Architecture, Development Model and Future Trends of Web Search Engines



Marcelo De Barros

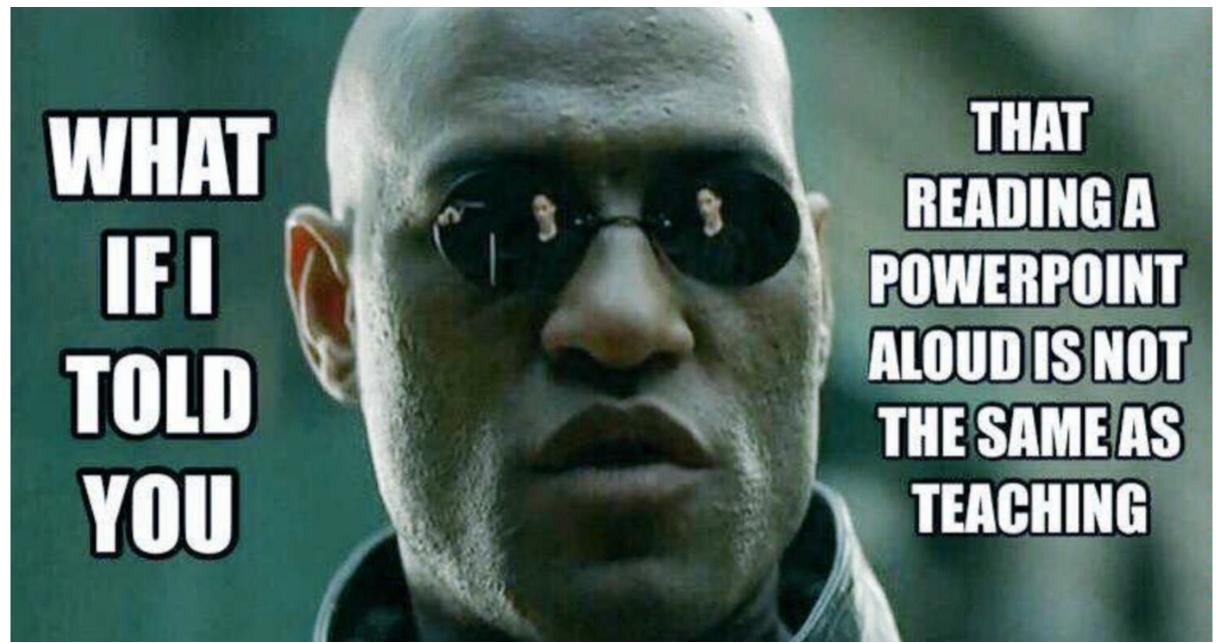
Bing UX Features and Shared Tools Team

Plan for the next hour

- I'll present you a simplified view of Search Engines architecture.
- I'll try **not** to use jargon without explaining it. Stop me if I forget.
- I'll talk about the future trends around search engines (my own opinion).
- You ask questions if you have them.
- If I can't answer, I'll follow up with someone who can.

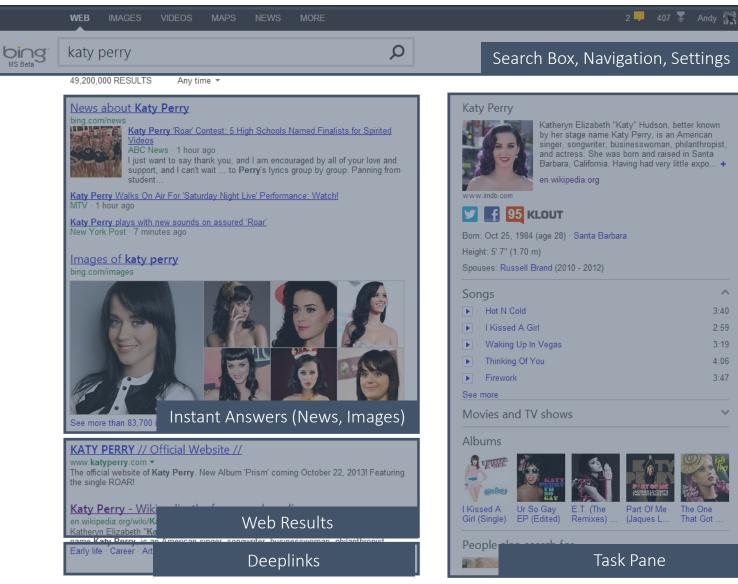
This deck represents an overview of Search Engines.

Some technical implementation details will be omitted on purpose ③

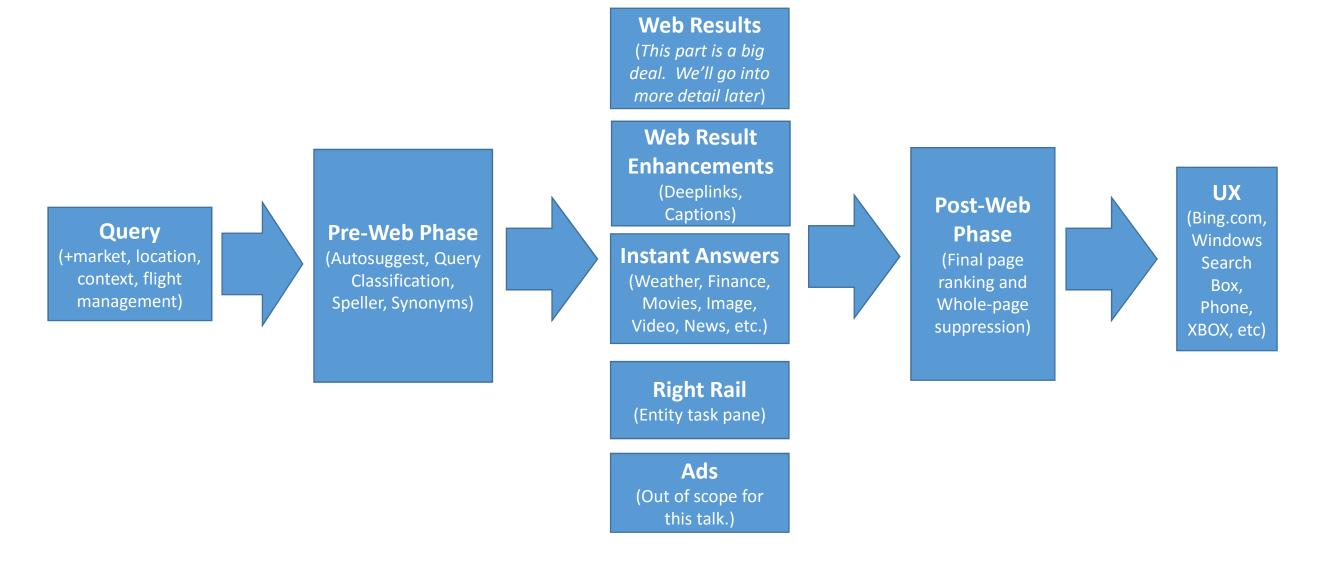


Microsoft confidential

The Anatomy of a Bing SERP – Search Engine Results Page



A Runtime Stack in One Slide



The User Query

- Things that a search engine might know even *before* we get to the web search:
 - Your query
 - Your entry point (Windows Search Box, Bing.com, Phone, XBOX, etc.)
 - Your market (country + language)
 - Your location (sometimes...)
 - Your past queries (sometimes...)
 - Your identity from logged-in experiences (sometimes)
 - Which flights (experiments) you are in
- Query Formulation via Autosuggest (traditional trie data structure)



Understanding Popular Pages

en.wikipedia.org/wiki/Katy Perry - Translate this page (0%)

• Search engines know a number of statistics about the pages:



Such information helps in ranking decisions, as well as caching decisions and placement decisions

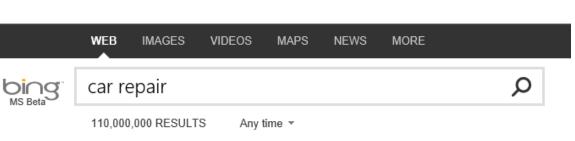
Query Rewriting: Spelling and Synonyms (pre-web)

- Spelling:
 - Dictionary (per language)
 - Logs (words proximity, clustering techniques, ranking within clusters)

	WEB IMAGES VIDEOS MAPS NEWS MORE		
	ktay perry	Q	
no bota	49,700,000 RESULTS Any time 👻		
	Including results for katy perry. Do you want results only for ktay perry?	5	perfecct
	 katy perry Ranker: 27.379032 Features - Local Norams - Udyssey ktay perry Ranker: 13.263762 Original Features - Local NGrams - Odyssey 		Web Images Videos Maps News Explore
	 katy parry Ranker: 4.836921 Features - Local NGrams - Odyssey katy perry' Ranker: 3.503151 Features - Local NGrams - Odyssey 		90,500,000 RESULTS Any time -
	5. katyperry Ranker: 3.23323 Features - Local NGrams - Odyssey 6. katey perry Ranker: 2.711165 Features - Local NGrams - Odyssey 7. kathy perry Ranker: 2.006864 Features - Local NGrams - Odyssey		Including results for <i>perfect</i> . Do you want results only for perfecct?
	Features - Local NGrams - Odyssey 8. katie perry Ranker: 1.353943 Features - Local NGrams - Odyssey 9. kay perry Ranker: -1.88421 Features - Local NGrams - Odyssey		

Query Rewriting: Spelling and Synonyms (pre-web)

- Synonyms:
 - Clustering techniques



DIY Auto Repair Help - Car Maintenance, Troubleshooting, ... autorepair.about.com -

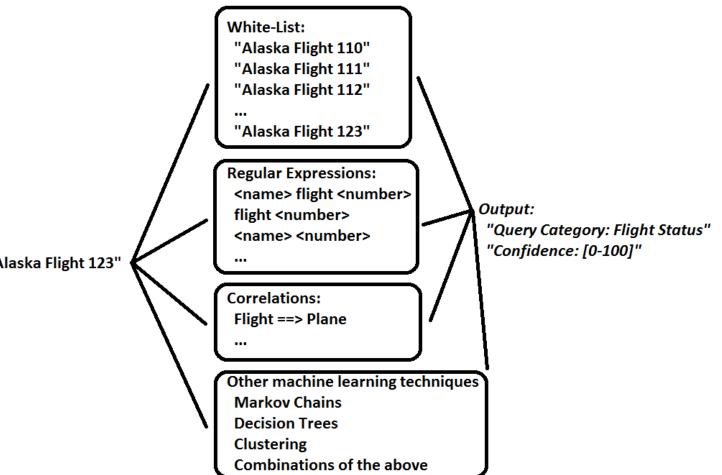
You can do your own **auto repairs** by following out easy step-by-step do-it-yourself tutorials which show you how to diagnose, troubleshoot, **repair**, fix, modify and ... Fix It Yourself \cdot Troubleshooting \cdot DIY Repairs \cdot My Car Won't Start

Term	<u>Candidate</u>	Tier1 Score		
[-] car				
	car	N/A		
	auto	999966		
	automobile	999885		
	automotive	999431		
	carr	996370		
	cars	999991		
	motor vehicle	918356		
	truck	999848		
	van	998102		
	vehicle	999862		
	vehicles	999752		
[-] repair				
	repair	N/A		
	fix	999851		
	fixing	1000000		
	maintenance	999855		
	owners	999849		
		000707		

Query Rewriting: Query Classification (pre-web)

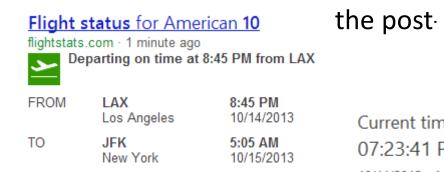
- Query Classification:
 - Fast Classifiers
 - White-List
 - Regular Expressions
 - Correlations
 - Other techniques

Query="Alaska Flight 123"



Instant Answers: a Federated Model

- After pre-web components are run, the query is federated out (dispatched) to dozens of Answer Services
- Anybody can ship an Answer service, and any answer can trigger for any query
- Answers vary widely in complexity. Some (like Flights/News/Stocks) have up-to-the-minute data requirements. Some (like Image/Video) have full indices and relevance stacks.



Cooking School near Seattle, Washington

bing.com/local

Blue Ribbon Cooking LLC 2501 Fairview Ave E, Seattle · (206) 328-2442

Directions

Nu Culinary

Directions

6523 California Ave Sw. Seattle · (206) 932-3855

Directions · ***** 1 review



4925 Ne 86th St, Seattle · (206) 525-7537

Bruno Mars height 5 feet 5 inches (1.65 meters) Find out more on: Freebase

Current time in Dubai, Dubai, United Arab Emirates 07:23:41 PM 10/14/2013 · Arabian Standard Time

Videos of video how to install a stereo bing.com/videos



YouTube

How to Install Your How-To Install a How to wire / Install a Car St Own Car ... YouTube YouTube

How to install an Car Stereo Svs... aftermarket car YouTube



Movies near Topeka, KS

msn.com/movies

ISIDIOU

age ra Weather in Hyderabad, India

Captain Phillips ★★★★★ · Drama · PG-13 2hr 14min

***** · Horror · PG-13

Insidious: Chapter 2

1hr 45min



Rush Action R · 2hr 03min

Runner Runner ***** · Suspe THURSDAR. R · 1hr 31min

Instant Answers: Targeted Experiences

- Instant answers are a great way to meet users' demands
- Users no longer have patience for the traditional blue links $\ensuremath{\mathfrak{S}}$

x^10-3x^3+9=0	periodic table	fibonacci in python		
Web Images Videos Maps News	Web Images Videos Maps News Explore	Web Images Videos Maps News Explore		
- ·	12,700,000 RESULTS Any time 👻	Also try: Fibonacci Sequence Python Code · Python Fibonacci Recursion · Fibon 169.000 RESULTS Any time •		
17,400,000 RESULTS Any time -	Periodic table			
Polynomial equation solver	Enter an element name or symbol Find element	Fibonacci Python ~		
$x^{10-3x^{3+9=0}}$ Standard form: $x^{10} - 3x^{3} + 9 = 0$	Chemical group Physical state Discovered Found on Earth Density H B B B C N O F Ne	<pre>1 def calculate_fib(n): 2 fib, tmp = 0, 1 3 for i in range(n): 4 fib, tmp = tmp, fib + tmp 5 return fib 6 7 ifname == "main": 8 print calculate_fib(5), 9 print calculate fib(10),</pre>		
Solutions based on Jenkins–Traub algorithm: x ₁ = -1.213921 - 0.454848i	11 Na 12 V 21 V 22 V 23 V 24 V 25 V 26 V 27 V 28 V 29 V 20 V 30 V 34 V 35 V 36 V 37 38 V 39 V 40 V 42 V 43 V 44 V 45 V 46 V 47 V 48 V 49 V 50 V 51 V 52 V 54 V 7b 57 V 7 V 10 V	<pre>print calculate_fib(15), print calculate_fib(15),</pre>		
x₂ = −1.213921 + 0.454848i	55 56 72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 CS Ba Hf Ta W Re Os Ir Pt Au Hg TI Pb Bi Po At Rn			
x₃ = −0.660282 − 0.958082i	87 Fr 88 Ra 104 Rf 105 Db 106 Db 107 Db 108 Hs 109 Hs 110 Hs 111 Ds 112 Ds 113 Ds 114 Dup 115 Uup 117 Uup 118 Uup 117 Uup 118 Uup 117 Uup 118 Uup 117 Uup 118 Uup 116 Uup 117 Uup 118 Uup 116 Uup 117 Uup 118 Uup 116 Uup 117 Uup 118 Uup 118 Uup			
x₄ = −0.660282 + 0.958082i	La Ce Pr Nd Pm Sm Eu Gd Tb Dy Ho Er Tm Yb Lu 89 90 91 92 93 94 95 96 97 98 99 100 101 102 103 Ac Th Pa U Np Pu Am Cm Bk Cf Es Fm Mdd No Lr	Run code		
Show all V		5 55 610		
	Hydrogen Alkali metals Alkali-earth metals Transition metals Post-transition metal Metalloid Polyatomic nonmetal Diatomic nonmetal Noble gas Lanthanide series Actinide series Actinide series	Powered by HackerRank.com		

Entity (or Side) Pane

- The Entity Pane is a special kind of Instant Answer
- It pulls in content from various answers and displays it all together in one place
- Search engines keep a graph of entities on the Web

Gravity (2013)



Dr. Ryan Stone is a brilliant medical engineer on her first shuttle mission, with veteran astronaut Matt Kowalsky. But on a seemingly routine spacewalk, disaster strikes. The shuttle is destroyed, leaving Stone and Kowalsky completely alone - tethered to nothing but each other and spiraling out into the blackness. The deafening silence tells them th... +

Watch trailer: Moviefone Summary: PG-13 · 1hr 31min · SciFi/Fantasy Director: Alfonso Cuarón

Reviews	★★★★★ 90,261 ∧
IMDB	8.8/10
MSN	4.5/5 stars
Flixster	97% positive

Cast



Sandra	George
Bullock	Clooney
Dr. Ryan	Matt
Stone	Kowalsky

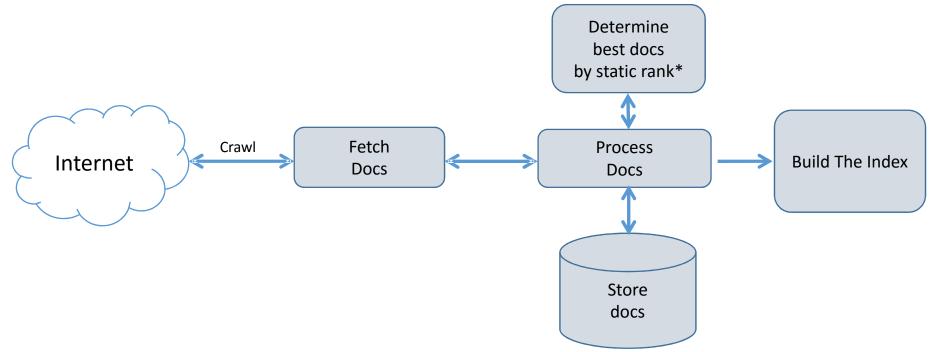
An Aside on Web Relevance

Web Relevance

- Objective: find the 10 (sometimes more, sometimes less) most relevant blue links for the query and put them in the right order on the page
- How this happens in 6 oversimplified steps:
 - 1. Acquire billions of web documents and index them "this is a hard problem from many angles, mainly from a scalability and storage standpoint"
 - 2. Match each user query to some possibly relevant web docs
 - 3. Use machine learning to rank the candidate web docs
 - 4. Return the top ten (give or take) to the user
 - 5. Do this globally
 - 6. Do this in a blink of an eye!!!

Where do the documents come from?

Generation of the Index = the process of crawling/storing docs and building the index



*Static Rank = the query-independent importance score that we assign to every document on the web

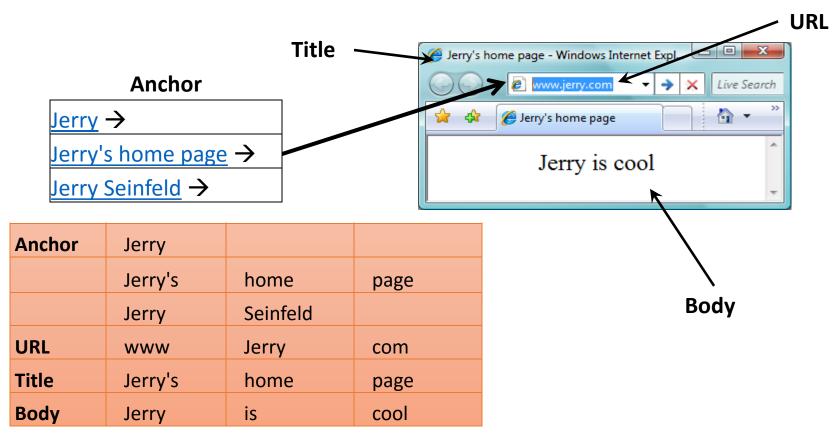
How are documents served for a query?

Index-Serve = the process of hosting docs and returning them for incoming queries

- Search Engines have multiple-tiers platform to balance for freshness, relevance, index depth, and cost. Bing for example:
 - Fresh tier
 - Millions of documents
 - Doc discovery to hosting takes <1min
 - Main tier
 - Billions of documents
 - Updates in <1 day
 - Depth tier
 - Many Billions documents
 - Updates in <7 days
- Includes both En-us docs as well as global docs

How is a query matched to documents?

There are four basic streams (text sources): Anchor, URL, Title, and Body (or AUTB)

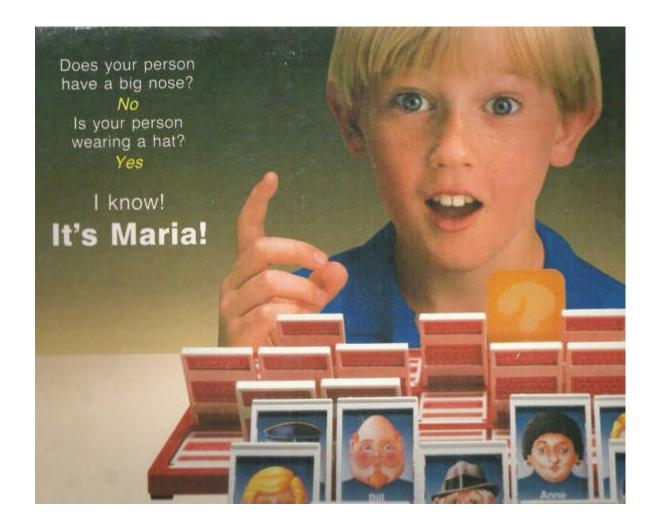


• AUTB is just the basics that Bing uses. Other engines might use other streams. We also rely heavily on Speller and Synonym expansion.

So Far, We Have a Big Pile of Documents

- We've matched a few thousand (or more) documents to your query
- Now we just need to get them in the right order
- How do we do that? Machine learning!

Machine Learning is like Guess Who



The Steps of Machine Learning

Machine learning helps a machine answer human questions (e.g. what are the best docs for this query?) by quantifying human questions into scores.

- 1. First, create some examples *where you know the right answer*. This is called **training data**.
- 2. Figure out some important *easy* and *quantifiable* questions to ask of those examples. These questions are called **features**.
- 3. Use the training data to "learn" how to get the known examples right by adjusting weights until the numbers work out. This is called the **training process**.
- 4. Then, for *new* examples, the system can take an educated guess at the right answer. This is called **generalization**.
- 5. Measure how well you do. Do this early and often.
- 6. Then go back and fix the problems. This is called **tuning**. Rinse and repeat.

Web Ranking Features

Features can be for the query, the doc, or both. Here are just a few examples of many that are used by different search engines:

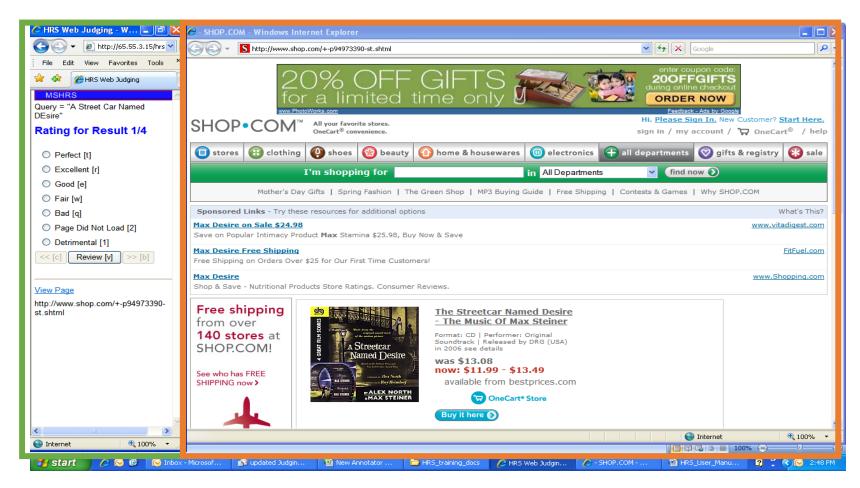
- Do the query and doc belong to the same **category**? (sports, movies, etc.)
- Do the query and doc come from the same **geographical origin**?
- How many times does the query term appear in the doc? (frequency)
- Does the query have any **known phrases**? e.g. {*star wars* trailer}
- How important is the doc? (Remember **static rank**?)
- We also look at **doc clicks**.
- Has the doc been classified as **junk/spam/adult**?
- What **query terms** have people used in the past to get to (click on) this doc? (queries association technique)
- And many, many more!

In the end, each query/doc pair gets a **dynamic rank** score. The docs are ordered by this score.

How do we gather training data?

Relevance Measurement: judges assess query/doc pairs on a five-point scale. This is used for both training and testing.

The process of pulling in the top N docs for a query and storing them is called **scraping**. We use these judgments to train, test and measure our rankers (machine learning models).

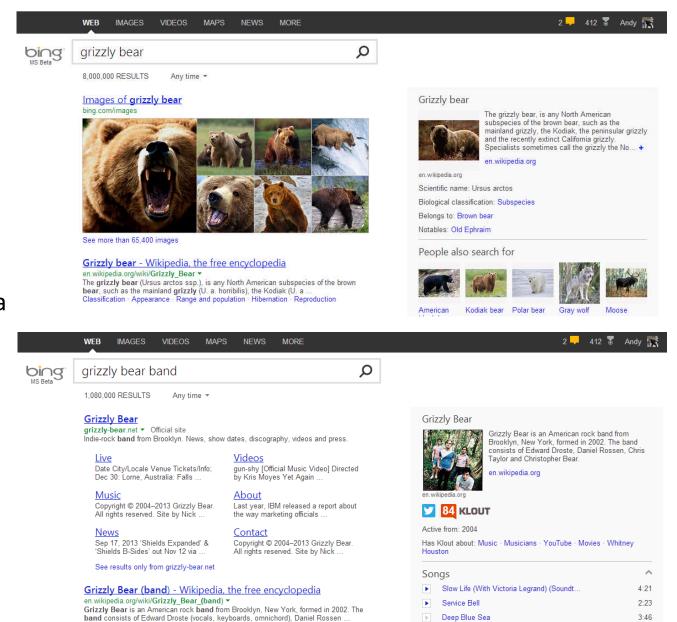


Bringing all together – recap!

- 1. A query comes in via one of several entry points
- 2. Some contextual information comes with the query
- 3. A few core services (e.g. Speller, Alteration) process the query
- 4. The query is "federated out" to Web, Answers, Task Pane, etc.
- 5. A subset of answers trigger for the query
- 6. The web ranker matches *many* documents and returns the top 10
- 7. All of this takes few milliseconds...
- 8. Now, we have a big pile of stuff waiting to be rendered on the page

Page Coherence

- It doesn't look good to show apples and oranges intertwined...
 - Jaguar: The Car? The Animal? The City?
- Need to apply **suppression**, and then
- Need to apply final ranking
- Coherence between web docs and answers is a key component
- Past data (user clicks) is also important
- The job of suppression is to **minimize defects**
 - A defect = irrelevant or otherwise bad content for a query
 - Components that perform poorly lose credibility
- The job of the final page ranking is to **push the best stuff to the top and the less-good stuff toward the bottom**
 - This is done via a metrics derived from click-info



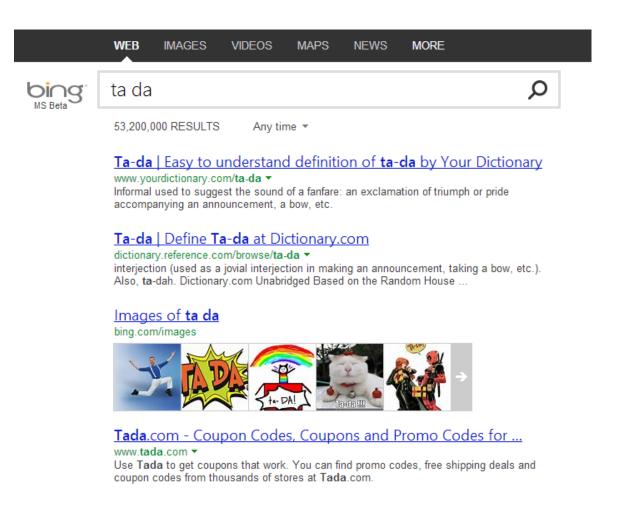
This Sona

3:39

History · Band members · Discography

UX (User Experience, or UI)

- After Whole-Page Relevance decides what to show, it passes the final content to the UX layer
- The content is rendered beautifully on the page
- The layout is customized by entrypoint, but the content is (mostly) the same
- UX Server: ASP.Net
- UX Client: Java Script (Libraries) + HTML + CSS3



Engineering Development Rhythm

Inner Dev Loop

- Feature development
- Concludes at checkin

Outer Dev Loop

- Build validation
- Concludes at PROD deployment

Monitoring

- Live Site quality
- Continuous

Flighting

• Controlled exposure of features

Development is composed of discrete states <u>https://www.youtube.com/watch?v=SiPtRjiCe4U</u>

Engineering Development Rhythm - Testing

Inner Dev Loop					
Mocked automation	Outer Dev Loo	p Monitoring			
Visual	automation	Exploratory	Flighting		
validation		testing	Pre-rotation		
Perf analysis		Auto- monitoring	validation		
		Feature parity			
		i cature parity			

Testing is composed of overlapping states

Engineering Development Rhythm

- Hundreds of engineers across many continents!
- Shipping multiple times a day (millions of lines of code):
 - Continuous Delivery → "your check-in will go to production soon!!!"
- Tens of thousands of automated tests
 - If any fails → don't ship
 - Don't write tests? Well, good luck shipping to hundreds of millions of users!!!
- Flight everything → Analyze the data → Ship or fail fast!!!

Flowers at 1-800-FLOWERS - Same Day Delivery Available.					
Y 1800flowers.com					
★★★★★ (183920 reviews) · 37,000+ followers on Twitter					
Same Day Delivery Available. 100% Satisfaction at 1-800-FLOWERS.					
Anniversary Flowers.	Best Selling Flowers.				
Rose Spectacular.	Birthday Flowers & Gifts.				
Fresh Cuts.	<u>Gift Baskets.</u>				

Guardrail Metrics	Treatment	Control	Delta [%]	Pval
Quick Back 20	0.2295	0.2281	0.0014 [0.60%]	< 0.001
Algo Pane Load Time(Overall PLT)	1212	1208	4.055 [0.34%]	< 0.001
Revenue /UU	1.088	1.075	0.0130 [1.21%]	< 0.001
Truncated Revenue / UU	0.8571	0.8504	0.0067 [0.79%]	< 0.001
Distinct Queries / UU	14.67	14.67	-	1.001
Average Log Record Size (in KB)	111.4	111.1	0.2545 [0.23%]	< 0.001



- Data: not every data is in the index...
 - Offline data other formats
 - Live data happening now, I mean, really, NOW !!!





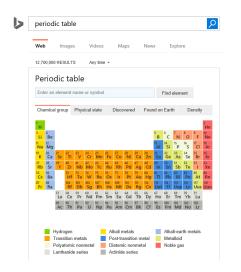
- •AI: it is only in its infancies
 - Image and Video Understanding
 - Media \rightarrow Features \rightarrow Syntax \rightarrow Semantics
 - Personal Assistant (Cortana, Siri, Google Now, Alexa)





- Fundamentals: more connections, less no patience
 - Internet of Things (IoT)
 - Availability across devices (phones, wearables, cars, things)
 - Poses unique User Interface challenges
 - Poses unique privacy concerns
 - Performance:
 - Not fast... but NOW!!!!
 - Fun experiment: slowdown flight → revenue hit!
 - Pushing the limits of techniques
 - Algorithms, distributed computation, hardware, networks, caching, programming languages, etc.
 - Faster data analysis
 - Data is becoming cheaper...
 - However, useful information from the massive data sets is hard!

- Collaboration: search is also about connecting services
 - No more blue-links: the answer must be right there!
 - Many specialized companies
 - Servicefication of platforms



Q&A