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# CART classification
library(rpart)
library(rpart.plot)
bank.df<-read.csv("UniversalBank.csv", header= T)</pre>
str(bank.df)
bank.df < bank.df[ , -c(1, 5)] # Drop ID and zip code columns.
# partition
set.seed(1)
train.index <- sample(c(1:dim(bank.df)[1]), dim(bank.df)[1]*0.6)</pre>
train.df <- bank.df[train.index, ]</pre>
valid.df <- bank.df[-train.index, ]</pre>
# classification tree
default.ct <- rpart(Personal.Loan ~ ., data = train.df, method = "class")</pre>
length(default.ct$frame$var[default.ct$frame$var == "<leaf>"])
# plot tree
prp(default.ct, type = 1, extra = 1, under = TRUE, split.font = 1, varlen = -10)
library(rattle)
fancyRpartPlot(default.ct)
# predict
predict(default.ct, valid.df)
# set argument type = "class" in predict() to generate predicted class membership.
default.ct.point.pred.train <- predict(default.ct,train.df,type = "class")</pre>
# generate confusion matrix for training data
library(caret)
library(ggplot2)
confusionMatrix(default.ct.point.pred.train, as.factor(train.df$Personal.Loan))
# generate confusion matrix for validation data
default.ct.point.pred.valid <- predict(default.ct,valid.df,type = "class")</pre>
confusionMatrix(default.ct.point.pred.valid, as.factor(valid.df$Personal.Loan))
### Tree Pruning
deeper.ct <- rpart(Personal.Loan ~ ., data = train.df, method = "class",</pre>
                   cp = 0, minsplit = 1)
# count number of leaves
length(deeper.ct$frame$var[deeper.ct$frame$var == "<leaf>"])
# plot tree
prp(deeper.ct, type = 1, extra = 1, under = TRUE, split.font = 1, varlen = -10,
    box.col=ifelse(deeper.ct$frame$var == "<leaf>", 'gray', 'white'))
# argument xval refers to the number of folds to use in rpart's built-in
# cross-validation procedure
# argument cp sets the smallest value for the complexity parameter.
cv.ct <- rpart(Personal.Loan ~ ., data = train.df, method = "class",</pre>
               cp = 0.00001, minsplit = 5, xval = 5)
# use printcp() to print the table.
printcp(cv.ct)
# prune by lower cp
pruned.ct \leftarrow prune(cv.ct, cp = 0.0169697)
length(pruned.ct$frame$var[pruned.ct$frame$var == "<leaf>"])
prp(pruned.ct, type = 1, extra = 1, split.font = 1, varlen = -10)
fancyRpartPlot(pruned.ct)
```

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# CART regression
car.df <- read.csv("ToyotaCorolla.csv")</pre>
#car.df<-read.csv(file.choose(), header= T)</pre>
str(car.df)
# preprocess
set.seed(1)
train.index \leftarrow sample(c(1:dim(car.df)[1]), dim(car.df)[1]*0.6)
valid.index <- setdiff(c(1:dim(car.df)[1]), train.index)</pre>
train.df <- car.df[train.index, ]</pre>
valid.df <- car.df[valid.index, ]</pre>
# regression tree:
tr <- rpart(Price ~ Age 08 04 + KM + Fuel Type +
               HP + Automatic + Doors + Quarterly Tax +
              Mfr_Guarantee + Guarantee_Period + Airco +
               Automatic airco + CD Player + Powered Windows +
               Sport Model + Tow Bar, data = train.df,
            method = "anova", minbucket = 1, maxdepth = 30, cp = 0.001)
prp(tr)
# errors
library(forecast)
library(ggplot2)
accuracy(predict(tr, train.df), train.df$Price)
accuracy(predict(tr, valid.df), valid.df$Price)
# shallower tree
tr.shallow <- rpart(Price ~ Age 08 04 + KM + Fuel Type +
                       HP + Automatic + Doors + Quarterly_Tax +
                       Mfr Guarantee + Guarantee Period + Airco +
                       Automatic_airco + CD_Player + Powered_Windows +
                       Sport Model + Tow Bar, data = train.df,
                     method = "anova")
prp(tr.shallow)
accuracy(predict(tr.shallow, train.df), train.df$Price)
accuracy(predict(tr.shallow, valid.df), valid.df$Price)
#Classification Tree
#Model for categorical price
bins <- seq(min(car.df$Price),</pre>
            max(car.df$Price),
             (max(car.df$Price) - min(car.df$Price))/20)
Binned Price <- .bincode(car.df$Price,</pre>
                          bins,
                          include.lowest = TRUE)
Binned Price <- as.factor(Binned Price)</pre>
train.df$Binned Price <- Binned Price[train.index]</pre>
valid.df$Binned Price <- Binned Price[valid.index]</pre>
tr.binned <- rpart(Binned_Price ~ Age_08_04 + KM + Fuel_Type +
                      HP + Automatic + Doors + Quarterly Tax +
                      Mfr Guarantee + Guarantee Period + Airco +
                      Automatic airco + CD Player + Powered Windows +
                      Sport Model + Tow Bar, data = train.df)
prp(tr.binned)
# predict price
new.record <- data.frame(Age 08 04 = 77, KM = 117000, Fuel Type = "Petrol", HP =
110, Automatic = 0, Doors = 5, Quarterly Tax = 100, Mfr Guarantee = 0, Guarantee Period =
3, Airco = 1, Automatic airco = 0, CD Player = 0, Powered Windows = 0, Sport Model =
0, Tow Bar = 1)
# regression model
price.tr <- predict(tr, newdata = new.record)</pre>
# classification model
price.tr.bin <- bins[predict(tr.binned, newdata = new.record, type = "class")]</pre>
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