A Content Management Perspective on Fact-Checking



Incha

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- Misinformation can stem from intentional deception effort, but also simply satire (Rubin et al. 2015) and "honest mistake"
- Fake news is as old as journalism, and hard to solve through political means



Motivation.

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Observation

We do not believe it is feasible to "automate" fact-checking or journalistic work; nor is it desirable, since journalistic content is authored by humans.Instead, we are interested in the role content management plays as a provider of models, algorithms and tools.

Outline

1 Computational Fact Checking

- 2 State ot the art
- **3** Perspectives



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Other definitions have been given (detail in the survey section.)



- For a claim to pass fact check, it must
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- Some claims may be crafted to mislead, i.e., look valid given some context or source that is either irrelevant or flawed.

The International Fact-Checking Network (IFCN), an organization sponsored by the Poynter Institute to "promote excellence in fact checking" 1 :

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- 5 A commitment to open and honest corrections.

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- Leads to the man-made part of the echo chamber.
- Automated recommendation system amplify the phenomenon, making it harder for users to break out of their filter bubbles.
- The actual impact of filter bubbles and echo chambers being studied[Garrett(2009), Garrett(2016)].

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- Timing matters : Emotionally engaging information, including rumors and misinformations, spread faster than corrections on social networks [Shin et al.(2017)Shin, Jian, Driscoll, and Bar].
 - If a verification may come too late, false information had time to "stick" in their audience's minds²
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The "arms race" between misinformation and fact checking is still on and probably will be so for a while.

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- NLP perspective, [Vlachos and Riedel(2014)] defines fact checking as the assignment of a truth value to a claim made in a particular context, on an ordinal scale, borrowed from Politifact's Truth-o-Meter.



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Definition (Fact checking [Babakar and Moy(2016)])

Fact checking is a four-stage process where media sources are

- (i) media sources are monitored,
- (ii) claims are spotted,
- (iii) claims are checked,

(iv) fact checking analysis results are created and published.

Overview

A very active research and development area today.

[Dagan et al.(2005)Dagan, Glickman, and Magnini, Álvaro Rodrigo	[Yu and Hatzivassiloglou(2003), Ennals et al.(2010)Ennals, Trushl

Legend Black boxes : tasks involved in fact checking Blue boxes : inputs and outputs

FIGURE – Fact checking tasks, ingredients, and relevant works : an overview.



End-to-end systems : DeFacto





End-to-end systems : DeFacto


End-to-end systems : ClaimBuster





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2016 Third Presidential Debate. Oct. 19, 2016, 9 p.m. EST

Chronological Order Order by Score

Least Check-worthy >=0.1>=0.2>=0.3>=0.--==0.5>=0.6>=0.7>=0.8>=0.9>=1.0 Most Check-worthy

is tougher. But they know what's going on. They know it better than anybody. They want strong borders. They feel we have to have strong borders. I was up in New Hampshire the other day. The biggest complaint they have -- it's with all of the problems going on in the world, many of the problems caused by Hillary Clinton and by Barack Obama All of the problems -- the single biggest problem is heroin that pours across our southern border. It's just pouring and destroying their youth. It's poisoning the blood of their youth and plenty of other people. We have to | Fackcheck this ders. We have to keep the drugs out of our country. We are -- right now, we're getting the drugs, they're getting the cash. We need strong borders. We need absolute -- we cannot give amnesty. Now, I want to build the wall. We need the wall Claim Matcher Claim Checker - Knowledge Bases Claim Checker - Search Engine Consulting the knowledge bases produced the We found the following claims which have been We found the following information after following results: professionally fact-checked. Check them out! processing some search engine results: Truth Rating Indeterminable Truth Rating True All of the problems -- the single biggest problem is heroin that pours across our southern border. It's just pouring and destroying their youth. What is all of the problems -- the single biggest "Heroin .. pours across our southern borders." problem? nilarity Rating 0.8320502943378437 Donald Trump URL SOURCE olitifact The single biggest problem in communication is the illusion that it has "I was up in New Hampshire the other day," Truth Rating True

The biggest complaint they have -- it's with all of the problems going on in the world, many of the problems caused by Hillary Clinton and by Barack Obama

GENIUS

JosueCaraballo

FILTER BY All Annotations

Is that true? What is the source?

report abuse





End-to-end systems : FullFact.org

Live Trends Tools BBC News BBC One BBC Two ITV Channel 4 PMOs Rec	
 ¹³³⁴³⁷ Economists have said GDP is rising. ¹³³⁴³⁷ Hospitals are struggling to provide essential gear, pi is the government gear, pi is the government cutting the number of beds in the National Health Service? 	Reference Reference <t< td=""></t<>
¹³³⁴³⁷ beds The beds ¹³³⁴³⁷ The average length of time for staying in the hospital has virtually Haft since the year 2000.	East daim matches • The government is cutting the number of hospital bads. This is right, the number of overnight bads in the English NHS has failen. Secont has iterating iterating interest than we sha the right daim to death of the interest of the interest we sha the right daim to death of the interest of the interest overnight data to death of the interest of the interes
 ¹⁵³⁴⁵⁷ We can look at the Labour Party 's record on this. ¹⁵³⁴⁵⁸ In the last six years of the last Labour government, 25,000 hospital beds were cut. 	East delim matches O the overcoge length of time for stoying in hospital has virtually holded since the year 2000. It host' quile howled but it has failen. The overcoge length of stoy overright was eight dops in 2000 and is about five dops now. Searchated (Intel) Jean Tha we that the opt cales to deat () (data)

End-to-end systems : FullFact.org



Read the factcheck

We've seen this claim 266 times since Jul 2016





End-to-end systems : CJ WorkBench

Workbench San Francisco contstruction pipeline Units built 〒 FII N/A total_unit 👃 Google Drive 😏 Twitter Data > Analyse - Filter by condition Mission Haves South SF 년 Column chart Q1 Q3 Q4 Neighborhoods 1014 Gilman Av **BP APPROVED** BVHP Area A.B 101 Donahue St 100 ALPINE TERRACE Castro/Upper Market

Data source

Reference and search.

Fixed reference database ClaimBus-

ter [Hassan et al.(2017)Hassan, Zhang, Arslan, Caraballo, Jimenez, Gawsane, Hasan, J DisputeFinder [Ennals et al.(2010)Ennals, Trushkowsky, and Agosta] (PolitiFact API); FullFact [Babakar and Moy(2016)] (internal manually checked claim DB); TruthTeller³ used claims manually checked by Factcheck.org; The Décodex plug-in developed by Le Monde also leverages their past fact checking analyses

Open database referencing FactMin-

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Search engines

DeFacto [Lehmann et al.(2012)Lehmann, Gerber, Morsey, and Ngomo, Gerber et al.(2015)Gerber, Esteves, Lehmann, Bühmann, Usbeck, Ngomo, and Speck] (search engines);

 Reference corpus Stance detection aims to determine whether a text in favor of a given target, against it, neutral or unrelated.

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Claim extraction

Definition (Claim detection [Levy et al.(2014)Levy, Bilu, Hershcovich, Aharoni, and

Given a topic (short sentence) and a set of relevant articles, their goal is to automatically extract concise statements from the articles that directly support or contests the given topic.

Claim extraction, w.r.t. to reference base.

- In [Levy et al.(2014)Levy, Bilu, Hershcovich, Aharoni, and Slonim], the authors provide manually labeled benchmark dataset and a classification approach based on a cascade of classifiers.
- DisputeFinder [Ennals et al.(2010)Ennals, Trushkowsky, and Agosta] indentifies within within text, claims covered by a reference database containing <u>disputed</u> claims.

In the extraction phase, phrases are transformed in bags of words through stop word removal, stemming etc. and matched against the database.

• FullFact ongoing work [Babakar and Moy(2016)] is researching claim extraction from text feeds, but also, audio and video streams;

A follow-up on De-

Facto [Gerber et al.(2015)Gerber, Esteves, Lehmann, Bühmann, Usbeck, Ngomo, and

Claim extraction (continued)

Claim check-worthiness.

- Using neural networks [Park and Cardie(2014)]and [Guggilla et al.(2016)Guggilla, Miller, and Gurevych] classify sentences as verifiable and unverifiable
- ClaimBuster [Hassan et al.(2015)Hassan, Li, and Tremayne, Hassan et al.(2017)Hassan, Zhang, Arslan, Caraballo, Jimenez, Gawsane, Hasan, Josep uses SVMs to monitors data sources such as social media, TV programs and websites. Classes include :
 - non-factual (e.g., opinions or subjective content);
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Separating fact from opinions.

- [Yu and Hatzivassiloglou(2003)] separates claims between facts and opinions with an opinion question answering system perspective.
- [Walker et al.(2012)Walker, Tree, Anand, Abbott, and King] for research on deliberation and debate



Claim accuracy assessment

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- ClaimBuster [Hassan et al.(2015)Hassan, Li, and Tremayne, Hassan et al.(2017)Hassan, Zhang, Arslan, Caraballo, Jimenez, Gawsane, Hasan, Josep
 - 1) Matches claims against previously checked claims from trusted repositories.
 - If no similar claim is found, reverts to Web search engines and question answering systems such as Wolfram Alpha.
 - 3) Combines the responses into a visual interface.

In a separate study, the authors show a strong correlation between the output of their system, and that claims checked by professionals at CNN and PolitiFact.

■ The Fast and Furious FactCheck Challenge⁵ proposed to classify news articles (not claims) among the categories TRUE, FALSE, SOMEWHAT TRUE and SOMEWHAT FALSE using any combination of human and automated tools;

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Putting claims into perspective

Query perturbation approach [Wu et al.(2014)Wu, Agarwal, Li, Yang, and Yu].



FIGURE – Strength and sensibility for the Giuliani's claim : "Adoptions went up 65 to 70 percent when [he] was mayor [of New York City]."

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 In [Wu et al.(2017a)Wu, Agarwal, Li, Yang, and Yu, Wu et al.(2017b)Wu, Gao, Agarwal, and Yang] the work is extended to support notions of diversity and representativity.

Julien: [Insert image]

Putting claims into perspective (continued)

 Data mining techniques are used in [Belfodil et al.(2017)Belfodil, Cazalens, Lamarre, and Plantevit] to highlight claims about behaviors in voting or rating contexts. Identifying groups of individuals and situations where their agreement significantly differs from usual.



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- Systems such

as [Sato et al.(2015)Sato, Yanai, Miyoshi, Yanase, Iwayama, Sun, and Niwa] can be used provide a general and balanced picture of a complex issue (e.g. "casinos increase criminality").

Crawls from a textual corpus in search for positive and negative stances, and outputs a summary of the articles for each stances.

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- **Structured Journalism**⁷ encourages journalists to publish database items directly
- Google introduced ClaimReview⁸ in 2017. It uses articles annotated with this metadata to display links in search result

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Publishing ClaimReview using MicroFormat

```
<div itemscope="" itemtype="http://schema.org/ClaimReview">
 An example paragraph reviewing a claim expressed in another document.
 <d1>
   <dt>Date published:</dt>
   <dd itemprop="datePublished">2014-07-23</dd>
   <dt>Review url:</dt>
   <dd itemprop="url">http://www.politifact.com/texas/statements/2014/jul/23/rick-perr
   <dt>Review by:</dt>
   <dd>
    <span itemprop="author" itemscope="" itemtype="http://schema.org/Organization">
        <span itemprop="name"><a itemprop="url" href="http://www.politifact.com/">Polit
        <img itemprop="image" src="http://static.politifact.com/mediapage/jpgs/politifa
        k itemprop="sameAs" href="http://twitter.com/politifact"/>
    </span>
   </dd>
 </dl>
 <h3>Claim reviewed:</h3>
   <blockquote itemprop="claimReviewed">
   More than 3,000 homicides were committed by 'illegal aliens' over the past six year
   </blockguote>
   <span itemprop="reviewRating" itemscope="" itemtype="http://schema.org/Rating">
     Rating: <span itemprop="ratingValue">1</span>
     (best score: <span itemprop="bestRating">6</span>),
     "<span itemprop="alternateName">True</span>".
     <img itemprop="image" src="http://static.politifact.com.s3.amazonaws.com/rulings/</pre>
```

ClaimReview as used in Google News





Outline

1 Computational Fact Checking

2 State ot the art

3 Perspectives

- 29 / 33 -

 Develop pluridisciplinarity Social and cognitive sciences useful to help devise psychologically effective fact-checking tools



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- Standardize Beyond "ClaimReview" more standards are needed to cover additional elements such as protocols, tools and functions.



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- Timely, sharp and balanced results.
- Avoiding frontal attack on one's convictions and beliefs.
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For instance, a misleading statement about the criminal activity of refugees in the countries receiving them, participates to a larger discussion about immigration and the way different political parties argue it should be handled. Thus, effective fact checking should focus more broadly on issues, as noted also in??.



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Educate

data literacy, envisioned as a set of math and statistic skills, through dedicated education modules at all levels



Data storage

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- Compliance with the standards Should adopt all standards of the trade, be open in the export and sharing of its inputs and outputs.



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