# Item Based KNN Collaborative Filtering

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#### Item Based CF

Memory Based Collaborative Filtering: Predict a rating for user u and movie m by using previous ratings

ltem Based: sim(m, m')

### **Problem Formulation**

### Rating Matrix

- Rows represent ratings given to each movie
- Columns represent ratings given by each user
- $ightharpoonup R_{m,u}$  represents rating given to movie m by user u

$$R = \begin{array}{c} u_1 & u_2 & u_* \\ m_* & 2 & 5 & ? \\ m_2 & 3 & ? & 3 \\ m_4 & ? & 4 & 2 \\ m_5 & 2 & 5 & ? \end{array}$$

#### Goal

Predict  $R_{m_*,u_*}$ , the missing rating for user  $u_*$  and movie  $m_*$  using the existing ratings

### Method

▶ Filter rating matrix to just contain rows of movies rated by  $u_*$ , along with the movie of interest,  $m_*$ 

### Method (Cont.)

▶ Replace missing ratings by the average rating of the user

# Method (Cont.)

Adjust each rating by the average rating of the user

We will further refer to this adjusted rating matrix simply as R

Find the k movies most **similar** to  $m_*$  and **predict**  $(m_*, u_*)$  to be the weighted average of the ratings given by  $u_*$  to these k movies

### Scoring Similar Movies

#### Adjusted Cosine Similarity Metric

$$sim(m_1, m_2) = \frac{\sum_{u \in A} (R_{m_1, u} - \bar{R}_u) (R_{m_2, u} - \bar{R}_u)}{\sum_{u \in A} (R_{m_1, u} - \bar{R}_u)^2 \sum_{u \in A} (R_{m_2, u} - \bar{R}_u)^2}$$

Where  $A=U(m_1)\cap U(m_2)=\{u:\ u\ \text{has rated both}\ m_1\ \text{and}\ m_2\ \}$ 

# Calculating Similarity

lacktriangle Want to calculate the similarity between  $m_*$  and all

$$m_i \in M(u_*)$$

▶ Break into subproblems. For each  $m_i \in M(u_*)$ 

- 1. Calculate  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})(R_{m_i,u} \bar{R_u})$
- 2. Calculate  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})^2$
- 3. Calculate  $\sum_{u \in A} (R_{m_i,u} \bar{R_u})^2$

### Step 1

- ightharpoonup Calculate  $\sum_{u\in A}(R_{m_*,u}-ar{R_u})(R_{m_i,u}-ar{R_u})$ 
  - lacktriangle Multiply adjusted rating matrix by row vector of ratings for  $m_*$

#### Resulting Vector

$$\begin{array}{l} m_*, m_* \\ m_1, m_* \\ m_2, m_* \\ m_3, m_* \end{array} \begin{pmatrix} (2 - \bar{u_1})^2 + (5 - \bar{u_2})^2 \\ (3 - \bar{u_1})(2 - \bar{u_1}) \\ (4 - \bar{u_2})(5 - \bar{u_2}) \\ (3 - \bar{u_1})(2 - \bar{u_1}) \end{pmatrix}$$

- ► Each entry is the sum of products of adjusted ratings between  $m_*$  and  $m_i \in M(u_*)$
- ▶ The only movie ratings considered are from users that have rated both  $m_*$  and  $m_i$
- ► Each entry equal to  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})(R_{m_i,u} \bar{R_u})$

### Step 2

# Calculate $\sum_{u \in A} (R_{m_*,u} - \bar{R_u})^2$

- ► Transpose the adjusted rating matrix: R<sup>T</sup>
- ► Transform the vector of ratings  $m_*$  into a matrix of same shape as  $R^T$  such that each column represents  $m_*$ :  $M_*$
- ► Multiply  $M_*$  by  $R^T$  elementwise, then divide by  $R^T$  elementwise

$$M_{*} \quad \begin{array}{ccccc} & m_{*} & m_{*} & m_{*} & m_{*} \\ u_{1} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} \\ 5 - \bar{u_{2}} & 5 - \bar{u_{2}} & 5 - \bar{u_{2}} & 5 - \bar{u_{2}} \\ u_{*} & 0 & 0 & 0 \end{array} \right) *$$

$$R^{T} = \begin{pmatrix} u_{1} & m_{1} & m_{2} & m_{3} \\ u_{1} & 2 - \bar{u_{1}} & 3 - \bar{u_{1}} & 0 & 3 - \bar{u_{1}} \\ 5 - \bar{u_{2}} & 0 & 4 - \bar{u_{2}} & 0 \\ u_{3} & 0 & 3 - \bar{u_{*}} & 2 - \bar{u_{*}} & 4 - \bar{u_{*}} \end{pmatrix}$$

In the resulting matrix, entry  $(u_i, m_i)$  is the adjusted rating of user  $u_i$  on movie  $m_*$  such that  $u_i$  has rated both  $m_i$  and  $m_*$ , 0 otherwise.

Now multiply the row vector of ratings for movie m<sub>∗</sub> by the resulting matrix

$$\begin{pmatrix} m_* \sim m_* & m_* \sim m_1 & m_* \sim m_2 & m_* \sim m_3 \\ u_1 & (2 - \bar{u_1} & 5 - \bar{u_2} & 0) * & u_2 & (5 - \bar{u_2} & 0 & 5 - \bar{u_2} & 0 \\ u_* & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} (2-\bar{u_1})^2 + (5-\bar{u_2})^2 & (2-\bar{u_1})^2 & (5-\bar{u_2})^2 & (2-\bar{u_1})^2 \end{pmatrix}$$

#### Resulting Vector

- ► Each entry represents the sum of squares of the adjusted ratings of  $m_*$  given by users that have rated both  $m_*$  and  $m_i$
- ▶ Each entry equal to  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})^2$



### Step 3

# Calculate $\sum_{u \in A} (R_{m_i,u} - \bar{R_u})^2$

- ▶ Transform the vector of ratings  $m_*$  into a matrix of same shape as  $R^T$  such that each column represents  $m_*$ :  $M_*$
- $\triangleright$  Transpose the adjusted rating matrix:  $R^T$
- Multiply R<sup>T</sup> by M<sub>\*</sub> elementwise, then divide by M<sub>\*</sub> elementwise

$$R_T = \begin{array}{ccccc} & m_* & m_1 & m_2 & m_3 \\ u_1 & 2 & 0 & 3 - \bar{u_1} \\ u_2 & 5 - \bar{u_2} & 0 & 4 - \bar{u_2} & 0 \\ u_* & 0 & 3 - \bar{u_*} & 2 - \bar{u_*} & 4 - \bar{u_*} \end{array} \right) *$$

$$M_{*} \quad \begin{array}{ccccc} m_{*} & m_{*} & m_{*} & m_{*} \\ u_{1} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} & 2 - \bar{u_{1}} \\ 5 - \bar{u_{2}} & 5 - \bar{u_{2}} & 5 - \bar{u_{2}} & 5 - \bar{u_{2}} \\ u_{*} & 0 & 0 & 0 \end{array} \right) *$$

In the resulting matrix, entry  $(u_i, m_i)$  is the adjusted rating of user  $u_i$  on movie  $m_i$  such that  $u_i$  has rated both  $m_i$  and  $m_*$ , 0 otherwise.

Now multiply the rating matrix by the above matrix

#### Resulting Matrix

- ▶ Entries along the diagnal represent the sum of squares of the adjusted ratings of  $m_i$  given by users that have rated both  $m_*$  and  $m_i$
- lacktriangle Entries along the diagnal represent  $\sum_{u\in A}(R_{m_i,u}-ar{R}_u)^2$



#### Prediction

- Able to calculate  $sim(m_*, m_i)$  for all  $m_i \in M(u_*)$  in a few simple matrix operations
- ▶ For all  $m_i \in M(u_*)$ 
  - ▶ Step 1 provides us  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})(R_{m_i,u} \bar{R_u})$
  - ► Step 2 proviedes us  $\sum_{u \in A} (R_{m_*,u} \bar{R_u})^2$
  - Step 3 providees us  $\sum_{u \in A} (R_{m_i,u} \bar{R}_u)^2$  along the diagnals of a matrix
  - ► Combine to get  $sim(m_*, m_i) = \frac{\sum_{u \in A} (R_{m_*, u} \bar{R}_u)(R_{m_i, u} \bar{R}_u)}{\sum_{u \in A} (R_{m_*, u} \bar{R}_u)^2 \sum_{u \in A} (R_{m_i, u} \bar{R}_u)^2}$
- Now can choose the k movies most similar to  $m_*$  (that are rated by  $u_*$ ) to predict  $R_{m_*,u_*}$

#### Prediction

Predict rating of user u for movie m by taking weighted average of the ratings given by the k movies most similar to m that have rating from u.

$$P_{m,u} = \frac{\sum_{m' \in N_u^K(m)} R_{m',u} sim(m,m')}{\sum_{m' \in N_u^K(m)} |sim(m,m')|}$$

 $N_u^K(m) = \{m' : m' \text{ belongs to the } k \text{ most similar movies of } m$ and u has rated  $m'\}$