

Logistic_Regression

```
#Top Songs Analysis
```

```
#importing dataset top10s and copying it to test data
```

```
data <- read.csv("C:/Users/rmadh/OneDrive/Desktop/Lecture_Notes/MVA/Top-Songs-Analysis-master/top10s.csv",header = TRUE)
```

```
View(data)
```

```
#Data Cleaning
```

```
y = data$pop
```

```
View(y)
```

```
max(y)
```

```
## [1] 99
```

```
mean(y)
```

```
## [1] 66.52073
```

```
max(y)
```

```
## [1] 99
```

```
rating <- cut(y, breaks = c(0,67,99),  
             labels = c("Below Average", "Above Average"),  
             right = FALSE, include.lowest = TRUE)
```

```
data['rating'] <- rating
```

```
View(data)
```

```
data_clean <- data[-c(433),]
```

```
View(data_clean)
```

```
#set.seed(422)
```

```
#split = sample.split(data_clean, SplitRatio=0.8)
```

```
#train = subset(data_clean, split == TRUE)
```

```
#test = subset(data_clean, split == FALSE)
```

```
#dim(train)
```

```
#View(train)
```

```
#View(test)
```

```
library(ggplot2)
```

```
fit_lg <- glm(rating~nrgy+dB+dur,data = data_clean, family = "binomial")
```

```
summary(fit_lg)
```

```
##
```

```
## Call:
```

```
## glm(formula = rating ~ nrgy + dB + dur, family = "binomial",
```

```
##      data = data_clean)
```

```

##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.6612  -1.2840   0.9274   1.0368   1.3749
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  3.126867   0.956132   3.270  0.00107 **
## nrgy        -0.017119   0.007002  -2.445  0.01449 *
## dB           0.115497   0.063324   1.824  0.06817 .
## dur         -0.004122   0.002488  -1.656  0.09766 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 816.49  on 601  degrees of freedom
## Residual deviance: 806.25  on 598  degrees of freedom
## AIC: 814.25
##
## Number of Fisher Scoring iterations: 4

## Now calculate the overall "Pseudo R-squared" and its p-value
ll.null <- fit_lg$null.deviance/-2
ll.proposed <- fit_lg$deviance/-2
## McFadden's Pseudo R^2 = [ LL(Null) - LL(Proposed) ] / LL(Null)
(ll.null - ll.proposed) / ll.null

## [1] 0.01254404

## The p-value for the R^2
1 - pchisq(2*(ll.proposed - ll.null), df=(length(fit_lg$coefficients)-1))

## [1] 0.01661637

lrm <- glm(rating~nrgy+dnce+dB+val+dur+spch,data = data_clean, family =
"binomial")
summary(lrm)

##
## Call:
## glm(formula = rating ~ nrgy + dnce + dB + val + dur + spch, family =
"binomial",
##      data = data_clean)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.8376  -1.2511   0.8648   1.0366   1.4465
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)

```

```

## (Intercept)  1.820856    1.087954    1.674 0.094199 .
## nrgy        -0.016671    0.007427   -2.245 0.024792 *
## dnce         0.026445    0.007693    3.438 0.000587 ***
## dB          0.132145    0.067972    1.944 0.051883 .
## val        -0.009256    0.004904   -1.887 0.059102 .
## dur        -0.003869    0.002580   -1.500 0.133676
## spch         0.011611    0.011764    0.987 0.323620
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 816.49  on 601  degrees of freedom
## Residual deviance: 793.67  on 595  degrees of freedom
## AIC: 807.67
##
## Number of Fisher Scoring iterations: 4

## Now calculate the overall "Pseudo R-squared" and its p-value
ll.null <- lrm$null.deviance/-2
ll.proposed <- lrm$deviance/-2
## McFadden's Pseudo R^2 = [ LL(Null) - LL(Proposed) ] / LL(Null)
(ll.null - ll.proposed) / ll.null

## [1] 0.02795105

## The p-value for the R^2
1 - pchisq(2*(ll.proposed - ll.null), df=(length(lrm$coefficients)-1))

## [1] 0.0008584091

#As we can observe if we split the dependent variable rating in 2 levels as
Above Average and Below Average
#the pseudo r-square value is very low and the p-value is low as well.
#Also dependent variable rating is not of binomial type and hence the
Logistic regression
#model does not fit for our data.

```