

# Movie Recommendation System

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Link: [https://github.com/helenapanchenko/DS\\_Internship\\_Glacier/tree/master/Week9](https://github.com/helenapanchenko/DS_Internship_Glacier/tree/master/Week9)

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## Problem Description

The video-on-demand streaming service is looking to develop a machine learning algorithm that can predict which movies a user will enjoy based on various factors such as genre, online ratings, and previous decisions. The primary objective is to create a system for movie recommendations.

## Data Preprocessing

### Data Cleansing

#### Dealing with Missing Values

We use only genome-scores, movies and ratings files in our project. None of these files contain missing values.

## Dealing with Outliers

We examined the 'relevance' column from the genome-scores file and the 'rating' column from ratings file.

**'relevance'**: In our case, outliers for the 'relevance' column are values less than 0 and greater than 1. We don't have such values and thus don't have outliers in the 'relevance' column.

**'rating'**: In our case, outliers for the 'rating' column are values less than 0.5 and greater than 5. We don't have such values and thus don't have outliers in the 'rating' column.

## Dealing with Duplicates

We have no duplicates in any of our files.

## Data Transformation

### Dealing with Categorical and Text Data

Only movies file contains categorical and textual data.

**'title'**: Column with textual data and has no value for our analysis. But each title contains a year the movie was released which can be useful for our model. Thus we extracted the release year from the 'title' column and kept it.

**'genres'**: Column with categorical data which we encoded using Multi-label Binarisation.

### Dealing with unnecessary columns

1. We dropped the **'timestamp'** column from the ratings file because it is irrelevant to our analysis.
2. We dropped the **'(no genres listed)'** column from the movies file to avoid repetition(zeros in every genre column already explain this case).

## Data Subsetting

For our analysis we use only used of users who left at least 5 ratings.