

Installing Packages

```
install.packages("e1071") install.packages("caTools") install.packages("caret")
```

Loading package

```
library(e1071) library(caTools) library(caret)
```

```
#LDA analysis library (MASS) df40 <- LDA df40Race <-  
-as.numeric(as.factor(df40Race)) set.seed(123) training.individuals <-  
df40Raceclass==test.transform$Race)
```

```
model <- lda(Race~., data = train.transform) model attributes(model) p1 <- predict(model,  
train.transform)classtab <- table(Predicted = p1, Actual = train.transformRace) tab  
sum(diag(tab))/sum(tab) p2 <- predict(model, test.transform)classtab1 <-  
-table(Predicted = p2, Actual = test.transformRace) tab1 sum(diag(tab1))/sum(tab1)
```

```
#Naive Bayes Test by Job df3 <- NBtest3 Job3 <- as.factor(df3$Job) df4=as.data.frame(df3)  
# Splitting data into train and test data split <- sample.split(df4, SplitRatio = 0.7) train_cl <-  
subset(df4, split == "TRUE") test_cl <- subset(df4, split == "FALSE")
```

Feature Scaling

```
train_scale <- scale(train_cl) test_scale <- scale(test_cl)
```

Fitting Naive Bayes Model to training dataset

```
set.seed(1) # Setting Seed classifier_cl <- naiveBayes(Job ~ ., data = train_cl) classifier_cl
```

Predicting on test data

```
y_pred <- predict(classifier_cl, newdata = test_cl)
```

Confusion Matrix

```
cm <- table(test_cl$Job, y_pred) cm
```

Model Evaluation

```
confusionMatrix(cm)
```

```
#Naive Bayes Test by Race (did not work) df10 <- NBtest df10Race <-
—as.numeric(as.factor(df10Race)) df10 df11=as.data.frame(df10) # Splitting data into
train and test data split <- sample.split(df11, SplitRatio = 0.7) train_cl2 <- subset(df11, split
== "TRUE") test_cl2 <- subset(df11, split == "FALSE")
```

Feature Scaling

```
train_scale2 <- scale(train_cl2) test_scale2 <- scale(test_cl2)
```

Fitting Naive Bayes Model

to training dataset

```
set.seed(1) # Setting Seed classifier_cl2 <- naiveBayes(Race ~ ., data = train_cl2)
classifier_cl2
```

Predicting on test data'

```
y_pred <- predict(classifier_cl2, newdata = test_cl2)
```

Confusion Matrix

```
cm <- table(test_cl2$Race, y_pred) cm
```

```
u <- union(y_pred, test_cl2) t <- table(factor(y_pred, u), factor(test_cl2, u))
confusionMatrix(t)
```

Model Evaluation

```
confusionMatrix(cm)
```

```
#KMeans cluster analysis library (tidyverse) library (cluster) library (factoextra) df30<-
LDA df30<- na.omit(df30) df30Race <- —as.numeric(as.factor(df30Race)) df30 <-
scale(df30) head(df30) distance <- get_dist(df30) fviz_dist(distance, gradient = list(low =
"#00AFBB", mid = "white", high = "#FC4E07")) k2 <- kmeans(df30, centers = 2, nstart = 25)
str(k2) k2 fviz_cluster(k2, data = df30) df30 %>% as_tibble() %>% mutate(cluster =
k2cluster, Race = row.names(LDA))cluster, Race = row.names(LDA)) %>%
ggplot(aes(Race, Hairstylist, color = factor(cluster), label = Race)) + geom_text() df30 %>%
as_tibble() %>% mutate(cluster = k2cluster, Race = row.names(LDA))cluster, Race =
row.names(LDA)) %>% ggplot(aes(Race, Teacher, color = factor(cluster), label = Race)) +
geom_text()
```

```
k3 <- kmeans(df30, centers = 3, nstart = 25) k4 <- kmeans(df30, centers = 4, nstart = 25) k5
<- kmeans(df30, centers = 5, nstart = 25)
```

plots to compare

```
p1 <- fviz_cluster(k2, geom = "point", data = df30) + ggtitle("k = 2") p2 <- fviz_cluster(k3,
geom = "point", data = df30) + ggtitle("k = 3") p3 <- fviz_cluster(k4, geom = "point", data =
df30) + ggtitle("k = 4") p4 <- fviz_cluster(k5, geom = "point", data = df30) + ggtitle("k = 5")
```

```
library(gridExtra) grid.arrange(p1, p2, p3, p4, nrow = 2)
```

```
df30 %>% as_tibble() %>% mutate(cluster = k5$cluster, Race = row.names(LDA)) %>%
ggplot(aes(Race, Engineer, color = factor(cluster), label = Race)) + geom_text()
```

```
set.seed(123)
```

```
fviz_nbclust(df30, kmeans, method = "wss")
```

```
set.seed(123) final <- kmeans(df30, 2, nstart = 25) print(final) fviz_cluster(final, data =
df30)
```

```
LDA %>% mutate(Cluster = final$cluster) %>% group_by(Cluster) %>%
summarise_all("mean")
```

```
library(MASS) #Original exploratory sections library(ggplot2) library(dplyr) filter
(Twogroups, Job == "Engineer") ggplot(Twogroups) + aes (x=Race, y = Total) +
geom_boxplot (fill = "orchid") + theme_minimal() hist(subset(Twogroups, Race
=="Was")$Total, main = "Total females, White/Asian", xlab = "In Engineering")
hist(subset(Twogroups, Race == "WOC")$Total, main = "Total females, Women of Color",
xlab = "In Engineering") shapiro.test(subset(Twogroups, Race == "Was")$Total)
shapiro.test(subset(Twogroups, Race == "WOC")$Total)
```

```
#Group means test test <- wilcox.test(Twogroups$Total ~ Twogroups$Race) test
```

```
filter (Twogroups2, Job == "Engineer") ggplot(Twogroups2) + aes (x=Race, y = Total) +
geom_boxplot (fill = "orchid") + theme_minimal() hist(subset(Twogroups2, Race
=="White")$Total, main = "Total females, White", xlab = "In Engineering")
hist(subset(Twogroups2, Race == "WOC")$Total, main = "Total females, Women of Color",
xlab = "In Engineering") shapiro.test(subset(Twogroups2, Race == "White")$Total)
shapiro.test(subset(Twogroups2, Race == "WOC")$Total) test <-
wilcox.test(Twogroups2$Total ~ Twogroups2$Race) test
```

```
hist(subset(Twogroups2, Job == "White")$Total, main = "Total females, White", xlab = "In
Engineering") res <- prop.test
```

```
#Exploratory charts library (tidyverse) ggplot(Twogroups, aes (x=Job, y = Total)) +
geom_boxplot() ggplot(Twogroups, aes (x=Job, y = Total, color = Race)) + geom_boxplot()
install.packages("ggplot2") library(ggplot2) subset (Twogroups2, Job == "Engineer")
ggplot(Twogroups2, aes(x = Race, y = Total)) + geom_bar(stat="identity", fill = "blue") +
```

```
xlab('WOC + Asian') + ylab('Total Female Engineers') + geom_text(aes(label =  
signif(sum(Total)), nudge_y = 2000000))
```