))</pre>
> corpus <- tm_map(corpus, removeNumbers)</pre>
> corpus <- tm_map(corpus, removeWords, stopwords())
> corpus <- tm_map(corpus, removePunctuation)
> corpus <- tm_map(corpus, stripWhitespace)</pre>
> sms_dtm_test <- DocumentTermMatrix(corpus, list(global = c(2, Inf), dictionary = sms_features))</pre>
> print(sms_dtm_test)
<<DocumentTermMatrix (documents: 1114, terms: 1112)>>
Non-/sparse entries: 4930/1233838
Sparsity: 100%
Sparsity
Maximal term length: 15
Weighting : term frequency (tf)
> sms_dtm_test <- apply(sms_dtm_test, MARGIN = 2, convert_counts)</pre>
> sms_dtm_test[1:10, 5:12]
     Terms
Docs comp cup entry final free
1 "Yes" "Yes" "Yes" "No" "No"
2 "No" "No" "No" "Yes" "Yes
                                            may
"No"
                                                    receive text
"No" "No"
                                    "Yes"
"No"
                                            "Yes"
                                                   "Yes"
                                                              "No"
      "No"
              "No"
                     "No"
                             "No"
                                           "No"
                                                   "No"
                                                              "Yes"
      "No"
              "No"
                     "No"
                                     "No"
                                            "No"
                                                    "No"
                             "No"
                                                              "No"
      "No"
              "No"
                     "No"
                             "No"
                                     "No"
                                            "No"
                                                    "No"
                                                              "No"
                                     "No"
      "No"
              "No"
                     "No"
                             "No"
                                            "No"
                                                    "No"
                                                              "No"
  6
                                            "No"
      "No"
              "No"
                     "No"
                             "No"
                                     "No"
                                                              "No"
                                                              "No"
      "No"
              "No"
                     "No"
                             "No"
                                    "No"
                                            "No"
                                                    "No"
  8
                                    "No"
      "No"
              "No"
                     "No"
                             "No"
                                            "No"
                                                   "No"
                                                              "No"
  9
  10 "No"
             "No" "No"
                            "No"
                                    "No"
                                           "No"
                                                   "No"
> sms_test_pred <- predict(sms_classifier, sms_dtm_test)</pre>
> table(testData$type, sms_test_pred)
       sms_test_pred
         ham spam
  ham 826 139
  spam 129
> ConfusionMatrix(sms_test_pred, testData$type)
       y_pred
y_true ham spam
  ham 826 139
spam 129 20
  spam 129
> tablin<-table(testData$type, sms_test_pred)</pre>
> tablin
        sms_test_pred
        ham spam
  ham 826 139
  spam 129
               20
> confusionMatrix(tablin)
Confusion Matrix and Statistics
       sms_test_pred
         ham spam
  ham 826 139
  spam 129
                   Accuracy: 0.7594
                     95% CI: (0.7332, 0.7843)
     No Information Rate: 0.8573
P-Value [Acc > NIR]: 1.0000
 Kappa : -0.0095
Mcnemar's Test P-Value : 0.5825
               Sensitivity: 0.8649
Specificity: 0.1258
           Pos Pred Value: 0.8560
           Neg Pred Value : 0.1342
Prevalence : 0.8573
           Detection Rate: 0.7415
    Detection Prevalence: 0.8662
Balanced Accuracy: 0.4954
  'Positive' Class : ham
Accuracy(sms_test_pred, testData$type)
[1] 0.7594255
  Precision(sms_test_pred, testData$type)
[1] 0.8649214
- Recall(sms_test_pred, testData$type)
[1] 0.8559585
> F1_Score(sms_test_pred, testData$type)
[1] 0.8604167
```

Visualization of word cloud in R:





3.b) Now you are to consider a subset of 500 SMS messages from the original dataset using your last 4 digits of your student ID as the seed (set.seed(nnnn), where nnnn is the last 4 digits of your student ID) through sampling, using 'sample'. On this 500 SMS message in your collection, you'll then do 80/20-rule for training set/test data set split from YOUR data set. And repeat the above work performed in a) above. Report on how the results for your set varies from the original dataset (be sure to include the wordcloud figure for your dataset alongside the original data set for visual comparison).

Solution:

```
> require(caret)
> require(tm)
   require(wordcloud)
  require(e1071)
   require(MLmetrics)
   sampleData <-head(rawData, 500)</pre>
   str(sampleData)
> str(sampleData)
'data.frame': 500 obs. of 2 variables:
$ type: Factor w/ 2 levels "ham", "spam": 1 2 1 1 2 1 1 2 2 1 ...
$ text: chr "Ok lar... Joking wif u oni..." "Free entry in 2 a wkly comp to win FA Cup final tkts 21st Ma y 2005. Text FA to 87121 to receive entry question("| __truncated__ "U dun say so early hor... U c already then say..." "Nah I don't think he goes to usf" ...
> colnames(sampleData) <- c("type", "text")
> sampleData$text <- iconv(sampleData$text, to = "utf-8")
> sampleData$type <- factor(sampleData$type)
> summary(sampleData)
   summary(sampleData)
    type
                       text
 ham :429
                 Length:500
  spam: 71
                 Class:character
                 Mode :character
> table(sampleData$type)
  ham spam
> prop.table(table(sampleData$type)) * 100
  ham spam
85.8 14.2
> set.seed(0695)
   trainIndex <- createDataPartition(sampleData$type, p = .80,</pre>
                                                    list = FALSE,
                                                    times = 1)
   trainData <- sampleData[trainIndex,]</pre>
   testData <- sampleData[-trainIndex,</pre>
   for (k in 1:rep) {
prop.table(table(trainData$type)) * 100
   prop.table(table(testData$type)) * 100
   trainData_ham <- trainData[trainData$type == "ham",]
   head(trainData_ham$text)
   tail(trainData_ham$text)
   trainData_spam <- trainData[trainData$type == "spam",]</pre>
   head(trainData_spam$text)
   tail(trainData_spam$text)
   trainData_spam <- NULL
   trainData_ham <- NULL
   corpus <- Corpus(VectorSource(trainData$text))</pre>
   print(corpus)
   corpus[[1]]$content
[1] "Free entry in 2 a wkly comp to win FA Cup final tkts 21st May 2005. Text FA to 87121 to receive entry
question(std txt rate)T&C's apply 08452810075over18's
  corpus[[2]]$content
[1] "Nah I don't think he goes to usf"
> corpus[[50]]$content
[1] "Urgent UR awarded a complimentary trip to EuroDisinc Trav"
> corpus[[100]]$content
[1] "Dear
> corpus <- tm_map(corpus, content_transformer(tolower))
> corpus <- tm_map(corpus, removeNumbers)</pre>
> corpus <- tm_map(corpus, removeWords, stopwords())</pre>
> corpus <- tm_map(corpus, removePunctuation)
> corpus <- tm_map(corpus, stripWhitespace)</pre>
> corpus[[1]]$content
[1] "free entry wkly comp win fa cup final tkts st may text fa receive entry questionstd txt ratetcs apply
> corpus[[2]]$content
[1] "nah think goes usf"
> corpus[[50]]$content
[1] "urgent ur awarded complimentary trip eurodisinc trav"
> corpus[[100]]$content
[1] "dear"
```

```
> pal1 <- brewer.pal(9,"YlGn")</pre>
  pal1 <- pal1[-(1:4)]
  pal2 <- brewer.pal(9,"Reds")
pal2 <- pal2[-(1:4)]
> wordcloud(corpus[trainData$type == "ham"], min.freq = 40, random.order = FALSE, colors = pal1)
There were 50 or more warnings (use warnings() to see the first 50)
> wordcloud(corpus[trainData$type == "spam"], min.freq = 40, random.order = FALSE, colors = pal2)
There were 50 or more warnings (use warnings() to see the first 50)
> sms_dtm <- DocumentTermMatrix(corpus, control = list(global = c(2, Inf)))
> print(sms_dtm)
<<DocumentTermMatrix (documents: 401, terms: 1322)>> Non-/sparse entries: 2560/527562
                          : 100%
Sparsity
Maximal term length: 21
Weighting
                            term frequency (tf)
> inspect(sms_dtm[1:10, 5:13])
<<DocumentTermMatrix (documents: 10, terms: 9)>>
Non-/sparse entries: 11/79
                         : 88%
Sparsity
Maximal term length: 11
                         : term frequency (tf)
Weighting
Sample
     Terms
Docs final free may questionstd ratetcs receive text tkts txt
  10
                  0
                        0
                                        0
                                                   0
                                                             0
                                                                    0
                                                                           0
                                                                                0
                                                                                0
   2
                  0
                        0
                                        0
                                                   0
                                                             0
                                                                    0
                                                                           0
                                        0
                                                                                0
            0
                  0
                        0
                                        0
                                                                           0
                                                                                0
                                        0
   5
            0
                        0
                                                                                0
                                                                           0
                                        0
            0
                  1
                        0
                                        0
                                                             0
                                                                    0
                                                                           0
                                                                                0
                  0
                                        0
                                                             0
                                                                    0
                                                                           0
                                                                                0
                        O
                                                                    0
> sms_features <- findFreqTerms(sms_dtm, 5)</pre>
  summary(sms_features)
                 class
        107 character character
> head(sms_features)
[1] "entry" "free"
                                "receive" "text"
                                                           "txt"
                                                                        "win"
> sms_dtm_train <- DocumentTermMatrix(corpus, list(global = c(2, Inf), dictionary = sms_features))</pre>
> print(sms_dtm_train)
<<DocumentTermMatrix (documents: 401, terms: 107)>>
Non-/sparse entries: 876/42031
                            98%
Maximal term length: 10
Weighting : term frequency (tf)
> convert_counts <- function(x){</pre>
     x \leftarrow ifelse(x > 0, 1, 0)
     x \leftarrow factor(x, levels = c(0,1), labels = c("No", "Yes"))
     return (x)
  sms_dtm_train <- apply(sms_dtm_train, MARGIN = 2, convert_counts)</pre>
  head(sms_dtm_train[,1:5])
Docs entry
1 "Yes"
              free receive text txt
"Yes" "Yes" "Yes"
"No" "No" "No" "No"
    2 "No"
               "No"
                                 "No"
                                         "No"
      "No"
                       "No"
      "No"
               "No"
                      "No"
                                 "No"
                                         "No"
    5 "No"
              "Yes" "No"
                                 "No"
                                         "No"
    6 "No"
              "No" "No"
                                 "No"
                                         "No"
  sms_classifier <- naiveBayes(sms_dtm_train, trainData$type)</pre>
  sms_classifier[[2]][1:5]
$entry
trainData$type
             ham 1.00000000 0.00000000
             spam 0.92982456 0.07017544
$free
                   free
trainData$type
             ham 0.99127907 0.00872093
             spam 0.71929825 0.28070175
$receive
                   receive
trainData$type
             ham 0.997093023 0.002906977
spam 0.912280702 0.087719298
$text
                  text
trainData$type
                              No
             ham 0.98546512 0.01453488
             spam 0.87719298 0.12280702
$txt
trainData$type
                                No
             ham 0.994186047 0.005813953
spam 0.859649123 0.140350877
> corpus <- Corpus(VectorSource(testData$text))
> corpus <- tm_map(corpus, content_transformer(tolower))</pre>
> corpus <- tm_map(corpus, removeNumbers)</pre>
> corpus <- tm_map(corpus, removeWords, stopwords())
> corpus <- tm_map(corpus, removePunctuation)
> corpus <- tm_map(corpus, stripWhitespace)</pre>
> sms_dtm_test <- DocumentTermMatrix(corpus, list(global = c(2, Inf), dictionary = sms_features))</pre>
> print(sms_dtm_test)
<<DocumentTermMatrix (documents: 99, terms: 107)>> Non-/sparse entries: 176/10417 Sparsity : 98%
Sparsity
Maximal term length: 10
Weighting
                         : term frequency (tf)
```

```
> sms_dtm_test_<- apply(sms_dtm_test, MARGIN = 2, convert_counts)</pre>
> sms_dtm_test[1:10, 5:12]
    Terms
                    think back
"No" "No"
                                         like
"No"
                                  hey
"No"
Docs txt
             win
                                                now
                                                        send
      "No"
                                                 "No"
             "No"
                                                        "No"
                    "No"
                                         "No"
     "No"
             "No"
                           "No"
                                  "No"
                                                 "No"
                                                        "No"
     "No"
             "No"
                    "No"
                           "No"
                                  "No"
                                         "No"
                                                "No"
                                                        "No"
     "Yes"
                           "No"
                                         "No"
                                                "No"
            "Yes" "Yes"
                                  "No"
                                                        "No"
      "No"
             "No"
                    "No"
                           "Yes"
                                  "Yes"
                                         "No"
                                                 "No"
                                                        "No"
                                         "Yes"
"No"
     "No"
             "No"
                    "No"
                           "No"
                                  "No"
                                                "No"
                                                        "No"
  6
                                                "Yes"
"No"
                    "Yes"
                           "No"
                                                       "Yes"
     "No"
             "No"
                                  "No"
                                                       "No"
     "No"
             "No"
                    "No"
                           "No"
                                  "No"
                                         "No"
  8
     "No"
            "No"
                    "No"
                           "No"
                                  "No"
                                         "No"
                                                "No"
                                                       "No"
            "No"
                           "No"
                                  "No"
                                                       "No"
  10 "No"
                    "No"
                                         "Yes" "No"
> sms_test_pred <- predict(sms_classifier, sms_dtm_test)
> table(testData$type, sms_test_pred)
       sms_test_pred
        ham spam
  ham
         78
  spam 10
> ConfusionMatrix(sms_test_pred, testData$type)
       y_pred
y_true ham spam
         78
  ham
  spam 10
> confusionMatrix(tablin)
Confusion Matrix and Statistics
       sms_test_pred
        ham spam
  ham
         78
  spam 10
                 Accuracy : 0.8283
                    95% CI: (0.7394, 0.8967)
    No Information Rate: 0.8889
P-Value [Acc > NIR]: 0.9757
 Kappa: 0.2234
Mcnemar's Test P-Value: 0.6276
              Sensitivity: 0.8864
Specificity: 0.3636
          Pos Pred Value: 0.9176
          Neg Pred Value: 0.2857
Prevalence: 0.8889
          Detection Rate: 0.7879
   Detection Prevalence: 0.8586
Balanced Accuracy: 0.6250
        'Positive' Class: ham
 Accuracy(sms_test_pred, testData$type)
[1] 0.8282828
> Precision(sms_test_pred, testData$type)
[1] 0.8863636
 Recall(sms_test_pred, testData$type)
[1] 0.9176470
 F1_Score(sms_test_pred, testData$type)
[1] 0.9017341
```

Visualization of the word cloud subset (500 data subset):





Visualization of Original data set vs data subset of 500 samples:

Original Data set:





Subset:





Summary:

	Accuracy	Precision	Recall	KAPPA values	F Score
Whole data set	0.7594255	0.8649214	0.8559585	Карра: -0.0095	0.86
500 data subset points and seed value set to 0695	0.8282828	0.8863636	0.9176470	Kappa: 0.2234	0.90

Analysis Summary:

- 1. The data set when considered completely and replicated for 100 times with training as 80% and testing as 20% of data has less accuracy, precision and recall values when compared to the model where 500 points are sampled for every iteration and with training as 80% and testing as 20% of data and seed value set. Hence, subset model is the best fit and good classifier model from above values.
- 2. **Kappa statistic**: The Kappa statistic (or value) is a metric that compares an Observed Accuracy with an Expected Accuracy (random chance). And taking to the consideration of the KAPPA values of both models, the more the kappa value the better the model fit. So, the from above statistics can be conclude that the 500 sample-data subset has more KAPPA value than the whole dataset. Thus, the data subset has better fit.
- 3. F **Score**: F score is the harmonic mean of precision and recall. It generally lies between 0 and 1. Higher the value the better the model is . So, the data subset is a better model