

Corroborating Information from Disagreeing Views

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February 4, 2010, *WSDM*

Motivating Example

What are the capital cities of European countries?

	France	Italy	Poland	Romania	Hungary
Alice	Paris	Rome	Warsaw	Bucharest	Budapest
Bob	?	Rome	Warsaw	Bucharest	Budapest
Charlie	Paris	Rome	Katowice	Bucharest	Budapest
David	Paris	Rome	Bratislava	Budapest	Sofia
Eve	Paris	Florence	Warsaw	Budapest	Sofia
Fred	Rome	?	?	Budapest	Sofia
George	Rome	?	?	?	Sofia

Voting

Information: redundancy

	France	Italy	Poland	Romania	Hungary
Alice	Paris	Rome	Warsaw	Bucharest	Budapest
Bob	?	Rome	Warsaw	Bucharest	Budapest
Charlie	Paris	Rome	Katowice	Bucharest	Budapest
David	Paris	Rome	Bratislava	Budapest	Sofia
Eve	Paris	Florence	Warsaw	Budapest	Sofia
Fred	Rome	?	?	Budapest	Sofia
George	Rome	?	?	?	Sofia
Frequence	P. 0.67 R. 0.33	R. 0.80 F. 0.20	W. 0.60 K. 0.20 B. 0.20	Buch. 0.50 Bud. 0.50	Bud. 0.43 S. 0.57

Evaluating Trustworthiness of Sources

Information: redundance, trustworthiness of sources (= average frequency of predicted correctness)

	France	Italy	Poland	Romania	Hungary	Trust
Alice	Paris	Rome	Warsaw	Bucharest	Budapest	0.60
Bob	?	Rome	Warsaw	Bucharest	Budapest	0.58
Charlie	Paris	Rome	Katowice	Bucharest	Budapest	0.52
David	Paris	Rome	Bratislava	Budapest	Sofia	0.55
Eve	Paris	Florence	Warsaw	Budapest	Sofia	0.51
Fred	Rome	?	?	Budapest	Sofia	0.47
George	Rome	?	?	?	Sofia	0.45
Frequency weighted by trust	P. 0.70 R. 0.30	R. 0.82 F. 0.18	W. 0.61 K. 0.19 B 0.20	Buch. 0.53 Bud. 0.47	Bud. 0.46 S. 0.54	

Iterative Fixpoint Computation

Information: redundance, trustworthiness of sources with iterative fixpoint computation

	France	Italy	Poland	Romania	Hungary	Trust
Alice	Paris	Rome	Warsaw	Bucharest	Budapest	0.65
Bob	?	Rome	Warsaw	Bucharest	Budapest	0.63
Charlie	Paris	Rome	Katowice	Bucharest	Budapest	0.57
David	Paris	Rome	Bratislava	Budapest	Sofia	0.54
Eve	Paris	Florence	Warsaw	Budapest	Sofia	0.49
Fred	Rome	?	?	Budapest	Sofia	0.39
George	Rome	?	?	?	Sofia	0.37
Frequence weighted by trust	P. 0.75 R. 0.25	R. 0.83 F. 0.17	W. 0.62 K. 0.20 B 0.19	Buch. 0.57 Bud. 0.43	Bud. 0.51 S. 0.49	

Context and problem

- **Context:**
 - Set of sources stating facts
 - (Possible) functional dependencies between facts
 - **Fully unsupervised setting:** we do not assume any information on truth values of facts or inherent trust in sources
- **Problem:** determine which facts are true and which facts are false
- **Real world applications:** query answering, source selection, data quality assessment on the web, making good use of the wisdom of crowds

Outline

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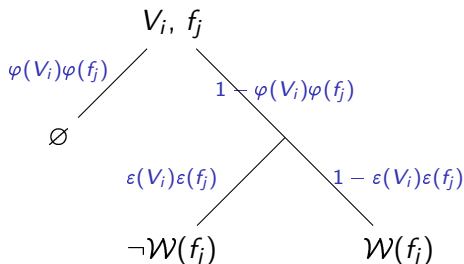
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General Model

- Set of facts $\mathcal{F} = \{f_1 \dots f_n\}$
 - Examples: “Paris is capital of France”, “Rome is capital of France”, “Rome is capital of Italy”
- Set of views (= sources) $\mathcal{V} = \{V_1 \dots V_m\}$, where a view is a partial mapping from \mathcal{F} to $\{T, F\}$
 - Example:
 - “Paris is capital of France” \wedge “Rome is capital of France”
- **Objective:** find the **most likely** real world \mathcal{W} given \mathcal{V} where the real world is a total mapping from \mathcal{F} to $\{T, F\}$
 - Example:
 - “Paris is capital of France” \wedge \neg “Rome is capital of France” \wedge “Rome is capital of Italy” \wedge ...

Generative Probabilistic Model



- $\varphi(V_i)\varphi(f_j)$: probability that V_i “forgets” f_j
- $\varepsilon(V_i)\varepsilon(f_j)$: probability that V_i “makes an error” on f_j
- Number of parameters: $n + 2(n + m)$
- Size of data: $\tilde{\varphi}nm$ with $\tilde{\varphi}$ the average forget rate

Obvious Approach

- **Method:** use this generative model to find the most likely parameters given the data
 - Inverse the generative model to compute the probability of a set of parameters given the data
 - Not practically applicable:
 - **Non-linearity** of the model and **boolean parameter** $\mathcal{W}(f_j)$
⇒ equations for inverting the generative model very complex
 - **Large number of parameters** (n and m can both be quite large)
⇒ Any exponential technique unpractical
- ⇒ Heuristic fix-point algorithms

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Baselines

Counting (does not look at negative statements, **popularity**)

$$\begin{cases} T & \text{if } \frac{|\{V_i : V_i(f_j) = T\}|}{\max_f |\{V_i : V_i(f) = T\}|} \geq \eta \\ F & \text{otherwise} \end{cases}$$

Voting (adapted only with negative statements)

$$\begin{cases} T & \text{if } \frac{|\{V_i : V_i(f_j) = T\}|}{|\{V_i : V_i(f_j) = T \vee V_i(f_j) = F\}|} \geq 0.5 \\ F & \text{otherwise} \end{cases}$$

TruthFinder [YHY07]: heuristic fix-point method from the literature

3-Estimates

- Iterative estimation of 3 kind of parameters:
 - truth value of facts
 - error rate or trustworthiness of sources
 - hardness of facts
- Tricky normalization to ensure stability

Functional dependencies

- So far, the models and algorithms are about positive and negative statements, without correlation between facts
- How to deal with functional dependencies (e.g., capital cities)?

pre-filtering: When a view states a value, all other values governed by this FD are considered **stated false**.
If I say that Paris is the capital of France, then I say that neither Rome nor Lyon nor ... is the capital of France.

post-filtering: Choose the **best answer** for a given FD.

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Datasets

- Synthetic dataset: large scale and highly customizable
- Real-world datasets:
 - General-knowledge quiz
 - Biology 6th-grade test
 - Search-engines results
 - Hubdub

Hubdub (2/2)

	Number of errors (no post-filtering)	Number of errors (with post-filtering)
Voting	278	292
Counting	340	327
TruthFinder	458	274
3-Estimates	272	270

General-Knowledge Quiz (1/2)

1. **Where is the city of Ushuaia located?**

- Don't know
- In Italy
- In Greece
- In Argentina
- In the Ivory Coast
- In Sweden
- In Malaysia

2. **What is the last word of all three parts of Dante's *Divine Comedy* (*Hell* — *Purgatory* — *Paradise*)?**

- Don't know
- "Stars" ("Stelle")
- "God" ("Dio")
- "Hope" ("Speranza")
- "Beatrice"

3. **Who discovered the planet Uranus?**

- Don't know
- Sir William Herschel (in 1781)
- Urbain Le Verrier (in 1846)
- Clyde Tombaugh (in 1930)
- Percival Lowell (in 1894)

<http://www.madore.org/~david/quizz/quizz1.html>

- 17 questions, 4 to 14 answers, 601 participants

General-Knowledge Quiz (2/2)

	Number of errors (no post-filtering)	Number of errors (with post-filtering)
Voting	11	6
Counting	12	6
TruthFinder	-	-
3-Estimates	9	0

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In brief

- We believe truth discovery is an important problem, we do not claim we have solved it completely
- Collection of **fix-point** methods (see paper), one of them (3-Estimates) performing remarkably and regularly well
- Cool real-world applications!

All code and datasets available from
<http://datacorrob.gforge.inria.fr/>

Thanks.


Webdam

Foundations of Web data management

Perspectives

- Exploiting **dependencies between sources** [DBES09]
- **Numerical values** ($1.77m$ and $1.78m$ cannot be seen as two completely contradictory statements for a height)
- No clear functional dependencies, but a **limited number of values** for a given object (e.g., phone numbers)
- **Pre-existing trust**, e.g., in a social network
- Clustering of facts, each source being trustworthy **for a given field**

References I

-  Xin Luna Dong, Laure Berti-Equille, and Divesh Srivastava. Integrating conflicting data: The role of source dependence. In *Proc. VLDB*, Lyon, France, August 2009.
-  Xiaoxin Yin, Jiawei Han, and Philip S. Yu. Truth discovery with multiple conflicting information providers on the Web. In *Proc. KDD*, San Jose, California, USA, August 2007.