

## mtb12

```
m12r<-read.csv("/Users/i Dolphin Online/Documents/12rv.csv")
library(randomForest)
```

```
## randomForest 4.6-14
```

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

```
m12rv<- randomForest(yield~ . ,data = m12r,importance = TRUE)
library(caret)
```

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

```
## Loading required package: ggplot2
```

```
##
```

```
## Attaching package: 'ggplot2'
```

```
## The following object is masked from 'package:randomForest':
```

```
##
```

```
##      margin
```

```
imp <- as.data.frame(varImp(m12rv))
imp <- data.frame(overall = imp$Overall, names = rownames(imp))
imp[order(imp$overall, decreasing = T),]
```

```
##      overall      names
## 1  10.252758 Total.Foreign.Investment.lag
## 9   9.511670      PKR.PER.USD.day.lag
## 53  9.035928      PK10YRPAV.FMAP.MA
## 24  8.391071      PK3YRPAV.FMAP.LAG
## 23  7.301245      PK2YRPAV.FMAP.LAG
## 2   7.193942      NET.RESERVES.WITH.SBP.lag
## 33  7.091402      PK20YRPAV.FMAP.LAG
## 26  6.940297      PK5YRPAV.FMAP.LAG
## 25  6.934011      PK4YRPAV.FMAP.LAG
## 3   6.833931      Workers..Remittances.lag
## 11  6.067255      TREASURY.CMT.LAG
## 32  5.947013      PK15YRPAV.FMAP.LAG
## 27  5.915033      PK6YRPAV.FMAP.LAG
## 5   5.886451      inflation.yoy
## 10  5.784331      LT.COMPOSITE.LAG
## 57  5.768475      KHIPKR2WD..MA
## 28  5.607992      PK7YRPAV.FMAP.LAG
```

```
## 56 5.506922 KHIPKRSWD..MA
## 4 5.486603 Trade.Deficit.lag
## 31 5.450170 PK10YRPAV.FMAP.LAG
## 29 5.439493 PK8YRPAV.FMAP.LAG
## 30 5.411812 PK9YRPAV.FMAP.LAG
## 39 5.247966 PK1MRPAV.FMAP.MA
## 22 5.161166 PK1YRPAV.FMAP.LAG
## 38 5.095318 PKSWRPAV.FMAP.MA
## 20 4.589725 PK6MRPAV.FMAP.LAG
## 58 4.559148 KHIPKR1MD..MA
## 59 4.470795 KHIPKR3MD..MA
## 13 4.459808 TREASURY.CMT.MA
## 45 4.331706 PK2YRPAV.FMAP.MA
## 54 4.251588 PK15YRPAV.FMAP.MA
## 7 4.249102 oil.prices.lag
## 14 4.231202 PKSWRPAV.FMAP.lag
## 6 4.158658 inflation.mom
## 12 4.052987 LT.COMPOSITE.MA
## 55 3.974345 PK20YRPAV.FMAP.MA
## 46 3.897783 PK3YRPAV.FMAP.MA
## 34 3.598405 KHIPKRSWD..LAG
## 50 3.562165 PK7YRPAV.FMAP.MA
## 19 3.506980 PK4MRPAV.FMAP.LAG
## 52 3.457336 PK9YRPAV.FMAP.MA
## 49 3.393734 PK6YRPAV.FMAP.MA
## 8 3.330059 OIL.PRICE.MA
## 21 3.247186 PK9MRPAV.FMAP.LAG
## 37 3.181901 KHIPKR6MD..LAG
## 16 2.973169 PK1MRPAV.FMAP.lag
## 15 2.763170 PK2WRPAV.FMAP.lag
## 40 2.737050 PK3MRPAV.FMAP.MA
## 61 2.634922 KHIPKR9MD..MA
## 62 2.586863 KHIPKR1YD..MA
## 60 2.462752 KHIPKR6MD..MA
## 51 2.388734 PK8YRPAV.FMAP.MA
## 35 2.205892 KHIPKR2WD..LAG
## 43 1.990580 PK9MRPAV.FMAP.MA
## 18 1.974998 PK3MRPAV.FMAP.lag
## 36 1.973904 KHIPKR1MD..LAG
## 48 1.866437 PK5YRPAV.FMAP.MA
## 41 1.816087 PK4MRPAV.FMAP.MA
## 17 1.807878 PK2MRPAV.FMAP.lag
## 44 1.575413 PK1YRPAV.FMAP.MA
## 42 1.571376 PK6MRPAV.FMAP.MA
## 47 1.341588 PK4YRPAV.FMAP.MA
```

```
m12c<-read.csv("/Users/i Dolphin Online/Documents/12c.csv")
tmp <- cor(m12c)
tmp[upper.tri(tmp)] <- 0
diag(tmp) <- 0
data.new <- m12c[,! apply(tmp,2,function(x) any(x > 0.8))]
head(data.new)
```

```
## Total.Foreign.Investment.lag NET.RESERVES.WITH.SBP.lag
```

```
## 1          270.1325          7862.6
## 2          270.1325          7862.6
## 3          270.1325          7862.6
## 4          270.1325          7862.6
## 5          270.1325          7862.6
## 6          270.1325          7862.6
##   Workers..Remittances.lag Trade.Deficit.lag PKR.PER.USD.day.lag KHIPKR1MD..MA
## 1          2302.02          -2061.011          146.3007          12.024
## 2          2302.02          -2061.011          146.3007          12.024
## 3          2302.02          -2061.011          146.3007          12.024
## 4          2302.02          -2061.011          146.3007          12.024
## 5          2302.02          -2061.011          146.3007          12.024
## 6          2302.02          -2061.011          157.0838          12.302
```

```
m12<-read.csv("/Users/i Dolphin Online/Documents/12.csv")
ind <- sample(2, nrow(m12), replace = TRUE, prob = c(0.7 , 0.3))
train12<-m12[ind==1,]
test12<-m12[ind==2,]
library(randomForest)
rf <-randomForest(yield~.,data=train12,mtry=4)
rf
```

```
##
## Call:
## randomForest(formula = yield ~ ., data = train12, mtry = 4)
##           Type of random forest: regression
##           Number of trees: 500
## No. of variables tried at each split: 4
##
##           Mean of squared residuals: 1.004854e-06
##           % Var explained: 99.82
```

```
pred_values2 = predict(rf,test12)
actual_values2 = test12$yield
library(Metrics)
```

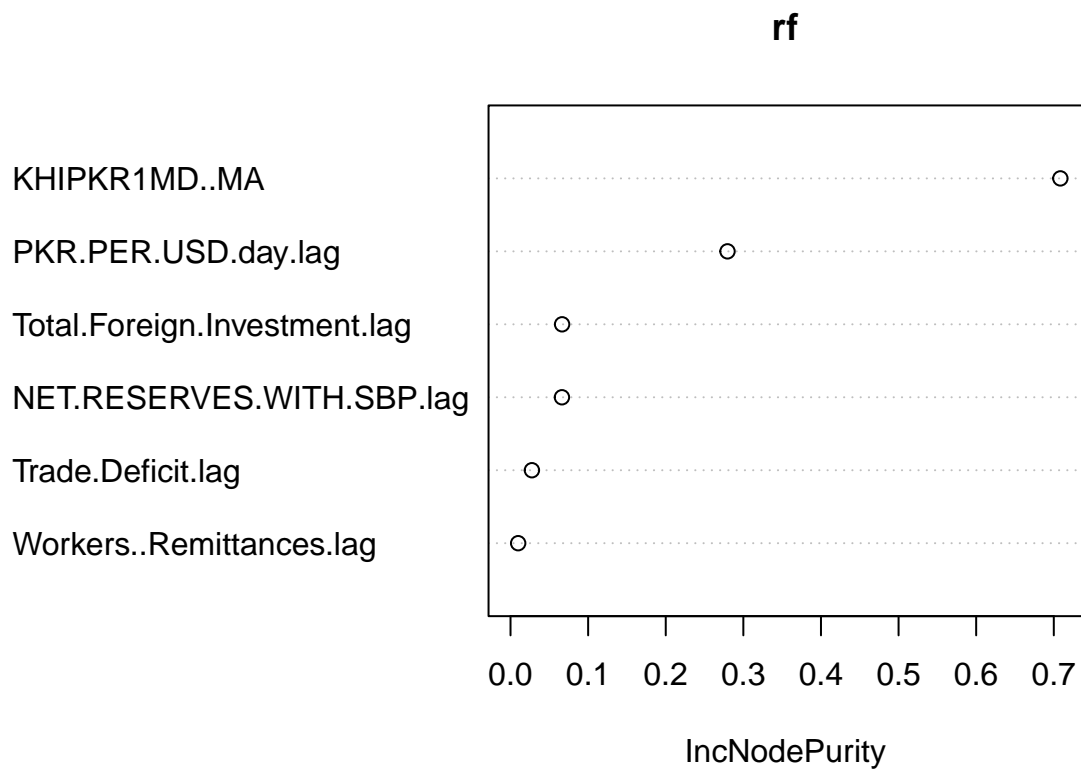
```
##
## Attaching package: 'Metrics'
```

```
## The following objects are masked from 'package:caret':
##
##   precision, recall
```

```
metrics_rmse = rmse(actual_values2,pred_values2)
print(metrics_rmse)
```

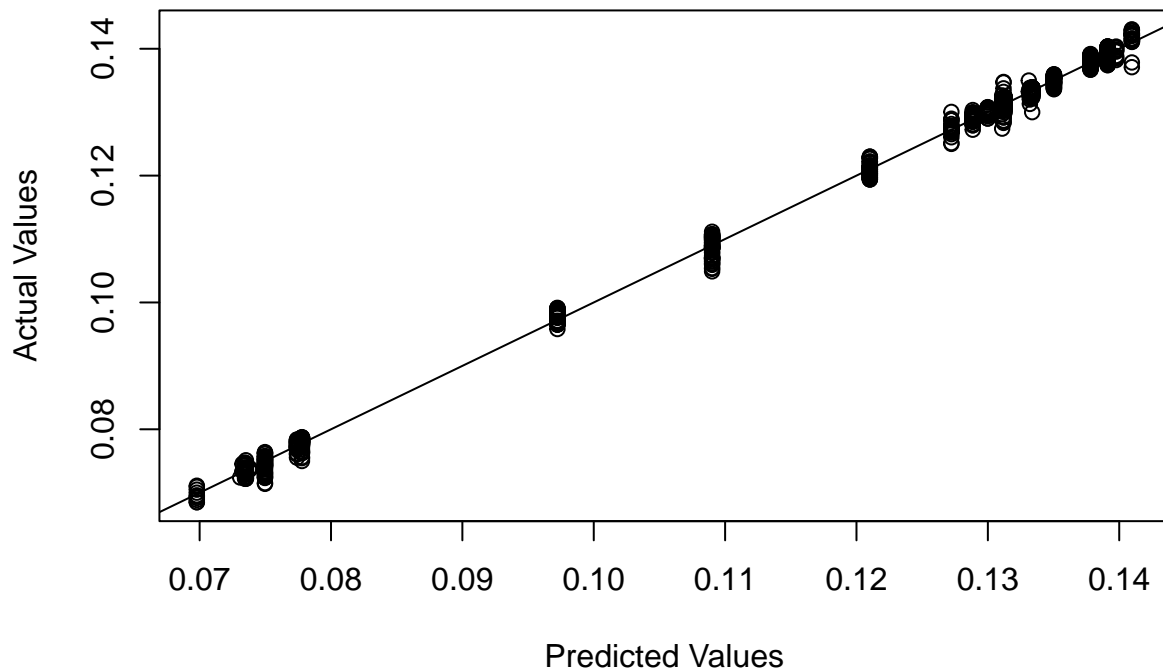
```
## [1] 0.000940249
```

```
varImpPlot(rf)
```



```
plot(x=pred_values2, y=test12$yield,
     xlab='Predicted Values',
     ylab='Actual Values',
     main='Predicted vs. Actual Values')
abline(a=0, b=1)
```

## Predicted vs. Actual Values



```
library(gbm)
```

```
## Loaded gbm 2.1.8
```

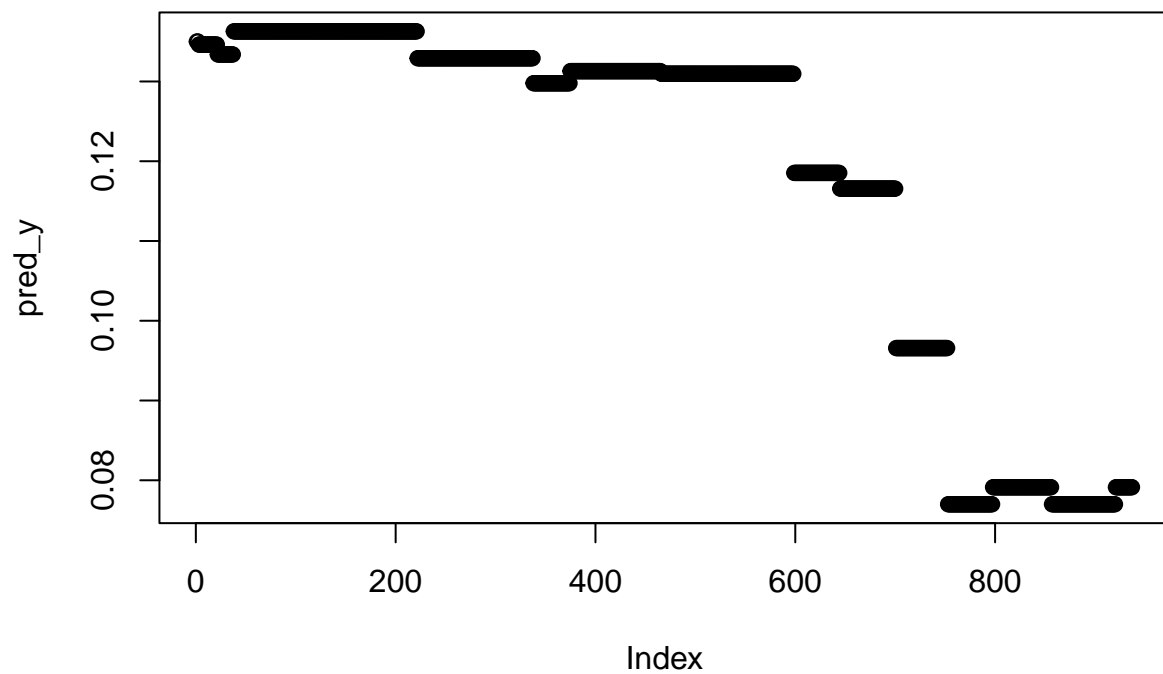
```
model_gbm <- gbm(yield ~ . , data = train12, distribution = "gaussian" , cv.folds = 10 , shrinkage = .01  
print(model_gbm)
```

```
## gbm(formula = yield ~ . , distribution = "gaussian", data = train12,  
##      n.trees = 500, n.minobsinnode = 10, shrinkage = 0.01, cv.folds = 10)  
## A gradient boosted model with gaussian loss function.  
## 500 iterations were performed.  
## The best cross-validation iteration was 500.  
## There were 6 predictors of which 5 had non-zero influence.
```

```
pred_y <- predict.gbm(model_gbm,test12)
```

```
## Using 500 trees...
```

```
plot(pred_y)
```



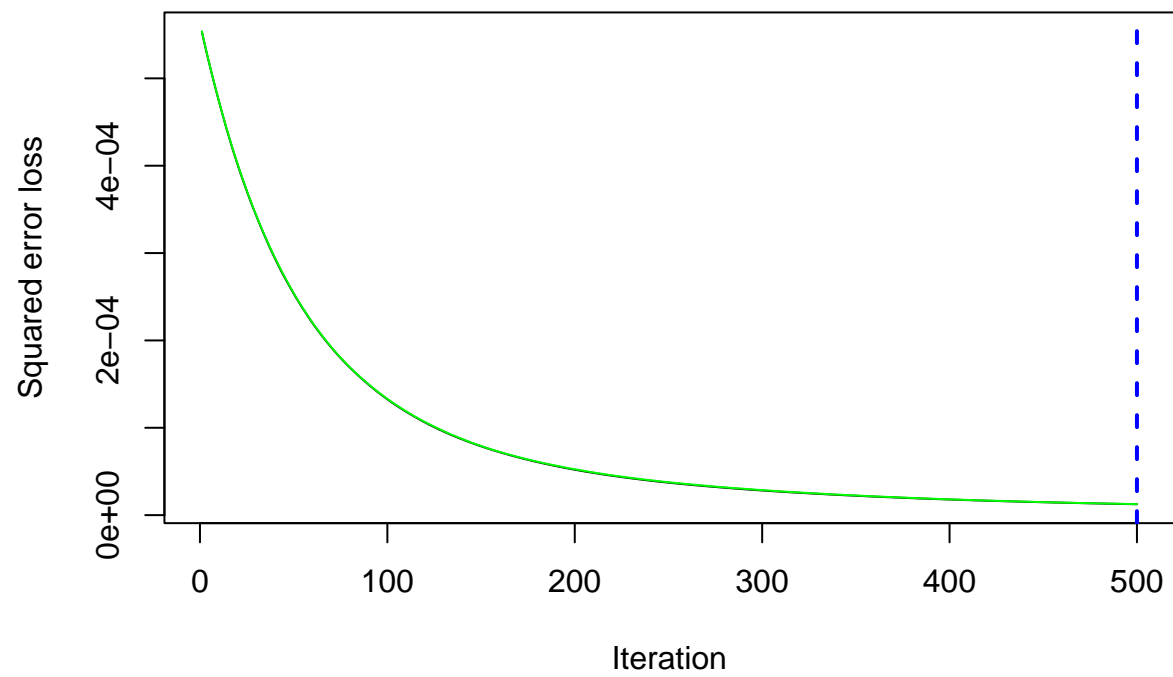
```
mse <- mse(test12$yield,pred_y)
print(mse)
```

```
## [1] 1.200455e-05
```

```
sqrt(mse)
```

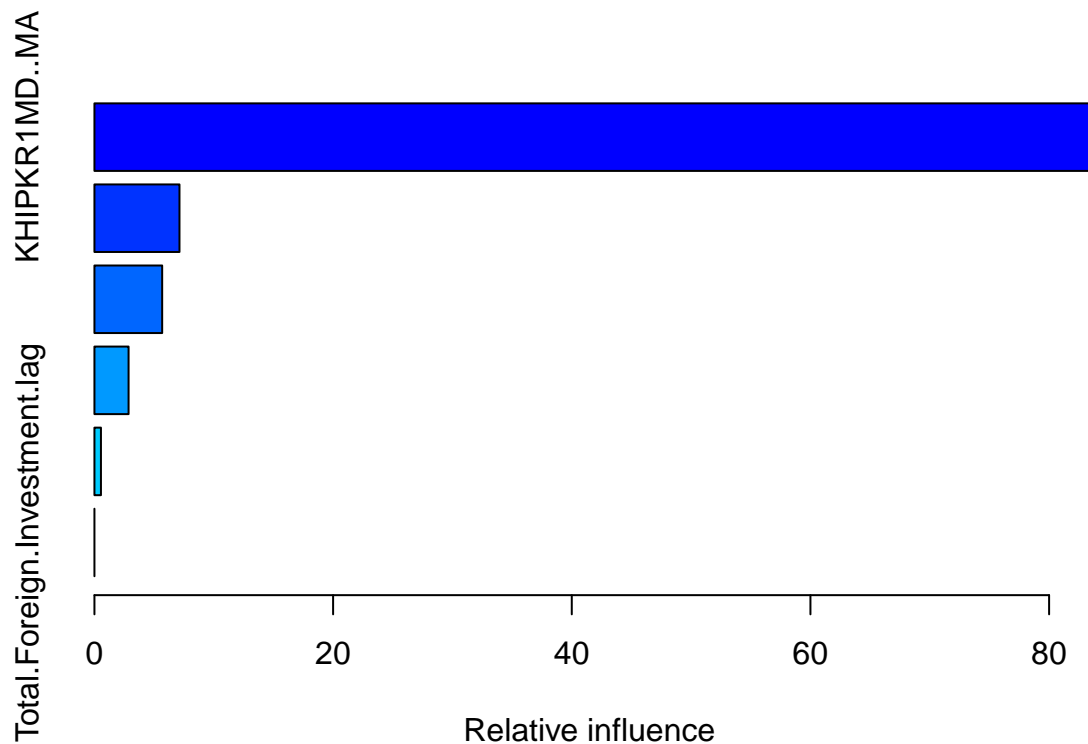
```
## [1] 0.003464758
```

```
gbm.perf(model_gbm,method="cv")
```



```
## [1] 500
```

```
summary(model_gbm)
```

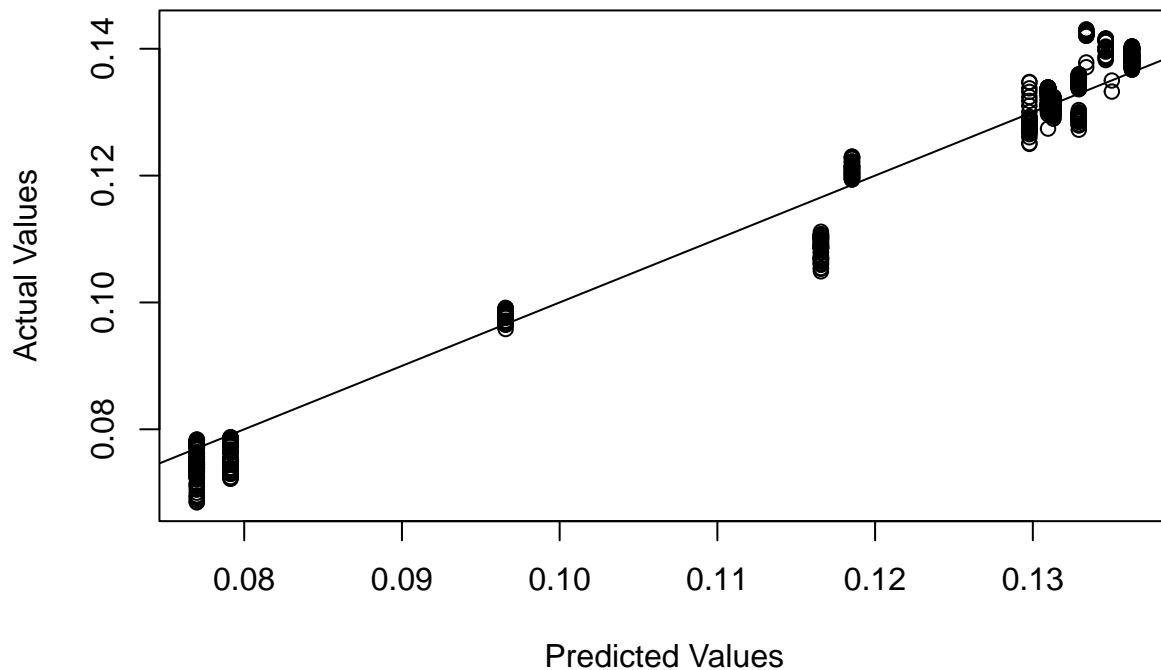


```
##                                var      rel.inf
## KHIPKR1MD..MA                KHIPKR1MD..MA 83.7917212
## PKR.PER.USD.day.lag          PKR.PER.USD.day.lag 7.1264160
## NET.RESERVES.WITH.SBP.lag    NET.RESERVES.WITH.SBP.lag 5.6760278
## Trade.Deficit.lag           Trade.Deficit.lag 2.8607267
## Workers..Remittances.lag     Workers..Remittances.lag 0.5451083
## Total.Foreign.Investment.lag Total.Foreign.Investment.lag 0.0000000
```

```
plot(x=pred_y, y=test12$yield,
     xlab='Predicted Values',
     ylab='Actual Values',
     main='Predicted vs. Actual Values')
abline(a=0, b=1)
```



## Predicted vs. Actual Values



```
train12[-1]<-scale(train12[-1])
test12[-1]<-scale(test12[-1])
library(e1071)
```

```
## Warning: package 'e1071' was built under R version 4.1.1
```

```
svmreg = svm(formula = yield~ .,
data = train12)
svmreg
```

```
##
## Call:
## svm(formula = yield ~ ., data = train12)
##
##
## Parameters:
##   SVM-Type:  eps-regression
##   SVM-Kernel: radial
##     cost:    1
##    gamma:   0.1666667
##   epsilon:  0.1
##
##
## Number of Support Vectors: 192
```

```
library(Metrics)
y_pred = predict(svmreg, newdata = test12[,2:7])
rmsesvm <- rmse(test12$yield,y_pred)
rmsesvm
```

```
## [1] 0.001630708
```

```
plot(x=y_pred, y=test12$yield,
     xlab='Predicted Values',
     ylab='Actual Values',
     main='Predicted vs. Actual Values')
abline(a=0, b=1)
```

