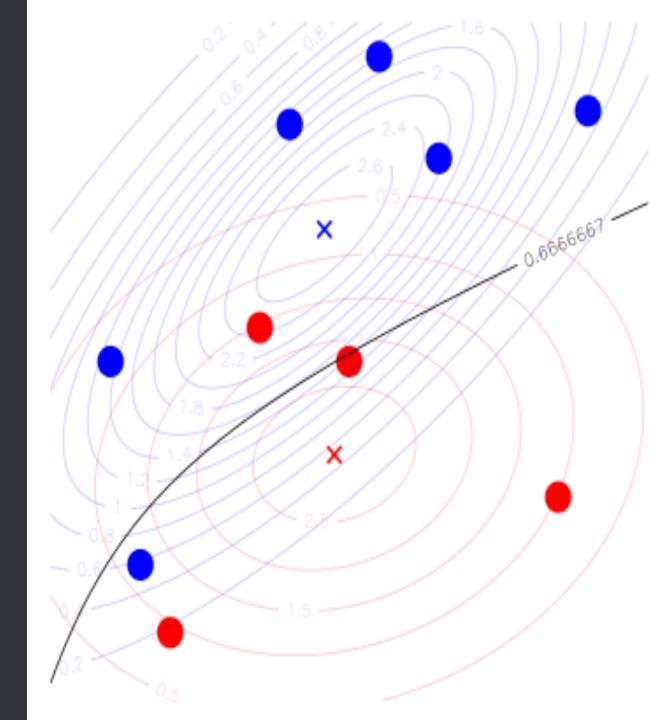
# WEEK 9:

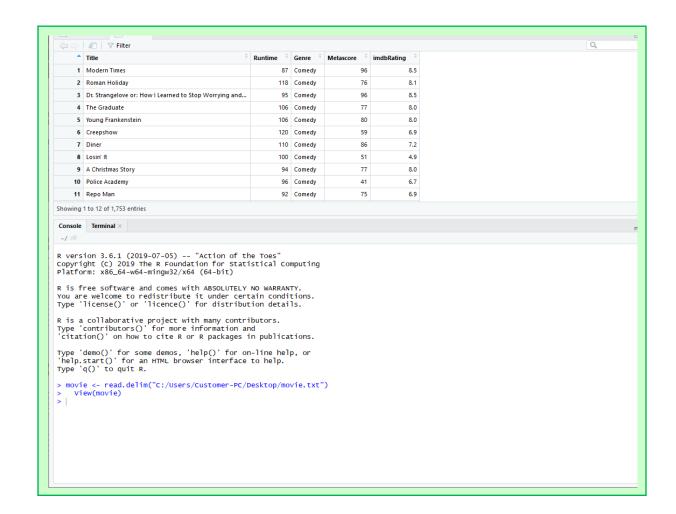
## INTRODUCTION TO DISCRIMINANT ANALYSIS



Tutorial Week 9 STA30005

 Data Preparation
 Splitting data (train / test)
 Applying LDA
 Visualisation
 Predictions

```
#Load the movie data file (in the tutorial 9 folder)
movie <- read.delim("LOCATION.txt")</pre>
# Show the data
View(movie)
# Install / Load the required packages
install.packages("tidyverse")
library(tidyverse)
install.packages("MASS")
library(MASS)
install.packages("klaR")
library(klaR)
# Set a seed value
set.seed(123)
```





Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions

We can separate the data into two subsets: a training set (building the model) and a testing set (evaluate the accuracy of the model). For convenience sake we will use a 50/50 split, using 50% of the data as the training set and the remaining 50% for the testing set.

```
# Split the data 50/50 training_sample <- sample(c(TRUE, FALSE), nrow(movie), replace = T, prob = c(0.5,0.5))
```

```
# Define the training data
train <- SATB[training_sample,]
# Define the testing data
test <- SATB[!training_sample,]
```

```
You are welcome to redistribute it under certain conditions.

Type 'license()' or 'licence()' for distribution details.

R is a collaborative project with many contributors.

Type 'contributors()' for more information and 'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or 'help.start()' for an HTML browser interface to help.

Type 'q()' to quit R.

> movie <- read.delim("C:/Users/Customer-PC/Desktop/movie.txt")

> View(movie)

> training_sample <- sample(c(TRUE, FALSE), nrow(movie), replace = T, prob = c(0.5,0.5))

> train <- movie[training_sample, ]

> test <- movie[training_sample, ]

> |
```

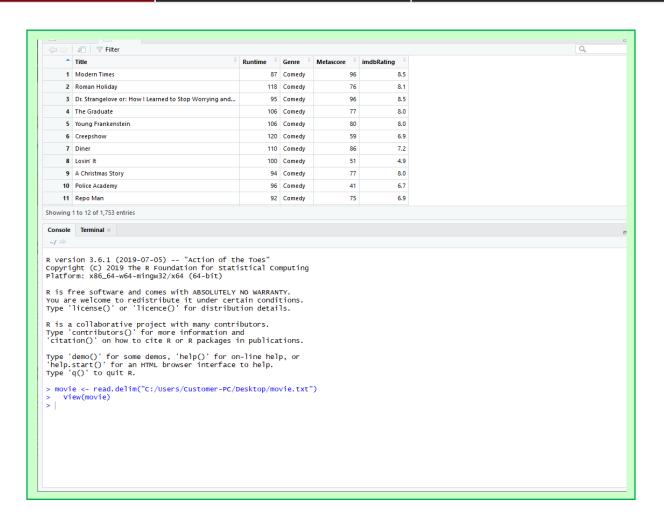


Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions

# Create an initial LDA model (m1) based upon the data m1<-lda(Genre ~ Runtime + Metascore + imdbRating, train)

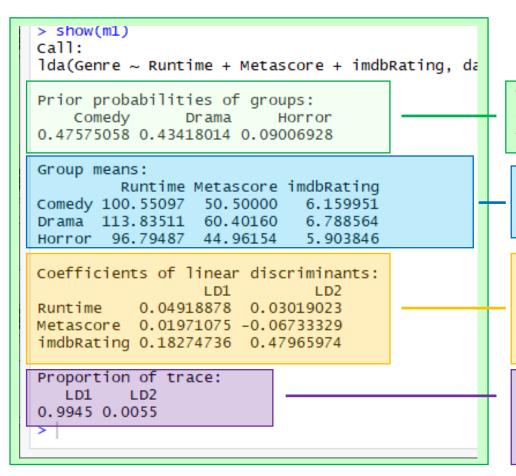
# Show the results show(m1)

```
> m1<-lda(Genre~Runtime+Metascore+imdbRating,train)
Error in lda(Genre ~ Runtime + Metascore + imdbRating, train) :
  could not find function "lda"
> library(MASS)
> m1<-lda(Genre~Runtime+Metascore+imdbRating,train)
> show(m1)
call:
lda(Genre ~ Runtime + Metascore + imdbRating, data = train)
Prior probabilities of groups:
    Comedy
                Drama
0.47575058 0.43418014 0.09006928
         Runtime Metascore imdbRating
Comedy 100.55097 50.50000
                            6.159951
Drama 113.83511 60.40160
                            6.788564
Horror 96.79487 44.96154
                            5.903846
Coefficients of linear discriminants:
                  LD1
          0.04918878 0.03019023
Metascore 0.01971075 -0.06733329
imdbRating 0.18274736 0.47965974
Proportion of trace:
   LD1
0.9945 0.0055
```





Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions



The Prior probabilities of groups show the probability of randomly selecting an observation from class the total training set (e.g. 47.6% chance to be comedy, 43.4% drama and 9.0% to be horror)

Shows the mean values for the different variables split by the classification factor (in this example: Genre)

```
LD1 = 0.05(Runtime) + 0.02(Metascore) + 0.18(imdbRating)
```

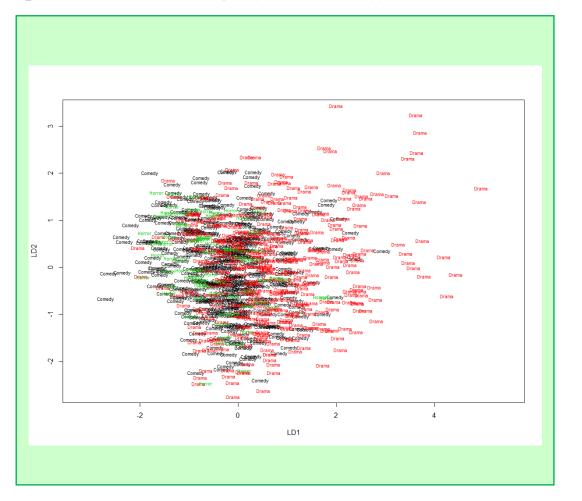
LD1 = 0.03(Runtime) - 0.07(Metascore) + 0.48(imdbRating)

The Proportions of trace describes the proportion of between-class variance that is explained by successive discriminant functions. As you can see LD1 explains 99.45% of the variance.

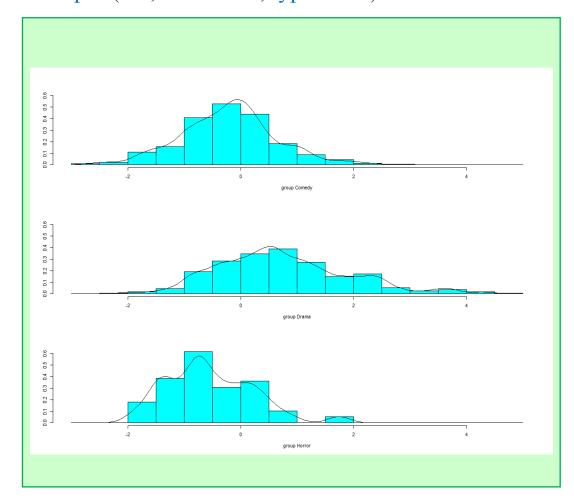


Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions

# Plot LD1 and LD2
plot(m1, col=as.integer(train\$Genre))



# Plot LD1 only
plot(m1, dimen = 1, type = "b")



Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions

Next let's evaluate the prediction accuracy of our model. First we'll run the model against the training set used to verify the model fits the data properly by using the command predict. The table output is a confusion matrix with the actual as the rows and the predicted as columns

```
# Compare model against test set
lda.train <- predict(m1)
train$lda <- lda.train$class
table(train$lda, train$Genre)
```

- The total number of correctly predicted observations is the sum of the diagonal (328 + 226 + 0 = 554). So this model fit the training data correctly for 63.97% of cases.
- Verifying the training set doesn't prove accuracy, but a poor fit to the training data could be a sign that the model isn't a good one.

```
Comedy
0.47575058 0.43418014 0.09006928
         Runtime Metascore imdbRating
Comedy 100.55097 50.50000
Drama 113.83511 60.40160
                             6.788564
Horror 96.79487 44.96154
Coefficients of linear discriminants:
          0.04918878 0.03019023
Metascore 0.01971075 -0.06733329
imdbRating 0.18274736 0.47965974
Proportion of trace:
   LD1
        LD2
0.9945 0.0055
> plot(m1, col=as.integer(train$Genre))
> plot(m1, dimen = 1, type = "b")
> lda.train <- predict(m1)</pre>
> train$1da <- lda.train$class
> table(train$lda, train$Genre)
         Comedy Drama Horror
  Comedy
                  226
                          10
  Drama
  Horror
```



Data Preparation Splitting data (train / test) Applying LDA Visualisation Predictions

Now let's run our test set against this model to determine its accuracy.

~/ @

#### # Compare model against test set

```
lda.test <- predict(m1, test)
test$lda <- lda.test$class
table(test$lda, test$Genre)</pre>
```

- The total number of correctly predicted observations is the sum of the diagonal (317 + 154 + 0 = 471).
- The overall accuracy is only 53.1%
- Therefore these three variables are not good at discriminating between movie genres (horror in particular was really bad)

```
Coefficients of linear discriminants:
          0.04918878 0.03019023
Metascore 0.01971075 -0.06733329
imdbRating 0.18274736 0.47965974
Proportion of trace:
         LD2
0.9945 0.0055
> plot(m1, col=as.integer(train$Genre))
> plot(m1, dimen = 1, type = "b")
> lda.train <- predict(m1)</pre>
> train$lda <- lda.train$class
> table(train$lda, train$Genre)
                           10
> lda.test <- predict(m1, test)</pre>
> table(test$lda, test$Genre)
  Comedy
                  220
                           10
  Horror
```





Now repeat the task with the IRIS data frame (base R data)

[this one discriminates quite well]

