Instituto Tecnológico de Tijuana

Nombre de Facultad

Ingeniería Informática

Proyecto / Tarea / Practica:

Practica 2 Unidad 3

Materia:

Minería de datos

Facilitador:

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Alumnos:

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Fecha:

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Código

```
1. # Importing the dataset
2. dataset <- read.csv('50_Startups.csv')
3.
4. # Encoding categorical data
5. dataset$State = factor(dataset$State,
6.                        levels = c('New
   York', 'California', 'Florida'),
7.                        labels = c(1,2,3))
8.
9. dataset
10.     # Splitting the dataset into the Training set and
   Test set
11.     # Install.packages('caTools')
12.     library(caTools)
13.     set.seed(123)
14.     split <- sample.split(dataset$Profit,
   SplitRatio = 0.8)
15.     training_set <- subset(dataset, split == TRUE)
16.     test_set <- subset(dataset, split == FALSE)
17.
18.     # Fitting Multiple Linear Regression to the Training
   set
19.     #regressor = lm(formula = Profit ~ R.D.Spend +
   Administration + Marketing.Spend + State)
20.     regressor = lm(formula = Profit ~ .,
21.                    data = training_set )
22.
23.     summary(regressor)
24.
25.     # Prediction the Test set results
26.     y_pred = predict(regressor, newdata = test_set)
27.     y_pred
28.
29.     # Assigment: visualize the siple liner regression
   model with R.D.Spend
30.
31.     # Building the optimal model using Backward
   Elimination
32.     regressor = lm(formula = Profit ~
   R.D.Spend + Administration + Marketing.Spend + State,
33.                    data = dataset )
34.     summary(regressor)
35.
36.     regressor = lm(formula = Profit ~
   R.D.Spend + Administration + Marketing.Spend,
37.                    data = dataset )
38.     summary(regressor)
39.
40.     regressor = lm(formula = Profit ~
   R.D.Spend + Marketing.Spend,
41.                    data = dataset )
42.     summary(regressor)
43.
```

```
44.     regressor = lm(formula = Profit ~
  R.D.Spend + Marketing.Spend,
45.                    data = dataset )
46.     summary(regressor)
47.
48.     y_pred = predict(regressor, newdata = test_set)
49.     y_pred
50.
51.     # Homework analise the follow atomation
  backwardElimination function
52.     backwardElimination <- function(x, sl) {
53.       numVars = length(x)
54.       for (i in c(1:numVars)){
55.         regressor = lm(formula = Profit ~ ., data = x)
56.         maxVar = max(coef(summary(regressor))[c(2:numVars
  ), "Pr(>|t|)"])
57.         if (maxVar > sl){
58.           j = which(coef(summary(regressor))[c(2:numVars)
  , "Pr(>|t|)"] == maxVar)
59.           x = x[, -j]
60.         }
61.         numVars = numVars - 1
62.       }
63.       return(summary(regressor))
64.     }
65.
66.     SL = 0.05
67.     #dataset = dataset[, c(1,2,3,4,5)]
68.     training_set
69.     backwardElimination(training_set, SL)
```

Explicación del "Automated Backward Elimination"

Código para automatizar la eliminación hacia atrás en p <.05 Las principales contribuciones sobre los métodos ya implementados en R se encuentran en el tratamiento de las interacciones. Primero eliminará todos los términos NS de mayor potencia (por ejemplo, con cuatro interacciones frente a tres). Además, mantiene NS términos de orden inferior de términos significativos de orden superior.