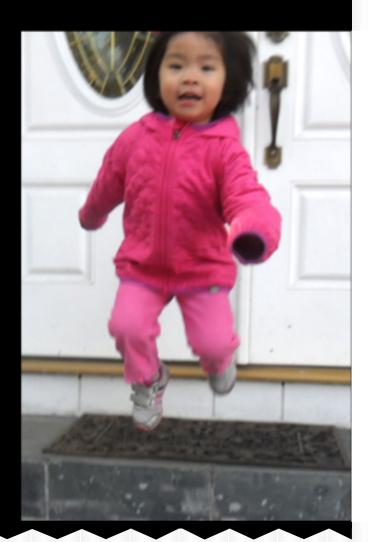
From Data Fusion to Knowledge Fusion

Xin Luna Dong, Google Inc. 9/6/2014 (a) APWeb'14

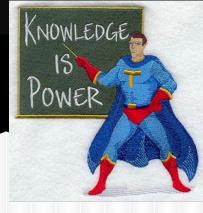


SONYA: A Big Project, A Fancy Machine, And A Cute Little Girl





Knowledge Is Power



Many Knowledge Bases (KB)











NELL: Never-Ending Language Learning









Google Knowledge Graph



Using KG in Search

Google



Web News Maps

Shopping

Images

More + Search tools

About 2,960,000 results (0.27 seconds)

岳麓书院:首页

ylsy.hnu.cn/ ▼ 轉為繁體網頁 Yuelu Academy ▼

在古木参天、浓荫蔽日、山光水色的湘江西岸,有一片典雅、庄重的古建筑群,这就是 为世人所瞩目的"四大书院"之一的岳麓书院。

4.5 ★★★★★ 22 Google reviews · Write a review

China, Hunan, Changsha, Yuelu, 麓山路 +86 731 8882 3764

岳麓书院-维基百科,自由的百科全书

zh.wikipedia.org/zh-hant/岳麓书院 * 轉為繁體網頁 Chinese Wikipedia *

岳麓书院位于中国湖南省长沙市岳麓山东麓,是中国古代四大书院之一,始建于北宋开 宝九年(976年),历经宋、元、明、清各个朝代,迨及晚清(1903年)改为湖南高等... 历史沿革 - 大事年表 - 书院学规 - 历代山长

News for 岳麓

"小社区大民生"领创和谐岳麓

光明网 - 5 hours ago

在坐拥15个街道2个镇、81个社区的岳麓区,其中有1个街道、2个社区荣获全国和谐社 区建设示范单位,有3个街道、24个社区分别荣获省、市精品、...

做客乘坐缆车高空扔垃圾长沙岳麓山很"受伤"

红网 - 21 hours ago

More news for 岳麓

岳麓书院_百度百科

baike.baidu.com/view/7288.htm ▼ 轉為繁體網頁 Baidu Baike ▼

北宋开宝九年(976),潭州太守朱洞在僧人办学的基础上,正式创立岳麓书院。嗣 后,历经宋、元、明、清各代,至清末光绪二十九年(1903)改为湖南高等学堂,尔后



岳麓書院

Directions

Write a review

Address: China, Hunan, Changsha, Yuelu, 麓山路

Phone: +86 731 8882 3764

Reviews

4.5 ★★★★★ 22 Google reviews

Are you the business owner?

Feedback

See results about

Yuelu District

Yuelu District is one of six urban districts of Changsha, the capital of Hunan province, China. It ...

Expanding KG— Extracting Knowledge from the Web

"In the near future, the web is going to be the master copy of human knowledge. We need to figure out ways to use that knowledge."

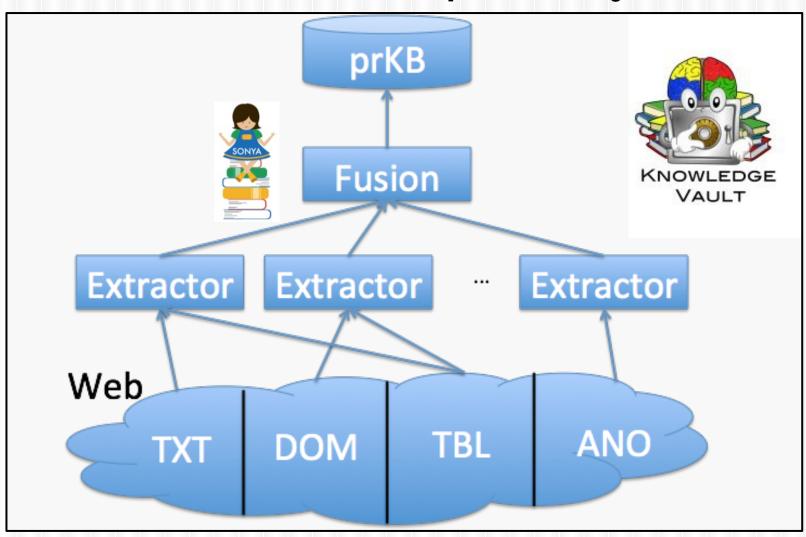
-Håkon Wium Lie

But-

- KG requires 99% accuracy for knowledge
- Web data is noisy and extraction is hard
- How to balance coverage and accuracy?

Knowledge Vault– Building a Probabilistic KB

[VLDB'2014, Sigmod'2014, KDD'2014]





News

Google's Knowledge Vault already contains 1.6 billion facts

FELICITY NELSON SATURDAY, 23 AUGUST 2014









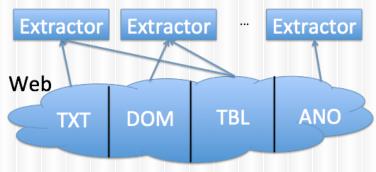


The automated, fact-harvesting bot will build up a collection of all human knowledge.

Outline



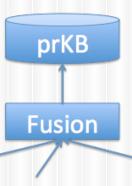
Knowledge extraction







Knowledge fusion





Interesting applications



Future directions

Knowledge Extraction I–Knowledge

- Triple: (subject, predicate, object) e.g., (Tom Cruise, date_of_birth, 7/3/1962)
 - Subject—a Freebase mid
 e.g., /m/07r1h
 - Predicate—predefined in Freebase; e.g., people/person/date_of_birth
 - Object—a Freebase mid, a string, a number, or a date.

Statistics for Extracted Triples

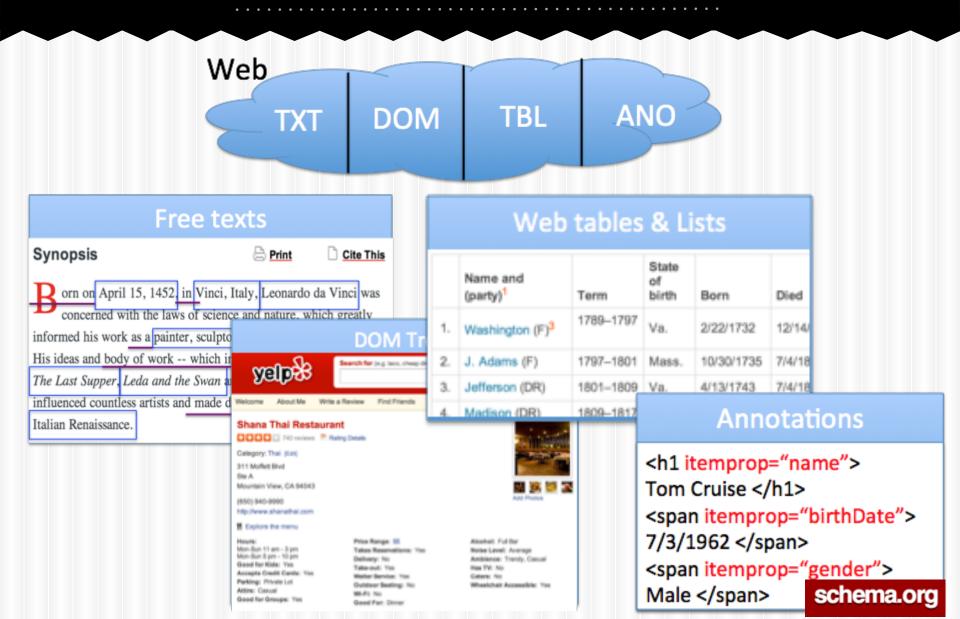
A large knowledge base

As of 11/2013

#Triples	1.6B (now 2.8B)
#Subjects (Entities)	43M
#Types	1.1K
#Predicates	4.5K
#Objects	102M

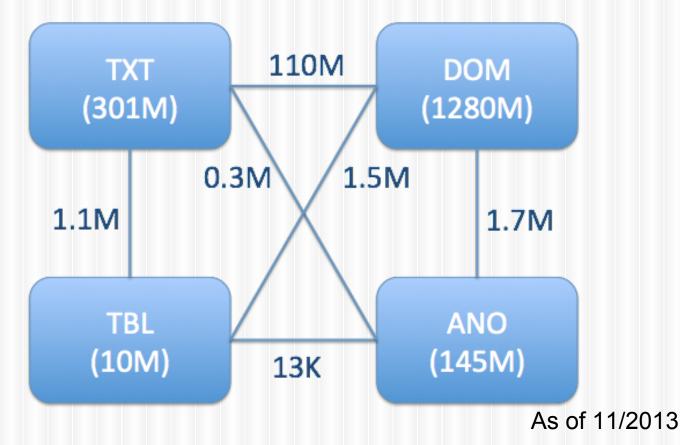
- Highly skewed data—fat heads, long tail
 - #Triples/type: 1–14M
 (location, organization, business)
 - o #Triples/entity: 1–2M (USA, UK, CA, NYC, TX)

Knowledge Extraction II—Sources



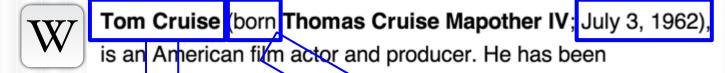
Statistics for Web Sources

- 1B+ Webpages over the Web
- Contribution is skewed: 1-50K



Knowledge Extraction III–Extractors

- Three tasks (any order, maybe combined)
 - I. Triple identification



II. Entity linkage

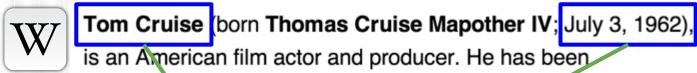


/people/person/date_ of_birth

III. Predicate linkage

Knowledge Extraction III–Extractors

Texts/DOM: distant supervision





Date of birth /people/person/date_of_birth

7/3/1962

Pattern 1: X "born" Y

→ (X, /people/person
/date of birth, Y)

- Web tables/lists: schema mapping
- Annotations: semi-automatic mapping

Statistics for Extractors

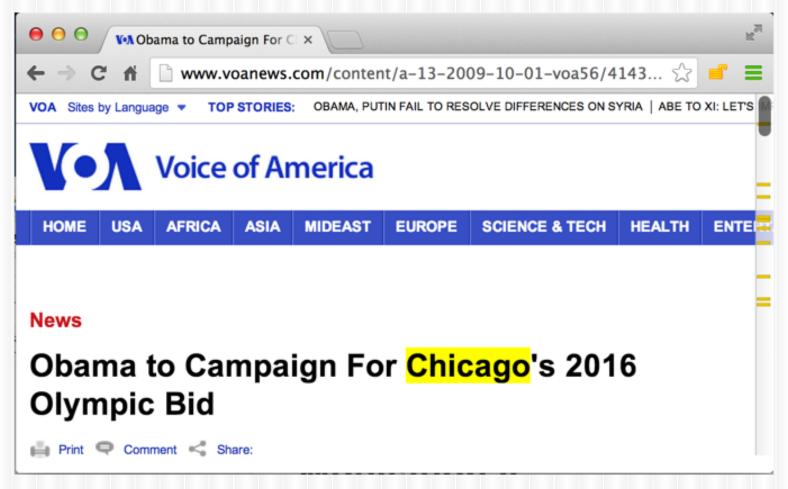
• 12 extractors; high variety

	#Triples	#Webpages	#Patterns	Accu	Accu (conf ≥ .7)
TXT1	274M	202M	4.8M	0.36	0.52
TXT2	31M	46M	3.7M	0.18	0.80
TXT3	8.8M	16M	1.5M	0.25	0.81
TXT4	2.9M	1.2M	0.1M	0.78	0.91
DOM1	804M	344M	(25.7M)	0.43	0.63
DOM2	431M	925M	No pat.	0.09	0.62
DOM3	45M	N/A	No pat.	0.58	0.93
DOM4	52M	7.8M	No pat.	0.26	0.34
DOM5	0.7M	0.5M	No pat.	0.13	No conf.
TBL1	3.1M	0.4M	No pat.	0.24	0.24
TBL2	7.4M	0.1M	No pat.	0.69	No conf.
ANO	145M	53M	No pat.	0.28	0.30

As of 11/2013

Errors Can Creep in at Every Stage

Extraction error: (Obama, nationality, Chicago)



Errors Can Creep in at Every Stage

Reconciliation error: (Obama, nationality, North America)



American
President
Barack Obama

influences from all over the world. There's a touch of this, a little smidgeon of that, a dash of something else, like when you're cooking." On the song "The Queen Is Back", Summer reveals her wry and witty self-awareness of her musical legacy and her public persona. "I'm making fun of myself," she admits. "There's irony. It's poking fun at the idea of being called a queen. That's a title that has followed me, followed me and followed me. We were sitting and writing and that title kept popping up in my mind and I'm thinking, 'Am I supposed to write this? Is this too arrogant to write?' But people call me 'the queen,' so I guess it's ok to refer to myself as what everybody else refers to me as. We started writing the song and thought it was kind of cute and funny." Summer wrote "The Queen Is Back" and "Mr. Music" with J.R. Rotem and Evan Bogart, the son of Casablanca Records founder Neil Bogart.

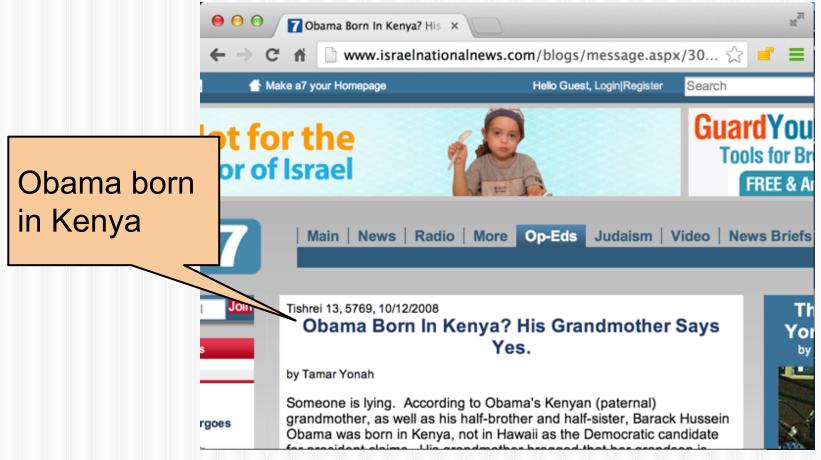
On Sember 11, 2009, Summer performed at the Nobel Peace Prize Concert in Oslo, Norway in honor of American President Barack Obama. She was backed by the Norwegian Radio Orchestra.

2010-12: Final recordings [edit source | edit beta]

On July 29, 2010, Summer gave an interview with Allvoices.com wherein she was asked if she would consider doing an album of standards. She said, I actually am, probably in September. I will begin work on a standards album. I will probably do an all-out dance album and a standards

Errors Can Creep in at Every Stage

Source data error: (Obama, nationality, Kenya)



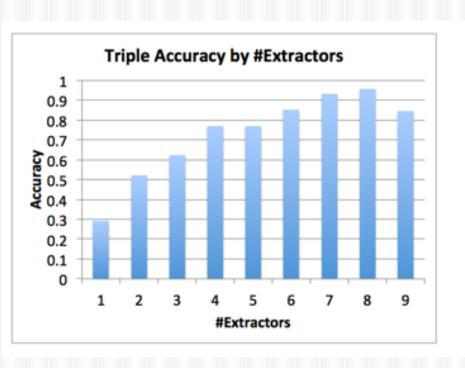
Knowledge Extraction IV—Quality

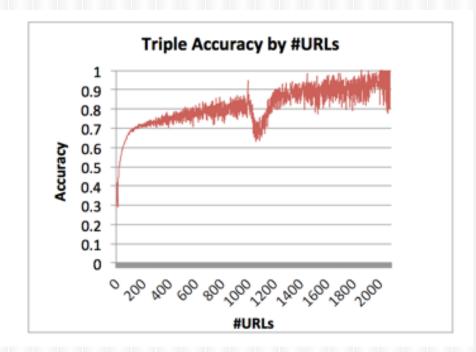
- Gold standard: Freebase
- LCWA (Local Closed-World Assumption)
 - o If (s,p,o) exists in FB: true
 - o Otherwise,
 - If (s,p) exists in FB: false (Freebase knowledge is locally complete)
 - Otherwise: UNKNOWN
- The gold standard contains about 40% of the triples

Statistics for Triple Correctness

- Overall accuracy: 30%
- Random sample on 25 false triples
 - Triple-identification errors: 11 (44%)
 - Entity-linkage errors: 11 (44%)
 - Predicate-linkage errors: 5 (20%)
 - Source-data errors: 1 (4%)

Statistics for Triple Correctness





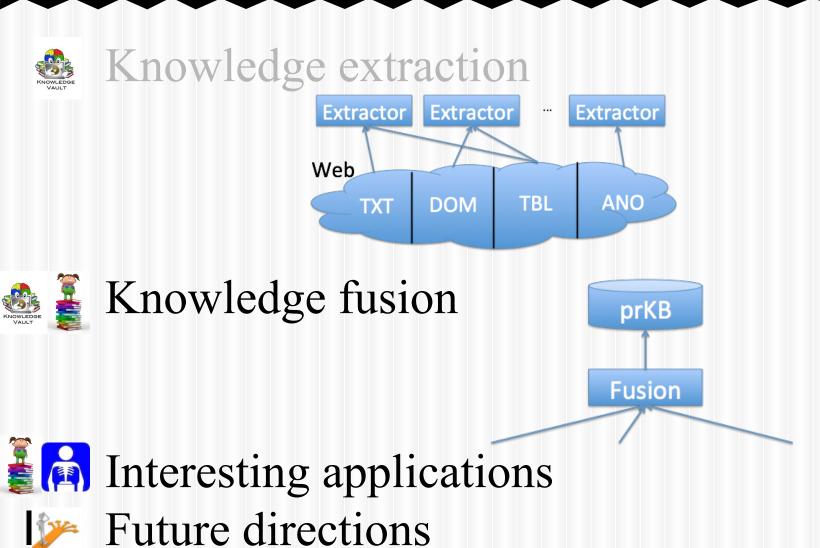
Statistics for Extractors

• 12 extractors; high variety

	#Triples	#Webpages	#Patterns	Accu	Accu (conf ≥ .7)
TXT1	274M	202M	4.8M	0.36	0.52
TXT2	31M	46M	3.7M	0.18	0.80
TXT3	8.8M	16M	1.5M	0.25	0.81
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TBL1	3.1M	0.4M	No pat.	0.24	0.24
TBL2	7.4M	0.1M	No pat.	0.69	No conf.
ANO	145M	53M	No pat.	0.28	0.30

As of 11/2013

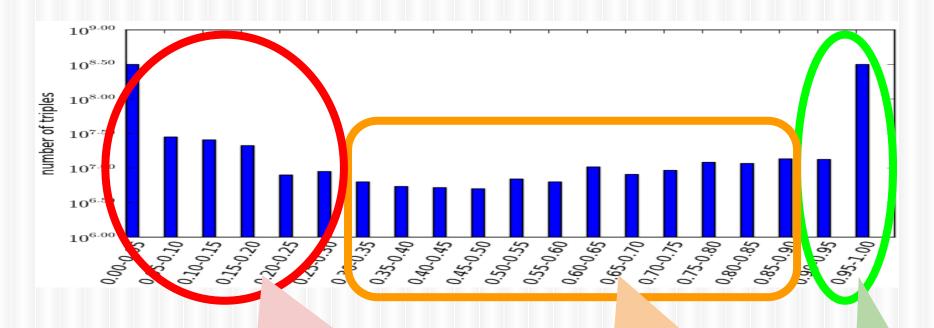
Outline



Goal: Judge Triple Correctness

- Input: Knowledge triples and their provenances (i.e., which extractor extracts from which source)
- Output: a probability in [0,1] for each triple
 - Probabilistic decisions
 vs. deterministic decisions

Usage of Probabilistic Knowledge

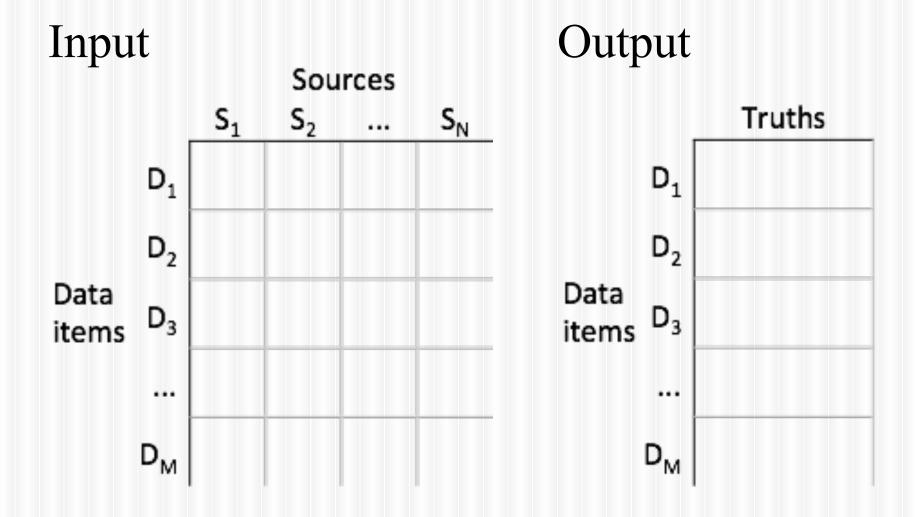


Negative training examples, and MANY EXCITING APPLICATIONS!!

Active learning, probabilistic inference, etc.

Upload to KG

Data Fusion-Definition



	Src1	Src2	Src3
Jagadish	UM	ATT	UM
Dewitt	MSR	MSR	UW
Bernstein	MSR	MSR	MSR
Carey	UCI	ATT	BEA
Franklin	UCB	UCB	UMD

	Src1	Src2	Src3
Jagadish	UM	ATT	UM
Dewitt	MSR	MSR	UW
Bernstein	MSR	MSR	MSR
Carey	UCI	ATT	BEA
Franklin	UCB	UCB	UMD

Voting--Trust the majority.

	Src1	Src2	Src3
Jagadish	UM	ATT	UM
Dewitt	MSR	MSR	UW
Bernstein	MSR	MSR	MSR
Carey	UCI	ATT	BEA
Franklin	UCB	UCB	UMD

	Src1	Src2	Src3
Jagadish	UM	ATT	UM
Dewitt	MSR	MSR	UW
Bernstein	MSR	MSR	MSR
Carey	UCI	ATT	BEA
Franklin	UCB	UCB	UMD

Quality-based--Give higher votes to more accurate sources.

Data Fusion-A Bayesian Model

[Dong et al., VLDB'09]

Q1. How to compute source accuracy?

Source Accuracy: A(S)

$$A(S) = \underset{v \in \overline{V}(S)}{Avg} P(v)$$

- \circ $\overline{V}(S)$ values provided by S
- P(v) pr of value v being true

Data Fusion-A Bayesian Model

[Dong et al., VLDB'09]

Q2. How to leverage accuracy in voting?

Input:

- -Data item D
- -Dom(D)= $\{v_0, v_1, ..., v_n\}$
- -Observation Φ on D

Output:

Pr(v_i true| Φ) for each i=0,..., n (sum up to 1)

According to the Bayes Rule, we need to know $Pr(\Phi|v_i true)$

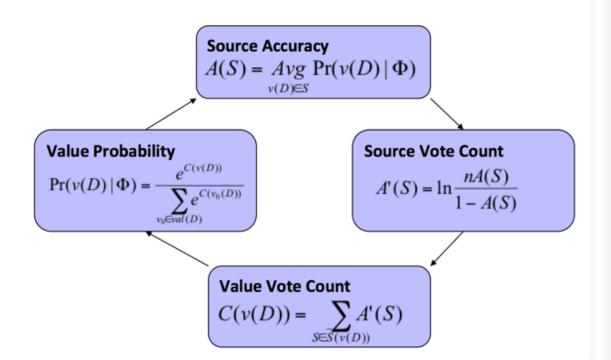
- Assuming independence of sources, we need to know Pr(Φ(S) |v_i true)
- If S provides v_i:
 Pr(Φ(S) |v_i true) =A(S)
- If S does not provide v_i:
 Pr(Φ(S) |v_i true) =(1-A(S))/n

Data Fusion-A Bayesian Model

[Dong et al., VLDB'09]

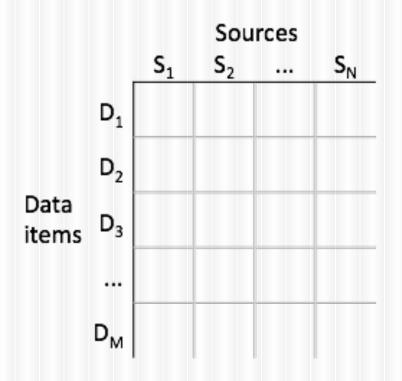
Q3. How to handle interdependence between source accuracy and value pr?

Continue until source accuracy converges

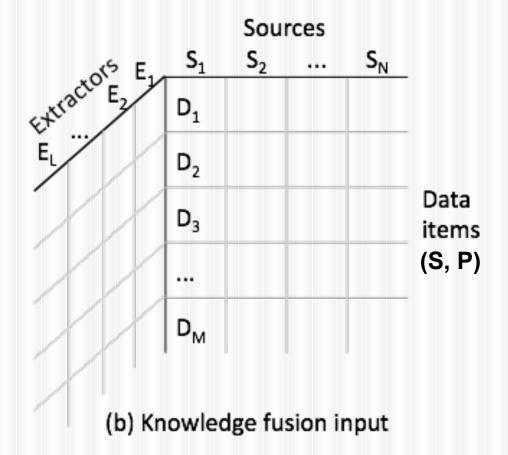


Knowledge Fusion Challenges

I. Input is three-dimensional

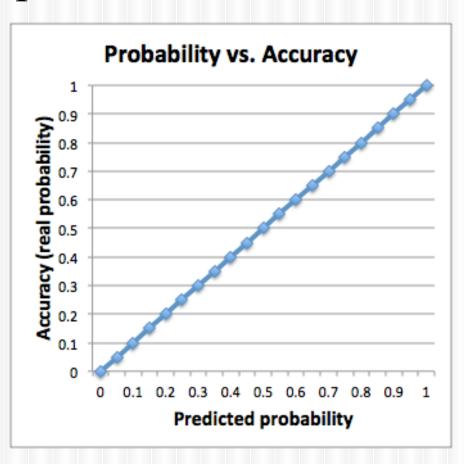


(a) Data fusion input



Knowledge Fusion Challenges

II. Output prs should be well-calibrated



Knowledge Fusion Challenges

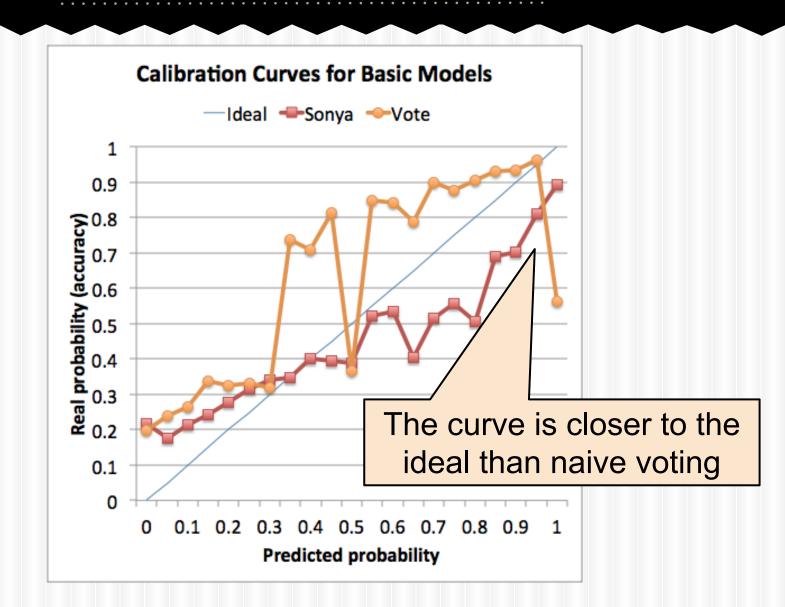
III. Data are of Web-scale

- Three orders of magnitude larger than currently published data-fusion applications
 - o Size: 1.1TB
 - \circ Sources: 170K \rightarrow 1B+
 - \circ Data items: 400K \rightarrow 375M
 - \circ Values: 18M \rightarrow 6.4B (1.6B unique)
- Data are highly skewed
 - o #Triples/Data-item: 1 2.7M
 - #Triples/Source: 1 50K

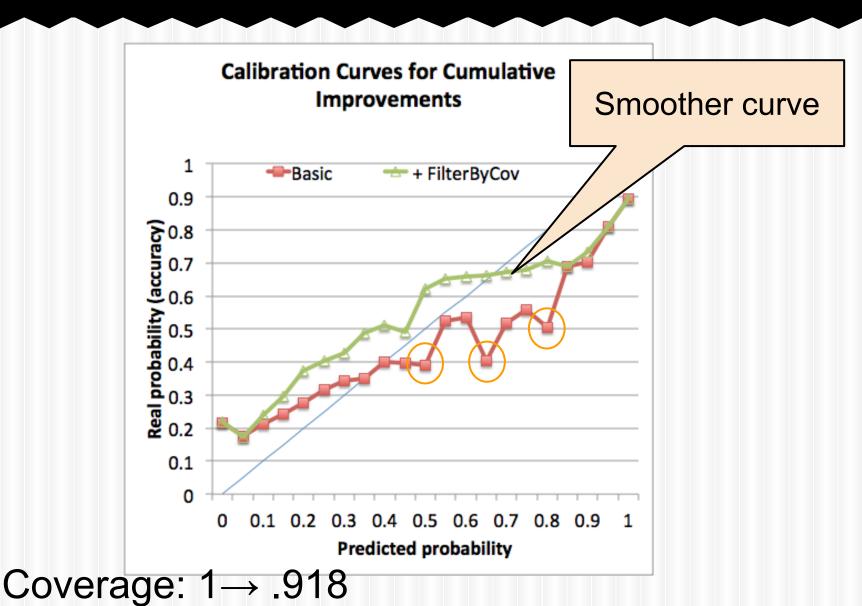
Knowledge Fusion Solutions

- Treat each (URL, Extractor) as a whole (provenance) for accuracy evaluation
- A series of refinements to improve probability calibration
- MapReduce Based Framework
 - Terminate in 5 rounds
 - Sample for too big data items or provenances

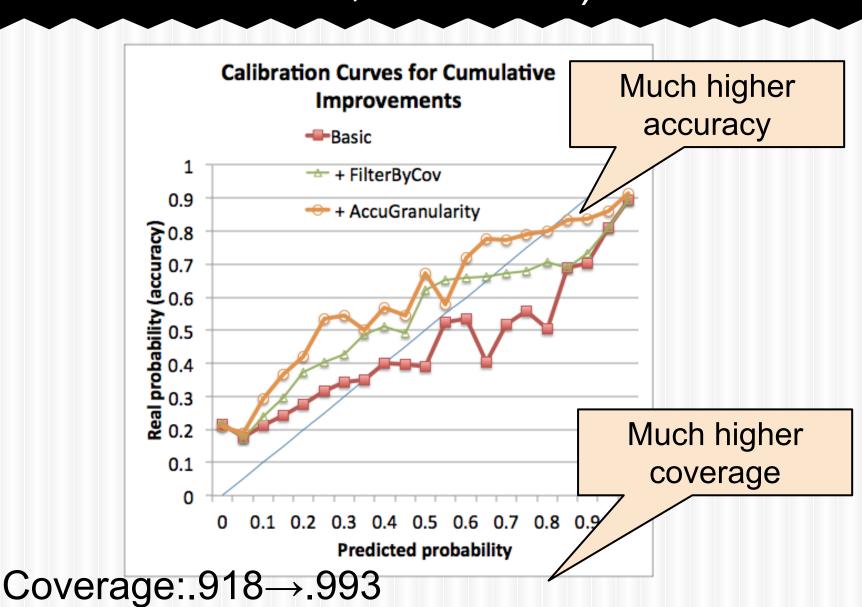
Basic Sonya Solution vs. Voting



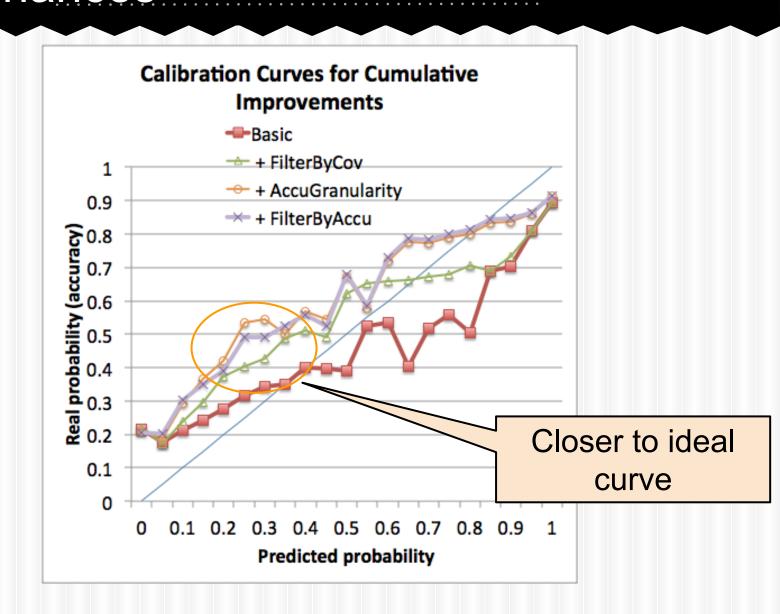
Refinement I. Ignore Low-Coverage Provenances



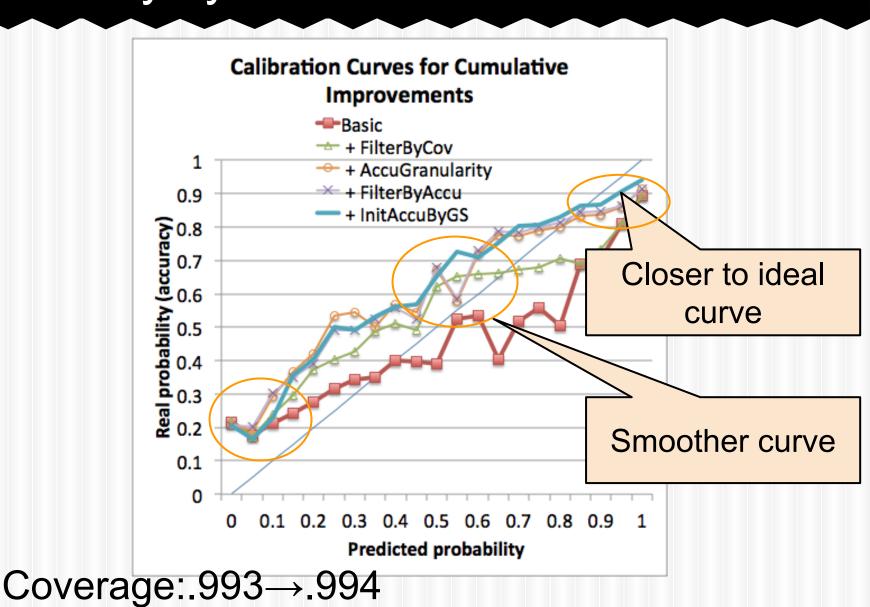
Refinement II. Granularity (URL->Site, Extractor->Pattern, Predicate)



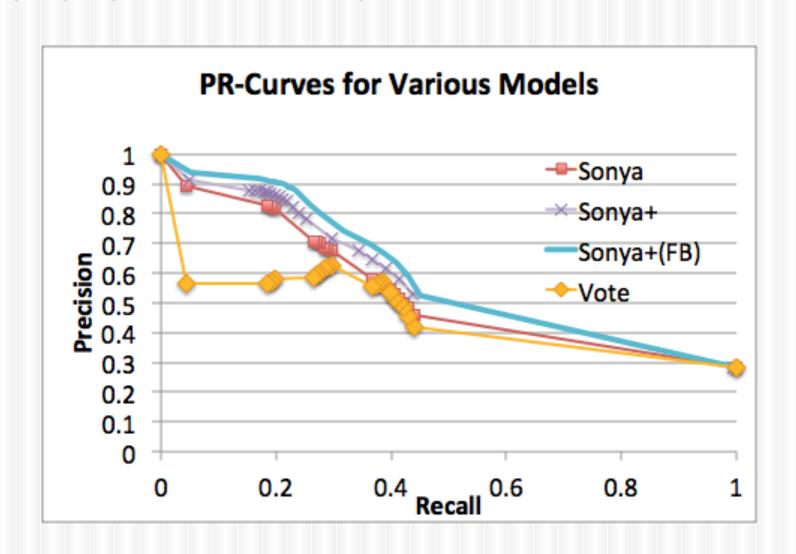
Refinement III. Ignore Low-Accuracy Provenances



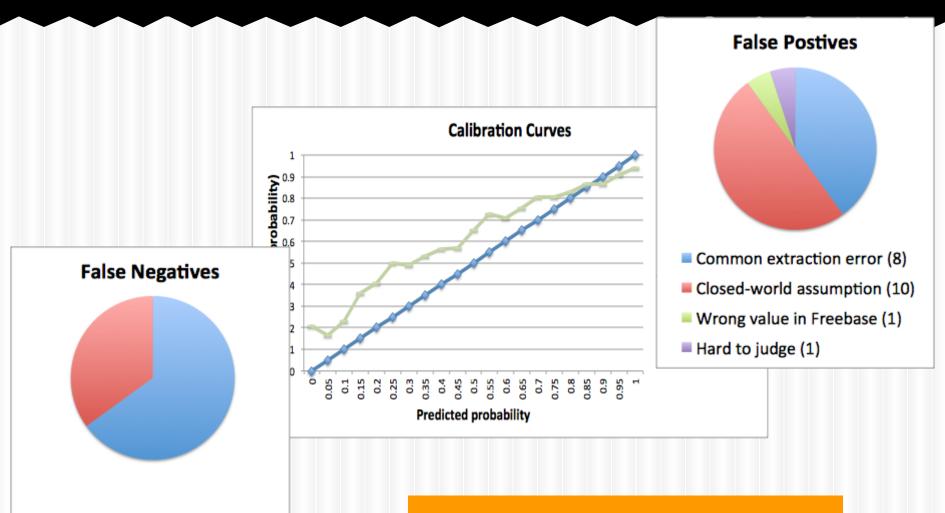
Refinement IV. Initiate Provenance Accuracy by FB



Precision-Recall Curve



Analysis of Errors



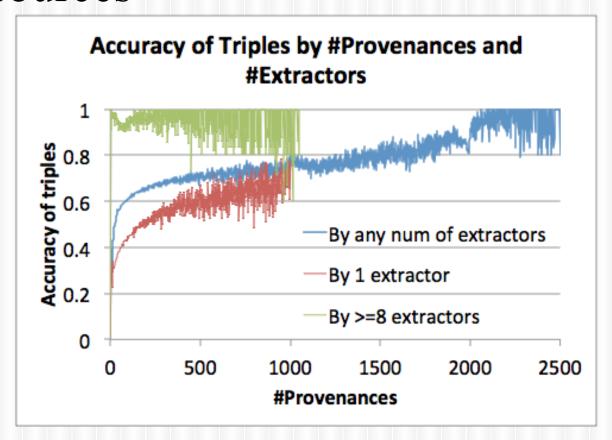
Multiple truths (13)

Specific/general value (7)

Future Directions!!!

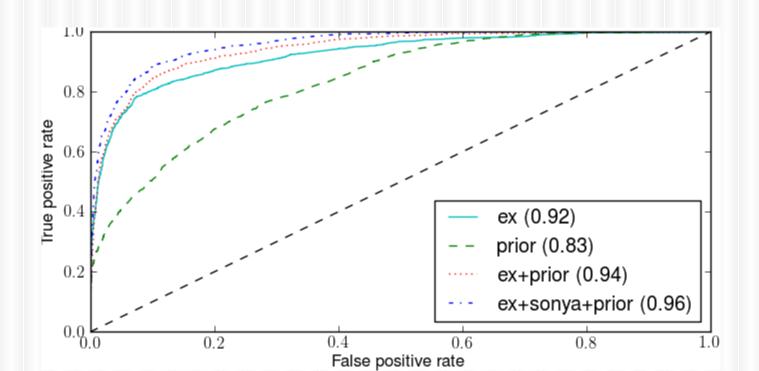
One Inherent Limitation

Cannot distinguish errors from extractors and from sources

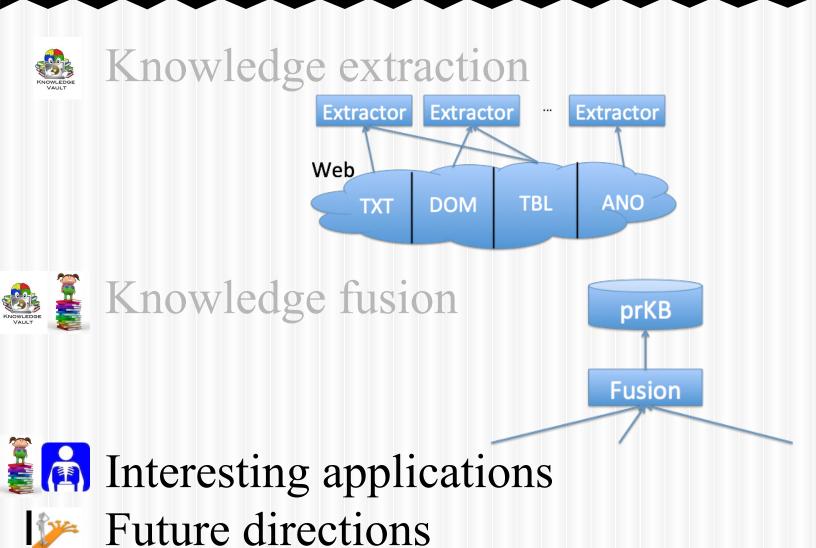


Other Fusion Techniques

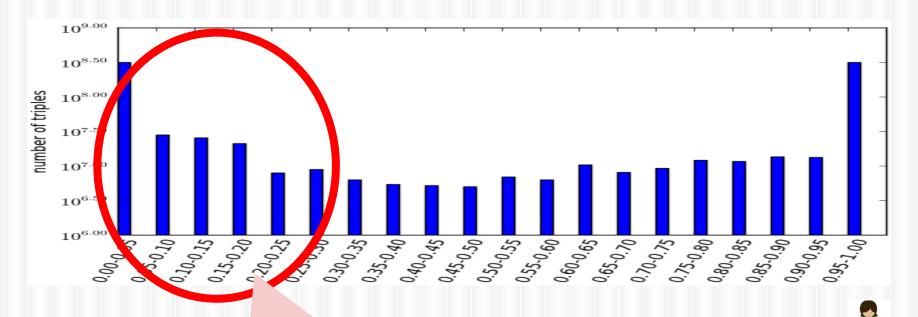
- Ex: Adaboost learning from extractions
- Prior: (A, parent_of, C), (B, parent_of, C)
 → (A, spouse of, B)



Outline



Usage of Probabilistic Knowledge



Negative training examples, and MANY EXCITING APPLICATIONS!!

- Source errors:
 trustworthiness evaluation
- Extraction errors: data abnormality diagnosis

Application I. A New Angle to Evaluate Web Source Quality

What we have now

- Page Rank: links between Websites/Webpages
- Log based: search log and clickthrough rate
- Web spam
- o etc.

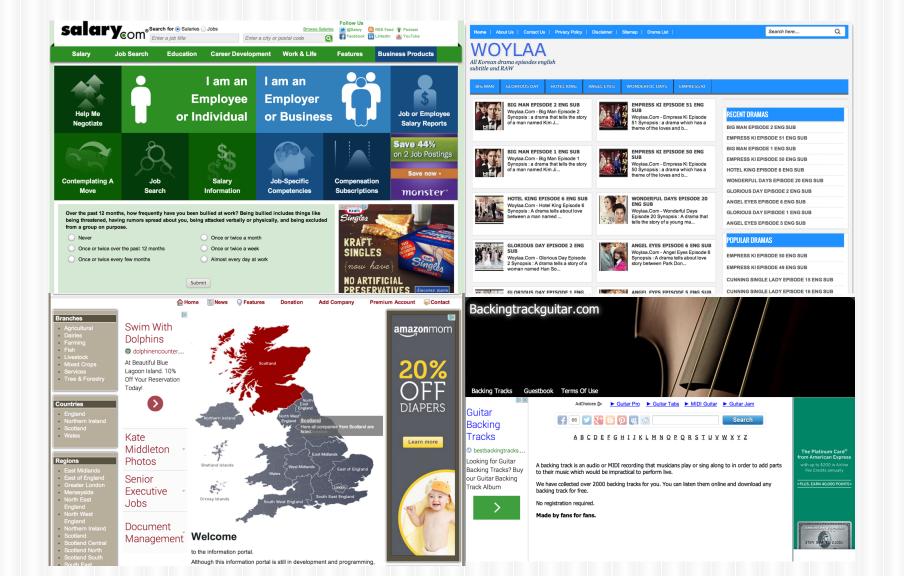
Popular Sources w. High Page Rank May Spread Gossip

14 out of 15 Gossip Websites have high page rank

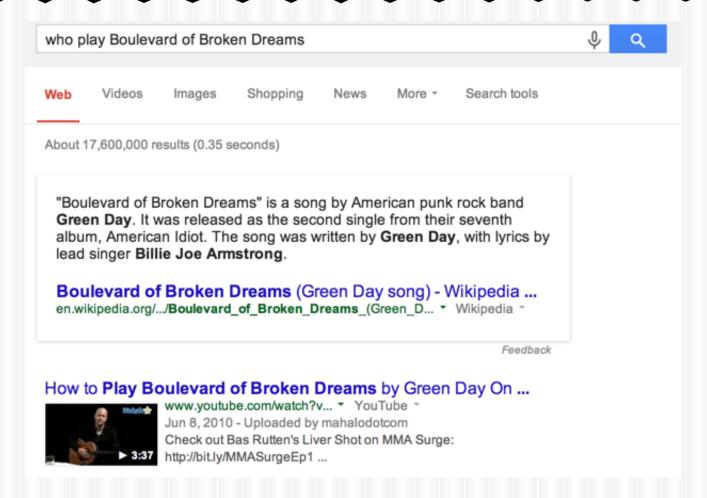
http://www.ebizmba. com/articles/gossip-websites

Domain			
www.eonline.com			
perezhilton.com			
radaronline.com			
www.zimbio.com			
mediatakeout.com			
gawker.com			
www.popsugar.com			
www.people.com			
www.tmz.com			
www.fishwrapper.com			
celebrity.yahoo.com			
wonderwall.msn.com			
hollywoodlife.com			
www.wetpaint.com			

Tale Sources w. Low Page Rank May Provide Valuable Info

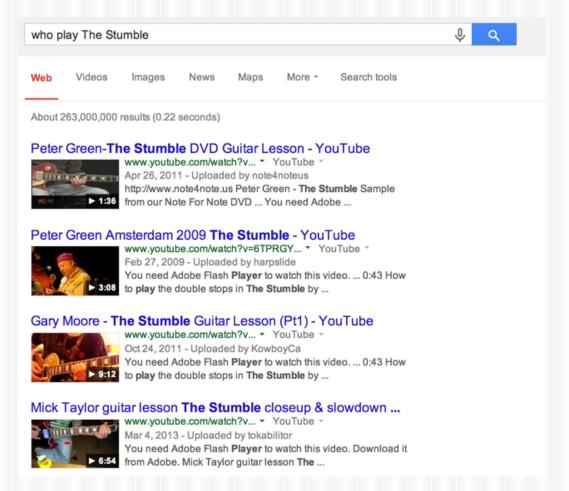


Tale Sources w. Low Page Rank May Provide Valuable Info



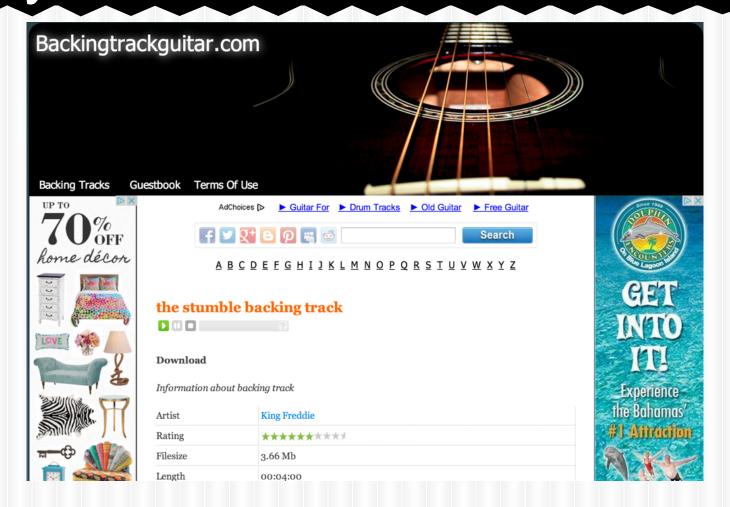
Good WebAnswer for an award-winning song

Tale Sources w. Low Page Rank May Provide Valuable Info



Missing WebAnswer for a not-so-popular song

Tale Sources w. Low Page Rank May Provide Valuable Info



Very precise info on guitar players but low Page Rank

Application I. A New Angle to **Evaluate Web Source Quality**





Q



Application I. A New Angle to Evaluate Web Source Quality





(USA), commonly referred to as the United States (US), America or simply the States, is a federal republic^{[10][11]} consisting of 50 states and a federal district. The 48 contiguous states and the federal district of Washington, D.C., are in central North America between Canada and Mexico. The state of Alaska is the northwestern part of North America and the state of Hawaii is an archipelago in the mid-Pacific. The country also has five populated and nine unpopulated territories in the Pacific and the Caribbean, At 3.79 million square miles (9.83 million km²) in total and with around 317



How to decide if a triple is indeed claimed by the source instead of an extraction error?

	0.8	
Triple 5	0.4	
Triple 6	0.8	
Triple 7	0.9	
Triple 8	1.0	
Triple 9	0.7	
Triple 10	0.2	
Accu	0.7	

Application I. A New Angle to **Evaluate Web Source Quality**

Create account & Log in

Q







(Obama, nationality, Bolivarianism) (many many such objects)

3 extractions (Pr_extCorr=0.01)

http://mathaba.net/news/?x=631316

http://www.laht.com/article.asp?ArticleId=329187&CategoryId=10717

http://www.iamericas.org/en/about-ioa/presidents-corner

Chávez Calls on Obama to Join Him in the Socialist Revolution

"Come on, Obama, align yourself with us on the way to socialism!" said the Venezuelan leader who this week also expropriated plants and lands of Venezuela's Polar, US firm Cargill, and Irish firm Smurfit. "Come on, it's the only way!"

By Jeremy Morgan Latin American Herald Tribune staff

CARACAS – President Hugo Chávez invited President Barack



Pr_tripleCorr=0

(Obama, nationality, Kenya)

- 2087 extractions:
 - Example of a correct extraction (Pr_extCorr=0.792):

http://beforeitsnews.com/obama-birthplace-controversy/2013/04/alabama-supreme-court-chief-justice-roy-moore-to-preside-over-obama-eligibility-case-2458624.html

2006: Obama In Kenya: I Am So Proud To Come Back Home - VIDEO HERE.

2007: Michelle Obama Declares Obama Is Kenyan And America Is Mean - VIDEO HERE.

2008: Michelle Obama Declares Barack Obama's Home Country Is Kenya - VIDEO HERE.

FLASHBACK: Obama Is The Original Birther! Obama In 1991 Stated In His Own Bio He Was Born In Kenya. DETAILS HERE.

Example of a wrong extraction (Pr extCorr=0.130):

http://www.monitor.co.ug/News/National/US+will+respect+winner+of+Kenya+election++Obama+says/-/688334/1685814/-/ksxagx/-/index.html

US will respect winner of Kenya election, Obama says

SHARE BOOKMARK □ PRINT RATING 常常常常常

Pr_tripleCorr=0 (not enough support)

(Obama, nationality, USA)

- 2481 extractions:
 - Example of a correct extraction (Pr_extCorr=0.999):

http://www.dogonews.com/2009/10/9/a-nobel-prize-for-our-awesome-president

Example of a wrong extraction (Pr_extCorr=0.261):

http://blogs.telegraph.co.uk/news/timstanley/100169248/barack-obamas-life-story-contains-myth-not-truth-says-biographer-so-why-did-the-media-report-it-as-truth/

Tim Stanley

Dr Tim Stanley is a historian of the United States. His new book about Hollywood politics is out in May. His personal website is www.timothystanley.co.uk and you can follow him on Twitter @timothy_stanley.

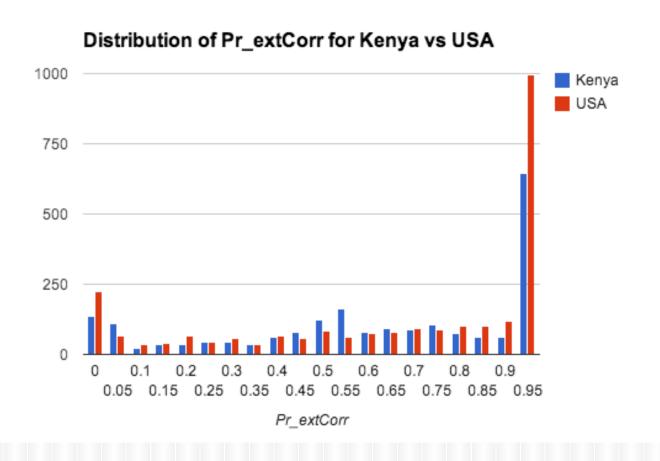




Barack Obama's life story contains 'myth, not truth', says biographer. So why did the media report it as truth?

Pr_tripleCorr=1 (Higher support)

Distribution of providers for Kenya and USA



Many gossip Web sites DO provide quite a lot of wrong factual information

Domain	#Triples	Sonya Score
www.eonline.com	12,871	0.363
perezhilton.com	46,912	0.427
radaronline.com	3,530	0.489
www.zimbio.com	2,464,452	0.530
mediatakeout.com	131	0.531
gawker.com	6,055	0.567
www.popsugar.com	1,805	0.576
www.people.com	16,886	0.585
www.tmz.com	8,149	0.621
www.fishwrapper.com	14	0.622
celebrity.yahoo.com	11,187	0.677
wonderwall.msn.com	2,524	0.684
hollywoodlife.com	4,536	0.689
www.wetpaint.com	19,284	0.730

Example for (URL, Predicate)

URL: https://ibirthdayworld.blogspot.com/2010/03/celebrity-birthdays-on-march-22.html

Predicate: date_of_birth

- #Facts = 42; Trustworthiness = 0.95



Source	Pred	Num_of_Triples	Accuracy
http://ibirthdayworld.blogspot.com/2010/03/celebrity-birthdays-on-march-22.html	/people/person/date_of_birth	42	0.953

Celebrity Birthdays On March 22



<-Marcel Marceau

Below are 124 famous people born on March 22.

Browse Gift Ideas - Browse Ecards

The names in brackets below are duplicate entries.

Aaron North was born on March 22, 1979. American guitarist.

Amy Stud was born on March 22, 1986. English singer-songwriter and musician.

Andreas Johnson was born on March 22, 1970. Swedish pop and rock singer-songwriter and musician.

Andrew Lloyd Webber was born on March 22, 1948. British composer of musicals.

Angelo Badalamenti was born on March 22, 1937. American composer.

Anja Kling was born on March 22, 1970. German actress.

Annabelle Apsion was born on March 22, 1963. English actress.

one Hyde was born on March 22, 1638. Wife of James II of England.

Anthony van Dyck was born on March 1599. Flemish Baroque artist.

Armin Hary was born on March 22, 1937. German athlete.

Avraham Fried was born on March 22, 1959. American singer-songwriter and musical entertainer.

Example for (URL, Predicate)

URL: https://ibirthdayworld.blogspot.com/2010/03/celebrity-birthdays-on-march-22.html

Predicate: date_of_birth

- #Facts = 42; Trustworthiness = 0.95

- Mistake: Anne Hyde (the URL says: 3/22/1638; Wiki/KG says:

3/12/1637)

Anne Hyde

Anne Hyde was Duchess of York and Albany as the first wife of James, Duke of York, later King James II and VII. Originally Anglican, her father was a lawyer. Wikipedia

Born: March 12, 1637, Windsor, United Kingdom

Died: March 31, 1671, London, United Kingdom **Spouse:** James II of England (m. 1660–1671)

Children: Anne, Queen of Great Britain, Mary II of England, More

Parents: Edward Hyde, 1st Earl of Clarendon, Frances Hyde,

Countess of Clarendon

Siblings: Laurence Hyde, 1st Earl of Rochester, Henry Hyde, 2nd Earl

of Clarendon



Application II. Provide An X-Ray for Extracted Data

- Goal: Help users analyze errors, changes and abnormalities in data
- Intuitions: cluster errors by features and return clusters with top error rates

Application II. Provide An X-Ray for Extracted Data

Cluster I.

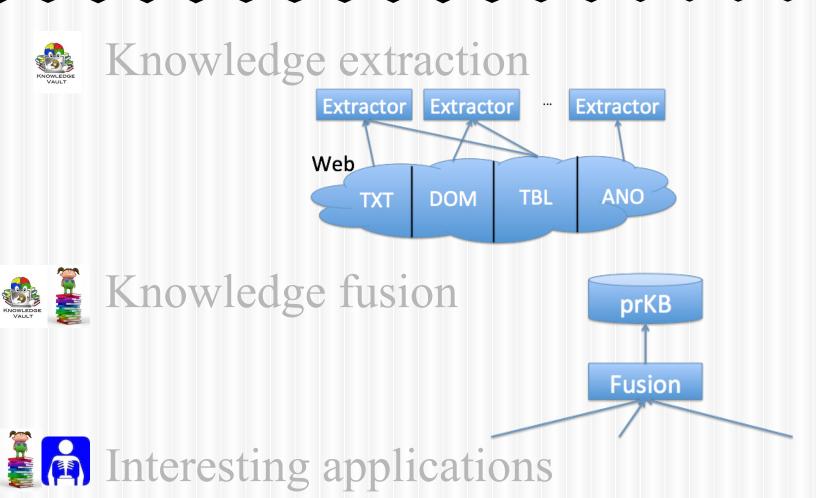
- Feature: (besoccor.com, date_of_birth, 1986_02_18)
- #Triples: 630; Errs: 100%
- Reason: default value

• Cluster 2.

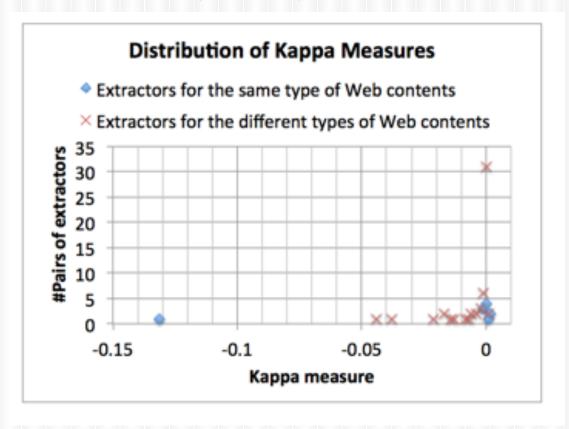
- Feature: (ExtractorX, pred: namesakes, obj:the county)
- #Triples: 4878; Errs: 99.8%
- E.g., [Salmon P. Chase, namesakes, The County]
- Contexts: The county was named for Salmon P. Chase, former senator and governor of Ohio
- Reason: Unresolved coreference

Outline

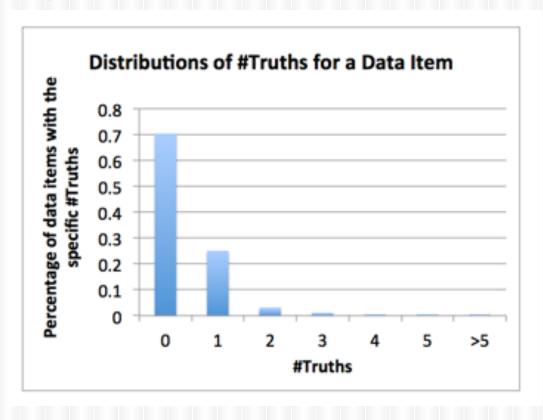
Future directions



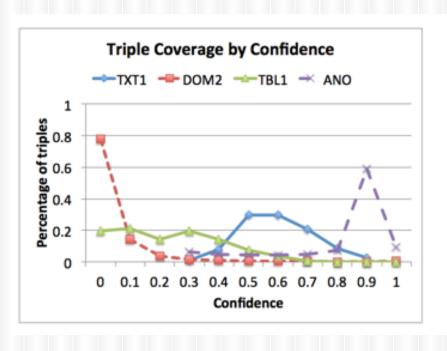
Assumption I. Independence between pairs of provenances (i.e., (URL, extractor))

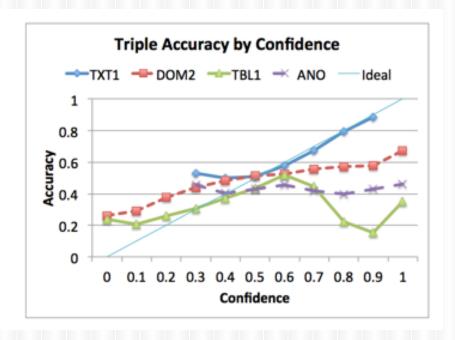


Assumption II. Single true object for each (sub, pred)



Assumption III. Extractions are deterministic





Assumption IV. Values (objects) are categorical

Assumption V. We have enough data to judge accuracy of each source Assumption VI. Local closed-world assumption in evaluation

Assumption VII. Global closed-world assumption--consider only existing entities and predicates in FB

WE NEED SOMETHING NEW!!!



TAKE AWAYS

A new area---Knowledge Fusion

 We can solve KF problem fairly well by adapting DF methods

Many interesting future directions for KF!

Many exciting applications for the prKB!!

Acknowledgement



Evgeniy Gabrilovich (Manager, need to say anything?)



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Kevin Murphy (Intelligent consultant)



Shaohua Sun (Critical representer to the outside world)



Wei Zhang (Fearless explorer of new ideas)

THANK YOU!

Questions?