The Promises and Pitfalls of Automatic Fact Checking

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ACL Workshop on NLP for Positive Impact
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*partial slide credit: Greta Warren

UNIVERSITY OF
COPENHAGEN

Fact checking – what is it?

Donald Trump

stated on February 18, 2025 in remarks to reporters at Mar-a-Lago:

Volodymyr Zelenskyy "started" the war in Ukraine with Russia.

FOREIGN POLICY | MILITARY | UKRAINE | RUSSIA | DONALD TRUMP

By Claire Cranford
February 19, 2025

By Louis Jacobson
February 19, 2025

Did Ukraine start its war with Russia, as President Donald Trump said? No, Russia invaded

IF YOUR TIME IS SHORT

- Media outlets worldwide covered Russia's February 2022 invasion of Ukraine and Russian President Vladimir Putin acknowledged it as a "special military operation," saying the offensive would "seek to demilitarize and denazify Ukraine."
- For years, Russia has sought to blame Ukrainian actions for its invasion.

See the sources for this fact-check

https://www.politifact.com/factchecks/2025/feb/19/donald-trump/did-ukraine-start-its-war-with-russia-as-president/#sources

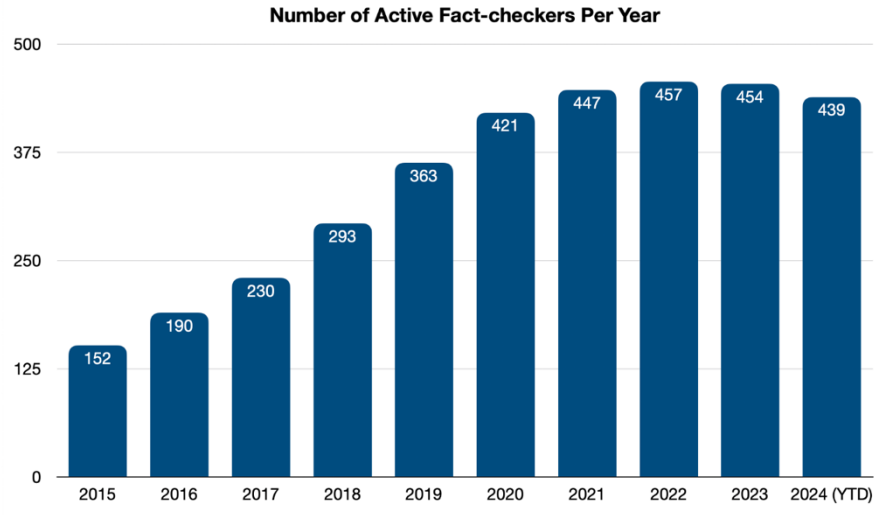
Fact checking – why is it important right now?

Global Risks Report 2025

WORLD ECONOMIC FORUM

Top 10 risks in the next 2 years

1st Misinformation and disinformation
2nd Extreme weather events
3rd State-based armed conflict
4th Societal polarization
5th Cyber espionage and warfare
6th Pollution
7th Inequality
8th Involuntary migration or displacement
9th Geoeconomic confrontation
10th Erosion of human rights and/or civic freedoms



The number of active fact-checkers per year, 2015 to 2024 (year-to-date). The Reporters' Lab continuously updates its counts based on the start and stop dates of the fact-checkers. That means our numbers are revised year-to-year. (Courtesy)

Fact checking – why is it important right now?

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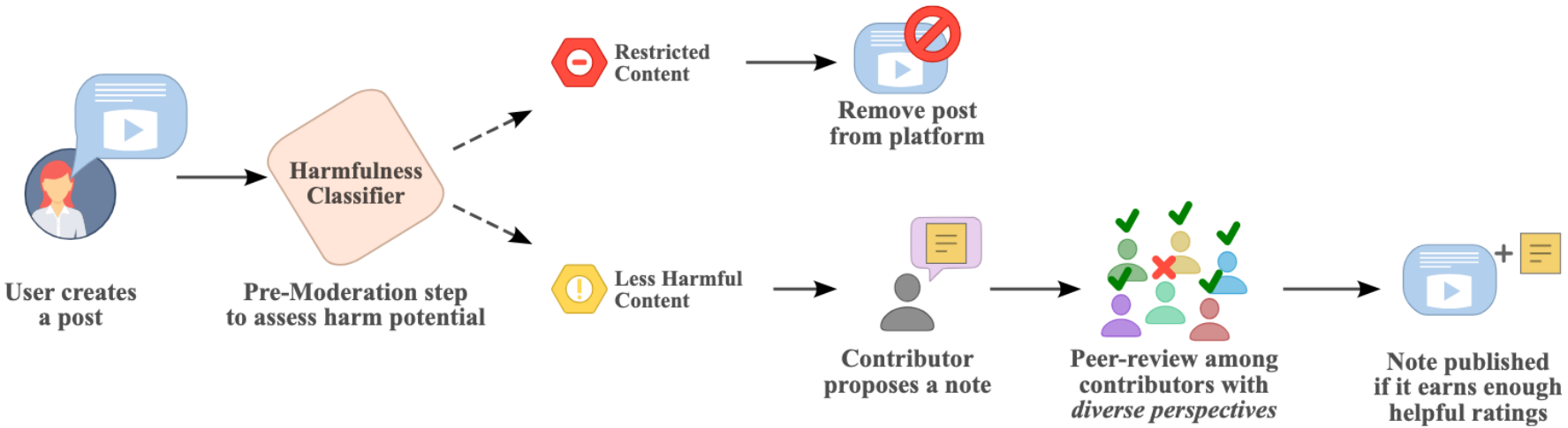
Meta's factchecking partners brace for layoffs

Meta has provided over $100m for certified organizations to conduct factchecks on its platforms since 2016

Ten factchecking outlets are listed by Meta as current partners in the US. Photograph: Jeff Chiu/AP

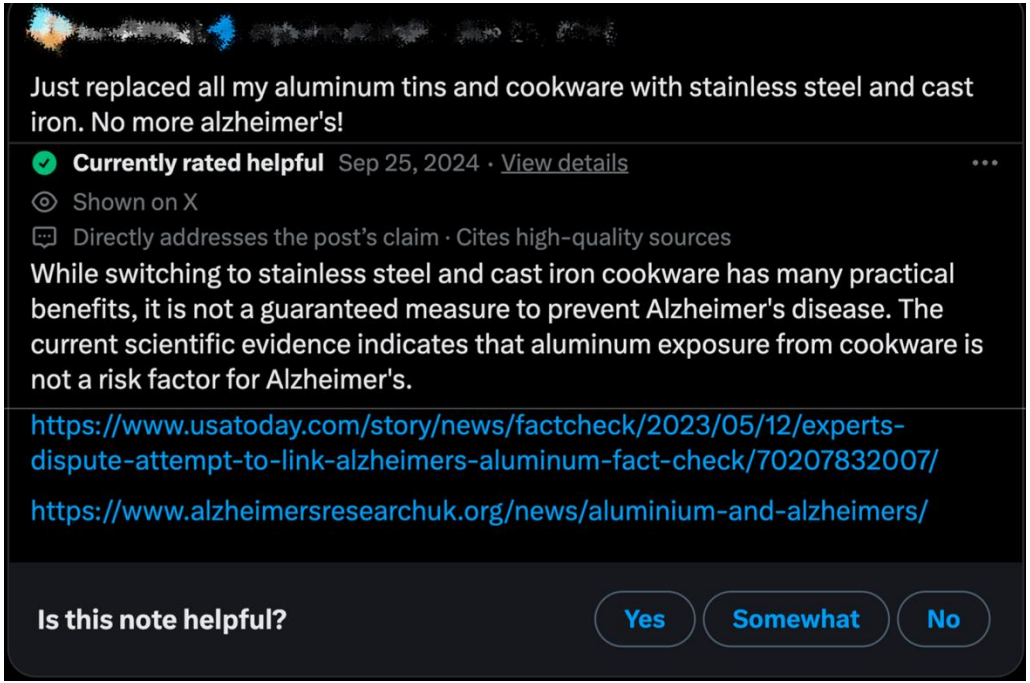
Community Notes (X / Twitter, Meta / Facebook, TikTok)



Moderation process:
(i) Pre-Moderation using AI classifiers: Restricted / blocked vs less harmful -> community moderation
(ii) Community Moderation: eligible volunteers propose additional context that undergoes peer review by other contributors with diverse perspectives before being published after a consensus is achieved

Augenstein et al. Community Moderation and the New Epistemology of Fact Checking on Social Media. CoRR, abs/2505.20067, May 2025.

Community Notes (X / Twitter, Meta / Facebook, TikTok)



Relation between fact checking and community notes

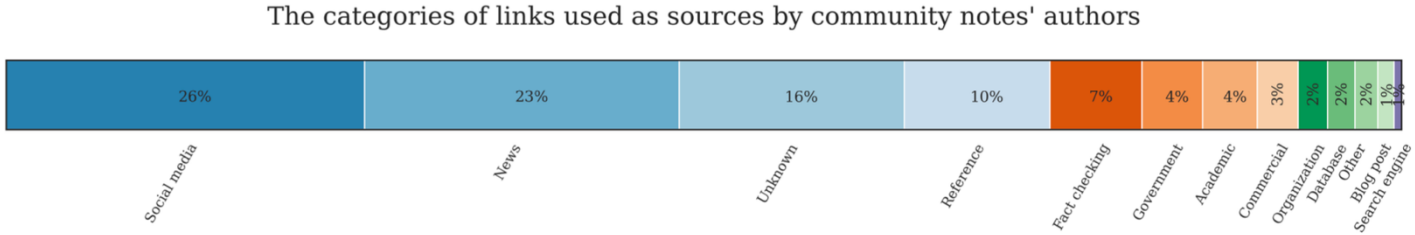


Figure 7: The categories of links used by Community notes' authors as a source, filtering for notes rated as "helpful".

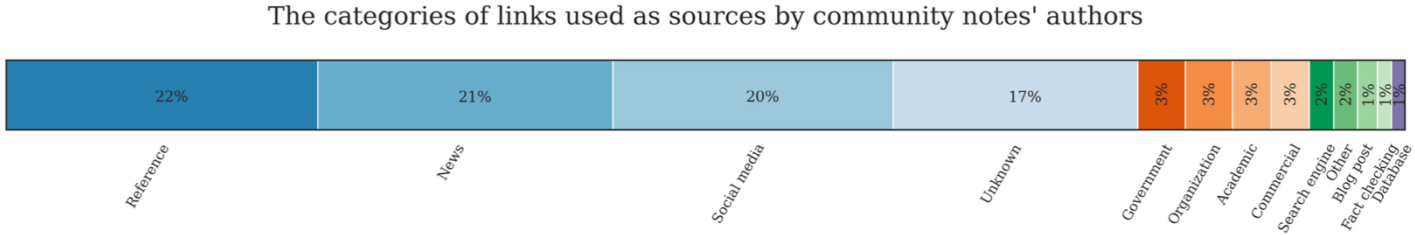


Figure 8: The categories of links used by Community notes' authors as a source, filtering for notes rated as "not helpful".

Nadav Borenstein*, Greta Warren*, Desmond Elliott, **Isabelle Augenstein**. Can Community Notes Replace Professional Fact-Checkers? In Proceedings of the 63rd Annual Meeting of the Association for Computational Linguistics (ACL 2025), July 2025.

Community Notes – does it work?

TECH · X

X's crowd-sourced 'Community Notes' fact checks fail to address flood of U.S. election misinformation, report says

BY BARBARA ORTUTAY AND THE ASSOCIATED PRESS
October 31, 2024 at 6:04 AM EDT

Workers install lighting on an "X" sign atop the company headquarters, formerly known as Twitter, in downtown San Francisco, July 28, 2023.
NOAH BERGER—AP

- *"Accurate notes correcting false and misleading claims about the U.S. elections were not displayed on **209 out of a sample of 283 posts deemed misleading** — or 74%"*
- *"Misleading posts that did not display Community Notes even when they were available included **false claims that the 2020 presidential election was stolen** and that **voting systems are unreliable**"*
- *"In the cases where Community Notes were displayed, the **original misleading posts received 13 times more views** than their accompanying notes"*

https://fortune.com/2024/10/31/x-community-notes-fact-checks-us-election-misinformation/

Community Notes – why does it not work?

- Only 11% of submitted notes reach ‘helpful’ status (i.e., shown to users) by achieving a cross-perspective
- Long time frame for notes to reach the algorithm’s required agreement level (15.5 hours on average)

➢ **False information has already spread**

- No expertise needed to become notes contributor
- Reliance on subjective helpfulness rather than objective facts
- Inadequate support and guardrails regarding explicit content

➢ **Key issues: speed, expertise, safety, adversarial attacks**

Nadav Borenstein*, Greta Warren*, Desmond Elliott, **Isabelle Augenstein**. Can Community Notes Replace Professional Fact-Checkers? In Proceedings of the 63rd Annual Meeting of the Association for Computational Linguistics (ACL 2025), July 2025.

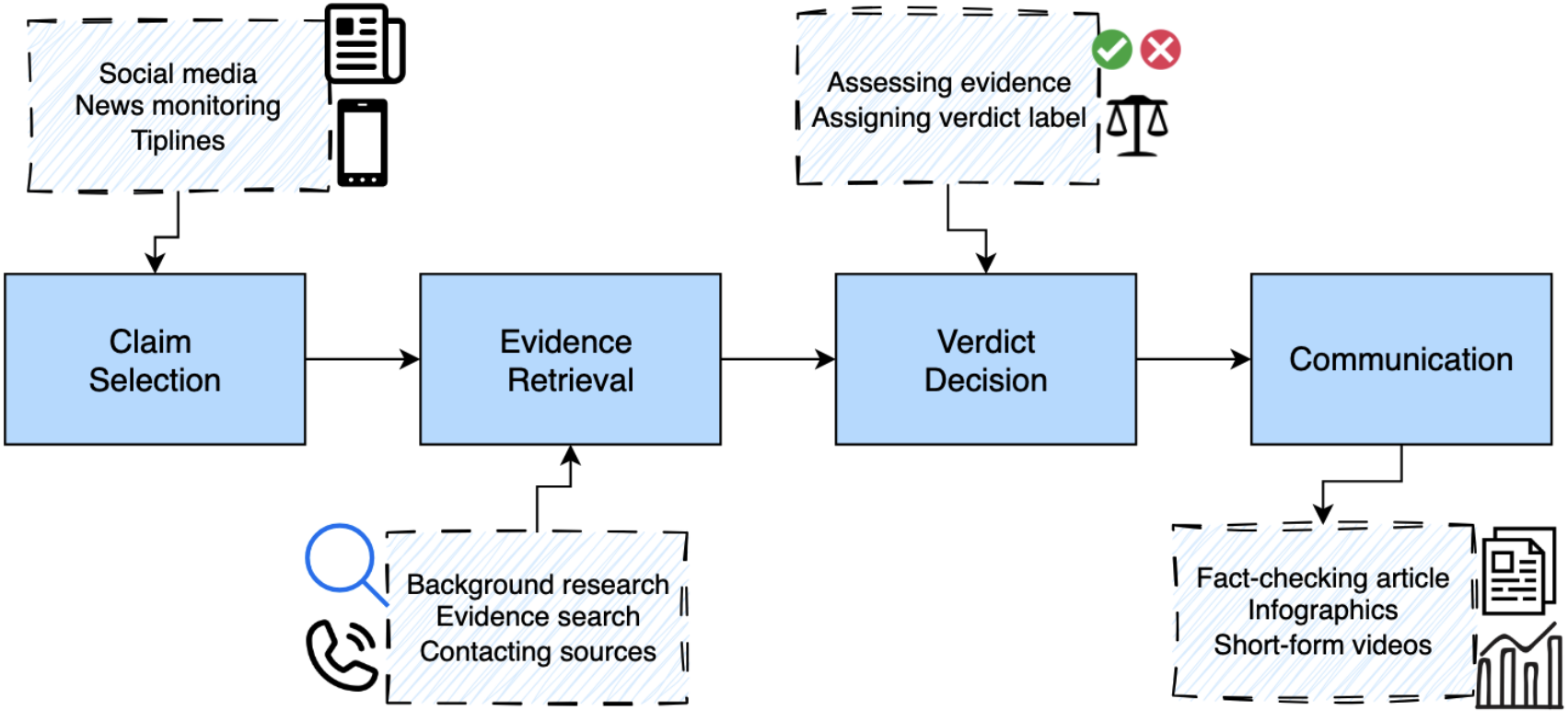
Community Notes: Recommendations

- Collaboration between **community and experts**
  - Workload distribution (repetitive claims vs high-risk claims)
  - Fact checkers as secondary reviewers of notes
  - Community flagging checkworthy claims

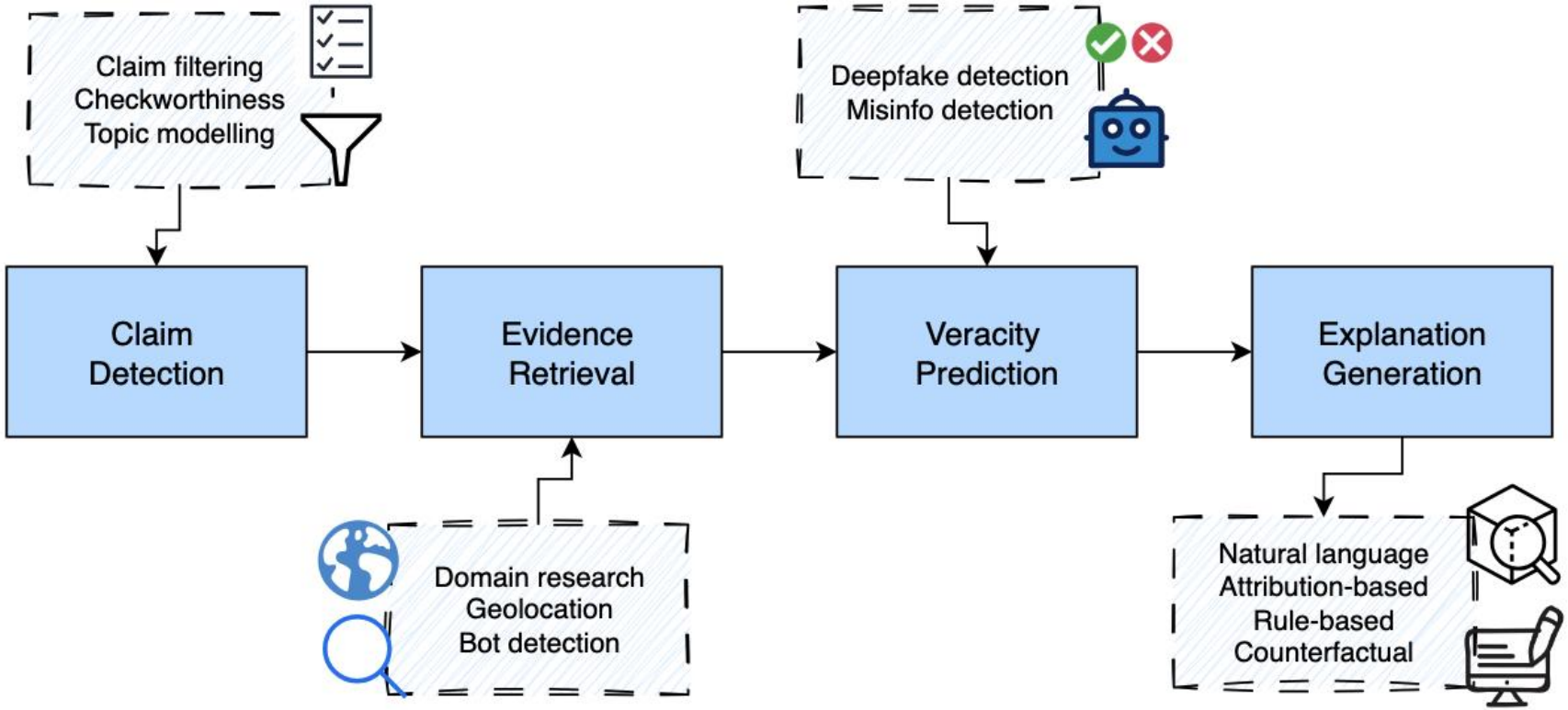
- Collaboration between **technology and the community**
  - Identify users likely to bring in diverse perspectives
  - Fusing community notes
  - Simulating crowd with AI agents (e.g. for sensitive content)
  - Handle previously checked notes with AI models

Augenstein et al. Community Moderation and the New Epistemology of Fact Checking on Social Media. CoRR, abs/2505.20067, May 2025.

Journalistic fact checking – how?

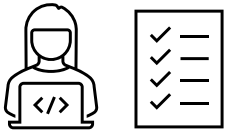


Explainable automatic fact checking – how?

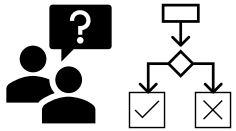


Explainable automatic fact checking

Methods disconnected
from fact-checking practice
(Schlichtkrull et al., 2023)



Desiderata shaped by
AI developers & researchers
(Das et al., 2023)



Ineffective for fact-checkers
& misleading for laypeople
(Schmitt et al., 2024; Lim et al., 2024)

Research Questions

How do fact-checkers explain their decisions and processes?

Where are explanations of automated fact-checking systems needed?

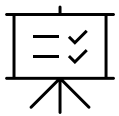
How can explanations of automated fact-checking systems address fact-checkers' explanation needs?

Greta Warren, Irina Shklovski, **Isabelle Augenstein**. Show Me the Work: Fact-Checkers' Requirements for Explainable Automated Fact-Checking. Conference on Human Factors in Computing Systems (CHI 2025), May 2025.

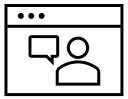
Method: Fact-checker interviews

10 interviews with fact-checkers in June & July 2024

5 women and 5 men from Europe, Africa, Asia, North America & South America



Pre-interview questionnaire



60-minute semi-structured interview

Bottom-up open coding → selective codes → Themes

Greta Warren, Irina Shklovski, **Isabelle Augenstein**. Show Me the Work: Fact-Checkers' Requirements for Explainable Automated Fact-Checking. Conference on Human Factors in Computing Systems (CHI 2025), May 2025.

Design implications: Source quality

Primary sources are gold-standard

Account for biases & positionality of secondary sources

Evidence quality, relevance and reliability must be assessed and **explained alongside the verdict**

Design implications: Nuanced verdicts

Pervasive misinformation often has a grain of truth

Detailed verdicts may be more effective & less polarising

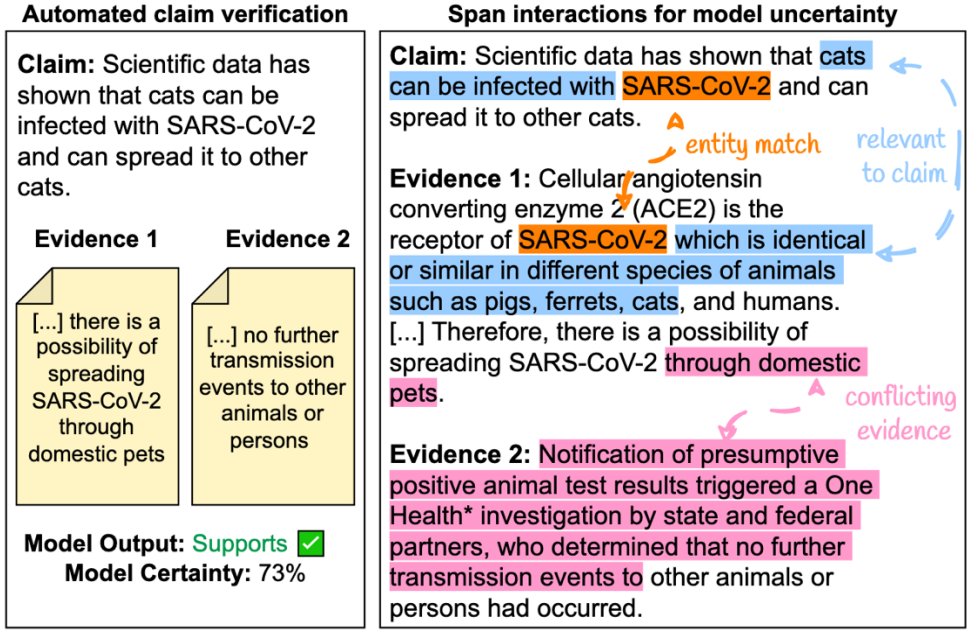
Explaining complex claims requires nuance **beyond binary true or false verdicts**

Design implications: Show the work

Explaining the **pathway to the verdict** is as important as the verdict itself

Greta Warren, Irina Shklovski, **Isabelle Augenstein**. Show Me the Work: Fact-Checkers' Requirements for Explainable Automated Fact-Checking. Conference on Human Factors in Computing Systems (CHI 2025), May 2025.

Show the work -> explaining sources of uncertainty



Natural language explanations of model uncertainty

The evidence in **Evidence 1** that "SARS-CoV-2" aligns with the statement in the **Claim** "SARS-CoV-2", confirming the virus's identity. This agreement slightly **reduces uncertainty** due to the **exact match**.

The evidence in **Evidence 2**, "which is identical or similar in different species of animals such as pigs, ferrets, cats," aligns with the **claim** "cats can be infected with". This agreement strengthens the claim by indicating that **cats are among the susceptible species**, **reducing uncertainty**.

However, the statement in **Evidence 1**, "through domestic pets." conflicts with the statement in **Evidence 2**, "Notification of presumptive positive animal test results triggered a One Health* investigation by state and federal partners, who determined that no further transmission events to other animals or persons had occurred." This disagreement **introduces significant uncertainty**, as it suggests that while cats can be infected, there is no evidence of them spreading the virus further, **contrary to the claim**.

Jingyi Sun, Greta Warren, Irina Shklovski, **Isabelle Augenstein**. Explaining Sources of Uncertainty in Automated Fact-Checking. CoRR, abs/2505.17855, May 2025.

Wrap-Up: Fact checkers needs vs. AI methods' limitations

Verifiable explanations
Issues with faithfulness and stability of feature attributions

Explaining uncertainty
Numerical percentages disconnected from human notions of uncertainty

Replicable explanations
Requires end-to-end fact-checking systems & alignment with fact-checker processes

Way forward: human-centered explainable fact checking

Aligning AI methods with fact-checker reasoning processes

Providing human-centred, useful explanations tailored to context and expertise

HCI & AI research is needed to integrate automated fact-checking into fact-checkers processes & ensure fact-checkers remain central

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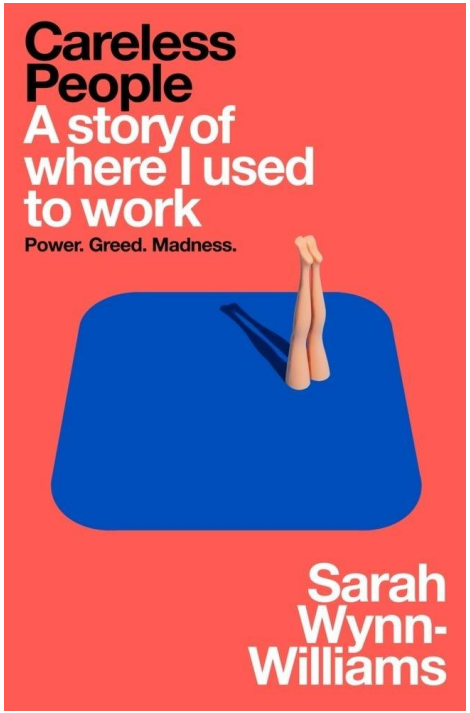
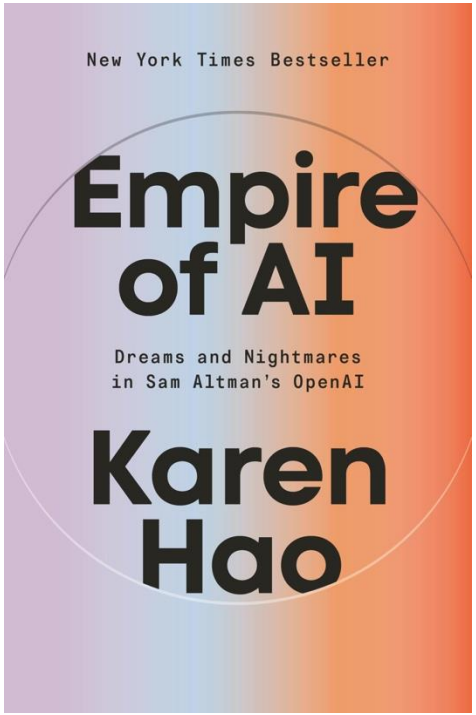
Book recommendations





CopeNLU Lab

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Isabelle's main research interests are natural language understanding, explainability and learning with limited training data.

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Pepa's research interests include the development, diagnostics, and application of explainability and interpretability techniques for NLP models.

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Dustin is a DDSA postdoctoral fellow, working on scientific natural language understanding and faithful text generation.

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Nadav Borenstein

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Nadav's research interests include improving the trustworthiness and usefulness of deep models in the NLP domain.

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Sarah broadly works in the area of computational social systems with a focus on news narrative and hate speech modelling. Her PhD at IIIT-Delhi was supported by fellowships from Google and PMRF.

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Erik's main research interests are question answering and explainability.

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Sara's research interests include explainable IR and NLP models, identifying biases in large text datasets, as well as working with social media data. She is a member of the DIKU ML section and IR group and co-advised by Isabelle.

Haeun Yu

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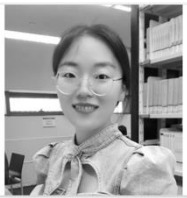
Haeun's main research interests include enhancing explainability in fact-checking and transparency of knowledge-enhanced LM.



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Sekh's research interests include explainability in fact checking and improving robustness and trustworthiness in NLP models.

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Lucas is an ELLIS PhD student at the University of Cambridge, supervised by Anna Corhonen and co-supervised by Isabelle. His research interests include machine learning, NLP and explainability.

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Thanks for your attention!

Questions?