Capstone Intial Results

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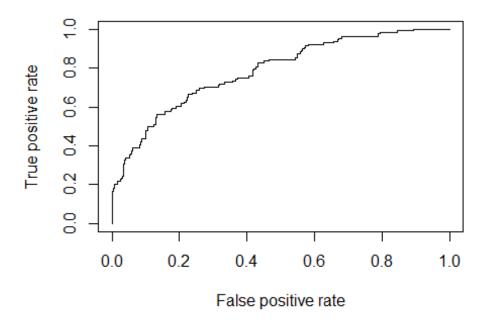
```
#Data Load
df<-
data.frame(read.csv("C:/Users/jatosan/Desktop/CAPSTONE/semifinal/secondtry/25
00 column gone 2016Q1.csv"))
#Data Cleaning and Preparation
#Removing "current" rows
df<-subset(df,!(df$loan_status=="Current"))</pre>
#Coerce lates, grace periods and charge offs to 1 and fully paid to 0
#Get loan status col name index
grep("loan_status", colnames(df))
## [1] 12
#change class of loan status to numeric to coerce loan status attributes
gracep<-which(df$loan_status=="In Grace Period")</pre>
late1<-which(df$loan_status=="Late (16-30 days)")</pre>
late2<-which(df$loan status=="Late (31-120 days)")</pre>
default<-which(df$loan status=="Charged Off")</pre>
loanstat1<-c(gracep,late1,late2,default)</pre>
fpaid<-which(df$loan status=="Fully Paid")</pre>
loanstat0<-fpaid</pre>
#change class of loan status to numeric to coerce
df$loan status<-as.numeric(df$loan status)</pre>
  #Coercing
  df[loanstat1,12]<-1</pre>
  df[loanstat0,12]<-0
#Remove percentage sign in df columns (int_rate, revol_util) and change to
numeric for correlation matrix
int_rateclean<-as.numeric(gsub("%","",df$int_rate))</pre>
```

```
df$int rate<-int rateclean</pre>
revol_utilcleann<-as.numeric(gsub("%","",df$revol_util))</pre>
df$revol util<-revol utilcleann</pre>
#Changing emp length to numeric values
emp_lengthc1<-gsub(" years | year|s|+","",df$emp_length)</pre>
emp_lengthc2<-sub("10+","10",emp_lengthc1,fixed = TRUE)
emp_lengthc3<-sub("< 1","0",emp_lengthc2,fixed = TRUE)</pre>
emp_lengthc4<-sub("n/a","0",emp_lengthc3,fixed = TRUE)</pre>
emp lengthc5<-as.numeric(emp_lengthc4)</pre>
df$emp length<-emp lengthc5</pre>
#Data Cleaning and Preparation continued
#sum(is.na(df$emp title))
#sum(is.na(df$mths since last deling)) #646
#sum(is.na(df$mths_since_last_record)) #1134
#sum(is.na(df$mths since last major derog)) #992
#sum(is.na(df$mths since recent bc dlq)) #1051
#sum(is.na(df$mths since recent ing)) #110
#sum(is.na(df$mths since recent revol deling))#894
#sum(is.na(df$il util)) #194
#Removing the columns with a sizable proportion of the rows are NAs
df<-subset(df ,select=-</pre>
c(mths_since_last_deling,mths_since_last_record,mths_since_last_major_derog,m
ths_since_last_major_derog, mths_since_recent_bc_dlq,
mths since recent inq,mths since recent revol delinq,revol util,mths since rc
nt_il,il_util,all_util,bc_open_to_buy,bc_util,mo_sin_old_il_acct,mths_since_r
ecent_bc,percent_bc_gt_75,last_credit_pull_d))
#Removing columns full of 0s
length(which(df$collections 12 mths ex med==0)) #1367
## [1] 1367
length(which(df$num_tl_90g_dpd_24m==0)) #1325
## [1] 1325
df<-subset(df,select=-c(collections 12 mths ex med,num tl 90g dpd 24m))
df<-subset(df,select=-</pre>
```

```
c(emp_title,earliest_cr_line,grade,sub_grade,purpose,addr_state,verification_
status))
#Logistic Regression Modelling
glm.lr<-glm(df$loan_status ~.,data=df,family = binomial,maxit=100)</pre>
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
vim<-varImp(glm.lr)</pre>
impr<-which((varImp(glm.lr)>0.5))
t1<-apply(vim,2,function(x) x[order(-x)])
t1
##
                                Overall
## tot_hi_cred_lim
                             4.05262350
## tot_cur_bal
                             3.79407180
## total_il_high_credit_limit 3.27648549
## total rev hi lim
                            3.08209871
## open_acc_6m
                            2.78473341
## term 60 months
                            2.29090931
## dti
                            2.26190275
## num_il_tl
                            2.17788010
## num op rev tl
                           1.82202045
## open act il
                           1.81723043
## int_rate
                           1.81031448
## revol bal
                           1.77497292
## total_acc
                            1.71890876
## num_rev_accts
                           1.67142248
## total bal il
                           1.58933020
## mort acc
                           1.53905150
## open acc
                            1.38736224
## num_tl_op_past_12m
                           1.26896910
## num_bc_tl
                            1.26352392
## home ownershipOWN
                            1.23850545
## mo sin rcnt rev tl op
                             1.21708862
## initial_list_statusw
                            1.18456016
## open_rv_12m
                            1.16348306
## delinq_2yrs
                            1.15159567
## avg_cur_bal
                            1.11024994
## home ownershipRENT
                            1.08996412
## installment
                            1.05932391
## open_il_12m
                            0.88198946
## max bal bc
                            0.72996867
## inq_last_6mths
                           0.67160571
## num_bc_sats
                            0.65752154
## pct tl nvr dlq
                            0.53953788
```

```
## loan amnt
                              0.49092166
## num sats
                              0.48391385
## pub_rec_bankruptcies
                              0.44627656
## pub rec
                              0.41694515
## num_actv_bc_tl
                             0.41480107
                             0.41085990
## num_actv_rev_tl
## inq last 12m
                             0.39753949
## total cu tl
                             0.35839992
## open_il_24m
                             0.32814589
## emp_length
                             0.26714819
## num_accts_ever_120_pd
## annual inc
                             0.25578847
## annual inc
                             0.24942979
## mo_sin_rcnt_tl
                             0.12248878
## total_bc_limit
                             0.10112516
## tot_coll_amt
                             0.08598267
## num_rev_tl_bal_gt_0
## total_bal_ex_mort
                             0.07103166
                             0.06980736
## open rv 24m
                             0.05337691
## out_prncp
                             0.03637960
## ing fi
                              0.03444153
## tax_liens
                              0.01842961
#Training and test data indexing
trainindex <- sample(1:nrow(df), 0.7 * nrow(df))</pre>
train.set <- df[trainindex,]</pre>
test.set <- df[-trainindex,]</pre>
#Prediction and accuracy
glm.pred<-predict(glm.lr,newdata=test.set,type='response')</pre>
glm.pred<-ifelse(glm.pred>0.5,1,0)
misclasserr<-mean(glm.pred ==test.set$loan_status)</pre>
print(paste('Accuracy',1-misclasserr))
## [1] "Accuracy 0.235154394299287"
glm<-glm.pred</pre>
#ROC curve, report on accuracies of models.
rocp<-predict(glm.lr,newdata=test.set,type="response")</pre>
pr<-prediction(rocp,test.set$loan status)</pre>
```

```
perf <- performance(pr, measure = "tpr", x.measure = "fpr")
plot(perf)</pre>
```



```
auc <- performance(pr, measure = "auc")
auc <- auc@y.values[[1]]
auc
## [1] 0.7859695</pre>
```