

PhishingAttack

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```
library("rpart")
library("ggplot2")
library("tidyverse")
```

```
## — Attaching packages — tidyverse 1.3.2 —
##
## ✓ tibble 3.1.8      ✓ dplyr 1.0.10
## ✓ tidyr 1.2.0      ✓ stringr 1.4.1
## ✓ readr 2.1.2      ✓ forcats 0.5.2
## ✓ purrr 0.3.4
## — Conflicts — tidyverse_conflicts() —
##
## ✖ dplyr::filter() masks stats::filter()
## ✖ dplyr::lag() masks stats::lag()
```

```
library("psych")
```

```
##
## Attaching package: 'psych'
##
## The following objects are masked from 'package:ggplot2':
##
## %+, alpha
```

```
library("corrplot")
```

```
## corrplot 0.92 loaded
```

```
library("RColorBrewer")
```

Load Dataset

```
q5=read.csv("/Users/abhinavram/Documents/IDS572 Data Mining/Assignment 1/Q5.csv")
```

Summary of dataset

```
summary(q5)
```

```
##           id           having_IP_Address      URL_Length      Shortining_Service
##  Min.      :      1      Min.      :-1.0000      Min.      :-1.0000      Min.      :-1.0000
```

##	1st Qu.: 2764	1st Qu.: -1.0000	1st Qu.: -1.0000	1st Qu.: 1.0000
##	Median : 5528	Median : 1.0000	Median : -1.0000	Median : 1.0000
##	Mean : 5528	Mean : 0.3138	Mean : -0.6332	Mean : 0.7388
##	3rd Qu.: 8292	3rd Qu.: 1.0000	3rd Qu.: -1.0000	3rd Qu.: 1.0000
##	Max. : 11055	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000
##	having_At_Symbol	double_slash_redirecting	Prefix_Suffix	having_Sub_Domain
##	Min. : -1.0000	Min. : -1.0000	Min. : -1.000	Min. : -1.00000
##	1st Qu.: 1.0000	1st Qu.: 1.0000	1st Qu.: -1.000	1st Qu.: -1.00000
##	Median : 1.0000	Median : 1.0000	Median : -1.000	Median : 0.00000
##	Mean : 0.7006	Mean : 0.7415	Mean : -0.735	Mean : 0.06395
##	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: -1.000	3rd Qu.: 1.00000
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.000	Max. : 1.00000
##	SSLfinal_State	Domain_registration_length	Favicon	
##	Min. : -1.0000	Min. : -1.0000	Min. : -1.0000	
##	1st Qu.: -1.0000	1st Qu.: -1.0000	1st Qu.: 1.0000	
##	Median : 1.0000	Median : -1.0000	Median : 1.0000	
##	Mean : 0.2509	Mean : -0.3368	Mean : 0.6286	
##	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000	
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	
##	port	HTTPS_token	Request_URL	URL_of_Anchor
##	Min. : -1.0000	Min. : -1.0000	Min. : -1.0000	Min. : -1.00000
##	1st Qu.: 1.0000	1st Qu.: 1.0000	1st Qu.: -1.0000	1st Qu.: -1.00000
##	Median : 1.0000	Median : 1.0000	Median : 1.0000	Median : 0.00000
##	Mean : 0.7283	Mean : 0.6751	Mean : 0.1868	Mean : -0.07653
##	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 0.00000
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.00000
##	Links_in_tags	SFH	Submitting_to_email	Abnormal_URL
##	Min. : -1.0000	Min. : -1.0000	Min. : -1.0000	Min. : -1.0000
##	1st Qu.: -1.0000	1st Qu.: -1.0000	1st Qu.: 1.0000	1st Qu.: 1.0000
##	Median : 0.0000	Median : -1.0000	Median : 1.0000	Median : 1.0000
##	Mean : -0.1181	Mean : -0.5957	Mean : 0.6356	Mean : 0.7053
##	3rd Qu.: 0.0000	3rd Qu.: -1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000
##	Redirect	on_mouseover	RightClick	popUpWidnow
##	Min. : 0.0000	Min. : -1.0000	Min. : -1.0000	Min. : -1.0000
##	1st Qu.: 0.0000	1st Qu.: 1.0000	1st Qu.: 1.0000	1st Qu.: 1.0000
##	Median : 0.0000	Median : 1.0000	Median : 1.0000	Median : 1.0000
##	Mean : 0.1157	Mean : 0.7621	Mean : 0.9139	Mean : 0.6134
##	3rd Qu.: 0.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000
##	Iframe	age_of_domain	DNSRecord	web_traffic
##	Min. : -1.0000	Min. : -1.00000	Min. : -1.0000	Min. : -1.0000
##	1st Qu.: 1.0000	1st Qu.: -1.00000	1st Qu.: -1.0000	1st Qu.: 0.0000
##	Median : 1.0000	Median : 1.00000	Median : 1.0000	Median : 1.0000
##	Mean : 0.8169	Mean : 0.06124	Mean : 0.3771	Mean : 0.2873
##	3rd Qu.: 1.0000	3rd Qu.: 1.00000	3rd Qu.: 1.0000	3rd Qu.: 1.0000
##	Max. : 1.0000	Max. : 1.00000	Max. : 1.0000	Max. : 1.0000
##	Page_Rank	Google_Index	Links_pointing_to_page	Statistical_report
##	Min. : -1.0000	Min. : -1.0000	Min. : -1.000	Min. : -1.0000
##	1st Qu.: -1.0000	1st Qu.: 1.0000	1st Qu.: 0.000	1st Qu.: 1.0000
##	Median : -1.0000	Median : 1.0000	Median : 0.000	Median : 1.0000
##	Mean : -0.4837	Mean : 0.7216	Mean : 0.344	Mean : 0.7196
##	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.000	3rd Qu.: 1.0000
##	Max. : 1.0000	Max. : 1.0000	Max. : 1.000	Max. : 1.0000

```
##      Result
## Min.      :-1.0000
## 1st Qu.: -1.0000
## Median :  1.0000
## Mean      : 0.1139
## 3rd Qu.:  1.0000
## Max.      :  1.0000
```

Checking for missing values

```
sapply(q5, function(x) sum(is.na(x)))
```

```
##              id          having_IP_Address
##              0                          0
##      URL_Length      Shortining_Service
##              0                          0
##      having_At_Symbol  double_slash_redirecting
##              0                          0
##      Prefix_Suffix     having_Sub_Domain
##              0                          0
##      SSLfinal_State  Domain_registration_length
##              0                          0
##      Favicon          port
##              0                          0
##      HTTPS_token      Request_URL
##              0                          0
##      URL_of_Anchor     Links_in_tags
##              0                          0
##      SFH              Submitting_to_email
##              0                          0
##      Abnormal_URL      Redirect
##              0                          0
##      on_mouseover      RightClick
##              0                          0
##      popUpWidnow       Iframe
##              0                          0
##      age_of_domain     DNSRecord
##              0                          0
##      web_traffic       Page_Rank
##              0                          0
##      Google_Index      Links_pointing_to_page
##              0                          0
##      Statistical_report      Result
##              0                          0
```

Split data into training and test

```

set.seed(1)
sample <- sample(c(TRUE, FALSE), nrow(q5), replace=TRUE, prob=c(0.7,0.3))
q5_train <- q5[sample, ]
q5_test  <- q5[!sample, ]

q5_train = select(q5_train, -id)
q5_test = select(q5_test, -id)

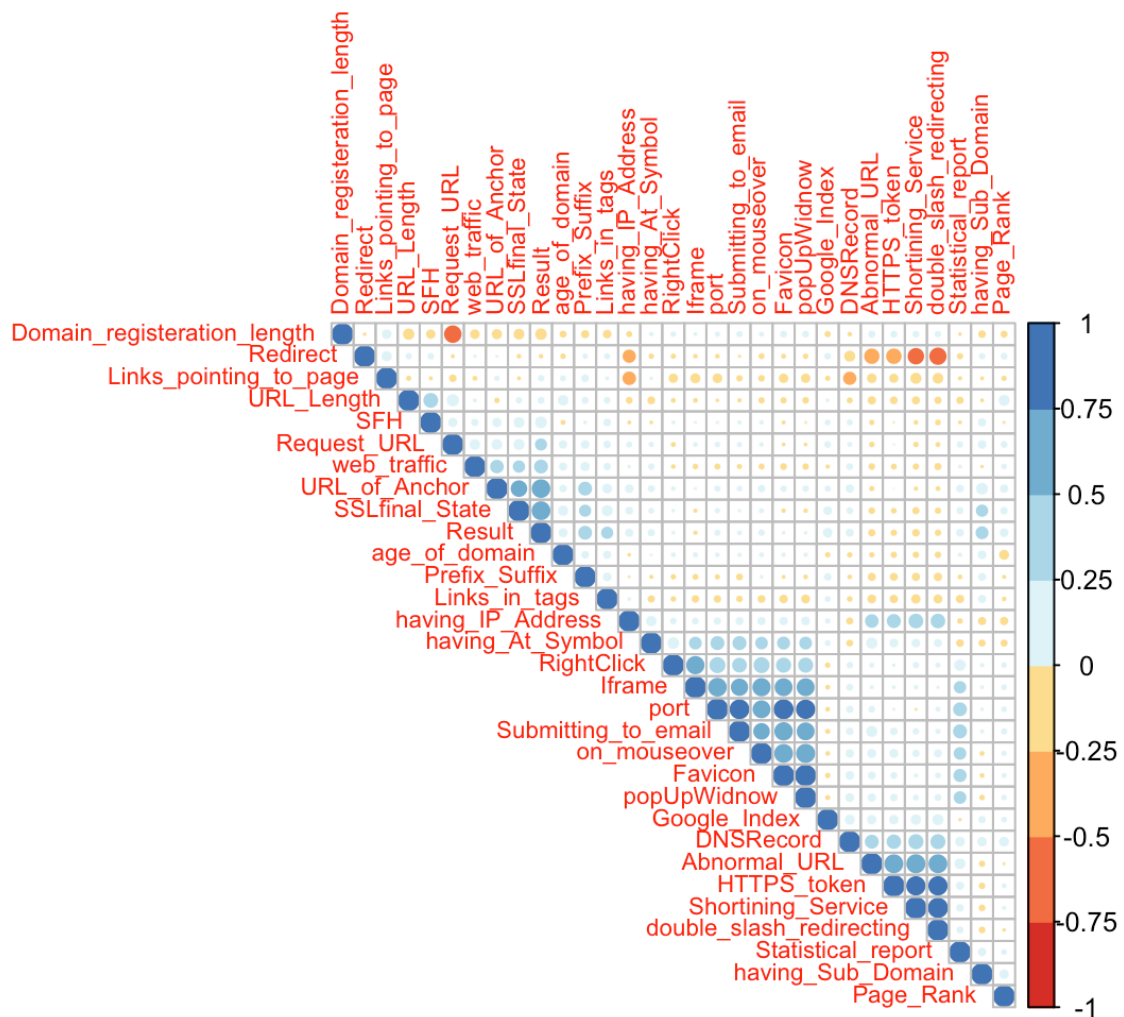
```

Correlation Plot

```

cor_graph_phishing <- cor(q5_train)
corrplot(cor_graph_phishing, type="upper", order= "hclust", tl.cex = 0.7, col=brewer
.pal(n=8, name="RdYlBu"))

```

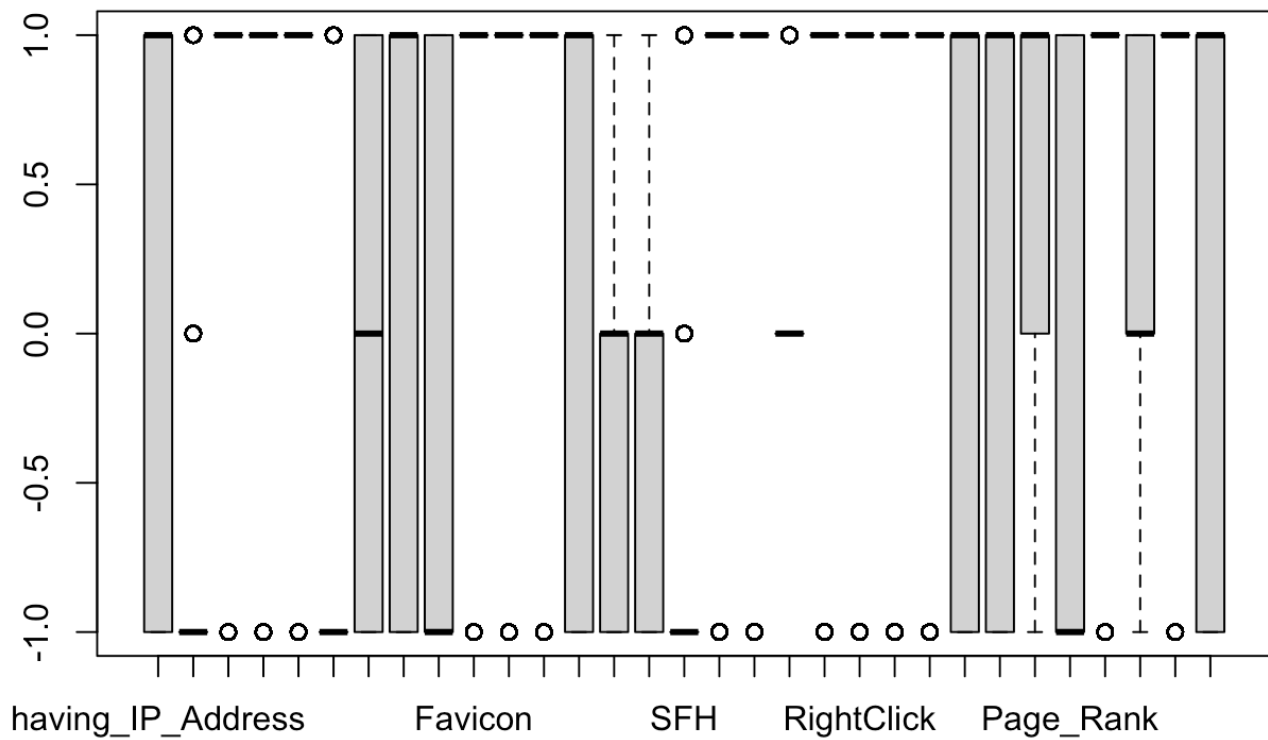


Box Plot

```

boxplot(q5_train)

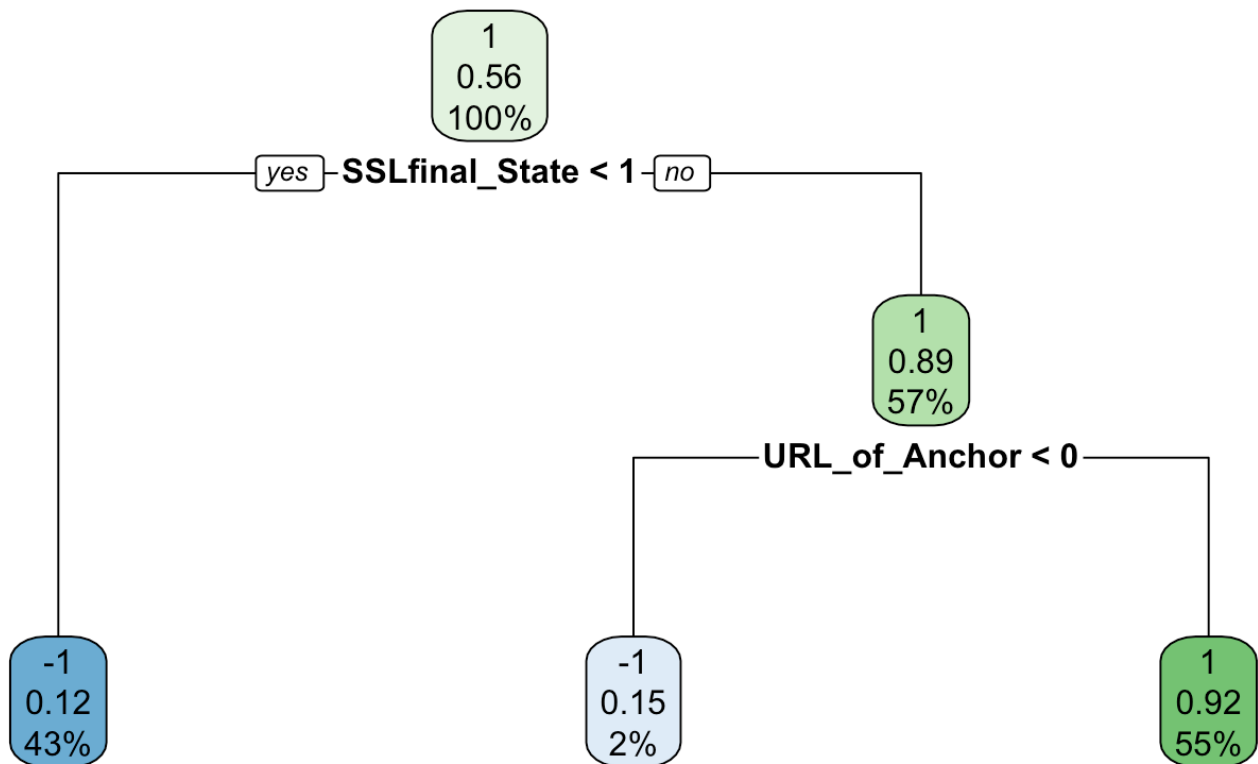
```



No outliers in this data

Decision Tree

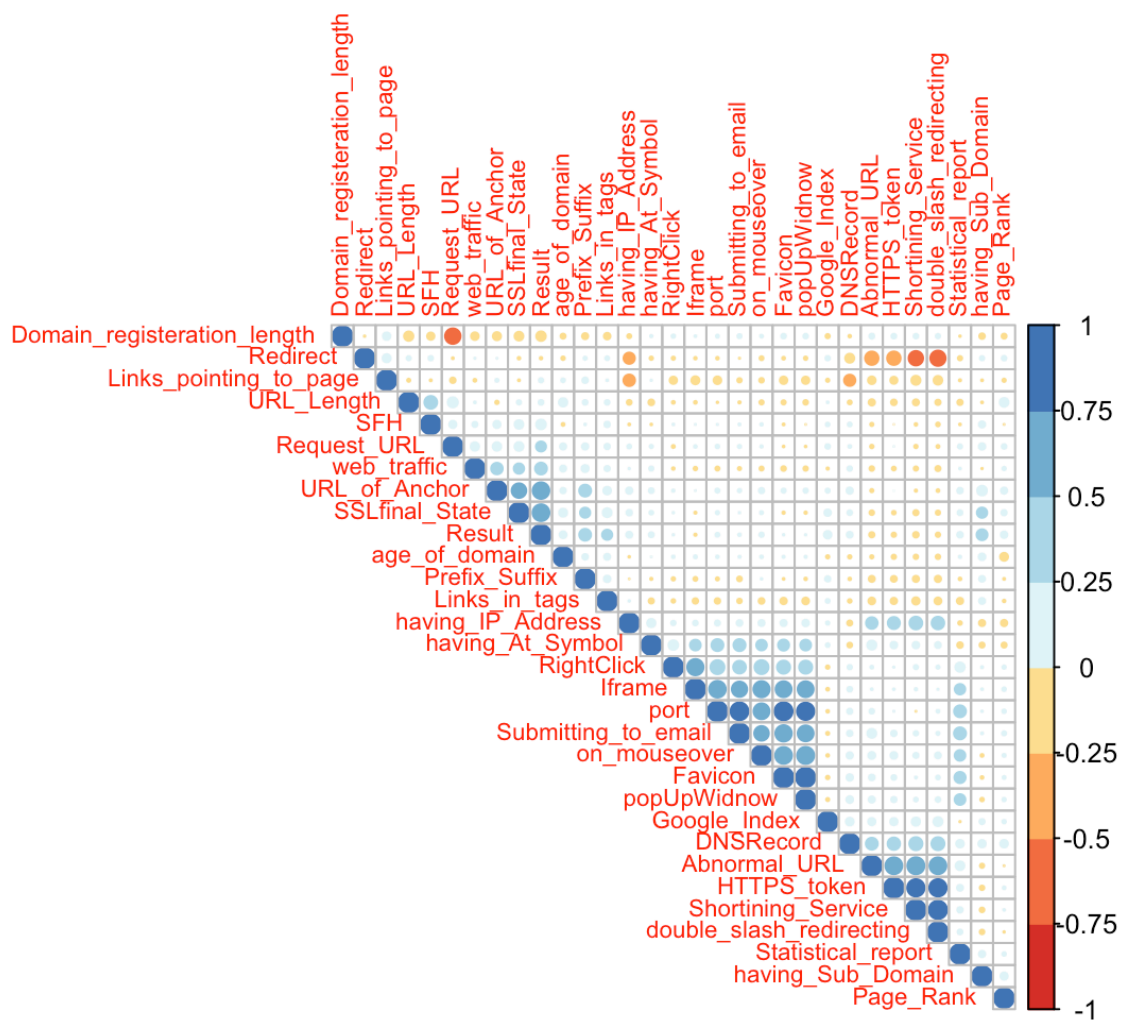
```
library(rpart.plot)
fit_phishing=rpart(Result~.,data=q5_train, parms = list(split="information"), method = 'class')
rpart.plot(fit_phishing, extra = 106)
```



```

q5_train$Result <- as.numeric(q5_train$Result)
cor_graph_phishing <- cor(q5_train)
corrplot(cor_graph_phishing, type="upper", order= "hclust", tl.cex = 0.7, col=brewer.
pal(n=8, name="RdYlBu"))

```



Confusion Matrix for combined and accuracy

```
t_pred_phishing=predict(fit_phishing, q5_train, type='class')
table(q5_train$Result, t_pred_phishing)
```

```
##      t_pred_phishing
##      -1      1
## -1 3070   339
##  1   410 3884
```

```
confusion_mat_phishing_dt = table(q5_train$Result, t_pred_phishing)
acc_phishing_dt = sum(diag(confusion_mat_phishing_dt))/sum(confusion_mat_phishing_dt)
print(confusion_mat_phishing_dt)
```

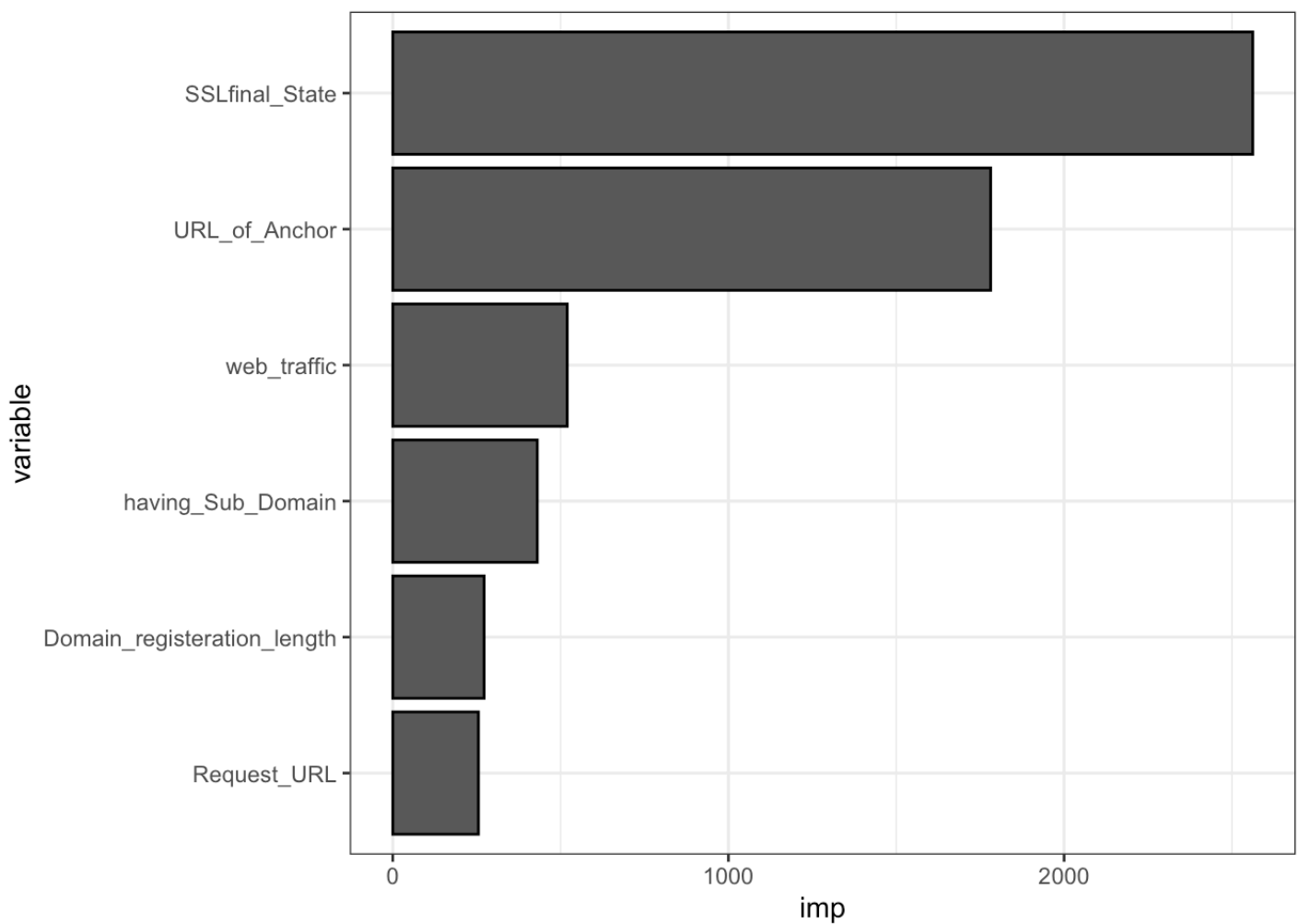
```
##      t_pred_phishing
##      -1      1
## -1 3070   339
##  1   410 3884
```

```
print(acc_phishing_dt)
```

```
## [1] 0.9027652
```

ggplot

```
df_phishing=data.frame(imp=fit_phishing$variable.importance)
df2_phishing=df_phishing %>%
  tibble::rownames_to_column()%>%
  dplyr::rename("variable"= rowname)%>%
  dplyr::arrange(imp)%>%
  dplyr::mutate(variable = forcats::fct_inorder(variable))
ggplot2::ggplot(df2_phishing) +
  geom_col(aes(x=variable, y=imp), col="black", show.legend = F) +
  coord_flip() +
  scale_fill_grey() +
  theme_bw()
```



Confusion Matrix for test

```
t_pred_phishing_test=predict(fit_phishing, q5_test, type='class')
table(q5_test$Result, t_pred_phishing_test)
```



```
##      t_pred_phishing_test
##      -1      1
##    -1 1355   134
##     1   153 1710
```

```
confusion_mat_test_phishing_dt = table(q5_test$Result, t_pred_phishing_test)
accTest_phishing_dt = sum(diag(confusion_mat_test_phishing_dt))/sum(confusion_mat_
test_phishing_dt)
print(confusion_mat_phishing_dt)
```

```
##      t_pred_phishing
##      -1      1
##    -1 3070   339
##     1   410 3884
```

```
print(accTest_phishing_dt)
```

```
## [1] 0.9143795
```

Random Forest

```
library(party)
```

```
## Loading required package: grid
```

```
## Loading required package: mvtnorm
```

```
## Loading required package: modeltools
```

```
## Loading required package: stats4
```

```
## Loading required package: strucchange
```

```
## Loading required package: zoo
```

```
##
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric
```

```
## Loading required package: sandwich
```

```
##  
## Attaching package: 'strucchange'
```

```
## The following object is masked from 'package:stringr':  
##  
##      boundary
```

```
library(randomForest)
```

```
## randomForest 4.7-1.1
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##  
## Attaching package: 'randomForest'
```

```
## The following object is masked from 'package:psych':  
##  
##      outlier
```

```
## The following object is masked from 'package:dplyr':  
##  
##      combine
```

```
## The following object is masked from 'package:ggplot2':  
##  
##      margin
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
##  
## Attaching package: 'caret'
```

```
## The following object is masked from 'package:purrr':  
##  
##      lift
```

```
q5_train$Result <- as.character(q5_train$Result)
q5_train$Result <- as.factor(q5_train$Result)
set.seed(1234)
fit_rf=randomForest(Result~ SSLfinal_State + URL_of_Anchor + web_traffic , data=q5_train, ntree=500,
                    importance=TRUE, proximity=TRUE)
```

View the forest results

```
print(fit_rf)
```

```
##
## Call:
## randomForest(formula = Result ~ SSLfinal_State + URL_of_Anchor + web_traf
fic, data = q5_train, ntree = 500, importance = TRUE, proximity = TRUE)
##
##           Type of random forest: classification
##           Number of trees: 500
## No. of variables tried at each split: 1
##
##           OOB estimate of  error rate: 8.72%
## Confusion matrix:
##      -1      1 class.error
## -1 3034   375   0.11000293
##  1   297 3997   0.06916628
```

Importance of each predictor.

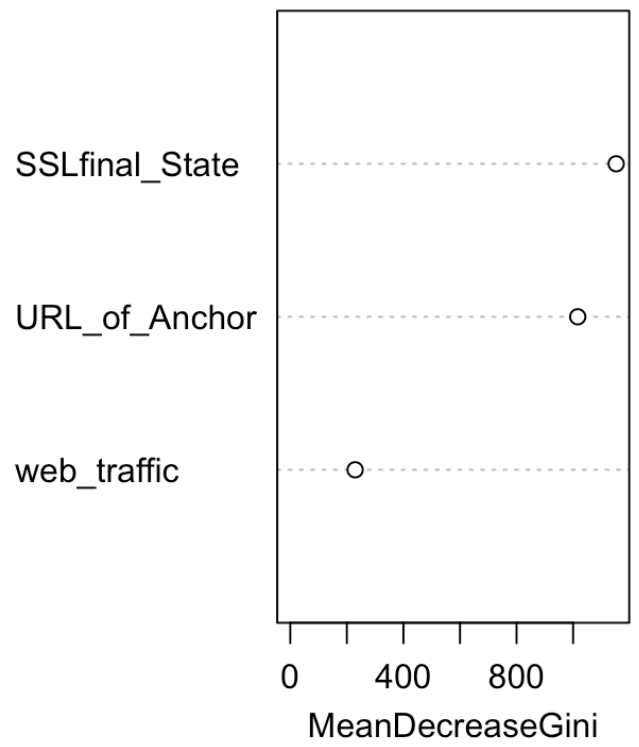
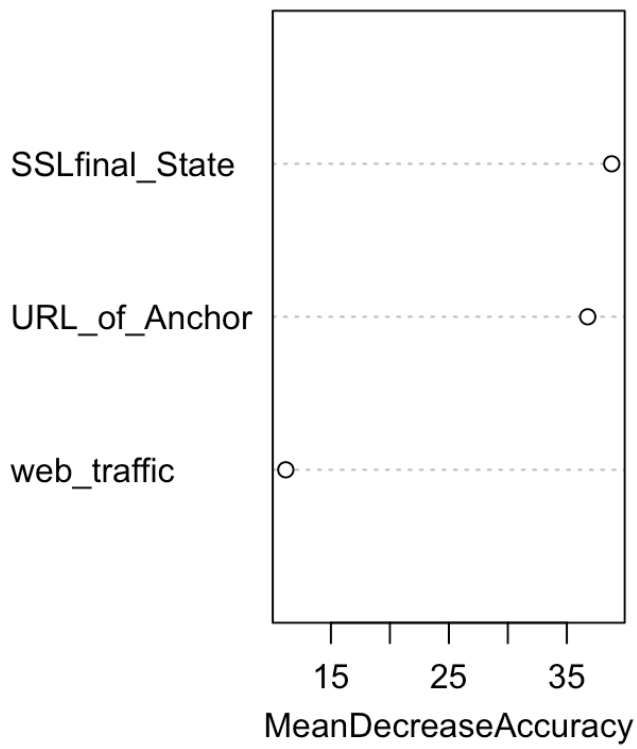
```
out.importance <- round(importance(fit_rf), 2)
print(out.importance )
```

```
##           -1      1 MeanDecreaseAccuracy MeanDecreaseGini
## SSLfinal_State 23.36 43.68                38.80          1152.25
## URL_of_Anchor  20.94 36.98                36.77          1016.31
## web_traffic     9.73  8.75                 11.17           229.32
```

Graph of RF Model

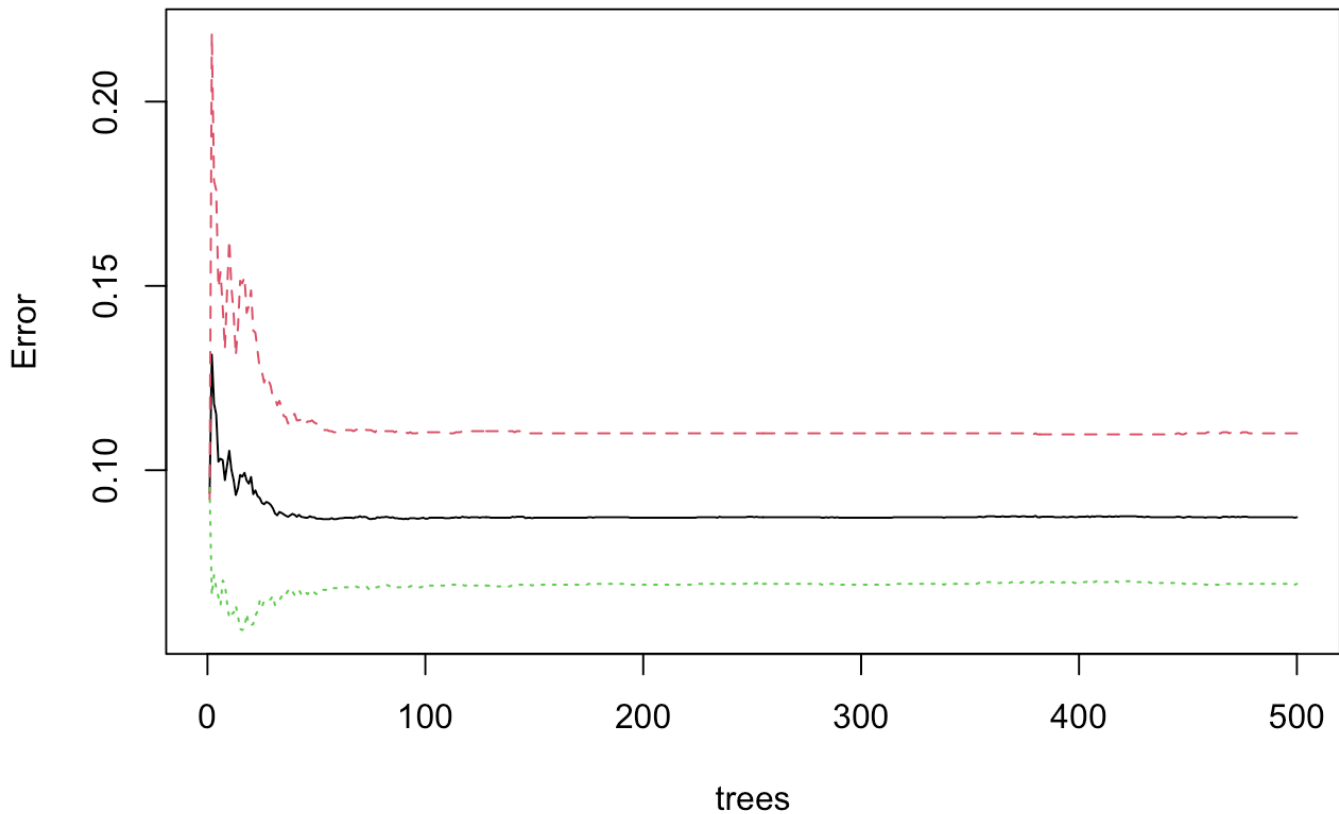
```
varImpPlot(fit_rf)
```

fit_rf



```
plot(fit_rf)
```

fit_rf



As we increase the number of trees, the error decreases exponentially. This is why we prefer Random Forest over Decision Tree. Moreover, for large quantities of data, Random Forest is more effective and quicker than Decision Tree.

From the above data we can observe %IncMSE and IncNodePurity where they define Mean Decrease Accuracy and Mean Decrease Gini respectively. With the former, it shows how much our accuracy will get affected without that particular variable and the latter shows how important a variable is (higher is better).

Accuracy of Decision Tree = 90.27% Accuracy of Random Forest = 91.57%