# Microsoft® Research Faculty Summit





# Bing: User Intent and Decision Engine

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## Overview

- Why Decision Engine
- Bing Demos
- Search Interaction model
- Data-driven Research Problems
- Q & A

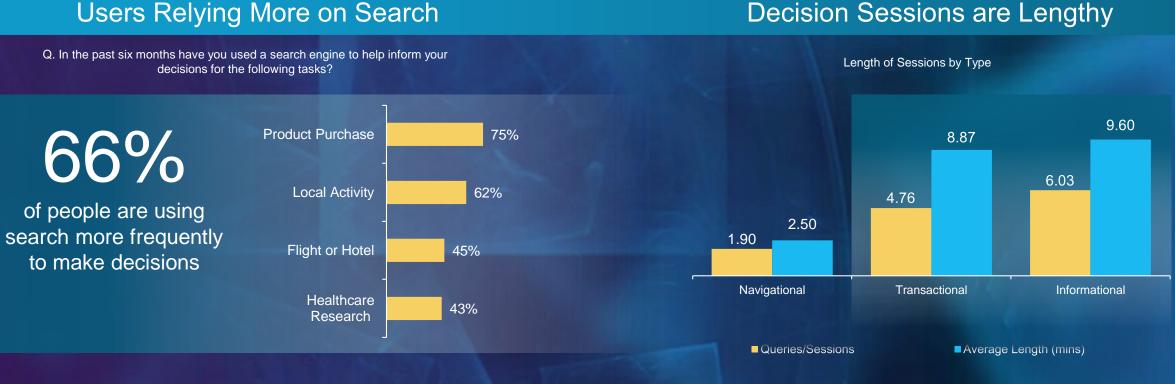


## **Opportunities for Search Innovation**





## **Turning to Search to Inform Decisions**



Users need help with tasks and making decisions

Complex task and decision sessions could be easier

Microsoft\*

Research

Sources: Microsoft Internal Research conducted by iPos 2009:; Live Search Session Analysis







#### **Frederick Savoye**

Senior Director Bing Product Marketing





# **Search User Interaction Model**

## Search User Interaction Model



### Search engine

- Objective: getting the user relevant information (a website)
- Model: getting out of search results page with a simple click
- Challenge: query URL matching

### Decision engine

- Objective: completing the task by fulfilling user intent
- Model: exploring search results by clicking and browsing
- Challenge: whole page relevance

## **Bing interaction model**

- Explore pane (or left rail)
  - TOC: verify user intent
  - Related search: expand user intent
  - Search history: remind user intent

### Task completion

- D-cards: showing structural information for the site
- Hover: showing more information for the site
- Simple tasks: e.g. Fedex tracking number







# **Data-driven Research Problems**

## Some Challenging Problems for Search



### A lot of data...

- Billions of query logs, documents, pictures, clicks, etc.
- Processing them is costly and takes time
- Statistical learning + distributed computing
  - Can we train 1 Billion samples (query URL pairs)
  - Within a few hours? No over-fitting?
- Two examples of Bing-MSR
  - "Bing-it-on" N-gram
  - Statistical model for log mining

## Statistical Language Model for Search



#### LM Applications in Search

- Query processing: alterations, expansions, suggestions
- Document processing: classification, clustering
- Matching and ranking
- Powerset.com
- Leading technology: N-gram
  - P(next word | N-1 preceding words)
  - Better "understanding" = model predicts better

## Challenges to build N-gram for Search



High quality model needs lots of data at web-scale:

- Billions of documents, trillions of words, PetaBytes of storage
- Smoothing:
  - How to deal with a very long tail
- Freshness:
  - Web contents are added and revised rapidly

Bing-MSR innovation: Constantly Adapting LM (CALM)

- Highly parallel algorithm designed for cloud computing
- Refine LM as soon as documents are crawled

## "Bing-It-On" Ngram Services



- We are sharing our resources
  - For details, go to <u>http://www.facebook.com/microsoftresearch</u> and follow "Bing-It-On Ngram"
- Compare Bing-It-On with Google's Release

	Google	Bing-It-On Ngram
Content Types	Document Body only	Body, Title, Anchor texts
Model Types	Raw Count only	Count and smoothed models
Highest order N	5	5
Training Size (Body)	~ 1.0 trillion words	> 1.3 trillion
# of 1-gram (Body)	13 million	1 billion
# of 5-gram (Body)	1 billion	237 billion
Availability	DVDs from LDC	On demand web services hosted by MS
Update	September 2006	Monthly

## Log Mining for Search



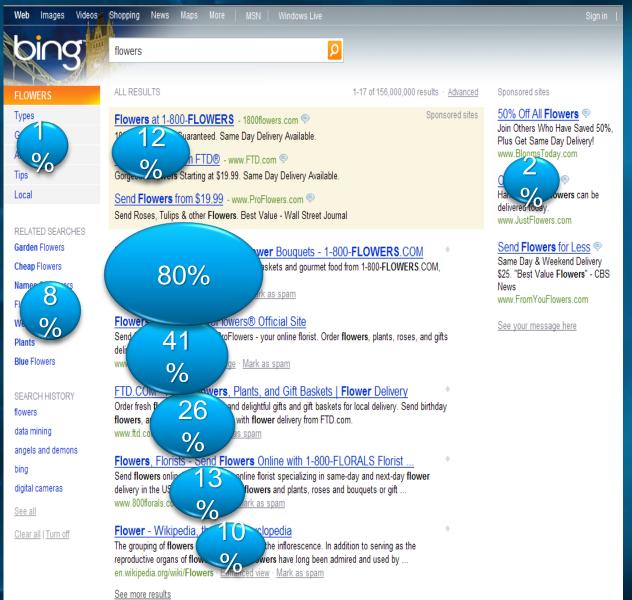
Data driven decision is first class citizen in search

- Toolbar and search logs: ~10 TB/Day each
- Bing uses log mining to
  - Understand what users want
  - Assess how we are doing
  - Quickly improve Bing
    - Query Processing
    - Ranking
    - User experience
- Examples:
  - Relevance inference using Bayesian Browsing Model (BBM)
  - User behavior understanding with Hidden Markov Model (HMM)

## Mining search log with HMM



- Search log records only clicked results
  - Skipped results are not explicitly recorded
- Hidden data mining
  - Model viewed results as Markov chain
  - Skipped results = hidden data
- How well does it work?
  - Very close to eye tracking data



### What's In the Name?









# Thank you!