Characterizing and Supporting Cross-Device Search Tasks

Yu Wang¹, Xiao Huang², Ryen White³

¹ Emory University, yuwang@emory.edu

- ² Microsoft Bing, xiaohua@microsoft.com
- ³ Microsoft Research, ryenw@microsoft.com

Motivation

- Multi-device usage is becoming common
- People can search anytime, anywhere

Smartphone



- We usually study one device at a time (primarily desktop)
- Here we examine cross-device searching ...

Search Activity over a Single Day



- Analyzing *desktop-only* one could observe some events
- Richer picture of behavior by considering multi-device use
- Focus on *switches* (transitions) between devices
- Our belief: Engine can help on post-switch device if it can anticipate post-switch task resumption

Our Definition of Device Switching

• Search sessions with 30 minute inactivity timeout



Challenges and Opportunities

- Challenges:
 - Switching is expensive for a user
 - User has to remember what has been searched on task
 - Re-typing is time consuming, sometimes very difficult if in motion

• Opportunities:

- How a search engine could help with switching
 - Predict cross-device task continuation
 - Use prediction to capitalize on between device downtime



- Why not just *always* use downtime?
 - Additional actions (e.g., run queries, crowdsourced answers) expensive
 - Only want to do it when confident that user will resume

Analyzing Cross-Device Search

- Subset of users who are signed in to Microsoft Bing
- Users who used both devices during one month period

Number of Days		31	
Number of Users		39,081	
	Desktop	709,610	
Number of Sessions	Mobile	301,028	
	Total	1,010,638	
	Desktop	3,023,582	
Number of Queries	Mobile	667,091	
	Total	3,690,673	
Number of Switches		158,324	

Transitions (within 6 hours)

	Desktop-to-Mobile Mobile-to-Deskto		
Same-query switch	10,480 (6.6%)	5,282 (3.3%)	
Different-query switch	69,441 (43.9%)	73,121 (46.2%)	

- Volume of Desktop-to-Mobile ≈ Mobile-to-Desktop
- Different query switches >> Same query switches
- 2x same-query D-to-M as M-to-D
 - Many desktop search tasks will be carried over to mobile.
 - More support for D-to-M is needed.

What do cross-device tasks look like?

Characterizing Cross-Device Search

- Focus on Desktop-to-Mobile
- Temporal:
 - When do users switch
 - How long elapses between pre- and post-switch
- Topical:
 - Topic shifts during switches
- Geospatial:
 - Physical location before and after device switch

Temporal

 Time between pre- and post-switch queries as a function of hour in the day, of pre-switch query



- Most switches initiated late afternoon, end early evening
- Gap between pre- and post-switch queries varies with time:
 - <u>Short gaps</u> are more likely late evening and early morning
 - Long gaps are more likely during work hours (9-6)
- Engine can use temporal features to predict task resumption

Topical

- Query topics estimated from Bing runtime classifiers
- Sustainability = Pr(topic post-switch | topic pre-switch)
- Lift over background (sustainability / overall topic popularity):

Most likely to			
be resumed			
post-switch,			
if pre-switch			

Least likely to be resumed post-switch, if pre-switch

Category	Lift		
Clothes and Shoes	82.240		
Weather	77.528		
Books	66.180		
Video Games	42.478		
Health	39.608		
Recipes	31.827		
Celebrities	30.342		
Restaurant	19.576		
Movie	18.305		
Sports	15.595		
Music	14.429		
Travel	7.805		
Location	5.364		
Image	3.467		
Local	3.117		
Navigational	1.710		

Purchasing (need to try on clothes/shoes) Weather forecasts Entertainment while mobile



General interest on mobile, popular irrespective of pre-switch topic

Geospatial

- Examine physical location before and after switch
- Caveat: Uses RevIP and cellphone provider geocoding
 - At town/city level, not GPS based
- 67% stay within same city, 33% move to different city
- Movement during post-switch session:

Single query	Multiple query session		
session	Moving session	Stationary session	
60.6%	5.3%	34.2%	

Must be moving quickly given how location is estimated

Can we predict cross-device tasks?

Predicting Cross-Device Search Tasks

 Predict whether the user will resume the task in the pre-switch session on another device



Different types of support offered at each (more later)

Prediction Experiment

Different features to predict cross-device task resumption



Prediction Experiment

- MART classifier
- Features
 - Behavioral, Topical, Temporal, Geospatial
- Cross-validation at the user level
- Training data
 - Automatic: Machine learned model using query similarity features
 - 17k judgments, 9.5% of the labeled switches were on same task
 - Human labeled: 5 judges reviewing pre- and post-switch behavior
 - 800 judgments, **15%** of the labeled switches were on same task
- Dropped nav. queries (personal freq > 5, global freq > 10)
 - Represent long-term *interests*, not search tasks

Feature Dictionary

Name	Description		
Features from Search History			
NumOfDesktopQuery ^B	Number of queries issued on desktop		
NumOfMobileQuery	Number of queries issued on mobile		
PercentageDesktopQuery ^B	Percentage of queries issued on desktop		
PercentageMobileQuery	Percentage of queries issued on mobile		
PercentageDesktopTime ^B	Percentage of searching time on desktop		
PercentageMobileTime	Percentage of searching time on mobile		
NumOfSession ^B	Number of search sessions		
NumOfContiguousSwitch	Number of contiguous cross-device search tasks		
NumOfRelevantCrossDevice	Number of search tasks appearing on both devices		
EntropyAvg	Average device entropy of same-task queries		
EntropySum	Total device entropy of same-task queries		
EntropyWeighted	Weighted device entropy of same-task queries		
Features from Pre-switch S	Sessions		
NumOfQuery ^B	Number of queries within session s_i		
TimeSpanPreSess ^B	Temporal length of session s_i (in minutes)		
NumOfLocationQuery ^B	Number of location-related queries in session s_i		
AvgDistancePreSess ^B	Average distance from current location to locations mentioned in session s_i		

Note: No cross-device *query similarity*features were included in the model
→ Also part of the automatic labeling

Features from Pre-switch Query			
GlobalFrequency	The historical frequency of q_i in the entire dataset		
PersonalFrequency	The frequency of q_i in personal search history S		
NumExactQueryDesktop ^B	Number of same queries as q_i on desktop in S		
NumExactQueryMobile	Number of same queries as q_i on mobile		
NumRelatedQueryDesktop ^B	Number of related queries as q_i on desktop		
NumRelatedQueryMobile	Number of related queries as q_i on mobile		
NumExactQuerySwitch	Number of switches that pre-switch query and post-		
	switch query are the same as q_i		
NumRelatedQuerySwitch	Number of switches that pre-switch query and post-		
	switch query are both relevant to q_i		
PreQueryContiguousSwitch	Number of contiguous cross-device tasks of q_i		
NumOfRelatedQueryInSess ^B	Number of queries relevant to q_i in session s_i		
NumOfTerm ^B	Number of terms in query q_i		
PreQueryCategory ^B	The search topic of query q_i		
PreQueryHour ^B	The hour component of t_i		
PreQueryDayofWeek ^B	The day of week of <i>t_i</i>		
IsWeekday ^B	Boolean, indicates t_i is weekday or weekend		
HasLocation ^B	Boolean, true if q_i contains location		
PreQueryDistance ^B	The distance from current location to location in q_i		
HasLocalService ^B	Boolean, true if q_i contains local service		
Features from the Transit	on		
TimeIntervalSwitch	The timespan between t_i and t_{i+1}		
GeoDistanceSwitch	The distance between where q_i and q_{i+1} are issued		
IsSameLocationSwitch	Boolean, true if q_i and q_{i+1} occur at the same place		
AvgSpeedSwitch	Average travelling speed during the switch		
Features from Post-switch Session			
TimeSpanPostSess	panPostSess The temporal length of session s_{i+1}		
PostQueryCategory	The search topic of query q_{i+1}		
PostQueryHour	The hour component of t_{i+1}		
GeoDistancePostSess	The distance travelled within session s_{i+1}		
AvgSpeedPostSess	Average travelling speed within session $\overline{s_{i+1}}$		

Findings (Automatic Labeling)

Feature Grouping	Accuracy	Positive Precision	Positive Recall	AUC
Baseline - Desktop Only	0.903	0.337	0.037	0.646
History	0.880 0.250 0.142	0.661		
+ Pre-switch Session	0.899	0.381	0.381 0.130	
+ Pre-switch Query	0.907**	0.504**	0.145**	0.757**
+ Transition	0.910** 0.544** 0.184**		0.781**	
+ Post-switch Session	0.910	0.568	0.169	0.806

Note: Similar findings for human labeling and when navigational queries retained (although smaller gains)



Feature Analysis

Top 10 features for predicting contiguous search tasks

Features		χ ²	Info Gain
	Number of related queries on mobile	491.23	0.0317
	Time span of the switch	378.00	0.0240
	Number of contiguous switch lead by the query (as pre- switch query)	342.97	0.0204
	Number of related queries appear as pre-switch queries	315.35	0.0208
	Average mobile moving speed during switch	295.77	0.0227
	Distance travelled during switch	270.39	0.0201
	Number of contiguous switches in user's history	235.27	0.0201
	Average device use entropy for all tasks in user's history	221.83	0.0154
	Number of related queries on desktop	172.74	0.0151
	If the pre-switch query and post-switch query are issued in the same city	95.81	0.0091



Enabled Scenario: Exploit Downtime

Being able to predict task resumption enables scenarios such as



What Can Engine do to Help?

 Search engine can perform actions on the users' behalf to *capitalize on the downtime* during the switch, e.g.,

Predict resumption at end of **pre-switch** session:

- Proactively save recent session state
- Try different ranking algorithms
- Pose the query to a question answering site
- Alert the user if better results found

Predict at start of **post-switch** session:

 Provide the user with the option to explicitly resume task on homepage



Summary and Takeaways

- Multi-device usage increasingly popular
- Cross-device search is prevalent
 - 15% of (non-navigational) switches are on same task
- Characterized some aspects of cross-device tasks
- Built predictive models of cross-device task resumption
- Provides a search engine with opportunity to help searchers by using between-device downtime