



Microsoft® Research

FacultySummit 2011

Cartagena, Colombia | May 18-20 | In partnership with COLCIENCIAS



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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



“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.

The New York Times, Aug 2006

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

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>

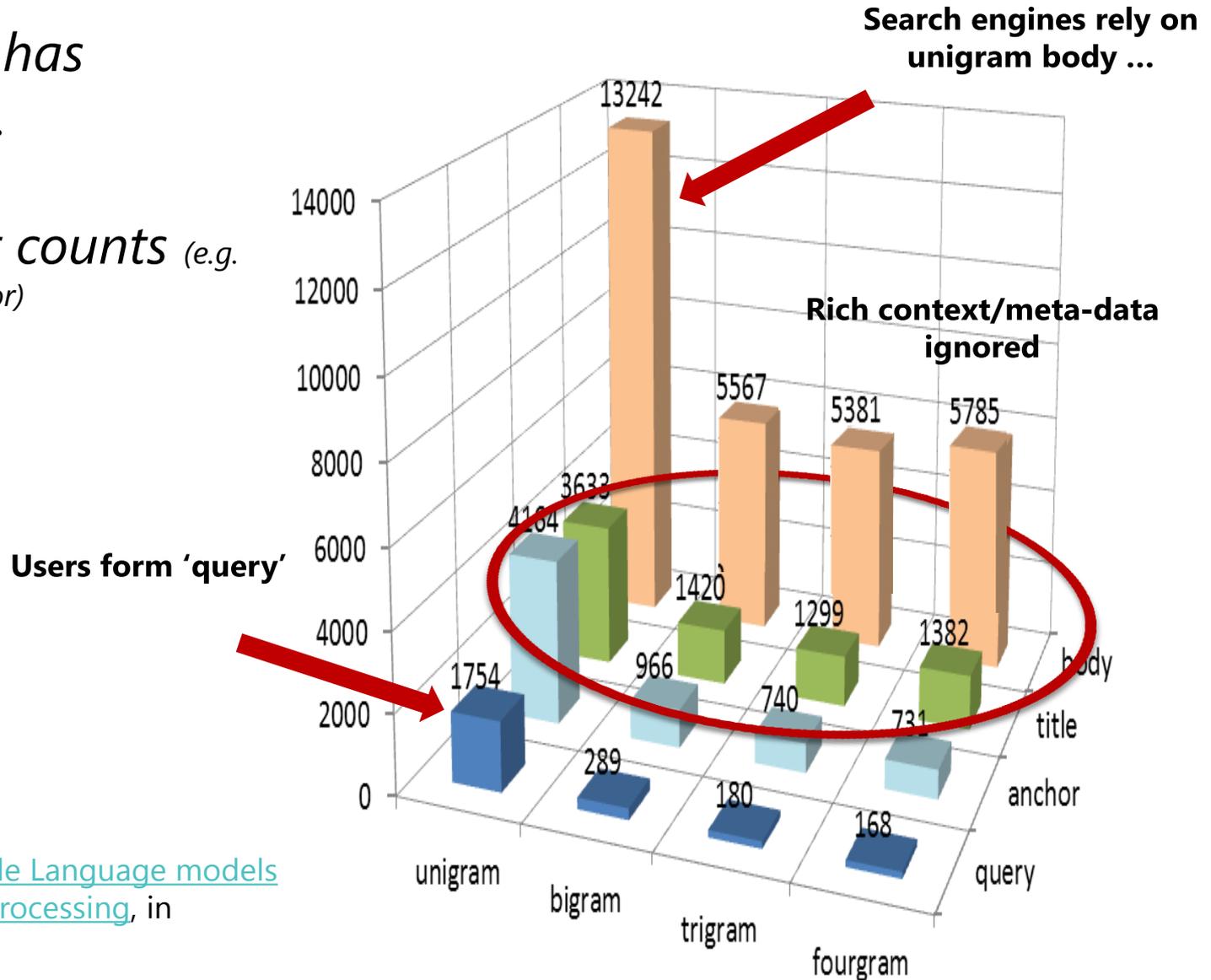
Leading technology in Search, Machine Translation,
Speech, Learning



Web N-Gram in Public Beta

Web data has structure...

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



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 #ripherunway #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 #Hereyugo #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 #dearfuturwife #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	Phrase	LgProbability
now playing	~	buenos dias	-7.412106
nowplaying	~	buenosdias	-9.258749
n ow playing	~	b uenos dias	-10.89817
now play in g	~	buenos dia s	-10.92766
now play ing	~	buenos di as	-10.96579

Total Time (ms): 312
Service Time (ms): 270

#parlezvousfrancais

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

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

VINOGRAPHY: a wine blog
Wine and food adventures in San Francisco and around the world

WINE REVIEWS | RESTAURANT REVIEWS | BOOK REVIEWS | RAMBLINGS & RANTS | WINE NEWS

Breaking Wine News: Bordeaux's Cos d'Estournel Buys Napa's Chateau Montelena

To those of you in the wine world paying attention to the dollar's stomach churning lows against the Euro, this news may come as little or no surprise. This morning, Chateau Cos d'Estournel announced its purchase of the historic **Chateau Montelena** in Napa. While not the first bit of investment from Bordeaux in the **Napa Valley** is certainly a significant one, given both the landmark historical status of Chateau Montelena as well as the prestige and success of Cos d'Estournel, whose star has certainly been rising in Bordeaux over the past decade.

Montelena became a world famous winery after its 1973 Chardonnay beat out French competitors in the famous **Judgement of Paris**

1976

Wine Spectator

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

'Chateau Montelena in Napa'
segmentation

article | discussion | edit this page | history

Chateau Montelena

From Wikipedia, the free encyclopedia

Coordinates: 38°52′16″N 122°59′13″W﻿ / ﻿38.87111°N 122.98750°W﻿ / 38.87111; -122.98750

Chateau Montelena is a Napa Valley winery most famous for winning the white wine section of the historic "Judgement of Paris" wine competition. Chateau Montelena's Chardonnay was in competition with nine other wines from France and California under blind testing. All 11 judges awarded their top scores to either the Chardonnays from Chateau Montelena or Chateau Winery, another Californian wine producer. Chateau Montelena was featured in the 2008 film *Bottle Shock*.

Contents [hide]

- History
 - 1.1 Terminated sale
- See also
- References
- External links

Chateau Montelena	
Location	Caldoga, California, USA
Appetation	Napa Valley AVA
Other labels	Potter Valley
Founded	1882
Key people	Jim Barrett, Vitner Bo Barrett, Winemaker Greg Ralston, Managing Director Dave Vella, Vineyard Manager
Cases/yr	30,000 - 36,000
Varietals	Chardonnay, Zinfandel, Cabernet Sauvignon, Riesling
Website	www.montelena.com#

'Chateau Montelena' as an
entity
in Wikipedia

Spelling Alteration for Web Search

<http://spellerchallenge.com>

SPELLER CHALLENGE

BING – MICROSOFT RESEARCH PARTNERSHIP

Speller Challenge

[Home](#) | [My Team](#) | [Rules](#) | [FAQ](#) | [Datasets](#) | [Leaderboard](#)



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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
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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

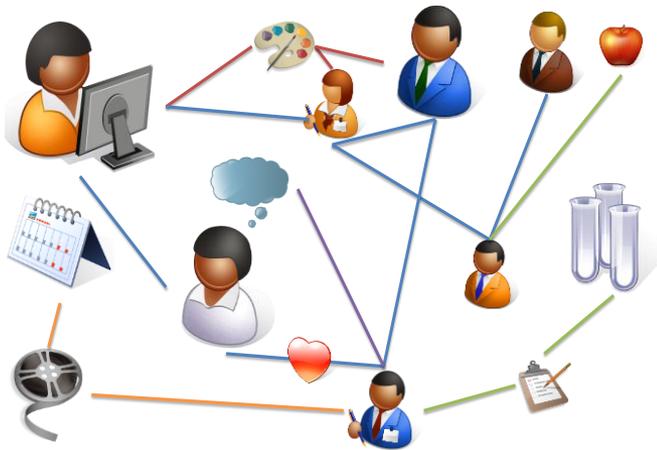
- From data release to data services
 - Allows to handle much bigger data sets
 - Allows to provide abstractions on the data (e.g. language models)
 - Allows to provide data compute capabilities
 - Allows for agile experimentation
- How to better engage with academia to drive data-driven research?
 - 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

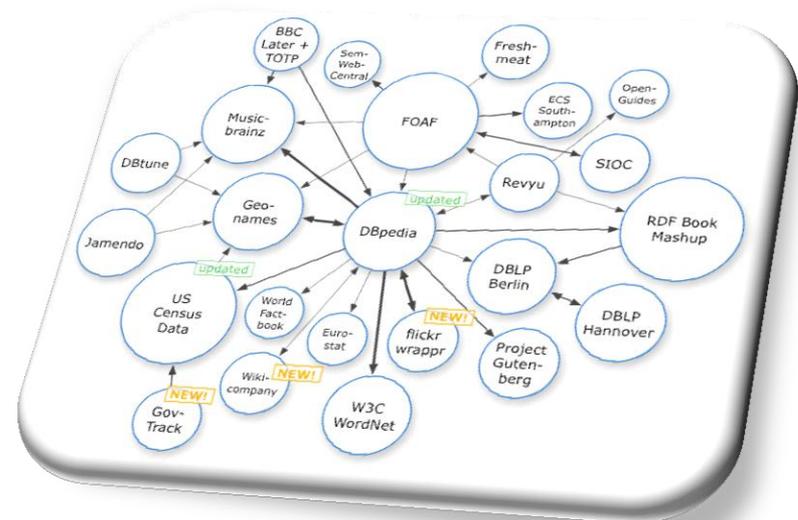
TOWARDS A KNOWLEDGE WORLD

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](#)

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

- 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. **Freebase**

Correctness is a probability.

Live with dirty data.

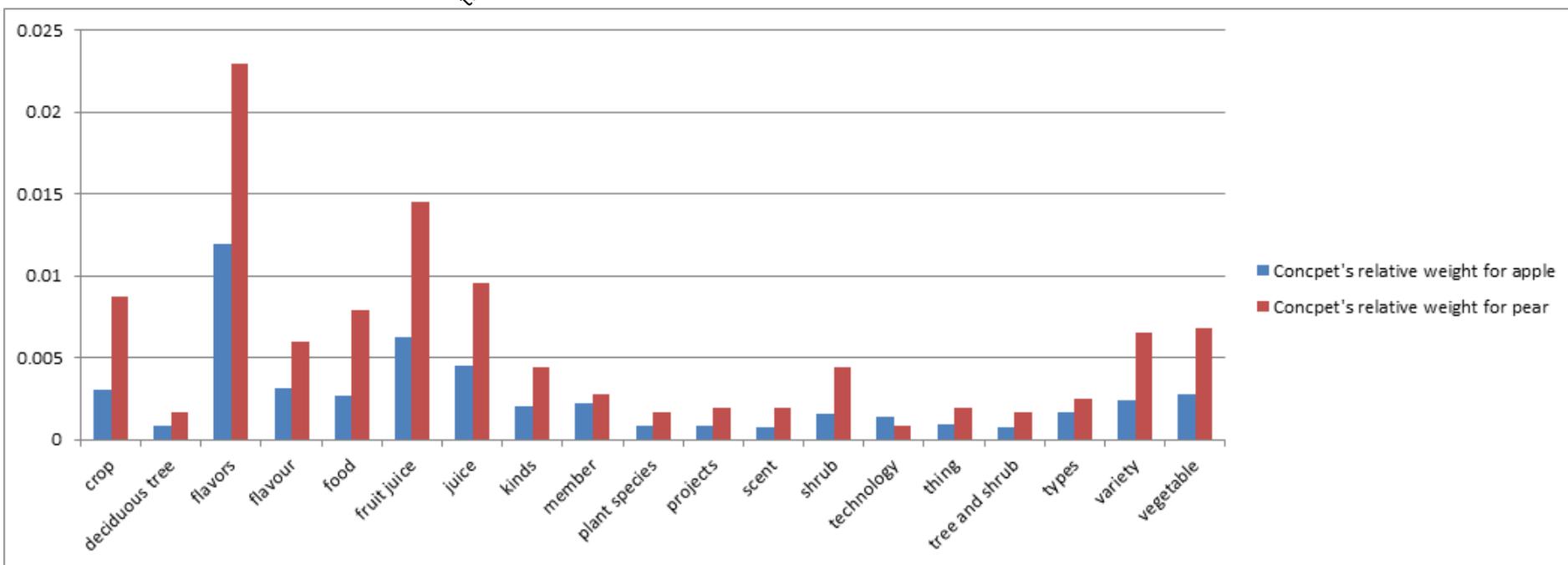
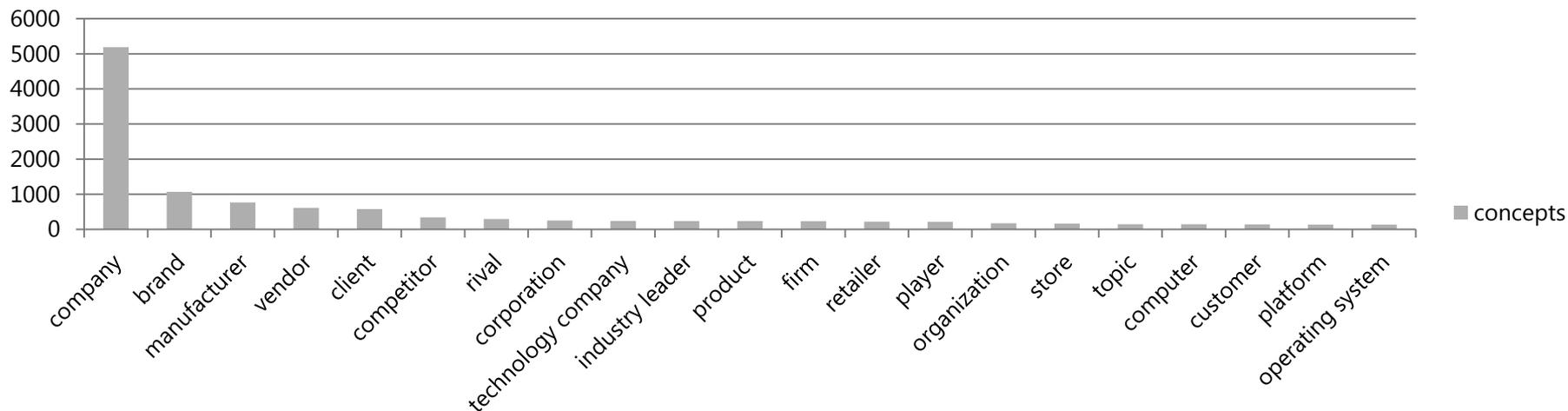
Dirty data is very useful.

Knowledge is black and white.

Clean up everything.

Dirty data is unusable.

Meaning of words in context (apple, pear)



Concept Search

The screenshot displays a web interface for a concept search tool. At the top left is a logo with the word "TRINITY" above it. A search bar contains the text "company". To the right of the search bar are two buttons: "Concept" with a dropdown arrow and "Search".

Below the search bar, on the left, is a section titled "Matched Results:" containing a list of related terms such as "large company", "leading company", "big company", "international company", "multinational company", "insurance company", "well-known company", "oil company", "local company", "global company", "private company", "technology company", "indian company", "U.S. company", "american company", "software company", "pharmaceutical company", "top company", "japanese company", "media company", "internet company", "cable company", and "successful company".

In the center is a word cloud visualization. A legend at the top of the word cloud indicates three categories: "Super Concept" (blue), "Sub Concept" (green), and "Similar Concept" (orange). The word "company" is the largest and is highlighted in yellow. Other prominent words include "organization" (blue), "information" (blue), "stakeholder" (blue), "legal entity" (blue), "entity" (blue), "group" (blue), "Personal Inf..." (blue), "item" (blue), "field" (blue), "institution" (blue), "factor" (blue), "party" (blue), "Dr data noun" (blue), "corporate body" (blue), "enterprise" (blue), "personal inf..." (blue), "world-famous..." (orange), "america's le..." (orange), "cell phone m..." (orange), "large US com" (orange), "PC company" (orange), "communicatio..." (orange), "large softwa..." (orange), "import vehicle" (orange), "arketer" (orange), "technology i..." (orange), "massive company" (orange), "global auto ..." (orange), "internat..." (orange), "technology g..." (orange), "big-name vendor" (orange), and "online gi..." (orange).

On the right side of the interface is a vertical list of attributes, each with a small upward-pointing arrow to its right. The attributes listed are: foundation, homepage, industry, company name, location, company logo, product, revenue, name, net income, founder, type, logo, parent, asset, and owner.

Table understanding

Web Images Videos Shopping News Maps More | MSN Hotmail

bing[™] MS Beta
br1009

American politicians birthday

Web Table More ▾

– 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*

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br1009

films budget 

Web **Table** Videos Ehow More ▾

– 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

- Data to drive innovation
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- How to better engage with academia to drive semantic computing research?
 - How does a cloud-based data and knowledge service approach change research?
- What else can industries do to help democratize semantic computing research?

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Acknowledgments

Jianfeng Gao, Zoubin Ghahramani, Mark Greaves,
Savas Parastatidis, Chris Thrasher, Haixun Wang,
Kuansan Wang, ...



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