Outline

- Preprocessing
  - denoise
  - supplement
- Pairwise Training
  - Max-margin optimization problem
- Multifaceted Factorization Models
  - Extend the SVD++
- Context-aware Ensemble
  - Logistic Regression

Preprocess: Session analysis

- Negative : Positive = 92 : 8 ?
  - not all the negative ratings imply that the users rejected to follow the recommended items
- Eliminating these “omitted” records is necessary
  - These negative samples can not indicate users' interests

Preprocess: Session analysis

- Session slicing according to the time interval

$$\Delta t_s(u) = t_{s+1}(u) - t_s(u)$$

$$\tau(u) = \frac{1}{2} * \left(\tau_0 + \frac{\sum_{s=0}^{m} \Delta \ddot{t}_s(u)}{|\Delta \ddot{t}(u)|} \right)$$

When $\Delta t_s(u) < 3600(s)$, we denote it as $\Delta \ddot{t}_s(u)$.

If $\Delta t_s(u) > \tau(u)$, the training records on time $t_{s+1}(u)$ and $t_s(u)$ will be seperated to different subsets $\psi_k(u)$ and $\psi_{k+1}(u)$.

- Select the right samples from the right session:

$$0 < \frac{|\psi_k^+(u)|}{|\psi_k(u)|} \leq \epsilon = 0.86$$

$$\sigma_s - \sigma_- \leq \pi_- = 0$$

$$\sigma_+ - \sigma_s \leq \pi_+ = 3$$

Preprocess: Session analysis

- Training dataset after preprocessing
  - Negative: 67,955,449 -> 7,594,443 (11.2%)
  - Positive: 5,253,828 ->4,999,118
- Benefits
  - improve precision (0.0037)
  - reduce computational complexity

Preprocess: Supply positive samples

- Lack of positive samples
  - An ideal pairwise training requires a good balance between the number of negative and positive samples
- Choose the users
  - users who have a far smaller number of positive samples than negative samples
- Generate the positive samples
  - Figure out from social graphs

$$\underset{i \in S(u) \cap A(u)}{\operatorname{argmax}} \left(\xi_0 N_{at}(u,i) + \xi_1 N_{retweet}(u,i) + \xi_2 N_{comment}(u,i)\right)$$

The procedure of data preprocessing