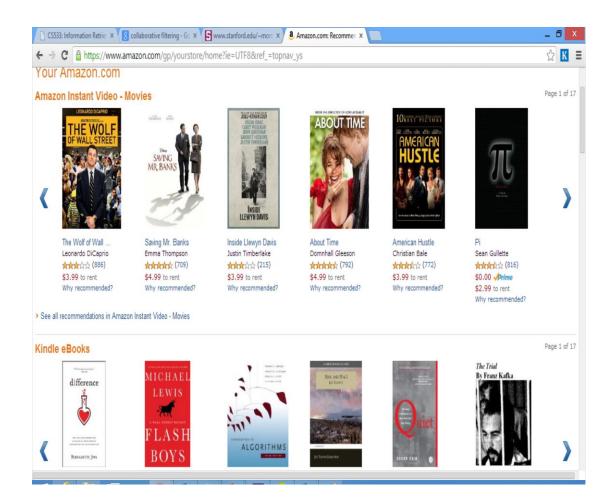
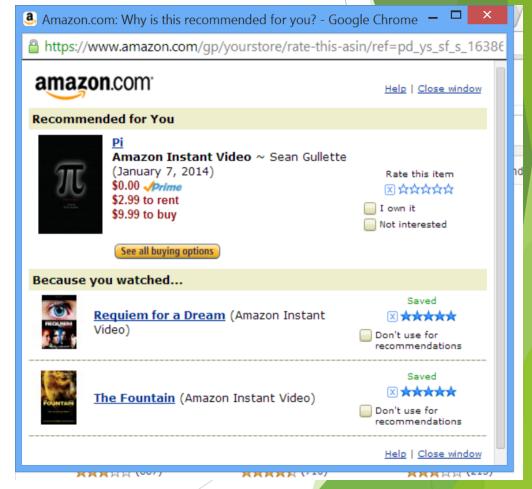
### Movie Recommendation System

Neslihan Bulut

#### Recommendation Systems

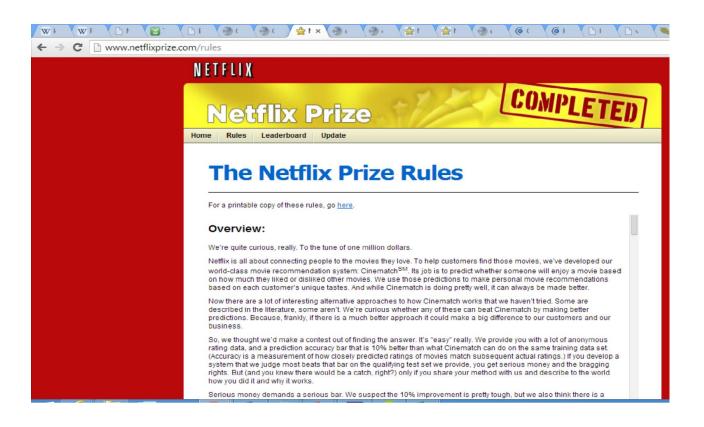




#### Collaborative Filtering

- Predictions on how well a user will like an item
- ► Has a training data of items, users and user votes
- Neighborhood methods are used; find similar users and make predictions based on their votes

#### **Netflix Data**



#### **Netflix Data**

- 3M votes for the training data
- ▶ 100K votes for testing the accuracy of the algorithm

# Algorithms: Collaborative Filtering (Breese et al.)

- $V_{i,i}$  vote of user i on item j
- $I_i$  = items for which user i has voted
- Mean vote for i is

$$\overline{v}_i = \frac{1}{|I_i|} \sum_{j \in I_i} v_{i,j}$$

Predicted vote for "active user" a is weighted sum

$$p_{a,j} = \overline{v}_a + \kappa \sum_{i=1}^n w(a,i)(v_{i,j} - \overline{v}_i)$$
 normalizer weights of  $n$  similar users

# Algorithms: Collaborative Filtering (Breese et al.) - Cont.

Pearson correlation coefficient (Resnick '94, Grouplens):

$$w(a,i) = \frac{\sum_{j} (v_{a,j} - \overline{v}_a)(v_{i,j} - \overline{v}_i)}{\sqrt{\sum_{j} (v_{a,j} - \overline{v}_a)^2 \sum_{j} (v_{i,j} - \overline{v}_i)^2}}$$

- Most similar users are selected to predict scores for unrated movies
  - ► Empirical Analysis of Predictive Algorithms for Collaborative Filtering, Breese, Heckerman, Kadie

#### **Evaluation**

- Test and training sets from the Netflix prize
- Compare the predictions from the algorithm with the actual vote data from the test set

### Thanks for listening!

Any questions?