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### Outline

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  - Dominant features
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### Introduction

- The user types "mo", the search engine suggests: "morning", "mother", "monastery", "moon", ...
- Two main goals:
  - Guess the intended query after just a few keystrokes
  - Rank the query suggestions
- Make the suggestions specific to the user

## Query Ranking Model

- x is the prefix of a query
- $\{F_1, F_2, \cdots, F_N\}$  set of user features
- Q set of candidate queries
- $oldsymbol{o} \forall q \in Q, \ p(q|F_1, F_2, \cdots, F_N)$
- $\circ$  score(q) =  $p(q) \cdot p(q|F_1) \cdot \cdots \cdot p(q|F_N)$



- Query logs from an online news site
- 380000 searches
- 250000 unique users (112000 registered)
  - onot reg: city, country, time
  - o reg: age, gender, industry, job, income

### Experiments

- Chronologically first 300000 are the training set
- We evaluate:
  - How often the correct query is in the top 3 suggestions
  - Which features are most helpful in ranking
  - How many keystrokes the user saves
- Baseline: rank the queries by most frequent first

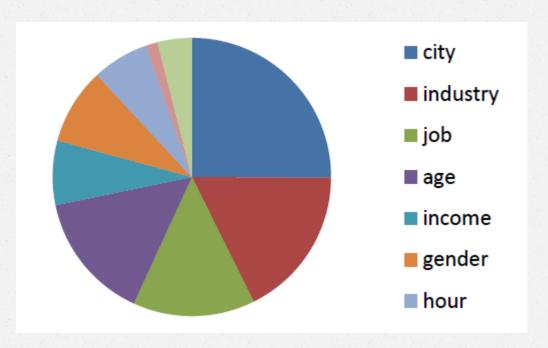




RANK OF SUGGESTION (%)					
FEATURES	Prefix	R1	R2	R3	TOP3
UNREG.	1	11.50	5.08	2.47	19.06
	2	23.44	10.12	5.39	38.96
	3	43.96	13.53	7.56	65.05
	4	56.87	14.59	5.30	76.78
REGISTERED	1	12.07	6.01	2.16	20.25
	2	25.56	9.86	4.90	40.33
	3	44.44	13.79	6.54	64.79
	4	57.71	13.44	4.86	76.02
BASELINE	1	9.06	6.19	2.34	17.60
	2	23.07	9.81	5.92	38.83
	3	42.94	14.55	6.59	64.05
	4	57.49	13.48	6.19	77.17

### Dominant Features

 $\underline{\text{Def}}: score(q) = p(q) \cdot p(q|F_1) \cdot \dots \cdot p(q|F_N)$   $\operatorname{argmax}_i[p(q|F_i)]$ 





- The user types the first character of q
  - The auto-completion algorithm is run
  - o If the correct answer is in the top 3 we assume success and we saved length(q) 1 keystrokes
  - Otherwise the user types one more character and we go again
- If after 4 characters the correct query is still not in the top 3, we assume failure
- On average 5.7 keystrokes are saved (40% less typing



- Using another approach for generating the candidate set Q, we can extend to query suggestions
- <u>Example</u>: A user has in the past searched for crossword, we suggest related queries: kenken, crossword puzzle, today's crossword, puzzle, daily crosswords, etc



- We developed a simple personalized queryautocompletion model
- 75% of the time we suggest the correct query
- The user has to type 40% less
- Personalization is useful especially if the prefix is short (1 or 2 characters)
- In the future we plan to extend the work to query suggestion and include the user's search history as a feature

# Thank you!

