



EACL 2026
MOROCCO

Palais Des Congres, Rabat

March 24 - 29, 2026



a place of mind

THE UNIVERSITY OF BRITISH COLUMBIA



BeDiscovER: The Benchmark of Discourse Understanding in the Era of Reasoning Language Models

Chuyuan Li, Giuseppe Carenini

Department of Computer Science
The University of British Columbia

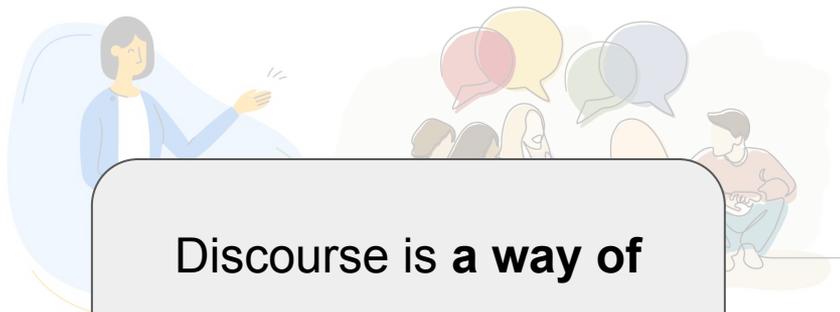
Why discourse understanding?

Speech

Dialogue

Scientific papers

News articles



Discourse is a way of speaking.

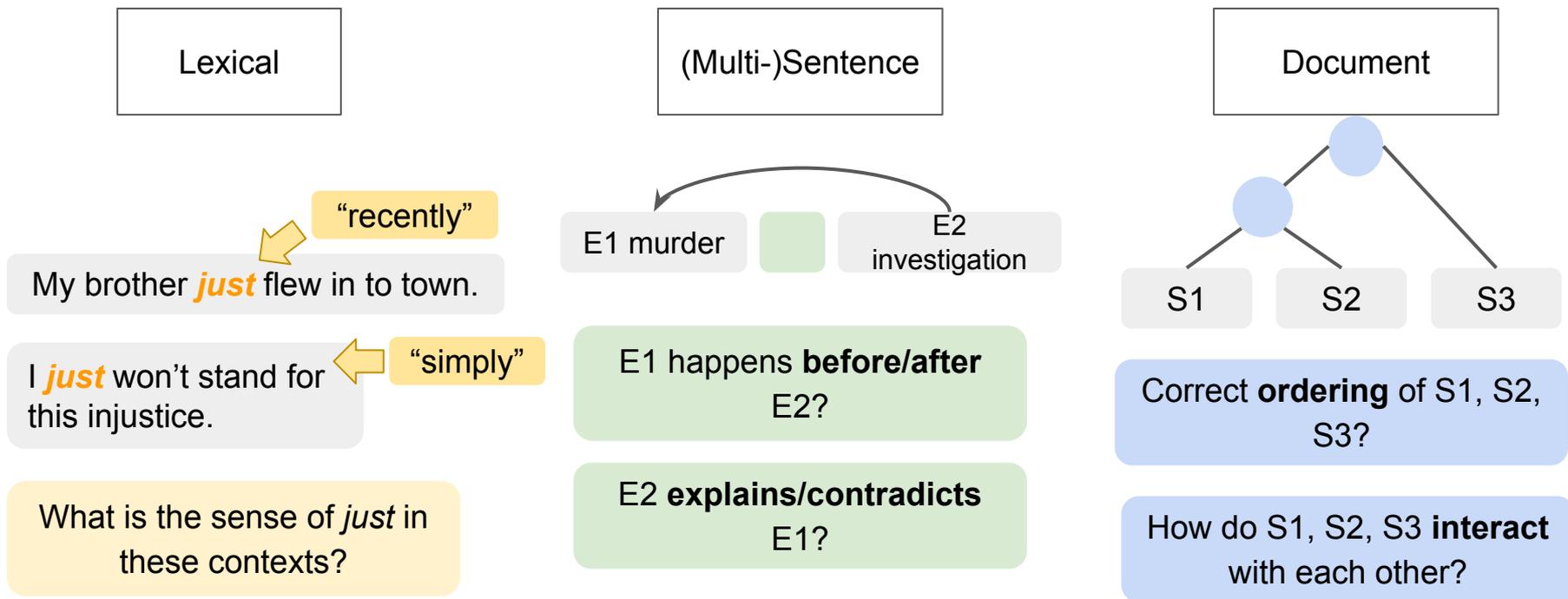


Rhetoric / Persuasion

But also:

- Text structure
- Communicative / behavior intentions
- Temporal reasoning
- Causal reasoning
- ...

Why discourse understanding? – a few examples



Why discourse understanding? – a few examples

Lexical

(Multi-)Sentence

Document

*Discourse understanding requires **lexical & semantic, temporal, rhetorical, commonsense...** knowledge.*

How well do modern LLMs understand discourse?

My brother *just* flew

“simply”

I *just* won't stand for this injustice.

S2

S3

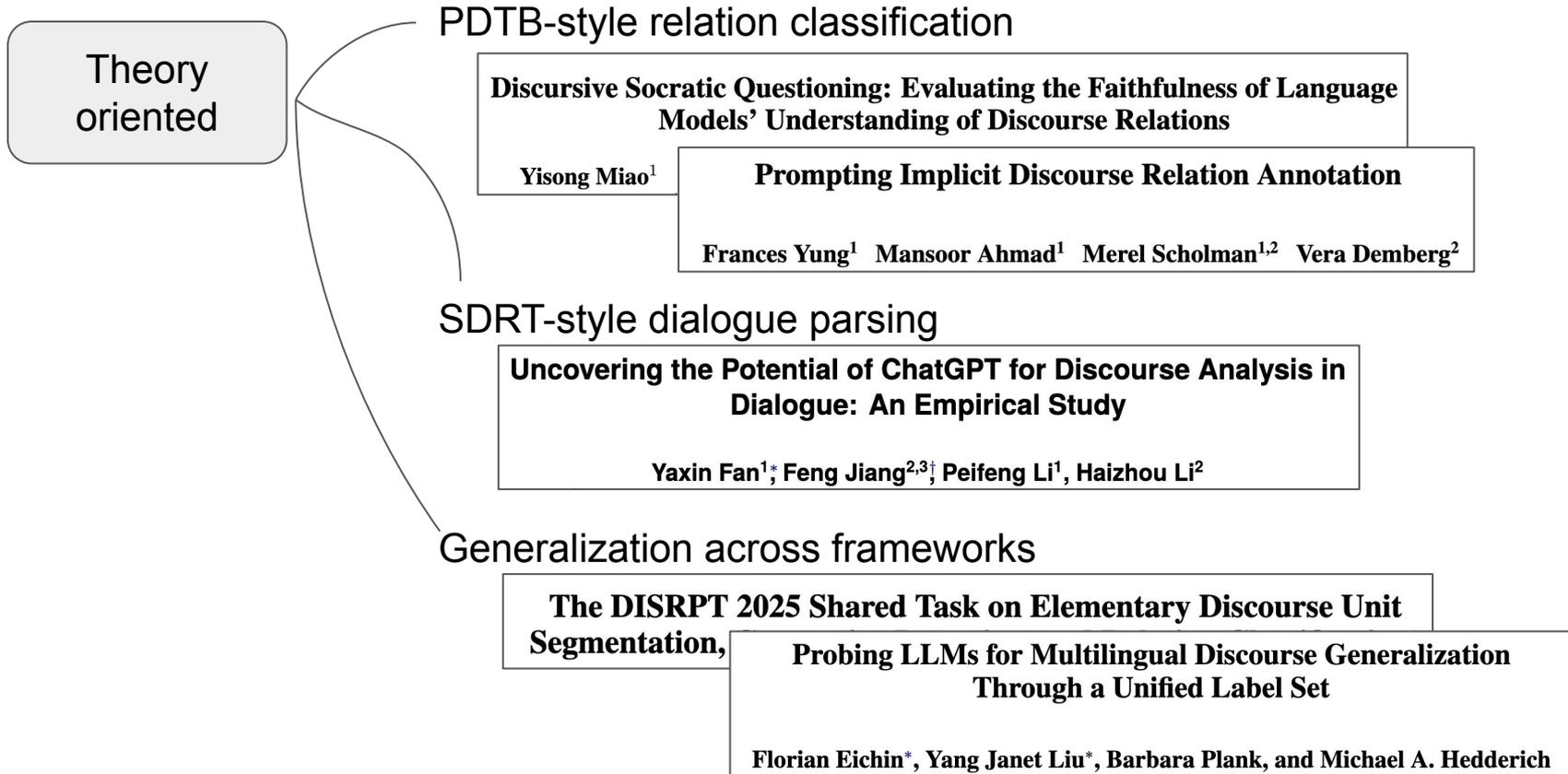
Correct ordering of S1, S2,

S3?

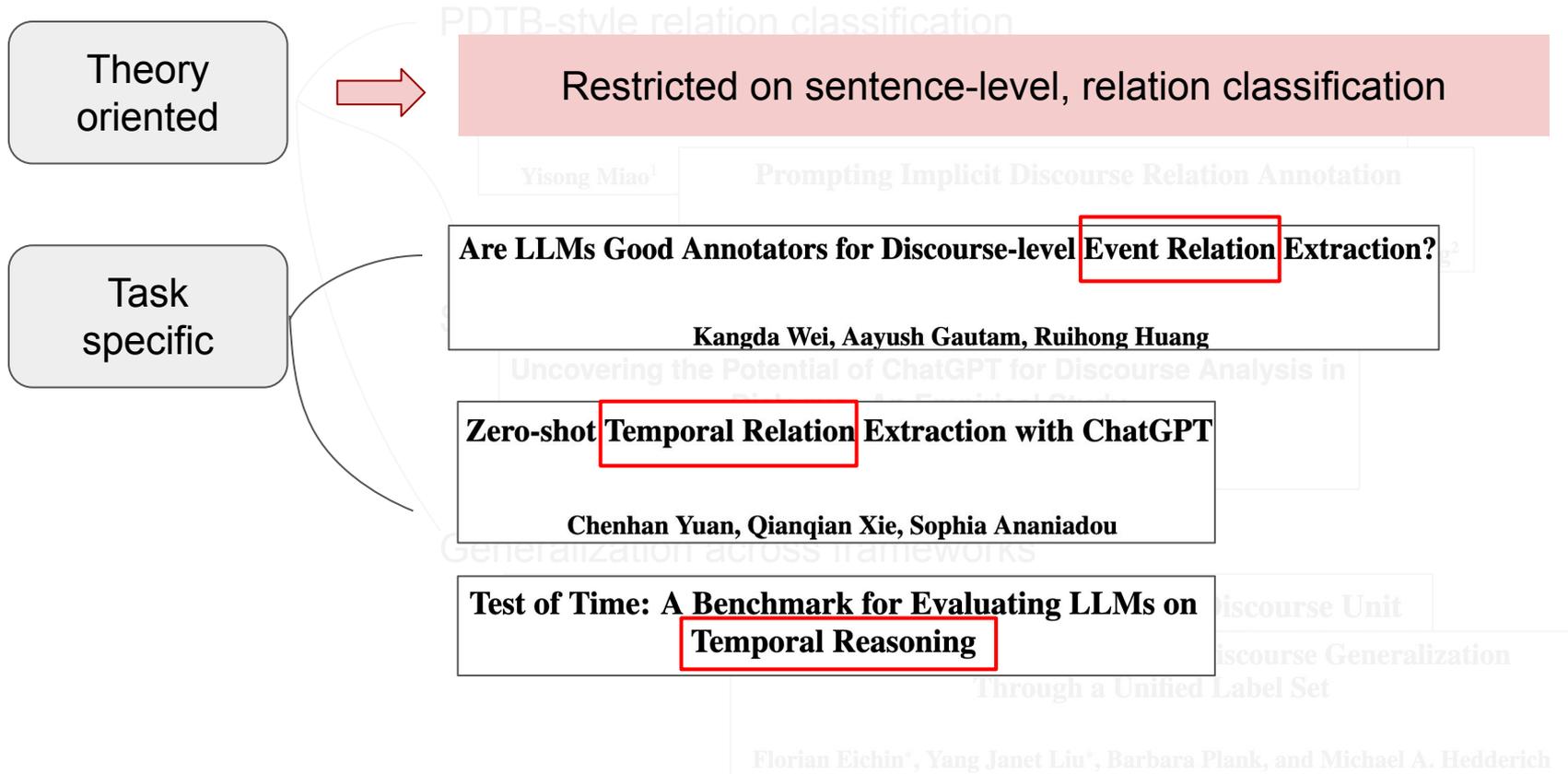
S1, S2, S3 interact with each other?



Why discourse understanding? – existing studies



Why discourse understanding? – existing studies



Why discourse understanding? – existing studies

Theory oriented



Restricted on sentence-level, relation classification

Task specific



Focus on certain aspects of discourse understanding

Discourse benchmarks

Disco-Bench: A Discourse-Aware Evaluation Benchmark for Language Modelling

Longyue Wang, Zefeng Du,
Donghuai Liu, Cai Deng, Dian Yu,
Haiyun Jiang, Yan Wang, Leyang
Cui, Shuming Shi, Zhaopeng Tu*

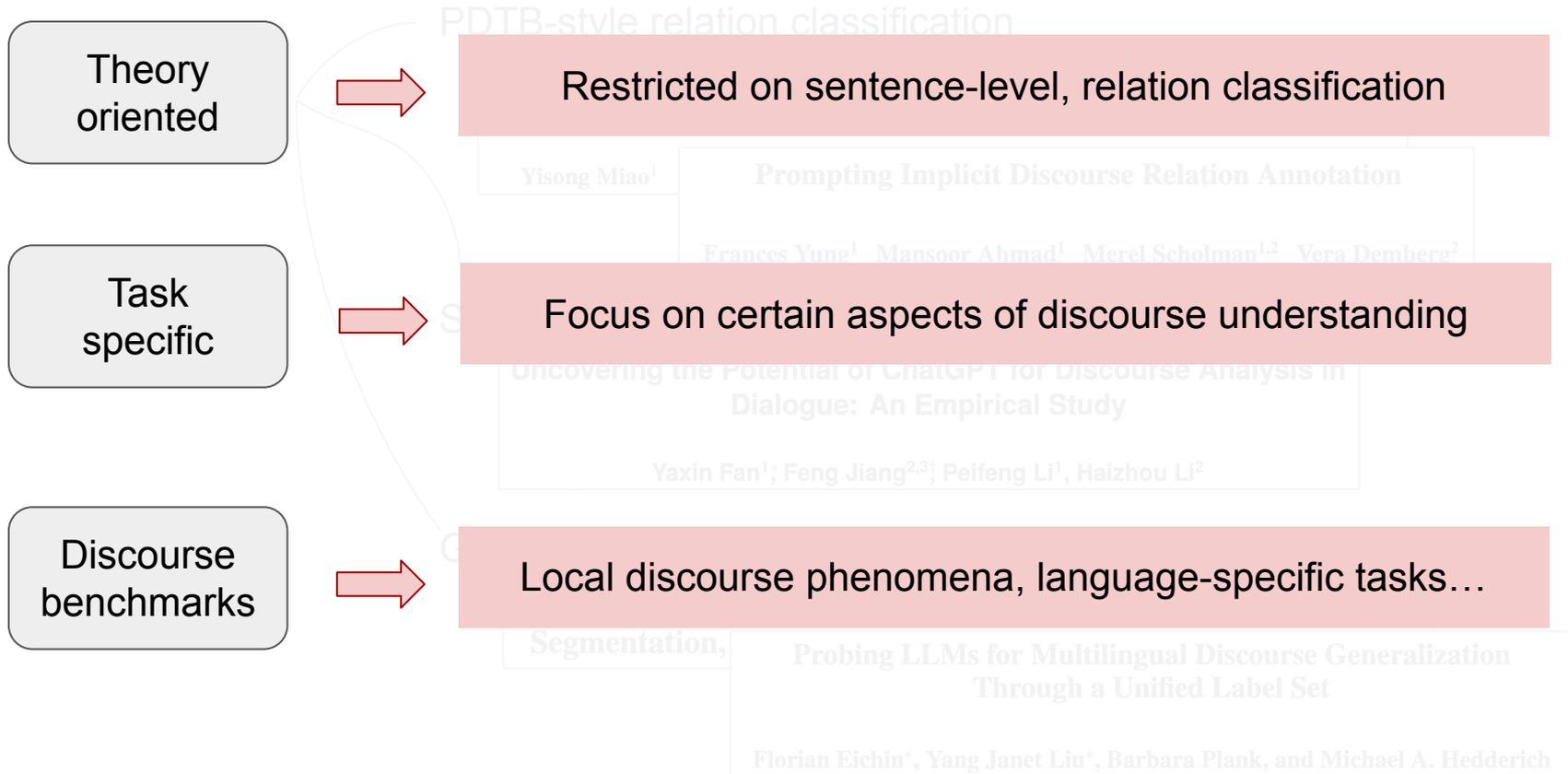
Evaluation Benchmarks and Learning Criteria for Discourse-Aware Sentence Representations

Mingda Chen^{2*} Zewei Chu^{1*} Kevin Gimpel²

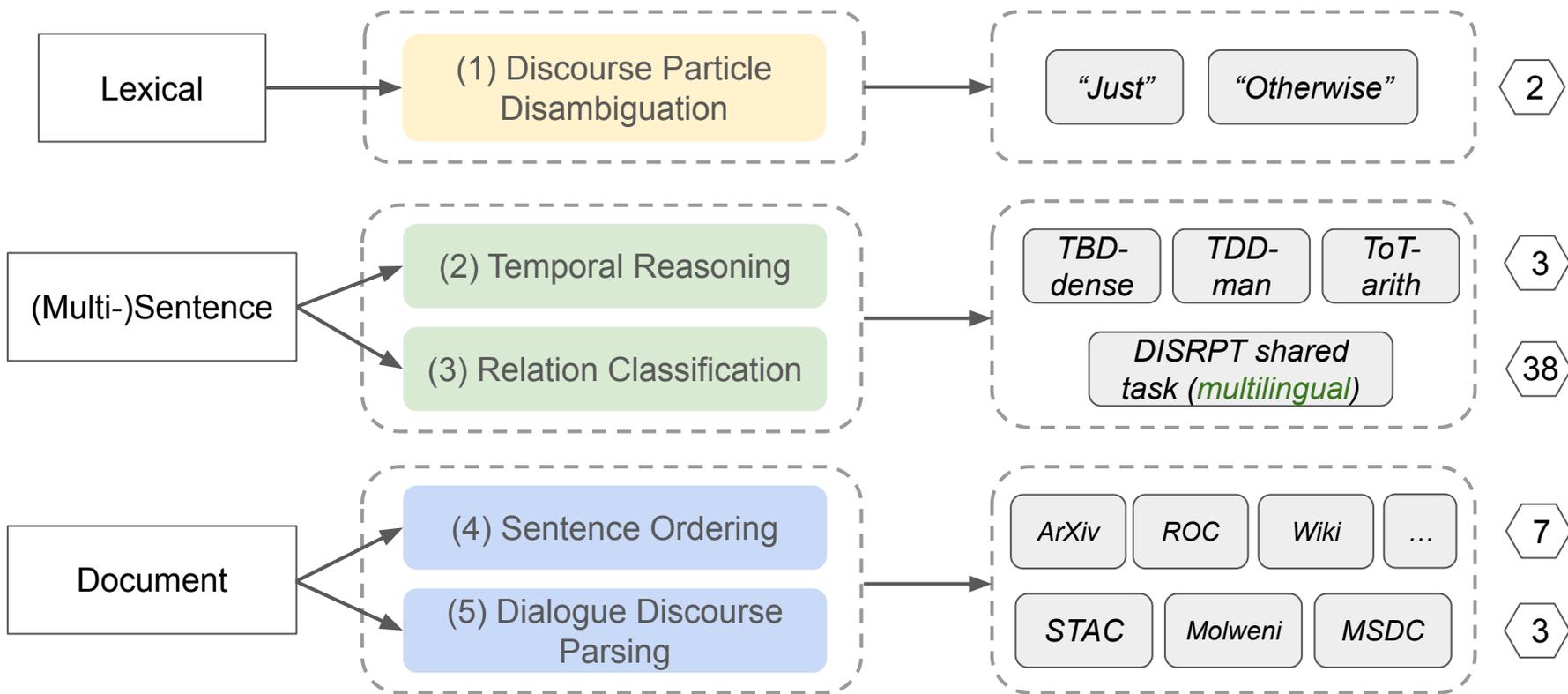
DiscoTrack: A Multilingual LLM Benchmark for Discourse Tracking

Lanni Bu Lauren Levine Amir Zeldes

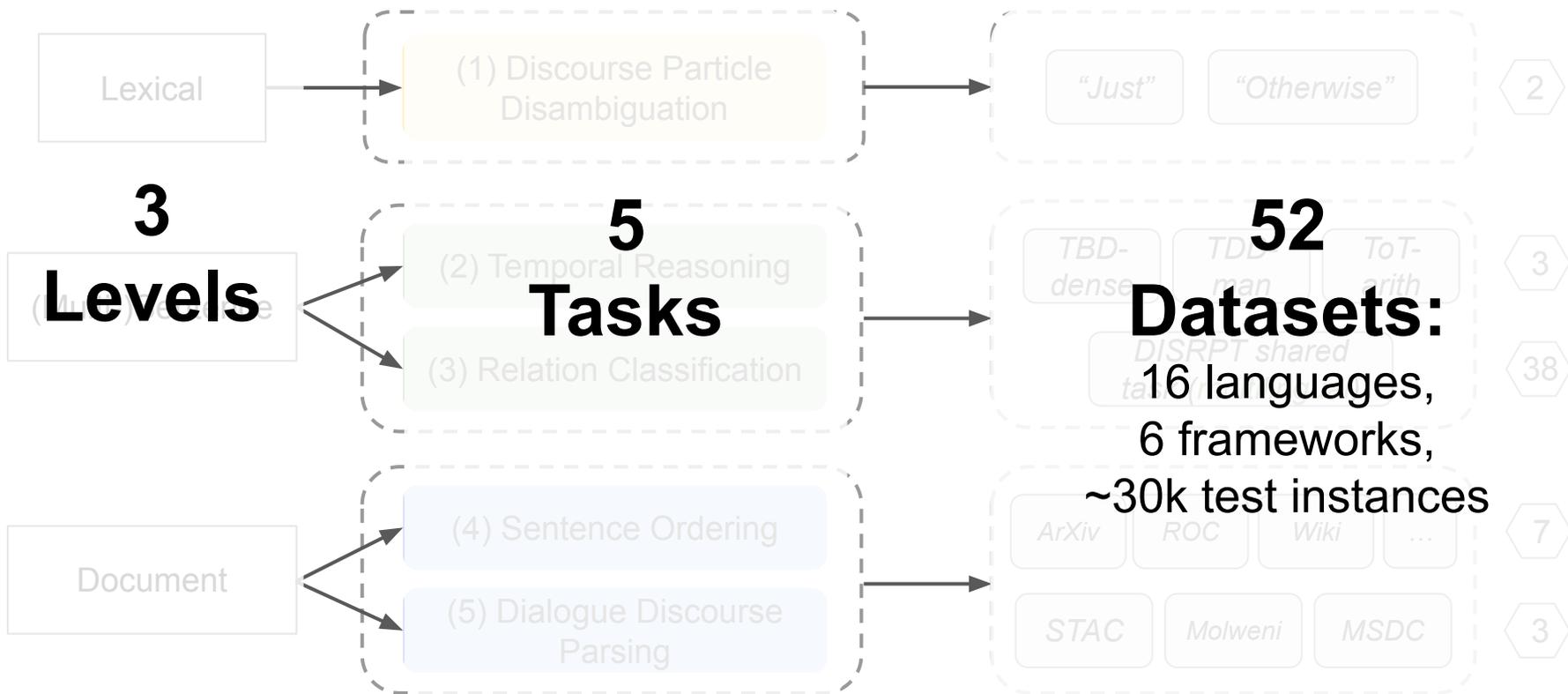
Why discourse understanding? – existing studies



Our proposal: BeDiscoverER!



Our proposal: BeDiscovER!



Open-ended Question-Answer Formatting

- **Unified evaluation pipeline**
- Classification tasks (1 2 3): fixed label space
- Parsing task (5): incremental generation task



Reasoning-oriented LLMs



GPT-5



Qwen3



DeepSeek-r1

Non reasoning-oriented LLMs



Llama-4

Qwen2.5

GPT4o

System prompt:

... Choose one of the following six labels: **[Exclusionary, Unelaboratory, Unexplanatory, Emphatic, Temporal, Adjective]**.

User prompt:

My brother *just* flew in to town.

Question: What is the function of the discourse marker “just” in the sentence above?



Temporal

Reasoning-oriented LLMs

Non reasoning-oriented LLMs



GPT-5



Qwen3



DeepSeek-r1

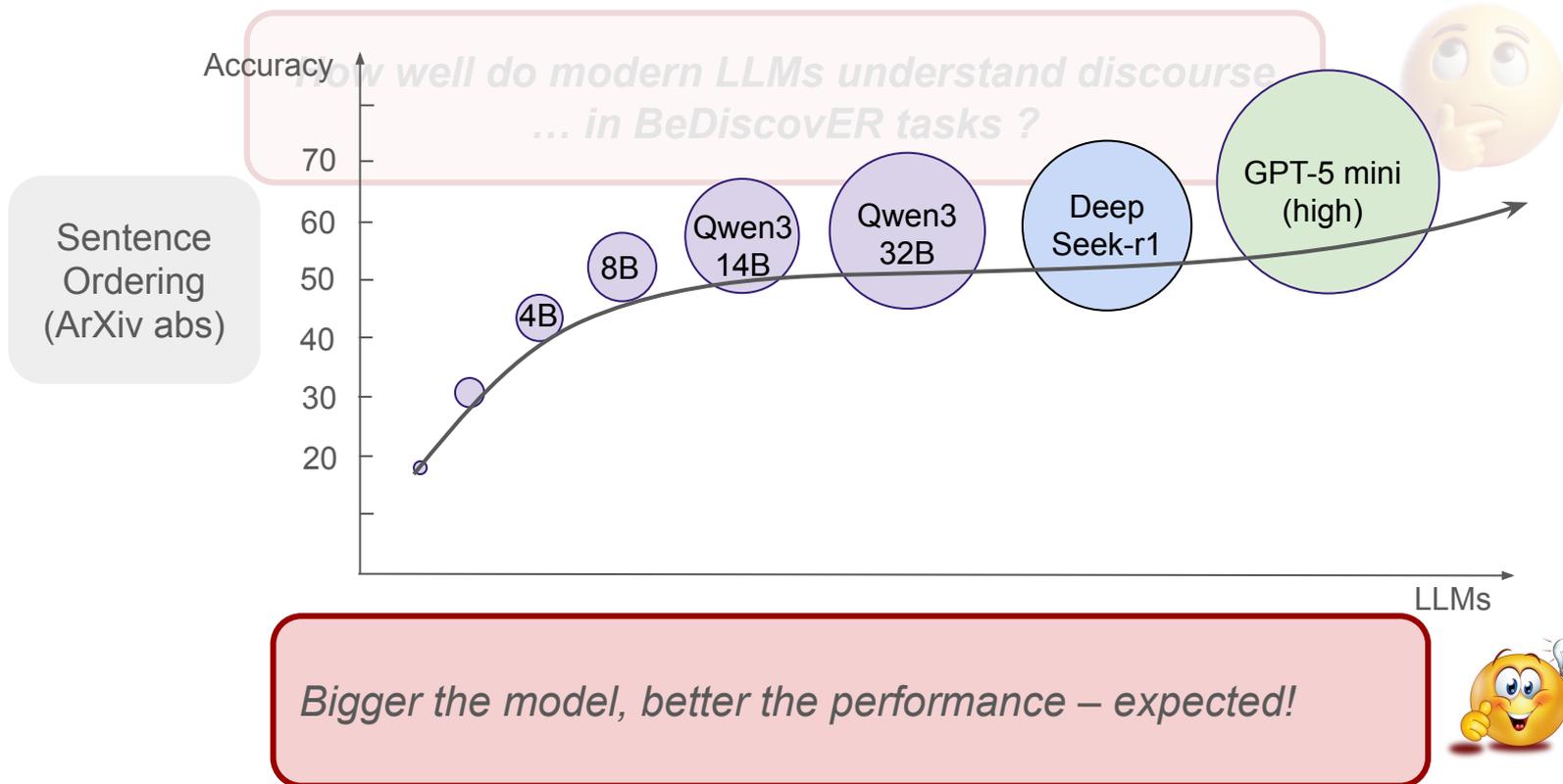


Llama-4

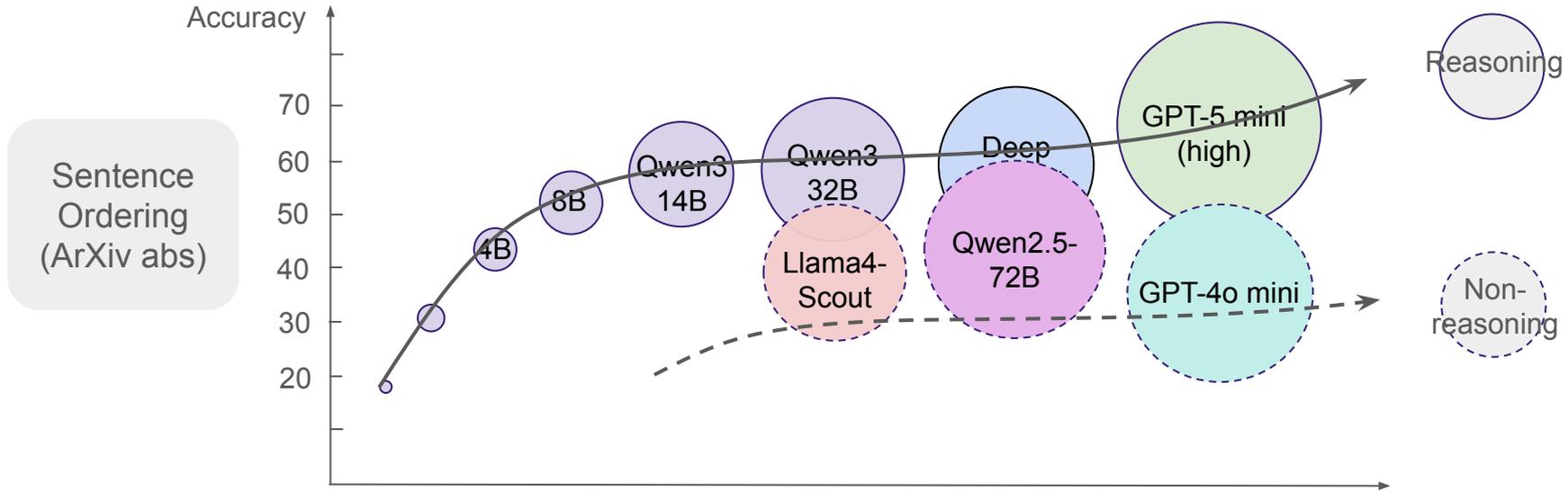
Qwen2.5

GPT4o

Performance: model scaling



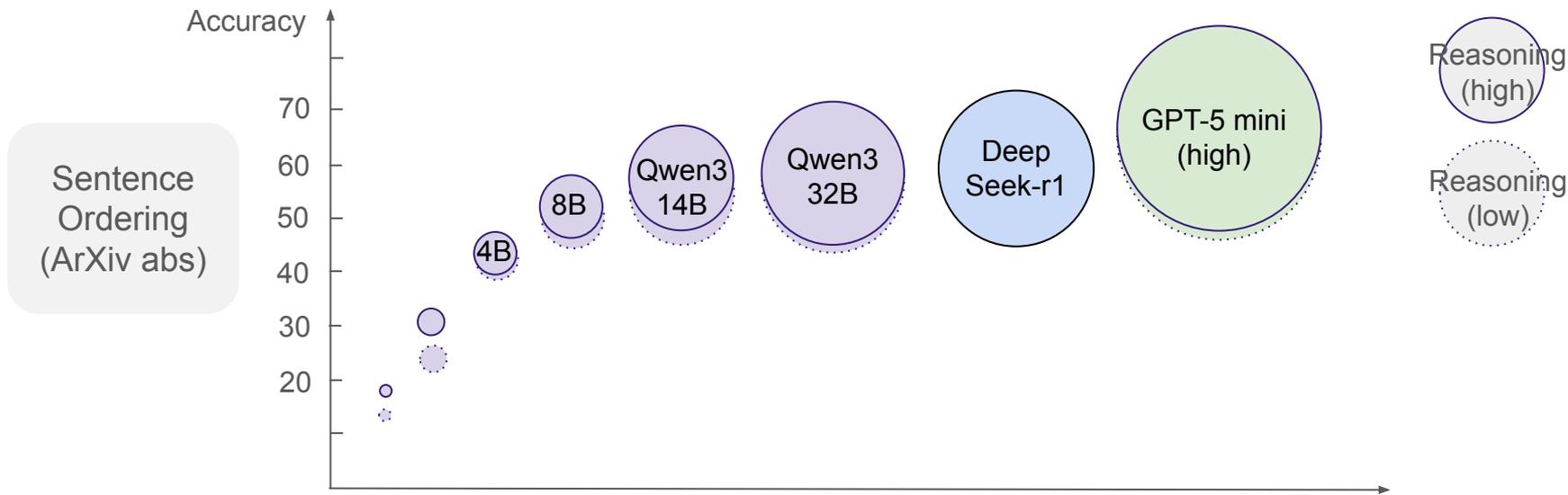
Performance: reasoning-oriented vs. non-reasoning LLMs



Reasoning-oriented LLMs outperform non-reasoning optimized LLMs – expected!

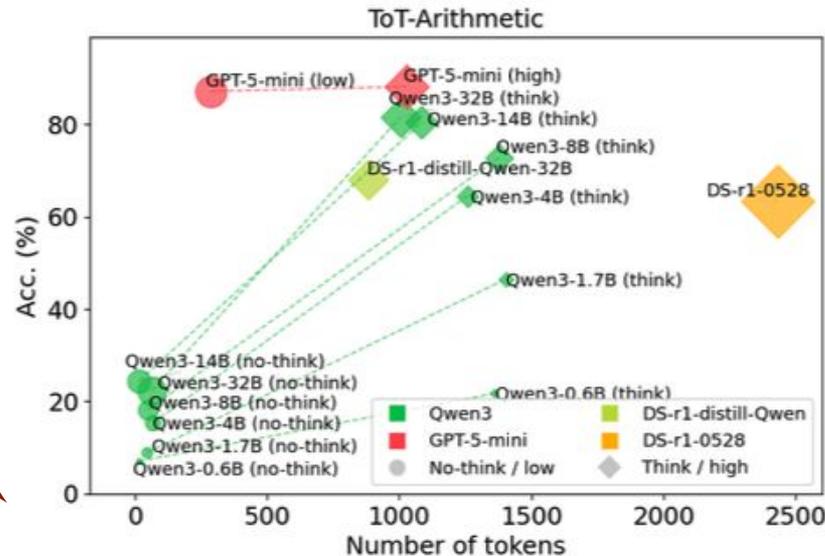
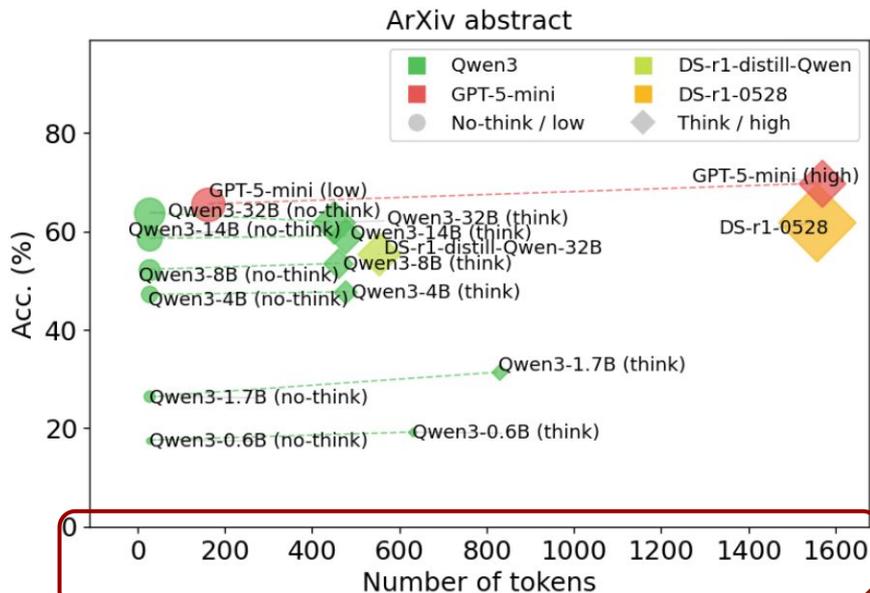


Performance: Higher reasoning effort, better result?



Higher thinking effort does not yield better outcome

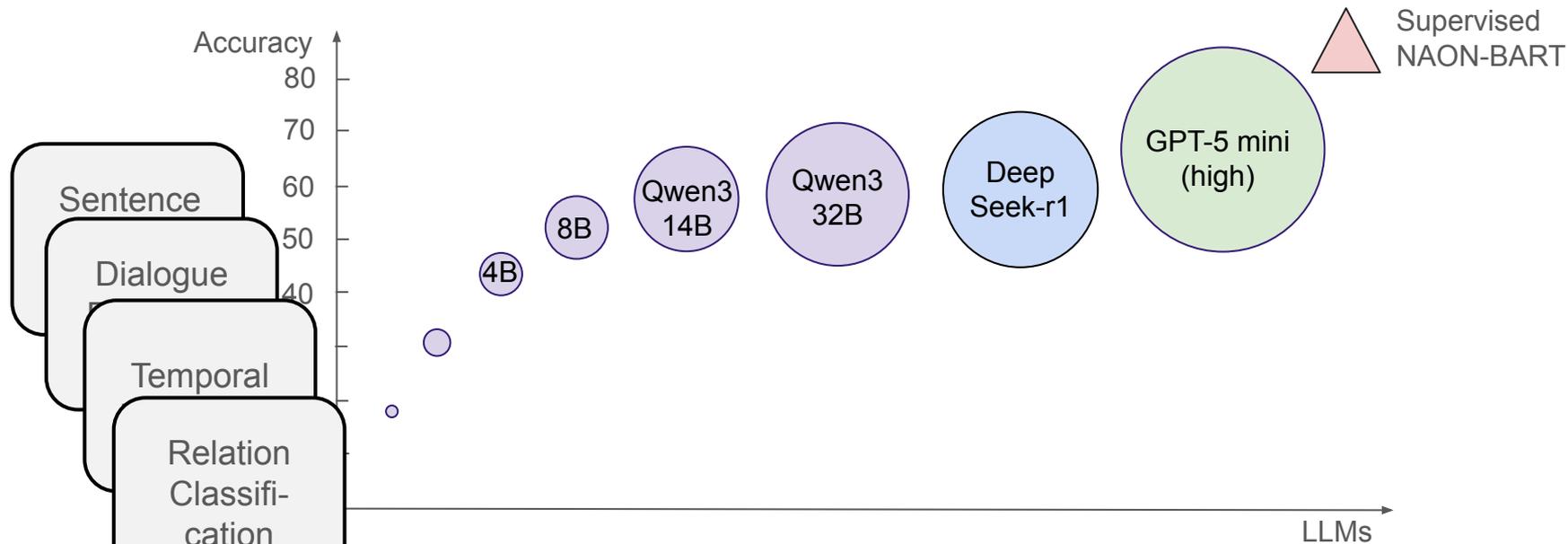
Performance: Higher reasoning effort, better result?



Models become verbose, yet, flat performance

*... except for **arithmetic temporal**: long reasoning lead to big gains!*

Performance: LLMs vs. supervised models



Reasoning-oriented LLMs show markedly lower performance (~10–30%) compared to supervised models.



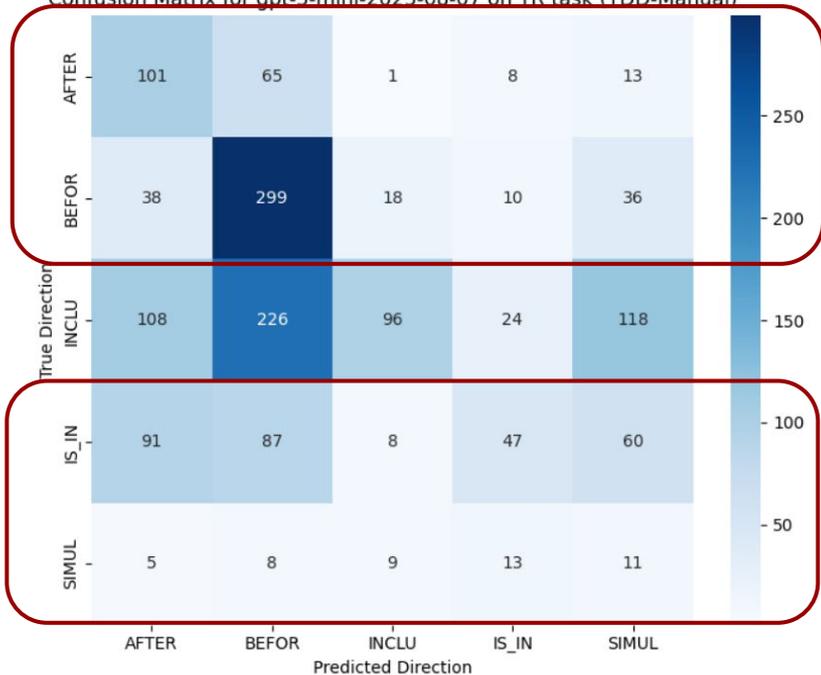
Analysis: Fine-grained sense disambiguation



Temporal Reasoning

Relation Classification

Confusion Matrix for gpt-5-mini-2025-08-07 on TR task (TDD-Manual)

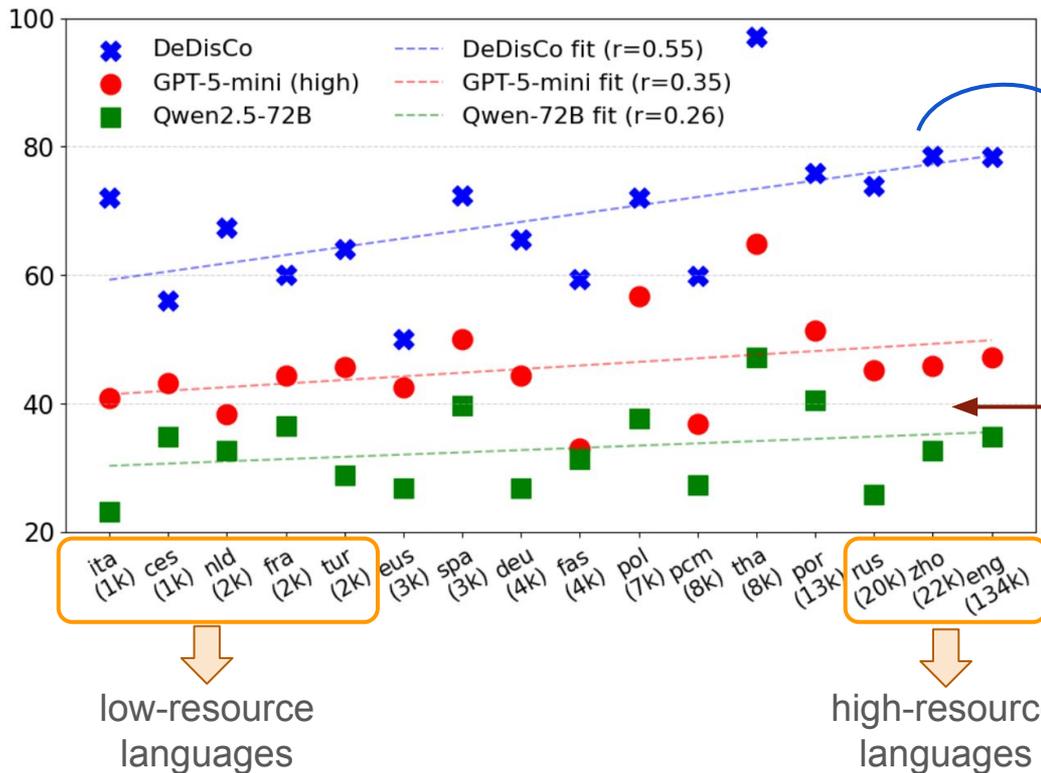


Identifies *before/after* relations

but fails to capture *overlap* or *containment*.

Analysis: Multilingual performance

Relation Classification



Supervised model show clear perform disparities (20%)

LLMs: nearly flat, lack of robust relation representation

Analysis: Inter-task correlation

Performance similarity
(Pearson r)

- For each task: aggregated results of all datasets
- Metric: accuracy
- Pairwise correlation: strong positive ($r \in [0.73, 0.98]$, $p < 0.05$)

- Lexicon & semantics
 - Rhetorical
 - Temporal
 - Logic
 - Arithmetic
- Commonsense

“Just”

Relation classification

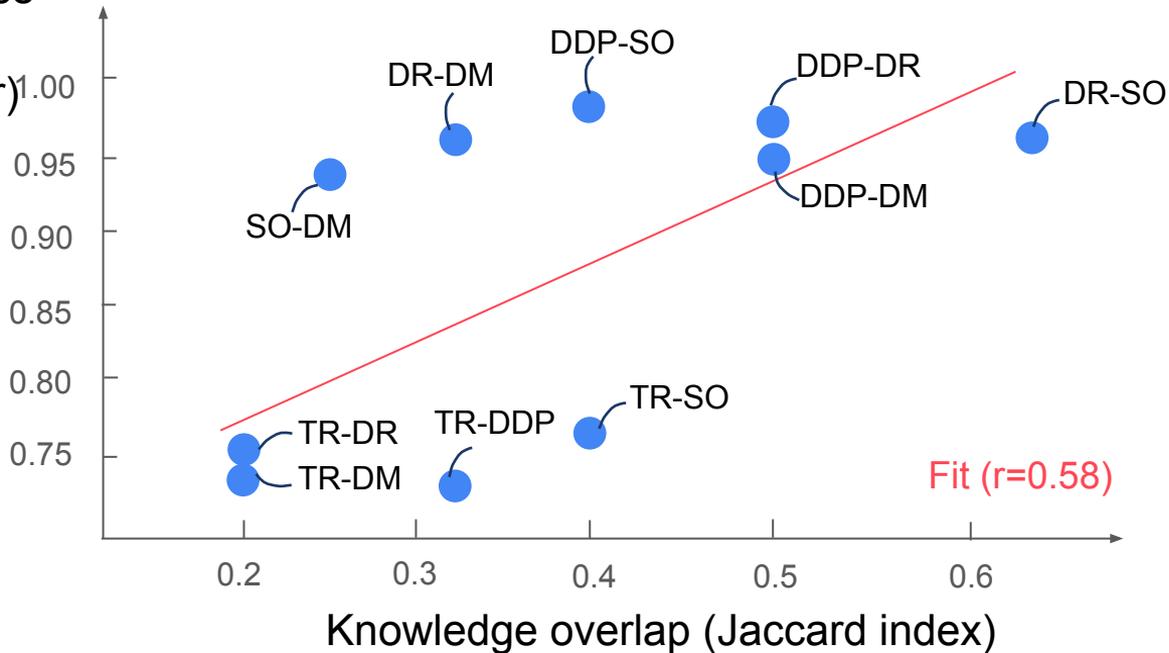
Overlap calculation:

$$Jaccard(A, B) = \frac{|A \cup B|}{|A \cap B|}$$

Knowledge overlap (Jaccard index)

Analysis: Inter-task correlation

Performance similarity (Pearson r)



- ❖ DDP – dialogue discourse parsing
- ❖ SO – sentence ordering
- ❖ TR – temporal
- ❖ DR – relation
- ❖ DM – markers

Summary: benchmark and evaluation baseline



Reasoning-oriented LLMs capture some discourse-level knowledge, especially **good in arithmetic** temporal reasoning.

Struggle with subtle semantic phenomena (e.g. rhetorical relations, for all languages)

Longer reasoning traces do not necessarily yield better outcomes
→ how to **improve the quality** of reasoning LLMs?

Data contamination? → **dynamic** benchmark, welcome contribution!



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Thank you and welcome to our Oral Presentation at Salle Le LIXUS, Wednesday March 25 at 14:30-16:00 if you have any question or would like to chat more!