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Semantic Computing for eScience

Evelyne Viegas
Director Semantic Computing

Outline

- My Background
  - Academia
  - Industry
- Data to drive innovation
  - Next Generation Innovations
  - From Data Release to Data Services
- From Data and Information to Knowledge
  - Towards a Knowledge World
  - From Data Services to Knowledge Services

Dr. Evelyne Viegas
Microsoft Research

Computational Semantics
Natural Language Processing
Intelligent Planning

Microsoft

Semantic Search

Language Processing for Help
Natural User Interactions

Microsoft Research

Semantic Computing

Data - Information - Knowledge - Intelligence

Beyond Data

Vision – Enable the *Next Generation Internet* by working with Academia, stakeholders from industry, government, and internet consumers/innovators to build an Intelligent Web, making sense of data via open innovation

DATA - INFORMATION - KNOWLEDGE - INTELLIGENCE

Data has become a first class citizen

IT'S A DATA-DRIVEN WORLD

It's a data-driven world

- Spell Checking
- Machine Translation
- Search queries + click through
- Online games skill matching

Data logs behaviours in more reliable ways than demographic studies or surveys to study/predict trends

(Banko and Brill, 2001) – effectiveness of statistical NLP techniques is highly susceptible to the **data size** used to develop them

(Norvig, 2008) – it is the **size of data**, not the sophistication of the algorithms that ultimately play the central role in modern NLP

Challenge in Data-driven Research

- Lot of the data needed for data-driven research in industry
  - Reason: scale; privacy, business sensitivity

How to make real world large scale data available to researchers to nurture innovation and perform valid experimentation, while maintaining privacy?

DATA RELEASE

DATA COMPUTE

DATA SERVICES

DATA ACADEMIC ENGAGEMENTS

Data for Open Innovation - Promises

Innovation

- By having access to real world data at scale, researchers can unveil **new** analysis or **research directions** based on shared assets and explore new questions

Science

- By allowing wider use of data, **repeatability of experiments** can be performed and data misrepresentations or faulty results avoided

Training

- Last but not least, real world large scale data is a powerful tool for **training the next generation of researchers**

Data for Open Innovation - Challenges

With web users becoming producers of information, leaving the footprint of their lives in digital trails, it is becoming easier for “data snoopers” to reconstruct the identity of an individual or an organization by cross linking information from different sources

A Face Is Exposed for Searcher No. 4417749

The New York Times, Aug 2006

Thelma Arnold's identity was betrayed by the records of her Web searches

"Search query data can contain the sum total of our work, interests, associations, desires, dreams, fantasies, and even darkest fears" said, Lauren Weinstein, a privacy advocate.

Accelerating Search in Academic Research Request for Proposals (RFPs)

Accelerating Search in Academic Research

Search RFP Awards

Search assets (15 million search queries + click through)
- PII (including inadvertent) removed
- Provided under a limited data licensing agreement

Increased quota to the Search API

Search Summit 2007

Search RFP06 projects review
The Quest for Assets – the Good the Bad and the Wanted

RFPs Program Feedback

- Search Summit 2007 asks:
  - Need more data, larger scale
  - Need to follow a user (privacy!)

- Beyond Search – Semantic Computing and Internet Economics 2009 new asks:
  - Need data access (as opposed to data release)
  - Compute power

Web N-gram Services

Access to up to *petabytes* of real world data

http://research.microsoft.com/web-ngram

Leading technology in Search, Machine Translation, Speech, Learning

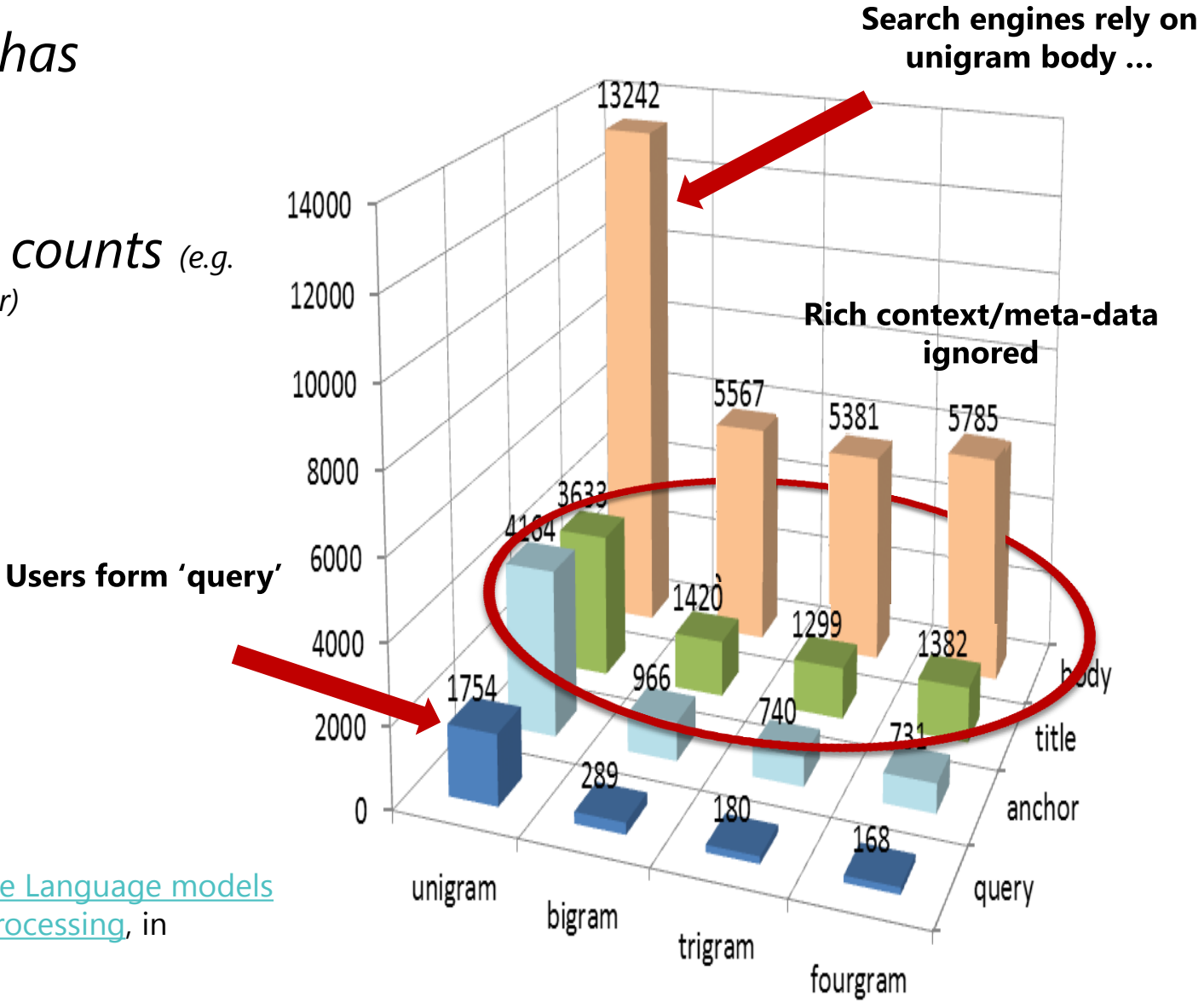
bing™

Microsoft Research

Web N-Gram in Public Beta

Web data has structure…

…and that counts *(e.g. Body, Title, Anchor)*



[Exploring Web Scale Language models for Search Query Processing](), in *WWW'2010*

Web N-gram Services

Content types

- Document Body, **Document Title, Anchor Texts**

Training size (Body)

- ***All* documents** for en-us indexed by Bing (no cut off)

Access

- **Hosted Services** by Microsoft

Updates

- **Periodical updates**

http://research.microsoft.com/web-ngram

Word Breaking examples

Enter a hash-tag phrase, and we will show the likely breakdown of sub-words. For instance, enter #nowplaying. More examples...

#whenifirstmet #nowplaying #wtfyoumean #thissummer #enoughisenough #ifirstmet #riptherunway #complimentgonebad #SMHyoureghetto #letmefindout #idoit2 #itaintmyfault #FlavoredCondoms #jayparkaom #ChrisBrownRocks #thingthatihate #nowplaying #whenifirstmet #hereugo #stpatricksday #thissummer #hiphopaintdead #idoit2 #sexisthebest #Lupequotes #WillYouEver #flavoredcondoms #whenimeetjustin #hcr #FF #Nowplaying #followfriday #howyouathug #youaintforme #OhJustLikeMe #NotMeThough #HCR #idoit2 #yeaisaidit #Advice #iloveitwhentrey #MarchMadness #TLS #ihatequotes #s1battle #nowplaying #howyouathug #uaintforme #youaintforme #WhenIfirstmet #whatsworse #WhenTwitterWasDown #howuathug #ChrisBrownonUstream #hereugo #TLS #justinbiebermyspace #idoit2 #HCR #willyouever #marchmadness #Hereyougo #nowplaying #imthekindofperson #FF #6wordstory #whitecusswords #whoelsenoticed #yeaisaidit #hcr #idoit2 #ss3forindonesia #Ohjustlikeme #blackcusswords #theboltonnews #ss2malaysia #FollowFriday #arashi #StopHatingDemi #mucoreSNSD #nowplaying #imthekindofperson #MJis #whitecusswords #OhJustLikeMe #idoit2 #thankstwitter4 #YourUnderArrest #hcr #BounceBackTeuk #inschool I #Imliableto #DontBeMadBut #becauseofbieber #ChrisBrownonUstream #hbu #nowplaying #dearfuturewife #imthekindofperson #musicmonday #Isitjustme #goseethedoctor #hcr #idoit2 #thankstwitter4 #MM #OhJustLikeMe #TLS #ohmySiWon #thatisall #ihatequotes #afmlmoment #biebermemories #tellmewhyumad

#yeaisaidit

Phrase	LgProbabilit
yea i said it	-9.345904

#whenifirstmet

Phrase	LgProba
when i first met	-6.97489
when ifirstmet	-10.34817
when ifirst met	-10.67689
when i firstmet	-11.09351
wheni first met	-11.1378

#nowplaying

Phrase	P
now playing	-4
nowplaying	-7
n ow playing	-9
now play in g	-7
now play ing	-7

buenosdias Go

Phrase	LgProbability
buenos dias	-7.412106
buenosdias	-9.258749
b uenos dias	-10.89817
buenos dia s	-10.92766
buenos di as	-10.96579

Total Time (ms): 312

Service Time (ms): 270

#parlezvousfrancais Go

Phrase	LgProbability
parlez vous francais	-7.901024
parlezvous francais	-11.01565
parlez vousfrancais	-11.23517
pa rlezvous francais	-11.30055
parlezvousfrancais	-11.41711

#w84u

Phrase	LgProbability
w8 4 u	-10.0969
w84u	-10.27723
w 84u	-10.69117
w 84 u	-10.7444
w 8 4 u	-11.06896

Multi-word Tag Cloud from Government Dataset Titles

Single Tag Cloud

Multi Tag Cloud

Ref: Dr. Li Ding, Rensselaer Polytechnic Institute

Query Segmentation

Body:

-18.64152 mike siwek | lawyer | mi
-19.66447 mike siwek lawyer | mi
-19.70832 mike siwek | lawyer mi
-20.3373 mike siwek lawyer mi
-20.60077 mike | siwek | lawyer | mi

Title:

-17.50179 mike siwek | lawyer mi
-17.92375 mike siwek | lawyer | mi
-18.0385 mike siwek lawyer mi
-18.46046 mike siwek lawyer | mi
-19.81768 mike | siwek | lawyer mi

Anchor:

-18.84468 mike siwek | lawyer | mi
-19.7035 mike siwek | lawyer mi
-20.96786 mike | siwek | lawyer | mi
-20.98327 mike siwek lawyer | mi
-21.82668 mike | siwek | lawyer mi

Impact with Microsoft Web N-gram Service

- Sheer power of data
  - Cross lingual documents are a way of life. N-grams seem to work on other languages
- Documents have structure and styles
  - A single document is written in many languages, with the document body, title and anchor text being all different languages that should be treated separately
  - Web has other languages such as those used for SMS. The N-gram Service works on this kind of language which opens up a lot of interesting research questions

Are we revisiting the concept of "language identification" as a means of identifying languages of different styles, and not so much on national languages (Wang et al., NAACL-HLT 2010)

Use Microsoft Web N-gram Services and get to Webscale

http://research.microsoft.com/web-ngram

webngram@microsoft.com

Free for *non* commercial research
Scaling to TeraBytes and PetaBytes
Regular data/feature updates
ISRC Research team to engage with

Available on Azure for the awardees of the NSF Program Solicitation Computing in the Cloud

Research papers (SIGIR 2010)

Implicit Search

-17.07376	Chateau Montelana in Napa
-17.28525	Chateau Montelana in Napa
-17.36758	Chateau Montelana in Napa
-17.49432	Chateau Montelana in Napa
-22.04415	Chateau Montelana in Napa
-22.10234	Chateau Montelana in Napa
-22.25322	Chateau Montelana in Napa
-22.39616	Chateau Montelana in Napa

'Chateau Montelena in Napa' segmentation

'Chateau Montelena' as an *entity* in Wikipedia

Spelling Alteration for Web Search
http://spellerchallenge.com

SPELLER CHALLENGE
BING – MICROSOFT RESEARCH PARTNERSHIP

Microsoft Research

Sign in

Speller Challenge

Home | My Team | Rules | FAQ | Datasets | Leaderboard

Find us on Facebook
Follow us on Twitter

bing

Microsoft Research in partnership with Bing is happy to launch the Speller Challenge

Do you have what it takes to build the best speller? Enter the Speller Challenge by developing a speller that generates the most plausible spelling alternatives for a search query.

In doing so, you can:

- Try out your speller using real world data;
- See how it compares to the rest of the community's spellers;
- Be eligible to win a cash prize.

Register Here

Speller Challenge

Start date: Dec 15, 2010
End date: May 27, 2011

Web Ngram Services used to create data set by participants

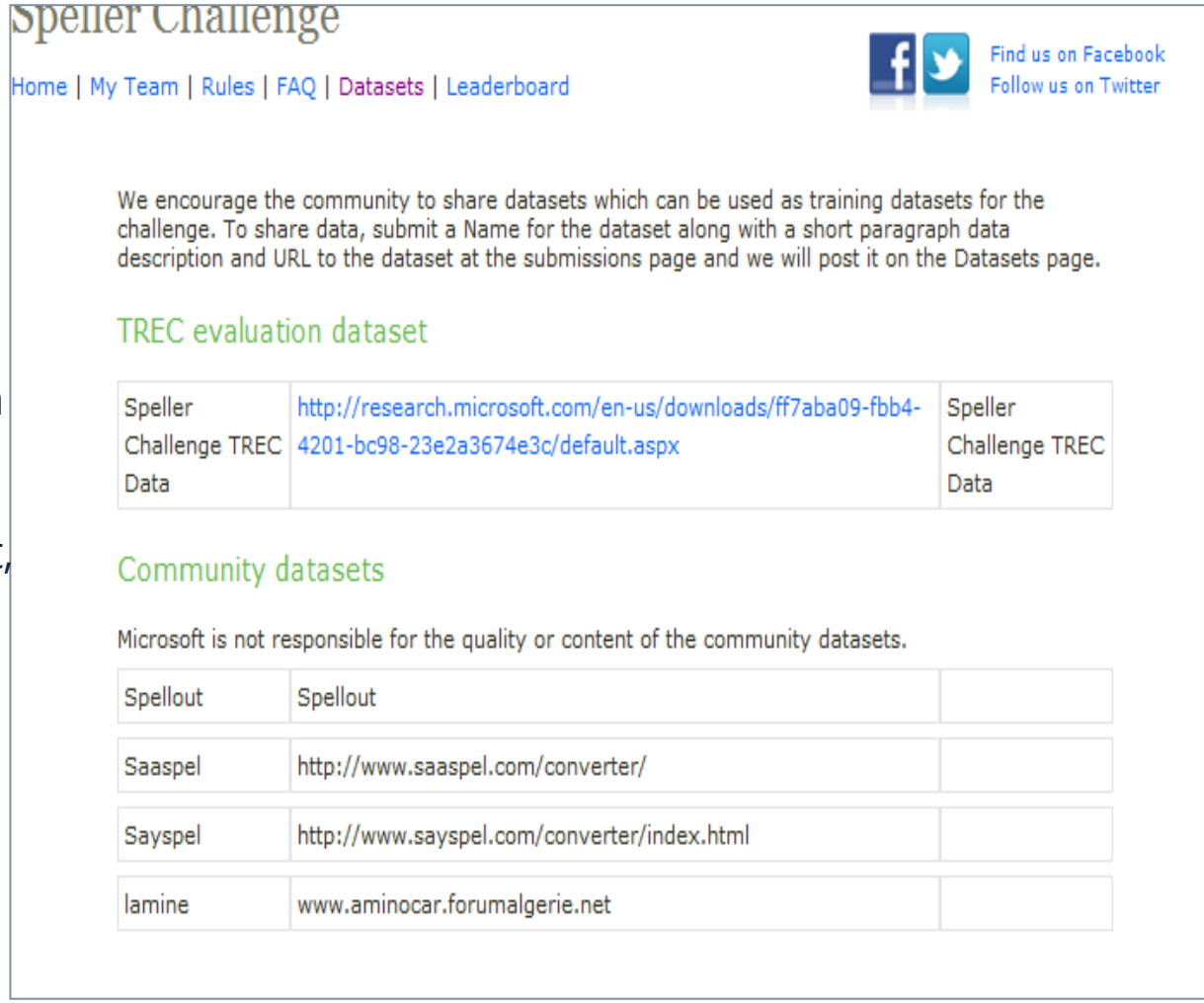
Automatic evaluation of the participants' spellers

Five Prizes to win

Learning - Community "shared data sets"

Speller Challenge TREC Data

Participants can make their data set available to the rest of the research community by providing a link to their data set, on the challenge "community datasets" page.



Speller Challenge

Home | My Team | Rules | FAQ | Datasets | Leaderboard

Find us on Facebook
Follow us on Twitter

We encourage the community to share datasets which can be used as training datasets for the challenge. To share data, submit a Name for the dataset along with a short paragraph data description and URL to the dataset at the submissions page and we will post it on the Datasets page.

TREC evaluation dataset

Speller Challenge TREC Data	http://research.microsoft.com/en-us/downloads/ff7aba09-fbb4-4201-bc98-23e2a3674e3c/default.aspx	Speller Challenge TREC Data

Community datasets

Microsoft is not responsible for the quality or content of the community datasets.

Spellout	Spellout	
Saaspel	http://www.saaspel.com/converter/	
Sayspel	http://www.sayspel.com/converter/index.html	
lamine	www.aminocar.forumalgerie.net	

Summary and Questions

- From data release to data services
  - Allows to handle much bigger data sets
  - Allows to provide abstractions on the data (e.g. language models)
  - Allows to provide data compute capabilities
  - Allows for agile experimentation

- How to better engage with academia to drive data-driven research?
  - How does a cloud-based data service approach change research?

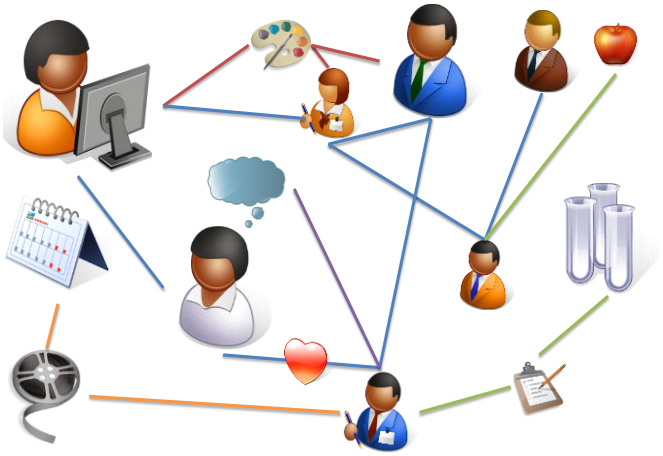
- What else can industries do to help democratize large scale data-driven research?

Contact: evelynev@microsoft.com

The world has become more connected

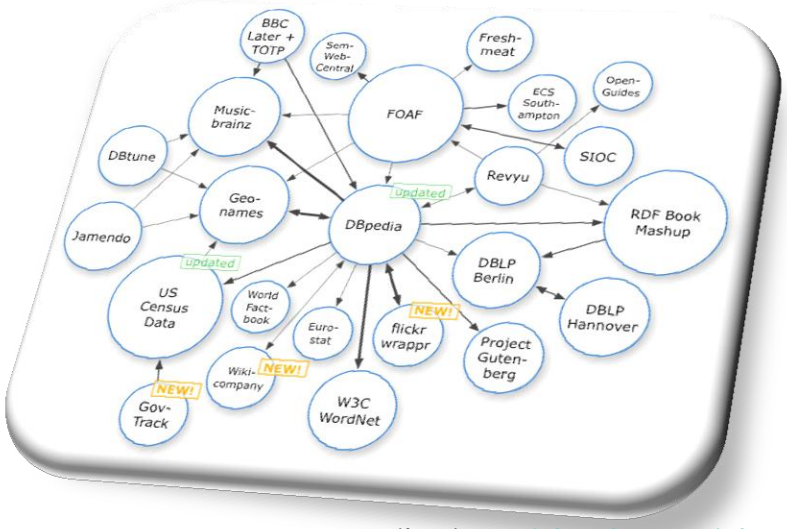
TOWARDS A KNOWLEDGE WORLD

A world where all data is linked…



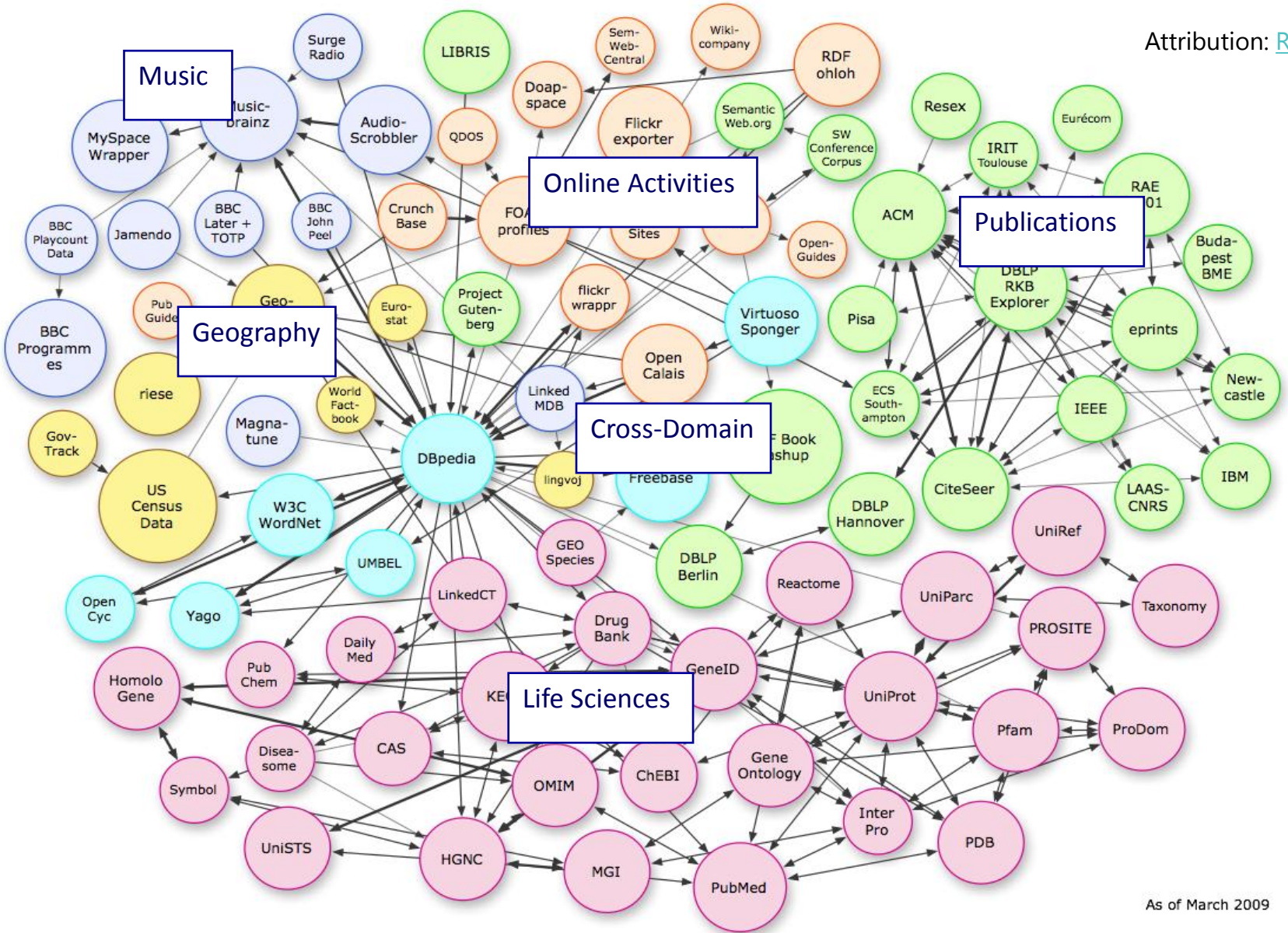
- Information inter-connected through machine-interpretable information (e.g. paper X **is about** star Y)

- Formats or “well-known” representations of data/information
- Pervasive access protocols are key (e.g. HTTP)
- Data/information is uniquely identified (e.g. URIs)
- Links/associations between data/information



Attribution: Richard Cyganiak

Linked Data



Attribution: Richard Cyganiak

As of March 2009

Semantics at Web Scale

- **Data, information is dynamic**
  - 450,000 changes per day in Wikipedia (en)
  - One new word created every 98 minutes (14.7 a day)

- **Is there a "right ontology?"**
  - Ontologies are abstractions
    - Ontologists make design choices all the time
  - Ontologies are application dependent
    - Machine Translation: *eat* vs. *comer* (INGEST concept)
    - Robotics: *eat* (task planning)

Probabilistic Modeling for Merging Knowledge Bases

- Yago - Ontology based on data from Wikipedia, WordNet and GeoNames with 400 M facts, 108 relations
- IMDb - Ontology based on data from the IMDb TV & movie site with 23M facts and 10 relations

- Challenges
  - Differences in information coverage
  - Text Mismatches (e.g. Kim_Novak vs. Novak, Kim)
  - Concept Mismatches (e.g. wasCreatedOnDate vs. hasProductionYear)
  - Granularity differences (e.g. one entity for one TV series vs. multiple entities per episode)

~20% entities exact match

Ontology Merging as a Probabilistic Inference and Learning Problem

- Are these two entities the same? Are these two relations the same?
- Can we learn the string transformations in an unsupervised manner from data?
- Given examples, can we learn a good cost function / probability model for matching?
- If two ontologies disagree, which is more reliable?
- How do we represent uncertainty in our ontologies?

Vision: a tool that automatically merges 2 ontologies

Zoubin Ghahramani, Univ. of Cambridge

ProBase

Large Scale Probabilistic Knowledge Base

Haixun Wang, MSRA

- Capture concepts (in our mental world)
- Quantify uncertainty (for reasoning)

Probase – 2.7 M concepts

automatically harnessed

Freebase – 2 K concepts

built by community effort

Cyc – 120 K concepts

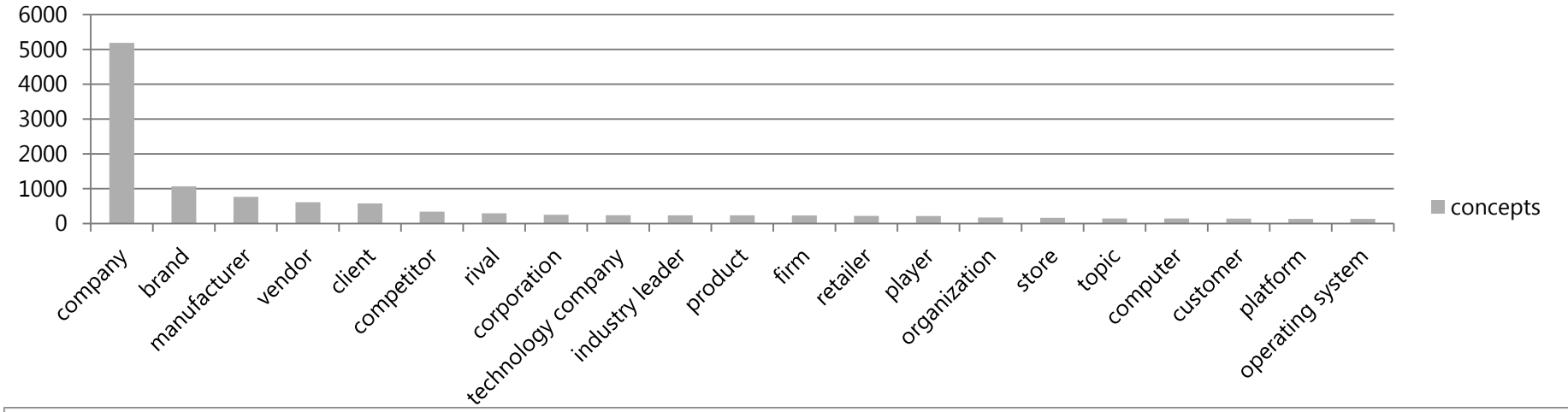
25 years human labor

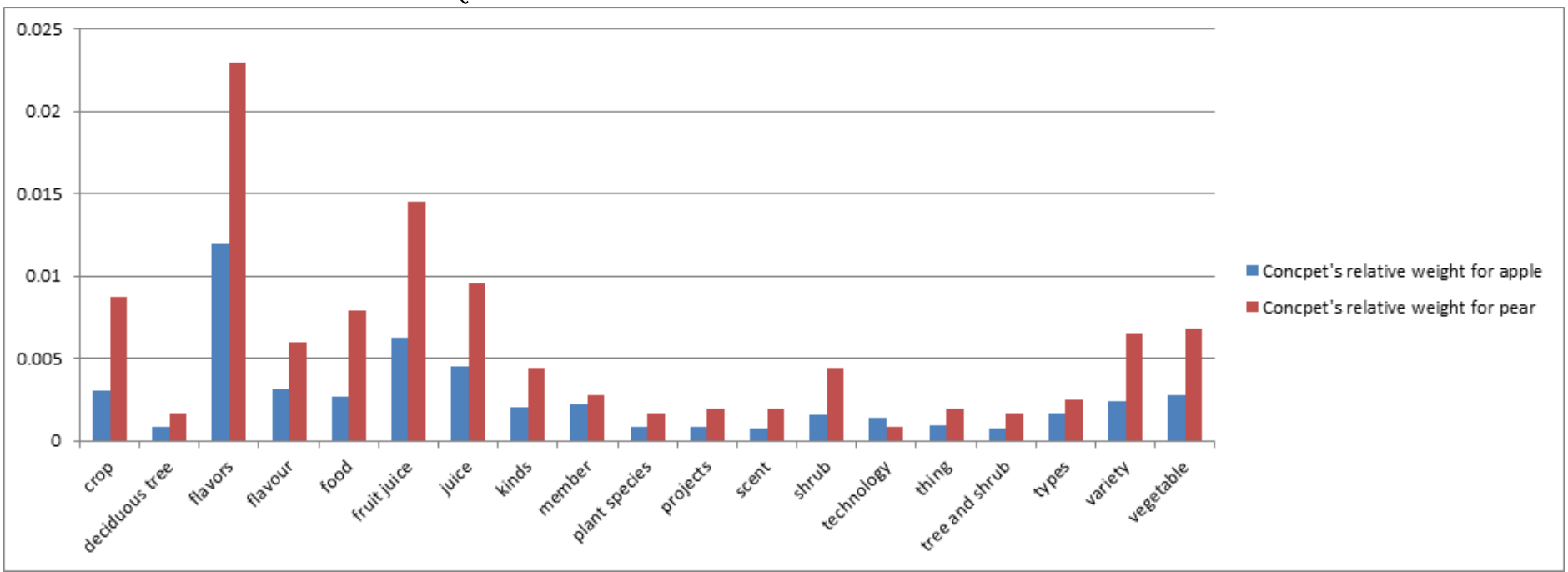
Uncertainty

Probase vs. Free**base**

Correctness is a probability.	**Knowledge is black and white.**
Live with dirty data.	**Clean up everything.**
Dirty data is very useful.	**Dirty data is unusable.**

Meaning of words in context (apple, pear)





Concept Search

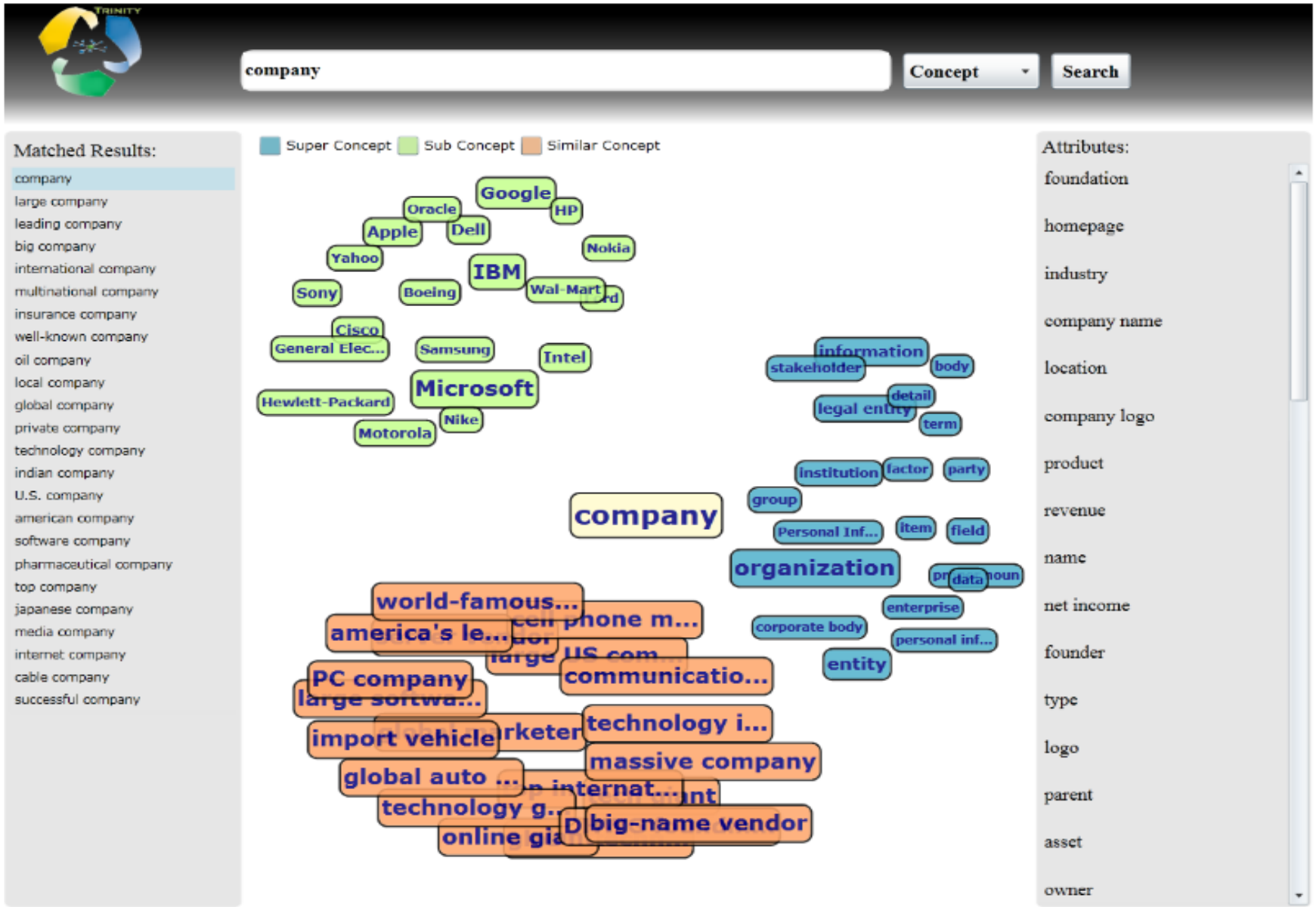
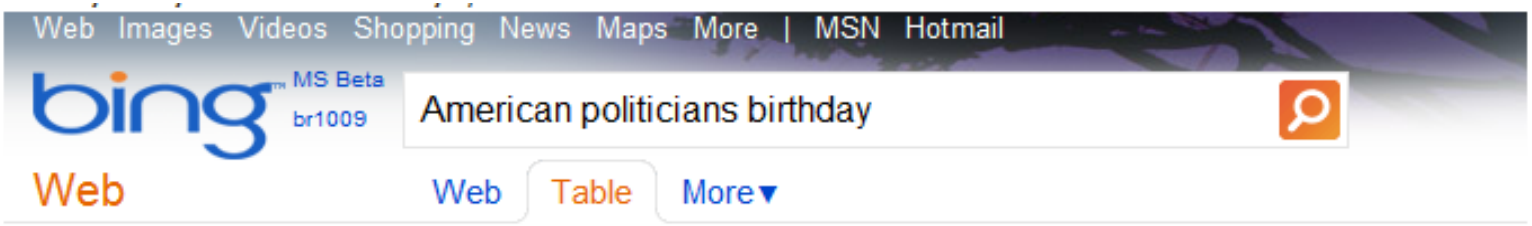


Table understanding



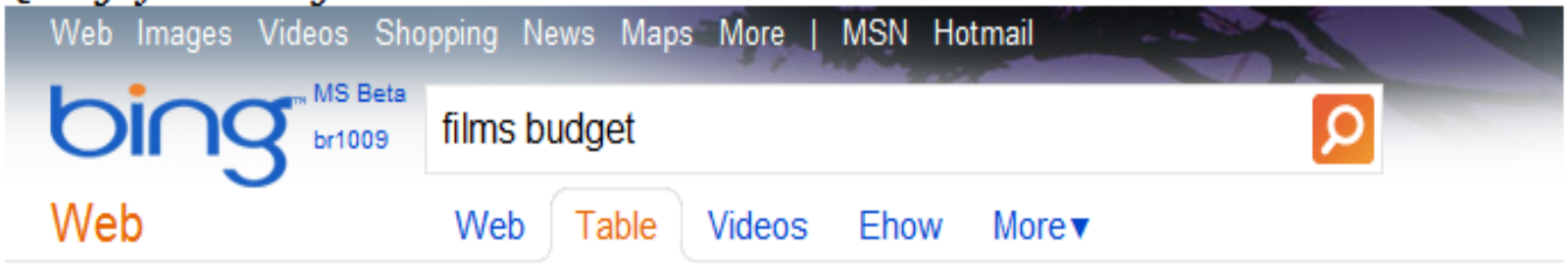
Shrink

Shrink table

Birth order	U.S. Vice President	Birthdate	Century	Order of office	Birthplace
39	Richard Nixon	January 9, 1913	20th	36	Yorba Linda , California
28	Theodore Roosevelt	October 27, 1858	19th	25	New York City , New York
46	Dan Quayle	February 4, 1947	20th	44	Indianapolis , Indiana
38	Hubert Humphrey	May 27, 1911	20th	38	Wallace , South Dakota
40	Gerald Ford	July 14, 1913	20th	40	Omaha , Nebraska
42	George H. W. Bush	June 12, 1924	20th	43	Milton , Massachusetts
44	Dick Cheney	January 30, 1941	20th	46	Lincoln , Nebraska
45	Joseph Biden	November 20, 1942	20th	47	Scranton , Pennsylvania
9	Martin Van Buren	December 5, 1782	18th	8	Kinderhook , New York

Table understanding

Query: films budget



Shrink

Shrink table

Year	Movie	Worldwide gross	**Budget**	Distributor	Director
1977	**Star Wars**	$ 782400000	$ 11000000	20th Century Fox	George Lucas
1997	**Titanic**	$ 1848813795	$ 200000000	Paramount Pictures	James Cameron
1993	**Jurassic Park**	$ 914691118	$ 95000000	Universal Studios	Steven Spielberg
1995	**Toy Story**	$ 365000000	$ 90000000	Walt Disney Pictures	John Lasseter
1972	**The Godfather**	$ 245066411	$ 6000000	Paramount Pictures	Francis Ford Coppola
2009	**Avatar**	$ 2606954237	$ 237000000	20th Century Fox	James Cameron
1975	**Jaws**	$ 470600000	$ 7000000	Universal Studios	Steven Spielberg
1996	**Independence Day**	$ 816969268	$ 75000000	20th Century Fox	Roland Emmerich
1998	**Armageddon**	$ 553709788	$ 140000000	Touchstone Pictures	Michael Bay

Summary

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- What else can industries do to help democratize semantic computing research?

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Acknowledgments

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