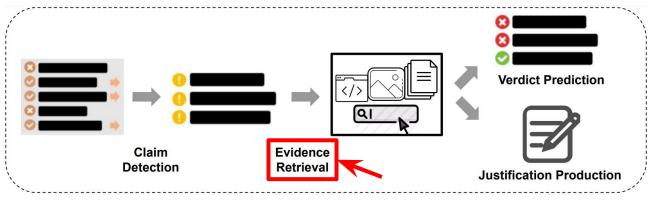
## When You Doubt, Abstain: A Study of Automated Fact-checking in Italian Under Domain Shift

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### Introduction

- Countering the **spread of misinformation**
- Automated tools



Guo et al., 2022. A survey on automated fact-checking. Transactions of the Association for Computational Linguistics.

- Ambiguity of claims rarely tackled >> Abstention?
- Italian language overlooked

#### **Fact-checking Data**

Italian portion of **X-Fact** (*Gupta and Srikumar, 2021*)

#### Sources:

- Pagella Politica (PP)
- Agenzia Giornalistica Italia (AGI)

#### Splits:

- Training set (943) PP
- Development set (125) PP
- Test set
  - *In-domain* (190) PP
  - Out-of-domain (160) AGI

Gupta and Srikumar, 2021. X-Fact: A New Benchmark Dataset for Multilingual Fact Checking. *Proceedings of the 59th Annual Meeting of the Association for Computational Linguistics*.

## **Claim Ambiguity**

Based on a preliminary assessment of the test portions of X-Fact

## Missing information

The claim **does not contain information** that <mark>calls for verification:</mark>

**Example from X-Fact** 

it. Di Battista e la guerra in Afghanistan.

en. Di Battista and the war in Afghanistan.

mostly-true

# Lack of context

The claim **does not provide enough context** (e.g., *who*, *when*, and *where*) or contains **underspecified language**, ill-defined terms and pronouns:

**Example from X-Fact** 

it. Siamo al nono mese consecutivo di riduzione degli sbarchi.

en. We are in the ninth consecutive month of reduced arrivals by sea.

<u>true</u>

## Discordant label

The statement has been **reported in a negated form** or as its opposite, but the label reflects the veracity of the original statement:

**Example from X-Fact** 

it. No, la Banca d'Italia non è controllata dalle banche private.

en. No, the Bank of Italy is not controlled by private banks.

#### partly-true

# Claim as question

The fact-checked statement has been **reported as a question**:

**Example from X-Fact** 

it. Davvero la triplice sede del Parlamento europeo costa oltre 200 milioni di euro l'anno?

en. Does the triple seat of the European Parliament really cost over 200 million euros per year?

partly-true

## **Claim Ambiguity**

Ambiguity class	PP test		AGI test		
	in-domain		out-c	of-domain	
Missing information	47	24.7%	6	3.8%	
Lack of context	13	6.8%	17	10.6%	
Discordant label	13	6.8%	0	0.0%	
Claim as question	31	16.3%	0	0.0%	
No ambiguity	86	45.3%	137	85.6%	
Total	190	100.0%	160	100.0%	

We focus on the highlighted subsets

### **Creation of Challenge Test Sets**

Goal: assess the performance of automated fact-checking under genre shift

Rewritten test sets in two different styles (for both we have 117 *in-domain* claims and 137 *out-of-domain* claims):

**News-like** (newspaper headline)

"Il M5S si conferma una delle principali forze politiche in Europa, secondo Di Maio."

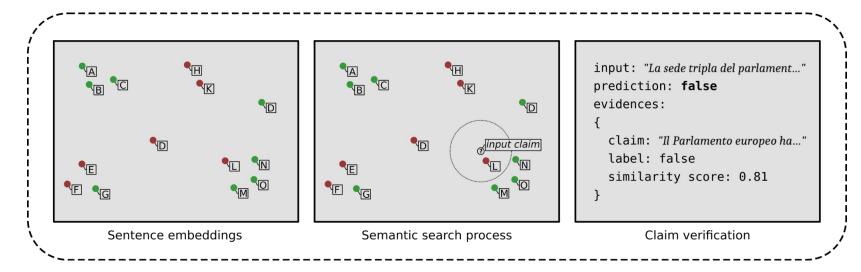
*The M5S is confirmed as one of the main political forces in Europe, according to Di Maio.* 

**Social-like** (social media post)

"Il #M5S è traa i partiti maggiori d'Europa!!!"

*The #M5S is amongg the major parties in Europe!!!* 





Semantic search method for evidence retrieval based on SentenceTransformers<sup>1</sup>

Hyperparameter  $\tau$  (cosine similarity threshold for abstention)

<sup>&</sup>lt;sup>1</sup> paraphrase-multilingual-MiniLM-L12-v2.



Two axes of variation:

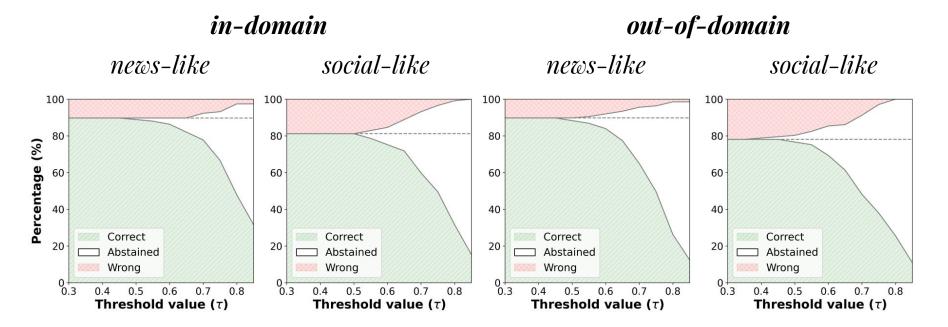
- **Source**: in-domain (ID) vs out-of-domain (OOD)
- Genre: news-like (NL) vs social-like (SL)

Two setups:

- **Controlled** (relevant information present in the evidence set)
- Non-controlled (relevant information may not be present in the evidence set)



#### **Controlled setups**



<mark>Results</mark>								
GENRE SHIFT		macro $F_1$ score			CO	correct (cor)		
			ID	OOD		ID	OOD	
Genre shift has a <mark>large impact</mark> on performance		$\mathbf{NL}$	0.86	0.82	$\mathbf{NL}$	0.86	0.84	
		$\bullet$ SL	0.74	0.67	$\operatorname{SL}$	0.75	0.69	
macro F1 score		abs	tention	ı (abs)	e	error (e	err)	
• 1			ID	OOD		ID	OOD	
in-domain:	0.86 > 0.74 (-0.12)	NL	0.03	0.08	$\mathbf{NL}$	0.10	0.08	
out-of-domain:	0.82 💊 0.67 (-0.15)	$\operatorname{SL}$	0.09	0.16	$\operatorname{SL}$	0.15	0.15	



Fact-checking <mark>sources do matter</mark>, too

macro F <sub>1</sub> score						
	ID —	►OOD				
$\mathbf{NL}$	0.86	0.82				
$\mathbf{SL}$	0.74	0.67				

correct (cor)						
	ID	OOD				
NL	0.86	0.84				
$\operatorname{SL}$	0.75	0.69				

	macro F1 score
news-like:	0.86 <b>\</b> 0.82 ( <b>-0.04</b> )
social-like:	0.74 <b>\</b> 0.67 ( <b>-0.07</b> )

abst	tention	(abs)	
	ID	OOD	
$\rm NL$	0.03	0.08	
$\operatorname{SL}$	0.09	0.16	

error (err)					
	ID	OOD			
NL	0.10	0.08			
$\operatorname{SL}$	0.15	0.15			

ABSTENTION $ma \leftarrow rer F_1 \\ D \\ OOD$ score $correct (correct ($	<mark>Results</mark>							
Abstention helps in reducing errors       NL $0.86$ $0.82$ NL $0.86$ $0.84$ SL $0.74$ $0.67$ SL $0.75$ $0.69$ news-like: $0.10 \ 0.08$ (-0.02) $D_{1D}$ $D_{00D}$ $D_{1D}$ $D_{00D}$ NL $0.03$ $0.08$ $D_{1D}$ $D_{00D}$ $D_{1D}$ $D_{00D}$	ABSTENTION		macro F <sub>1</sub> score			correct (cor)		
Abstention neuros in reducing       SL $0.74$ $0.67$ SL $0.75$ $0.69$ error rate $ocod$ $abstention$ $(abs)$ $ocod$ $vertor (err)$ $ID$ $ocod$ $ID$ $ocod$ $vertor (err)$ $ID$ $ocod$ $Ocod$ $ID$ $ocod$ $Ocod$ $ID$ $ocod$				ID	OOD		ID	OOD
errors       SL       0.74       0.67       SL       0.75       0.69         news-like:       0.10 \ 0.08 (-0.02)       abs:-ntion       (abs)       (abs) </th <th>Abstention <mark>he</mark>l</th> <th>lps in reducing</th> <th>NL</th> <th>0.86</th> <th>0.82</th> <th><math>\mathbf{NL}</math></th> <th>0.86</th> <th>0.84</th>	Abstention <mark>he</mark> l	lps in reducing	NL	0.86	0.82	$\mathbf{NL}$	0.86	0.84
news-like: $0.10 \ge 0.08 (-0.02)$ ID $0.03$ $0.08$ NL $0.10$ $0.08$			$\operatorname{SL}$	0.74	0.67	$\operatorname{SL}$	0.75	0.69
news-like: $0.10 \ge 0.08 (-0.02)$ ID $0.03$ $0.08$ NL $0.10$ $0.08$			aha	tontion	(aba)		rror (a	)
<i>news-like:</i> 0.10 \ 0.08 (-0.02) NL 0.03 0.08 NL 0.10 0.08		error rate	abs		· /			,
NL 0.03 0.08 NL 0.10 0.08	neros_like.	0.10 > 0.08 (-0.02)		ID	OOD		ID	OOD
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	social-like:	$0.15 \rightarrow 0.15$	$\operatorname{SL}$	0.09	0.16	$\operatorname{SL}$	0.15	0.15
						n		



Causes of errors:

- 33.0% **correct evidence discarded** (wrong evidence with higher similarity)
- 22.9% correct evidence not found  $\rightarrow$  abstention
- 44.0% correct evidence not found  $\rightarrow$  wrong label



### **Conclusion**

- Genres and sources have a large impact on performance
- Abstention to cope with lack of sufficient evidence
- Semantic similarity for transparent fact-checking

Future work:

- Automating claim ambiguity categorization
- Assessing system efficacy with intended users



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*https://github.com/jo-valer/fact-checking-ita-abstention*