

# SemEval 2025 Shared Task 7: Multilingual and Crosslingual Fact-Checked Claim Retrieval

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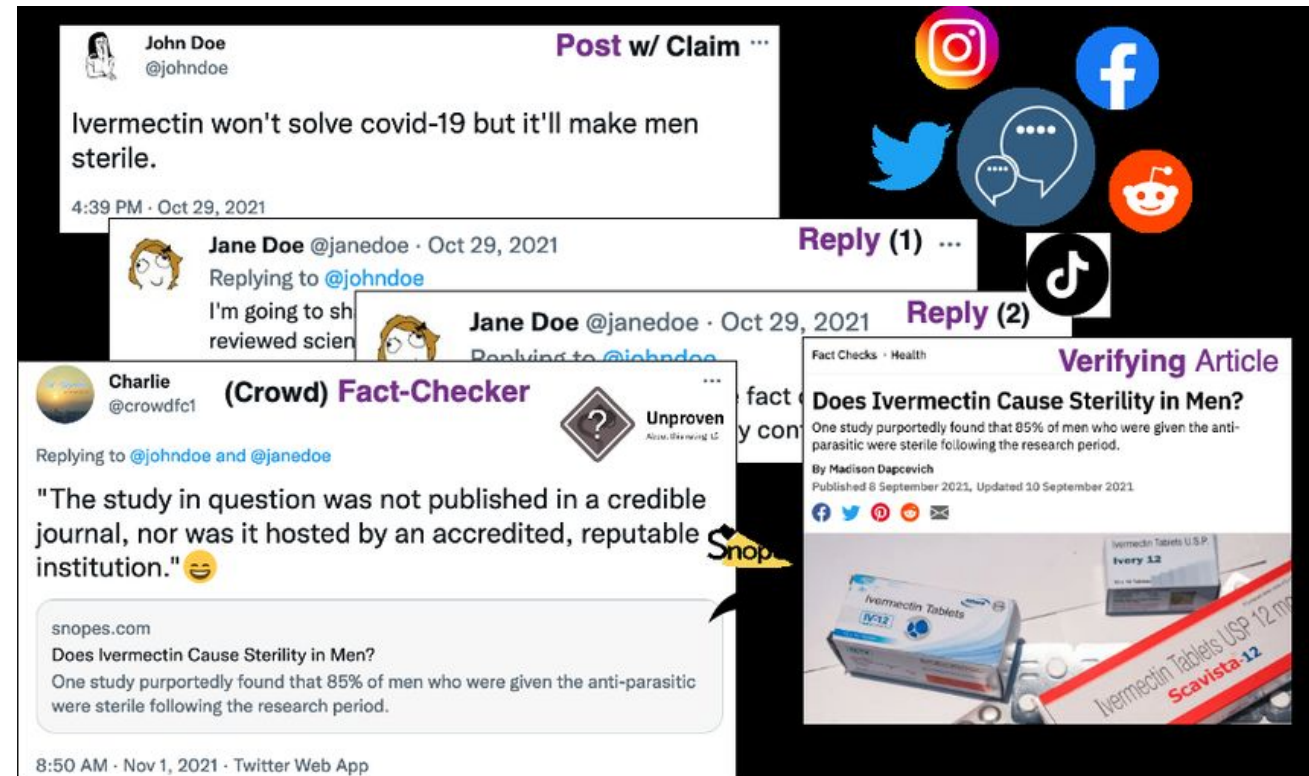
# Main Goal

Efficiently identifying previously fact-checked claims across multiple languages.

The media has not pixelated the Mercadona logo on the truck that appears in this video of a rescue in Valencia

A video is circulating on social media showing the rescue of a Mercadona truck that was trapped in a flood caused by the passage the Dana in Spain and whose logo was allegedly pixelated by the media.

- However, the Valencia Provincial Firefighters Consortium has told Newtral.es that it was they who distributed the images with the brand blurred for reasons of privacy of the victims.



# Dataset: Example

	8626
:-----:	
ocr	odas Don't talk on the phone while on charge SUL
verdicts	Missing context
text	Do not talk on the phone, while it is charging!!
lan	eng
fb	1
tw	0
ig	0
full_text	odas Don't talk on the phone while on charge SUL Do not talk on the phone, while it is charging!!
gs	[138780]

	138780
:-----:	
claim	Using a mobile phone while it charges is dangerous
instances	[' <a href="https://factcheck.afp.com/misleading-video-risks-charging-phone-spreads-worldwide#4af86ceff67a6d6e81b54bf60a3bb0c6">https://factcheck.afp.com/misleading-video-risks-charging-phone-spreads-worldwide#4af86ceff67a6d6e81b54bf60a3bb0c6</a> ']
title	Misleading video on risks from charging phone spreads worldwide
full_text	Misleading video on risks from charging phone spreads worldwide Using a mobile phone while it charges is dangerous

Link

# Dataset: Example

## Misleading video on risks from charging phone spreads worldwide

This article is more than one year old. | Published on March 4, 2021 at 19:22 | ⌚ 3 min read |

By [Alex CADIER, AFP USA](#)

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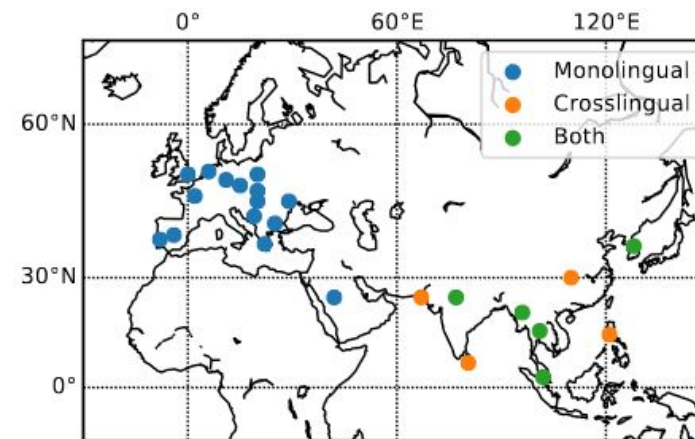
**A video shared hundreds of thousands of times warns against using mobile phones while they charge because “electric waves” will pass through your body. This advice is misleading, according to experts who say other appliances would also exhibit similar harmless voltage leakage, while the real danger -- though slim -- could come from faulty batteries and low-quality chargers.**

“Don’t talk on the phone while on charge” warns the video, shared more than 50,000 times on [Facebook](#) and more than 600,000 times on [TikTok](#). It has spread in the [United States](#), [Canada](#), [Nigeria](#), and [the Philippines](#).

# Dataset: MultiClaim

The **MultiClaim dataset** was generated by collecting:

1. **Fact-checks:** A total of **205,751 fact-checks from 142 organizations in 39 languages**, including claims, titles, publication dates, and URLs.
2. **Social Media Posts:** **28,092 posts in 27 languages** from platforms like Facebook, Instagram, and Twitter, selected and linked to relevant fact-checks by professional fact-checkers.
3. **Pairs:** **31,305 post-fact-check pairs** (26,774 monolingual and 4,212 crosslingual), created based on professional fact-checker assignments.





# Dataset: Provided Splits

- 153,743 fact checks
- 24,431 posts
- 25,743 post-fact-check pairs

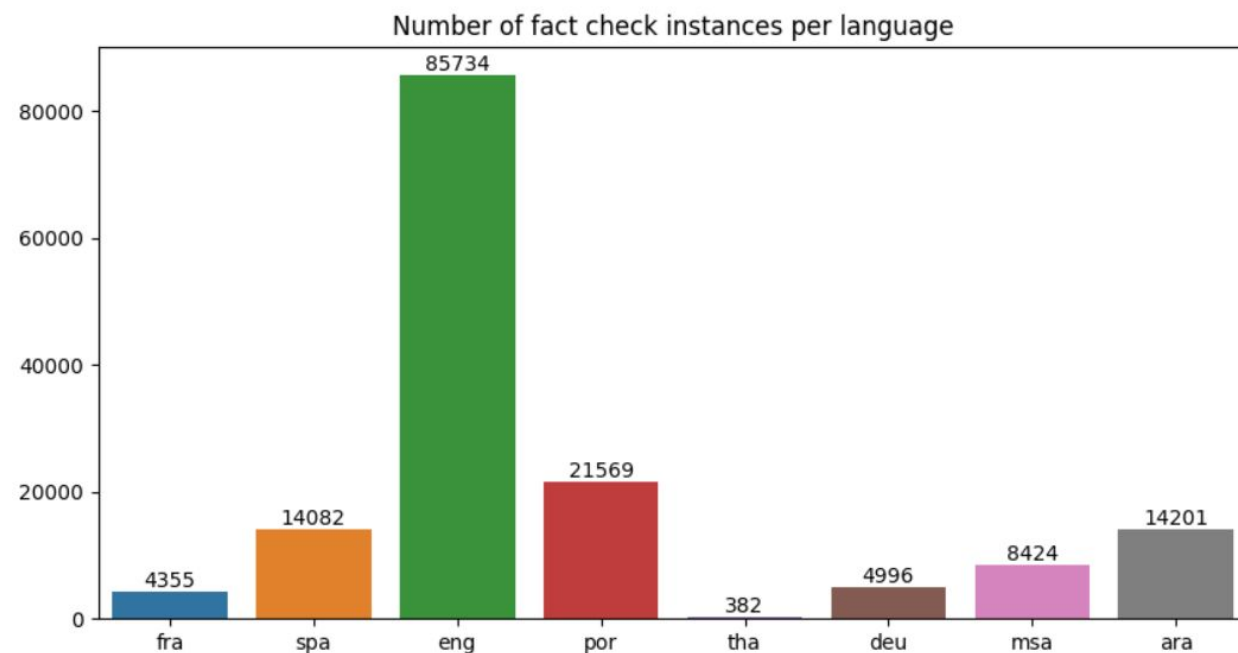
Total Posts (24,431)

Monolingual  
18,907  
17,016 + 1,891

Crosslingual  
5,524  
4,972 + 552

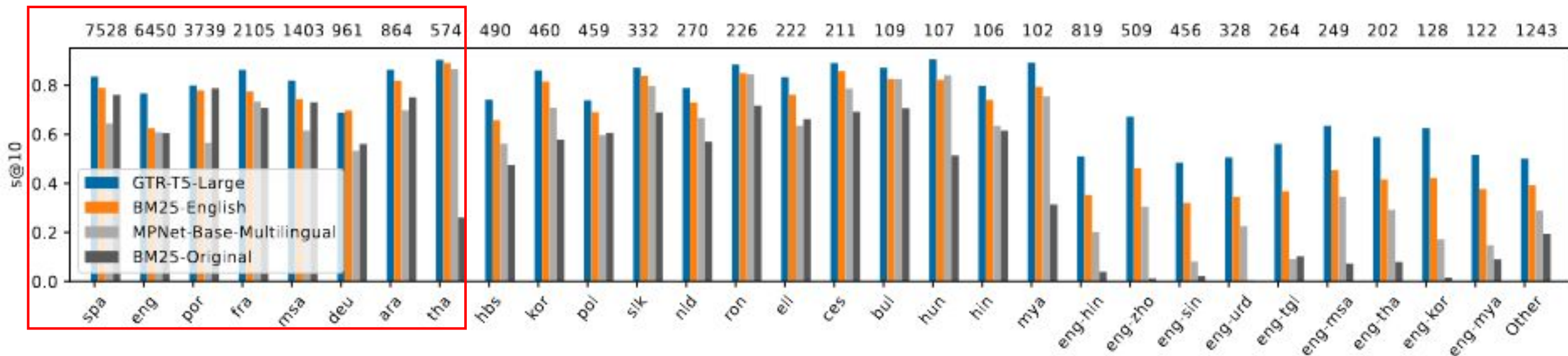
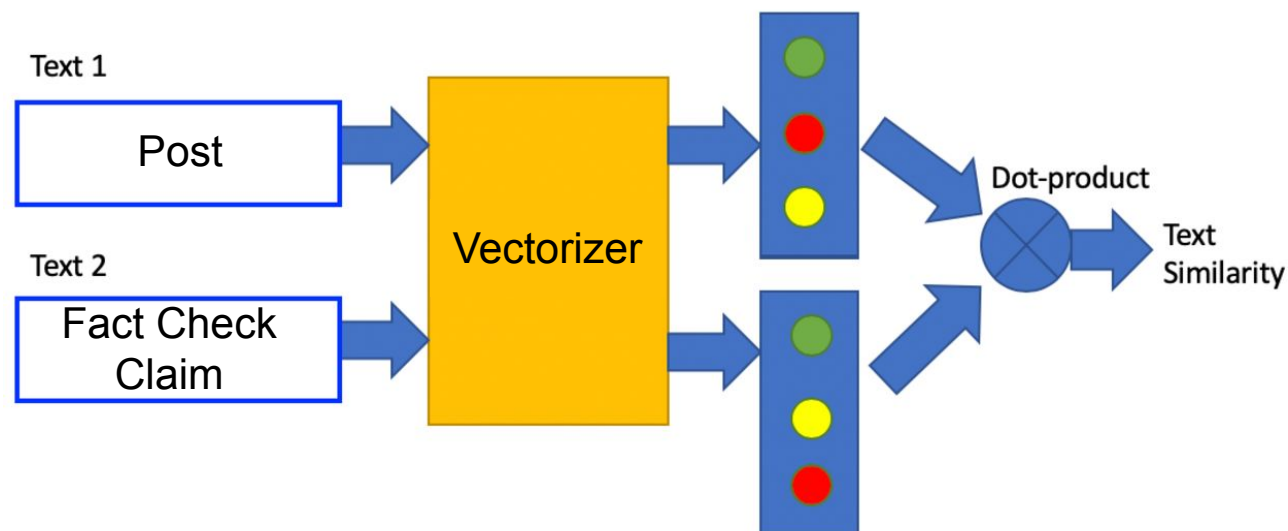
Number of posts per language

	fra	spa	eng	por	tha	deu	msa	ara
train	1596	5628	4351	2571	465	667	1062	676
dev	188	615	478	302	42	83	105	78



# Previous Approaches: Retrieval Systems

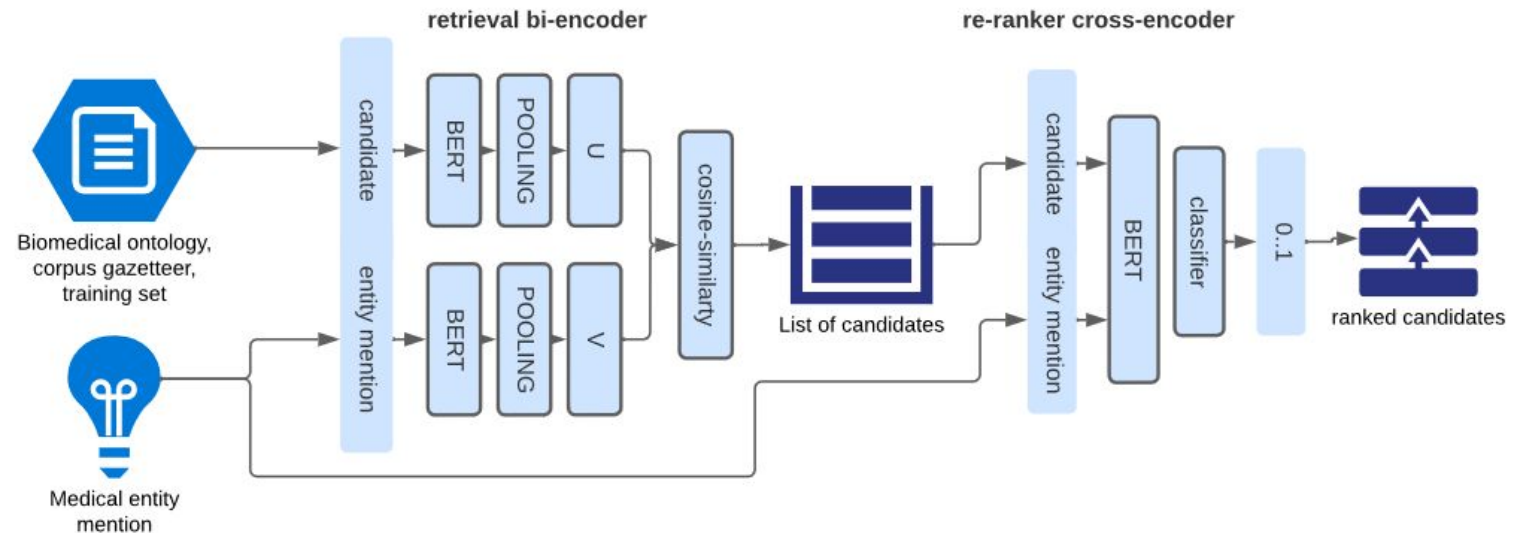
- **Sparse Retrieval:** BM25 vectorizer (tf-idf based; No training + fast inference)
- **Dense Retrieval:** Multilingual embeddings (Pre-trained + slower inference)



# Proposed Approaches: Contrastive Learning

## Contrastive Learning Approach

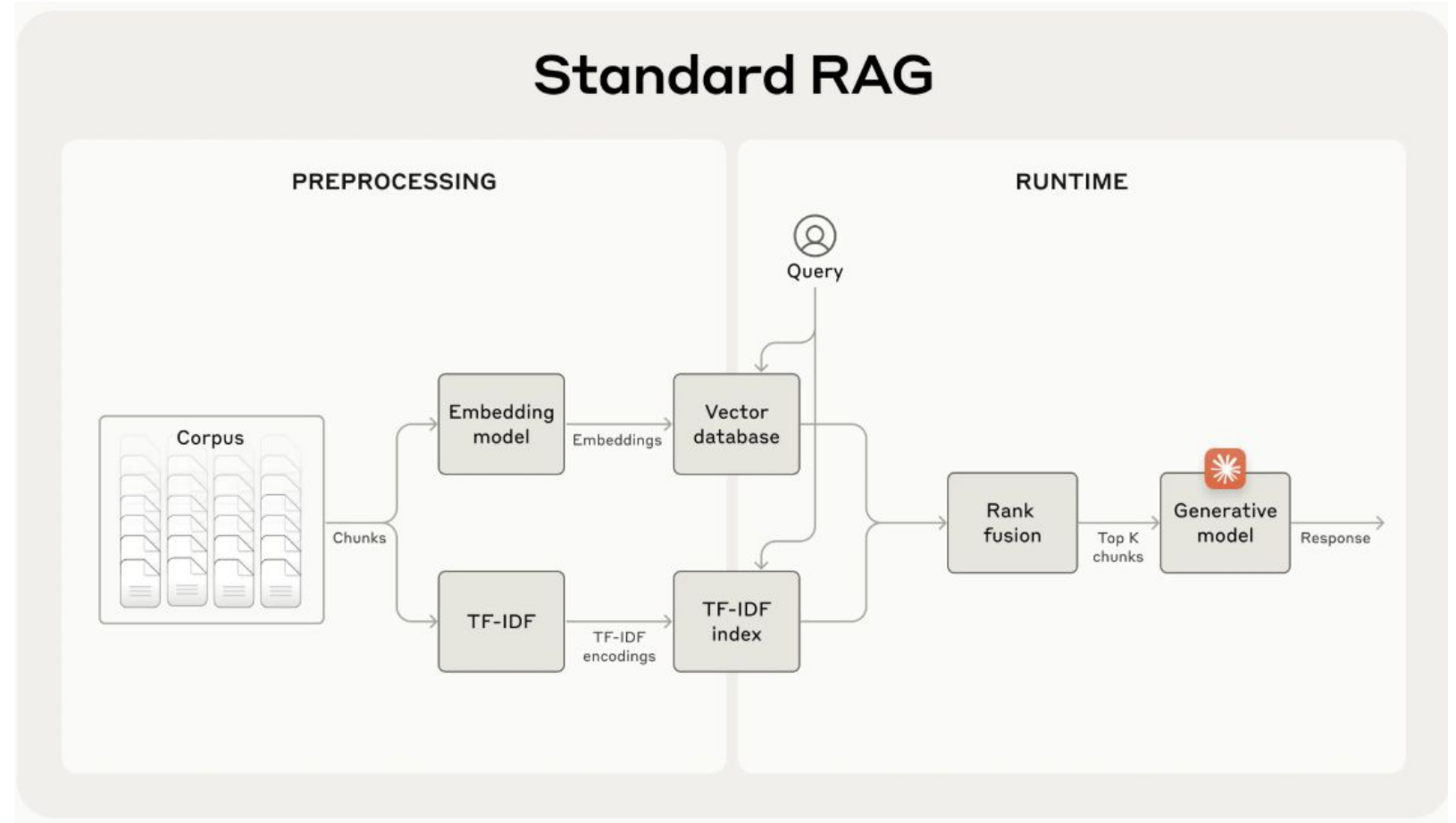
- Biencoder selection:
  - BM25
  - Multilingual TEMs
  - English Translations + TEMs
  - Hybrid
- Training Re-ranker using contrastive learning
  - Triplets Generation
    - Negative Sampling





# Proposed Approaches: Generative Models

- K-shot Naïve Generation
- Instruction Fine-tuning
- RAG



# References

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