Investigating the Learning Process in Job Search: A Longitudinal Study

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Background: Search as Learning

- Users learn new knowledge and search strategies during search
 - Exploratory search
 - Completing complex search tasks
- Understanding the learning process in search tasks can enable better support for complex search tasks
- The learning process in search has been studied in IIR
 - Kuhlthau's Information Search Process (ISP) model (Kuhlthau, 1993)
 - Vakkari's 3-stage model for the learning process in search (Vakkari, 2001)(Vakkari, 2016)

Restructuring • Conceptual changes







Assimilation

 Instantiating conceptual structure

Background: Job Search

- Job seeking: a complex search task
- Search, evaluate, and compare available jobs
- The process lasts for days or weeks
- The collection of available jobs may change during the process
- Job seekers learn over time about available jobs and how to search for a job
- Investigating the how their search behavior changes during the job seeking process is essential for building a better job search engine

Motivation

- Revisit the learning process in search for job search tasks
 - The information need is relatively stable
 - We can investigate behavioral changes within a single task by analyzing job seeker's queries within a time window
- Conduct a log-based, longitudinal study for job search
- Discuss how the empirical findings support or contradict existing models of the learning process in search

Research Questions

- **RQ1**: How is job seekers' search behavior characterized?
- **RQ2:** How does the behavior change over time?
- **RQ3:** Does information consumption (result clicks) and response behavior (application lodging) change over time?

Methodology

- Data collection
 - ~125,000 queries from thousands of randomly sampled users of SEEK
 - No personally identifiable information was available or used in experiments
 - Contains:
 - Queries
 - # clicks, click-through rate, application rate
 - Filters on job classification, work type, job location, salary range, and posting time
- Partition the query log into job search tasks
- Assumption: it is unlikely that a job seeker will be involved in multiple job search tasks at the same time
- Used a 14-day gap in logged actions as boundary of distinct search tasks
- Investigate the changes of search behavior over time
 - Divide the process of each search task into three stages of equal time
 - Compare users' querying, click, and application behavior in three stages
 - Conduct one-way ANOVA to test whether the behavioral measures differ between stages

Characterizing Job Search Tasks

- The statistics of job search tasks
- Job search is complex
 - lasts for 8.6 days
 - submits 11.1 queries
 - 45.75% of the tasks exceed one day in duration

All search tasks		longer than one day		
11,267		5,159		
8.56	(16.04)	17.51	(20.36)	
11.14	(35.29)	20.75	(50.39)	
15.66	(47.02)	29.13	(66.83)	
268.63	(837.17)	503.35	(1191.65)	
	8.56 11.14 15.66	11,267 8.56 (16.04) 11.14 (35.29) 15.66 (47.02)	All search tasks longer the 11,267 8.56 (16.04) 17.51 11.14 (35.29) 20.75 15.66 (47.02) 29.13	

Search tasks last

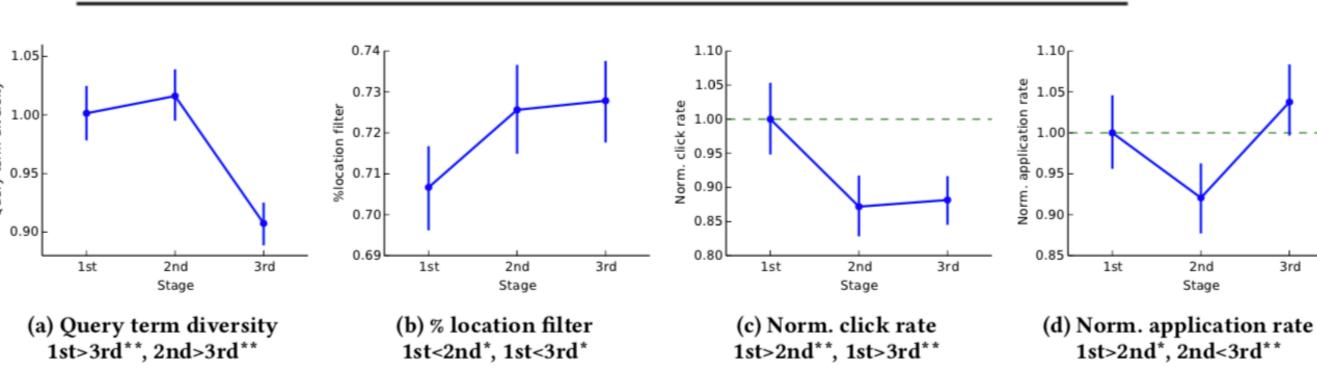
- The statistics of querying behavior
 - Job seekers tend to use short queries
 - The query diversity ,#{unique terms},
 - The location filter is widely used in job search

	All search	55 (1.58) 1.69 (2.11)			
# search tasks	11,	267	5,159		
Length in characters Length in terms Query term diversity % classification filter % location filter % date filter % work type filter % salary filter	11.95 1.65 0.923 0.239 0.704 0.105 0.099 0.105			`	

Changes in User Behavior over Time

The statistics of search behavior in three stages (n = 5, $159 \times 3 = 15$, 477)

	1 st stage	2 nd stage	3 rd stage	F-value	p-value	
Querying Behavior						
Length in characters	12.263	12.304	12.325	0.03	0.974	
Length in terms	1.680	1.693	1.692	0.05	0.953	
Query term diversity	1.001	1.016	0.908	33.21	< 0.001	
%classification filter	0.240	0.237	0.234	0.38	0.686	
%location filter	0.707	0.726	0.728	4.87	0.008	
%date filter	0.105	0.110	0.109	0.42	0.658	
%work type filter	0.098	0.110	0.109	2.78	0.062	
%salary filter	0.107	0.117	0.116	1.67	0.189	
Clicking and Application Behavior						
Norm. click rate	1.000	0.872	0.882	10.44	< 0.001	
Norm. application rate	1.000	0.921	1.038	7.30	0.001	



^{*/**} indicate the difference is significant at $\alpha = 0.05/0.01$ level with a post hoc comparison using the Tukey's HSD test.

Discussion

- Vakkari's model predicts that at the last assimilation stage, users will:
- Have a clear usefulness criteria
- Have a lower click rate
- and a higher use/selection ratio (i.e. a higher application rate)
- Supported by our findings
- Use more unique query terms
- Increase the specificity of terms
- Use more synonyms

Not supported by our findings

- Extends existing model with a fourth stage: monitoring
- After learning knowledge of how to find jobs in a particular field, the job seeker monitors the updates of new jobs with relative static queries
- Other search tasks may also have such a monitoring stage
- The search system for a dynamic collection should detect and provide better support for this stage

Conclusions

- **RQ1**: How is job seekers' search behavior characterized?
- Job search is a complex search task that requires multiple queries over a long period of time to complete
- Job seekers like to use short queries with different filters
- **RQ2:** How does the behavior change over time?
- Job seekers tend to use more filters and less diverse query terms over time
- RQ3: Does information consumption (result clicks) and response behavior (application lodging) change over time?
 - The click rate decreases while the application rate increases over time