

report_3

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```
library(readxl)
library(MASS)
library(caret)
library(rpart)
library(rpart.plot)
library(randomForest)
library(glmnet)
library(xgboost)
```

Cleaning the Data

```
df <- read_xls('economy.xls', sheet='2011-2019 NACE 2')
df <- df[c(4,6,11,30,34),]

df <- df[,-c(1,3)]

my_months <- c('JAN','FEB','MARCH','APRIL','MAY','JUNE','JULY','AUG','SEP','OCT','NOV','DEC')
my_years <- c(2011:2019)
df_months = c()
j = 0
for (i in 2:length(df)) {

  if((i-2)%12==0) {
    j = j + 1
    df_months = c(df_months, paste0(my_years[j], '_', my_months))
  }
}

df_months = head(df_months,-6)

df <- as.data.frame(t(df))
cols <- df[1,]
cols = as.character(unlist(cols))
df <- df[-c(1),]
colnames(df) <- as.character(cols)
rownames(df) <- df_months
df[] <- sapply(df[],function(x) as.numeric(as.character(x)))

df$`Total industry` = df$`Total industry` * 1000000
df$`mining and quarrying` = df$`mining and quarrying` * 1000000
df$manufacturing = df$manufacturing * 1000000
df$`Electricity, gas, steam and air conditioning supply` = df$`Electricity, gas, steam and air conditioning supply`
df$`Water supply, sewerage, waste management and remediation activities` = df$`Water supply, sewerage, waste management and remediation activities`
```

T-1 = T

```
total = df$`Total industry`  
names = rownames(df)  
df = df[-102,]  
total = total[-1]  
names = names[-1]  
rownames(df) = names  
df$`Total industry`=total
```

```
head(df)
```

```
##          Total industry mining and quarrying manufacturing  
## 2011_FEB      68950700000      13164400000      32688300000  
## 2011_MARCH     76904600000      12337100000      38793800000  
## 2011_APRIL     75019400000      13334900000      48051400000  
## 2011_MAY       82815300000      15377500000      46263600000  
## 2011_JUNE      87199000000      14881400000      53648900000  
## 2011_JULY      78206100000      14519800000      58033900000  
##          Electricity, gas, steam and air conditioning supply  
## 2011_FEB                      16059900000  
## 2011_MARCH                     16327300000  
## 2011_APRIL                     14049200000  
## 2011_MAY                       11845900000  
## 2011_JUNE                      12772400000  
## 2011_JULY                      13069300000  
##          Water supply, sewerage, waste management and remediation activites  
## 2011_FEB                      1404200000  
## 2011_MARCH                     1492500000  
## 2011_APRIL                     1469100000  
## 2011_MAY                       1532400000  
## 2011_JUNE                      1512600000  
## 2011_JULY                      1575900000
```

Splitting the data into *train.85* and *test.15*

```
index <- sample(1:nrow(df),round(0.85*nrow(df)))  
  
train <- df[index,]  
test <- df[-index,]  
  
n <- names(train)
```

```
df$`Total industry`
```

```
## [1] 68950700000 76904600000 75019400000 82815300000 87199000000  
## [6] 78206100000 82340200000 91564700000 89221300000 93969200000  
## [11] 102601000000 74121200000 80296600000 85180900000 83637700000  
## [16] 95768500000 94252400000 95346800000 93504300000 94322000000  
## [21] 98881600000 107133800000 118030000000 95473900000 93167800000
```

```
## [26] 99722300000 85554000000 90153900000 103338200000 95634300000
## [31] 101106900000 110464000000 118994600000 116280000000 130688800000
## [36] 87916800000 87808200000 98631100000 95404300000 101988000000
## [41] 107586700000 114318300000 105561000000 119757000000 121128900000
## [46] 120359200000 127696200000 88660200000 99366900000 106852900000
## [51] 104876300000 105518700000 114280400000 110330100000 113534000000
## [56] 118789900000 117244200000 117806100000 129815800000 94430400000
## [61] 99675100000 111202400000 115680100000 112485600000 126229400000
## [66] 117745900000 121268100000 129288300000 124218000000 133420400000
## [71] 143400300000 110043500000 124444200000 139515900000 132504200000
## [76] 145671500000 140072100000 144456900000 147917100000 161314700000
## [81] 179763200000 186876600000 223798900000 130511600000 131634500000
## [86] 153538600000 133147900000 143461700000 152379300000 155670300000
## [91] 154683600000 163500600000 194080500000 189553000000 233100000000
## [96] 124934500000 136840600000 160751900000 151122400000 159462400000
## [101] 167388400000
```

```
f <- as.formula(`Total industry` ~ `mining and quarrying` + manufacturing + `Electricity, gas, steam and
```

Prediction Models

Linear Regression

```
attach(df)
fit <- lm(`Total industry` ~ ., train)

pred1 <- predict(fit, newdata = test)

RMSE1 <- RMSE(test$`Total industry`, pred1)

MAE1 <- MAE(test$`Total industry`, pred1)
```

```
## [1] "RMSE for Linear Regression: 9225345439.22755"
```

```
## [1] "MAE for Linear Regression: 6717603196.47461"
```

K- Nearest Neighbors

```
ctrl <- trainControl(method = "cv", number = 5)

knn_c <- train(f, data = train, method = "knn",
              trControl = ctrl, preProcess = c("center", "scale"), tuneLength = 5)
```

```
model_predict_test = predict(knn_c, newdata = test)
# which.min(c(sqrt(mean(abs(model_predict_test - test$main)^2)),sqrt(mean((test$main- predict(fit, test.

RMSE_KNN <- RMSE(test$`Total industry`, model_predict_test)

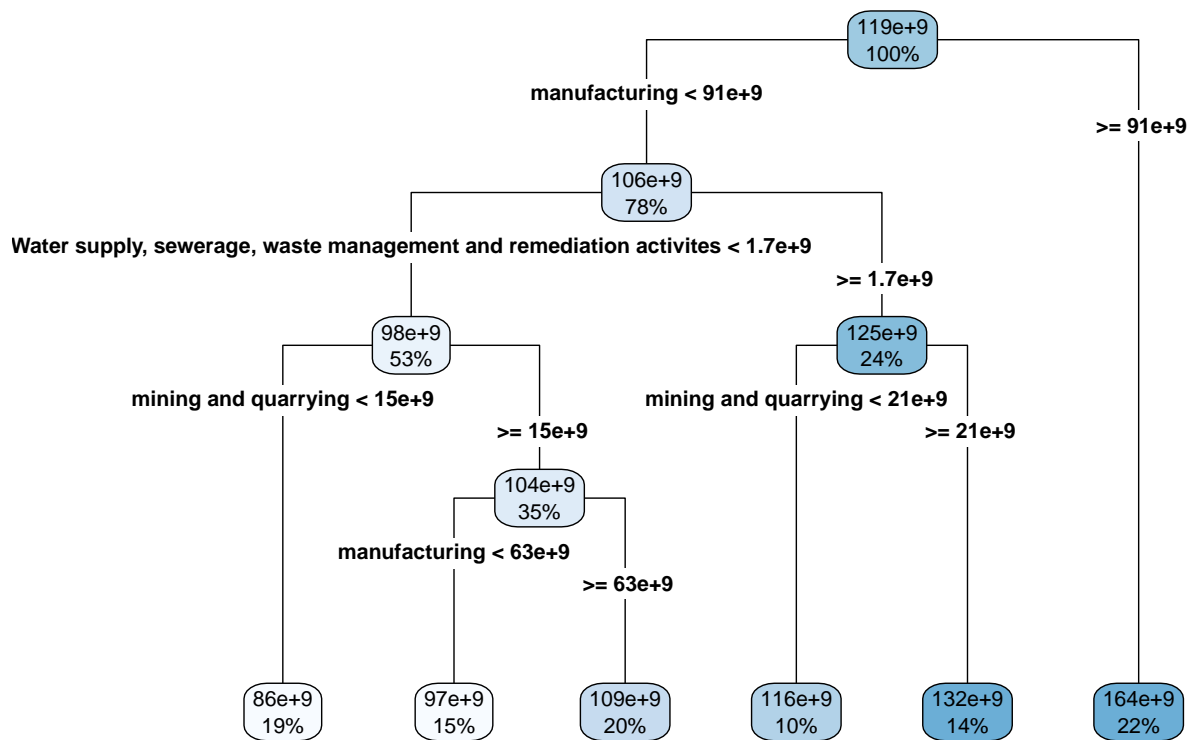
MAE_KNN <- MAE(test$`Total industry`, model_predict_test)
```

```
## [1] "RMSE for KNN: 8790092379.18225"
```

```
## [1] "MAE for KNN: 7417820952.38095"
```

Tree

```
my_model <- rpart(f,subset = index, data= df)
```



```
predictions<-predict(my_model,newdata=test)
```

```
RMSE_Tree <- RMSE(predictions, test$`Total industry`)
```

```
MAE_Tree <- MAE(predictions, test$`Total industry`)
```

```
## [1] "RMSE for Tree: 14055441347.3879"
```

```
## [1] "MAE for Tree: 9145335838.36098"
```

forrest

```
# set.seed(1)
# bag.black <- randomForest(f,data=df, subset=index,importance =TRUE)
#
# prediction_forest = predict(bag.black, newdata=test[1,])
```

```
# getTree(bag.black,1,labelVar=TRUE)
```

```
# MAE_forest <- MAE(prediction_forest, test$main)
# RMSE_forest <- RMSE(prediction_forest, test$main)
```

Ridge Regression

```
x = model.matrix(f, data = df)[,c(-1)]
y = df$`Total industry`

grid = 10^seq(10,-2, length = 100)

ridge.mod = glmnet(x[index,], y[index], alpha = 0, lambda = grid,
                   thresh = 1e-12)
ridge.pred = predict(ridge.mod, s = 4, newx = x[-index,])

RMSE_ridge <- RMSE(ridge.pred, y[-index])

MAE_ridge <- MAE(ridge.pred, y[-index])
```

```
## [1] "RMSE for Ridge Regression: 9225345436.8015"
```

```
## [1] "MAE for Ridge Regression: 6717603194.20622"
```

Lasso Regression

```
set.seed(1)

lasso.mod=glmnet(x[index,],y[index],alpha=1,lambda=grid)
cv.out=cv.glmnet(x[index,],y[index],alpha=1)

bestlam=cv.out$lambda.min

lasso.pred=predict(lasso.mod,s=bestlam,newx=x[-index,])
RMSE_lasso <- RMSE(lasso.pred, y[-index])

MAE_lasso <- MAE(lasso.pred, y[-index])
```

```
## [1] "RMSE for Lasso Regression: 9316509959.38094"
```

```
## [1] "MAE for Lasso Regression: 6416709007.17746"
```

Extreme Gradient Boosting

```
set.seed(1)
dtrain2 <- xgb.DMatrix(data = x[index,], label = y[index])
dtest2 <- xgb.DMatrix(data = x[-index,], label = y[-index])
watchlist <- list(train= dtrain2, test= dtest2)
set.seed(1)
bst2 <- xgb.train(data= dtrain2, max.depth=10, eta=0.09, nrounds=120, watchlist=watchlist,
                  base_score = 0.1)

# xgb_test <- predict(bst2, data.matrix(test[, -c(1)]))

# RMSE_xgboost <- RMSE(test$main, xgb_test)
#
# MAE_xgboost <- MAE(test$main, xgb_test)
```

```
## [1] 3
```

```
## [1] "the Algorithm with the least error is: 9225345436.8015"
```

```
options(scipen = 999)
## Ridge Regression
mean_error(difference(ridge.pred, test$`Total industry`))
```

```
## [1] 0.06130604
```

```
# Linear Regression
mean_error(difference(pred1, test$`Total industry`))
```

```
## [1] 0.06130604
```