

Classification Model

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loading required packages

```
library(caret)
library(dplyr)
library(corrplot)
library(glmnet)
library(tidyverse)
```

Reading the data file

```
data<-read.csv("train_v3.csv")
```

Clearing the Null values and checking for percentage of null values in each column

#Replacing all NA values with 0

```
data1 <- data %>% mutate_all(funs(replace_na(.,0)))
```

Calculate the percentage of 0 values in each column

```
null_percent <- apply(data1 == 0, 2, mean)
```

Get the names of the columns with less than or equal to 30% of 0(null) values

```
cols <- names(null_percent[null_percent <= 0.3])
```

Create a new data frame with the selected columns

```
data2 <- data1[, cols]
```

#Check if the columns with more than 30% null values are deleted

```
sums<-(colSums(data2==0)/nrow(data2))*100
```

```

#Adding loss column back to data

data3<-cbind(data2,loss=data1$loss)

# Create a new column called 'default' with a value of 1 is loss is above 0 and 0 is loss is 0

data3$default <- ifelse(data3$loss == 0, 0, 1)

```

Reading cleaned data file

```
data<-data3
```

Clearing columns with near zero variance,high correlation and median imputation for missing values for smooth running in the model

```

data<-data[ ,-(data$f736,data$f764)]

preProcessModel <- preProcess(data, method = c("nzv", "corr", "medianImpute"))

data1 <- predict(preProcessModel, data)

```

After data cleaning we are left with 261 columns

Building a lasso model

```

#Creating a matrix

X <- as.matrix(data1[ ,-(261)])

#Setting default column to factor levels

Y <- as.vector(as.factor(data1$default))

#Building the model with 10 fold cross-validation and performance measure as AUC curve

model<- cv.glmnet(X, Y, alpha = 1, family = "binomial", nfolds = 10, type.measure = "auc")

summary(model)

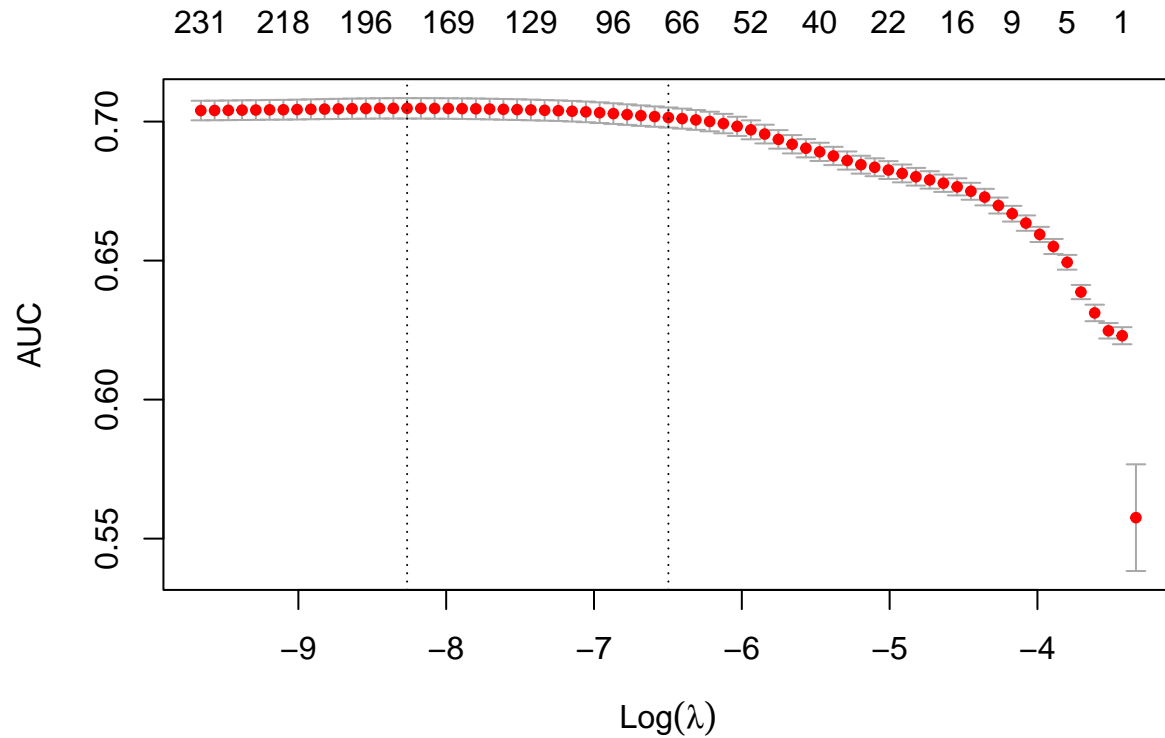
```

```

##           Length Class  Mode
## lambda      69      -none- numeric
## cvm          69      -none- numeric
## cvsd         69      -none- numeric
## cvup         69      -none- numeric
## cvlo         69      -none- numeric
## nzero        69      -none- numeric
## call         7       -none- call
## name         1       -none- character
## glmnet.fit   13      lognet list
## lambda.min   1       -none- numeric
## lambda.1se   1       -none- numeric
## index        2       -none- numeric

```

```
plot(model)
```



```
#Finding minimum value for lambda
```

```
model$lambda.min
```

```
## [1] 0.0002575271
```

```
#Saving the coefficients for lambda minimum
```

```
coefs <- coef(model, s = "lambda.min")
```

```
# Taking the coefficient values into a data frame for processing
```

```
coefs <- data.frame(name = coefs@Dimnames[[1]][coefs@i + 1], coefficient = coefs@x)
```

```
# Rounding to the absolute value of all the coefficients in the model
```

```
coefs$coefficient <- abs(coefs$coefficient)
```

```
# Finding the largest coefficient in the model
```

```
coefs <- coefs[order(coefs$coefficient, decreasing = TRUE), ]
```

```

# Taking out the intercept from the data frame

coefs <- coefs[-1, ]

#Converting the coefficients to a vector type

coefs <- as.vector(coefs$name)

#Combining the default column back to the coefficients

coefs <- c(coefs,"default")

data_new<-select(data1,coefs)

```

Splitting data into 85% training and 15% validation

```

#Splitting data into Training and Validation

set.seed(6782)

Split_data <- createDataPartition(data_new$default,p=.85,list=FALSE,times=1)

Training <- data_new[Split_data,]

Validation <- data_new[-Split_data,]

```

Building a Gradient Boosting Machine model

```

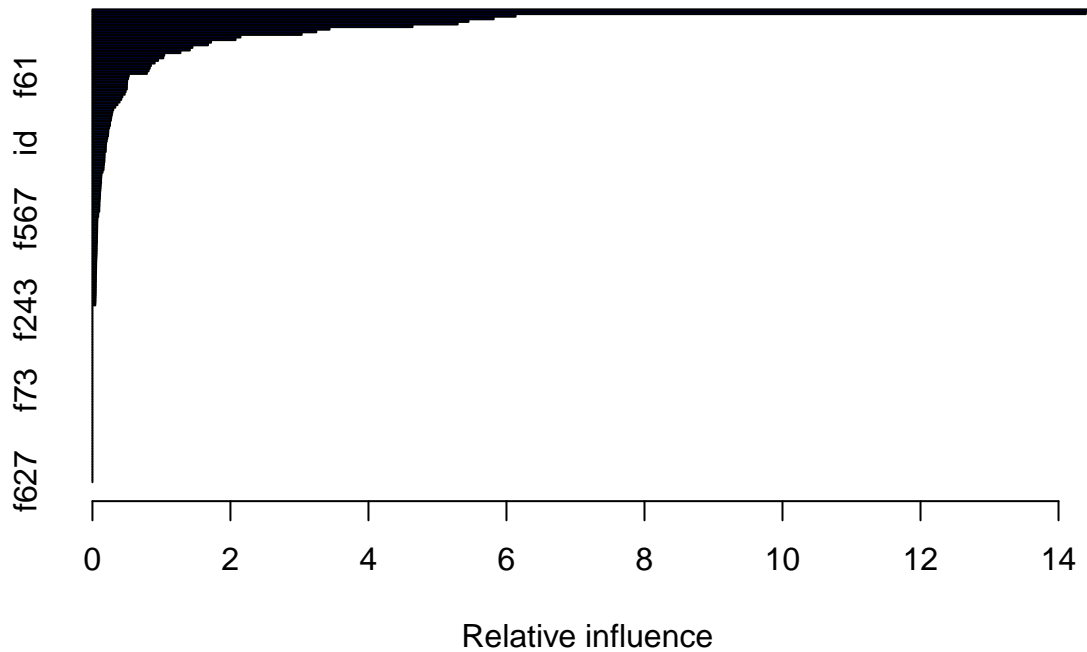
library(gbm)

#Cross-validation with 10 folds and number of trees are set to 50

gbm_model <- gbm(default ~ ., data = Training, distribution = "bernoulli",
  n.trees = 100, interaction.depth = 9, shrinkage = 0.02,
  bag.fraction = 0.8, train.fraction = 0.85,
  n.minobsinnode = 15,cv.folds = 10)

summary(gbm_model)

```



##	var	rel.inf
##	f322 f322	14.49036550
##	f67 f67	14.39655693
##	f674 f674	6.13149462
##	f64 f64	5.81482166
##	f765 f765	5.44950283
##	f471 f471	5.29261823
##	f630 f630	4.63841978
##	f514 f514	3.43592335
##	f13 f13	3.24401161
##	f468 f468	3.03193368
##	f251 f251	2.14494023
##	f221 f221	2.07397569
##	f434 f434	1.71871484
##	f652 f652	1.67426590
##	f260 f260	1.44928864
##	f268 f268	1.41078450
##	f32 f32	1.27521745
##	f461 f461	1.04039864
##	f411 f411	1.02690701
##	f333 f333	0.95094625
##	f229 f229	0.90164913
##	f261 f261	0.84726504
##	f763 f763	0.83084047
##	f272 f272	0.81366463
##	f677 f677	0.78891803

```

## f61    f61  0.52681463
## f213 f213  0.51716062
## f278 f278  0.50325031
## f536 f536  0.50126347
## f19    f19  0.49877305
## f202 f202  0.49856051
## f420 f420  0.48107377
## f672 f672  0.46772978
## f768 f768  0.43415335
## f146 f146  0.41775577
## f470 f470  0.39293360
## f269 f269  0.36242621
## f421 f421  0.33082845
## f270 f270  0.30184962
## f233 f233  0.28888770
## f144 f144  0.28168502
## f183 f183  0.27265754
## f277 f277  0.26779860
## f54    f54  0.25499490
## f669 f669  0.25422297
## f673 f673  0.24206397
## f94    f94  0.23024905
## f699 f699  0.22849870
## f739 f739  0.22605869
## f238 f238  0.21479759
## f662 f662  0.20769932
## id     id  0.19823772
## f341 f341  0.19511532
## f131 f131  0.19304432
## f524 f524  0.19268205
## f143 f143  0.17733356
## f374 f374  0.17602127
## f628 f628  0.17373100
## f742 f742  0.17295467
## f618 f618  0.16745975
## f198 f198  0.16287088
## f636 f636  0.16065409
## f139 f139  0.14680565
## f150 f150  0.13092641
## f734 f734  0.12842166
## f479 f479  0.12517945
## f99    f99  0.12514775
## f347 f347  0.12115458
## f16    f16  0.11838786
## f638 f638  0.11511888
## f620 f620  0.11085250
## f774 f774  0.11006228
## f90    f90  0.10848512
## f217 f217  0.10490524
## f330 f330  0.10188927
## f590 f590  0.10161796
## f170 f170  0.09917514
## f647 f647  0.09897650
## f637 f637  0.08474235

```

f450 f450 0.08344579
f567 f567 0.07299141
f631 f631 0.07080612
f646 f646 0.06945600
f587 f587 0.06919687
f756 f756 0.06851326
f661 f661 0.06818486
f129 f129 0.06681447
f199 f199 0.06623383
f589 f589 0.06548066
f349 f349 0.06538572
f133 f133 0.06353138
f84 f84 0.06187323
f653 f653 0.06169639
f740 f740 0.06085932
f664 f664 0.05967448
f775 f775 0.05770165
f110 f110 0.05569849
f161 f161 0.05401770
f533 f533 0.05382205
f104 f104 0.05281992
f102 f102 0.05241601
f649 f649 0.05137703
f651 f651 0.05067050
f76 f76 0.05027029
f412 f412 0.05014326
f760 f760 0.04916410
f100 f100 0.04900660
f203 f203 0.04849675
f659 f659 0.04803993
f522 f522 0.04703855
f80 f80 0.04598116
f383 f383 0.04440169
f598 f598 0.04398964
f609 f609 0.04123584
f57 f57 0.00000000
f243 f243 0.00000000
f249 f249 0.00000000
f248 f248 0.00000000
f130 f130 0.00000000
f188 f188 0.00000000
f70 f70 0.00000000
f316 f316 0.00000000
f65 f65 0.00000000
f71 f71 0.00000000
f82 f82 0.00000000
f385 f385 0.00000000
f81 f81 0.00000000
f151 f151 0.00000000
f153 f153 0.00000000
f158 f158 0.00000000
f645 f645 0.00000000
f340 f340 0.00000000
f321 f321 0.00000000

```
## f384 f384 0.00000000
## f458 f458 0.00000000
## f220 f220 0.00000000
## f218 f218 0.00000000
## f173 f173 0.00000000
## f3      f3 0.00000000
## f289 f289 0.00000000
## f755 f755 0.00000000
## f588 f588 0.00000000
## f331 f331 0.00000000
## f279 f279 0.00000000
## f444 f444 0.00000000
## f601 f601 0.00000000
## f73     f73 0.00000000
## f320 f320 0.00000000
## f350 f350 0.00000000
## f160 f160 0.00000000
## f518 f518 0.00000000
## f425 f425 0.00000000
## f442 f442 0.00000000
## f367 f367 0.00000000
## f725 f725 0.00000000
## f436 f436 0.00000000
## f44     f44 0.00000000
## f733 f733 0.00000000
## f422 f422 0.00000000
## f612 f612 0.00000000
## f288 f288 0.00000000
## f614 f614 0.00000000
## f189 f189 0.00000000
## f358 f358 0.00000000
## f212 f212 0.00000000
## f5      f5 0.00000000
## f403 f403 0.00000000
## f650 f650 0.00000000
## f648 f648 0.00000000
## f378 f378 0.00000000
## f654 f654 0.00000000
## f680 f680 0.00000000
## f285 f285 0.00000000
## f428 f428 0.00000000
## f4      f4 0.00000000
## f679 f679 0.00000000
## f743 f743 0.00000000
## f735 f735 0.00000000
## f526 f526 0.00000000
## f715 f715 0.00000000
## f287 f287 0.00000000
## f627 f627 0.00000000
```

```
predictions <- predict(gbm_model, newdata = Validation, n.trees = 100, type = "response")
#Threshold value is set to 50 %
```



```

predictions<-ifelse(predictions>0.5,1,0)

# Building Confusion Matrix

Final_data<-cbind(Validation,predictions)

Final_data$predictions<-as.factor(Final_data$predictions)

Final_data$default<-as.factor(Final_data$default)

confusionMatrix<-confusionMatrix(Final_data$default,Final_data$predictions)
confusionMatrix

```

```

## Confusion Matrix and Statistics
##
##           Reference
## Prediction      0      1
##           0 10920      0
##           1  1079      1
##
##           Accuracy : 0.9101
##           95% CI : (0.9048, 0.9151)
##       No Information Rate : 0.9999
##       P-Value [Acc > NIR] : 1
##
##           Kappa : 0.0017
##
##  Mcnemar's Test P-Value : <2e-16
##
##           Sensitivity : 0.9100758
##           Specificity : 1.0000000
##           Pos Pred Value : 1.0000000
##           Neg Pred Value : 0.0009259
##           Prevalence : 0.9999167
##           Detection Rate : 0.9100000
##       Detection Prevalence : 0.9100000
##           Balanced Accuracy : 0.9550379
##
##           'Positive' Class : 0
##

```

```

#Precision<- TP/TP+FP = 0.005%

```

Here Precision was found to be very low .This was expected because from our EDA we have seen that the data has imbalanced classes of default to non-default with almost 1:10 ratio. Therefore we will use stratified sampling for splitting and run the model again

```

data_sample<-data_new

# Calculate the smallest class size

min_length <- min(table(data_sample$default))

```

```
# Sample the minority class to the size of the smallest class

balanced_spam <- data_sample %>%group_by(default) %>%sample_n(min_length) %>%ungroup()

# Check the class distribution after balancing

table(balanced_spam$default)
```

```
##
##      0      1
## 7379 7379
```

Splitting data into 15% test and 85% train within our balanced data

```
set.seed(6782)

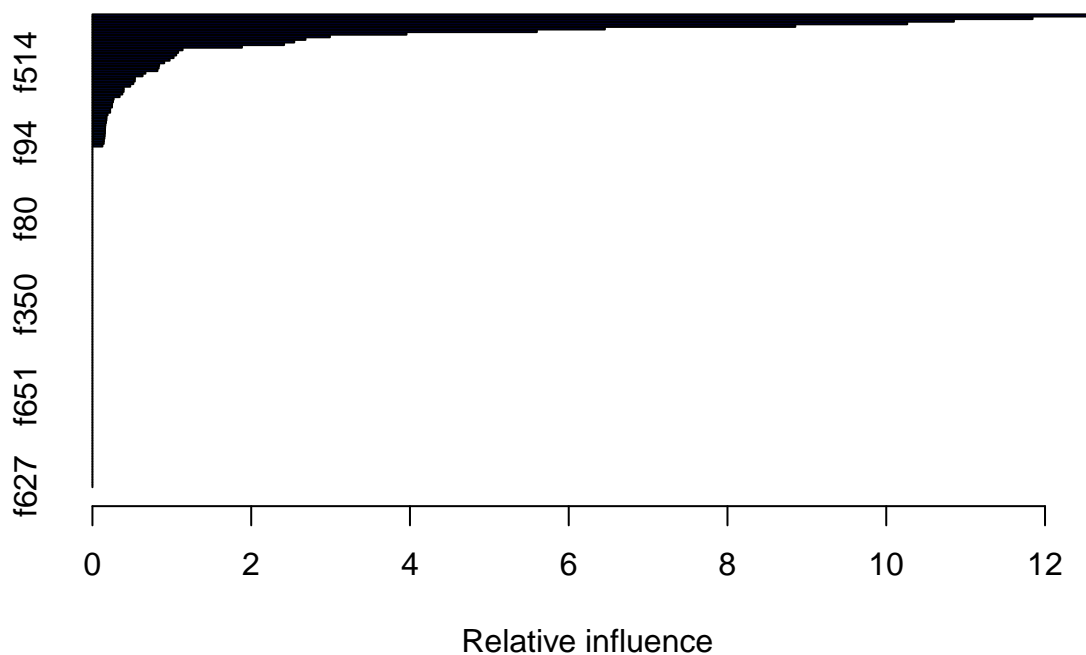
Split_data_sample <- createDataPartition(balanced_spam$default,p=.85,list=FALSE,times=1)

Training_sample <- balanced_spam[Split_data_sample,]

Validation_sample <- balanced_spam[-Split_data_sample,]
```

Building the model again using the new representative dataset

```
library(gbm)
gbm_model <- gbm(default ~ ., data = Training_sample, distribution = "bernoulli",
  n.trees = 50, interaction.depth = 5, shrinkage = 0.02,
  bag.fraction = 0.80, train.fraction = 0.85,
  n.minobsinnode = 25,cv.folds = 10)
summary(gbm_model)
```



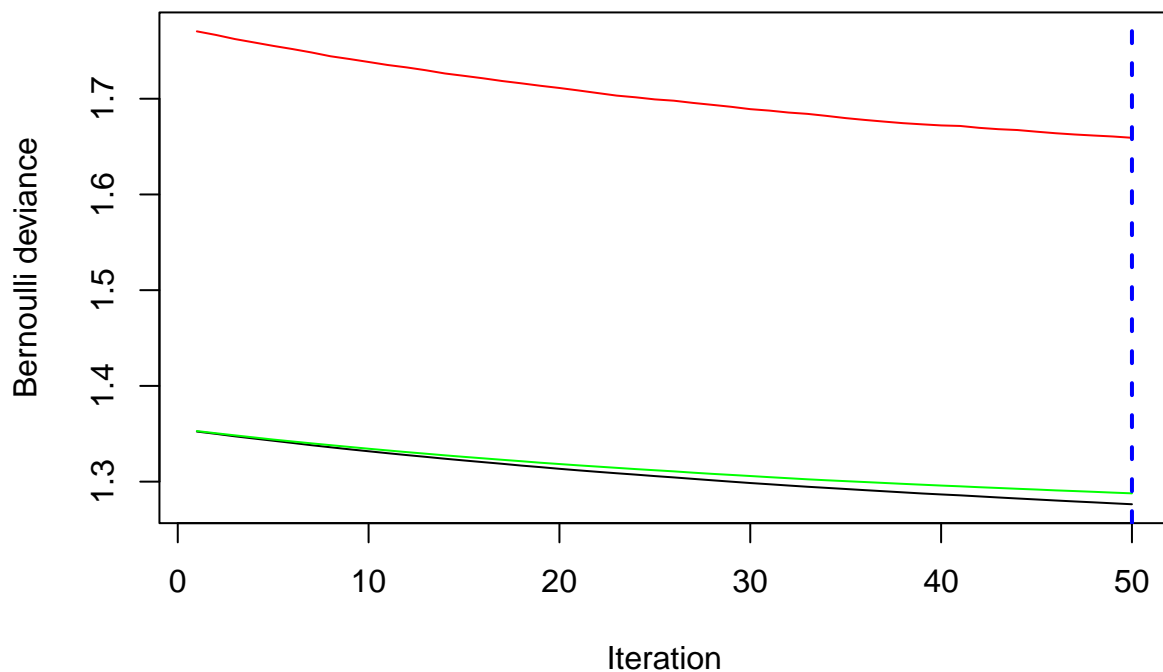
##	var	rel.inf
##	f630 f630	12.5964053
##	f67 f67	11.8433098
##	f322 f322	10.8521082
##	f64 f64	10.2639996
##	f674 f674	8.8543818
##	f765 f765	6.4551375
##	f32 f32	5.5985196
##	f13 f13	3.9561965
##	f229 f229	2.9913778
##	f19 f19	2.6839004
##	f768 f768	2.5424172
##	f378 f378	2.4140177
##	f260 f260	1.8831030
##	f763 f763	1.1353925
##	f251 f251	1.0793114
##	f268 f268	1.0551172
##	f84 f84	1.0223973
##	f269 f269	0.9721860
##	f514 f514	0.9039666
##	f213 f213	0.8399960
##	f183 f183	0.8340425
##	f270 f270	0.8185144
##	f278 f278	0.6673766
##	f188 f188	0.6316940
##	f221 f221	0.5350747

f238 f238 0.5339548
f65 f65 0.5150226
f99 f99 0.4761546
f442 f442 0.3940122
f233 f233 0.3931776
f198 f198 0.3766796
f461 f461 0.3424858
f411 f411 0.2669712
f129 f129 0.2587350
f316 f316 0.2474813
f100 f100 0.2463719
f203 f203 0.2231855
f277 f277 0.2229400
f434 f434 0.1908153
f652 f652 0.1778407
f479 f479 0.1761464
f272 f272 0.1707954
f333 f333 0.1633231
f734 f734 0.1594331
f420 f420 0.1583693
f662 f662 0.1574581
f468 f468 0.1547471
f699 f699 0.1509614
f589 f589 0.1443054
f94 f94 0.1425667
f261 f261 0.1261224
f131 f131 0.0000000
f471 f471 0.0000000
f57 f57 0.0000000
f243 f243 0.0000000
f249 f249 0.0000000
f760 f760 0.0000000
f61 f61 0.0000000
f248 f248 0.0000000
f536 f536 0.0000000
f774 f774 0.0000000
f130 f130 0.0000000
f756 f756 0.0000000
f70 f70 0.0000000
f71 f71 0.0000000
f82 f82 0.0000000
f590 f590 0.0000000
f637 f637 0.0000000
f598 f598 0.0000000
f102 f102 0.0000000
f90 f90 0.0000000
f620 f620 0.0000000
f385 f385 0.0000000
f81 f81 0.0000000
f110 f110 0.0000000
f151 f151 0.0000000
f153 f153 0.0000000
f158 f158 0.0000000
f80 f80 0.0000000

```
## f631 f631 0.0000000
## f645 f645 0.0000000
## f340 f340 0.0000000
## f321 f321 0.0000000
## f384 f384 0.0000000
## f458 f458 0.0000000
## f450 f450 0.0000000
## f220 f220 0.0000000
## f218 f218 0.0000000
## f173 f173 0.0000000
## f533 f533 0.0000000
## f3      f3 0.0000000
## f330 f330 0.0000000
## f289 f289 0.0000000
## f638 f638 0.0000000
## f170 f170 0.0000000
## f755 f755 0.0000000
## f588 f588 0.0000000
## f331 f331 0.0000000
## f279 f279 0.0000000
## f374 f374 0.0000000
## f150 f150 0.0000000
## f444 f444 0.0000000
## f601 f601 0.0000000
## f341 f341 0.0000000
## f73    f73 0.0000000
## f522 f522 0.0000000
## f775 f775 0.0000000
## f524 f524 0.0000000
## f677 f677 0.0000000
## f609 f609 0.0000000
## f320 f320 0.0000000
## f350 f350 0.0000000
## f160 f160 0.0000000
## f518 f518 0.0000000
## f104 f104 0.0000000
## f425 f425 0.0000000
## f647 f647 0.0000000
## f367 f367 0.0000000
## f143 f143 0.0000000
## f725 f725 0.0000000
## f436 f436 0.0000000
## f44    f44 0.0000000
## f672 f672 0.0000000
## f733 f733 0.0000000
## f422 f422 0.0000000
## f54    f54 0.0000000
## f739 f739 0.0000000
## f76    f76 0.0000000
## f612 f612 0.0000000
## f288 f288 0.0000000
## f646 f646 0.0000000
## f614 f614 0.0000000
## f383 f383 0.0000000
```

```
## f349 f349 0.0000000
## f189 f189 0.0000000
## f161 f161 0.0000000
## f358 f358 0.0000000
## f412 f412 0.0000000
## f673 f673 0.0000000
## f212 f212 0.0000000
## f5      f5 0.0000000
## f403 f403 0.0000000
## f669 f669 0.0000000
## f650 f650 0.0000000
## f648 f648 0.0000000
## f618 f618 0.0000000
## f651 f651 0.0000000
## f144 f144 0.0000000
## f199 f199 0.0000000
## f654 f654 0.0000000
## f653 f653 0.0000000
## f661 f661 0.0000000
## f664 f664 0.0000000
## f680 f680 0.0000000
## f649 f649 0.0000000
## f740 f740 0.0000000
## f285 f285 0.0000000
## f428 f428 0.0000000
## f742 f742 0.0000000
## f4      f4 0.0000000
## f679 f679 0.0000000
## f587 f587 0.0000000
## f139 f139 0.0000000
## id     id 0.0000000
## f743 f743 0.0000000
## f202 f202 0.0000000
## f636 f636 0.0000000
## f133 f133 0.0000000
## f146 f146 0.0000000
## f421 f421 0.0000000
## f470 f470 0.0000000
## f16   f16 0.0000000
## f735 f735 0.0000000
## f526 f526 0.0000000
## f217 f217 0.0000000
## f715 f715 0.0000000
## f287 f287 0.0000000
## f628 f628 0.0000000
## f347 f347 0.0000000
## f659 f659 0.0000000
## f567 f567 0.0000000
## f627 f627 0.0000000
```

```
cv_error<-gbm.perf(gbm_model,method="cv")
```



```
cv_error
```

```
## [1] 50
```

Prediction on Validation data

```
predictions_sample <- predict(gbm_model, newdata = Validation_sample, n.trees = 50, type = "response")
predictions_sample
```

```
##      [1] 0.3512700 0.4583707 0.2847233 0.3483735 0.3073960 0.5240308 0.3334525
##      [8] 0.3236956 0.5142390 0.3808043 0.2382272 0.3549159 0.2773079 0.4165826
##     [15] 0.4804397 0.3539431 0.4292882 0.4366508 0.4090311 0.2776910 0.3649510
##     [22] 0.4287243 0.3638741 0.3848456 0.2656934 0.3978761 0.3468707 0.4608265
##     [29] 0.4259265 0.4345325 0.5081809 0.5265465 0.4204526 0.4574839 0.5177636
##     [36] 0.3474000 0.3251357 0.4866832 0.3903400 0.3943205 0.3954220 0.3082572
##     [43] 0.5294977 0.3130981 0.3250128 0.3831457 0.4857314 0.3736070 0.4369706
##     [50] 0.4545811 0.4168306 0.4666406 0.3242163 0.4670334 0.3719814 0.4571148
##     [57] 0.3192908 0.4356544 0.3689215 0.4407491 0.4237964 0.2394223 0.4850796
##     [64] 0.3786924 0.5133822 0.2450227 0.5088532 0.4428922 0.3557342 0.3810464
##     [71] 0.3931159 0.2447817 0.4356498 0.3438653 0.4530715 0.5067687 0.4597439
##     [78] 0.2414107 0.3125177 0.4058313 0.2768254 0.3917835 0.4209054 0.4866623
##     [85] 0.4061073 0.4949465 0.4915450 0.4192337 0.3712030 0.4803570 0.3277115
##     [92] 0.4329843 0.2942027 0.2519695 0.4699136 0.4793094 0.3644111 0.3999837
##     [99] 0.4840759 0.3231180 0.4540910 0.3720394 0.4381432 0.5065283 0.3626155
##    [106] 0.4560494 0.3660617 0.3245538 0.3592933 0.4132843 0.3195020 0.3948955
```

```

## [113] 0.3271387 0.4742971 0.3495219 0.4261178 0.4395426 0.3385281 0.3933607
## [120] 0.2936933 0.3910811 0.4023028 0.4940776 0.4924153 0.2753641 0.3671475
## [127] 0.2382272 0.4638132 0.3628118 0.2969628 0.3220651 0.4142224 0.3894989
## [134] 0.4570212 0.5248322 0.3528418 0.3912169 0.4161747 0.3927500 0.3869069
## [141] 0.2400447 0.5055612 0.3948634 0.4043886 0.5069293 0.4551695 0.2420100
## [148] 0.5141201 0.3497600 0.4619601 0.4375734 0.4821414 0.3456718 0.4345012
## [155] 0.4790800 0.3116946 0.3479682 0.4155604 0.4718020 0.3605601 0.3925107
## [162] 0.4499931 0.4458981 0.4221770 0.4246055 0.4453083 0.2634520 0.4093659
## [169] 0.4009920 0.2598138 0.3261679 0.4113935 0.4708064 0.5126999 0.3572988
## [176] 0.4013459 0.4555195 0.4375413 0.3165300 0.4192005 0.4110214 0.3522558
## [183] 0.4058093 0.3827428 0.3037422 0.5200278 0.4540590 0.3004622 0.3717632
## [190] 0.2791624 0.3082728 0.4172809 0.2438578 0.4125276 0.4111089 0.5173644
## [197] 0.2753641 0.4548461 0.3878689 0.2639613 0.4582006 0.3710194 0.4803948
## [204] 0.4982989 0.4847316 0.4392195 0.4242028 0.2477462 0.3288168 0.3284359
## [211] 0.2571387 0.4773820 0.3363819 0.3325763 0.4324652 0.3309363 0.3584380
## [218] 0.3407672 0.3779328 0.3400191 0.4096745 0.2437166 0.3321559 0.4237482
## [225] 0.3999099 0.4704797 0.5200868 0.4667755 0.3723371 0.4267105 0.5162579
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## [379] 0.2464278 0.4528668 0.2600254 0.3919465 0.3418595 0.3348705 0.4641403
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## [477] 0.4481693 0.2579972 0.5212058 0.4248606 0.3984348 0.3295803 0.3142301
## [484] 0.3420842 0.4999264 0.5007025 0.3919819 0.4006212 0.4764213 0.2692907

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##	[491]	0.3771093	0.4512149	0.2981010	0.3810468	0.5046286	0.2455183	0.5231736
##	[498]	0.3802299	0.5050066	0.3563375	0.3856645	0.4551864	0.3906302	0.3392899
##	[505]	0.2993418	0.4506932	0.3592041	0.4141145	0.2399801	0.3055869	0.3173533
##	[512]	0.3948859	0.3101451	0.3335658	0.3929449	0.4001602	0.3803478	0.3346113
##	[519]	0.4492294	0.5027883	0.3007819	0.4367864	0.3531881	0.3049706	0.3184450
##	[526]	0.3975359	0.4673576	0.4516352	0.4261396	0.3787419	0.5119983	0.4210645
##	[533]	0.4450468	0.3516989	0.3470123	0.4109872	0.2399801	0.4960605	0.5265465
##	[540]	0.3923269	0.4108157	0.4157836	0.3370637	0.4005438	0.4912185	0.3262255
##	[547]	0.3902464	0.3278050	0.2802088	0.3770987	0.2753641	0.4643588	0.3356262
##	[554]	0.3203239	0.2582717	0.5236356	0.2778158	0.2420100	0.5129614	0.3465537
##	[561]	0.5000772	0.3435013	0.4582569	0.3960363	0.3848979	0.4716167	0.2975809
##	[568]	0.3874415	0.2981479	0.4759867	0.4587729	0.2766578	0.4527743	0.5229459
##	[575]	0.4066587	0.3844333	0.3564335	0.3614361	0.4166857	0.3589267	0.3167023
##	[582]	0.4590312	0.5039448	0.4153774	0.4250336	0.3644588	0.4027226	0.4110934
##	[589]	0.4917987	0.5258747	0.3074915	0.2549761	0.3843957	0.4992313	0.4015744
##	[596]	0.3715409	0.4169450	0.4656082	0.2838563	0.4525420	0.3580461	0.5027396
##	[603]	0.3785115	0.3489566	0.4259044	0.4351252	0.3944208	0.3860728	0.3059155
##	[610]	0.3087929	0.3236401	0.2896218	0.4728937	0.2405491	0.3700041	0.3327159
##	[617]	0.3637795	0.4085509	0.4518760	0.3764338	0.3584297	0.3634281	0.4809136
##	[624]	0.5033802	0.4221960	0.5051682	0.4174194	0.5294977	0.3456002	0.4432307
##	[631]	0.3639885	0.4446697	0.5132881	0.4735305	0.5240966	0.3268022	0.4983437
##	[638]	0.3672467	0.4453838	0.3463242	0.3936499	0.4162482	0.4128578	0.3836946
##	[645]	0.4371136	0.4561388	0.4170762	0.2399801	0.4986327	0.3474531	0.4106730
##	[652]	0.3861515	0.5036643	0.4259458	0.4001020	0.3533366	0.4471631	0.4689092
##	[659]	0.2977913	0.4997165	0.3413683	0.3888459	0.2619634	0.3213267	0.4082077
##	[666]	0.4073644	0.4446159	0.3812781	0.2991098	0.4705978	0.5044069	0.4687487
##	[673]	0.4680947	0.4161368	0.3341160	0.4297134	0.3336975	0.3479039	0.4865422
##	[680]	0.3758547	0.2457063	0.3735124	0.3622568	0.3257279	0.4572445	0.3334735
##	[687]	0.4551341	0.4026538	0.4166676	0.4061603	0.3302436	0.3535785	0.4043858
##	[694]	0.2576974	0.3684254	0.4936305	0.2956640	0.4522445	0.4794016	0.4703132
##	[701]	0.3747297	0.3455953	0.3487374	0.4958495	0.3914526	0.4974733	0.3901674
##	[708]	0.4391199	0.2667349	0.2419451	0.4592592	0.3793887	0.3178674	0.3107502
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##	[729]	0.2883270	0.2726204	0.3174260	0.3159777	0.3033960	0.4494999	0.3060827
##	[736]	0.4817139	0.4106430	0.3345684	0.3514405	0.3907653	0.3907653	0.4036781
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##	[904]	0.3083377	0.4543257	0.4398191	0.3546974	0.3261192	0.4798689	0.4073293
##	[911]	0.5167487	0.4602957	0.3196201	0.2941371	0.4455149	0.2435745	0.2868507
##	[918]	0.5077438	0.3702176	0.2665334	0.2591919	0.3884880	0.3757138	0.3629023
##	[925]	0.4946752	0.4503453	0.3700975	0.3681222	0.4664419	0.4156678	0.4390276
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##	[939]	0.3905040	0.4855181	0.3934391	0.4684133	0.4021051	0.3315627	0.3234118
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##	[953]	0.4013408	0.4565773	0.3045476	0.4004006	0.4328632	0.3751297	0.3649322
##	[960]	0.4242888	0.2849196	0.4074260	0.5017140	0.3582364	0.2691134	0.4594790
##	[967]	0.4343655	0.4722297	0.4055025	0.3425517	0.2685743	0.3147917	0.4079865
##	[974]	0.4755898	0.4880900	0.3937331	0.4859236	0.3506486	0.3678827	0.3202926
##	[981]	0.3283532	0.3033951	0.4848632	0.4400110	0.2443180	0.4785080	0.4870988
##	[988]	0.4730223	0.5029135	0.4051949	0.4349404	0.4375916	0.2711699	0.3698553
##	[995]	0.3630051	0.4035365	0.4355128	0.3208483	0.4481964	0.2733270	0.3640800
##	[1002]	0.2420100	0.3377047	0.5287065	0.2480347	0.5178466	0.2577651	0.3440275
##	[1009]	0.5266852	0.4577362	0.5159199	0.5080937	0.2592893	0.2897177	0.2445142
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##	[1037]	0.2526220	0.2809878	0.4194796	0.4821540	0.3595168	0.3939206	0.3590378
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##	[1065]	0.4038713	0.4102671	0.4975707	0.2463826	0.3191005	0.3254122	0.4032830
##	[1072]	0.3226357	0.4155857	0.2733017	0.2980812	0.4316982	0.2988155	0.5333677
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##	[1100]	0.5075970	0.4971207	0.4076870	0.4674370	0.2584108	0.4426853	0.4522018
##	[1107]	0.4934538	0.4938624	0.3607824	0.3325883	0.4054611	0.3783555	0.3116668
##	[1114]	0.4905883	0.4687612	0.4181315	0.4476711	0.4411853	0.4261953	0.4368084
##	[1121]	0.5106912	0.4573537	0.3997417	0.3225039	0.5039620	0.4107025	0.3496522
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##	[1135]	0.5265711	0.3659875	0.3798718	0.4553578	0.5081402	0.5209020	0.4105511
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##	[1163]	0.5078539	0.4746200	0.4505417	0.4478311	0.4517256	0.2879764	0.4504458
##	[1170]	0.5114462	0.3414067	0.4525820	0.5130151	0.4653291	0.4718020	0.5193898
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##	[1212]	0.3203811	0.3190947	0.4396091	0.3395999	0.3799711	0.4056901	0.3915390
##	[1219]	0.4244635	0.4450716	0.3488210	0.3202807	0.2859445	0.4424051	0.3227229
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## [1443] 0.4966476 0.5203898 0.4899053 0.5041477 0.4784914 0.3710425 0.5260924
## [1450] 0.4279688 0.2724120 0.4837881 0.3572090 0.5127337 0.3346410 0.5127122
## [1457] 0.4242668 0.4215793 0.5073499 0.3925053 0.4537279 0.4404301 0.5096892
## [1464] 0.3406326 0.4843224 0.5048804 0.4918989 0.4822941 0.4871762 0.4509765
## [1471] 0.4350172 0.5029455 0.5140243 0.4624439 0.4977704 0.4529631 0.5093439
## [1478] 0.4424088 0.3763829 0.4144751 0.4556798 0.4594789 0.3541800 0.4329267
## [1485] 0.5034282 0.4517216 0.3209753 0.3125037 0.4812543 0.4952512 0.4687624
## [1492] 0.3804504 0.4080649 0.4255314 0.4334630 0.5151542 0.4919205 0.4924121
## [1499] 0.5013548 0.5160143 0.4406557 0.4849261 0.3256026 0.4413585 0.3317001
## [1506] 0.4441477 0.5208584 0.4198989 0.4082633 0.4958894 0.4941410 0.4601938
## [1513] 0.4998443 0.4546887 0.4857538 0.5056371 0.3948827 0.4056826 0.4225661
## [1520] 0.4387862 0.4358096 0.5184171 0.4732543 0.4504168 0.5048444 0.3213924
## [1527] 0.4870699 0.3786128 0.4150528 0.3758158 0.4861151 0.5044462 0.4123774
## [1534] 0.4352723 0.4634393 0.4541350 0.3873280 0.5201398 0.4542791 0.3741547
## [1541] 0.4993524 0.4261520 0.3301408 0.4439425 0.4767146 0.4073490 0.5212280
## [1548] 0.4480909 0.4535537 0.3942577 0.3043656 0.4668100 0.5021309 0.5200084
## [1555] 0.5104996 0.3113638 0.4787613 0.3270482 0.3868366 0.4342617 0.4466205
## [1562] 0.4210531 0.4049825 0.2731248 0.5175241 0.4612187 0.4485574 0.4015450
## [1569] 0.5250513 0.4125412 0.4622647 0.5058962 0.4793798 0.4800249 0.5260967
## [1576] 0.4084730 0.4377917 0.2884272 0.4997759 0.4228947 0.5008051 0.5089459
## [1583] 0.4477784 0.4644696 0.5277670 0.3754511 0.4497152 0.4576562 0.4418189
## [1590] 0.3477611 0.4598618 0.5248961 0.4356735 0.4369958 0.5006402 0.5109260
## [1597] 0.4053919 0.4949930 0.3585773 0.2999635 0.3372020 0.4069694 0.5046490
## [1604] 0.4020873 0.5191684 0.4972065 0.5283023 0.5223960 0.4129596 0.5200084
## [1611] 0.4240173 0.5212058 0.3407619 0.4179764 0.4457141 0.4936089 0.3481961
## [1618] 0.4185368 0.5221738 0.2867243 0.3462467 0.4661301 0.4499796 0.5191380

```

##	[1625]	0.5223960	0.5277670	0.3838713	0.3979241	0.3845763	0.5110165	0.4577745
##	[1632]	0.4040396	0.5215344	0.5054188	0.3318778	0.4956593	0.3318281	0.4155314
##	[1639]	0.3149034	0.5001885	0.3912766	0.5217243	0.4475816	0.3905089	0.5031136
##	[1646]	0.4876507	0.4100389	0.3895370	0.5179508	0.4859382	0.5297480	0.4551478
##	[1653]	0.4995104	0.3943315	0.3514473	0.4593389	0.5277670	0.3889434	0.4639519
##	[1660]	0.5027747	0.4732947	0.4220048	0.3413078	0.3152034	0.4748484	0.4910379
##	[1667]	0.4533196	0.5188056	0.4803517	0.4032232	0.3645862	0.3862831	0.4348337
##	[1674]	0.4322972	0.5236356	0.4090486	0.5078071	0.4520435	0.4525950	0.3569424
##	[1681]	0.5214238	0.5150377	0.4960598	0.3708088	0.4985417	0.4414871	0.3694571
##	[1688]	0.4267901	0.4002712	0.4121760	0.4936437	0.3212546	0.4305096	0.5135035
##	[1695]	0.3963896	0.4200318	0.2611146	0.5074820	0.5240308	0.5265711	0.4581389
##	[1702]	0.4250385	0.4119115	0.5183511	0.4436773	0.4007029	0.5100954	0.4844638
##	[1709]	0.5200641	0.4201751	0.5142041	0.4373039	0.5212571	0.4577064	0.3610803
##	[1716]	0.4174009	0.4095036	0.3869852	0.3948178	0.3852682	0.3714093	0.4446295
##	[1723]	0.5173112	0.3904721	0.3974721	0.5201488	0.4920389	0.4528984	0.4863321
##	[1730]	0.4074186	0.4276824	0.3266865	0.4727408	0.4367946	0.4028661	0.4061248
##	[1737]	0.4786525	0.4719120	0.3039735	0.3154286	0.3594266	0.5219193	0.4780580
##	[1744]	0.4617646	0.5051834	0.5091959	0.5176695	0.4759463	0.3992711	0.5196317
##	[1751]	0.3946399	0.4655299	0.4052073	0.4082876	0.2933727	0.4235894	0.5026665
##	[1758]	0.4751446	0.3332413	0.4815183	0.4433187	0.3365595	0.3697489	0.3575471
##	[1765]	0.3903277	0.4659068	0.4250650	0.3849458	0.5014552	0.3735846	0.4364550
##	[1772]	0.4384164	0.3653702	0.4515301	0.4463038	0.5198717	0.5223960	0.3550373
##	[1779]	0.5097759	0.4289733	0.5107439	0.3386797	0.4360107	0.3130279	0.3775043
##	[1786]	0.5277670	0.4310861	0.3835082	0.4718020	0.4545102	0.4611574	0.5120182
##	[1793]	0.5213635	0.4951256	0.4951466	0.4042681	0.4836244	0.4565611	0.3594547
##	[1800]	0.3143242	0.3770253	0.4208540	0.4342203	0.5080937	0.4118950	0.4203947
##	[1807]	0.4976509	0.4437480	0.4079953	0.5107307	0.4066230	0.4372846	0.4901143
##	[1814]	0.4497604	0.5167847	0.5118191	0.3858443	0.4820172	0.5091473	0.4248919
##	[1821]	0.4951290	0.3550760	0.4045657	0.3890482	0.5098001	0.4184643	0.4440948
##	[1828]	0.4946667	0.5144490	0.3529215	0.5032972	0.5021548	0.4548017	0.5240966
##	[1835]	0.4556950	0.4654161	0.3479853	0.4326849	0.3824252	0.5163304	0.4553038
##	[1842]	0.4397953	0.5064623	0.4890727	0.3207490	0.3889101	0.4896468	0.4343006
##	[1849]	0.3585623	0.4165826	0.4770064	0.3850337	0.5122684	0.3064394	0.4651830
##	[1856]	0.4296355	0.4359713	0.5091691	0.4309280	0.4026815	0.5277670	0.3087265
##	[1863]	0.3013120	0.4458148	0.4493573	0.4091712	0.4703479	0.5163295	0.4979158
##	[1870]	0.4475638	0.4292322	0.4594780	0.4463142	0.5183919	0.4983775	0.4317023
##	[1877]	0.3657432	0.3905366	0.4824330	0.3778525	0.4990532	0.3907164	0.4461972
##	[1884]	0.3572679	0.4942015	0.4191359	0.4602716	0.3728530	0.4065995	0.4723876
##	[1891]	0.2925247	0.4670268	0.3980566	0.3557082	0.3273305	0.3402681	0.4465846
##	[1898]	0.5029980	0.5103263	0.4446621	0.4311359	0.4499107	0.5184136	0.4482264
##	[1905]	0.3435591	0.4036853	0.2719229	0.5233632	0.3722168	0.4930096	0.5241212
##	[1912]	0.4921866	0.5037354	0.3285624	0.3962073	0.4753412	0.5212571	0.3797314
##	[1919]	0.4667473	0.4926130	0.3589247	0.4972083	0.5169122	0.4541131	0.3876446
##	[1926]	0.4602626	0.4043588	0.4065517	0.5029455	0.3572730	0.3834898	0.5199008
##	[1933]	0.3350587	0.2978913	0.2526993	0.5236356	0.3097355	0.4921671	0.4395644
##	[1940]	0.4263772	0.4014018	0.5100863	0.3788018	0.4609233	0.4616837	0.4620810
##	[1947]	0.4139310	0.4300873	0.4401410	0.4395842	0.4471475	0.4729101	0.4806438
##	[1954]	0.4412190	0.4310938	0.5018921	0.4959105	0.4834192	0.5249707	0.4641964
##	[1961]	0.3514369	0.3705732	0.3748734	0.5212058	0.4788730	0.4229896	0.4744556
##	[1968]	0.4517719	0.4101421	0.5067991	0.5106667	0.4578660	0.3904136	0.5052157
##	[1975]	0.5128578	0.4472007	0.3975229	0.3897610	0.5065958	0.5018276	0.5002392
##	[1982]	0.4833914	0.5253177	0.4004174	0.4423822	0.4420089	0.5048214	0.3103310
##	[1989]	0.4317535	0.5189537	0.3896964	0.3432867	0.4171938	0.5253503	0.5021290
##	[1996]	0.3344425	0.5083434	0.4410658	0.3546063	0.4789949	0.3014026	0.3965599

```
## [2003] 0.4080739 0.3797217 0.4116581 0.3976703 0.4255272 0.4888568 0.3785775
## [2010] 0.3933660 0.4965192 0.5149856 0.4465957 0.4865149 0.5121116 0.2630291
## [2017] 0.4644833 0.3106212 0.4749776 0.5155384 0.3217429 0.3662761 0.4955474
## [2024] 0.4803587 0.4065280 0.4721465 0.3171899 0.4692601 0.5126628 0.5004642
## [2031] 0.5209020 0.3319260 0.4865726 0.3480320 0.3883719 0.3990147 0.4492459
## [2038] 0.4919254 0.3072939 0.5144350 0.5212571 0.4141456 0.4096180 0.4805248
## [2045] 0.4533987 0.5183949 0.3715966 0.5200974 0.4217298 0.4834091 0.4138096
## [2052] 0.4916418 0.4015661 0.4077723 0.3907653 0.4963692 0.3437809 0.4132648
## [2059] 0.3515854 0.4943797 0.3064700 0.5082259 0.3397400 0.3745714 0.4791788
## [2066] 0.3936707 0.4192036 0.3846652 0.5029413 0.4880423 0.4537894 0.5034929
## [2073] 0.4455376 0.4906925 0.4775747 0.5086723 0.5223960 0.5273607 0.4916731
## [2080] 0.5256378 0.4330602 0.4453570 0.3882133 0.3914404 0.4537279 0.4584422
## [2087] 0.3646563 0.3952638 0.4421872 0.4982729 0.4320770 0.4382397 0.3605990
## [2094] 0.4065666 0.3058969 0.3692252 0.3551358 0.4213501 0.2862315 0.3340529
## [2101] 0.4751144 0.3689430 0.4484602 0.4249923 0.5210601 0.4589301 0.5191676
## [2108] 0.3599598 0.4044857 0.4961386 0.4889267 0.5178989 0.3091333 0.4866749
## [2115] 0.5054158 0.3763911 0.4417171 0.5050683 0.4972325 0.4469730 0.3226085
## [2122] 0.5178989 0.4246372 0.4338195 0.4818553 0.4342770 0.5146331 0.4655531
## [2129] 0.4029186 0.4842781 0.4838781 0.4835916 0.5172865 0.4314756 0.4498697
## [2136] 0.2909402 0.5200598 0.3935388 0.4485244 0.3052052 0.5247393 0.4571644
## [2143] 0.4074055 0.3923029 0.4995140 0.4851179 0.3313621 0.5073830 0.4962912
## [2150] 0.4742469 0.4311456 0.4371466 0.3165552 0.3186550 0.4236628 0.4549449
## [2157] 0.5150992 0.3718849 0.3603331 0.3254122 0.4563500 0.2840436 0.5170422
## [2164] 0.4272580 0.4501518 0.3770963 0.5212058 0.4429468 0.3739283 0.3820310
## [2171] 0.4395728 0.2811430 0.4440070 0.4223111 0.3893688 0.4688476 0.4749514
## [2178] 0.4301926 0.4991201 0.4329813 0.4983901 0.4553344 0.5249813 0.4709925
## [2185] 0.5142212 0.5241286 0.2514505 0.4742971 0.5057014 0.3521234 0.3501947
## [2192] 0.3102723 0.3894873 0.4096069 0.3984634 0.4472556 0.4352035 0.3944463
## [2199] 0.4560962 0.3186417 0.4233926 0.4318051 0.4198159 0.5223960 0.4284945
## [2206] 0.4902417 0.4681692 0.4494077 0.5265711 0.3982494 0.4162528 0.4262469
```

Here the threshold value is set to be 35% ,that we are becoming more pickier while selecting customer

```
predictions_sample<-ifelse(predictions_sample>0.35,1,0)
```

```
Final_data_sample<-cbind(Validation_sample,predictions_sample)
```

```
Final_data_sample$predictions_sample<-as.factor(Final_data_sample$predictions_sample)
```

```
Final_data_sample$default<-as.factor(Final_data_sample$default)
```

#Building Confusion Matrix

```
confusionMatrix_sample<-confusionMatrix(Final_data_sample$default,Final_data_sample$predictions_sample)
```

```
confusionMatrix_sample
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
```

```
## Prediction    0    1
```

```
##              0 319 787
```

```
##          1 120 986
##
##          Accuracy : 0.59
##          95% CI : (0.5691, 0.6106)
##    No Information Rate : 0.8015
##    P-Value [Acc > NIR] : 1
##
##          Kappa : 0.1799
##
##    McNemar's Test P-Value : <2e-16
##
##          Sensitivity : 0.7267
##          Specificity : 0.5561
##    Pos Pred Value : 0.2884
##    Neg Pred Value : 0.8915
##          Prevalence : 0.1985
##    Detection Rate : 0.1442
##    Detection Prevalence : 0.5000
##    Balanced Accuracy : 0.6414
##
##    'Positive' Class : 0
##
```

Calculating Precision for the new model

Precision <- TP / (TP + FP) = 1004 / (1004 + 102) = 90.77%

Running the model on test data

```
Test_data<-read.csv("testv3.csv")

Test_Prediction<-predict(gbm_model, newdata = Test_data, n.trees = 50, type = "response")

Test_Prediction<-ifelse(Test_Prediction>0.35,1,0)

Test_datafile<- cbind(Test_data,Test_Prediction)

Default_test<- subset(Test_datafile, Test_datafile$Test_Prediction == 1)

#Writing data frame to a new csv file containing the list of customers who defaults the loan
#write.csv(Default_test,file="defaulted_test_customers.csv")
```