

Truth Discovery – FTS talk

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- I'm Joe
- First year PhD
- Topic for now is *truth discovery*
- This talk is preparation for a seminar – feedback welcome
- Feel free to ask questions throughout

What is Truth Discovery?

- Lots of information is available today, from many different sources
 - The web
 - Social media platforms (Twitter, Facebook, ...)
 - Crowdsourcing systems
- People often *disagree* with what is true. Who should we trust in this case, and what should we believe?
- **Truth discovery:** find *true facts* and *trustworthy data sources* when faced with conflicting information.

- Background and context to the problem
- Existing work in this area
- My (preliminary) work:
 - How is it different?
 - What have I actually done?

- **Background and context to the problem**
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Setting the scene

- Information can be collected from *data sources*
 - Websites
 - Individuals
 - Crowdsourcing participants
 - Sensors
- A piece of information relates to an *object*
 - A real-world entity or question
 - E.g. How much does the UK send to the EU per week? What will the temperature be in Cardiff tomorrow?
- Different sources can provide different ‘facts’ for the same object
 - Conflicting statistics
 - Much conflict over ‘facts’ in politics
 - Low-quality sensors
- Can result from *poor or incomplete knowledge, or deliberate misinformation*

Example

- fullfact.org is an 'independent fact checking charity'

The image displays three fact-checking cards from fullfact.org, each with a different background color and layout. Each card includes a 'WHAT WAS CLAIMED' section, an 'OUR VERDICT' section, and a 'Read more...' button.

Card 1 (Teal background):
WHAT WAS CLAIMED: Repeated claim by Conservative Party Published widely across the media
“Labour’s spending plans would cost £1.2 trillion over five years.”
OUR VERDICT: Labour hasn’t published its 2019 manifesto yet, so nobody knows. This is largely based on Labour’s previous policy announcements. Many of the figures behind this estimate are uncertain or based on flawed assumptions.
Read more...

Card 2 (Pink background):
WHAT WAS CLAIMED: Claim by David Lommy MP Shared on Twitter
“Over 200,000 nurses have resigned since 2010.”
OUR VERDICT: Incorrect. This figure covers all nurses and health visitors who left the NHS in England, including those who retired or died in service. Overall the number of nurses working in the NHS is higher than in 2010.
Read more...

Card 3 (Dark Grey background):
WHAT WAS CLAIMED: Published by Liberal Democrats Leaflet sent to voters in London
“The Guardian said “Lib Dems winning and on the up after by-election victory” in August 2019.”
OUR VERDICT: This is misleading. The quote is an edited version of the Guardian headline, which made clear that the words originally came from Jo Swinson, not the Guardian itself.
Read more...

Figure 1: fullfact.org screenshot

Setting the scene (II)

- Even assuming fact-checkers are available, are they themselves to be trusted?
- Need automatic methods for finding true information
- Naive approach: take the information claimed by the most sources, i.e. perform a *vote*
- Will this work? Things to consider...
 - Large number of people today are claiming vaccines are harmful
 - A study investigated the spread of news on Twitter¹: “Falsehood diffused significantly farther, faster, deeper, and more broadly than the truth in all categories of information”
 - Some websites copy content from each other

¹Soroush Vosoughi, Deb Roy, and Sinan Aral. “The spread of true and false news online”. In: Science 359.6380 (2018)

Setting the scene (III)

- Trouble with voting is that all sources are *equally weighted*
- It would be better to use *trust* information
 - Trustworthy sources are given more weight
 - Won't get misled by an untrustworthy majority
- Central goals of truth discovery:
 - Identify *trustworthy sources* and *believable facts*, such that trustworthy sources claim believable facts and vice versa

What does it mean to be trustworthy?

- The notion of trust is extremely important in daily life
- Trust has been studied in the social sciences, but does not have an agreed upon formal definition
- Some authors distinguish between *trust*, *reputation* and *reliability*²
- Trust in daily life is often *personalised*
 - Trustworthiness is in the eye of the beholder

²Mohammad Momani and Subhash Challa. "Survey of trust models in different network domains". In: CoRR abs/1010.0168 (2010).

What does it mean to be trustworthy? (II)

- In contrast, truth discovery methods often seek a *global* notion of trustworthiness
- Different interpretations of trustworthiness exist in the truth discovery literature:
 - Probabilistic interpretation
 - Weights in optimisation-based methods
 - Heuristics
- Note: measures of trust are not comparable between algorithms

- Background and context to the problem
- **Existing work in this area**
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- Resolving conflicts in information is not new
 - *Data fusion* considers how to combine data from multiple sources, including conflict resolution
 - *Belief revision* considers how to update existing beliefs based on new (possibly conflicting) information
- Truth discovery is distinguished by its consideration of trustworthiness
- Many algorithms proposed in recent years
 - Mostly *unsupervised*: no ground truths for objects, and no known trustworthiness values
 - Mostly *iterative*: compute trust and belief scores iteratively until convergence

Example algorithm: *Sums*

- Perhaps one of the simplest algorithms is *Sums*
- Assigns each source s a sequence of *trust scores* $(T_n(s))_{n \in \mathbb{N}}$, and each fact f a sequence of *belief scores* $(T_n(f))_{n \in \mathbb{N}}$.
- Initially all scores are 0.5
- Update algorithm is as follows:
 - For each source s :
 - $T_{n+1}(s) \leftarrow \sum_{f \in \text{facts}(s)} T_n(f)$
 - For each fact f :
 - $T_{n+1}(f) \leftarrow \sum_{s \in \text{src}_n(f)} T_{n+1}(s)$
 - Divide each trust and belief score by the maximum
- Repeat until convergence

- Background and context to the problem
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Potential issues with existing work

- Lots of good algorithms, *but...*
- Many algorithms are somewhat opaque – difficult to see what the algorithm is actually *doing*
- Have to be evaluated empirically: run on a test dataset and compute accuracy
- This can make it difficult to compare algorithms:
 - Accuracy calculation depends on the dataset used
 - Algorithms may perform better or worse on different datasets
- Would be useful to have some theory behind truth discovery
 - Understand what they are doing by looking at theoretical properties
 - Make more principled comparisons
 - Deeper understanding of the problem (eventually...)

- Theoretical analysis has been done for related problems in *social choice*
- **Social choice:** aggregate the preferences of multiple agents in a 'fair' way to form a social preference
 - e.g. voting: how can votes be aggregated to choose the winner of an election?
 - In our case, aggregating claims from multiple sources
- The *axiomatic approach* is popular in social choice
 - Formulate *axioms* which describe intuitively desirable properties of voting rules
 - E.g. if everyone votes for the same person, they should be elected

- Can evaluate and compare algorithms by checking which axioms are satisfied
- Common goals are *impossibility results* and *characterisation results*
- E.g. voting has *Arrow's Impossibility Theorem*
 - Three seemingly good axioms cannot hold at the same time
 - Highlights fundamental problem with voting
- E.g. Altman and Tennenholtz³ characterised PageRank from Google
 - A ranking system coincides with PageRank iff it satisfies these axioms...
- Idea: can we give truth discovery an axiomatic treatment?

³Alon Altman and Moshe Tennenholtz. 2005. Ranking systems: the PageRank axioms.

- Background and context to the problem
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- **My (preliminary) work:**
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What have I been doing?

- Applying axiomatic approach to truth discovery
- Defined formal framework
- Formulated some axioms
- Had a look at some existing algorithms against my axioms

The framework: what is the input to the truth discovery?

- I consider a *very basic* form of truth discovery
- We have a finite set of sources \mathcal{S} , facts \mathcal{F} and objects \mathcal{O}
- Input to the problem (the dataset) is called a *truth discovery network*, and is defined as a graph

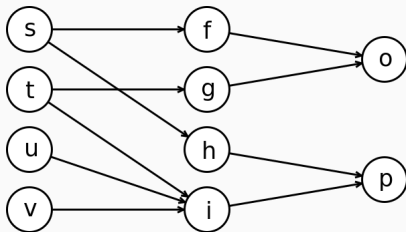


Figure 2: Example network

- Representing input as a graph is already common in the literature

Truth discovery network definition

From the paper...

Definition

A *truth discovery network* is a directed graph $N = (V, E)$ where $V = \mathcal{S} \cup \mathcal{F} \cup \mathcal{O}$, and $E \subseteq (\mathcal{S} \times \mathcal{F}) \cup (\mathcal{F} \times \mathcal{O})$ has the following properties:

1. For each $f \in \mathcal{F}$ there is a unique $o \in \mathcal{O}$ with $(f, o) \in E$, denoted $\text{obj}_N(f)$. That is, each fact is associated with exactly one object.
2. For $s \in \mathcal{S}$ and $o \in \mathcal{O}$, there is at most one directed path from s to o . That is, sources cannot claim multiple facts for a single object.
3. $(\mathcal{S} \times \mathcal{F}) \cap E$ is non-empty. That is, at least one claim is made.

We will say that s *claims* f when $(s, f) \in E$. Let \mathcal{N} denote the set of all TD networks.

The framework: what is the output?

- Most algorithms give output as numeric *trust scores* and *belief scores*
- Since scores are not comparable across algorithms, we are only concerned with the *ranking* that is induced by the scores
- Output should be therefore be a pair of rankings:
 - Source ranking tells us who is *more trustworthy*
 - Fact ranking tells us which fact is *more believable*
- In the previous example, *Sums* gives the rankings

$$s < u = v < t$$

$$f = h < g < i$$

- Algorithms are represented in the framework as functions, and are called *truth discovery operators*

From the paper...

Notation

For a set X , let $\mathcal{L}(X)$ denote the set of all total preorders on X , i.e. the set of transitive, reflexive and complete binary relations on X .

Definition

A *truth discovery operator* T is a mapping $T : \mathcal{N} \rightarrow \mathcal{L}(\mathcal{S}) \times \mathcal{L}(\mathcal{F})$. We shall write $T(N) = (\sqsubseteq_N^T, \preceq_N^T)$, i.e. \sqsubseteq_N^T is a total preorder on \mathcal{S} and \preceq_N^T is a total preorder on \mathcal{F} .

- The framework provides the definitions required to formally state axioms
- Most axioms adapted from social choice
- I will only mention the important ones...

- Recall that axioms are supposed to represent intuitive *desirable properties* of operators
- A key principle of truth discovery is that trustworthy sources make believable claims, and vice versa
- The trust and belief rankings need to *cohere* in this sense
- This idea is hard to pin down in general, but we can do so in specific cases...

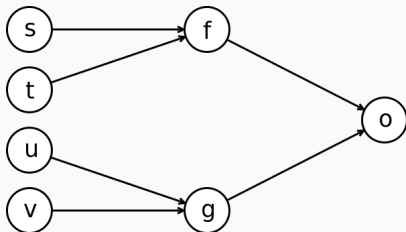


Figure 3: Coherence motivating example

- This idea comes from axiomatic analysis of ranking systems under the name *transitivity*⁴
- We consider this the most important axiom

⁴Alon Altman and Moshe Tennenholtz. 2008. Axiomatic Foundations for Ranking Systems

Definition

Let T be a TD operator, N be a TD network and $Y, Y' \subseteq \mathcal{F}$. We shall say Y is *less believable* than Y' with respect to N and T if there is a bijection $\phi : Y \rightarrow Y'$ such that $f \preceq_N^T \phi(f)$ for each $f \in Y$, and $\hat{f} \prec_N^T \phi(\hat{f})$ for some $\hat{f} \in Y$.

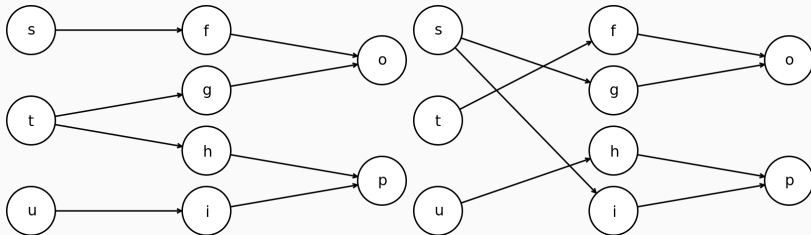
For $X, X' \subseteq \mathcal{S}$ we define X *less trustworthy* than X' with respect to N and T in a similar way.

Axiom (Coherence)

For any network N , $\text{facts}_N(s_1)$ *less believable* than $\text{facts}_N(s_2)$ implies $s_1 \sqsubset_N^T s_2$, and $\text{src}_N(f_1)$ *less trustworthy* than $\text{src}_N(f_2)$ implies $f_1 \prec_N^T f_2$.

Symmetry

- Rankings should depend on the *structure* of the network, not the *names* of sources and facts
- Consider swapping s with t and h with i :



- The structure is the same in each case, just different labels
- We should have $s \sqsubseteq_N t$ iff $t \sqsubseteq_{N'} s$ and $h \preceq_N i$ iff $i \preceq_{N'} h$
- Prevents operators being biased towards or against particular sources

Definition

Two TD networks N and N' are *equivalent* if there is a graph isomorphism π between them that preserves sources, facts and objects, i.e., $\pi(s) \in \mathcal{S}$, $\pi(f) \in \mathcal{F}$ and $\pi(o) \in \mathcal{O}$ for all $s \in \mathcal{S}$, $f \in \mathcal{F}$ and $o \in \mathcal{O}$. In such case we write $\pi(N)$ for N' .

Axiom (Symmetry)

Let N and $N' = \pi(N)$ be equivalent networks. Then for all $s_1, s_2 \in \mathcal{S}$, $f_1, f_2 \in \mathcal{F}$, we have $s_1 \sqsubseteq_N^T s_2$ iff $\pi(s_1) \sqsubseteq_{N'}^T \pi(s_2)$ and $f_1 \preceq_N^T f_2$ iff $\pi(f_1) \preceq_{N'}^T \pi(f_2)$.

Monotonicity

- We don't want *Voting*, but more support is better in some sense...
- If f is at least as believable as g and extra support for f comes in, f becomes *strictly* more believable

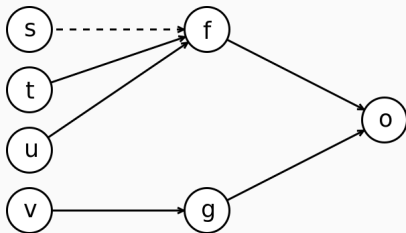


Figure 4: Monotonicity motivating example

Axiom (Monotonicity)

Suppose $N \in \mathcal{N}$, $s \in \mathcal{S}$, $f \in \mathcal{F} \setminus \text{facts}_N(s)$. Write E for the set of edges in N , and let N' be the network in which s claims f ; i.e. the network with edge set

$$E' = \{(s, f)\} \cup E \setminus \{(s, g) : g \neq f, \text{obj}_N(g) = \text{obj}_N(f)\}$$

Then for all $g \neq f$, $g \preceq_N^T f$ implies $g \prec_{N'}^T f$.

Independence

- Notion of *independence* is important: the ranking of a source/fact should only depend on the stuff that is relevant to it

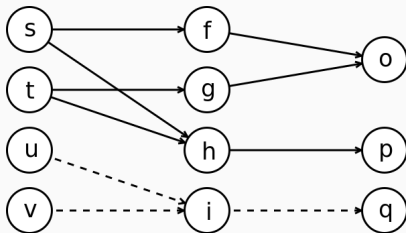
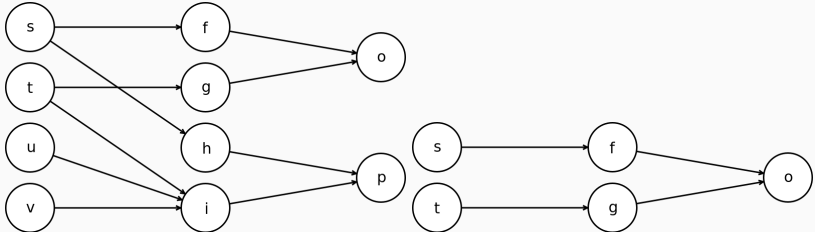


Figure 5: Independence motivating example

- e.g. are *u* and *v* relevant to *s*?

Per-object Independence (POI)

- First stab at independence, obtained by translating social choice (esp. voting) versions of independence
- If facts and sources for object o are the same in N and N' , the ranking of o 's facts is the same



Axiom

Let $o \in \mathcal{O}$ and write $\text{obj}_N^{-1}(o) \subseteq \mathcal{F}$ for the set of facts for o in a network N . Suppose N_1, N_2 are networks such that $F_o = \text{obj}_{N_1}^{-1}(o) = \text{obj}_{N_2}^{-1}(o)$ and $\text{src}_{N_1}(f) = \text{src}_{N_2}(f)$ for each $f \in F_o$. Then the restrictions of $\preceq_{N_1}^T$ and $\preceq_{N_2}^T$ to F_o are equal; that is, $f_1 \preceq_{N_1}^T f_2$ iff $f_1 \preceq_{N_2}^T f_2$ for all $f_1, f_2 \in F_o$.

Is POI a good idea?

- POI means we cannot use inter-object links
- With Symmetry and Monotonicity, this is very bad: it implies *Voting* behaviour within the facts for each object

Theorem

Let T be any operator satisfying Symmetry, Monotonicity and POI.

Then for any network N , object o and facts f, g for o , we have

$$f \preceq_N^T g \text{ iff } |\text{src}_N(f)| \leq |\text{src}_N(g)|$$

- Note: It is possible to strengthen POI – to what I call *Strong Independence* – to get *Voting*-like behaviour for *any* two facts: we have found an axiomatic characterisation of *Voting*

- Remember Coherence is our key axiom, which *Voting* fails
- Symmetry, Monotonicity and POI imply *Voting*-like behaviour
- Symmetry, Monotonicity, POI and Coherence? **No**

Theorem

There is no operator satisfying Coherence, Symmetry, Monotonicity and POI.

Impossibility (II)

- Counterexample is shown below

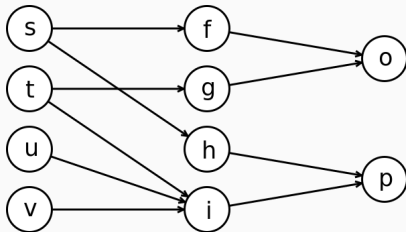


Figure 6: Counterexample used in the proof

- As far as we know, this is the first impossibility result for truth discovery

Final Independence axiom

- POI is *not* desirable since it rules out using indirect links
- Our final version of independence is very weak: two nodes are relevant to each other if there is a path between them, i.e. if they are in the same *connected component* of the graph

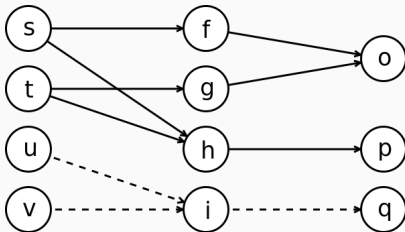


Figure 7: Independence example

Axiom (Independence)

For any TD networks N_1, N_2 with a common connected component G , the restrictions of $\sqsubseteq_{N_1}^T$ and $\sqsubseteq_{N_2}^T$ to $G \cap \mathcal{S}$ are equal, and the restrictions of $\preceq_{N_1}^T$ and $\preceq_{N_2}^T$ to $G \cap \mathcal{F}$ are equal; that is, $s_1 \sqsubseteq_{N_1}^T s_2$ iff $s_1 \sqsubseteq_{N_2}^T s_2$ and $f_1 \preceq_{N_1}^T f_2$ iff $f_1 \preceq_{N_2}^T f_2$ for $s_1, s_2 \in G \cap \mathcal{S}$ and $f_1, f_2 \in G \cap \mathcal{F}$.

Satisfaction of the axioms

- Those are the important axioms. Are they satisfied by actual truth discovery algorithms?

	Voting	SC-Voting	Sums	U-Sums
Coherence	X	X	✓	✓
Symmetry	✓	✓	✓	✓
Mon.	✓	✓	X	?
POI	✓	✓	X	X
Indep.	✓	X	X	✓

Table 1: Satisfaction of the axioms for the various operators

- Thanks for listening
- Questions?