

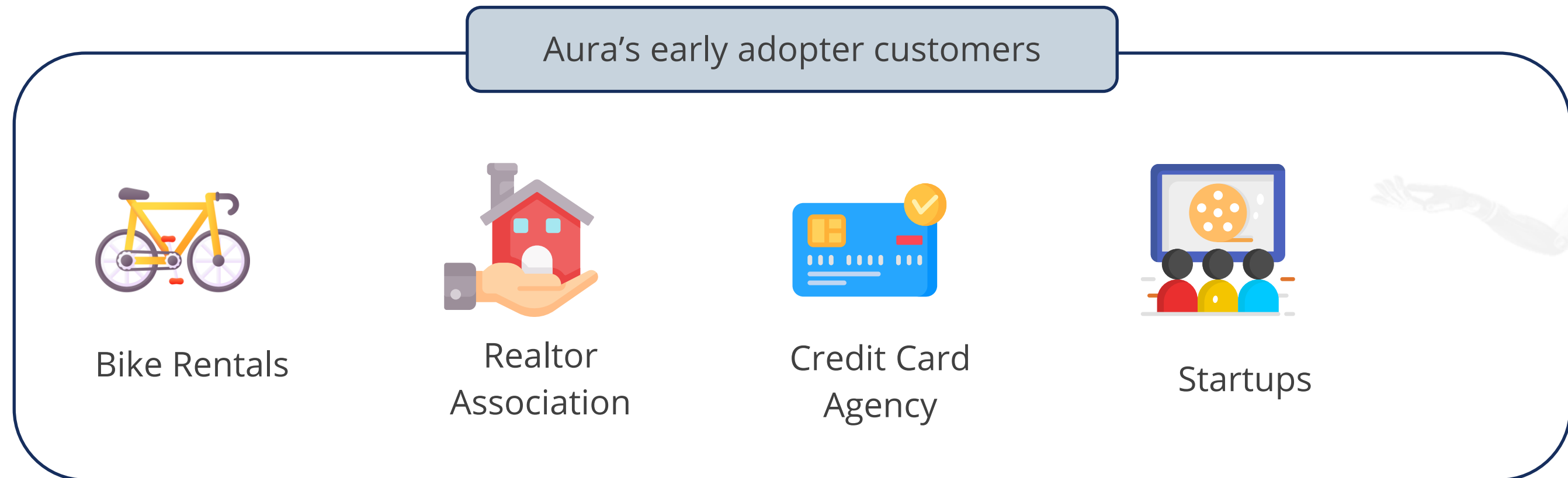


## Capstone Session 8

## Machine Learning for Modeling

# Machine Learning End Goal

The intelligence provided by Aura will help customers make decisions for their omnichannel marketing and customer acquisition programs.



# Project Statement

Aura must do the following:



Predict bike-sharing demand

Classify incomes

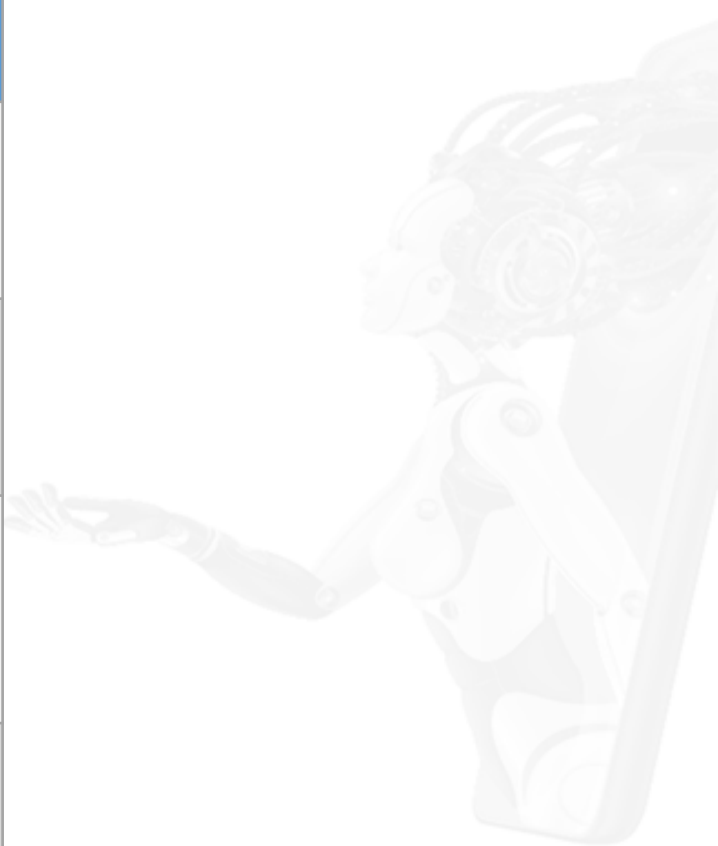
Cluster credit card users

Build a recommendation engine

# Week 8: Dataset Description

ratings.csv

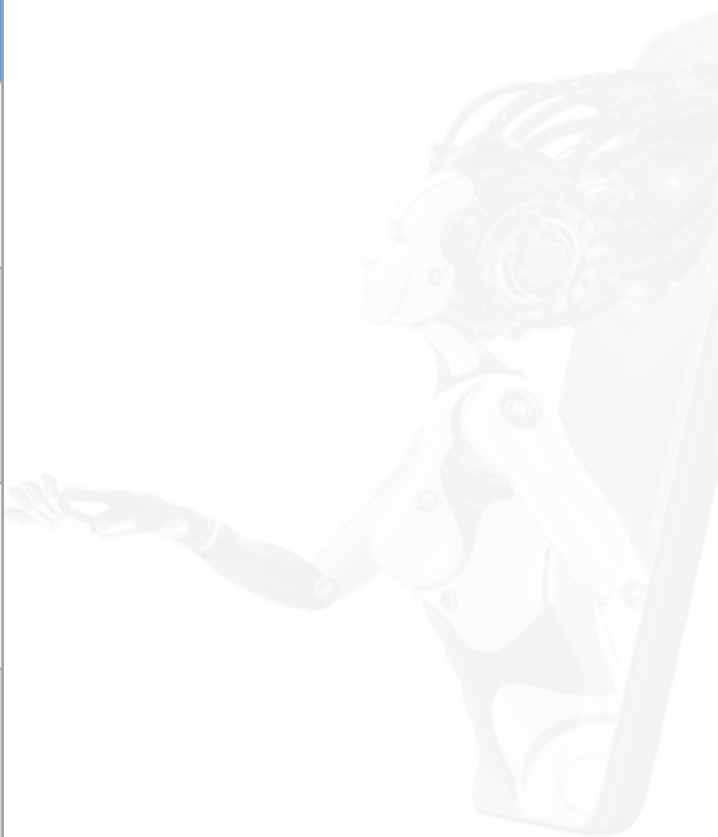
| Variable  | Description   |
|-----------|---|
| userId    | A unique Id represents every user.                                  |
| movieId   | Unique ID representing a movie                                      |
| rating    | Represents the rating given by the user to the corresponding movie. |
| timestamp | The time at which the rating was recorded.                          |



# Week 8: Dataset Description

movies.csv

| Variable  | Description  |
|-----------|--|
| movieId   | A unique Id represents every user.                           |
| title     | Movie name which is represented by the corresponding movieId |
| genres    | Represents the category of the movie.                        |
| timestamp | The time at which the rating was recorded.                   |

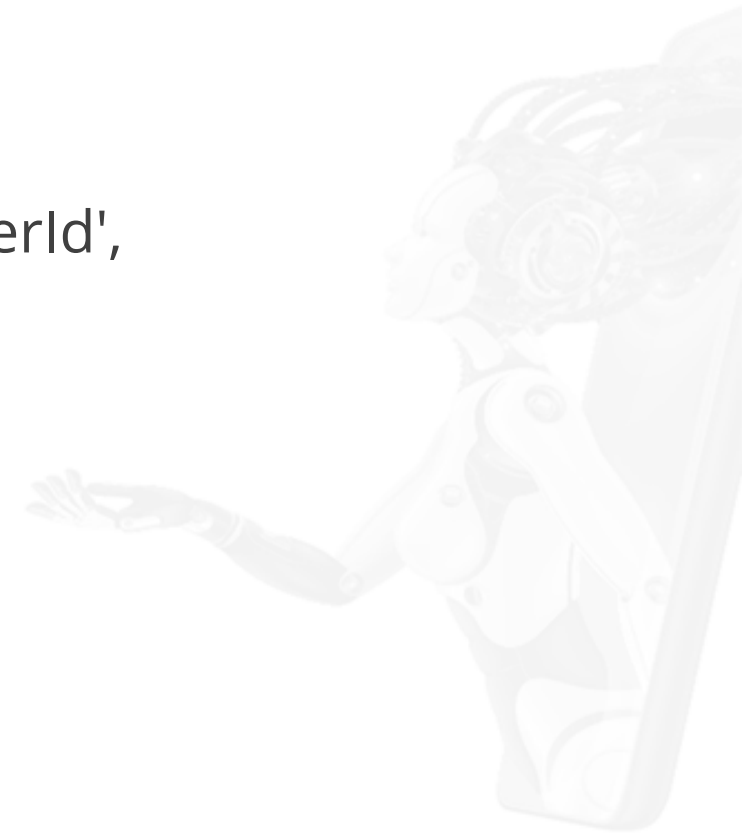


## Week 8

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**Task:** Study the various Recommendation Techniques for recommending movies using movies.csv, ratings.csv datasets

- Load movies.csv and ratings.csv dataset
- Merge both data frames on movieid
- Create User-Item Matrix (Hint: Use pandas pivot\_table method with index = 'userId', columns = 'title', values = 'rating' )





## Week 8

- Perform User-based Collaborative Filtering
  - Fill the row-wise NaNs in the User-Item Matrix with the corresponding user's mean ratings, and find the Pearson correlation between users
  - Choose the correlation of all users with User 1 only.
  - Sort the user 1 correlation in the descending order
  - Drop the NaN values generated in the correlation matrix
  - Choose the top 50 users that are highly correlated to User 1
  - Predict the rating that User 1 might give for the movie with movieid 32 based on the top 50 user correlation matrix

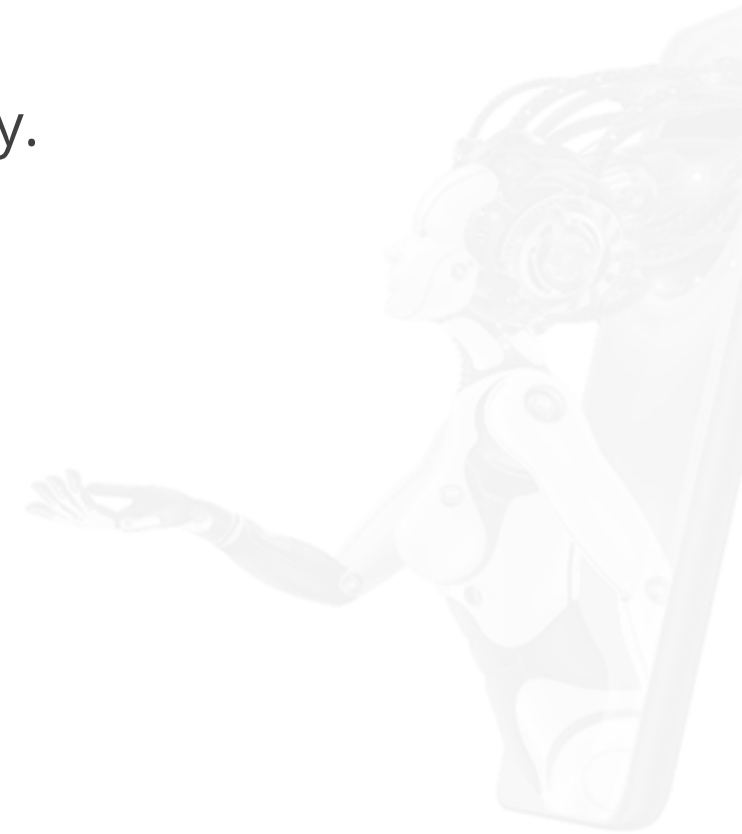
(Hint: Predicted rating =  $\text{sum of } [(\text{weights}) * (\text{ratings})] / \text{sum of } (\text{weights})$  . Here, weights is the correlation of the corresponding user with the first user.). That is, the predicted rating is calculated as the weighted average of k similar users



## Week 8

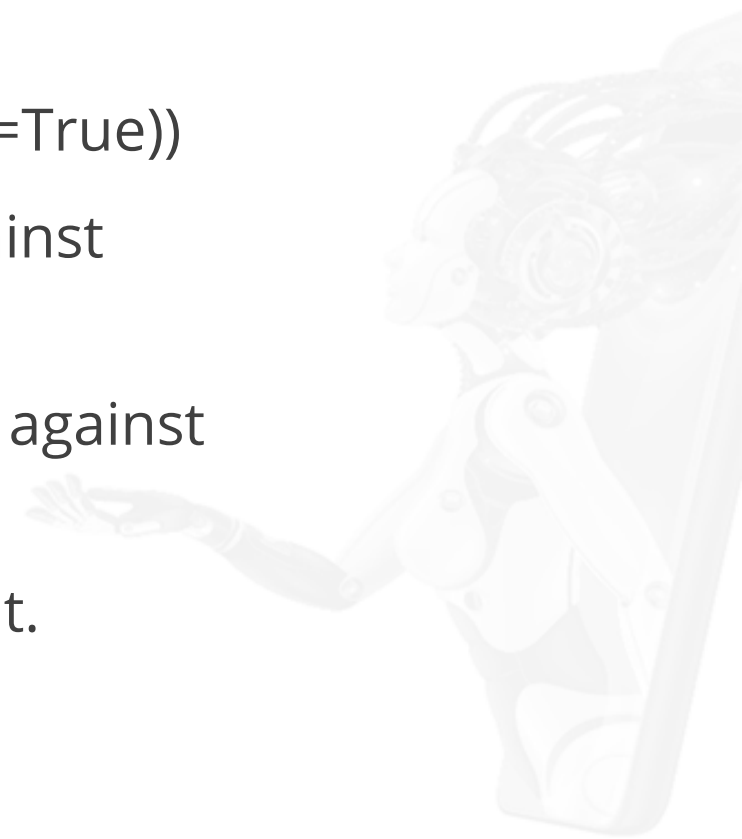
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- Perform Item-based Collaborative Filtering
  - Fill the column-wise NaN's in the User-Item Matrix with the corresponding movie's mean ratings, and find Pearson correlation between movies
  - Choose the correlation of all movies with the movie Jurassic Park (1993) only.
  - Sort the Jurassic Park movie correlation in descending order
  - Drop the NaN values generated in the correlation matrix
  - Find the most 10 movies similar to the movie Jurassic Park (1993)



## Week 8

- Perform KNNBasic, SVD, NMF Model-based Collaborative Filtering
  - Initialize KNNBasic with similarity configuration as Mean Squared Distance Similarity (msd), 20 neighbors and cross-validate 5 folds against measure RMSE.  
(Hint: `cross_validate(algo=algo, data=data, measures=['RMSE'], cv=5, verbose=True)`)
  - Initialize Singular Value Decomposition (SVD) and cross-validate 5 folds against measure RMSE.
  - Initialize Non-Negative Matrix Factorization (NMF) and cross-validate 5 folds against measure RMSE.
  - Print best score and best params from Cross Validate on all the models built.



**Thank You**