

QUERY RECOMMENDATIONS FOR INTERACTIVE DATABASE EXPLORATION

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Motivation



- Scientific disciplines use relational DBMS for storage and retrieval of information
 - Biologists (e.g. UCSC Genome, BMRB)
 - Astronomers (e.g. Skyserver)
 - Chemists (e.g. PubChem)
- DBs are accessible online by users with diverse information needs
- Typical users do interactive exploration

Motivation (cont'd)



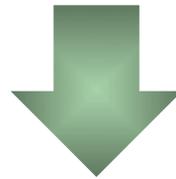
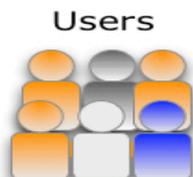
- Typical users are not SQL experts
- Scientific datasets increase in size
- Users may miss interesting information
 - They do not write the “right” query
 - They are not aware of all parts of the database

Our goal: Assist users in finding useful information

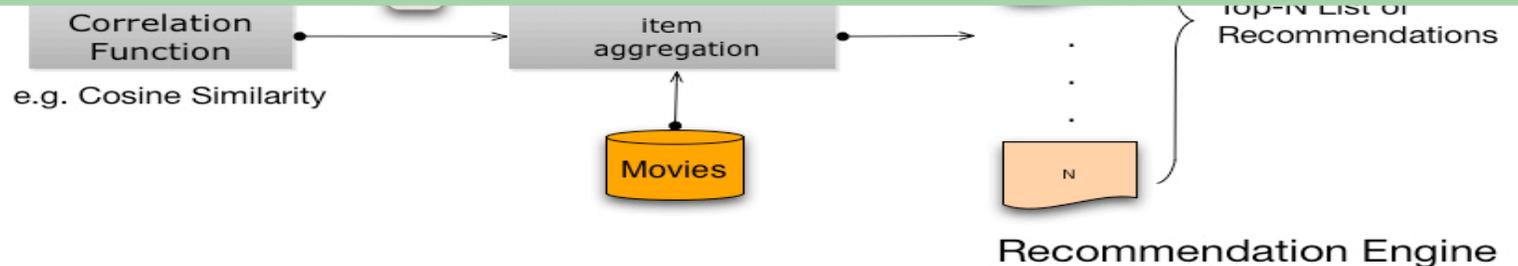
Web Collaborative Filtering

Example: Movie Recommendations

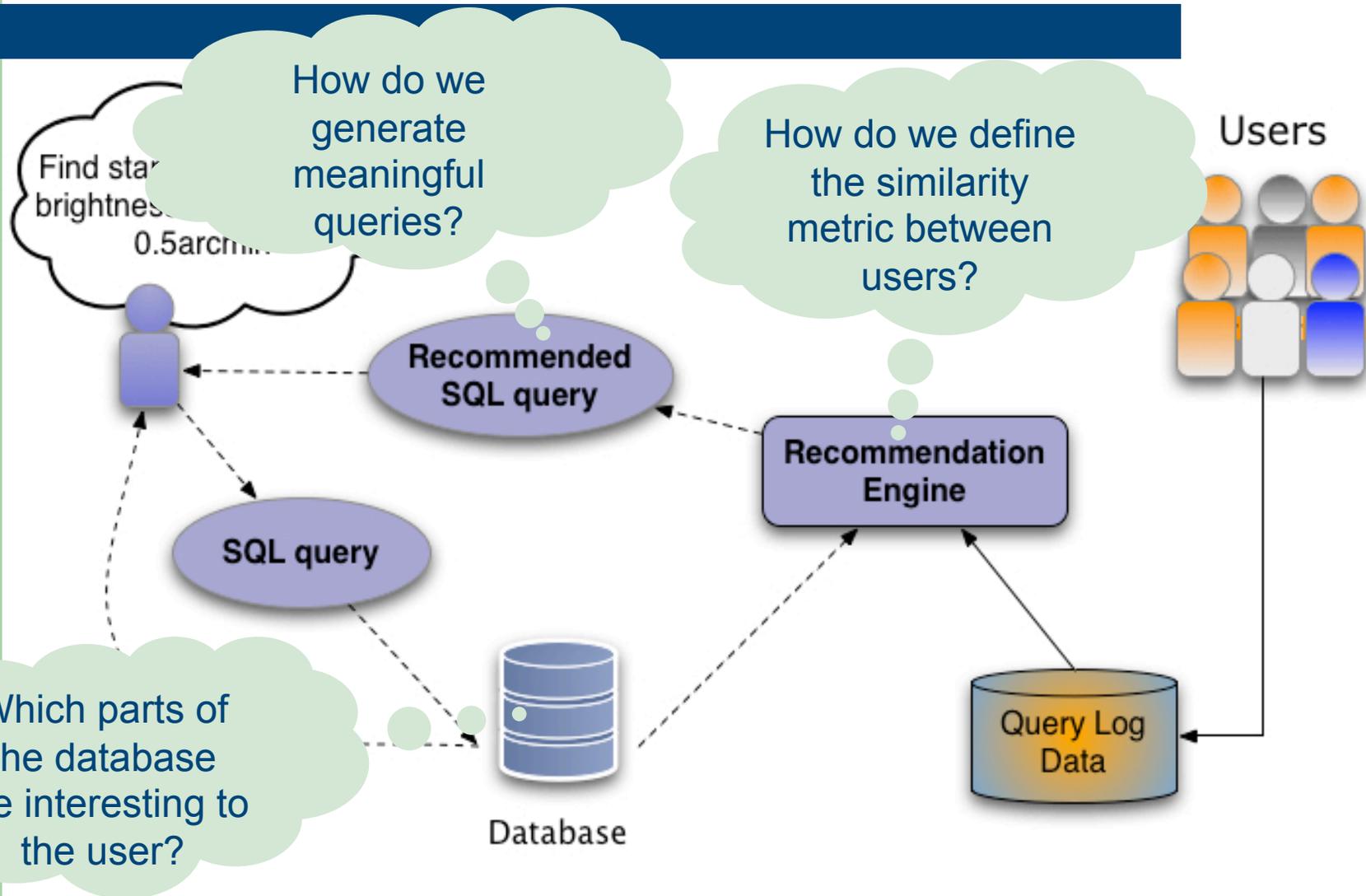
If Alice and Bob **both like movie X** and Alice **likes movie Y**
then
Bob is likely to be interested in **seeing movie Y**



If Alice and Bob **both query data X** and Alice **queries data Y**
then
Bob is likely to be interested in **querying data Y**



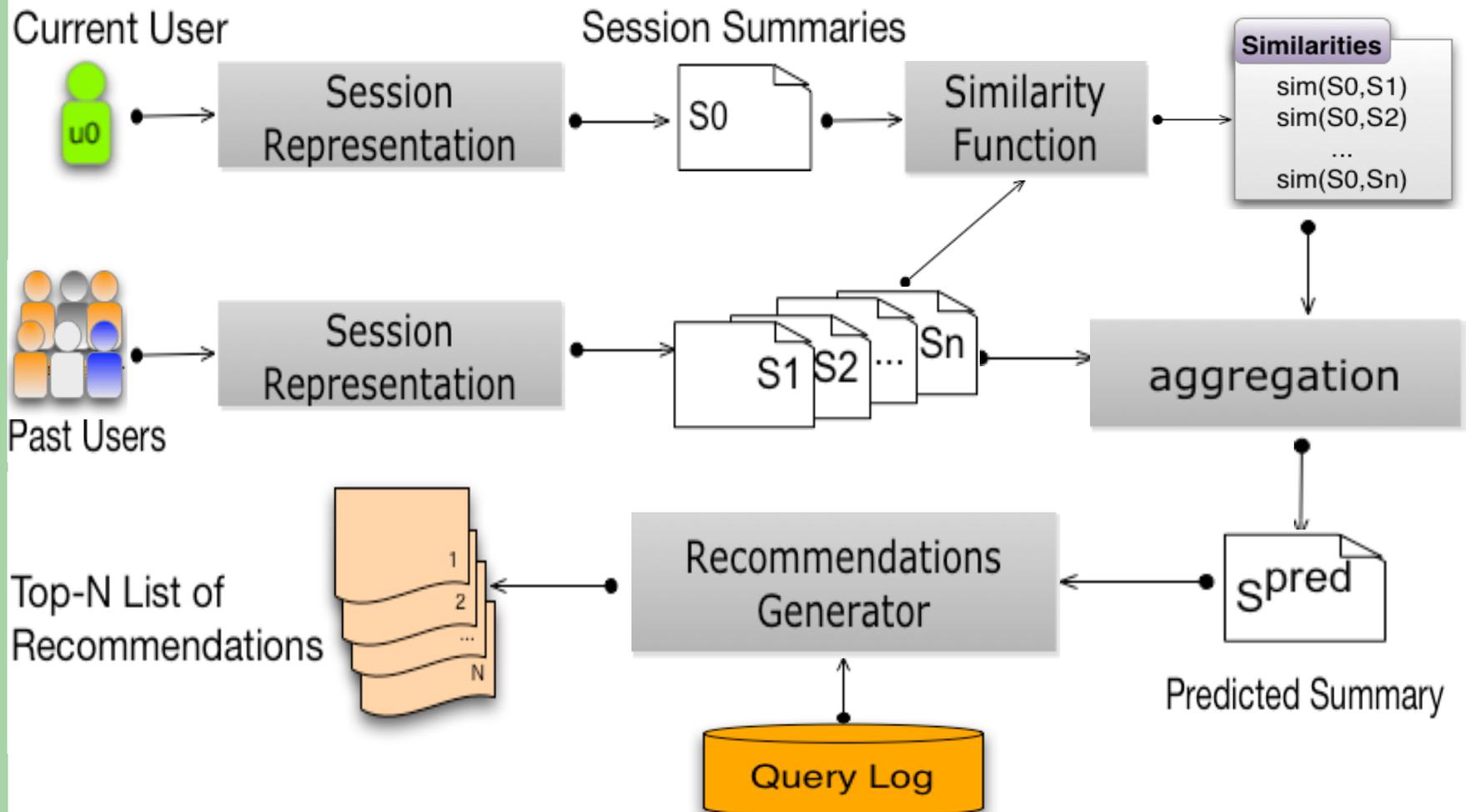
System Architecture



Roadmap

- Introduction
- QueRIE Recommendation Framework
- Experiments
- Conclusions

Conceptual Framework



Session Summaries

R	a	b
	y	3
	s	4
	w	3
	r	2

L	a	c
	y	9
	s	3
	s	5
	t	8



q1: $R \bowtie_{R.a=L.a} L$

q2: $\sigma_{R.b=4} (F \bowtie_{R.a=L.a} L)$

Binary Weighting Scheme

q1 = $\langle 1, 1, 0, 0, 1, 1, 1, 0 \rangle$

q2 = $\langle 0, 1, 0, 0, 0, 1, 1, 0 \rangle$

s0 = $\langle 1, 2, 0, 0, 1, 2, 2, 0 \rangle_2$

Result Weighting Scheme

q1 = $\langle 0.33, 0.33, 0, 0, 0.33, 0.33, 0.33, 0 \rangle$

q2 = $\langle 0, 0.50, 0, 0, 0, 0.50, 0.50, 0 \rangle$

s0 = $\langle 0.33, 0.83, 0, 0, 0.33, 0.83, 0.83, 0 \rangle$

Similarity Function

- Vector-space similarity functions can be used
 - Cosine Similarity

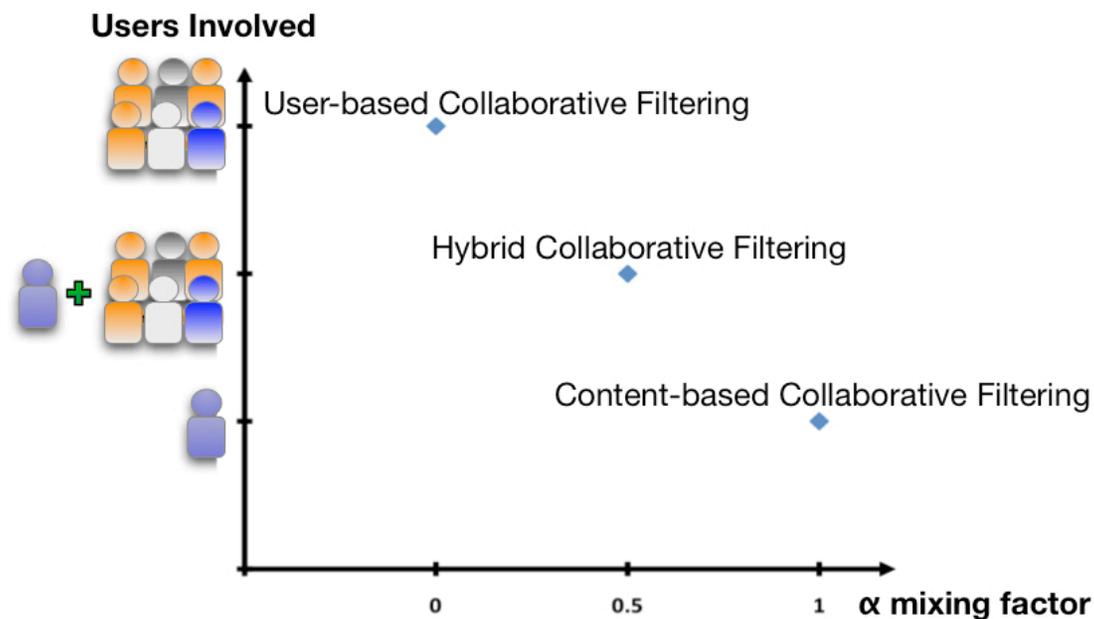
$$\text{sim}(uA, uB) = \frac{uA \bullet uB}{\|uA\| * \|uB\|}$$

- High similarity means that users are interested in the same parts of the database

Predicted Summary

$$u^{pred} = \alpha * u + (1 - \alpha) * \frac{\sum_{1 \leq i \leq h} sim(u, u_i) * u_i}{\sum_{1 \leq i \leq h} sim(u, u_i)}$$

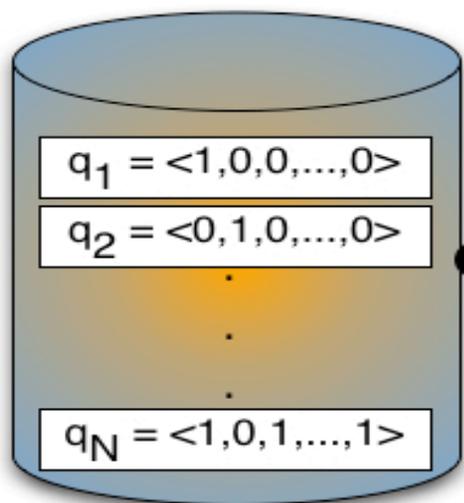
where α is the “mixing factor” $\alpha \in [0, 1]$



Generating Recommendations



Use queries of past users



Query Log Data

$$u^{\text{pred}} = \langle 1, 0, 0, \dots, 0 \rangle$$

Similarity Function
 (u^{pred}, q_i)

$$\text{rank}(q_1) = \text{sim}(u^{\text{pred}}, q_1)$$

$$\text{rank}(q_2) = \text{sim}(u^{\text{pred}}, q_2)$$

$$\text{rank}(q_N) = \text{sim}(u^{\text{pred}}, q_N)$$

Return Top-K
Queries

Roadmap

- Introduction
- **QueRIE Recommendation Framework**
- Experiments
- Conclusions

Experimental Setup

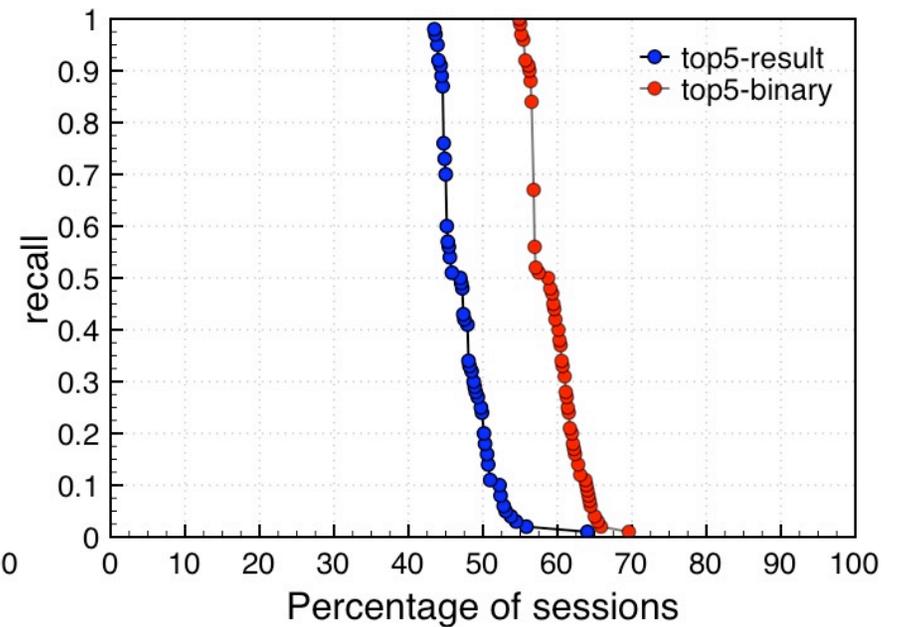
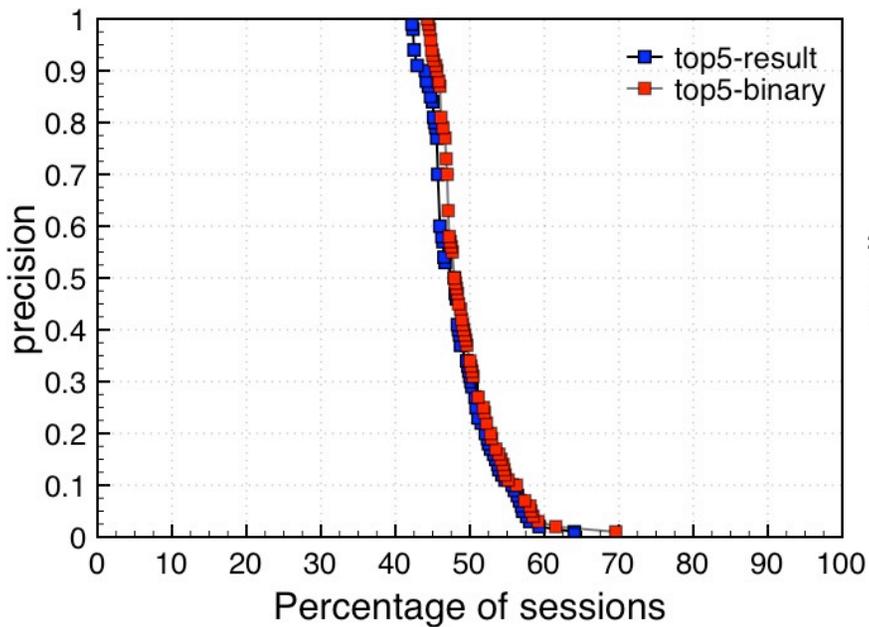
- SkyServer Dataset

Database Size	2.6TB
#Sessions	720
#Queries	6713
#Distinct Queries	4037
Avg. number of queries per session	9.3
Min. number of queries per session	3

- Evaluation Metrics: Precision and Recall

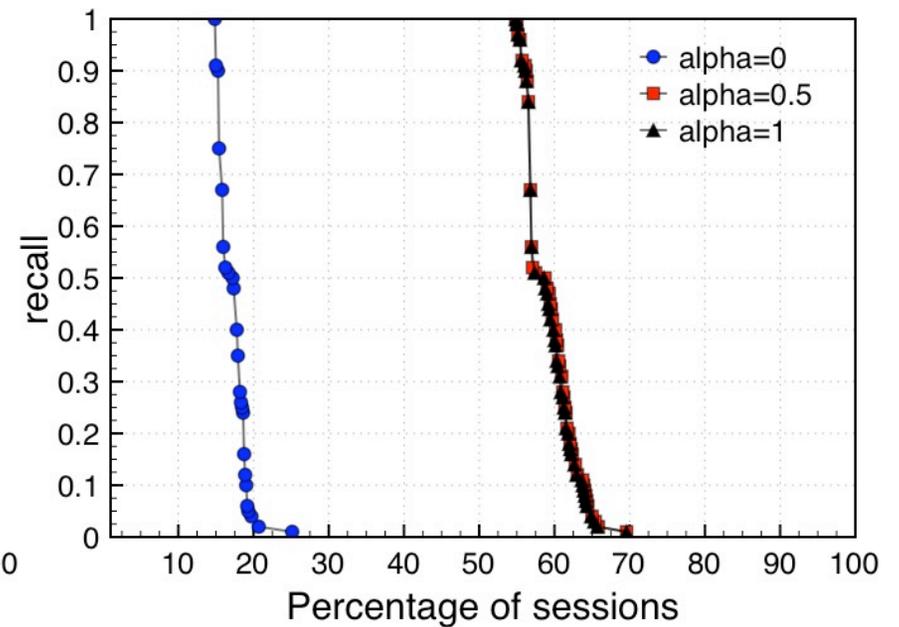
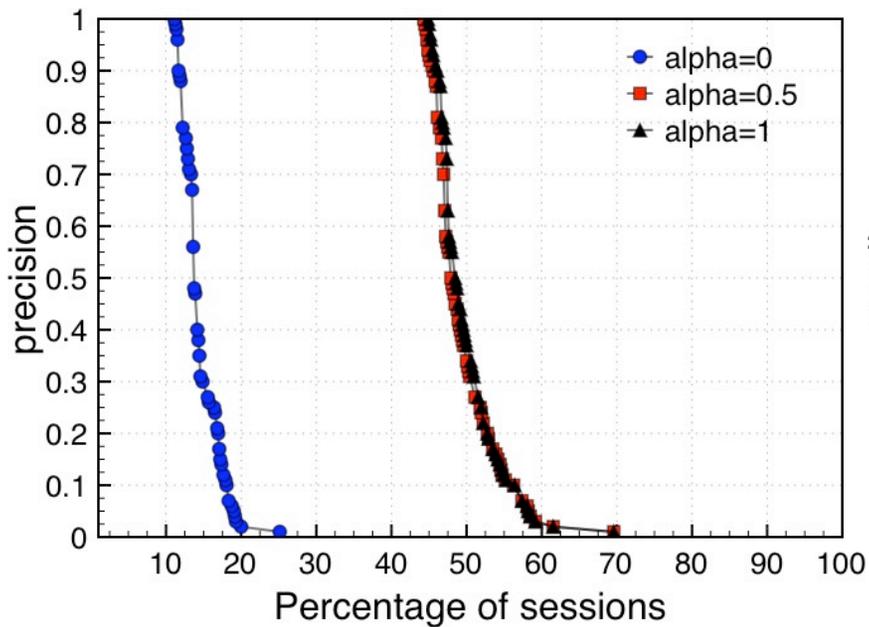
- **High precision:** most witnesses of the recommended query are witnesses in the actual query.
- **High Recall:** most witnesses of the actual query are witnesses in the recommended query.

Binary vs Result Weighting Schemes



Binary outperforms Result Weighting Scheme

Effect of mixing factor α



Hybrid Collaborative Filtering yields better results

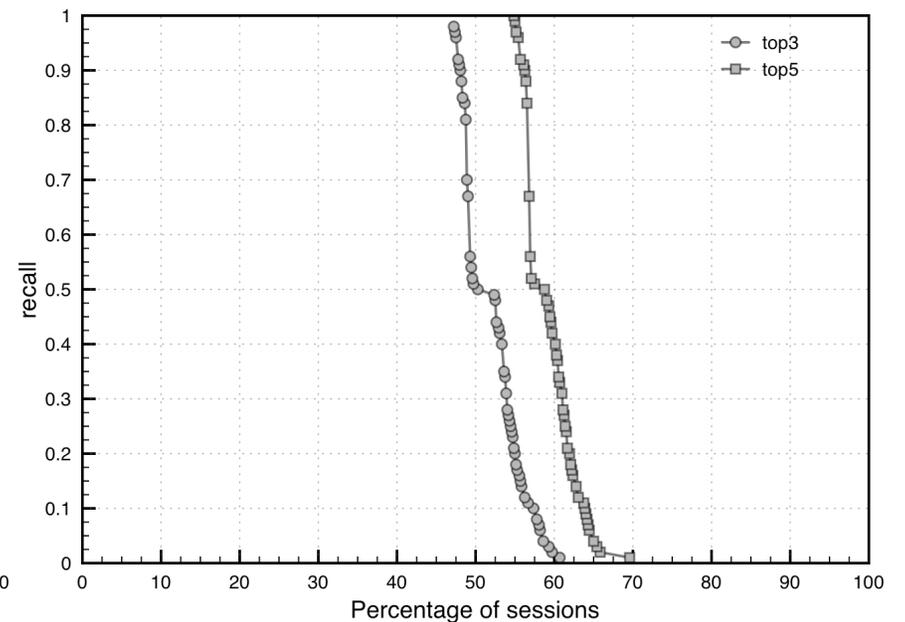
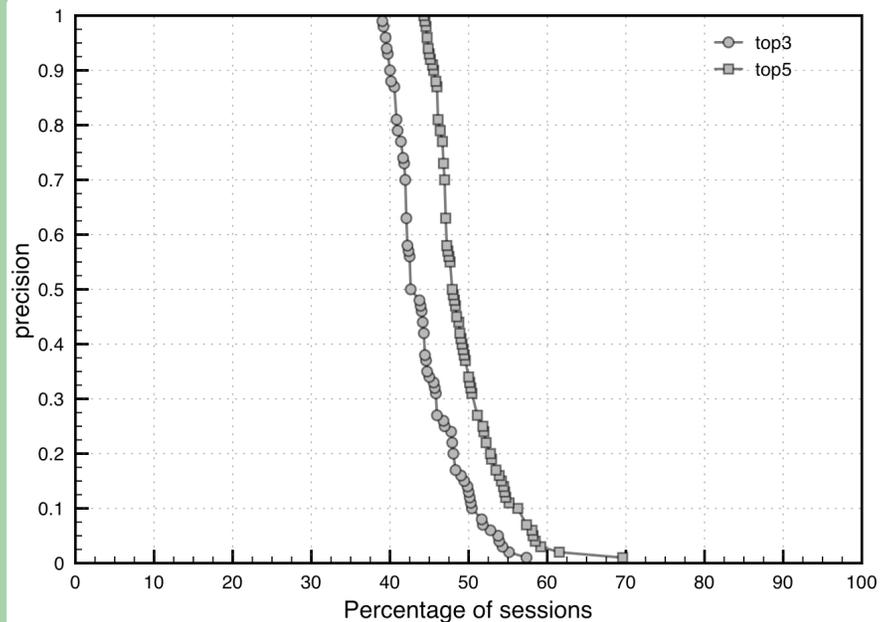
Conclusions

- Scientists need help in exploring databases
- Query recommendations can be an effective tool in guiding exploration
- Collaborative filtering provides a natural method to generate recommendations
- Experiments show promising results on real-world datasets
- Ongoing Work:
 - Performance improvement
 - Use of approximation techniques

Thank you



Top-3 vs Top-5 Binary Weights



The bigger recommendation set the higher accuracy