Recommender System for Movies

Big Data Project

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https://github.com/filippoguerranti/recommender_system

Outline

- 1. Dataset
- 2. Collaborative filtering
- 3. Content based

Dataset

Dataset

MovieLens

5-star rating and free-text tagging activity about movies

9742

movies

movies.csv

610

users

only IDs

3683

tags

tags.csv

100863

ratings

ratings.csv

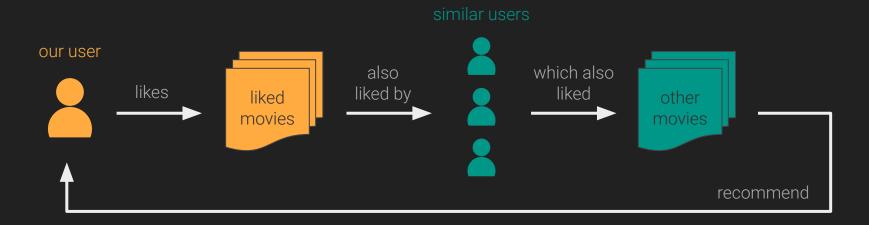
Files

	movield	title	genres
	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy
movies.csv	2	Jumanji (1995)	Adventure Children Fantasy
MOV100.00V	3	Grumpier Old Men (1995)	Comedy Romance
	userId	movield	rating
	1	1	4.0
ratings.csv *	1	3	4.0
1 4 6 2 1 1 9 0 1 0 0 1	1	6	4.0
	userld	movield	tag
	2	60756	funny
tags.csv *	2	60756	Highly quotable
eago.oov	2	60756	will ferrell

Collaborative Filtering

Collaborative Filtering

Idea: find movies recommendations for a user identified by userId based on the ratings of users that have shown similar behaviour (similar ratings).



Collaborative Filtering

Libraries: pandas, numpy, surprise

Reader

Dataset

• SVD

KNNBaseline

- accuracy
- model selection.KFold

Outline:

- Implementation
- Results
- Predictions

- * different algorithms
- * 5-fold cross-validation
- * accuracy measures

Algorithms:

KNN

- * Cosine similarity and Baseline (KNN-cosine-baseline)
- * Pearson's Coefficient and Baseline (KNN-pearson-baseline)

• SVD

- * N. epochs: 20 | Learning Rate: 0.005 (SVD-20ep-.005lr)
- * N. epochs: 50 | Learning Rate: 0.003 (SVD-50ep-.003lr)

KNN

Goal: estimate the rating that user x will give to item i

Idea: find the k most similar items to item i that are rated by x

$$\hat{r}_{xi} = b_{xi} + \frac{\sum_{j \in N(i;x)} sim(i,j) \cdot (r_{xj} - b_{xj})}{\sum_{j \in N(i;x)} sim(i,j)}$$

sim(i,j) similarity measure between item **i** and item **j**

 b_{xi} baseline estimate for user **x** and item **i** $b_{xi} = \mu + b_x + b_i$

SVD

Goal: estimate the rating that user x will give to item i

Idea: use dimensionality reduction to find the latent factors

$$\hat{r}_{xi} = \sum_{s} q_{is} \cdot p_{fs}$$

Find P and Q: by the Stochastic Gradient Descent method

$$\min_{P,Q} \sum_{training} (r_{xi} - q_i p_x)^2 + \lambda \left[\sum_{x} ||p_x||^2 + \sum_{i} ||q_i||^2 \right]$$

5-fold cross-validation:

For each algorithm:

For each fold:

- * **split** the dataset into training set and test set
- * train the algorithm on the training set
- * **test** the algorithm on the test set
- * measure the accuracy of the current fold

Accuracy measures:

Root Mean Squared Error (RMSE)

$$RMSE = \sqrt{\sum_{test} \frac{(\hat{r}_{xi} - r_{xi})^2}{N}}$$

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Results

Algorithm	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Mean
KNN-cosine-baseline	0.869872	0.883382	0.879242	0.874197	0.879286	0.877196
KNN-pearson-baseline	0.879939	0.875910	0.875314	0.878922	0.877261	0.877469
SVD-20ep0051r	0.870731	0.884309	0.871147	0.866316	0.877215	0.873944
SVD-50ep0031r	0.868532	0.875322	0.879058	0.879815	0.868313	0.874208

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Predictions

Top 5 recommendations for user 238

movieId	est_rating	title	genres
3275	4.55	Boondock Saints, The (2000)	Action Crime Drama Thriller
5690	4 . 52	Grave of the Fireflies (Hotaru no haka) (1988)	Animation Drama War
3972	4.48	Legend of Drunken Master, The (Jui kuen II) (1	Action Comedy
7371	4.44	Dogville (2003)	Drama Mystery Thriller
7008	4.44	Last Tango in Paris (Ultimo tango a Parigi) (1	Drama Romance

Content based

Content Based

Idea: find movies recommendations for a user identified by userId based on the similarity between the user profile and the movie profile (content).



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Content Based

Libraries: pandas, numpy, sklearn

- feature_extraction.text.TfidfVectorizer
- feature_extraction.text.CountVectorizer
- metric.pairwise.cosine_similarity
- decomposition.TruncatedSVD

Outline:

- Implementation
- Predictions

- * Movies class
- * Users class
- * ContentBased class

Movies class: takes in input the movies and the tags DataFrames and creates the movies profiles.

```
def __init__(self, movies_df, tags_df, tfidf=True, lsa=True, n_components=40)
def __create_movies_profiles(self, movies_df, tags_df)
def __create_movies_terms(self, tfidf=True)
def __latent_semantic_analysis(self, n_components=40)
def movie_vector(self, movieId)
```

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Users class: takes in input the ratings DataFrame and an instance of Movies class and creates the users profiles.

```
def __init__(self, ratings_df, movies_instance)
def __create_users_movies_dict(self, ratings_df)
def __create_users_vectors(self, mv)
def user_vector(self, userId)
```

ContentBased class: takes in input an instance of Users class and an instance of Movies class and creates the recommendations.

```
def __init__(self, users_instance, movies_instance)
def recommend(self, userId, n_recommendations=10)
```

Recommendations are created by computing the **cosine similarity** between the user profile and all the movies profiles.

Predictions

Top 5 recommendations for user 238

movieId	similarity	title	genres
5628	0.894	Wasabi (2001)	Action Comedy Crime Drama Thriller
5027	0.894	Another 48 Hrs. (1990)	Action Comedy Crime Drama Thriller
1432	0.894	Metro (1997)	Action Comedy Crime Drama Thriller
145	0.894	Bad Boys (1995)	Action Comedy Crime Drama Thriller
20	0.894	Money Train (1995)	Action Comedy Crime Drama Thriller

Comparison

	genres	title	est_rating	movieId
Collaborative filtering	Action Crime Drama Thriller	Boondock Saints, The (2000)	4.55	3275
	Animation Drama War	Grave of the Fireflies (Hotaru no haka) (1988)	4.52	5690
	Action Comedy	Legend of Drunken Master, The (Jui kuen II) (1	4.48	3972
	Drama Mystery Thriller	Dogville (2003)	4.44	7371
	Drama Romance	Last Tango in Paris (Ultimo tango a Parigi) (1	4.44	7008
	genres	title	similarity	movieId
Content	on Comedy Crime Drama Thriller	Wasabi (2001) Act	0.894	5628
Content	on Comedy Crime Drama Thriller	Another 48 Hrs. (1990) Act:	0.894	5027
based	on Comedy Crime Drama Thriller		0.894 0.894	5027 1432
		Metro (1997) Act :		
	on Comedy Crime Drama Thriller	Metro (1997) Act:	0.894	1432

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Code can be found at https://github.com/filippoguerranti/recommender_system