## From Big Data to Big Knowledge

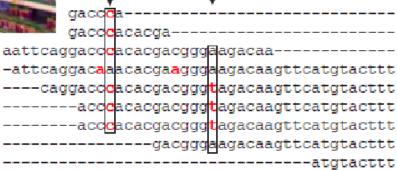
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Joint work with Luna Dong, Evgeniy Gabrilovich, Geremy Heitz, Wilko Horn, Panos Ipeirotis, Ni Lao, Wei-Lwun Lu, Thomas Strohmann, Shaohua Sun, Chun How Tan, Robert West, Wei Zhang, and others



## Big Data is everywhere







#### From Big Data to Big Knowledge

We are drowning in information and starving for knowledge. --- John Naisbitt.

- What does all this data "mean"?
- Words are ambiguous.
- e.g., "Taj Mahal"





We need to move from "strings" to "things".



## Google's Knowledge Graph

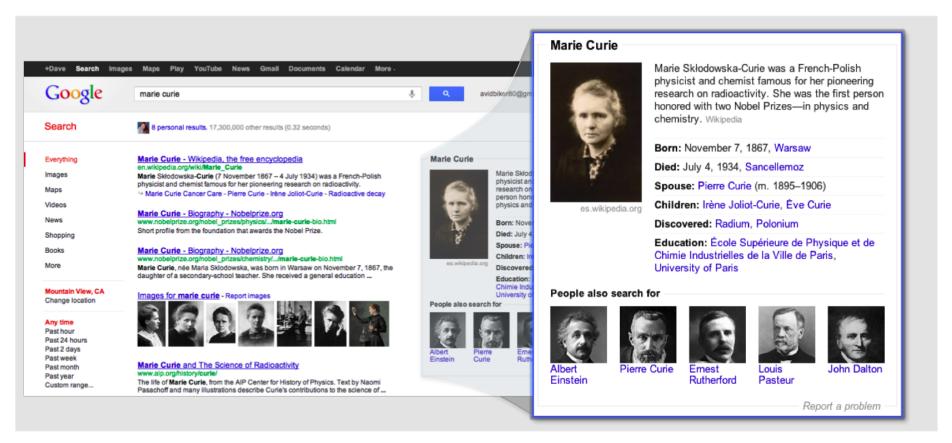


- 500M nodes (entities)
- 3.5B edges (facts)
- 1500 node types
- 35k edge types
- Extension of Freebase.com

Source: Brian Karlak, Google Faculty Summit, China, Dec 2012



#### **Knowledge Panels**





#### Freebase is created by merging many data sources

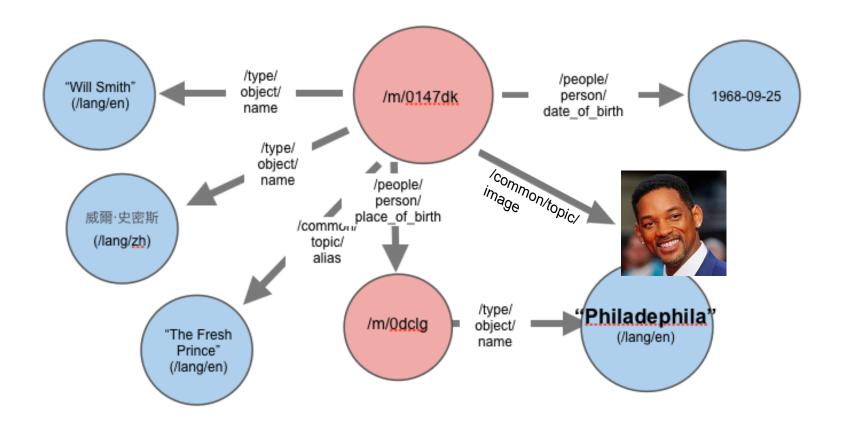


Massive entity linkage problem!

Source: John Giannandrea, CIKM 2011 industry talk



#### A fragment of Freebase (in RDF format)



Source: Brian Karlak, Google Faculty Summit, China, Dec 2012

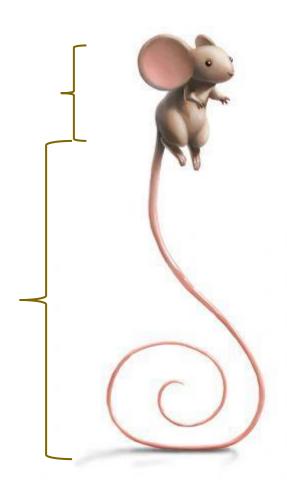
#### The long tail of knowledge



 Freebase is large, but still very incomplete:

Relation	% unknown	
	in Freebase	
Profession	68%	
Place of birth	71%	
Nationality	75%	
Education	91%	
Spouse	92%	
Parents	94%	

- We need automatic knowledge base construction methods
  - cf AKBC workshop at CIKM.



http://www.flickr.com/photos/sandreli/4691045841/

#### Outline



- From strings to things
- Reading the web
- Asking the web
- Asking people
- Open issues

#### Machine reading



- There are many academic groups (e.g., CMU, UW, MPI) that have developed methods to extract facts from large text corpora.
- At Google, we have developed a similar system, except it is 10x bigger.
- In addition, we use "prior knowledge" to help reduce the error rate.



Template matching methods

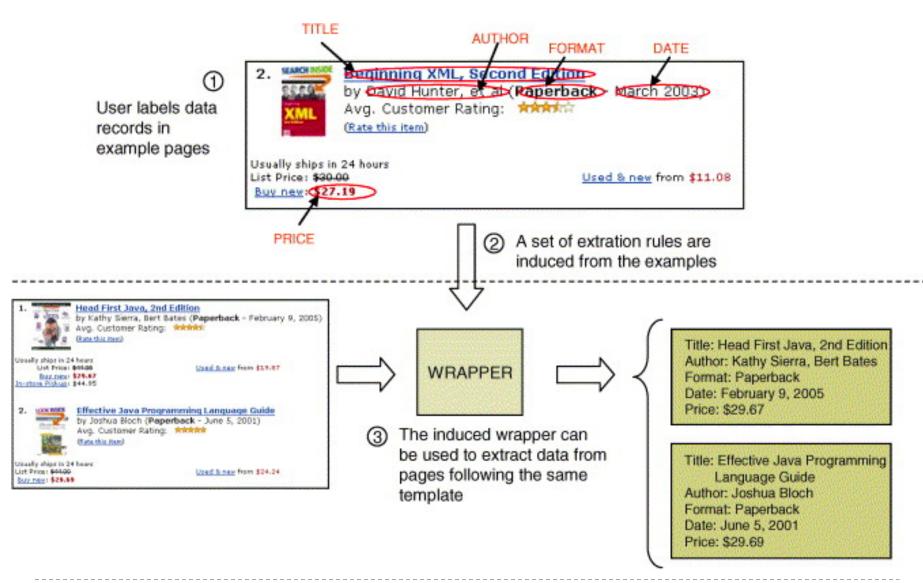


 Machine learning (binary classifiers trained on text / parse tree features)

$$\Pr(\text{predicate } p | \text{features } x) = \frac{1}{1+e^{-\theta^T f(x,p)}}$$

#### Wrapper induction





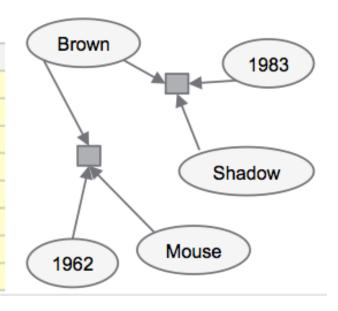
#### Fact extraction from tables



#### Caldecott Medal

From Wikipedia, the free encyclopedia

	<b>\$</b>	Illustrator \$	Title
	1999	Mary Azarian	Snowflake Bentley
	1954	Ludwig Bemelmans	Madeline's Rescue
	1983	Marcia Brown	Shadow
4	1962	Marcia Brown	Once a Mouse
4	1955	Marcia Brown	Cinderella, or the Little Glass Slipper
	1943	Virginia Lee Burton	The Little House
	1980	Barbara Cooney	Ox-Cart Man
	1959	Barbara Cooney	Chanticleer and the Fox



Need to create hidden column containing CVT or blank node, to represent the 3-tuple

#### Webmaster annotation



```
<section> Hello, my name is John Doe, I am a graduate research assistant at
the University of Dreams. My friends call me Johnny.
You can visit my homepage at <a href="http://www.JohnnyD.com">www.JohnnyD.com</a>.
I live at 1234 Peach Drive Warner Robins, Georgia./section>
```

```
<section itemscope itemtype="http://data-vocabulary.org/Person">
        Hello, my name is
        <span itemprop="name">John Doe</span>,
        I am a
        <span itemprop="title">graduate research assistant</span>
        at the
        <span itemprop="affiliation">University of Dreams</span>.
        My friends call me
        <span itemprop="nickname">Johnny</span>.
        You can visit my homepage at
        <a href="http://www.JohnnyD.com" itemprop="url">www.JohnnyD.com</a>.
        <section itemprop="address"</pre>
                 itemscope itemtype="http://data-vocabulary.org/Address">
                I live at
                <span itemprop="street-address">1234 Peach Drive</span>
                <span itemprop="locality">Warner Robins</span>
                <span itemprop="region">Georgia</span>.
        </section>
</section>
```

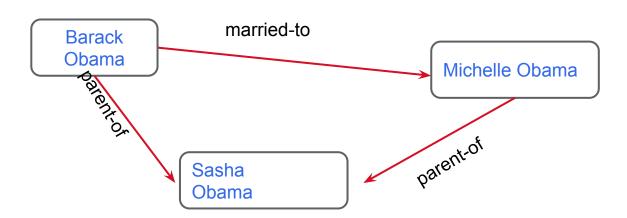
Example taken from <a href="http://en.wikipedia.org/wiki/Microdata">http://en.wikipedia.org/wiki/Microdata</a> (HTML)

## Predicting facts given prior knowledge



 Perform association rule mining\* on Freebase graph, to find noisy rules (features passed to a learned classifier).

$$\forall x, y. \exists z. \text{parent-of}(x, z) \land \text{parent-of}(y, z) \Rightarrow \text{married}(x, y)$$

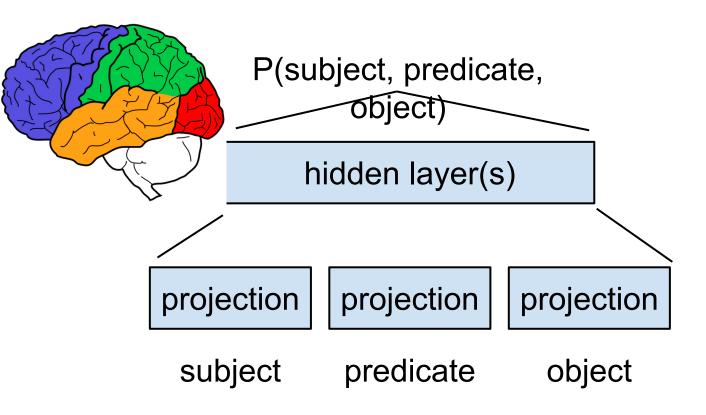


\* "Random Walk Inference and Learning in A Large Scale Knowledge Base", Ni Lao et al, 2011

#### A "neural" prior model



 Train a deep neural network\* to predict the probability of arbitrary facts, cf. tensor factorization.

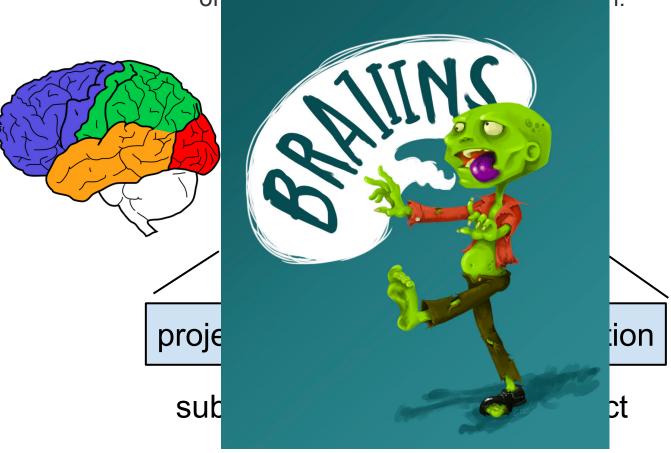


<sup>\*</sup> Similar to "Learning Structured Embeddings of Knowledge Bases", Bordes et al, 2011

## A "neural" prior model - Halloween version



Train a deep neural network\* to predict the probability of arbitrary factor of tancor factorization.

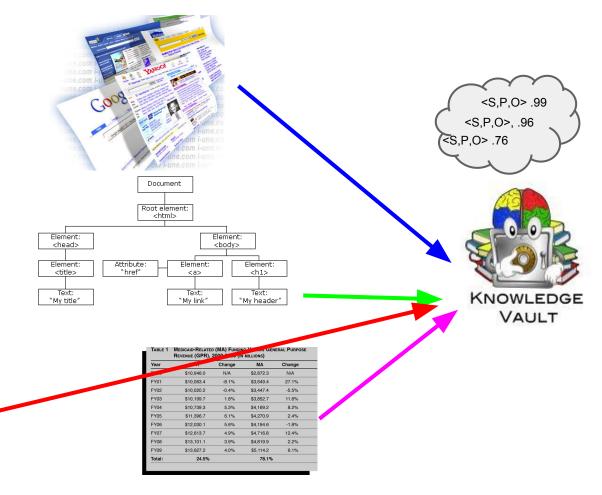




## Knowledge Vault\* fuses all these signals together

- Data from web
  - Unstructured text
  - Semi-structuredDOM trees
  - Structured WebTables
- "Prior" data from FB

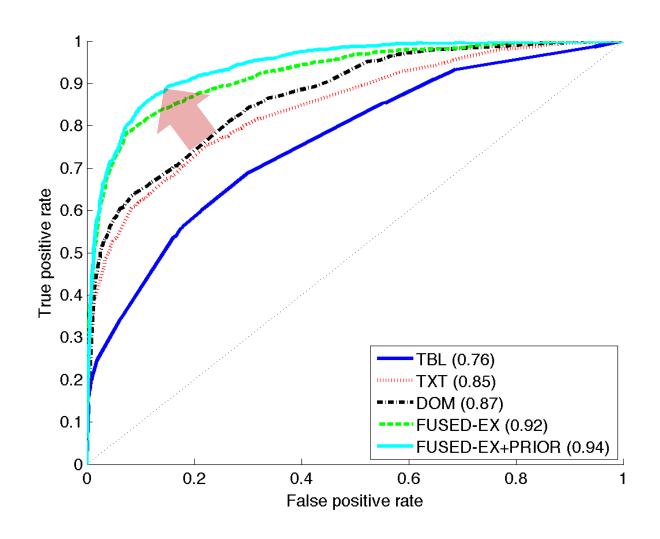




★ Details in a paper submitted to WWW'14 (Dong et al)

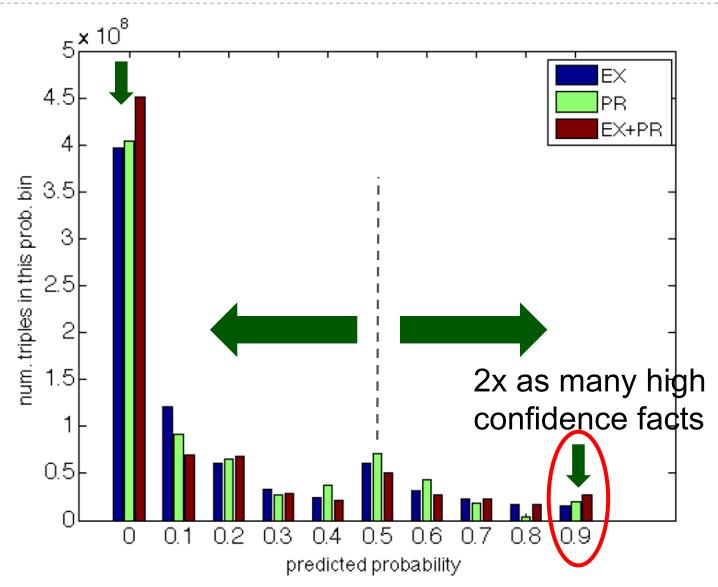
#### Benefits of information fusion





## Benefits of prior knowledge





#### Example: <Barry Richter, studied at, UW-Madison>

"In the fall of 1989, Richter accepted a scholarship to the University of Wisconsin, where he played for four years and earned numerous individual accolades ..."

"The Polar Caps' cause has been helped by the impact of knowledgeable coaches such as Andringa, Byce and former UW teammates Chris Tancill and Barry Richter."

→ Fused extraction confidence: 0.14

Prior knowledge:

- <Barry Richter, born in, Madison>
- <Barry Richter, lived in, Madison>
- → Final belief (fused with prior): 0.61

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## Knowledge based completion using Question Answering

- Even after large-scale machine reading of the web, many facts are still unknown.
- We can use web-based question-answering to perform targeted completion of missing attributes (pull vs push model).
- Main issue: what questions should we ask?

★ Details in a paper submitted to WWW'14 (West et al)

#### The importance of asking the right question



#### Who is the mother of Frank Zappa





#### The Mothers of Invention - Wikipedia, the free encyclopedia

en wikipedia.org/wiki/The Mothers\_of\_Invention \*

The **Mothers** of Invention were an American rock band from California that served as the backing musicians for **Frant Zappa**, a self-taught composer and ...

History - Personnel - Discography - References

#### Frank Zappa - Wikipedia, the free encyclopedia

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Jump to 1970: Rebirth of The **Mothers** and filmmaking - [edit]. **Frank Zappa** in Paris, early 1970s. Later in 1970, Zappa formed a new version of The ...

Discography - Moon Zappa - Diva Zappa - Gail Zappa



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Who is the mother of Frank Zappa Baltimore Maryland





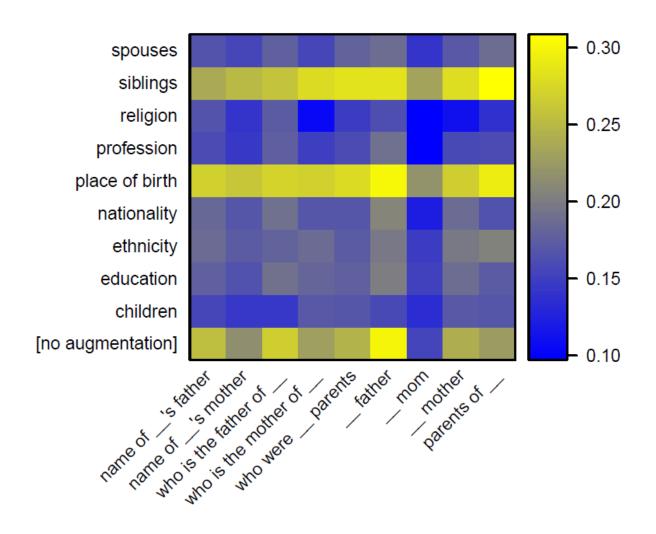
Frank Zappa - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Frank Zappa \*

Frank Vincent Zappa was born in Saltimore, Maryland, on December 21, 1940. His mother, Rose Marie (née Colimore), ) as of Italian and French ancestry; his ...

## Learning which questions to ask





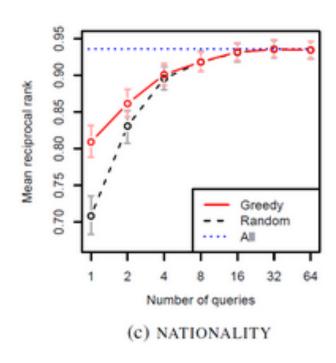
Color = mean reciprocal rank of true answer





## How many questions should we ask?

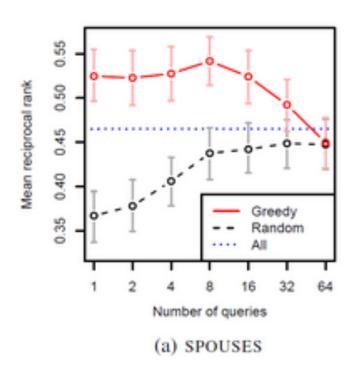




Performance increases, then plateaus

## Asking too many questions can hurt performance

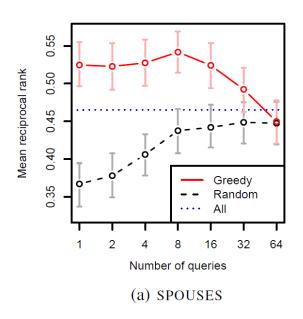


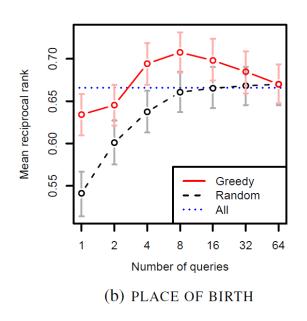


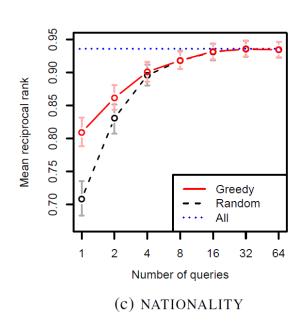
#### Performance gets worse

## Why does performance differ?





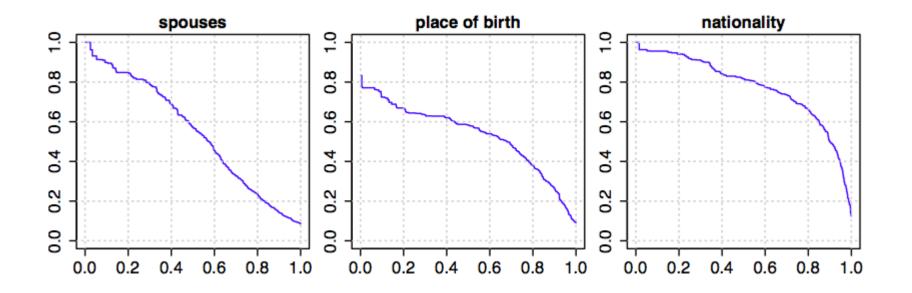




## Open class

Closed class





- About 25% of the high confidence facts were not discovered by the "read the web" approach.
- Accuracy is higher for closed-class predicates.

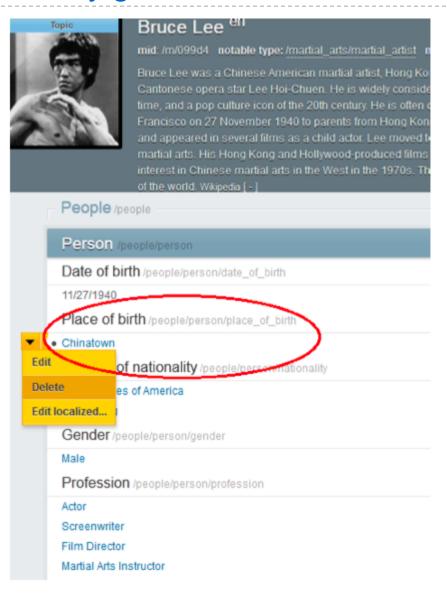
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#### Freebase is community generated/ edited





#### Knowledge panel feedback





georgia o'keeffe artist



## Georgia O'Keeffe

Artist

Georgia Totto O'Keeffe was an American artist. Born near Sun Prairie, Wisconsin, O'Keeffe first came to the attention of the New York art community in 1916. Wikipedia

Born: November 15, 1887, Sun Prairie, WI

Died: March 6, 1986, Santa Fe, NM

Nationality: American

Period: American modernism

Spouse: Alfred Stieglitz (m. 1924-1946)

Artwork: Cow's Skull: Red, White, and Blue, Sky Above

Clouds IV, More

#### Knowledge panel feedback



georgia o'keeffe artist



Click any fact to locate it on the web. Click **Wrong?** to report a problem. **You can also provide general feedback.** Cancel



#### Wrong?

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Wrong?

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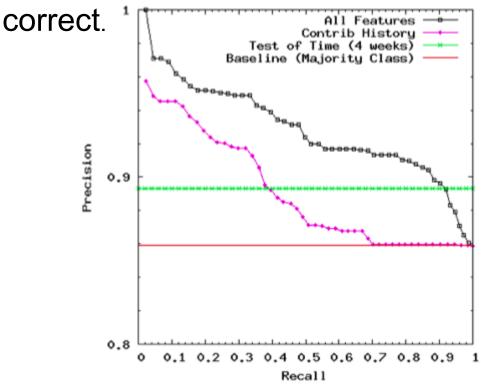
Wrong? Spouse: Alfred Stieglitz (m. 1924–1946)

Wrong? Artwork: Cow's Skull: Red, White, and Blue, Sky Above

Clouds IV, More



Use a binary classifier, trained on features derived from user contribution history, to predict the probability the contribution is



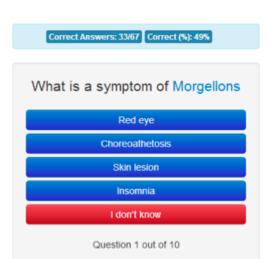
★ Details in a paper submitted to WSDM'14 (Tan et al)

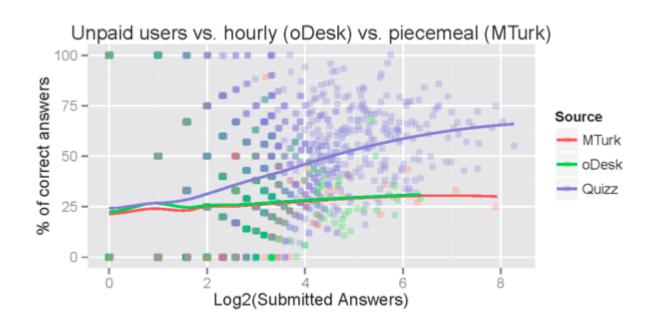


Place an ad asking users to take a quiz. Use ad optimization system to figure out which kinds of users to show the ad to.

#### Quiz on disease symptoms

Test how well you can recognize various disease symptoms www.quizz.us





★ Details in a paper submitted to WWW'14 (Ipeirotis and Gabrilovich)

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#### New entities

/m/?





"The Polar Caps' cause has been helped by the impact of knowledgeable coaches such as Andringa, Byce and former UW teammates Chris Tancill and Barry Richter."

/m/02ql38b

40M entities in Freebase, still missing many!

#### **New relations**



/people/person/education . /education/educational\_institute

In the fall of 1989, Richter accepted a scholarship to the University of Wisconsin, where he played for four years and earned numerous individual accolades ..."

/people/person/?

35k types of relations in Freebase, still missing many!

## Implicitly stated information



Joanne Schieble was just twenty-three and attending graduate school in Wisconsin when she learned she was pregnant. Her father didn't approve of her relationship with a Syrian-born graduate student, and social customs in the 1950s frowned on a woman having a child outside of marriage. To avoid the glare, Schieble moved to San Francisco and was taken in by a doctor who took care of unwed mothers and helped arrange adoptions. Originally, a lawyer and his wife agreed to adopt the new baby. But when the child was born on February 24, 1955, they changed their minds. Clara and Paul Jobs, a modest San Francisco couple with some high school education, had been waiting for a baby. When the call came in the middle of the night, they jumped at the chance to adopt the newborn, and they named him Steven Paul.

- <Joanne Schieble, /people/person/parents, Steve Jobs>
- <Steve Jobs, /people/person/date-of-birth, 2/24/55>

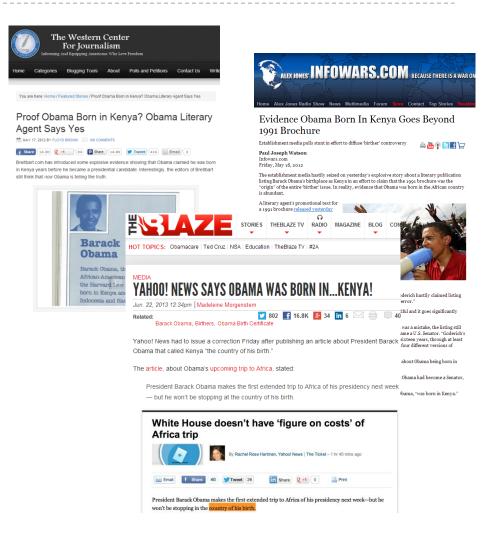
#### Assessing trustworthiness of sources







Create account & Log in



#### Fictional contexts



</en/abraham\_lincoln,</li>/people/person/profession,/en/vampire\_hunter> ?



#### Summary



KNOWLEDGE VAULT

- 1. Knowledge Vault is the largest repository of automatically extracted structured knowledge on the planet.
- 2. We can extract more information by asking the right questions from the web and/or people.
- 3. We are only extracting a small fraction of the facts on the web.