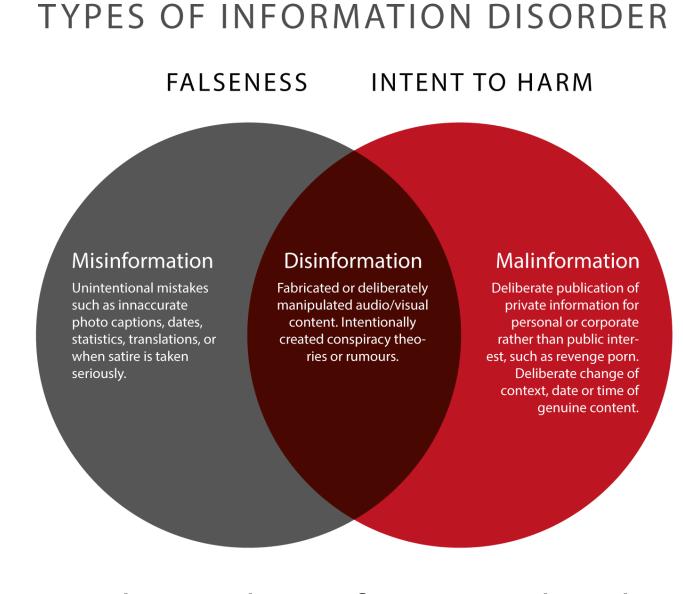


A Survey on Stance Detection for Mis- and Disinformation Identification Momchil Hardalov^{1,2} Arnav Arora^{1,3} Preslav Nakov^{1,4} Isabelle Augenstein^{1,3} ¹Checkstep Research ²Sofia University "St. Kliment Ohridski" ³University of Copenhagen ⁴Qatar Computing Research Institute, HBKU ☑ hardalov@fmi.uni-sofia.bg, arnav@checkstep.com,

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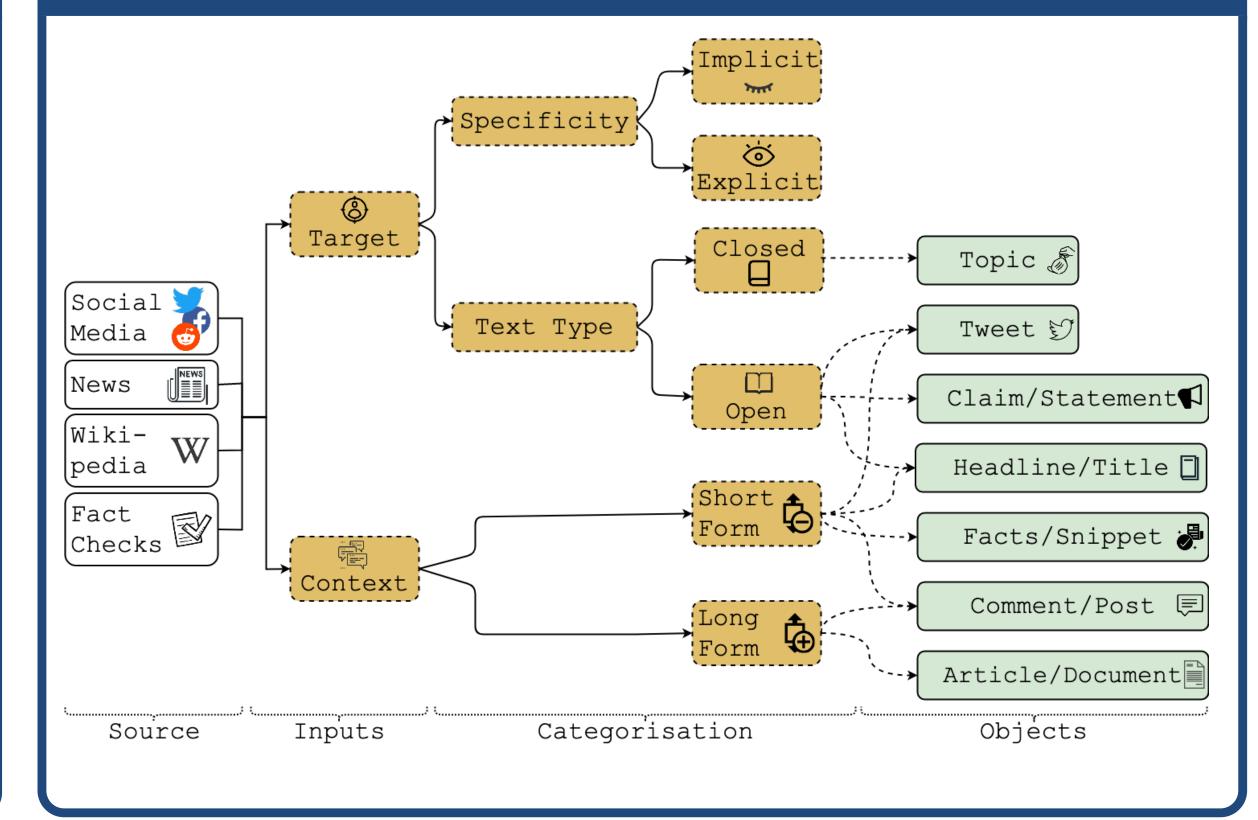


Stance Detection and Mis/Disinformation



Def. (Stance): "A public act by a social actor, achieved dialogically through overt communicative means, of simultaneously evaluating objects, positioning subjects (self and others), and aligning with other subjects, with respect to any salient dimension of the sociocultural field" Du Bois (2007)

Types of Stance Settings



Understanding information disorder Wardle (2020)

Def. (Stance Detection): "For an input in the form of a piece of text and a target pair, stance detection is a classification problem where the stance of the author of the text is sought in the form of a category label from this set: Favor, Against, Neither." Küçük and Can (2020)

Summary

What is included:

- Holistic overview of the role that different formulations of **stance detection** play in the **detection of** false content
- Settings for mis- and identifidisinformation cation to which stance detection has **successfully been** applied (datasets and task formulations)

Stance Detection Datasets for Mis- and Disinformation Detection						
<i>Sources</i> : 🎔 Twitter, 💷 News, Wikipedia, 🚭 Reddit.				<i>Evidence</i> : 🖹 Single, 🎞 Multiple, Q Thread.		
Dataset	Source(s)	Target	Context	Evidence	#Instances	Task
English Datasets						
Rumour Has It Qazvinian (2011)	Y	Topic	Tweet		10K	Rumours
PHEME Zubiaga (2016)	Y	Claim	Tweet	Q	4.5K	Rumours
Emergent Ferreira and Vlachos (2016)		Headline	Article		2.6K	Rumours
FNC-1 Pomerleau and Rao (2017)		Headline	Article		75K	Fake news
RumourEval '17 Derczynski (2017)	Y	Implicit	Tweet		7.1K	Rumours
FEVER Thorne (2018)	W	Claim	Facts		185K	Fact-checking
<i>Snopes</i> Hanselowski (2019)	Snopes	Claim	Snippets		19.5K	Fact-checking
RumourEval '19 Gorrell (2019)	🥩 🥑	Implicit	Post		8.5K	Rumours

- state-of-the-art • Current approaches, systems, and **applications**
- Lessons learned and important **future trends**

Out of scope:

- Surveying stance detection holistically, without a specific focus on veracity
- Other closely related NLP tasks, e.g., sentiment anal-**YSIS**

COVIDLies Hossain (2020) Claim Tweet W *TabFact* Wenhu et al. (2020) WikiTable Statement **Non-English Datasets** Arabic FC Baly et al. (2018) Claim Document DAST (Danish) Lillie (2019) 6 Submission Comment Croatian Bošnjak and Karan (2019) Title Comment ANS (Arabic) Khouja (2020) Claim Title Ara(bic)Stance Alhindi (2021) Claim Title

Fact-checking Rumour Claim verifiability Claim verification Claim verification

Misconceptions

Fact-checking

Lessons Learned and Future Trends

\mathbf{a}_{\mathbf{a}}^{\mathbf{a}} Integration:

Q

6.8K

118K

3K

3K

0.9K

3.8K

4K

We argue for a tighter integration between stance and fact-checking.

Dataset Size:

- A major limitation when training models
- The vast majority of datasets contain only a few thousand examples

Multilinguality:

- Only a handful of multilingual datasets

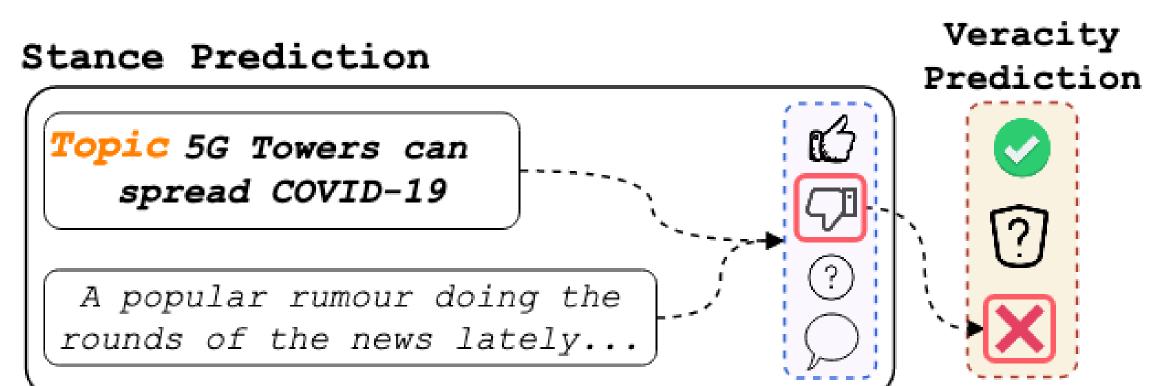
C Data Mixing:

- Can compensate for small dataset sizes
- Not trivial, task definitions and label inventories vary (Schiller, 2021; Hardalov, 2021)

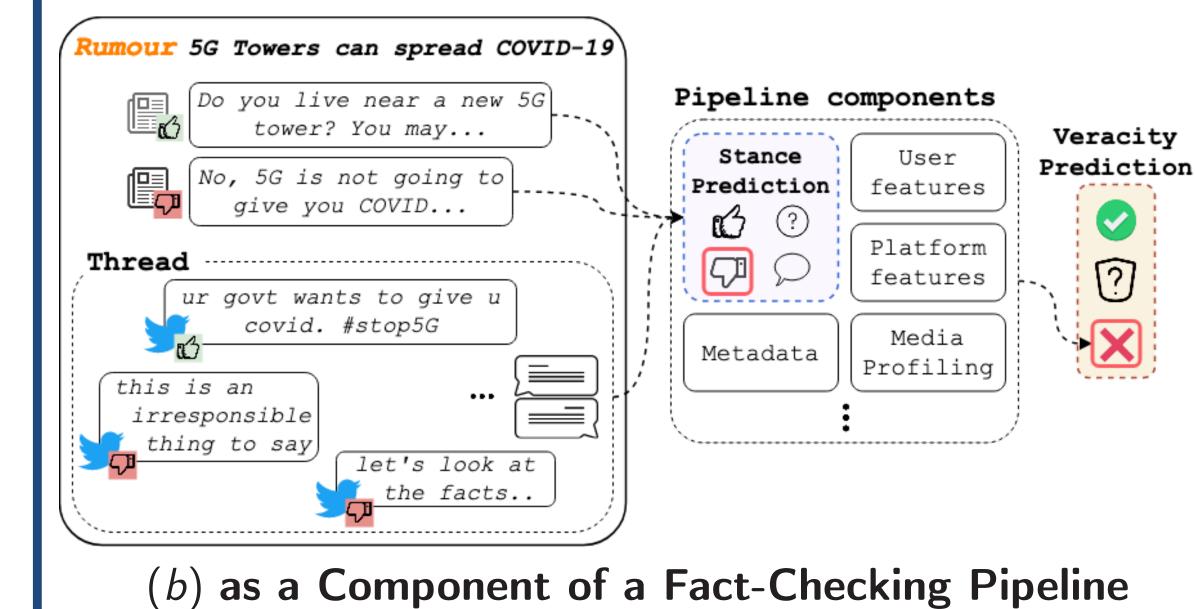
Multimodal Content:

– Spreading misdisinformation and through multiple modalities is becoming increasingly popular (e.g., deepfakes, *memetic warfare*) - The wisdom of the crowd in social media can be an additional information source

Stance Detection Formulations



(a) as Fact-Checking



Small in size and do not offer cross-lingual settings

exception: Vamvas and Sennrich (2020)

- Cultural norms play a crucial role

Shades of Truth:

- Missing notion of stance detection, but fact-checking goes beyond true/false
- Fine-grained labels are common for the related task of Sentiment Analysis

Explainability:

- Crucial step towards adopting fully automated fact-checking
- There is a need for holistic and realistic explanations of how a fact-checking model arrived at its prediction

Modelling the Context:

- The background of the stance-taker, e.g., previous activity, network, interests
- Characteristics of the target, e.g., funding, previously known biases, credibility

Label Semantics:

- With time, the definition of stance and the label inventories have evolved
- The labels can share semantic similarities, but there can be mismatches in the label definitions (Hardalov, 2021; Momchil et al., 2022)