Chapter 12: Use Case Notebook for Instructors

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Load packages	
library(bbmle)	

Use Case: Profit Forecasting using Linear Regression

Forecasting profits, revenues, debt levels, and a whole host of other metrics and analytics is crucial in business. Linear regression provides the most widely used tool kit for forecasting. The dataset gives us a snapshot of annual data for a range of startup companies: research and development spending (R&D Spend), Administration expenses, Marketing Spend, tells us the State the startup is based in, and the Profit generated:

```
df = read.csv("../../data/50_Startups.csv")
startups = sample(1:nrow(df),3)
head(df)
```

```
##
     R.D.Spend Administration Marketing.Spend
                                                      State Profit
## 1
        165349
                        136898
                                         471784
                                                   New York 192262
## 2
        162598
                        151378
                                         443899 California 191792
## 3
                                         407935
        153442
                        101146
                                                   Florida 191050
## 4
        144372
                        118672
                                         383200
                                                   New York 182902
## 5
        142107
                         91392
                                         366168
                                                   Florida 166188
## 6
        131877
                         99815
                                         362861
                                                   New York 156991
```

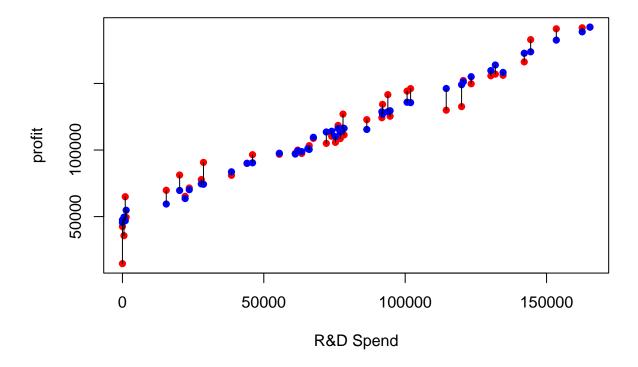
New Startups

We will randomly select three new startups we want to assess from the dataframe. We will build the regression model based on 47 startups (we will term it training data) and then test how well the model predicts for the three new startups (we will term it the test data):

```
df_test = df[startups,]
df_train =df[-startups,]
```

Building the Model

```
reg1 = lm(Profit ~ .-State, data=df_train)
summary(reg1)
##
## lm(formula = Profit ~ . - State, data = df_train)
##
## Residuals:
     Min
            10 Median
                          3Q
                               Max
## -32765 -4413
                  -76
                        6699
                            18034
## Coefficients:
                   Estimate Std. Error t value
                                                        Pr(>|t|)
                                                     0.0000000023 ***
## (Intercept)
                 51352.6134 6832.4708
                                        7.52
                                        17.09 < 0.000000000000000 ***
## R.D.Spend
                     0.8179
                               0.0479
## Administration
                    -0.0429
                               0.0541
                                        -0.79
                                                            0.43
## Marketing.Spend
                     0.0246
                               0.0171
                                        1.44
                                                            0.16
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 9410 on 43 degrees of freedom
## Multiple R-squared: 0.951, Adjusted R-squared: 0.948
plot(df_train$R.D.Spend,df_train$Profit,col="red",pch=16,xlab="R&D Spend",ylab="profit")
points(df_train$R.D.Spend,reg1$fitted.values,col="blue",pch=16)
segments(df_train$R.D.Spend,df_train$Profit,df_train$R.D.Spend,reg1$fitted.values)
```

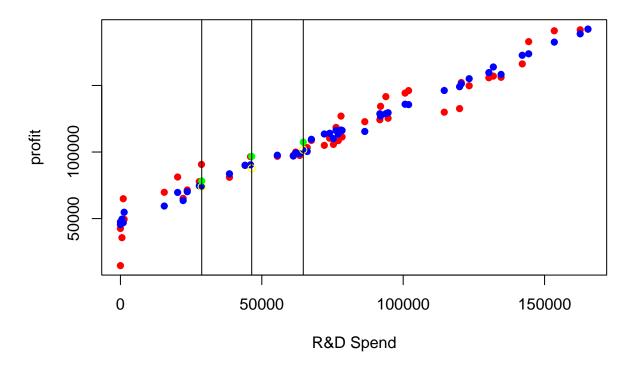


Forecast Startup Profits

```
df_test$prediction = predict.lm(reg1,df_test)
df_test
```

```
R.D.Spend Administration Marketing.Spend
##
                                                     State Profit prediction
## 35
          46426
                        157694
                                         210798 California 96713
                                                                        87746
          64665
                        139553
## 26
                                         137963 California 107404
                                                                       101648
## 41
          28754
                        118546
                                         172796 California 78240
                                                                        74037
```

```
plot(df_train$R.D.Spend,df_train$Profit,col="red",pch=16,xlab="R&D Spend",ylab="profit")
points(df_train$R.D.Spend,reg1$fitted.values,col="blue",pch=16)
#segments(df_train$R.D.Spend,df_train$Profit,df_train$R.D.Spend,reg1$fitted.values)
points(df_test$R.D.Spend,df_test$prediction,col="yellow",pch=)
points(df_test$R.D.Spend,df_test$Profit,col="green",pch=16)
#segments(df_test$R.D.Spend,df_test$Profit,df_test$R.D.Spend,df_test$prediction,lwd = 5)
abline(v = df_test$R.D.Spend)
```



Dealing with Categorical Data

```
library(caret)

## Loading required package: ggplot2

## Loading required package: lattice

dummy = dummyVars(" ~ .", data=df)
final_df = data.frame(predict(dummy, newdata=df))
head(final_df)
```

```
##
      {\tt R.D.Spend} \ \ {\tt Administration} \ \ {\tt Marketing.Spend} \ \ {\tt StateCalifornia} \ \ {\tt StateFlorida}
## 1
         165349
                           136898
                                               471784
                                                                                        0
                                                                        0
## 2
         162598
                           151378
                                               443899
                                                                        1
                                                                                        0
                                                                        0
         153442
                                               407935
                                                                                        1
## 3
                           101146
## 4
         144372
                                               383200
                                                                        0
                                                                                        0
                            118672
## 5
         142107
                             91392
                                               366168
                                                                        0
                                                                                        1
## 6
         131877
                             99815
                                               362861
      StateNew.York Profit
## 1
                    1 192262
## 2
                    0 191792
## 3
                    0 191050
## 4
                    1 182902
## 5
                    0 166188
## 6
                    1 156991
```