

## Logistic Regression–Carseats Sales.R

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```
#LOGISTIC REGRESSION for binary classification
library(ISLR)

## Warning: package 'ISLR' was built under R version 4.0.2

data("Carseats")
attach(Carseats)
str(Carseats)

## 'data.frame':  400 obs. of  11 variables:
## $ Sales      : num  9.5 11.22 10.06 7.4 4.15 ...
## $ CompPrice  : num  138 111 113 117 141 124 115 136 132 132 ...
## $ Income     : num   73 48 35 100 64 113 105 81 110 113 ...
## $ Advertising: num   11 16 10 4 3 13 0 15 0 0 ...
## $ Population : num  276 260 269 466 340 501 45 425 108 131 ...
## $ Price      : num  120 83 80 97 128 72 108 120 124 124 ...
## $ ShelfLoc   : Factor w/ 3 levels "Bad","Good","Medium": 1 2 3 3 1 1 3 2
## $ Age        : num   42 65 59 55 38 78 71 67 76 76 ...
## $ Education  : num   17 10 12 14 13 16 15 10 10 17 ...
## $ Urban      : Factor w/ 2 levels "No","Yes": 2 2 2 2 2 1 2 2 1 1 ...
## $ US         : Factor w/ 2 levels "No","Yes": 2 2 2 2 1 2 1 2 1 2 ...

set.seed(256)

#create new categorical variable
High <- as.factor(ifelse(Sales >= 8, "YES", "NO")) #categorical variable w/ 2
levels
Data <- data.frame(Carseats, High) #new df with High variable included
Data <- Data[, -1] #removes 1st column "Sales"
colnames(Data)[11] <- "Target" #change name to Last (11th) column to Target
head(Data)

##   CompPrice Income Advertising Population Price ShelfLoc Age Education Urban   US Target
## 1      138     73          11        276    120      Bad  42         17   Yes   Yes   YES
## 2      111     48          16        260     83     Good  65         10   Yes   Yes   YES
## 3      113     35          10        269     80   Medium  59         12   Yes   Yes   YES
## 4      117    100           4        466     97   Medium  55         14   Yes   Yes    NO
## 5      141     64           3        340    128     Bad  38         13   Yes   No    NO
## 6      124    113          13        501     72     Bad  78         16   No    Yes   YES

indx <- sample(2,nrow(Data), replace=T, prob = c(0.8, 0.2))
train <- Data[indx ==1, ]
test <- Data[indx ==2, ]
```

```

#glm - generalized linear model (~)
#glm(categorical target ~ inputs, data= train, family= "binomial")

logitModel <- glm(Target ~ . , data = train, family = "binomial")
summary(logitModel)

##
## Call:
## glm(formula = Target ~ . , family = "binomial", data = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.54086  -0.29416  -0.05406   0.16124   3.00877
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -5.042977   2.647769  -1.905   0.0568 .
## CompPrice     0.165231   0.025361   6.515 7.26e-11 ***
## Income        0.035677   0.009113   3.915 9.04e-05 ***
## Advertising   0.320174   0.059138   5.414 6.16e-08 ***
## Population    -0.002236   0.001649  -1.356   0.1751
## Price        -0.161611   0.021128  -7.649 2.02e-14 ***
## ShelveLocGood  7.827315   1.073606   7.291 3.08e-13 ***
## ShelveLocMedium 2.932128   0.693583   4.228 2.36e-05 ***
## Age          -0.076811   0.015677  -4.900 9.60e-07 ***
## Education    -0.015547   0.082481  -0.188   0.8505
## UrbanYes     -0.396721   0.481712  -0.824   0.4102
## USYes        -0.666006   0.652072  -1.021   0.3071
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 453.77  on 336  degrees of freedom
## Residual deviance: 149.73  on 325  degrees of freedom
## AIC: 173.73
##
## Number of Fisher Scoring iterations: 7

#Deviance: measure of goodness of fit of a glm : -2 Log (Likelihood)
#higher number - worse fit
#Null deviance: deviance of model with NO input variables, only intercept
#Residual deviance: deviance of full model.

predictions <- predict(logitModel, newdata = test)
#predicted log of odds

predictions <- predict(logitModel, newdata = test, type="response")
#***** probability of being in class YES

```

```
Class <- ifelse(predictions >= 0.5, "YES", "NO")
```

```
Class
```

```
##      6      19      20      31      33      41      42      43      45      49      54      61
63
## "YES" "YES" "YES" "YES" "NO" "NO" "NO" "YES" "NO" "NO" "NO" "YES"
"NO"
##      65      69      86      98     100     108     118     131     138     150     156     160
167
## "YES" "YES" "NO" "YES" "NO" "YES" "NO" "YES" "NO" "YES" "NO" "YES"
"NO"
##     168     172     187     189     191     199     204     211     223     238     245     247
250
## "NO" "YES" "YES" "NO" "YES" "NO" "NO" "NO" "NO" "NO" "YES" "YES"
"NO"
##     256     262     266     267     268     274     277     280     296     304     314     316
318
## "NO" "NO" "NO" "YES" "NO" "YES" "NO" "NO" "NO" "YES" "YES" "NO"
"NO"
##     333     339     351     352     356     371     375     383     384     385     388
## "NO" "NO" "YES" "YES" "NO" "NO" "NO" "NO" "YES" "YES" "YES"
```

```
test$Target == Class
```

```
##      6      19      20      31      33      41      42      43      45      49      54      61
63
## TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
TRUE
##      65      69      86      98     100     108     118     131     138     150     156     160
167
## FALSE TRUE FALSE FALSE  TRUE  TRUE FALSE  TRUE  TRUE  TRUE  TRUE  TRUE
TRUE
##     168     172     187     189     191     199     204     211     223     238     245     247
250
## TRUE  TRUE  TRUE FALSE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE FALSE  TRUE FALSE
TRUE
##     256     262     266     267     268     274     277     280     296     304     314     316
318
## TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
TRUE
##     333     339     351     352     356     371     375     383     384     385     388
## TRUE  TRUE  TRUE  TRUE  TRUE  TRUE FALSE  TRUE  TRUE  TRUE  TRUE
```