```
title: "Qualifying 6"
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date: "December 22, 2014"
output: pdf document
1). Data exploration using visualizations
```{r,echo=FALSE}
rm(list=ls())
data = read.table("http://archive.ics.uci.edu/ml/machine-learning-
databases/adult/adult.data",
 sep=",",header=F,col.names=c("age", "type employer", "fnlwgt",
"education",
 "education num", "marital", "occupation", "relationship",
"race", "sex",
 "capital gain", "capital loss", "hr per week", "country",
"income"),
 fill=FALSE,strip.white=T)
testdata = read.table("http://archive.ics.uci.edu/ml/machine-learning-
databases/adult/adult.data",
 sep=",",header=F,col.names=c("age", "type employer", "fnlwgt",
"education",
 "education num", "marital", "occupation", "relationship",
"race", "sex",
 "capital gain", "capital loss", "hr per week", "country",
"income"),
 fill=FALSE,strip.white=T)
data[["education num"]]=NULL
data[["fnlwqt"]]=NULL
data$type_employer = as.character(data$type employer)
data$occupation = as.character(data$occupation)
data$country = as.character(data$country)
data$education = as.character(data$education)
data$race = as.character(data$race)
data$marital = as.character(data$marital)
data$type employer = gsub("^Federal-gov", "Federal-
Govt",data$type employer)
data$type employer = gsub("^Local-gov", "Other-Govt", data$type employer)
data$type employer = gsub("^State-gov","Other-Govt",data$type employer)
data$type_employer = gsub("^Private","Private",data$type_employer)
data$type employer = gsub("^Self-emp-inc", "Self-
Employed",data$type employer)
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data$type_employer = gsub("^Self-emp-not-inc","Self-
Employed",data$type_employer)
data$type_employer = gsub("^Without-pay","Not-
Working",data$type_employer)
data$type_employer = gsub("^Never-worked","Not-
Working",data$type_employer)
data$occupation = gsub("^Adm-clerical", "Admin", data$occupation)
data$occupation = gsub("^Armed-Forces", "Military", data$occupation)
data$occupation = gsub("^Craft-repair", "Blue-Collar", data$occupation)
data$occupation = gsub("^Exec-managerial","White-
Collar", data (occupation)
data$occupation = gsub("^Farming-fishing","Blue-Collar",data$occupation)
data$occupation = gsub("^Handlers-cleaners", "Blue-
Collar", data$occupation)
data$occupation = gsub("^Machine-op-inspct", "Blue-
Collar",data$occupation)
data$occupation = gsub("^0ther-service","Service",data$occupation)
data$occupation = gsub("^Priv-house-serv", "Service", data$occupation)
data$occupation = gsub("^Prof-specialty", "Professional", data$occupation)
data$occupation = gsub("^Protective-serv","Other-
Occupations", data (occupation)
data$occupation = gsub("^Sales", "Sales", data$occupation)
data$occupation = gsub("^Tech-support","Other-
Occupations", data (occupation)
data$occupation = gsub("^Transport-moving","Blue-
Collar",data$occupation)
data$country[data$country=="Cambodia"] = "SE-Asia"
data$country[data$country=="Canada"] = "British-Commonwealth"
data$country[data$country=="China"] = "China"
data$country[data$country=="Columbia"] = "South-America"
data$country[data$country=="Cuba"] = "Other"
data$country[data$country=="Dominican-Republic"] = "Latin-America"
data$country[data$country=="Ecuador"] = "South-America"
data$country[data$country=="El-Salvador"] = "South-America"
data$country[data$country=="England"] = "British-Commonwealth"
data$country[data$country=="France"] = "Euro_1"
data$country[data$country=="Germany"] = "Euro_1"
data$country[data$country=="Greece"] = "Euro_2"
data$country[data$country=="Guatemala"] = "Latin-America"
data$country[data$country=="Haiti"] = "Latin-America"
data$country=="Holand-Netherlands"] = "Euro_1"
data$country[data$country=="Honduras"] = "Latin-America"
data$country[data$country=="Hong"] = "China"
data$country[data$country=="Hungary"] = "Euro_2"
data$country[data$country=="India"] = "British-Commonwealth"
data$country[data$country=="Iran"] = "Other"
data$country[data$country=="Ireland"] = "British-Commonwealth"
data$country[data$country=="Italy"] = "Euro_1"
```

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data$country[data$country=="Jamaica"] = "Latin-America"
data$country[data$country=="Japan"] = "Other"
data$country[data$country=="Laos"] = "SE-Asia"
data$country[data$country=="Mexico"] = "Latin-America"
data$country[data$country=="Nicaragua"] = "Latin-America"
data$country[data$country=="Outlying-US(Guam-USVI-etc)"] = "Latin-
America"
data$country[data$country=="Peru"] = "South-America"
data$country[data$country=="Philippines"] = "SE-Asia"
data$country[data$country=="Poland"] = "Euro_2"
data$country[data$country=="Portugal"] = "Euro_2"
data$country[data$country=="Puerto-Rico"] = "Latin-America"
data$country[data$country=="Scotland"] = "British-Commonwealth"
data$country[data$country=="South"] = "Euro_2"
data$country[data$country=="Taiwan"] = "China"
data$country[data$country=="Thailand"] = "SE-Asia"
data$country[data$country=="Trinadad&Tobago"] = "Latin-America"
data$country[data$country=="United-States"] = "United-States"
data$country[data$country=="Vietnam"] = "SE-Asia"
data$country[data$country=="Yugoslavia"] = "Euro_2"
data$education = gsub("^10th", "Dropout", data$education)
data$education = gsub("^11th", "Dropout", data$education)
data$education = gsub("^12th", "Dropout", data$education)
data$education = gsub("^1st-4th","Dropout",data$education)
data$education = gsub("^5th-6th", "Dropout", data$education)
data$education = gsub("^7th-8th","Dropout",data$education)
data$education = gsub("^9th", "Dropout", data$education)
data$education = gsub("^Assoc-acdm", "Associates", data$education)
data$education = gsub("^Assoc-voc", "Associates", data$education)
data$education = gsub("^Bachelors", "Bachelors", data$education)
data$education = gsub("^Doctorate", "Doctorate", data$education)
data$education = gsub("^HS-Grad","HS-Graduate",data$education)
data$education = gsub("^Masters", "Masters", data$education)
data$education = gsub("^Preschool","Dropout",data$education)
data$education = gsub("^Prof-school", "Prof-School", data$education)
data$education = gsub("^Some-college","HS-Graduate",data$education)
data$marital[data$marital=="Never-married"] = "Never-Married"
data$marital[data$marital=="Married-AF-spouse"] = "Married"
data$marital[data$marital=="Married-civ-spouse"] = "Married"
data$marital[data$marital=="Married-spouse-absent"] = "Not-Married"
data$marital[data$marital=="Separated"] = "Not-Married"
data$marital[data$marital=="Divorced"] = "Not-Married"
data$marital[data$marital=="Widowed"] = "Widowed"
data$race[data$race=="White"] = "White"
data$race[data$race=="Black"] = "Black"
data$race[data$race=="Amer-Indian-Eskimo"] = "Amer-Indian"
data$race[data$race=="Asian-Pac-Islander"] = "Asian"
data$race[data$race=="Other"] = "Other"
```

```
data[["capital_gain"]] <- ordered(cut(data$capital_gain,c(-Inf, 0,</pre>
 median(data[["capital_gain"]][data[["capital_gain"]] >0]),
 Inf)),labels = c("None", "Low", "High"))
data[["capital_loss"]] <- ordered(cut(data$capital_loss,c(-Inf, 0,</pre>
 median(data[["capital_loss"]][data[["capital_loss"]] >0]),
 Inf)), labels = c("None", "Low", "High"))
is.na(data) = data=='?'
is.na(data) = data==' ?'
data = na.omit(data)
data$marital = factor(data$marital)
data$education = factor(data$education)
data$country = factor(data$country)
data$type_employer = factor(data$type_employer)
data$occupation = factor(data$occupation)
data$race = factor(data$race)
data$sex = factor(data$sex)
data$relationship = factor(data$relationship)
data$income = as.factor(ifelse(data$income==data$income[1],0,1))
data$age = scale(data$age)
data$hr_per_week = scale(data$hr_per_week)
dummy <-dummyVars(income~.-age -hr_per_week,data=data)</pre>
newdata<-as.data.frame(predict(dummy,data))</pre>
newdata<-cbind(newdata,data$age,data$hr_per_week)</pre>
```{r,echo=FALSE}
plot(as.factor(testdata$education),testdata$education_num,main="Education
 vs Education Num", xlab="Education level", ylab="Numecial value for
education")
newdata <-newdata[,-nearZeroVar(newdata)]</pre>
corrrelation <- cor(newdata)</pre>
newdata<-newdata[,-findCorrelation(corrrelation,cutoff=0.75)]</pre>
library(corrplot)
corrplot(corrrelation, order = "hclust")
2)Outliars
```{r,echo=FALSE}
boxplot(data$age,main="Box plot for age")
boxplot(data$hr_per_week,main="Box plot for hours per week")
```

```
3)skewness
```{r,echo=FALSE}
library(e1071)
#Capital gain
boxplot(testdata$capital_gain,main="Box plot for capital gain")
skewness(testdata$capital_gain)
hist(testdata$capital_gain, main="Histogram for Capital gain")
barplot(table(data$capital_gain),main="Capital Gain")
skewness(testdata$capital_loss)
hist(testdata$capital loss,main="Histogram for Capital Loss")
barplot(table(data$capital_loss), main="Capital Loss")
#Here I block capital gains and losses, rather than do a transformation.
Both variables are heavily skewed to the point that I think a numerical
transformation would not have been appropriate. So I choose to block
them into "None", "Low", and "High". For both variables, none means they
don't play the market. Low means they have some investments. High means
they have significant investments.
```{r,echo=FALSE}
par(mfrow=c(2,2))
barplot(table(data$type_employer), main="type_employer")
barplot(table(data$education), main="education")
barplot(table(data$marital),main="martial")
barplot(table(data$occupation), main="occupaion")
barplot(table(data$relationship), main="relationship")
dev.off()
par(mfrow=c(2,2))
barplot(table(data$race),main="race")
barplot(table(data$sex),main="sex")
barplot(table(data$capital_gain),main="capital gain")
barplot(table(data$capital_loss), main="capital loss")
dev.off()
par(mfrow=c(1,2))
barplot(table(data$country), main="country")
barplot(table(data$income), main="income")
data <-data[,-nearZeroVar(data)]</pre>
- - -
```{r,echo=FALSE}
dev.off()
testdata[ testdata == "?" ] = NA
image(is.na(testdata), main = "Missing Values", xlab = "Observation",
ylab = "Variable", xaxt = "n", yaxt = "n", bty = "n")
axis(1, seq(0, 1, length.out = nrow(testdata)), 1:nrow(testdata), col =
```

```
"white")
testdata[,15]
table(testdata$type employer)
table(testdata$education)
table(testdata$marital)
table(testdata$occupation)
table(testdata$relationship)
table(testdata$race)
table(testdata$sex)
table(testdata$country)
- - -
```{r,echo=FALSE}
#removed the missing data, still there are over 30000
7)
```{r,echo=FALSE}
# AUC of ROC will be used as a classification statistic.
8)
```{r,echo=FALSE}

barplot(table(data$income), main="Predictor frequency distribution")
#unbalnced
#describe the preprocessing
#13 predictors are left
\``{r,echo=FALSE}
dummy <-dummyVars(income~.-age -hr_per_week -capital gain -</pre>
capital_loss,data=data)
dummy_data<-predict(dummy,data)</pre>
dummy data<-as.data.frame(dummy data)</pre>
trainIndex <- createDataPartition(data$income, p = .75,list =</pre>
FALSE, times = 1)
y<-data$income
x<-data[,-10]
trainX<-x[trainIndex,]</pre>
trainY<-factor(y[trainIndex])</pre>
```

```
testX<- x[-trainIndex,]
testY<-factor(y[-trainIndex])</pre>
dummy <-dummyVars(income~.-age -hr_per_week -capital_gain -</pre>
capital loss,data=x)
dummy <-dummyVars(income~.,data=data)</pre>
dummy_data<-predict(dummy,data)</pre>
dummy_data<-as.data.frame(dummy_data)</pre>
xD<-dummy data[,-40]
xD<-xD[,-nearZeroVar(xD)]</pre>
trainXD<-xD[trainIndex,]</pre>
testXD<- xD[-trainIndex,]</pre>
Logistic Regression
```{r,echo=FALSE}
ctrl <- trainControl(method = "LGOCV",</pre>
                       summaryFunction = twoClassSummary,
                       classProbs = TRUE,
                       savePredictions = TRUE)
set.seed(476)
lrModel<- train(trainX,y = trainY,method = "qlm",metric =</pre>
"ROC", trControl = ctrl)
lrModel
#confusionMatrix(predict(lrModel,testX),testY,positive="1")
LDA
```{r,echo=FALSE}
ldaModel<- train(x = trainXD,y = as.factor(trainY),method = "lda",metric</pre>
= "ROC", trControl = ctrl)
#confusionMatrix(predict(ldaModel,testX),testY,positive ="1")
ldaModel
PLSDA
```{r,echo=FALSE}
library(pls)
plsFit <- train(x = trainXD,y = trainY,method = "pls",tuneGrid =</pre>
expand.grid(.ncomp = 1:10),metric = "ROC",trControl = ctrl)
plsFit
plsImpGrant <- varImp(plsFit, scale = FALSE)</pre>
plsImpGrant
Penalized Models - GLMnet
```{r,echo=FALSE}
qlmnGrid \leftarrow expand.grid(.alpha = c(0,.2, .6, .8),.lambda=c(.1, .2,
length = 20))
```

```
glmnTuned <- train(trainXD,y = trainY,method = "glmnet",tuneGrid =</pre>
glmnGrid,metric = "ROC",trControl = ctrl)
glmnTuned
Nearest Shrunken Centrods
```{r,echo=FALSE}
nscGrid <- data.frame(.threshold = 0:25)</pre>
set.seed(476)
nscTuned <- train(x = trainXD,y = trainY,method = "pam",preProc =</pre>
c("center", "scale"),tuneGrid = nscGrid,metric = "ROC",trControl = ctrl)
varImp(nscTuned, scale = FALSE)
11) Non Linear models
Non linear discriminant analysis
```{r,echo=FALSE}
set.seed(476)
mdaFit <- train(trainX, trainY, method = "mda", metric = "ROC", tuneGrid =</pre>
expand.grid(.subclasses = 1:8),trControl = ctrl)
mdaFit
NeuralNet
```{r,echo=FALSE}
nnetGrid \leftarrow expand.grid(.size = 1:5,.decay = c(0,0.001,0.1,1))
maxSize <- max(nnetGrid$.size)</pre>
numWts < -1*(maxSize * (9 + 1) + maxSize + 1)
set.seed(476)
nnetFit <- train(x = trainX,y = trainY,</pre>
                  method = "nnet",
                   metric = "ROC",
                  tuneGrid = nnetGrid,
                  trace = FALSE,
                  maxit = 10000,
                  MaxNWts = numWts,
                  ## ctrl was defined in the previous chapter
                   trControl = ctrl)
nnetFit
. . .
FDA
```

```
```{r,echo=FALSE}
fdaFit <- train(trainX, trainY,method = "fda",metric = "ROC",trControl =</pre>
fdaFit
plsImp <- varImp(fdaFit,scale=FALSE)</pre>
```{r,echo=FALSE}
library(e1071)
library(kernlab)
library(klaR)
set.seed(202)
sigmaRange <- sigest(as.matrix(trainXD),na.action = na.omit)</pre>
svmRGrid <- expand.grid(.sigma = sigmaRange[1],.C = 2^(seq(-2, 2)))</pre>
set.seed(476)
svmRModel <- train(trainXD, trainY,</pre>
                     method = "svmRadial",
                     metric = "ROC",
                     tuneGrid = svmRGrid,
                     fit = FALSE,
                     trControl = ctrl)
svmRModel
varImp(svmRModel, scale = FALSE)
KNN
```{r,echo=FALSE}
set.seed(476)
knnFit <- train(trainXD, trainY,</pre>
 method = "knn",
 metric = "ROC",
 tuneGrid = data.frame(.k = c(3,5,9)),
 trControl = ctrl)
knnFit
NaiveBayes
```{r,echo=FALSE}
library(klaR)
ctrl1 <- trainControl(</pre>
                      summaryFunction = twoClassSummary,
                      classProbs = TRUE,
                      savePredictions = TRUE)
nb<-naiveBayes(trainX, trainY)</pre>
NBFit <- train(trainX, trainY,method = "nb",metric = "ROC",trControl =</pre>
ctrl1)
confusionMatrix(predict(nb, testX),testY)
```