

Accountable and Robust Automatic Fact Checking

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ECIR

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UNIVERSITY OF COPENHAGEN

Fact Checking

False Information Online



False claims such as 'drinking bleach or pure alcohol can cure the coronavirus infections': on the contrary, drinking bleach or pure alcohol can be very harmful. **Belgium's Poison Control Centre has recorded an increase of 15% in the number of bleach-related incidents.**



Conspiracy theories, such as the claim that coronavirus is 'an infection caused by the world's elites for reducing population growth'. The scientific evidence is clear: the virus comes from a family of viruses originating in animals that include other viruses such as SARS and MERS.



Claims that '5G installations would be spreading the virus'. These theories had no specific substantiation and led to attacks on masts.



Types of False Information



Cleuci de Oliveira
@CLEUCI



A correction for the ages, by Brazilian news magazine
Veja

"The candidate likes to spend his free time reading
Tolstoy, and not watching Toy Story, as originally
reported"



Macaco Aéreo @macacoaereo

Eduardo Jorge disse que não fuma maconha, prefere Tolstói. A
reportagem da Veja entendeu errado, e o mal-entendido deu
origem ao melhor Erramos deste século.
veja.abril.com.br/politica/eduar...

*Ao contrário do que informava esta nota, o candidato do PV à Presidência,
Eduardo Jorge, não afirmou que tem entre seus passatempos assistir ao
desenho Toy Story. O verde referia-se ao escritor russo Leon Tolstói (1828-1910),
autor de clássicos como Anna Karenina e Guerra e Paz. O site de VEJA pede
desculpas aos seus leitores e ao candidato.*

3:38 AM · Sep 7, 2018



18K



5.4K



Copy link to Tweet

Types of False Information

Linkbait Title Generator

Enter a subject and get link bait title ideas.

- 10 ways marketers are making you addicted to fact checking
- the most boring article about fact checking you'll ever read
- 10 ways fact checking can suck the life out of you
- 11 ways investing in fact checking can make you a millionaire

<http://www.contentrow.com/tools/link-bait-title-generator>

Types of False Information

Mars indfører indrejseforbud efter fund af britisk virusvariant

📅 23 februar, 2021

📁 Udland, Videnskab



Foto: Merlinus74, Bigstock

Den britiske mutation B.117 er blevet fundet på Mars efter, at en videnskabelig mission fra Jorden har nedsat udstyr på planeten. For at forhindre uhæmmet smittespredning indfører planeten totalforbud mod indrejse, indtil videre i fire uger.

<https://rokokoposten.dk/2021/02/23/mars-indfoerer-indrejseforbud-efter-fund-af-britisk-virusvariant/>

Types of False Information

BIRTH CONTROL MAKES WOMEN UNATTRACTIVE AND CRAZY



The media never stops banging on about women's health, particularly in the wake of the disgusting revelations about Planned Parenthood. They're always telling us that "women's bodies" and "women's choices" should be paramount. But just how healthy are the solutions to unwanted pregnancy that they propose?

SIGN UP FOR OUR NEWSLETTER

email address

SUBMIT



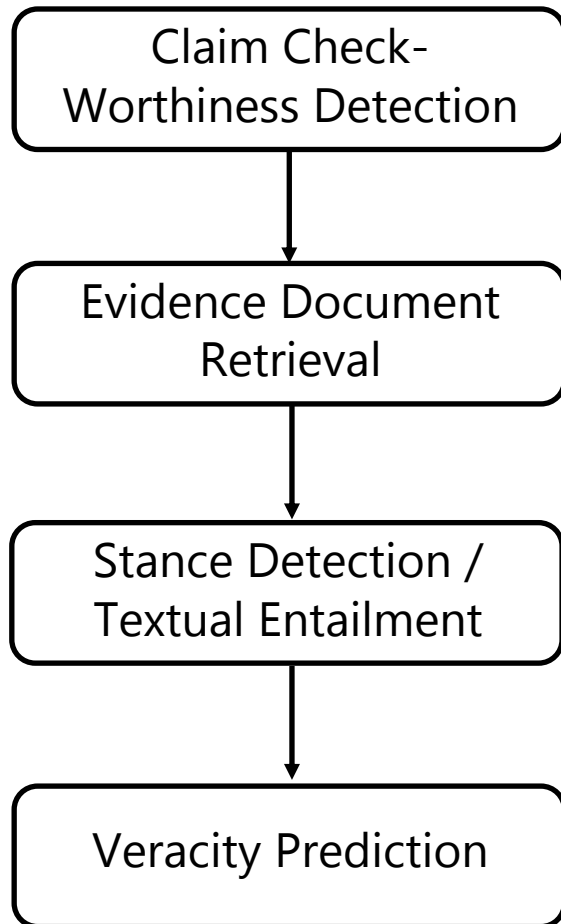
Types of False Information

- Disinformation:
 - Intentionally false, spread deliberately
- Misinformation:
 - Unintentionally false information
- Clickbait:
 - Exaggerating information and under-delivering it
- Satire:
 - Intentionally false for humorous purposes
- Biased Reporting:
 - Reporting only some of the facts to serve an agenda

Types of False Information

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Full Fact Checking Pipeline

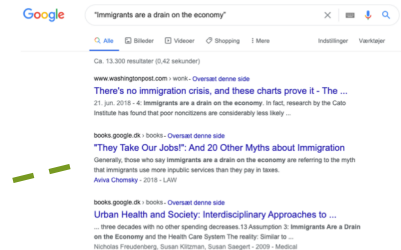


"Immigrants are a drain on the economy"

not check-worthy

check-worthy

"Immigrants are a drain on the economy"



"Immigrants are a drain on the economy", "EU immigrants have a positive impact on public finances"

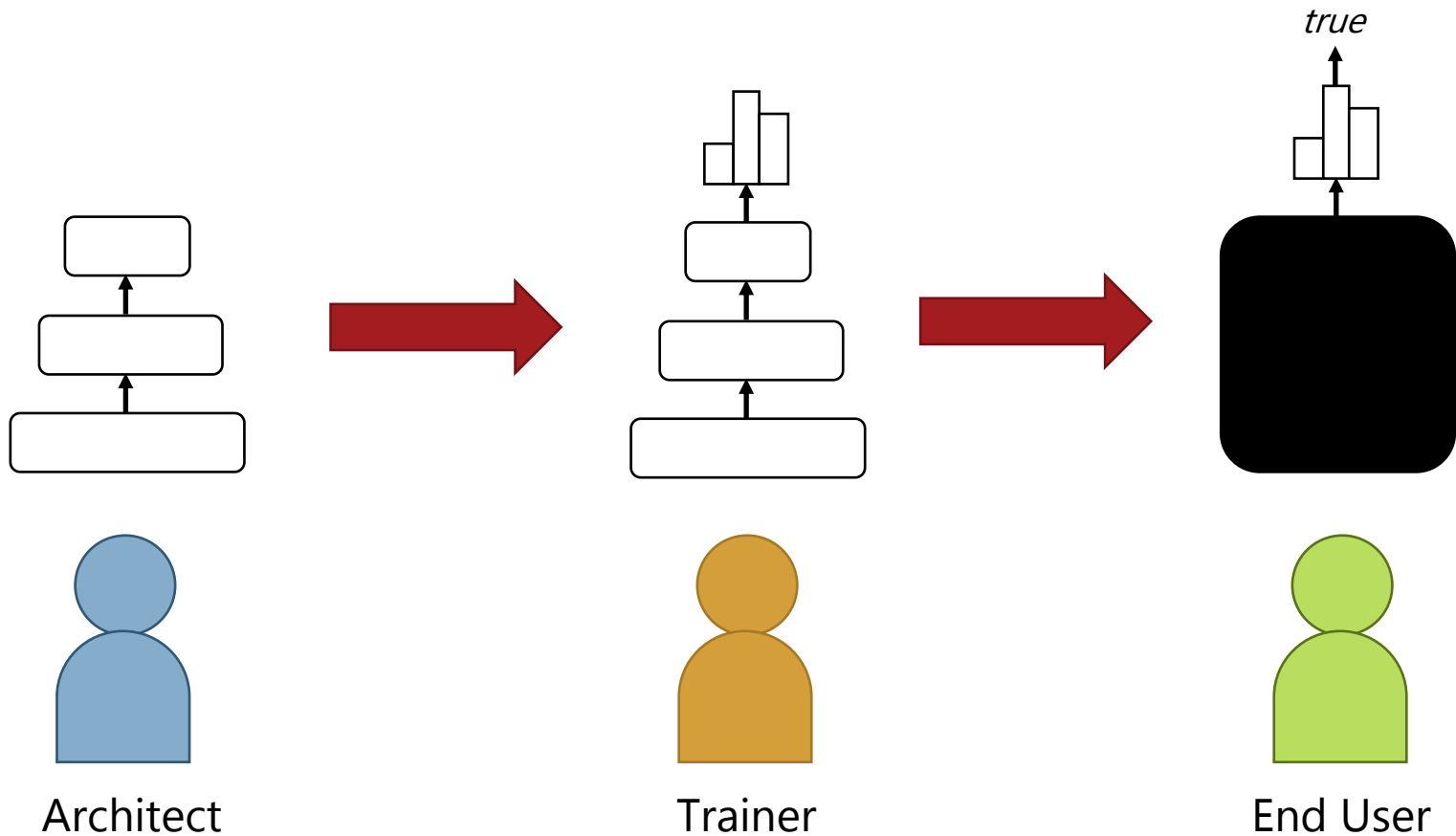
positive
neutral
negative

"Immigrants are a drain on the economy"

true
not enough info
false

Explainability

Explainability – what is it and why do we need it?



Terminology borrowed from Strobel et al. (2018), "LSTMVis: A tool for visual analysis of hidden state dynamics in recurrent neural networks. IEEE transactions on visualization and computer graphics."

Explainability – what is it and why do we need it?

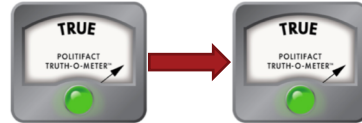
Right prediction

Wrong prediction

Right reasons

Claim: "In the COVID-19 crisis, 'only 20% of African Americans had jobs where they could work from home.'"

Evidence: "20% of black workers said they could work from home in their primary job, compared to 30% of white workers."



Claim: "Children don't seem to be getting this virus."

Evidence: "There have been no reported incidents of infection in children."



Claim: "Taylor Swift had a fatal car accident."

Reason: overfitting to spurious patterns (celebrity death hoaxes are common)



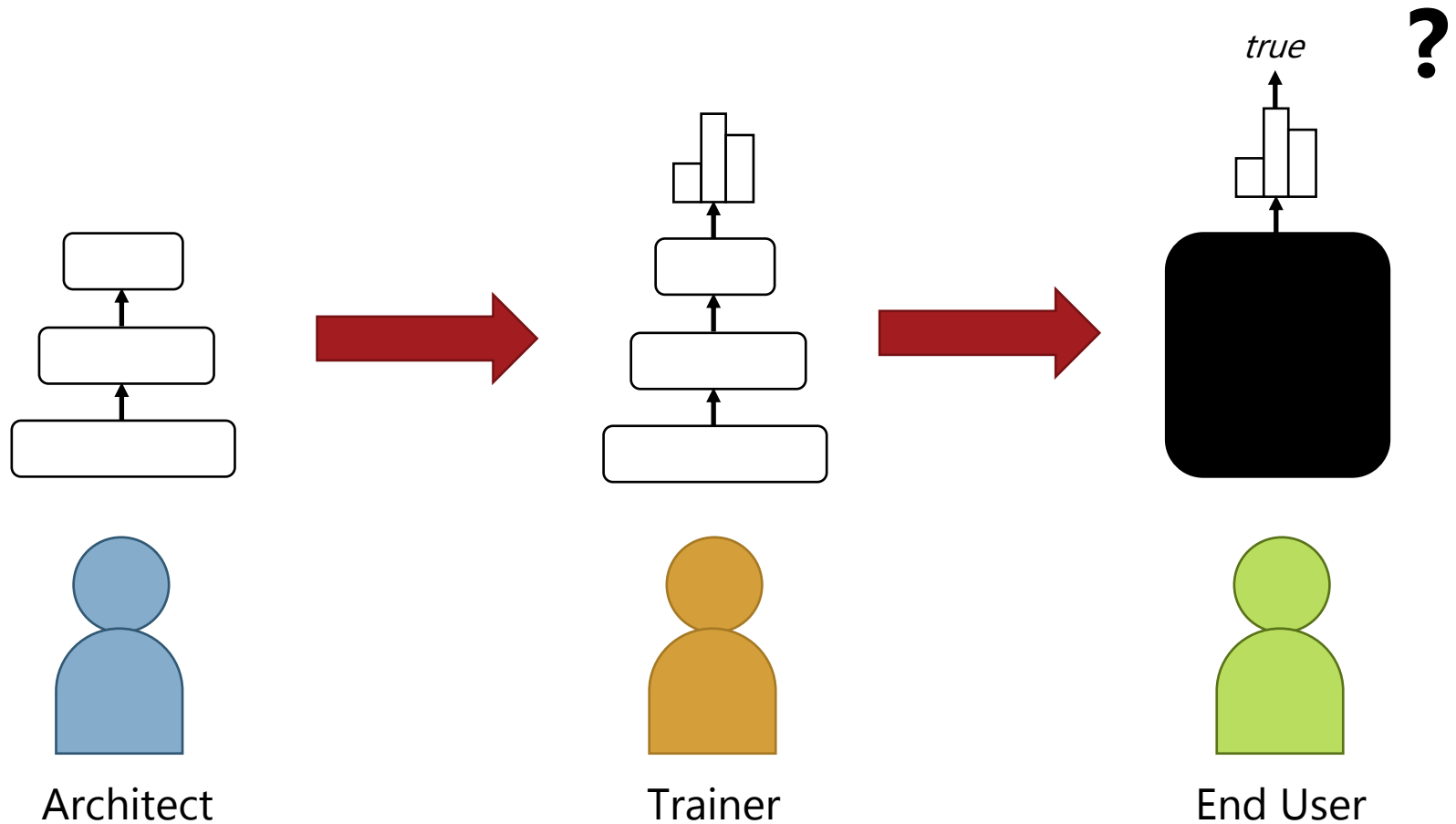
Claim: "Michael Jackson is still alive, appears in daughter's selfie."

Reason: overfitting to spurious patterns (celebrity death hoaxes are common)

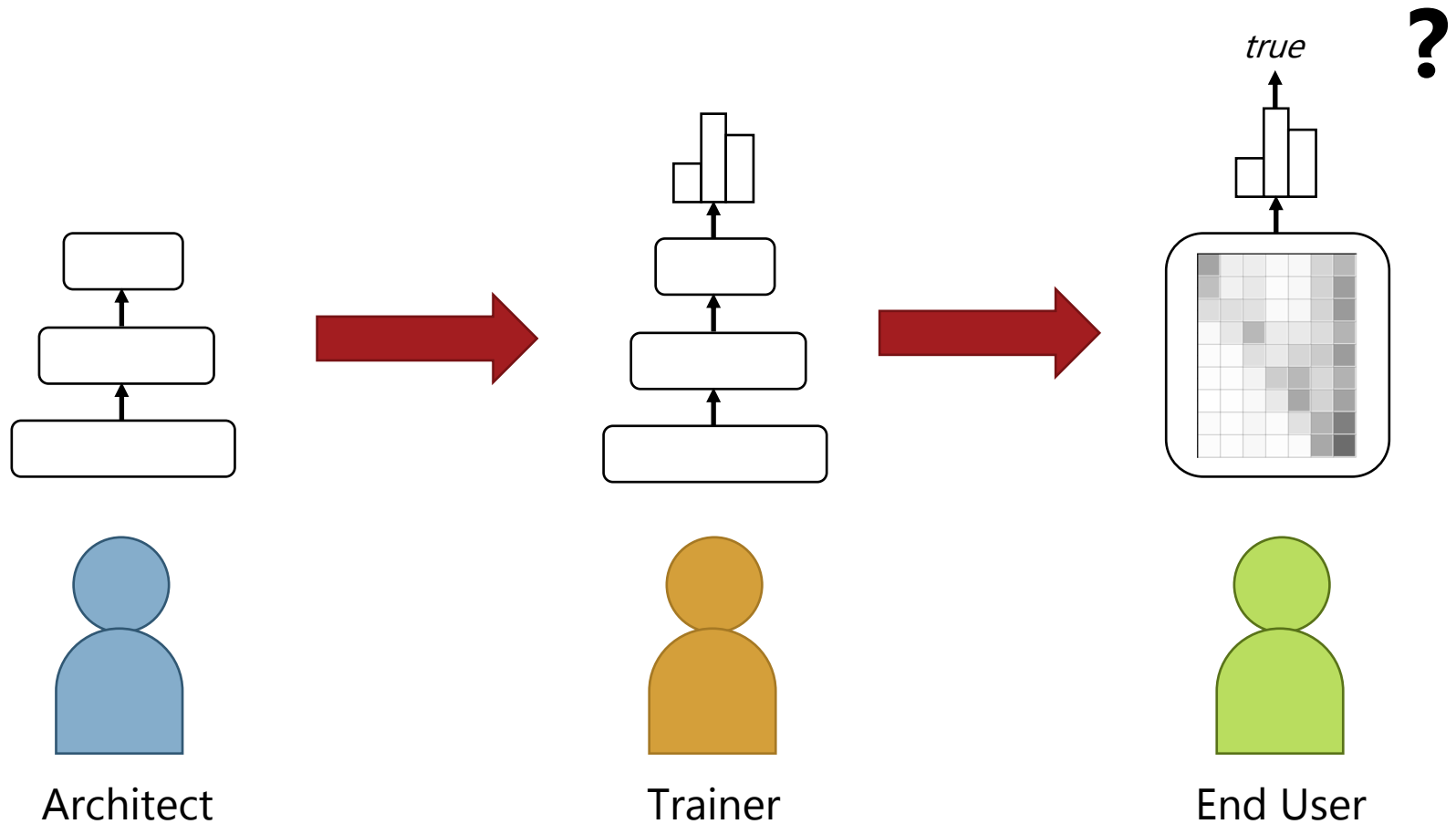


Wrong reasons

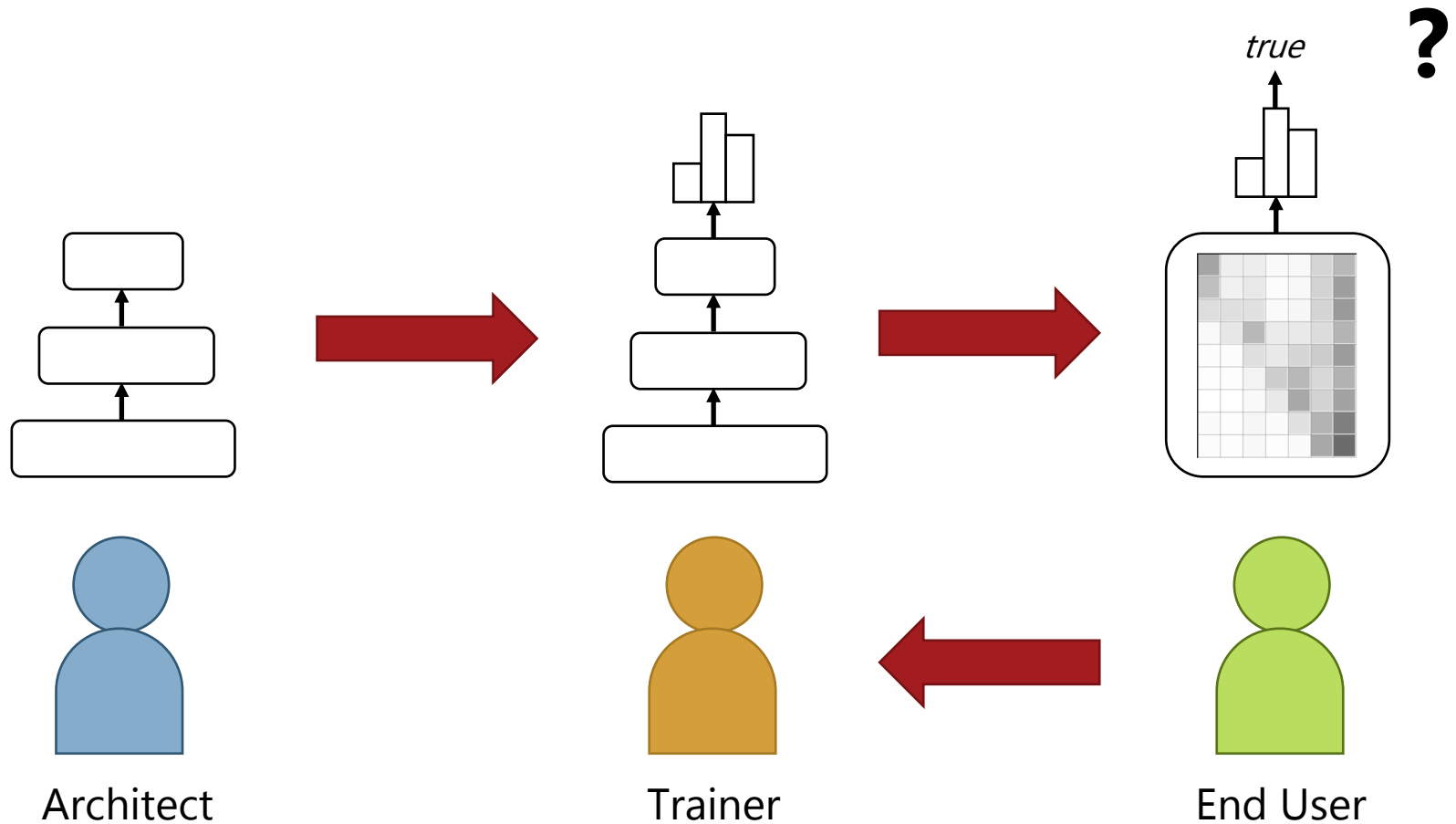
Explainability – what is it and why do we need it?



Explainability – what is it and why do we need it?



Explainability – what is it and why do we need it?



Overview of Today's Talk

- **Introduction**

- Fact checking – detecting false information online
- Explainability – making the right prediction for the right reasons

- **Diagnostic Properties**

- Evaluating explanations using different diagnostic properties
- Diagnostics-guided explanation generation

- **Generating Free-Text Explanations for Fact Checking**

- Supervised generation of explanations
- Unsupervised post-editing of explanations

- **Detecting Vulnerabilities of Fact Checking Models**

- Prediction without sufficient evidence

- **Wrap-up**

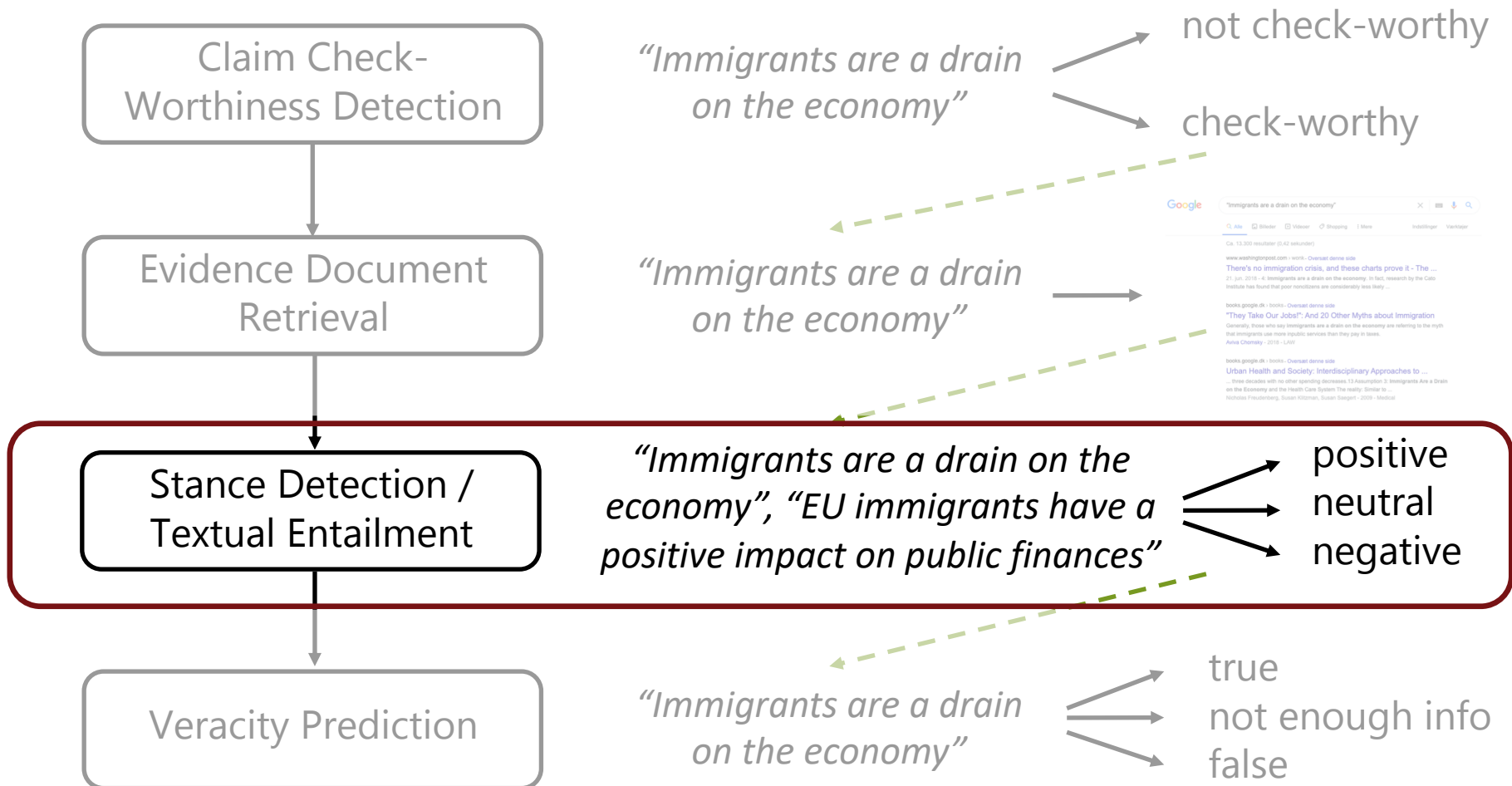
Diagnostic Properties

A Diagnostic Study of Explainability Techniques for Text Classification

Pepa Atanasova, Jakob Grue Simonsen,
Christina Lioma, Isabelle Augenstein

EMNLP 2020

A Typical Fact Checking Pipeline



Explanation via Rationale Selection -- Sentences

Claim: The Faroe Islands are no longer part of a kingdom.

Evidence document: The Faroe Islands, also called the Faeroes, is an archipelago between the Norwegian Sea and the North Atlantic, about halfway between Norway and Iceland, 200 mi north-northwest of Scotland. The Islands are an autonomous country within the Kingdom of Denmark. Its area is about 1,400 km² with a population of 50,030 in April 2017.
(...)

Label: refutes

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(...)

Label: refutes

Explanation via Rationale Selection -- Words

Hypothesis: An adult dressed in black holds a stick.

Premise: An adult is walking away empty-handedly.

Label: contradiction

Oana-Maria Camburu, Tim Rocktäschel, Thomas Lukasiewicz, Phil Blunsom (2018). e-SNLI: Natural Language Inference with Natural Language Explanations. In Proc. of NeurIPS.

<https://proceedings.neurips.cc/paper/2018/hash/4c7a167bb329bd92580a99ce422d6fa6-Abstract.html>

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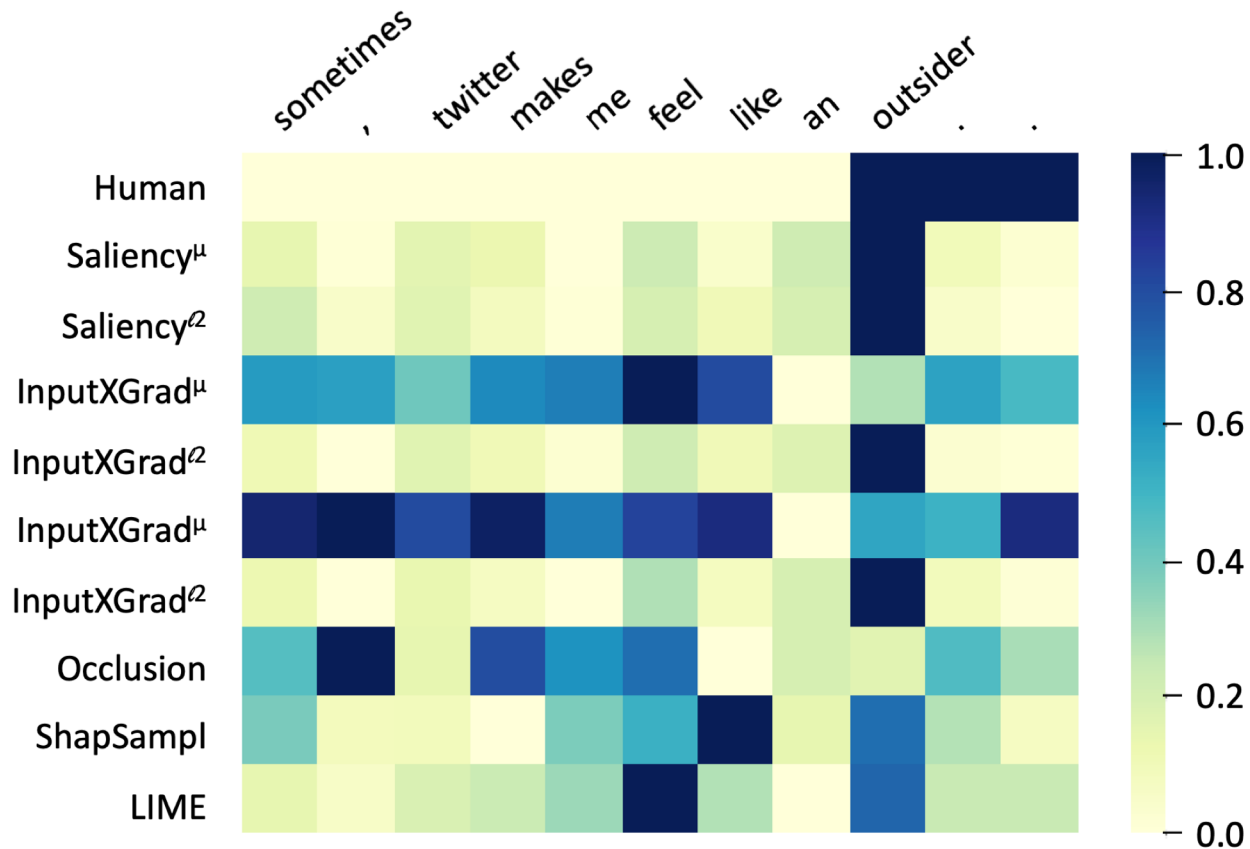
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Post-Hoc Explainability Methods via Rationale Selection



Example from Twitter Sentiment Extraction (TSE) dataset

Post-Hoc Explainability Methods via Rationale Selection

Gradient-Based Approaches

- Compute gradient of input w.r.t. output
- Different aggregation methods used to produce one score per input token (mean average, L2 norm aggregation)
- *Saliency, InputX-Gradient, Guided Backpropagation*

Perturbation-Based Approaches

- Replace tokens in input with other tokens to compute their relative contributions
- *Occlusion, Shapley Value Sampling*

Simplification-Based Approaches

- Train local linear models to approximate local decision boundaries
- *LIME*

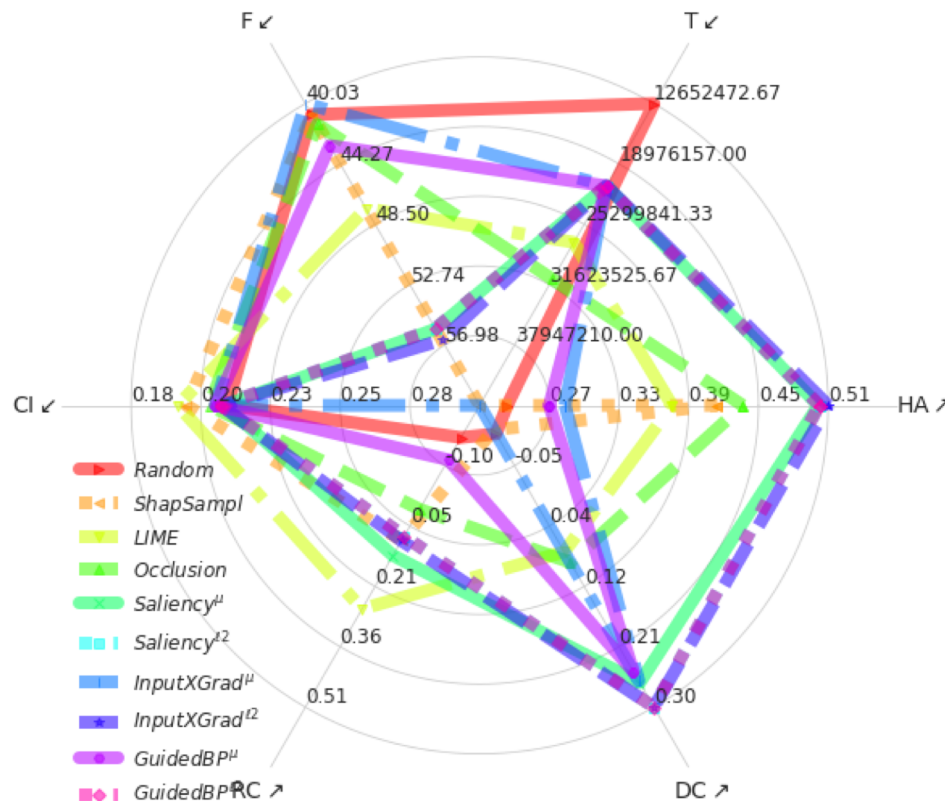
Post-Hoc Explainability via Rationale Selection: Research Questions

- How can explainability methods be evaluated?
 - Proposal: set of **diagnostic properties**
- What are characteristics of different explainability methods?
- How do explanations for models with different architectures differ?
- How do automatically and manually generated explanations differ?

Post-Hoc Explainability Methods via Rationale Selection: Diagnostic Properties

- **Agreement with Human Rationales (HA)**
 - Degree of overlap between human and automatic saliency scores
- **Faithfulness (F)**
 - Mask most salient tokens, measure drop in performance
- **Rationale Consistency (RC)**
 - Difference between explanations for models trained with different random seeds, with model with random weights
- **Dataset Consistency (DC)**
 - Difference between explanations for similar instances
- **Confidence Indication (CI)**
 - Predictive power of produced explanations for model's confidence

Post-Hoc Explainability Methods via Rationale Selection: Selected Results

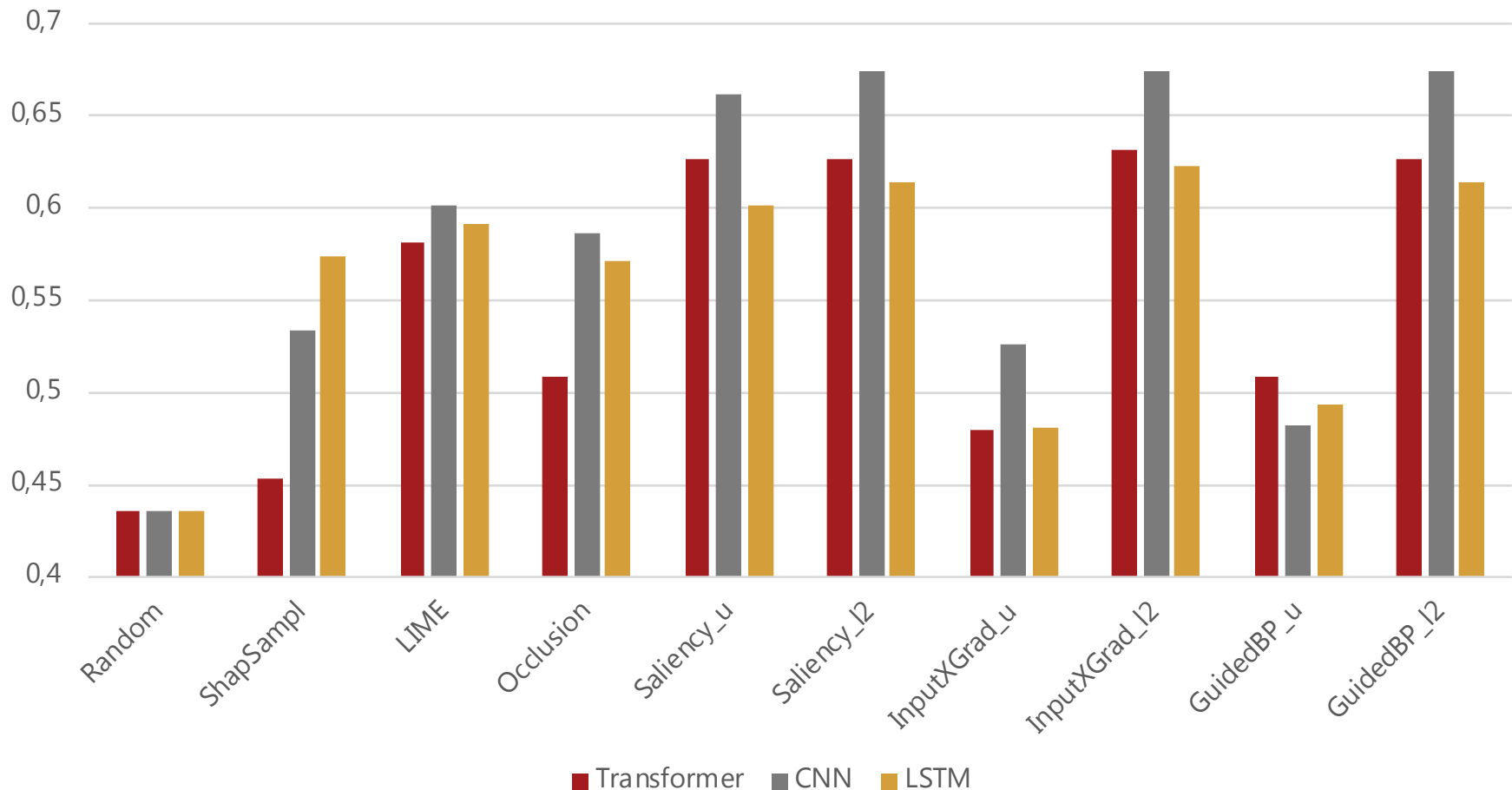


HA: Agreement with human rationales
F: Faithfulness
RC: Rationale Consistency
DC: Dataset Consistency
CI: Confidence indication
T: Computing Time

Spider chart for Transformer model on e-SNLI

Post-Hoc Explainability Methods via Rationale Selection: Aggregated Results

Mean of diagnostic property measures for e-SNLI



Summary

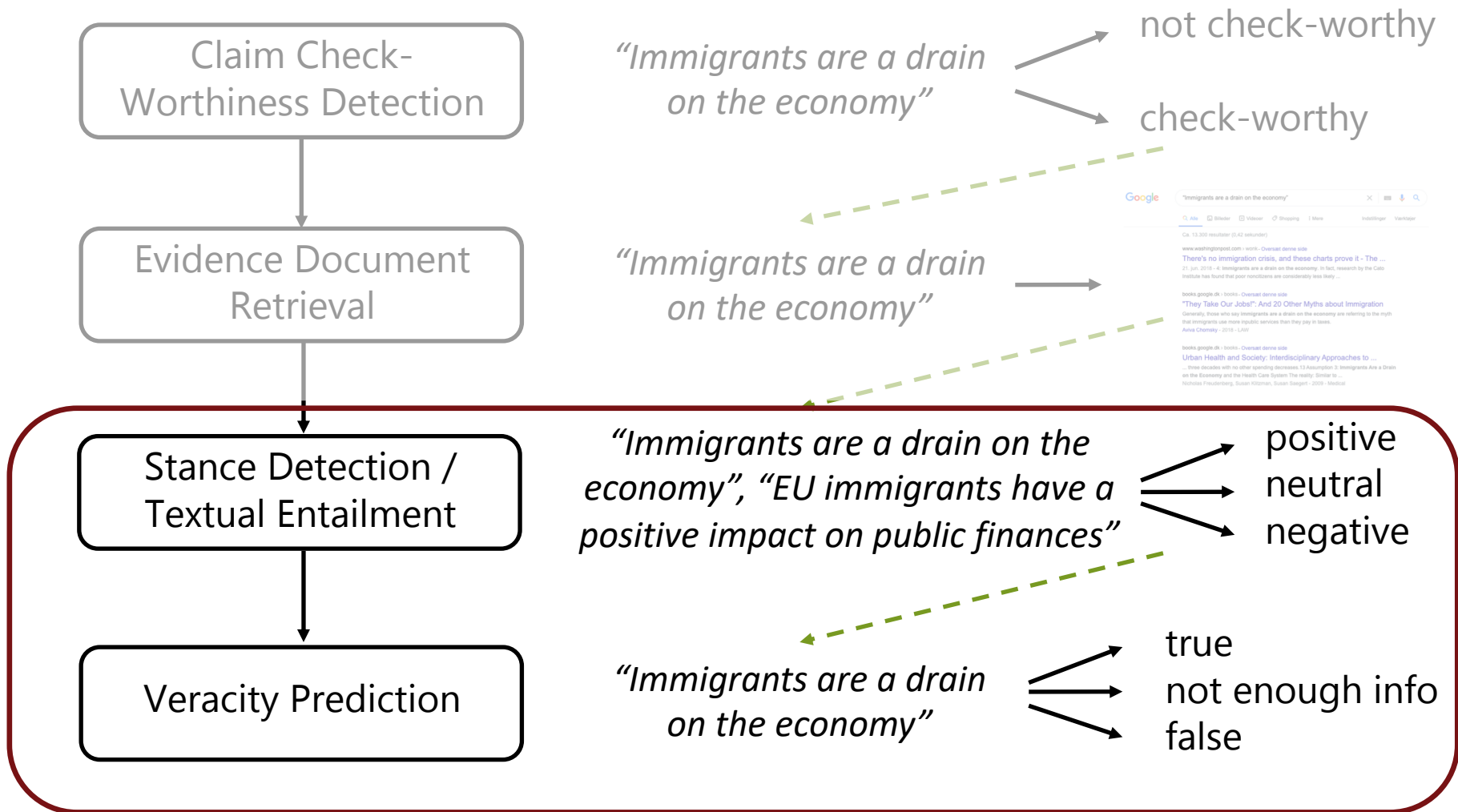
- Different explainability techniques
 - Produce different explanations
 - More than one explanation can be correct
 - Different aspects to what makes a good explanation
- Diagnostic properties allow one to assess different aspects of explainability techniques

Diagnostics-Guided Explanation Generation

Pepa Atanasova, Jakob Grue Simonsen,
Christina Lioma, Isabelle Augenstein

AAAI 2022

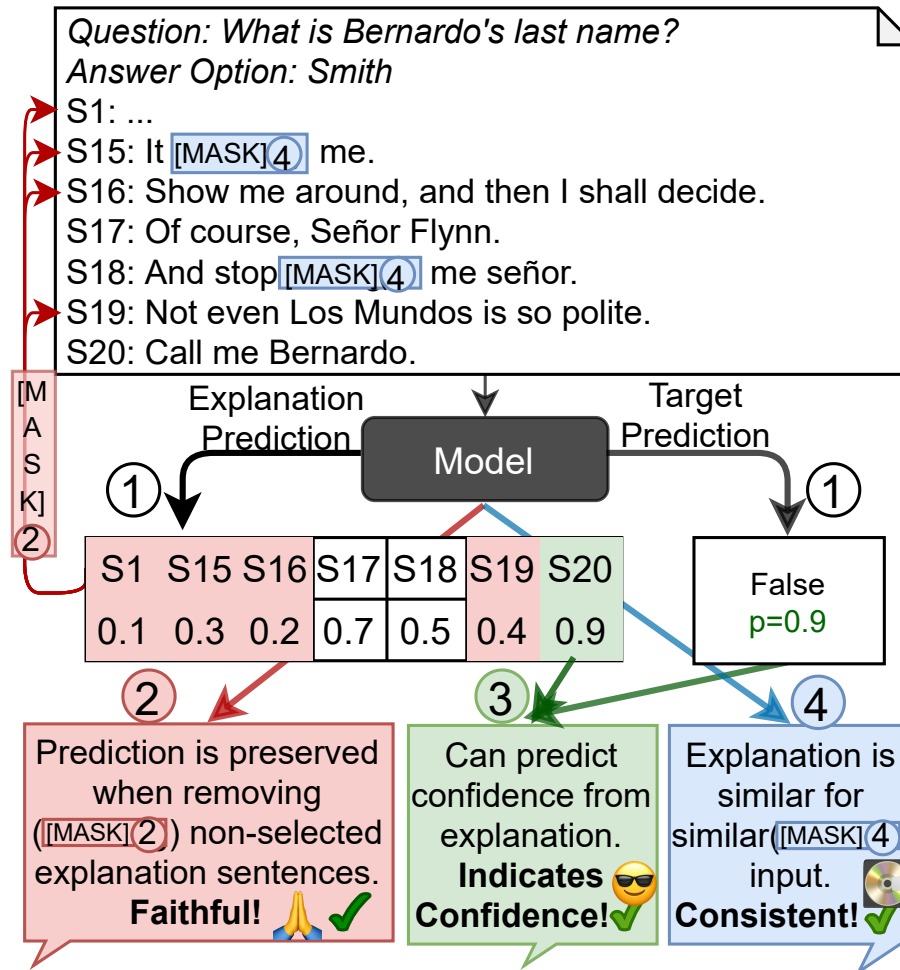
A Typical Fact Checking Pipeline



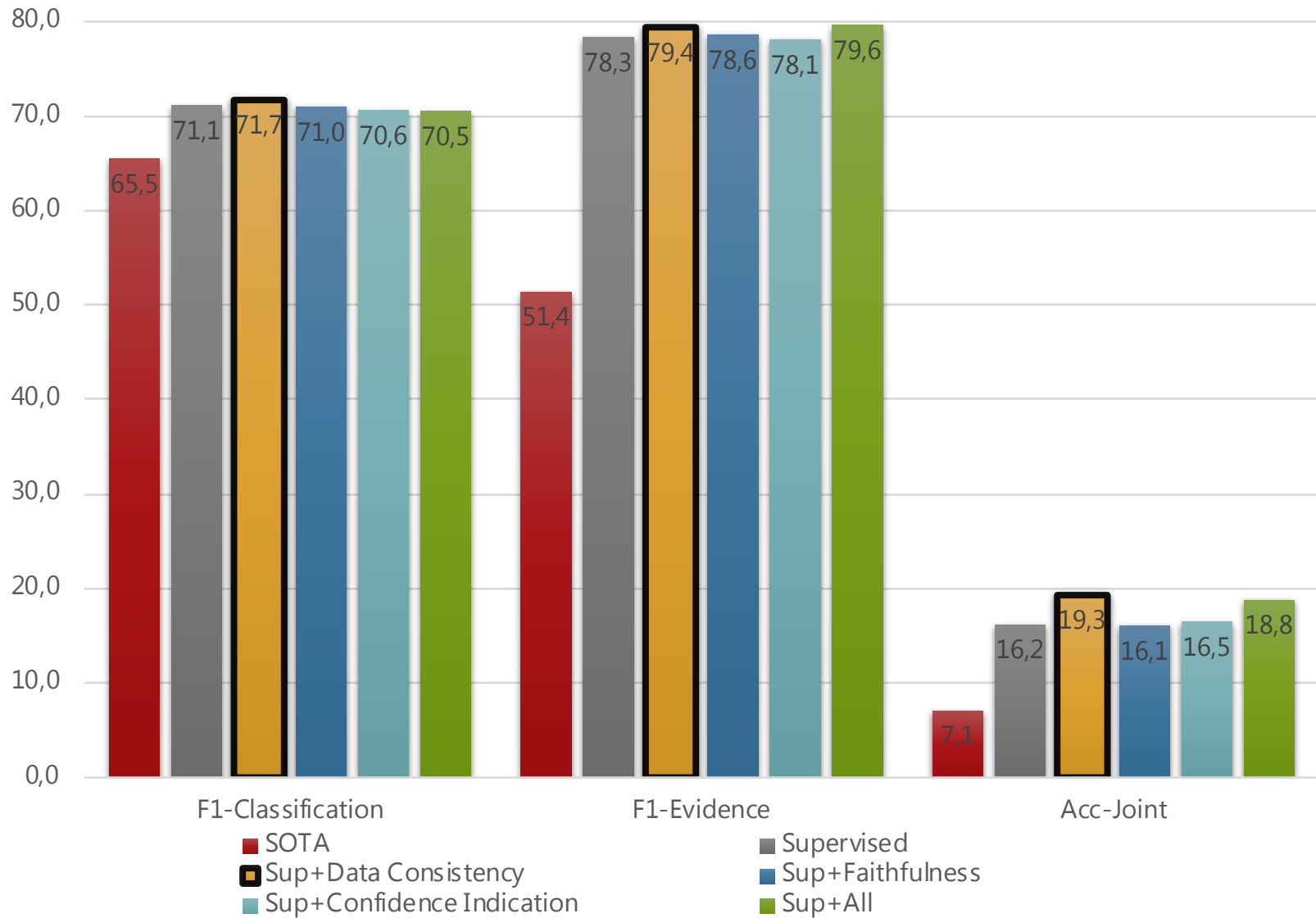
Diagnostic-Guided Explanations

- Explanations currently suffer from:
 - 🙏 lack of faithfulness to the underlying model
 - 📀 lack of consistency for similar data points
 - 😎 lack of confidence in explanation predictions
- We present the first method to learn those diagnostic properties in an unsupervised way
- We directly optimise for them to improve the quality of generated explanations

Diagnostic-Guided Explanations



Selected Results (MultiRC)



Effects of Diagnostic Property Objectives



Data Consistency

- Improves explanations by removing sentences unrelated to the target prediction.

Question: What colors are definitely used in the picture Lucy drew?

Answer: Yellow and purple **Label:** True

Supervised

Pred: True, $p=.98$

E: She makes sure to draw her mom named Martha wearing a purple dress, because that is her favorite. She draws many yellow feathers for her pet bird named Andy. She draws a picture of her family.

Remove unrelated

Supervised + DC

Pred: True, $p=.99$

E: She makes sure to draw her mom named Martha wearing a purple dress, because that is her favorite. She draws many yellow feathers for her pet bird named Andy.

Keep related to target prediction

Effects of Diagnostic Property Objectives



Faithfulness

- Improves the explanations to reflect the rationale used to predict the target for instances classified correctly by the supervised model.

Claim: Zoey Deutch did not portray Rosemarie Hathaway in Vampire Academy. **Label:** REFUTE

Supervised

Pred: Refute, $p=.99$

E: Zoey Francis Thompson Deutch (born November 10, 1994) is an American actress.

Does not lead to correct prediction

Supervised + F

Pred: Refute, $p=.99$

E: She is known for portraying Rosemarie "Rose" Hathaway in Vampire Academy(2014), Beverly in the Richard Link later film Everybody Wants Some!!

Effects of Diagnostic Property Objectives

😎 Confidence Indication

- Re-calibrates the prediction probabilities of generated explanations and predicted target tasks.
- Does not change many target predictions.

Label: Positive

E (both Sup-Sup-CI): For me, they calibrated my creativity as a child; they are masterful, original works of art that mix moving stories with what were astonishing special effects at the time (and they still hold up pretty well).

Predicted Sup: **negative**, $p=.99$

Predicted Sup+CI: positive, $p=.99$

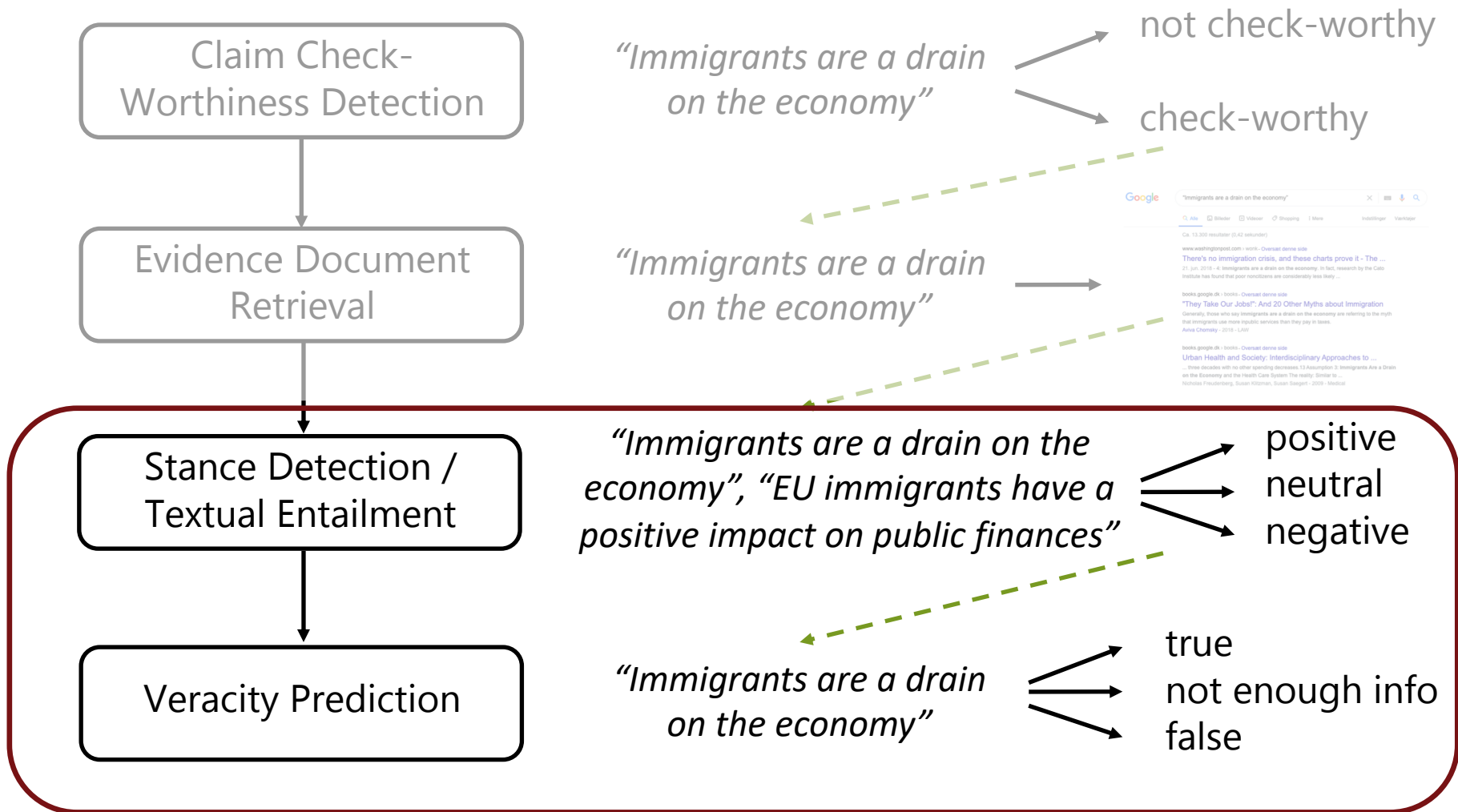
Not aligned with the high confidence of the explanation for the positive class.

Generating Fact Checking Explanations

Pepa Atanasova, Jakob Grue Simonsen,
Christina Lioma, Isabelle Augenstein

ACL 2020

A Typical Fact Checking Pipeline



Extracted Wikipedia Sentences as Explanations

- FEVER dataset (Thorne et al., 2018)
 - Claims are re-written sentences from Wikipedia.
 - Explanation consists of evidence sentences extracted from Wikipedia pages.
 - Biggest dataset for fact checking
 - Brings small or no improvements for veracity prediction with real-world claims (Wadden et al., 2020, Ostrowski et al., 2020)

Claim: The Rodney King riots took place in the most populous county in the USA.

[wiki/Los Angeles Riots]

The 1992 Los Angeles riots, also known as the Rodney King riots were a series of riots, lootings, arsons, and civil disturbances that occurred in Los Angeles County, California in April and May 1992.

[wiki/Los Angeles County]

Los Angeles County, officially the County of Los Angeles, is the most populous county in the USA.

Verdict: Supported

Thorne et al., 2018 "FEVER: a large-scale dataset for Fact Extraction and VERification"

Thorne et al., 2021 "Evidence-based Verification for Real World Information Needs"

Ostrowski et al., 2021 "Multi-Hop Fact Checking of Political Claims"

Wadden et al., 2020 "Fact or Fiction: Verifying Scientific Claims"

Real World Fact Checking

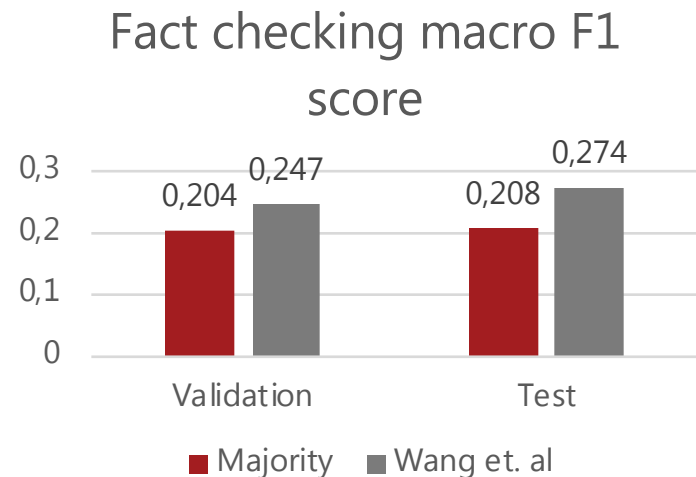
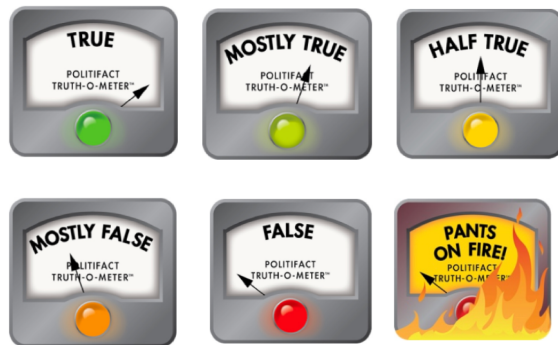
Statement: "The last quarter, it was just announced, our gross domestic product was below zero. Who ever heard of this? Its never below zero."

Speaker: Donald Trump

Context: presidential announcement speech

Label: Pants on Fire

Dataset Statistics	
Training set size	10,269
Validation set size	1,284
Testing set size	1,283
Avg. statement length (tokens)	17.9
Top-3 Speaker Affiliations	
Democrats	4,150
Republicans	5,687
None (e.g., FB posts)	2,185



Real-World Fact Checking Explanations

Donald Trump's claim that US tested more than all countries combined is Parts on Five wrong

IF YOUR TIME IS SHORT

- The United States has tested more than 1 million people for COVID-19.
- The United States has tested more than 1 million people for COVID-19.
- The United States has tested more than 1 million people for COVID-19.

According to a report from the White House, the United States has tested more than 1 million people for COVID-19. This is a claim that has been widely reported in the media.

But is it true? Let's take a closer look at the numbers.

First, let's look at the United States. According to the White House, the United States has tested more than 1 million people for COVID-19.

Now let's look at the rest of the world. According to the World Health Organization, the total number of COVID-19 tests performed worldwide is approximately 1.5 million.

So, if we add up the number of tests performed in the United States and the rest of the world, we get a total of approximately 2.5 million tests.

But wait, there's more. The White House report also states that the United States has tested more than 1 million people for COVID-19. This is a claim that has been widely reported in the media.

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Our ruling

Trump claimed that the United States has tested more than every country combined."

There is no reasonable way to conclude that the American system has run more diagnostics than "all other major countries combined." Just by adding up a few other nations' totals, you can quickly see Trump's claim fall apart.

Ruling
Comments

Generating Explanations from Ruling Comments

Claim:

We've tested more than every country combined.

Ruling Comments:

Responding to weeks of criticism over his administration's COVID-19 response, President Donald Trump claimed at a White House briefing that the United States has well surpassed other countries in testing people for the virus.

"We've tested more than every country combined,"

Trump said April 27 [...] We emailed the White House for comment but never heard back, so we turned to the data. Trump's claim didn't stand up to scrutiny.

In raw numbers, the United States has tested more people than any other individual country — but nowhere near more than "every country combined" or, as he said in his tweet, more than "all major countries combined."[...] The United States has a far bigger population than many of the "major countries" Trump often mentions. **So it could have run far more tests but still have a much larger burden ahead than do nations like Germany, France or Canada.**[...]

Joint Model

Veracity Label

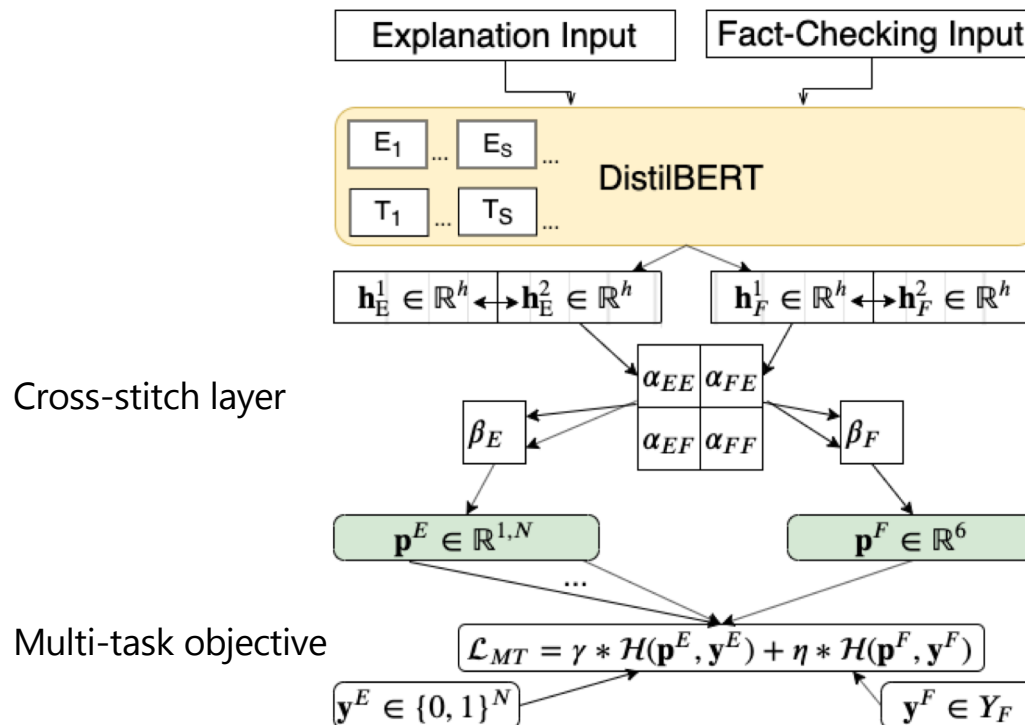
Justification/
Explanation

Related Studies on Generating Explanations

- *Camburu et. al; Rajani et. al* generate abstractive explanations
 - Short input text and explanations;
 - Large amount of annotated data.
 - Real world fact checking datasets are of limited size and the input consists of long documents
 - We take advantage of the LIAR-PLUS dataset:
 - Use the summary of the ruling comments as a gold explanation;
 - Formulate the problem as extractive summarization.
-
- *Camburu, Oana-Maria, Tim Rocktäschel, Thomas Lukasiewicz, and Phil Blunsom. "e-SNLI: Natural language inference with natural language explanations." In Advances in Neural Information Processing Systems,. 2018.*
 - *Rajani, Nazneen Fatema, Bryan McCann, Caiming Xiong, and Richard Socher. "Explain Yourself! Leveraging Language Models for Commonsense Reasoning." In Proceedings of the 57th Annual Meeting of the Association for Computational Linguistics, pp. 4932-4942. 2019.*

Joint Explanation and Veracity Prediction

Which sentences should be selected for the explanation?



Generating Free Text Explanations: Automatic Evaluation Measures

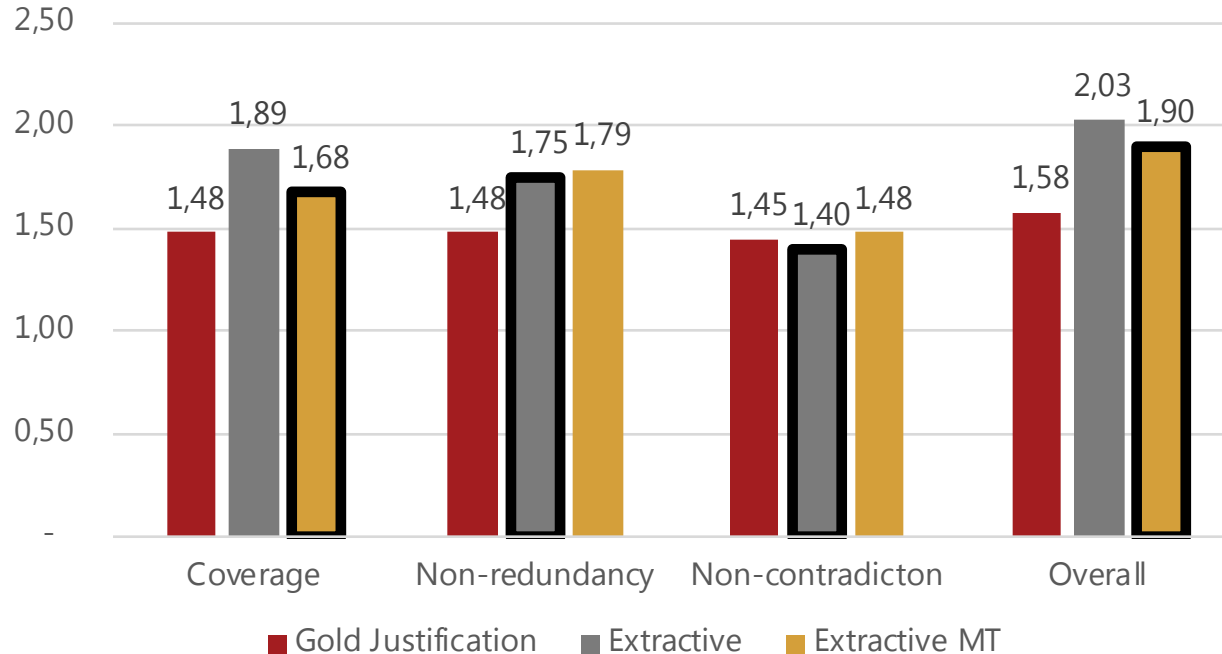
- ***Indirectly: Fact Checking Performance***
 - ***F1***. Predicting fact checking labels: using a joint model, from generated explanations, gold explanations, etc.
- ***Directly: Explanation Quality***
 - ***ROUGE-N***. Overlap of n-grams between the oracle and the generated summaries.
 - ***ROUGE-L***. Longest Common Subsequence (LCS) between the oracle and the generated summaries.

Generating Free Text Explanations: Manual Evaluation Measures

- ***Explanation Quality***
 - ***Coverage***. The explanation contains important, salient information and does not miss any points important to the fact check.
 - ***Non-redundancy***. The explanation does not contain any information that is redundant/repeated/not relevant to the claim or the fact check.
 - ***Non-contradiction***. The explanation does not contain any pieces of information contradictory to the claim or the fact check.
 - ***Overall***. Overall explanation quality.
- ***Explanation Informativeness***. Veracity label for a claim provided based on the automatically generated explanations only.

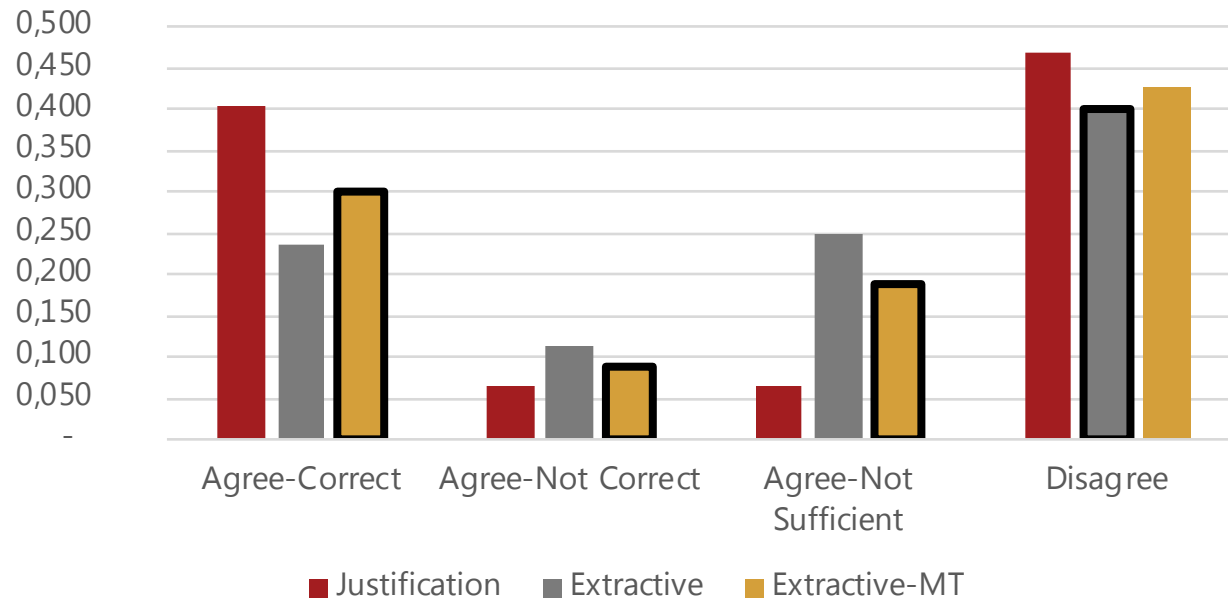
Selected Results: Explanation Quality

Mean Average Rank (MAR).
Lower MAR is better! (higher rank)



Explanation Informativeness

Manual veracity labelling, given a particular explanation as percentages of the dis/agreeing annotator predictions.



Summary

- First study on generating real-world veracity explanations
- Jointly training veracity prediction and explanation
 - improves the performance of the classification system
 - improves the coverage and overall performance of the generated explanations

Generating Fluent Fact Checking Explanations with Unsupervised Post- Editing

Shailza Jolly, Pepa Atanasova, Isabelle
Augenstein

Preprint, December 2021

Ongoing Work: Unsupervised Post-Editing

Claim	Label: False
EU suspends delivery of 10 million masks over quality issues.	
Explanation from Ruling Comments	
<p>After a first batch of 1.5 million masks was shipped to 17 of the 27 member states and Britain, 600,000 items did not have European certificates and medical standards. As part of its efforts to tackle the COVID-19 crisis, this month the EU's executive arm started dispatching the masks to health care workers. (R) It was set to be distributed in weekly installments over six weeks. (D) "We have decided to suspend future deliveries of these masks," Commission health spokesman Stefan De Keersmaecker said. (P)</p>	
Post-Edited Explanation	
<p>As part of its efforts to tackle the COVID-19 crisis, this month the EU's executive arm started dispatching the masks to health care workers. (R) After a first batch of 1.5 million masks was shipped to 17 of the 27 member states and Britain, 600,000 items did not have European certificates and did not comply with (I) medical standards. The Commission has decided to stop future deliveries of these masks, De Keersmaecker said. (P)</p>	

Fig. 1. Example of a post-edited explanation from PubHealth that was initially extracted from RCs. We illustrate four post-editing steps: reordering (R), insertion (I), deletion (D), and paraphrasing (P).

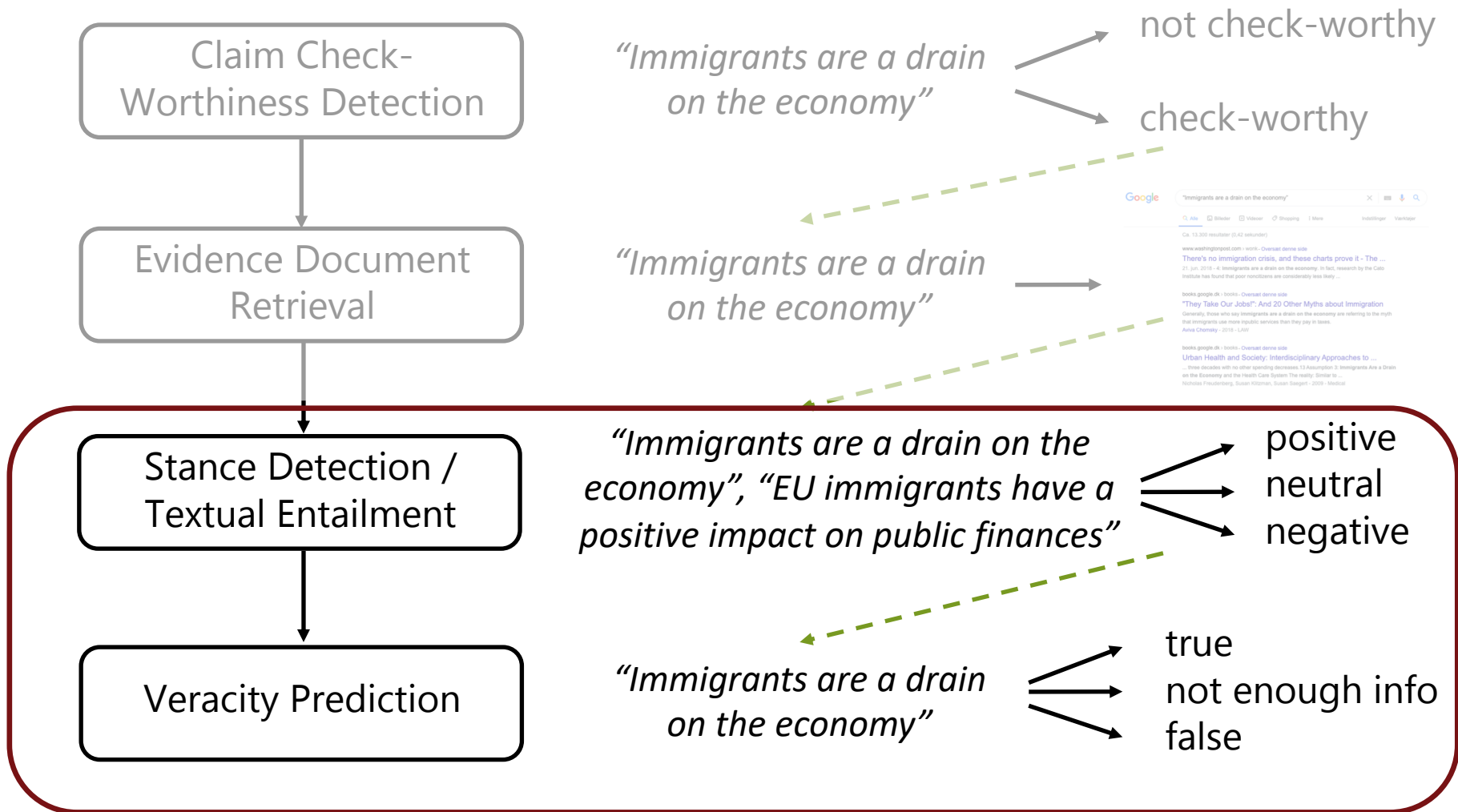
Vulnerabilities of Fact Checking Models

Fact Checking with Insufficient Evidence

Pepa Atanasova, Jakob Grue Simonsen,
Christina Lioma, Isabelle Augenstein

TACL, Vol 10 (2022), to appear

A Typical Fact Checking Pipeline

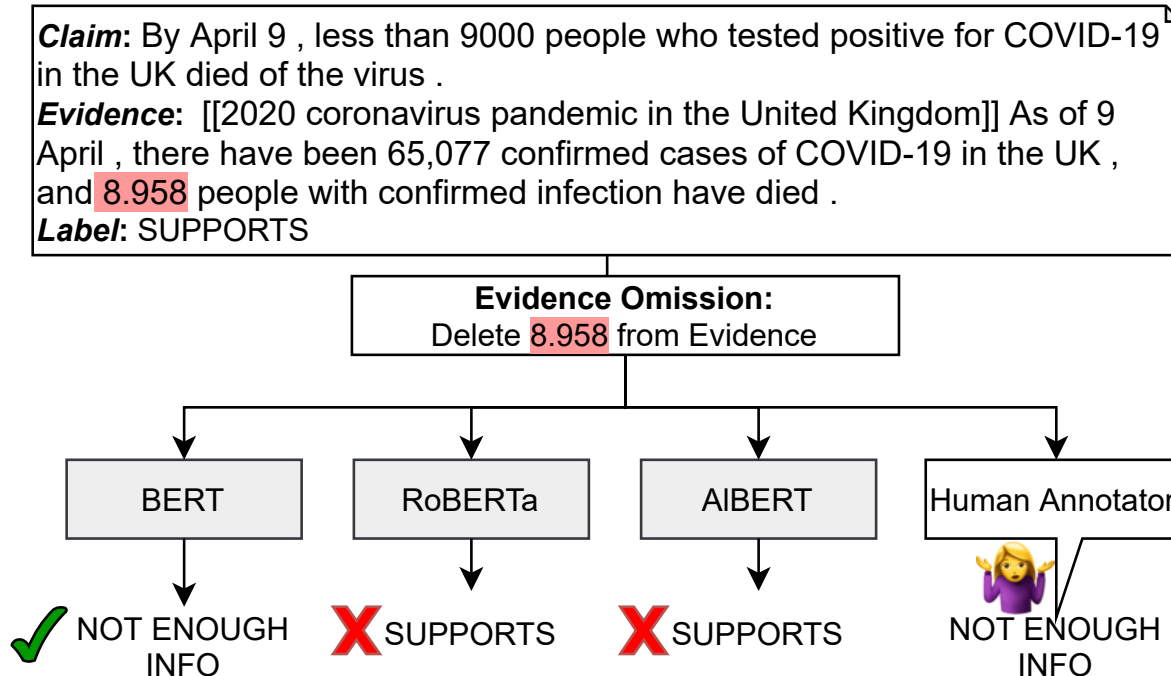


Known Vulnerabilities of Fact Checking Models

- Claim-only bias
 - 61.7 accuracy of a BERT given only the claim
- Lexical biases
 - E.g. negation and the REFUTES class
- Prior proposed solutions:
 - FEVERSymmetric dataset to measure the bias of fact checking models
 - Re-weighted training objective
 - Increase the importance of claims with different labels containing those phrases

Known Vulnerabilities of Fact Checking Models

- Prediction with insufficient evidence
 - SufficientFacts dataset – fluency-preserving removal of constituents, e.g. adjective/adverb/noun/number modifiers, prepositional phrases
 - Models still predict the original label in most cases:



Source Datasets

Dataset/Size	Example
FEVER 145,449 train 999,999 dev 999,999 test	Label: REFUTES ($\in \{\text{SUPPORTS, REFUTES, NOT ENOUGH INFO}\}$) Claim: Sindh borders Indian states and is in India. Evidence: [Sindh] Sindh is home to a large portion of Pakistan's industrial sector and contains two of Pakistan's commercial seaports – Port Bin Qasim and the Karachi Port.
Vitamin C 370,653 train 63,054 dev 55,197 test	Label: SUPPORTS ($\in \{\text{SUPPORTS, REFUTES, NOT ENOUGH INFO}\}$) Claim: Westlife sold more than 1 million video albums and made over 23.5 million sales in the UK. Evidence: [Westlife] According to the British Phonographic Industry (BPI), Westlife has been certified for 13 million albums, 1.3 million video albums, and 9.8 million singles, with a total of more than 24 million combined sales in the UK.
HoVer 18,171 train 1818 dev 4,000 test	Label: NOT SUPPORTED ($\in \{\text{SUPPORTS, NOT SUPPORTS}=(\text{REFUTES}+\text{NOT ENOUGH INFO})\}$) Claim: 2000's Reason Is Treason is the second single release from a British rock band that are not from England. The band known for the early 90's album Novelty (album) are not from England either. Evidence: [Kasabian] Kasabian are an English rock band formed in Leicester in 1997. [Jawbox] Jawbox was an American alternative rock band from Washington, D.C., United States. [Reason Is Treason] "Reason Is Treason" is the second single release from British rock band Kasabian. [Novelty (album)] Novelty is an album from the early 90's by Jawbox.



Evidence Omission

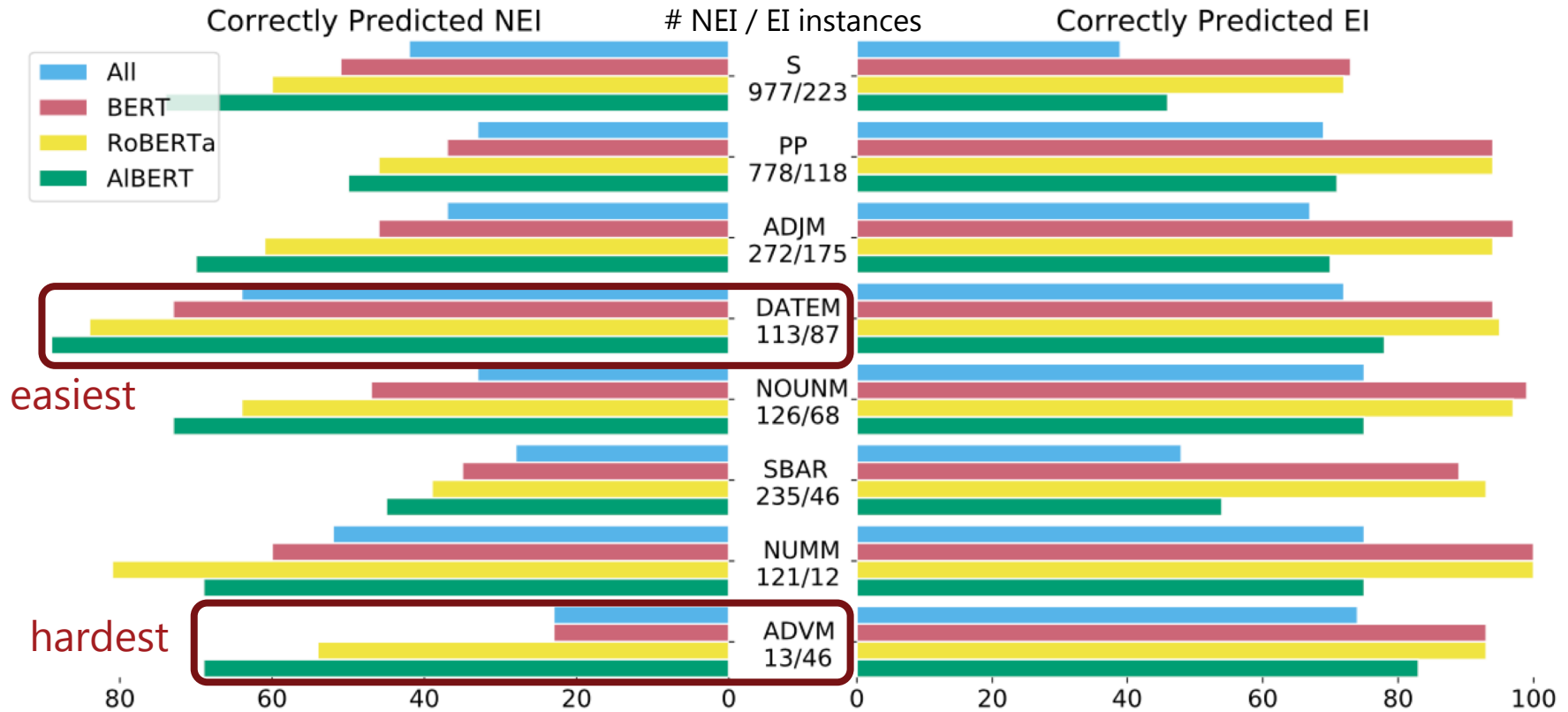
Type	L	Claim	Evidence
S	R	The Endless River is an album by a band formed in 1967.	[[The Endless River]] The Endless River is the fifteenth and final studio album by the English rock band Pink Floyd. [[Pink Floyd]] Pink Floyd were founded in 1965 by students ...
PP	R	Uranium-235 was discovered by Arthur Jeffrey Dempster in 2005.	[[Uranium-235]] It was discovered in 1935 by Arthur Jeffrey Dempster.
NOUNM	S	Vedam is a drama film.	[[Vedam (film)]] Vedam is a 2010 Telugu language Indian drama film written and directed by Radhakrishna Jagarlamudi, starring Allu Arjun ...
ADJM	S	Christa McAuliffe taught social studies at Concord High School.	[[Christa McAuliffe]] She took a teaching position as a social studies teacher at Concord High School in New Hampshire in 1983.
ADVM	S	Richard Rutowski heavily revised the screenplay for Natural Born Killers.	[[Natural Born Killers]] The film is based on an original screenplay that was heavily revised by writer David Veloz , associate producer Richard Rutowski ...
NUMM	S	Being sentenced to federal prison is something that happened to Efraim Diveroli.	[[Efraim Diveroli]] Diveroli was sentenced to four years in federal prison .
DATEM	R	Colombiana was released in 2001.	[[Colombiana]] Colombiana is a 2011 French action film ...
SBAR	R	North Vietnam existed from 1945 to 1978.	[[North Vietnam]] North Vietnam, was a state in South-east Asia which existed from 1945 to 1976.

Sentence (S), Prepositional Phrase (PP), Noun Modifier (NOUNM), Adjective Modifier (ADJM), Adverb Modifier (ADVM), Number Modifier (NUMM), Date Modifier (DATEM), Subordinate Clause (SBAR)

SufficientFacts Dataset Construction

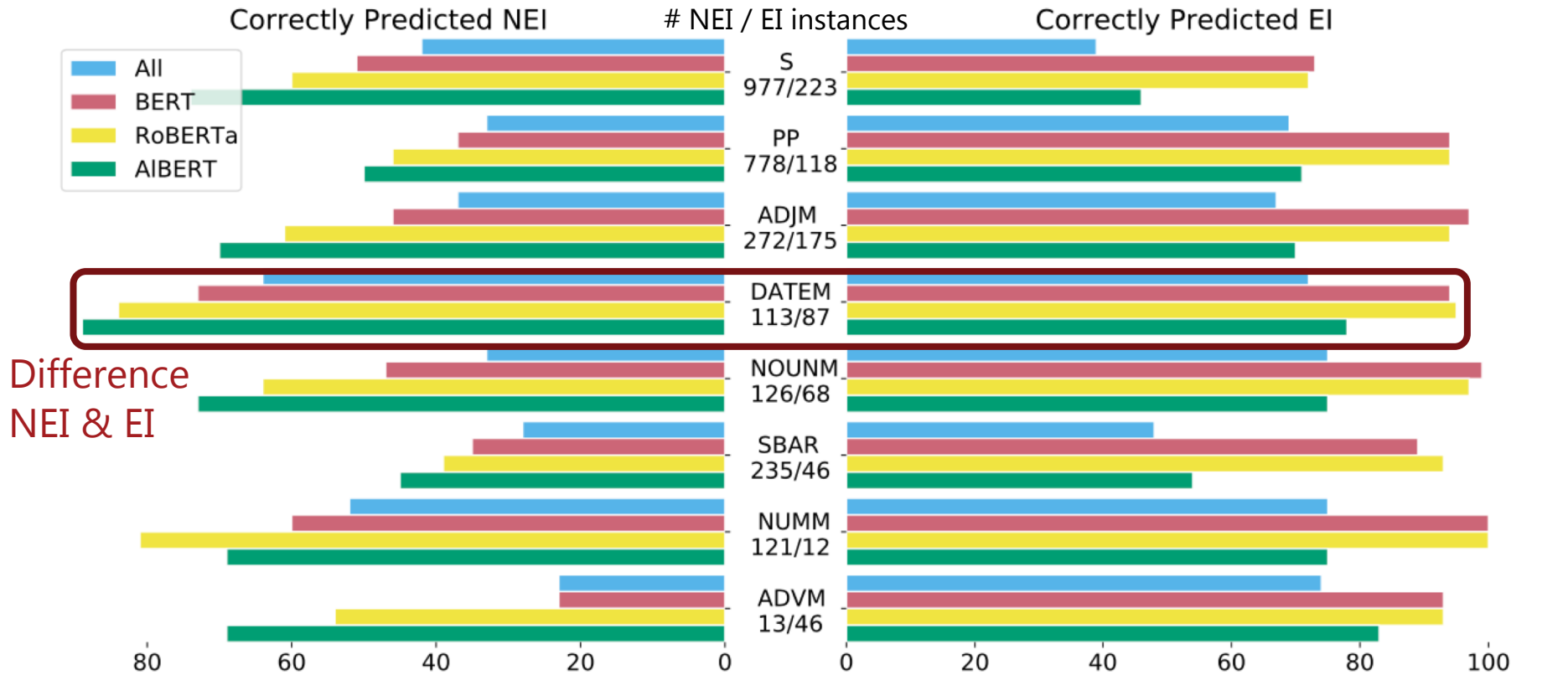
- Ensemble of 3 trained Transformer-based FC models (BERT, RoBERTa, ALBERT)
- Predict labels for instances from FC datasets (FEVER, HoVer, VitaminC) from which information has been removed
- Manually annotate instances for which:
 - Models agree that evidence is sufficient
 - Agree that evidence is insufficient
 - Disagree
- Overall findings from manual annotation
 - Model disagreements are mostly for NEI instances
 - When models agree that information is insufficient, they are correct in 97.2% of cases

Fine-Grained Analysis By Removal Type



All = all models agree on prediction; higher proportion of correct predictions -> better

Fine-Grained Analysis By Removal Type



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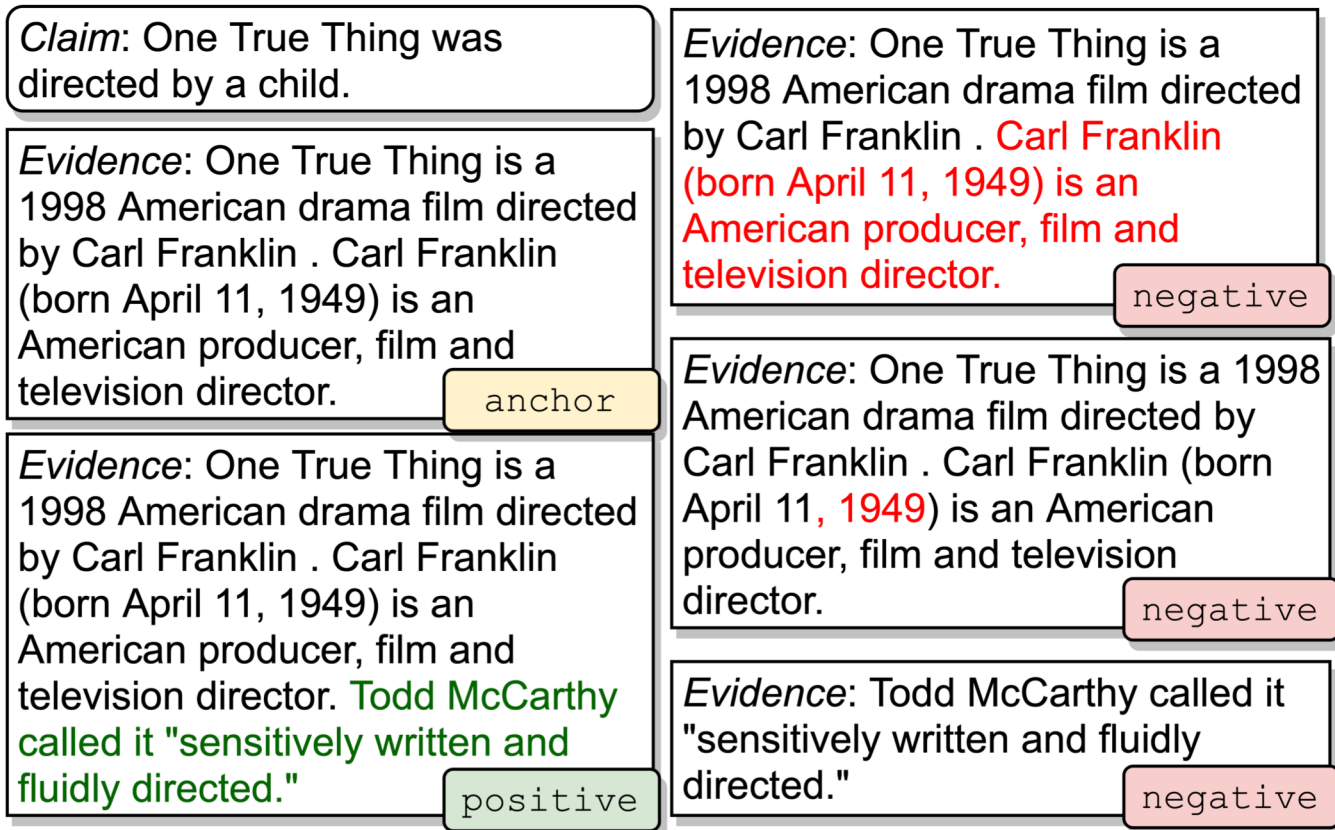
New Task: Evidence Omission Detection

- Given an instance, find a **distractor sentence** from the document of the gold evidence that is most semantically similar to claim (word overlap)
- Append distractor sentence to original evidence texts to serve as anchors
- **Omit information** from original evidence texts (using evidence omission procedure)

New Task: Evidence Omission Detection

- For each of the 3 trained Transformer-based FC models:
 - Select candidates instances for which *the other two supervised models previously predicted NEI*
 - Two types of **positive instances**:
 - Sentences from original evidence without distractor sentence
 - Sentence from original evidence + distractor sentence with highest word overlap with the claim, with one constituent omitted
 - **Negative instances**: original evidence with constituents/sentences removed that all three models predict as NEI

New Task: Evidence Omission Detection

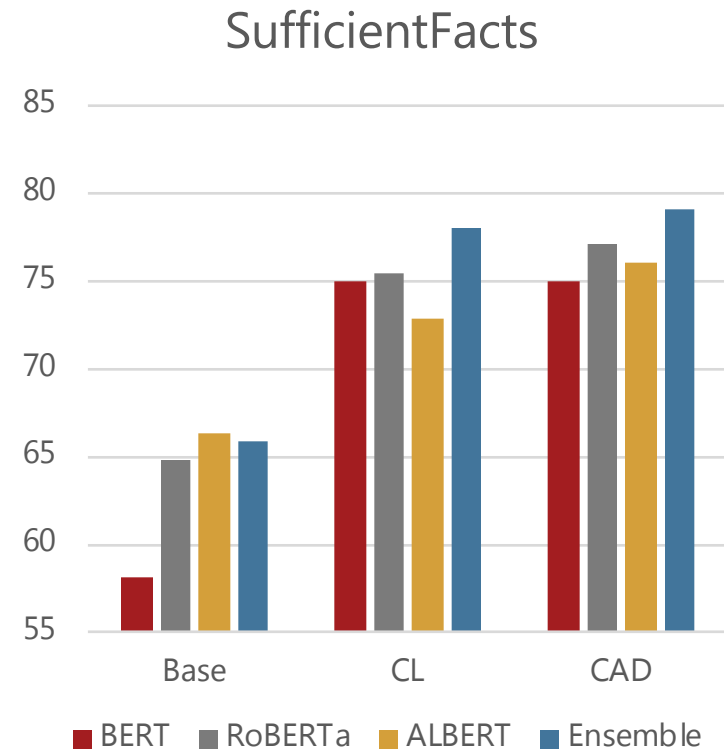
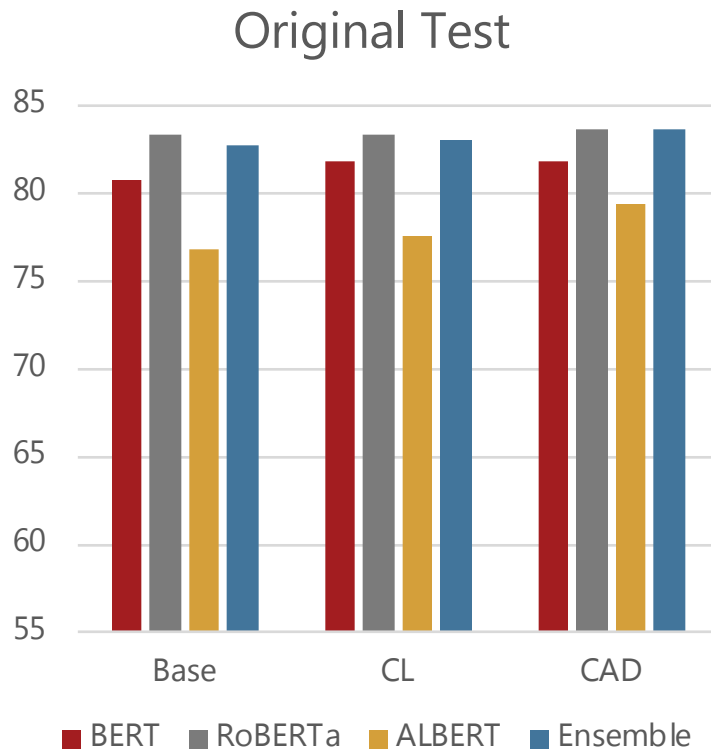


Contrastive instances: **negative ones**, where models agree that the remaining evidence is insufficient; **positive ones**, where a distractor sentence with high lexical overlap is added

Fact Checking with Evidence Omission Detection

- Contrastive learning (CL):
 - Auxiliary task:
 - bring the anchor and the positive instance closer in the representation space
 - drive the anchor and the negative instances further apart
- Counterfactual data augmentation (CAD):
 - Augment fact checking datasets with instances from Evidence Omission Detection dataset
 - Treat it as the same task

Selected Results for Fact Checking on HoVer



Summary of Findings

- Comprehensive study of FC with omitted information
 - Fluency-preserving omission methods
- FC models struggle most when adverbial modifiers are removed; least when date modifiers are removed
- Significant differences between Transformer architectures
 - ALBERT better at correctly predicting NEI
 - BERT and RoBERTa better at correctly predicting with EI

Summary of Findings

- SufficientFacts is challenging test bed for FC
 - Performance ~30 F1 points lower than on standard test set
- Training on additional data for evidence omission detection task improves performance on original test set and on SufficientFacts



Wrap-Up

Overall Take-Aways

- **Why** explainability?
 - understanding if a model is right for the right reasons
- Generated explanations can **help users understand**:
 - inner workings of a model (model understanding)
 - how a model arrived at a prediction (decision understanding)
- Explainability can **enable**:
 - transparency for end users
 - human-in-the-loop model development
 - human-in-the-loop data selection

Overall Take-Aways

- **Diagnostic Properties**

- Enable automatic evaluation of explanations
- Additional training objectives can improve the properties of the explanations and the general model performance

- **Generating Free-Text Explanations**

- Complex task; we propose first solutions, framing the task as summarisation
- How to generate fact checking explanations from evidence documents directly?
- How to collect a less noisy dataset of real-world fact checks?

Overall Take-Aways

- **Detecting Vulnerabilities of Fact Checking Models**
 - Fact checking models contain many vulnerabilities
 - Easily fooled with lexical trigger words
 - Easily tricked into making predictions without sufficient evidence
 - Training on adversarially generated instances improves robustness
 - Training on instances with omitted information improves robustness
 - How to generate explanations that are useful both for debugging and for the end user?

Thank you!

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Thanks to my PhD students and colleagues!



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Presented Papers

Pepa Atanasova, Jakob Grue Simonsen, Christina Lioma, **Isabelle Augenstein**. *A Diagnostic Study of Explainability Techniques for Text Classification*. In Proceedings of EMNLP 2020.

Pepa Atanasova, Jakob Grue Simonsen, Christina Lioma, **Isabelle Augenstein**. *Diagnostics-Guided Explanation Generation*. In Proceedings of AAAI 2022.

Pepa Atanasova, Jakob Grue Simonsen, Christina Lioma, **Isabelle Augenstein**. *Generating Fact Checking Explanations*. In Proceedings of ACL 2020.

Shailza Jolly, Pepa Atanasova, **Isabelle Augenstein**. *Generating Fluent Fact Checking Explanations with Unsupervised Post-Editing*. Preprint, abs/2112.06924, December 2021.

Pepa Atanasova, Jakob Grue Simonsen, Christina Lioma, **Isabelle Augenstein**. *Fact Checking with Insufficient Evidence*. TACL, Volume 10 (2022), to appear.

Other Recent Relevant Papers

Pepa Atanasova, Dustin Wright, **Isabelle Augenstein**. *Generating Label Cohesive and Well-Formed Adversarial Claims*. In Proceedings of EMNLP 2020.

Dustin Wright, David Wadden, Kyle Lo, Bailey Kuehl, **Isabelle Augenstein**, Lucy Lu Wang. *Generating Scientific Claims for Automatic Scientific Fact Checking*. In Proceedings of AAAI 2022.

Dustin Wright, **Isabelle Augenstein**. *Semi-Supervised Exaggeration Detection of Health Science Press Releases*. In Proceedings of EMNLP 2021.

Shailza Jolly, Pepa Atanasova, **Isabelle Augenstein**. *Time-Aware Evidence Ranking for Fact-Checking*. Journal of Web Semantics, Special Issue on Content Credibility, Volume 71, November 2021.

Wojciech Ostrowski, Arnav Arora, Pepa Atanasova, **Isabelle Augenstein**. *Multi-Hop Fact Checking of Political Claims*. In Proceedings of IJCAI 2021.