

dfkinit2b at CheckThat! 2025: Leveraging LLMs and Ensemble of Methods for Multilingual Claim Normalization

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Chikkala, Natalia Skachkova, Jing Yang, Veronika
Solopova, Vera Schmitt, Simon Ostermann

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CLEF CheckThat! 2025, 11.09.2025

ivan.vykopal@kinit.sk



Motivation

- Misinformation is spreading rapidly on social media

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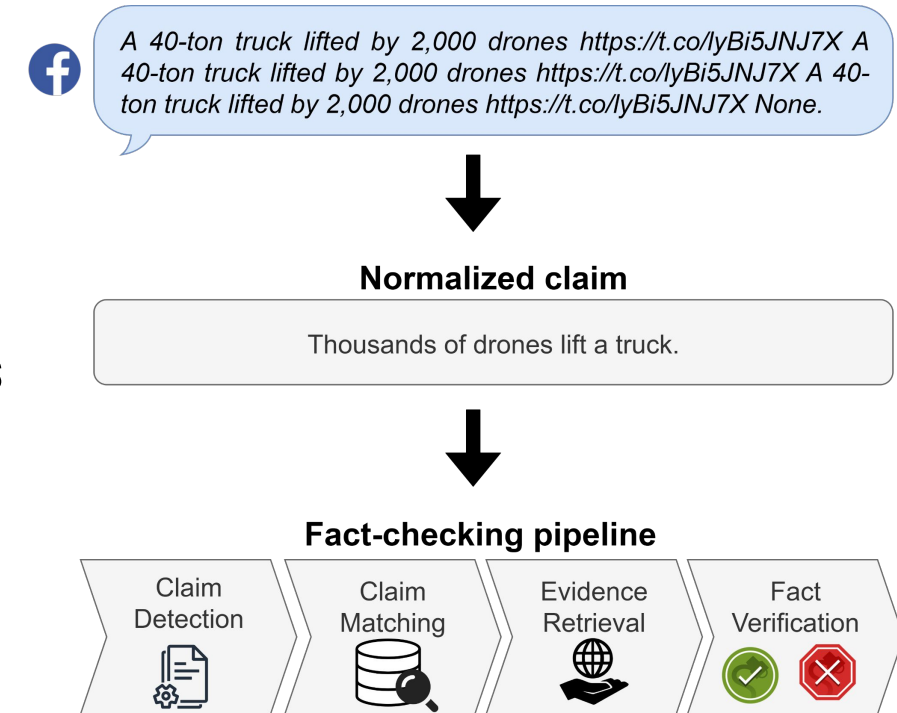
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- **Posts are noisy, informal and often ambiguous**



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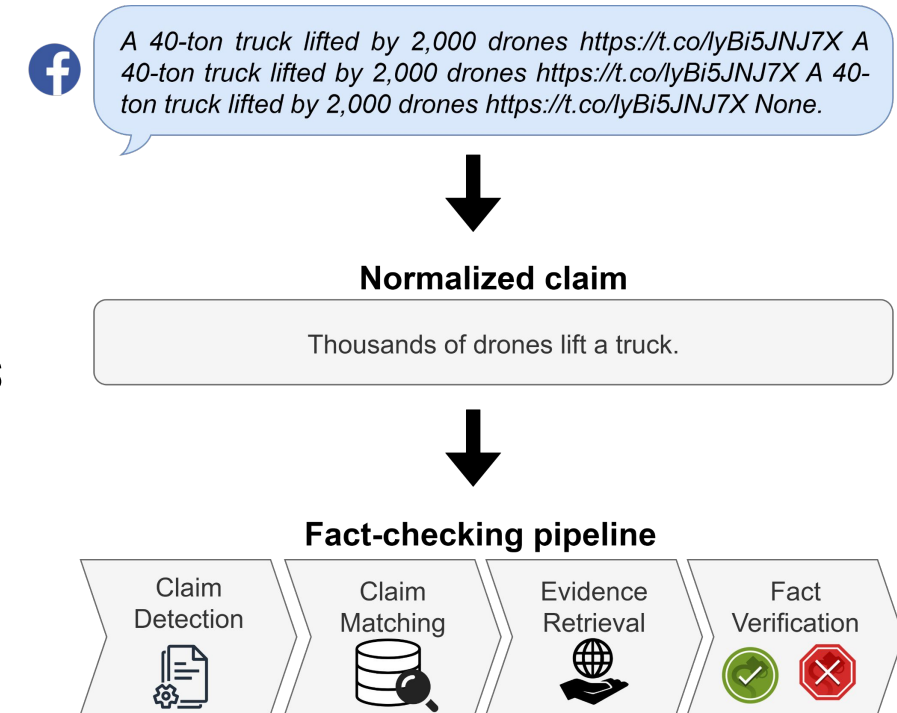
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- Misinformation is spreading rapidly on social media
- **Posts are noisy, informal and often ambiguous**
- Need for ***Claim Normalization*** to support automated fact-checking
- Multilingual dimension adds complexity



CheckThat! 2025 Task 2 Overview

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Two settings:

- Monolingual: 13 languages with full datasets (train, dev, test)


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Monolingual



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Two settings:

- Monolingual: 13 languages with full datasets (train, dev, test)
- Zero-shot: 7 unseen languages, test set only

Normalized claim

Thousands of drones lift a truck.

Monolingual



Zero-shot



Data & Pre-Processing

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Challenges

Data & Pre-Processing

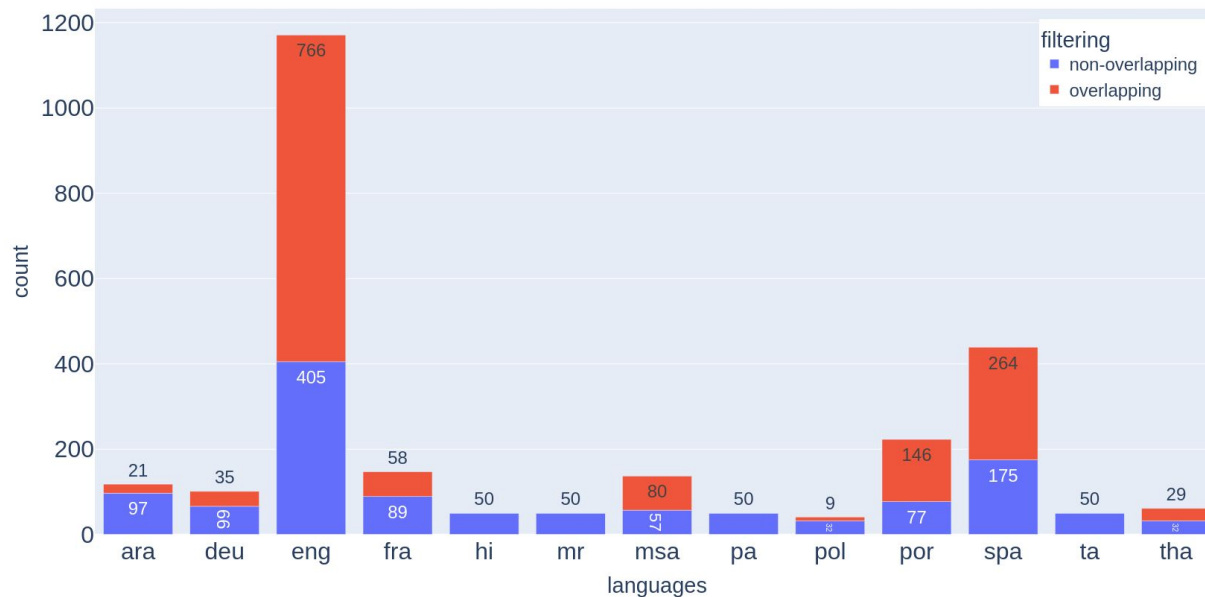
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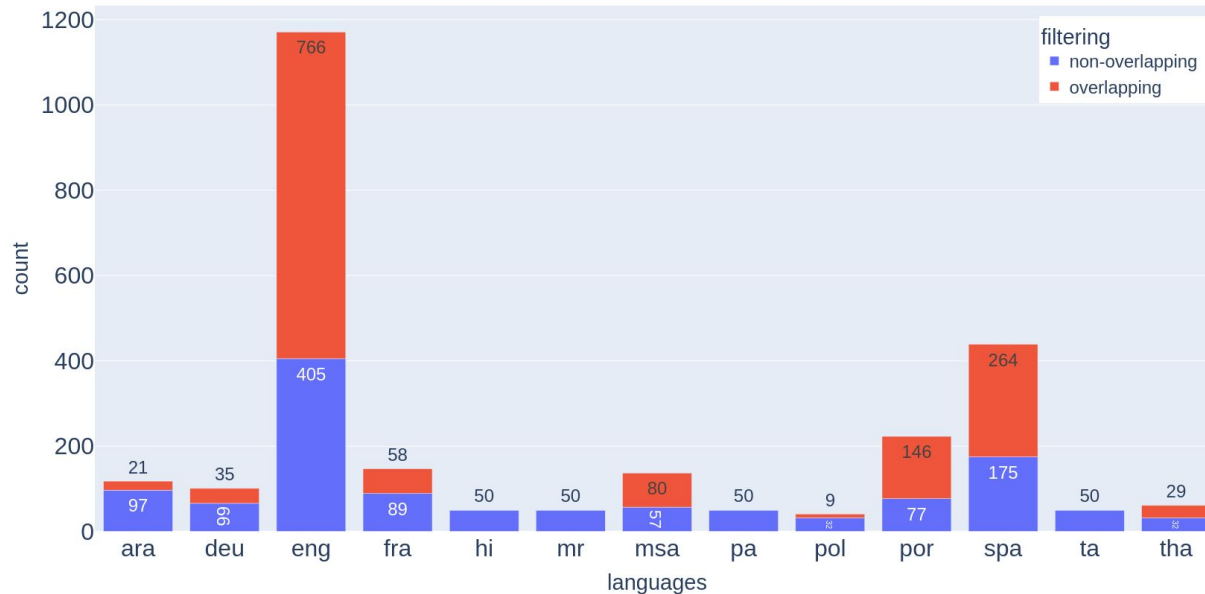
Claim overlap between the gold train and development data.

Challenges

- Mixed languages in post/claim
- Duplicates across train/dev

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Claim overlap between the gold train and development data.

Pre-processing: de-duplication & similarity filtering (discarded if $\cos. \text{sim} < 0.05$)

Challenges

- Mixed languages in post/claim
- Duplicates across train/dev

Models & Approaches

- 9 LLMs

Model	# Params	# Langs	Citation	LoRA	Prompting
Llama3.1 Instruct	405 B	8	Grattafiori et al. [27]		✓
Llama3.1 Nemotron Ultra	253 B	8	Bercovich et al. [28]		✓
Qwen2.5 Instruct	72 B	29	Yang et al. [29]		✓
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Gemma3 IT	27 B	140+	Team et al. [31]	✓	✓
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Models & Approaches

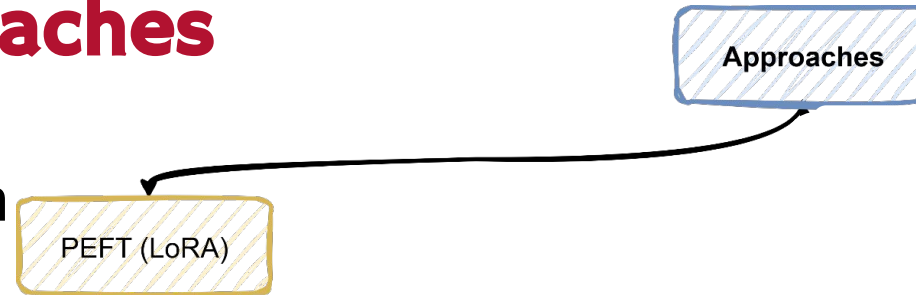


- 9 LLMs and 3 main approaches

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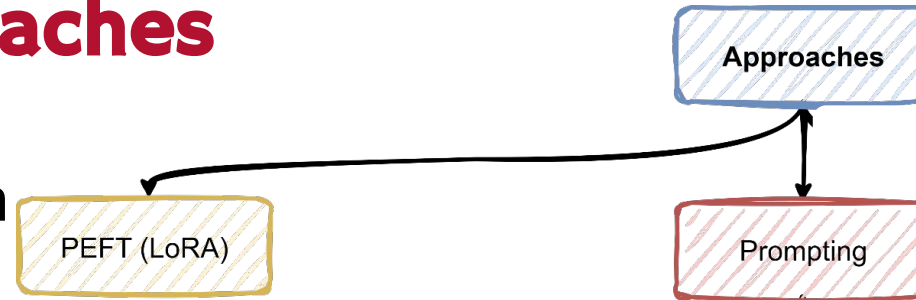
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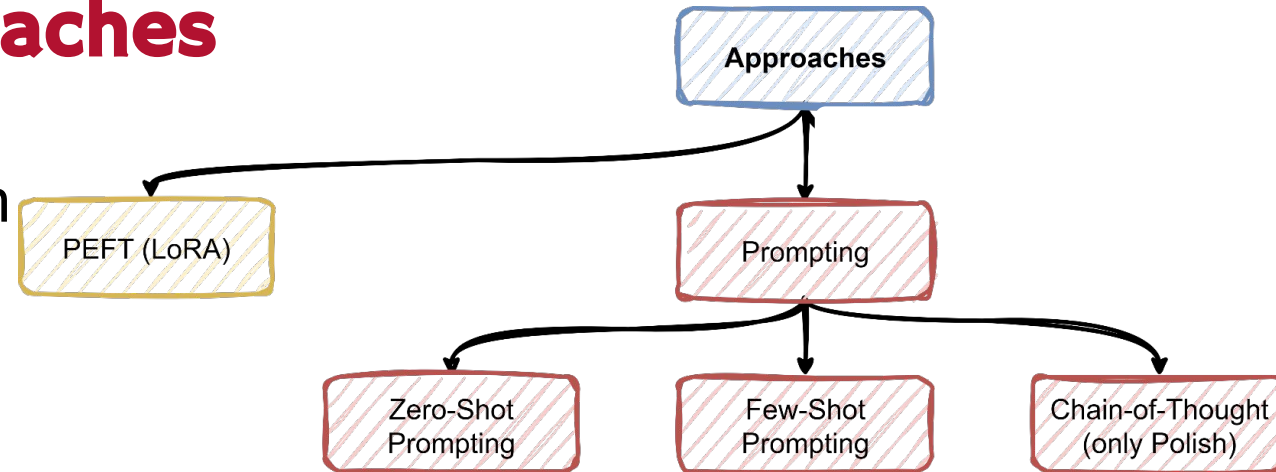
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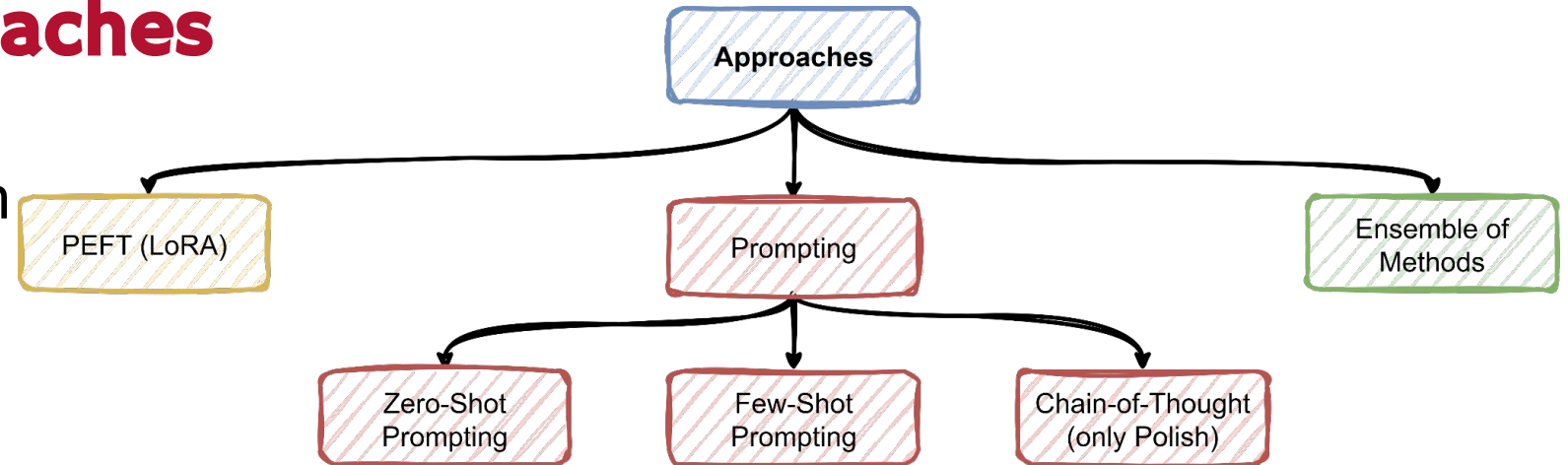
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 - Translation-based augmentation (translation from English data)

Prompting Experiments

Zero-Shot Prompting

- Task description + characteristics of normalized claims

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Few-Shot Prompting

- Demonstration selection via similarity (mGTE-base)
- Experimented with 1, 2, 5 and 10 examples

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Translated Prompts (Zero- & Few-Shot)

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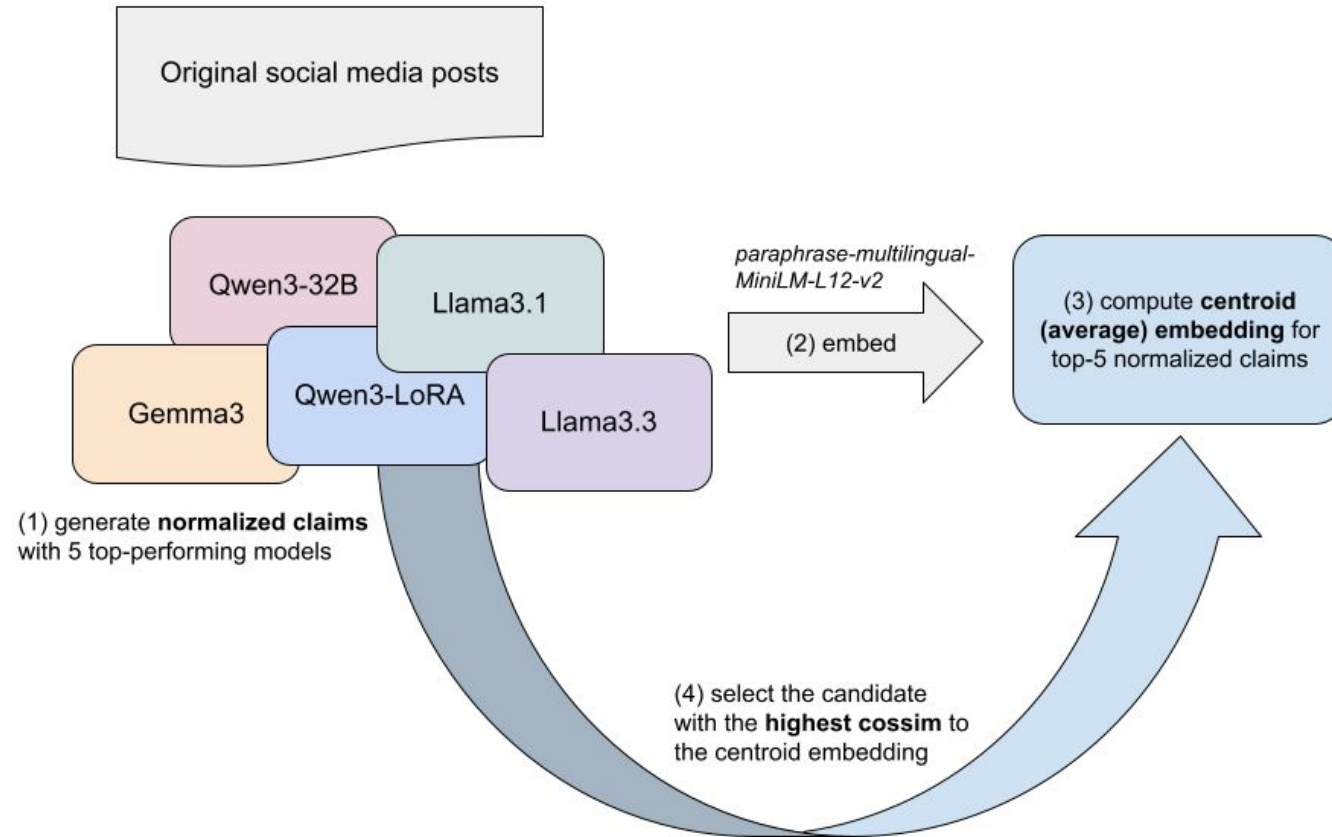
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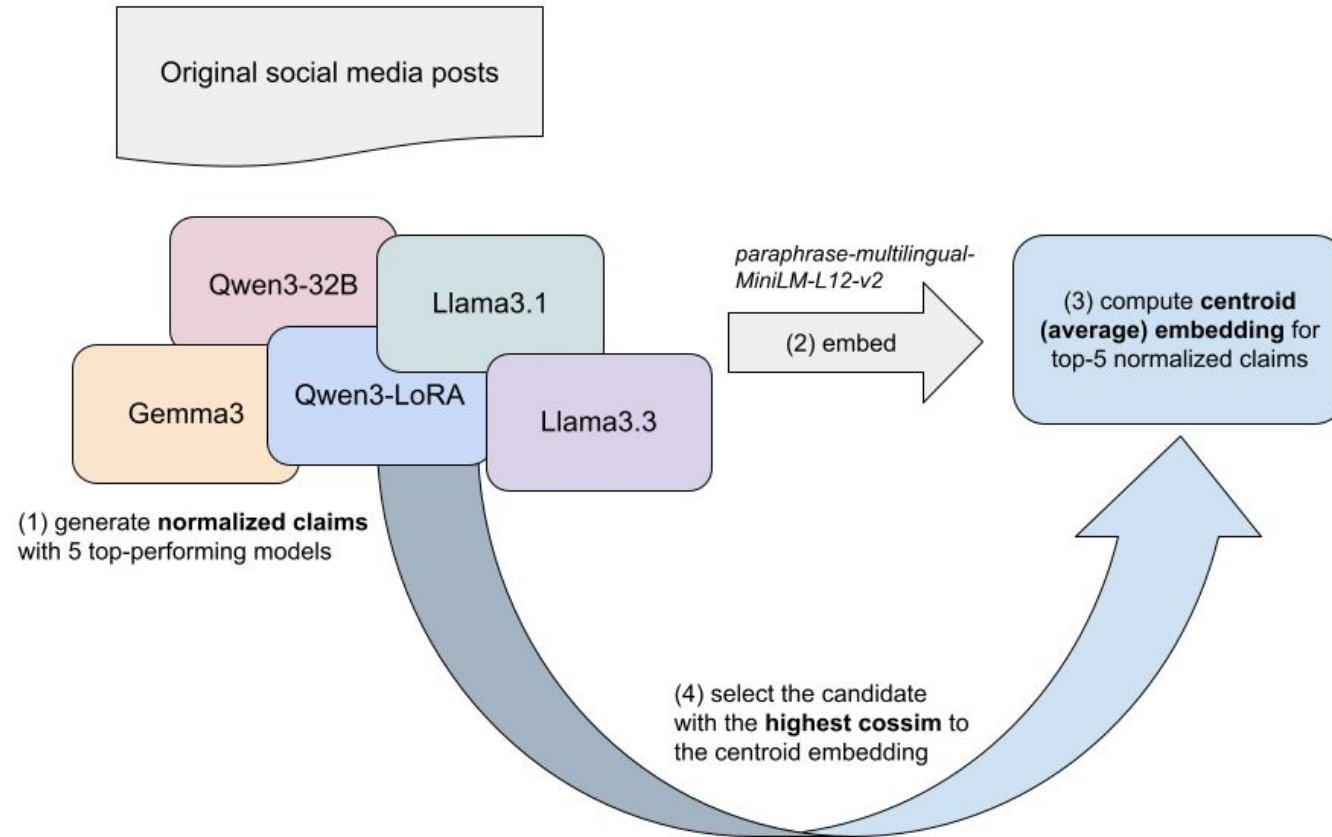
Experiments with Polish

- *Polish-CoT* vs. *Few-shot prompting*
- Experiments also with the Polish LLM

Ensemble of Methods



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Monolingual Setting

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Fine-tuning results for Gemma3 27B evaluated on the official development set. The first row shows the zero-shot prompting results for comparison.

Approach	ara	deu	eng	fra	hi	mr	msa	pa	pol	por	spa	ta	tha
Zero-shot	0.305	0.161	0.244	0.265	0.224	0.275	0.219	0.311	0.194	0.294	0.268	0.340	0.054
LoRA-target	0.361	0.298	0.658	0.439	0.290	0.311	0.599	0.352	0.267	0.509	0.518	0.450	0.217
LoRA-all-balanced	0.390	0.293	N/A	0.454	0.285	0.287	0.570	0.309	0.265	0.510	0.531	0.438	0.213
LoRA-translated	0.379	0.302	N/A	0.430	0.254	0.297	0.551	0.290	0.236	0.497	0.509	0.332	0.175
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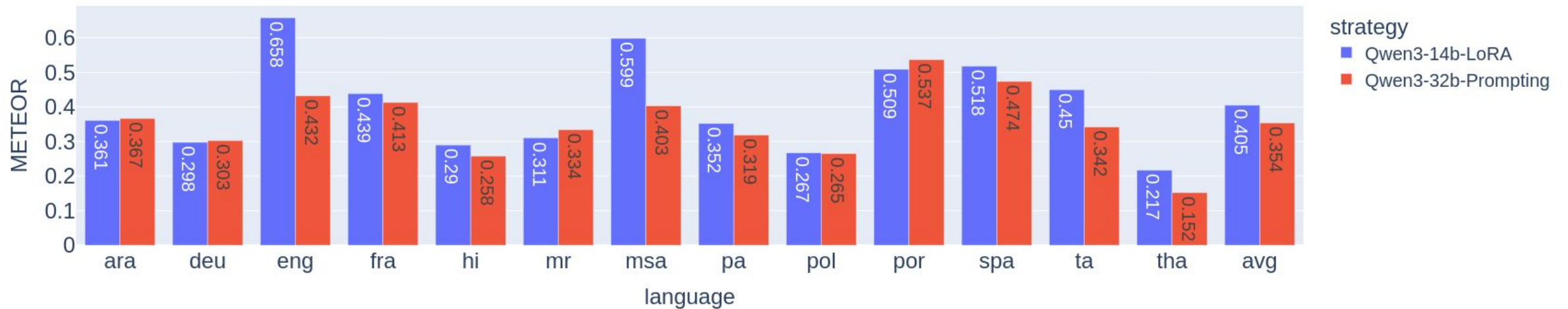
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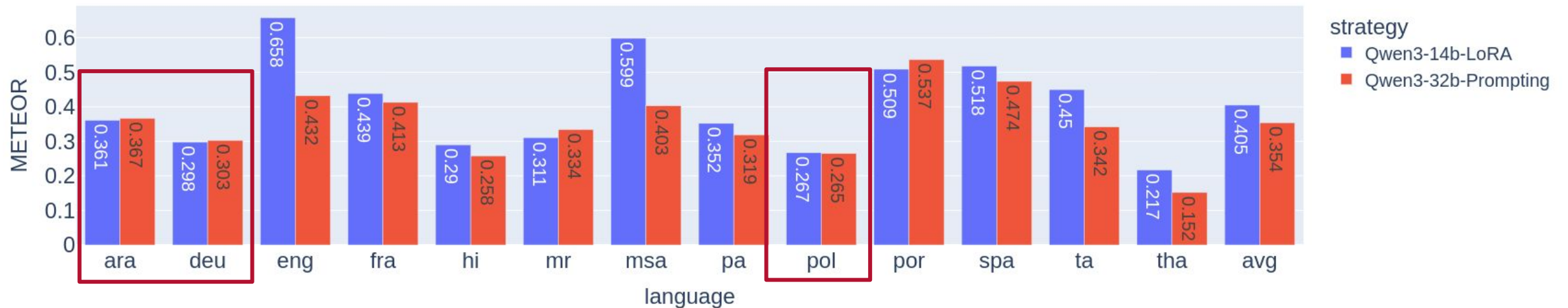
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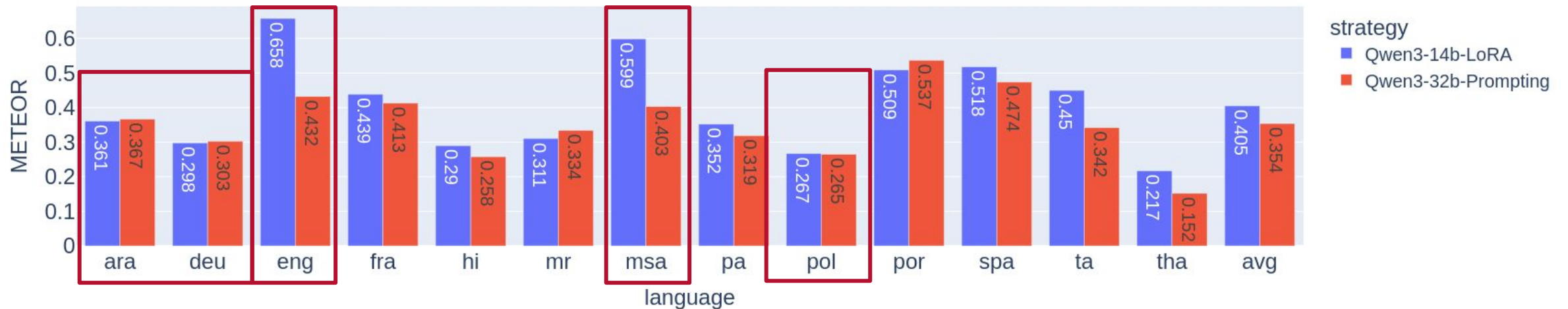
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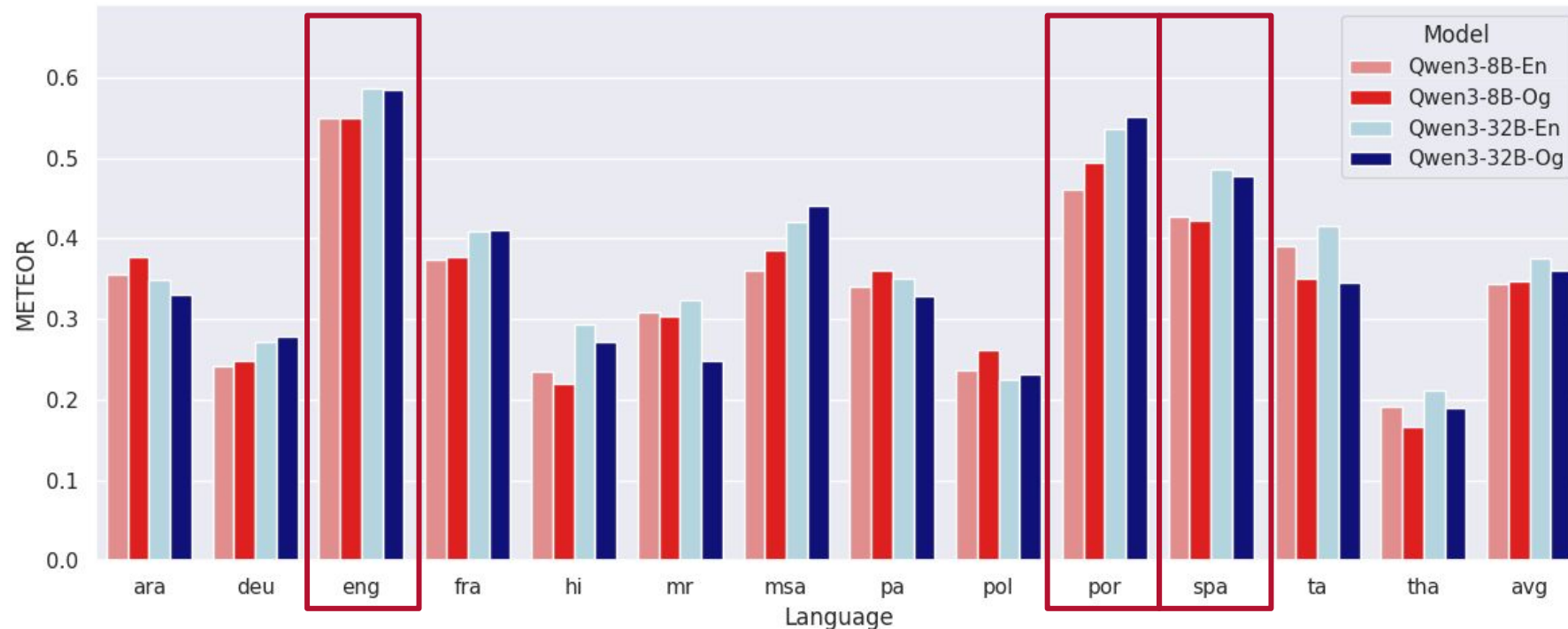
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LoRA offers negligible gains for some languages but larger improvements for others



High-resource European languages demonstrated consistently strong performance



Performance of Qwen3 in the monolingual setting on the dev set using 10-shot prompting with unfiltered data for sample selection.

Multilingual models outperform the Polish LLM

Model	Prompt Type	Dev Set	Test Set
Bielik Instuct v2.3	Polish-CoT	0.198	N/A
	3-shot	0.282	0.297
Llama3.1 Nemotron Ultra	3-shot	0.254	N/A
	10-shot	0.296	0.347
Llama3.1 405B	10-shot	0.271	0.393
	20-shot	N/A	0.396



Overall Results

We ranked first on all 7 zero-shot languages and on 6 out of 13 languages in the monolingual setting

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Language	Arabic (ara)	German (deu)	English (eng)	French (fra)	Hindi (hi)	Marathi (mr)	Indonesian (msa)	Punjabi (pa)	Polish (pol)	Portuguese (por)
Best Score	0.504	0.386	0.457	0.527	0.328	0.389	0.565	0.331	0.407	0.577
Our Score	0.504	0.347	0.457	0.470	0.328	0.389	0.502	0.331	0.396	0.574
Δ (Ours vs Best)	0	-0.039	0	-0.057	0	0	-0.063	0	-0.011	-0.003
Our Strategy	Ensemble	Qwen3-32b	Ensemble	Qwen3 _{LoRA}	Ensemble	Qwen3 _{LoRA}	Qwen3 _{LoRA}	Qwen3-8b	Llama3.1	Ensemble
Our Rank	1	2	1	2	1	1	2	1	2	2

Language	Spanish (spa)	Tamil (ta)	Thai (tha)	Bengali _{zero} (ben)	Czech _{zero} (ces)	Greek _{zero} (ell)	Korean _{zero} (kor)	Dutch _{zero} (ndl)	Romanian _{zero} (ron)	Telugu _{zero} (te)
Best Score	0.608	0.632	0.586	0.378	0.252	0.262	0.134	0.200	0.295	0.526
Our Score	0.554	0.632	0.300	0.378	0.252	0.262	0.134	0.200	0.295	0.526
Δ (Ours vs Best)	-0.054	0	-0.286	0	0	0	0	0	0	0
Our Strategy	Ensemble	Qwen3 _{LoRA}	Ensemble	Ensemble	Gemma3	Ensemble	Gemma3	Ensemble	Ensemble	Ensemble
Our Rank	2	1	3	1	1	1	1	1	1	1

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Our Rank	2	1	3	1	1	1	1	1	1	1

We ranked first on all 7 zero-shot languages and on 6 out of 13 languages in the monolingual setting

Language	Arabic (ara)	German (deu)	English (eng)	French (fra)	Hindi (hi)	Marathi (mr)	Indonesian (msa)	Punjabi (pa)	Polish (pol)	Portuguese (por)
Best Score	0.504	0.386	0.457	0.527	0.328	0.389	0.565	0.331	0.407	0.577
Our Score	0.504	0.347	0.457	0.470	0.328	0.389	0.502	0.331	0.396	0.574
Δ (Ours vs Best)	0	-0.039	0	-0.057	0	0	-0.063	0	-0.011	-0.003
Our Strategy	Ensemble	Qwen3-32b	Ensemble	Qwen3 _{LoRA}	Ensemble	Qwen3 _{LoRA}	Qwen3 _{LoRA}	Qwen3-8b	Llama3.1	Ensemble
Our Rank	1	2	1	2	1	1	2	1	2	2

Language	Spanish (spa)	Tamil (ta)	Thai (tha)	Bengali _{zero} (ben)	Czech _{zero} (ces)	Greek _{zero} (ell)	Korean _{zero} (kor)	Dutch _{zero} (ndl)	Romanian _{zero} (ron)	Telugu _{zero} (te)
Best Score	0.608	0.632	0.586	0.378	0.252	0.262	0.134	0.200	0.295	0.526
Our Score	0.554	0.632	0.300	0.378	0.252	0.262	0.134	0.200	0.295	0.526
Δ (Ours vs Best)	-0.054	0	-0.286	0	0	0	0	0	0	0
Our Strategy	Ensemble	Qwen3 _{LoRA}	Ensemble	Ensemble	Gemma3	Ensemble	Gemma3	Ensemble	Ensemble	Ensemble
Our Rank	2	1	3	1	1	1	1	1	1	1

Best strategies – Gemma3 prompting & Ensemble method

Key Findings & Conclusion

- LoRA adapters are efficient for low-resource languages

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- LoRA adapters are efficient for low-resource languages
- Prompting works well with high-resource languages
- English prompts outperform target language in zero-shot
- Ensemble smooths out model variance and achieves the best performance in 11 out of 20 languages (*5 out 7 cases in the zero-shot setting*)

Current Challenges

- Dataset imbalance and overlap between splits

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- Missing context (especially visual, e.g., images or videos)

Future Directions

- Use LLMs for data augmentation

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- Use LLMs for data augmentation
- Evaluate the impact of normalized claims in fact-checking tasks

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