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**WEB OF DATA**

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# Future Research Issues: Recommending Tasks to Search Engine Users

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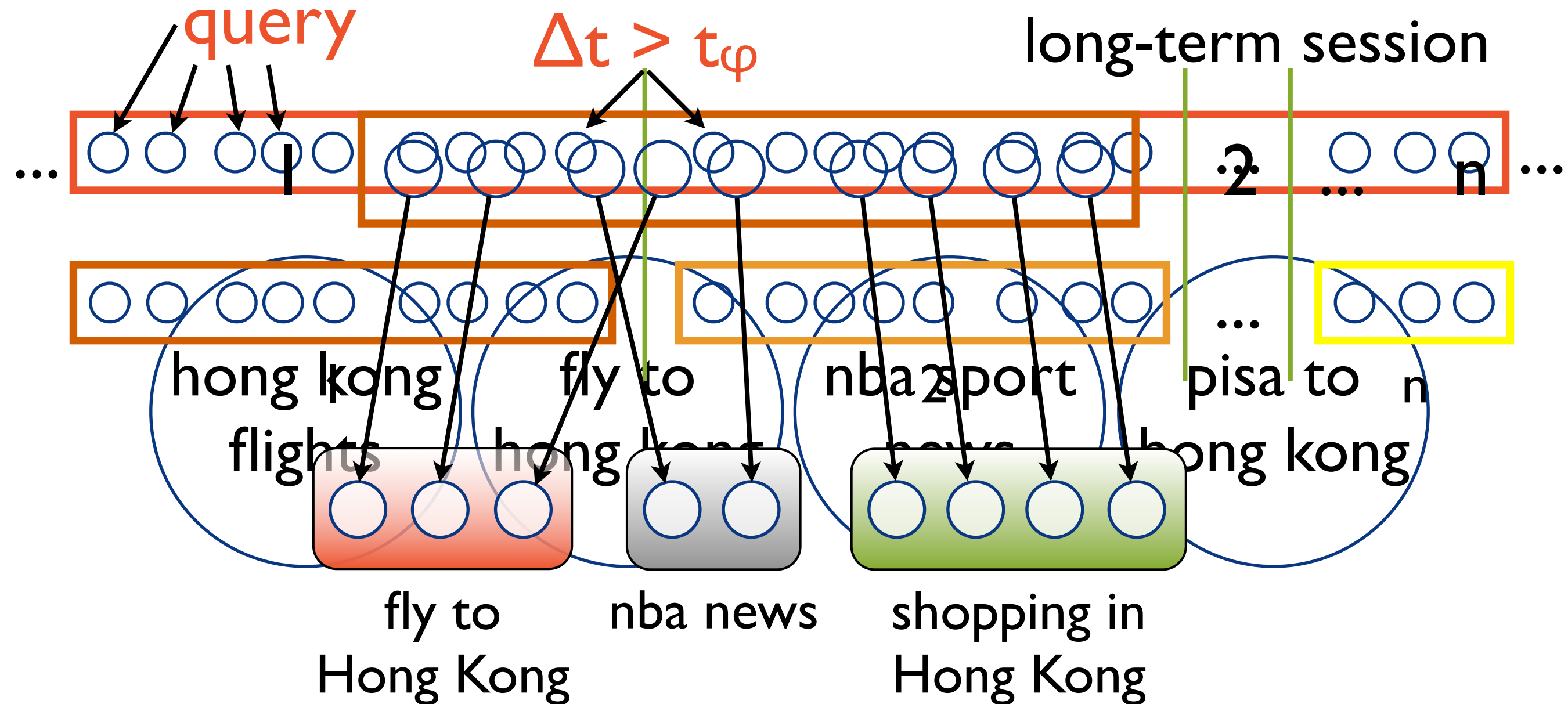
# Background

- From Web **task**
  - A “template” for representing any (atomic) activity that can be achieved by exploiting the information available on the Web, e.g., “find a recipe”, “book a flight”, “read news”, etc.
- To Web **mission**
  - Each single search task may subsume a complex task, namely a mission, that the user aims to accomplish throughout the SE.
- **Task/Query Recommendation**
  - Common query suggestions can be classified as **intra-task recommendations** (query rewriting, specialization, generalization, etc.)
  - We argue that people are also interested in task-oriented (query) suggestions, which can bring us to provide **inter-task recommendations**, i.e. related to another task in a given mission

# Example

- Example of inter-task suggestion
  - Alice starts interacting with her favorite SE by submitting the query “**new york hotel**”, i.e. a query belonging to a simple search task related to *the booking of a hotel room in New York*.
  - Current query suggestion mechanisms provide alternative related queries, by only focusing on the task behind this original single query (**intra-task query suggestions**), such as “cheap new york hotels”, “times square hotel”, “waldorf astoria”, etc.
  - Assume that you can recognize that the current Alice's task is included in a mission, including more tasks, concerned with “**planning a travel to New York**”
  - This means to recommend to Alice other tasks whose underpinning queries look like: “**mta subway**”, or “**broadway shows**”, or “**JFK airport shuttle**” (**inter-task query suggestion**)

# QC-htc: from long-term sessions to task-based sessions



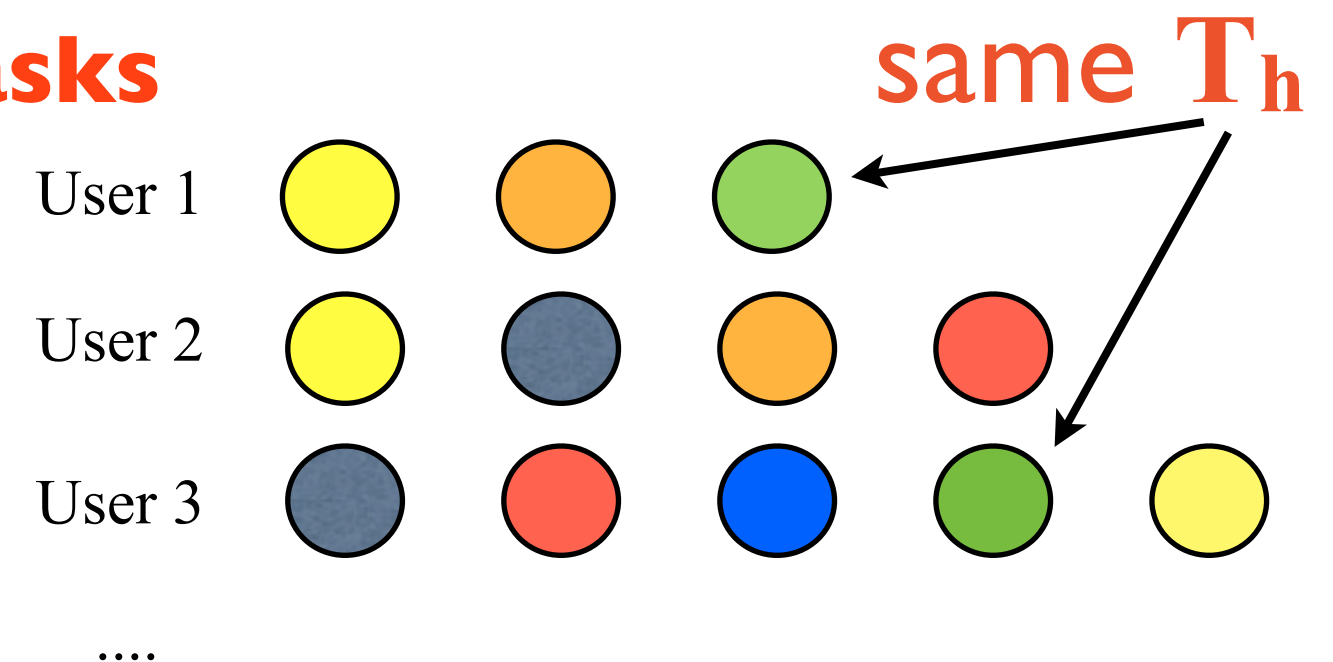
# Crowd-based Task Synthesis

- We already used an **unsupervised strategy** to **identify tasks** in the long-term sessions of the different users
- We still use an **unsupervised method** to **identify tasks common to many users**
- we further use a **cluster tool** to identify “similar” **tasks performed by distinct users** just identified by the previous method
- eventually replacing each task  $\theta_i^j \in \Theta_i$  in a long-term session of a user with a synthesized task  $T_h$



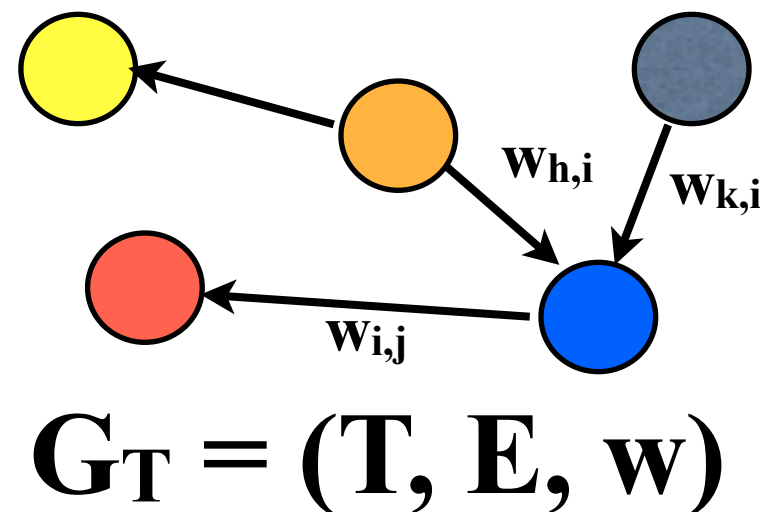
# Crowd-based Task Synthesis

- Each **synthesized task  $T_h$**  can be considered as a **representative** for an **aggregation** composed of **similar tasks**, performed by several **distinct users**
  - We can rewrite each task-oriented session in terms of the new tasks identifiers:  $T_h$   
where  $T_h = \{T_1, \dots, T_K\}$
- The various **long term sessions** thus become **sets/sequences of synthesized tasks**



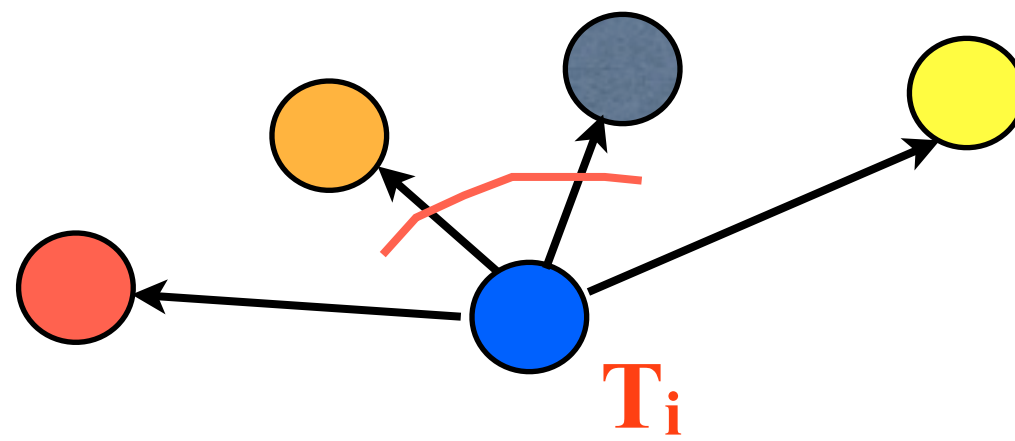
# Task-based Model Generation

- Produce a **Task Recommendation Model**
  - a **weighted directed graph**  $G_T = (T, E, w)$ , where the weighting function  $w(.)$  measures the “**inter-task relatedness**”
  - if they are **related**, they are probably part of the **same mission**



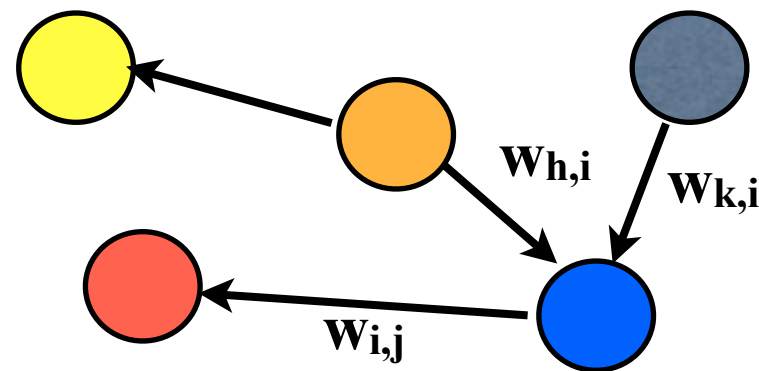
# Task-based Recommendation

- Generate a **Task-oriented Recommendations**
  - given a user who is interested in (has just performed) a **task  $T_i$**
  - retrieve from  **$G_T$**  the set  **$R_m(T_i)$** , which includes the  **$m$ -top related nodes/tasks** to  **$T_i$**
  - the graph nodes in  **$R_m(T_i)$**  are **directly connected** to node  **$T_i$**  and are the  **$m$**  ones labeled with the **highest weights**





# How to Generate the Model

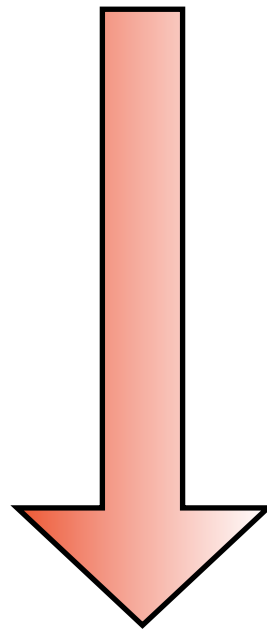


$$G_T = (T, E, w)$$

- Various methods to generate edges in  $G_T$  and the associated weights
- **Random-based (baseline)**: an edge for each pair, whose weights are uniform
- **Sequence-based**: the frequency of the pairs wrt a given support threshold, by considering the relative order in the original sequences
- **Association-Rule based (support)**: the frequency of the rule wrt a given support threshold. We do not consider the relative order in the original sequences to extract the rules
- **Association-Rule based (confidence)**: the confidence of the rules wrt a given confidence threshold. We do not consider the relative order in the original sequences to extract the rules

# Data Set: AOL 2006 Query Log

Original Data Set



Sample Data Set

- ✓ 3-months collection
- ✓ ~20M queries
- ✓ ~657K users

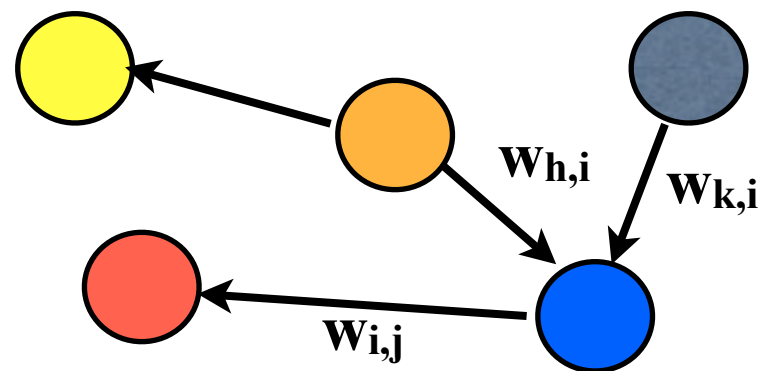


- ✓ Top-600 longest user sessions
- ✓ ~58K queries
- ✓ avg 14 queries per user/day
- ✓ two subsets A and B



- ✓ A : 500 user sessions (**training**)
- ✓ B : 100 user sessions (**test**)

# Experimental results



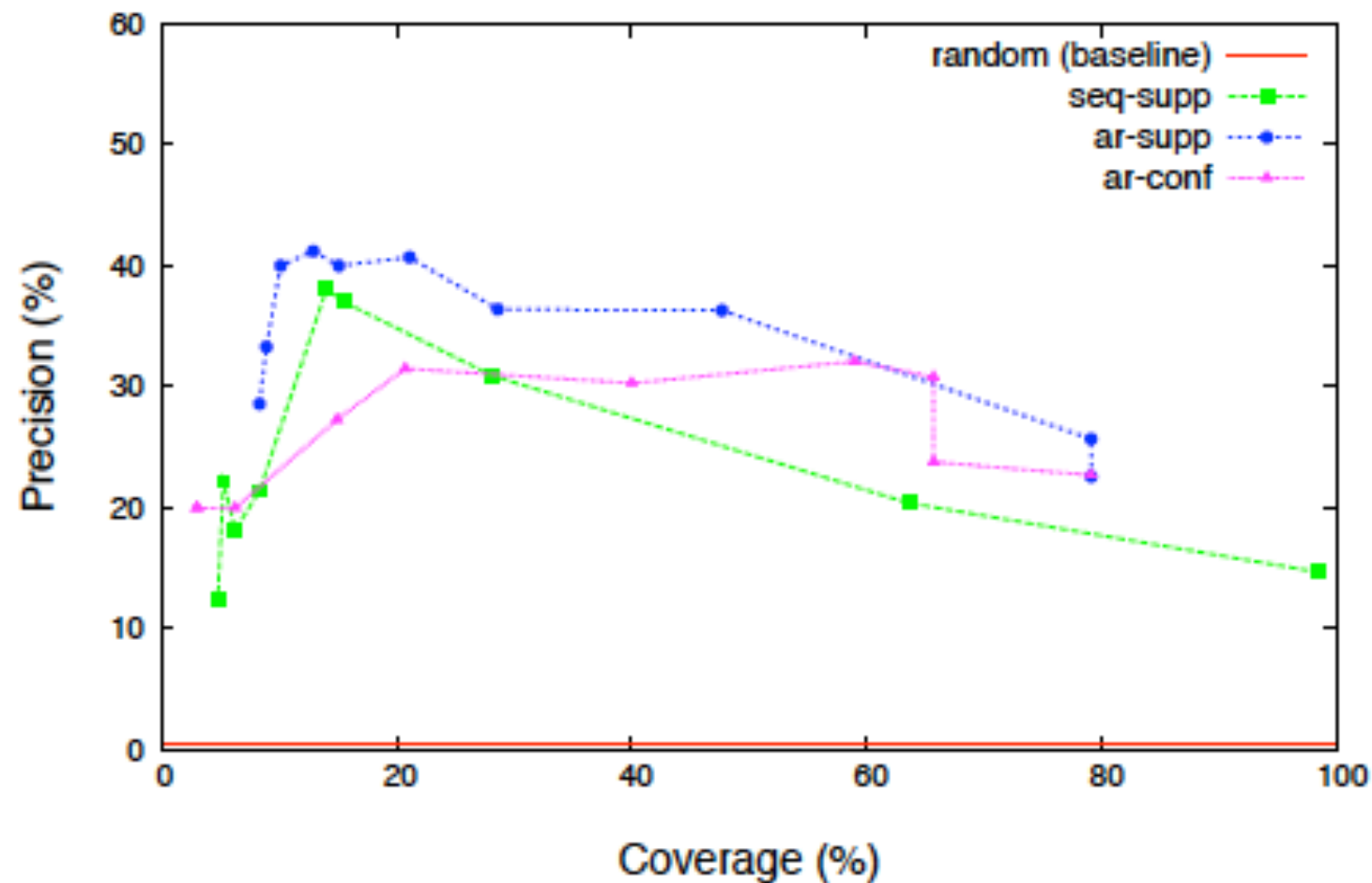
$$G_T = (T, E, w)$$

- We used the log subset **B** for evaluation (test query log)
  - we divided **each long term session in B** (with synthesized tasks) into a **1/3 prefix** and **2/3 suffix**
  - the prefix is used to retrieve from  $G_T$  the sets  $R_m(T_i)$
  - for each  $T_i$  belonging to the 1/3 prefix of each session in  $S$  in  $B$ , retrieve  $R_m(\{T_i \mid T_i \text{ in } S\})$
- We measured **precision** (proportion of suggestions that actually occur in the 2/3 suffix) and **coverage** (proportion of tasks in the 1/3 prefix that are able to provide at least one suggestion)
  - changing the weighting in each model, by tuning the corresponding parameters, modifies the coverage ...
  - we thus plot **precision vs coverage** to permit the different models to be fairly compared

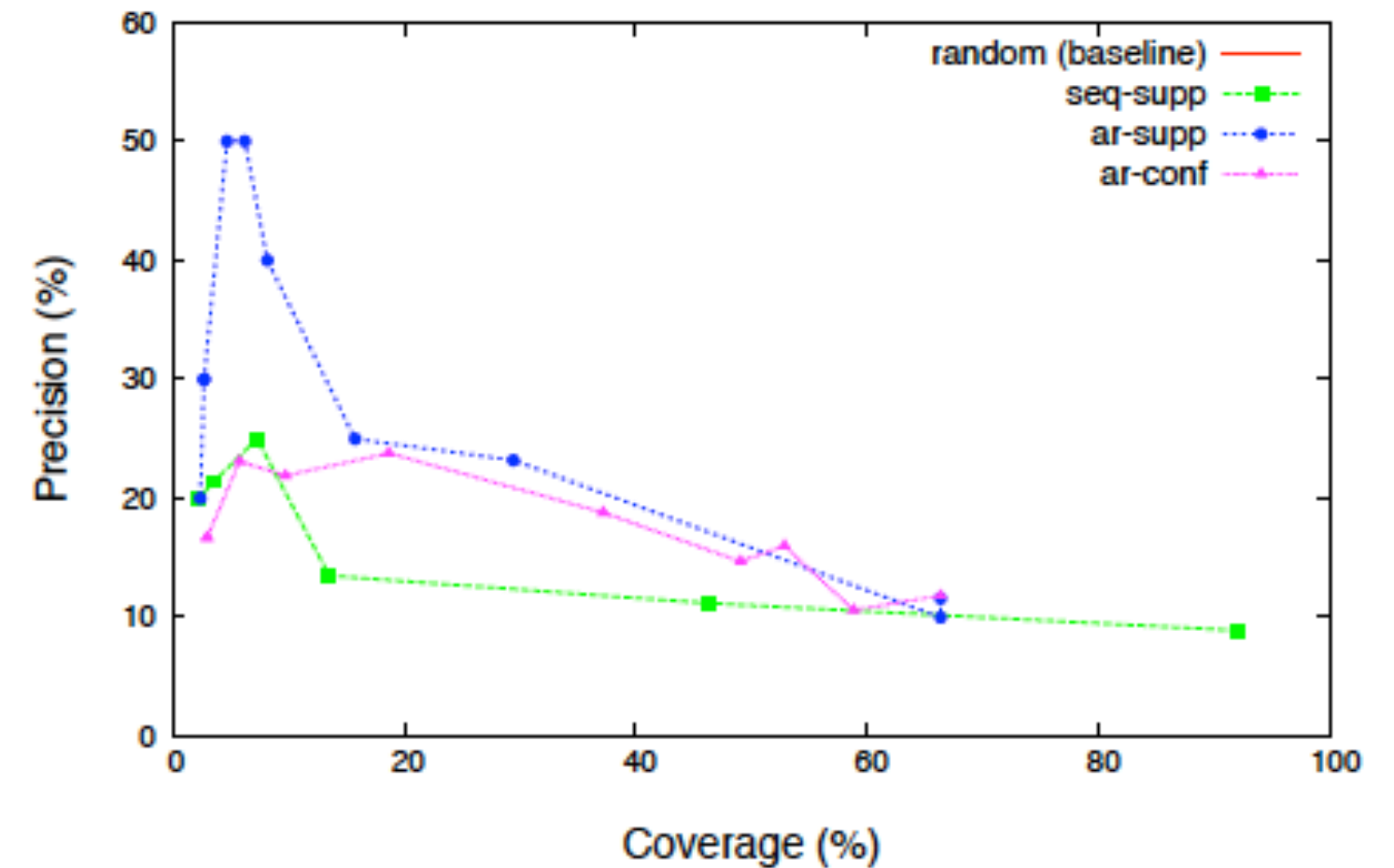
# Experimental results

## Recommendation Models

Top-1 suggestions in medium-size sessions



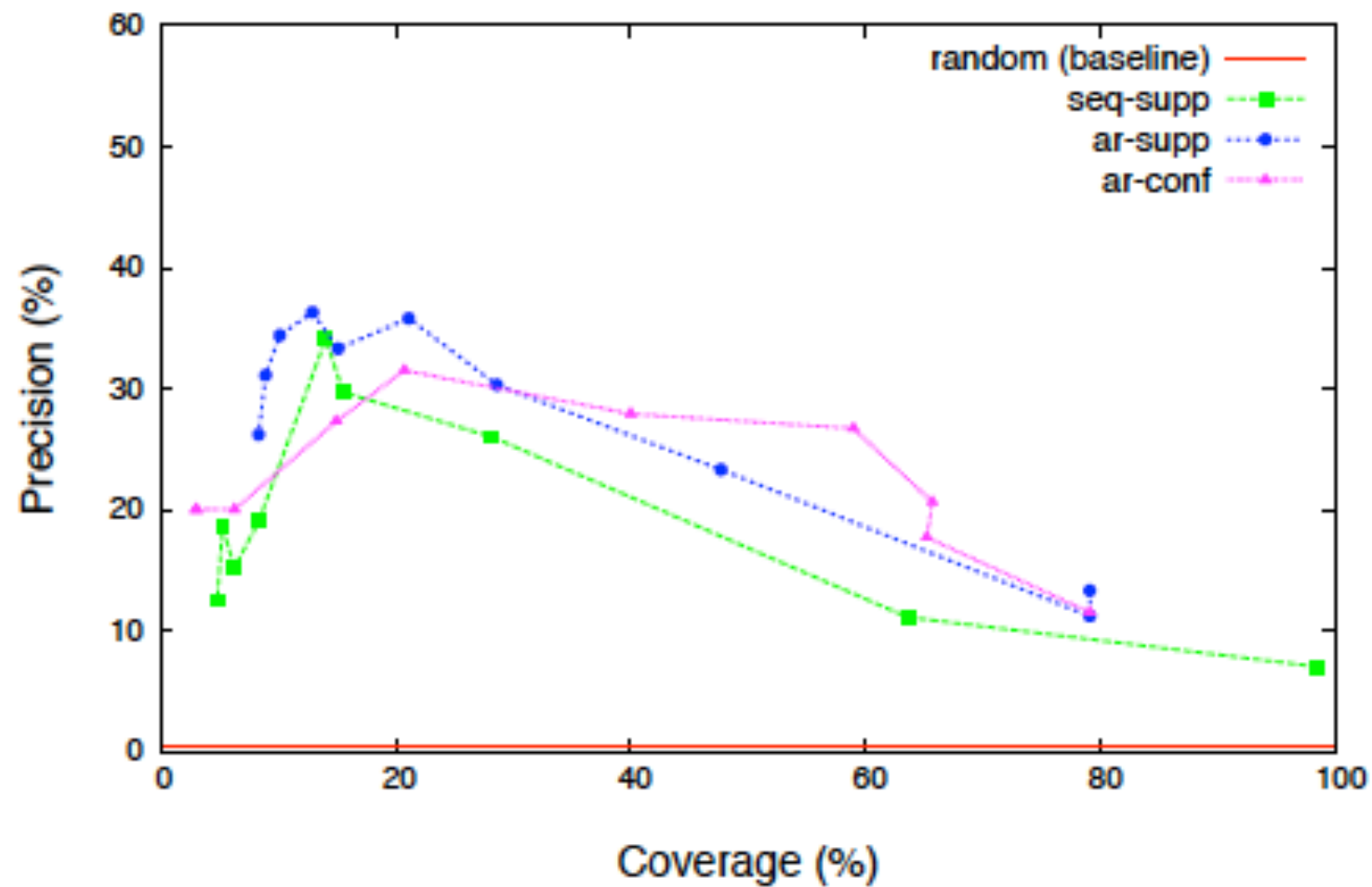
Top-1 suggestions in large-size sessions



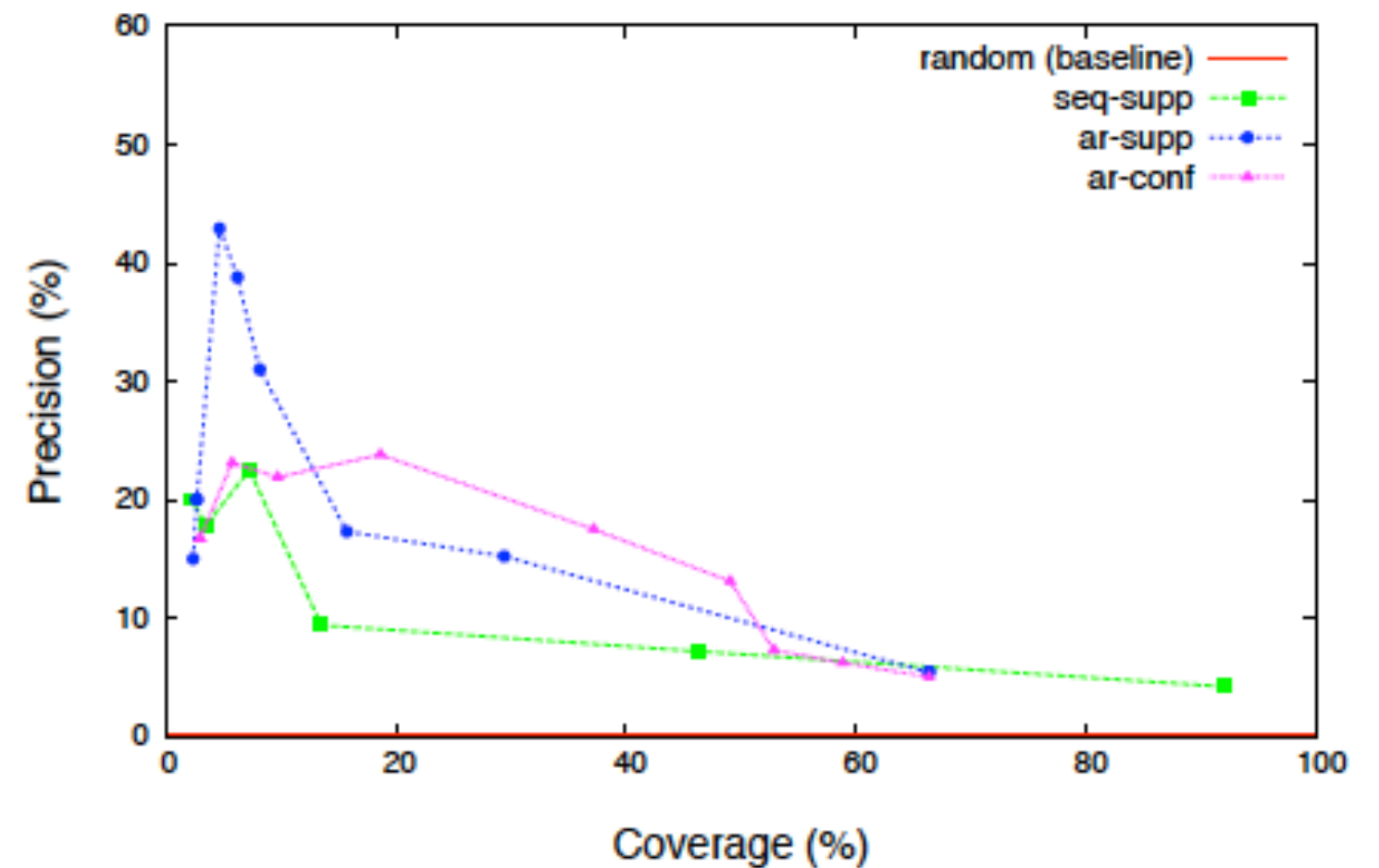
# Experimental results

## Recommendation Models

Top-3 suggestions in medium-size sessions



Top-3 suggestions in large-size sessions



# Anecdotal Evidence

| Performed Task/Query | Recommended Tasks             |
|----------------------|-------------------------------|
| Home Furnitures      | Home Gardening                |
| beach house          | <u>cottage garden</u>         |
| ...                  | <u>cottage garden roses</u>   |
| ...                  | <i>decor garden</i>           |
| beach house vanity   | <i>best garden blogs</i>      |
|                      | <i>vegetable garden ideas</i> |
|                      | <i>open garden</i>            |
|                      | antiques store*               |
|                      | book stores*                  |
| Kitchen Decor        | Kitchen Supplies              |
| dining room          | <u>stoves</u>                 |
| ...                  | <i>country stoves</i>         |
| ...                  | <i>country cooking stoves</i> |
|                      | country music gossips*        |
|                      | canyon country parkway*       |

actually  
performed  
queries

Table 1: Recommended and performed tasks.



# Anecdotal Evidence

| Performed Task/Query | Recommended Tasks             |
|----------------------|-------------------------------|
| Child Entertainment  | Child Games                   |
|                      | <i>baby shower games</i>      |
| fables               | <i>baby horse</i>             |
| ...                  | Child Clothes                 |
| ...                  | <i>baby gap</i>               |
| baby fables          | Child Health                  |
|                      | <i>baby emotional disease</i> |
|                      | cuddling couch picture*       |
|                      | husband became parents boy*   |
| University           | University Sports             |
| university           | <i>university sports</i>      |
| ...                  | <i>university basketball</i>  |
| ...                  | University Information        |
| duke university      | <i>university tuition</i>     |

Table 2: Recommended and *surprising* tasks.

# Questions?

