Collaborative Filtering

based on user similarity

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Task

Several users already rated an item

One user haven't rated it yet

 \rightarrow Will he like it?

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Goal: Predict his rating for this item

How: Similarity between users

Main issue

How to measure the similarity?

Efficiency of the algorithm?

Algorithm

message	# Ken	Lee	Meg	Nan
1	1	4	2	2
2	5	2	4	4
3			3	
4	2	5		5
5	4	1		1
6	?	2	5	?

Figure 5: a sample matrix of ratings.

Step 1: Sample matrix

Algorithm



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Figure 5: a sample matrix of ratings.

Step 1: Sample matrix

$$r_{KL} = \frac{Cov(K, L)}{\sigma_K \sigma_L}$$

$$= \frac{\sum_i (K_i - \overline{K})(L_i - \overline{L})}{\sqrt{\sum_i (K_i - \overline{K})^2} \sqrt{\sum_i (L_i - \overline{L})^2}}$$

$$= \frac{-2 - 2 - 2 - 2}{\sqrt{10} \sqrt{10}} = -0.8$$
Step 2: Define the covariance matrix

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$$K_{6_{\text{Pred}}} = \overline{K} + \frac{\sum (J_6 - \overline{J})r_{\text{KJ}}}{\sum |r_{\text{KJ}}|} =$$

$$3 + \frac{2r_{\text{KM}} - r_{\text{KL}}}{|r_{\text{KM}}| + |r_{\text{KL}}|} = 3 + \frac{2 - (-.8)}{|1| + |-.8|} = 4.56$$
Step 3: Define the rating based on the others users weighted rating

Introduction to the dataset

Dataset: MovieLens data set

collected by the GroupLens Research Project at the University of Minnesota

http://ict.ewi.tudelft.nl/~jun/CollaborativeFiltering.html

User	Movie	Grade
1	1	4
2	1	1
3	2	5

Extraction from the dataset

943 users 1682 movies Grades from 1 to 5 100 000 ratings

Histogram

