SCHOOL OF ELECTRICAL ENGINEERING AND COMPUTING

BIG DATA AND DATA ANALYTICS LAB PROJECT 5 (DATASET DESCRIPTION & PREPARATION)



This lab project is based on a dataset about movie success in 2014 and 2015 by Ahmad et al. (2015) which is available on the online platform by Lichman et al (2013). Download the file movidata.csv from Blackboard and then practice the following topics in preparations for Lab Project 5.

Preparation in Week 10

In preparation for Lab Project 5, load the moviedata.csv dataset. Use the dataset to practice the following topics:

- Logistic regressions: coefficient estimates, predicted probabilities, residuals, standard errors, confidence intervals, z-values, AIC, log likelihood
- Search for and watch online videos and blog entries about implementing a classifier (based on logistic regressions and classification trees; e.g., YouTube, R-Bloggers, Stackoverflow)
- Reporting of results in APA style
- Analysing and interpreting the results of logistic regressions
- Creating and interpreting a confusion matrix
- Sensitivity, specificity, and accuracy
- Area under the ROC
- Decision trees: regression trees and classification trees
- Pruning decision trees
- Analysing and interpreting the results of decision trees

REFERENCES

Ahmed M, Jahangir M, Afzal H, Majeed A, Siddiqi I. Using Crowd-source based features from social media and Conventional features to predict the movies popularity. In Smart City/ SocialCom/S ustainCom (SmartCity), 2015 IEEE International Conference on 2015 Dec 19 (pp. 273-278). IEEE. https://ieeexplore.ieee.org/document/7463737

Lichman, M. (2013). UCI Machine Learning Repository [http://archive.ics.uci.edu/ml]. Irvine, CA: University of California, School of Information and Computer Science.

DATASET

moviedata

Conventional and Social Media Movies 2014 and 2015

Description

A dataset about the success of movies in 2014 and 2015.

Usage

moviedata

Format

A data frame with 231 observations on the following 14 variables.

movie Name of the movie year Year of movie release ratings Rating of the movie (0-10)

genre Identifier for the genre of the movie (e.g., action, adventure, drama)

gross Gross world-wide income from the movie (in US\$)

budget Budget for the movie

Number of screens that the movie was initially launched in on the

opening weekend in the US

sequel A number indicating whether the movie is sequel or original

(individual) movie, where higher numbers indicate later sequels in a series. For instance, for Mission Impossible a sequel value of 5

indicates that this is the fifth movie in the series.

 $dummy_sequel$ 0 - Original movie

1 – Sequel movie

sentiment A sentiment score assessed through an analysis of tweets about the

movie on Twitter. 0 represents a neutral sentiment, a positive value represents a positive sentiment, and a negative value indicates a negative sentiment. The sentiment score for each movie was calculated by retrieving all tweets related to each movie, assigning the sentiment score to each of them and then aggregating the score.

viewsNumber of times the movie trailer was viewed on YouTubelikesNumber of likes the movie trailer received on YouTubedislikesNumber of dislikes the movie trailer received on YouTube

comments Number of times the movie trailer received a comment on YouTube aggregate_followers The aggregate number of actor followers: Equal to sum of followers

of top 3 cast from Twitter

Source

Ahmed M, Jahangir M, Afzal H, Majeed A, Siddiqi I. Using Crowd-source based features from social media and Conventional features to predict the movies popularity. In Smart City/SocialCom/S ustainCom (SmartCity), 2015 IEEE International Conference on 2015 Dec 19 (pp. 273-278). IEEE. https://ieeexplore.ieee.org/document/7463737

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