

Is Everything in Order?

A simple way to Order Sentences

Somnath Basu Roy Chowdhury* , Faeze Brahman*
Snigdha Chaturvedi

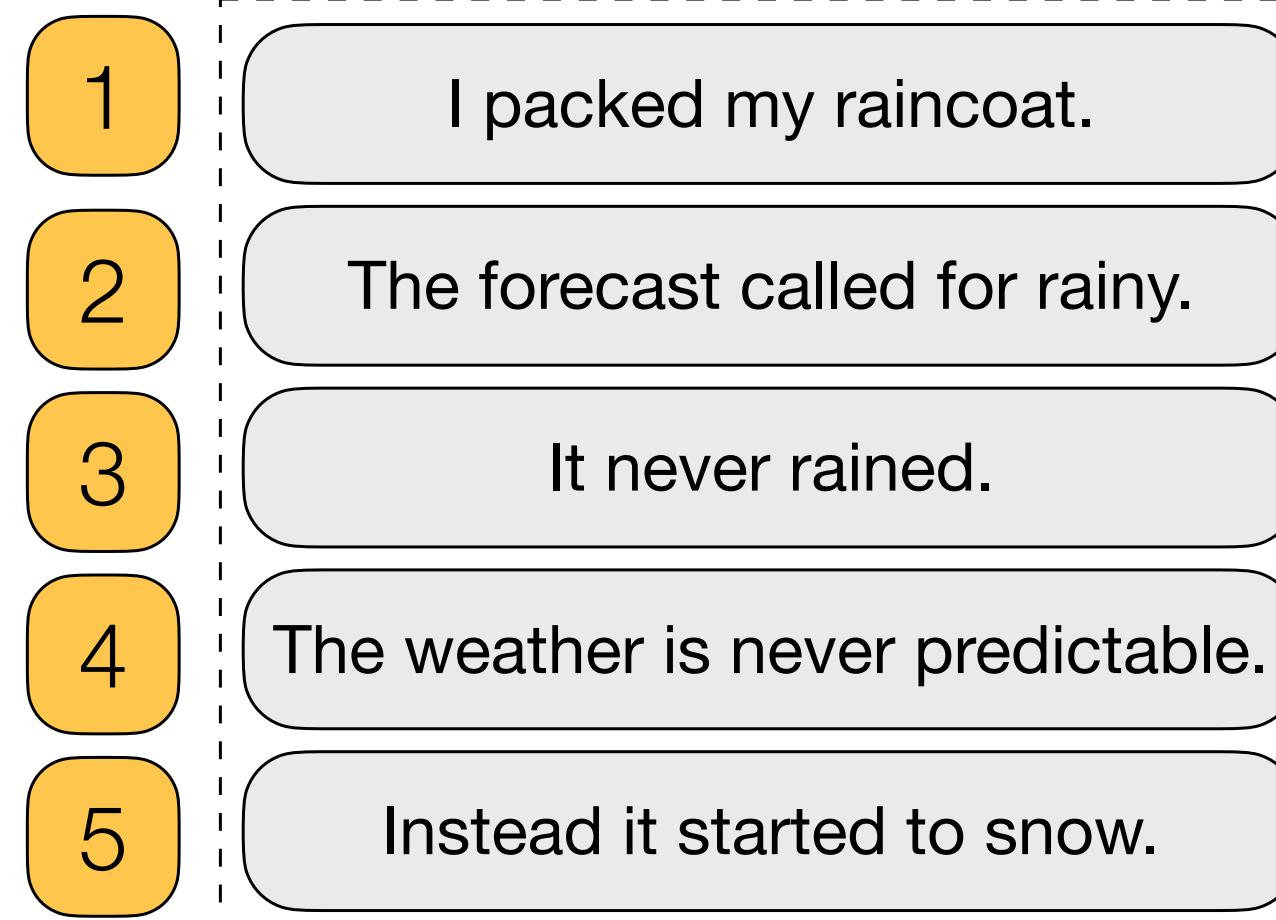


Motivation

Sentence Ordering Task:

- Organizing a shuffled set of sentences into a coherent text

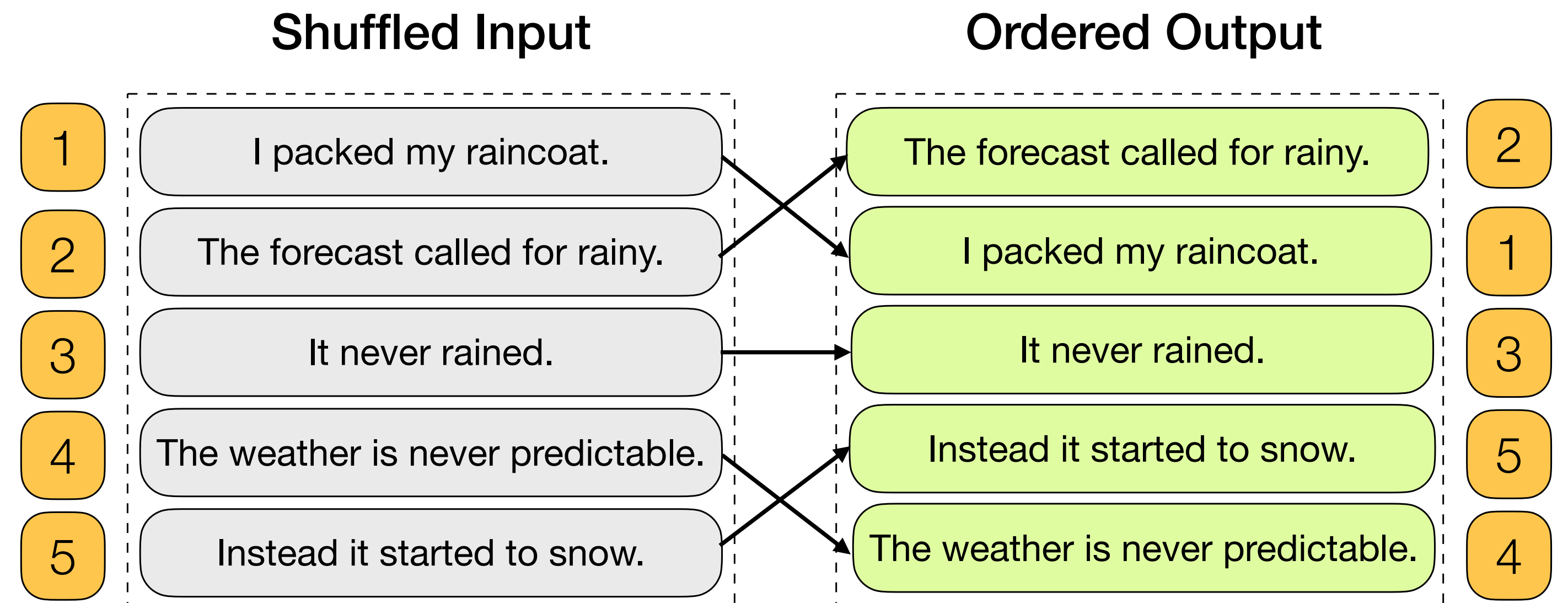
Shuffled Input



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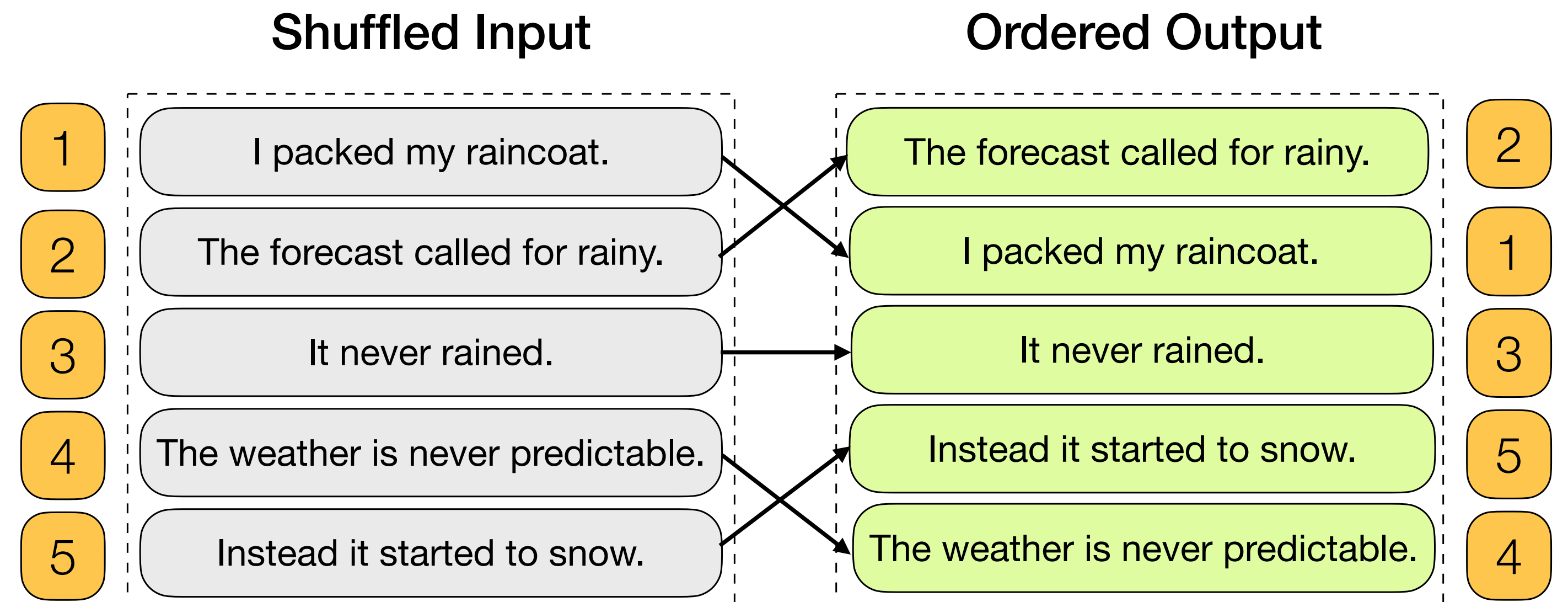
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Motivation

Sentence Ordering Task:

- Organizing a shuffled set of sentences into a coherent text
- Requires understanding of causal and temporal relations.
- Applications in NLG, QA, etc.



Prior Works

- **Pointer networks for Pairwise Ranking** (Gong et al., 2016, Logeswaran et al., 2018a, Cui et al., 2018, Yin et al., 2019, 2020)
- Solving a ranking problem (Chen et al., 2016)
- Constraint solving + topological sorting (Prabhumoye et al., 2020)
- SOTA: Novel Pointer Decoder with Deep relational module (Cui et al., 2020)

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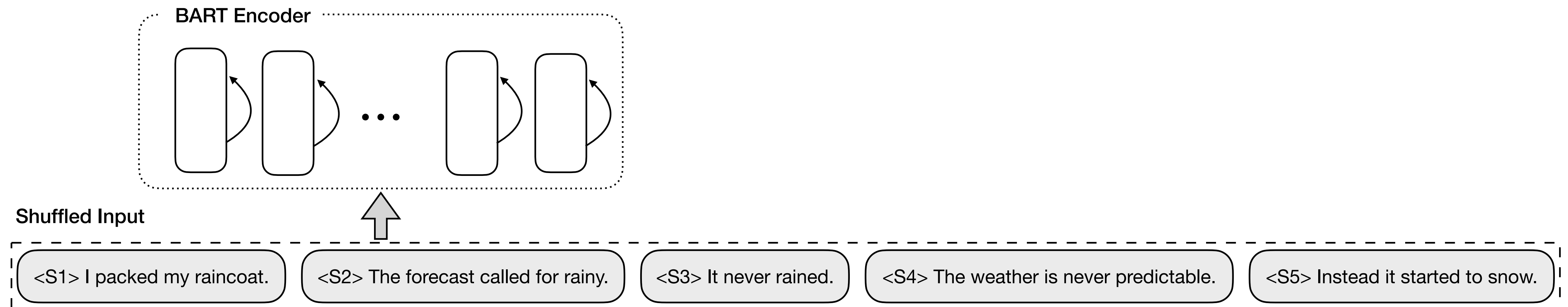
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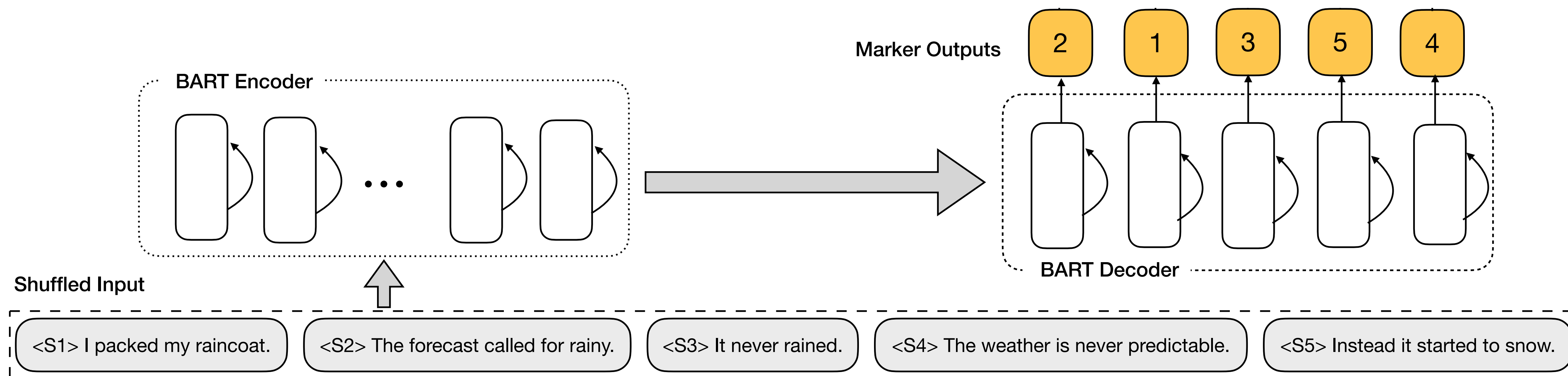
Re-order BART (Re-Bart)

- Solve the task as a conditional text-to-marker generation problem



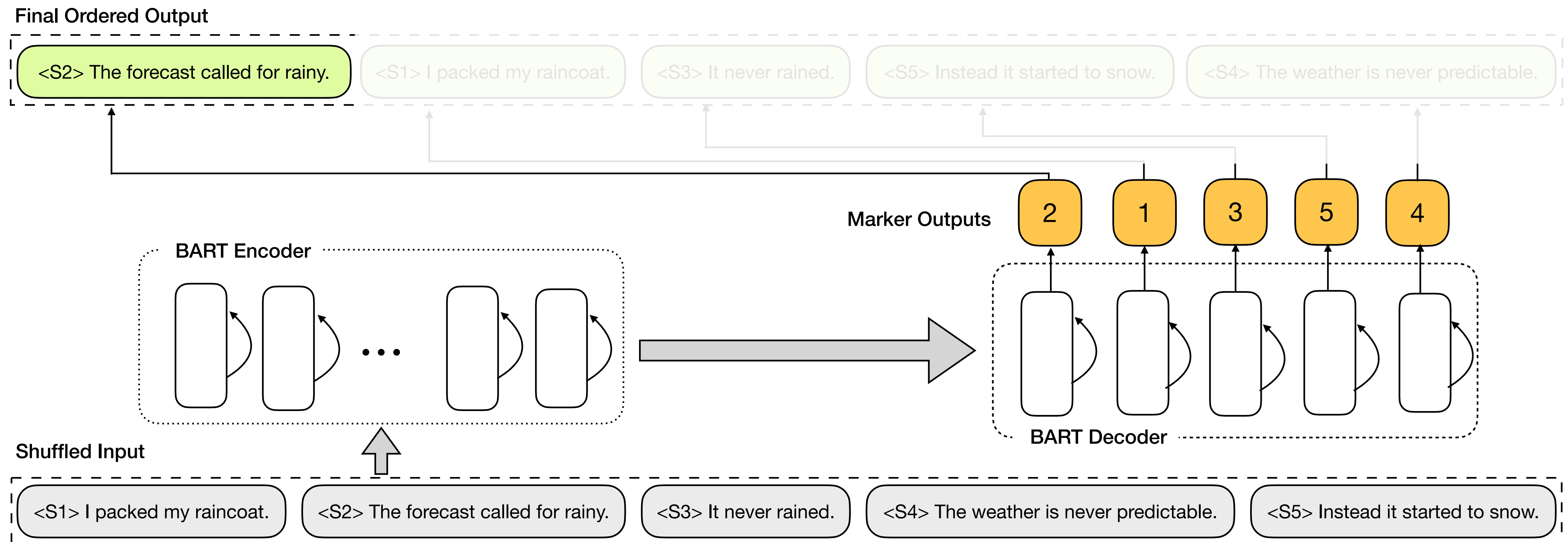
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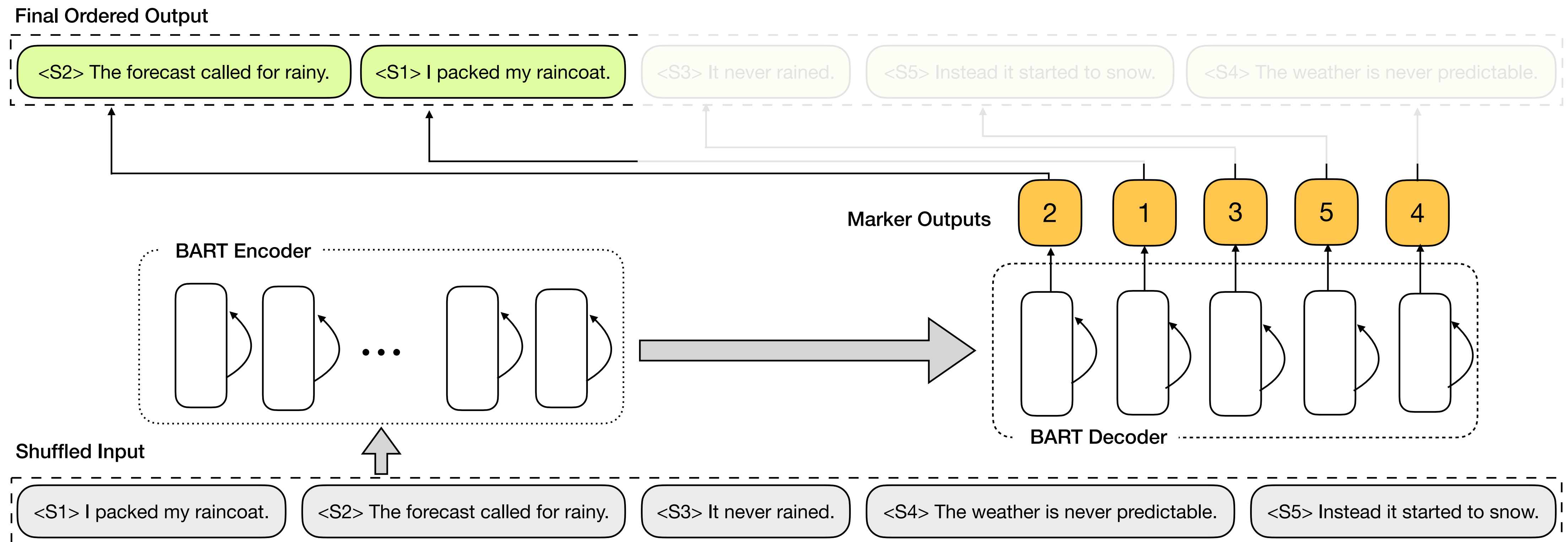
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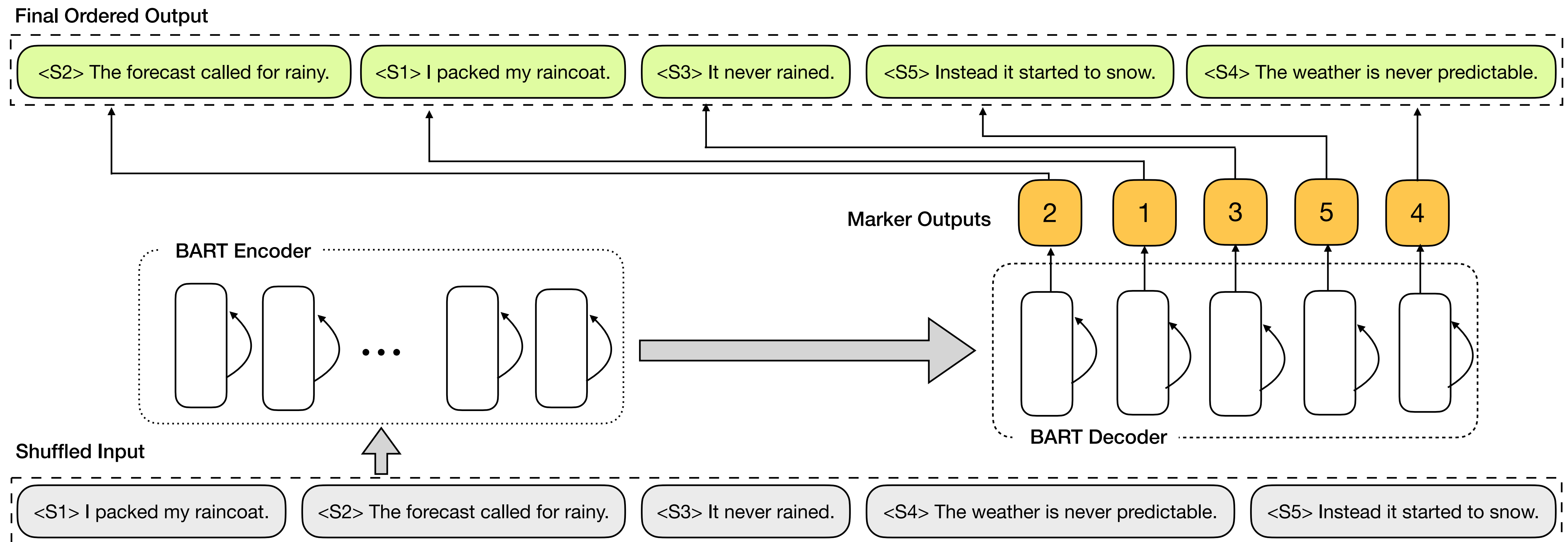
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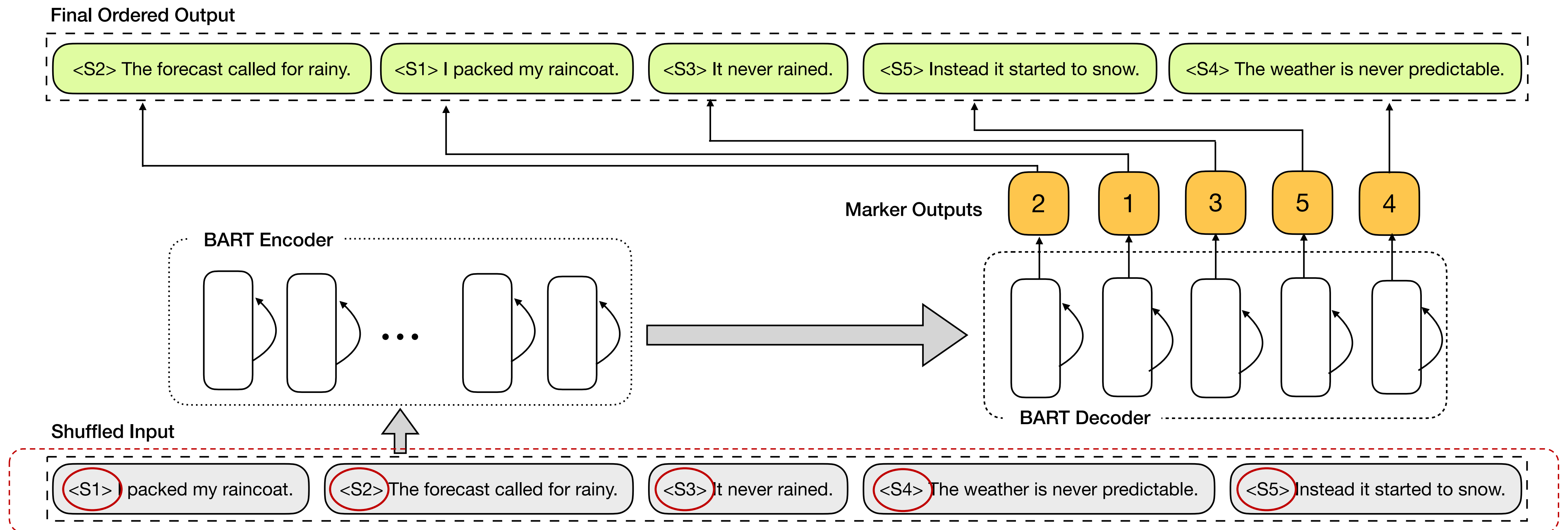
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What are the advantages of text-to-marker setup over text-to-text setup?

- Less susceptible to neural degeneration.
- Text-to-text often generate tokens that are not part of the input.

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Datasets

Paper Abstracts

- NeurIPS
- AAN
- ACL
- NSF Research Awards
- arXiv

Narratives

- ROCStoris
- SIND
- **Wikipedia Movie Plots**

Evaluation Metrics

- **Accuracy:** The fraction of output sentence positions predicted correctly
- **Perfect Match Ratio:** The fraction of sentence orders exactly matching with the correct order
- **Kendall's Tau:** The correlations between predicted and gold order

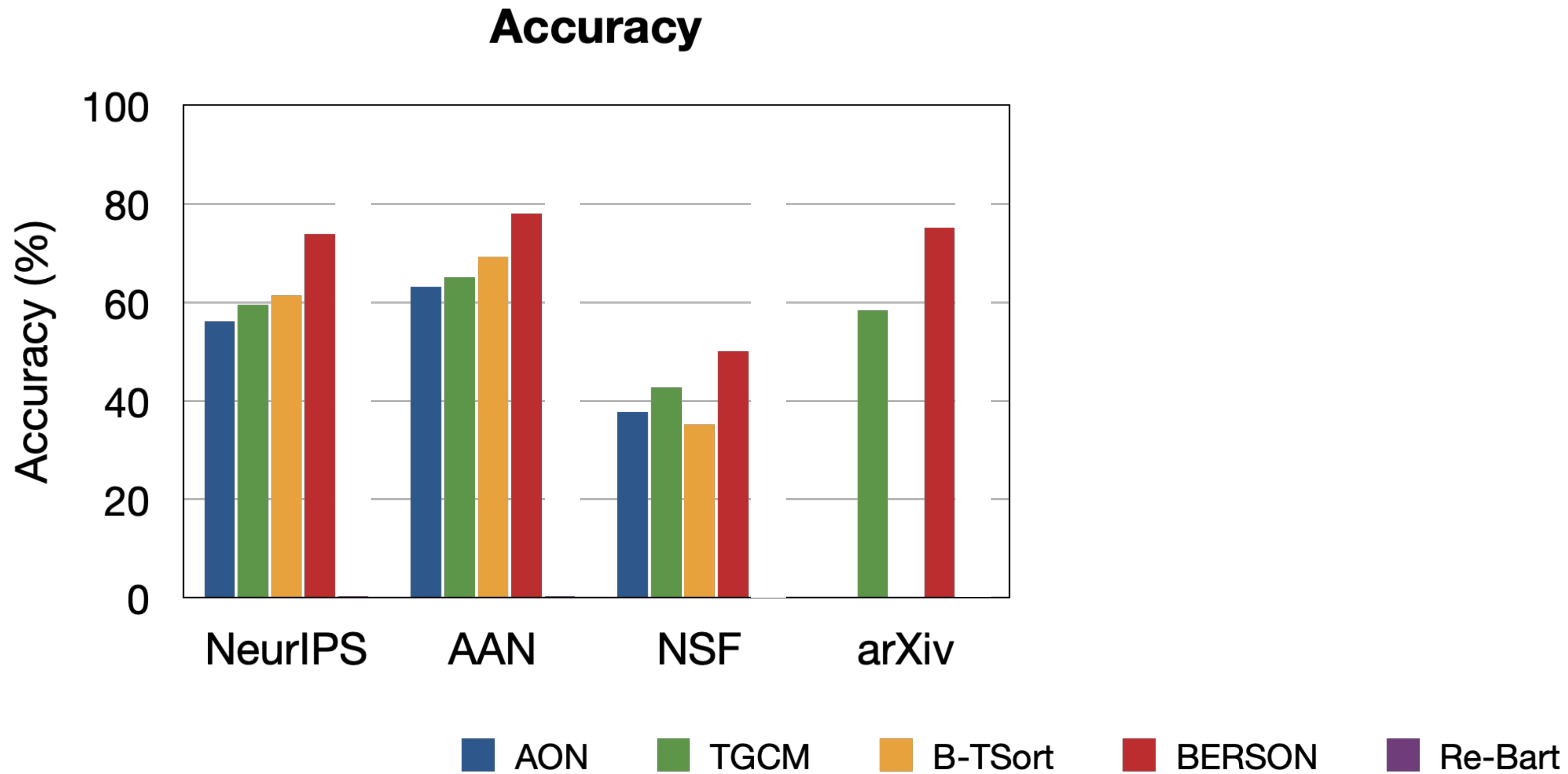
$$\tau = 1 - \frac{2 (\#invesrions)}{\binom{n}{2}}$$

Results

Paper Abstracts

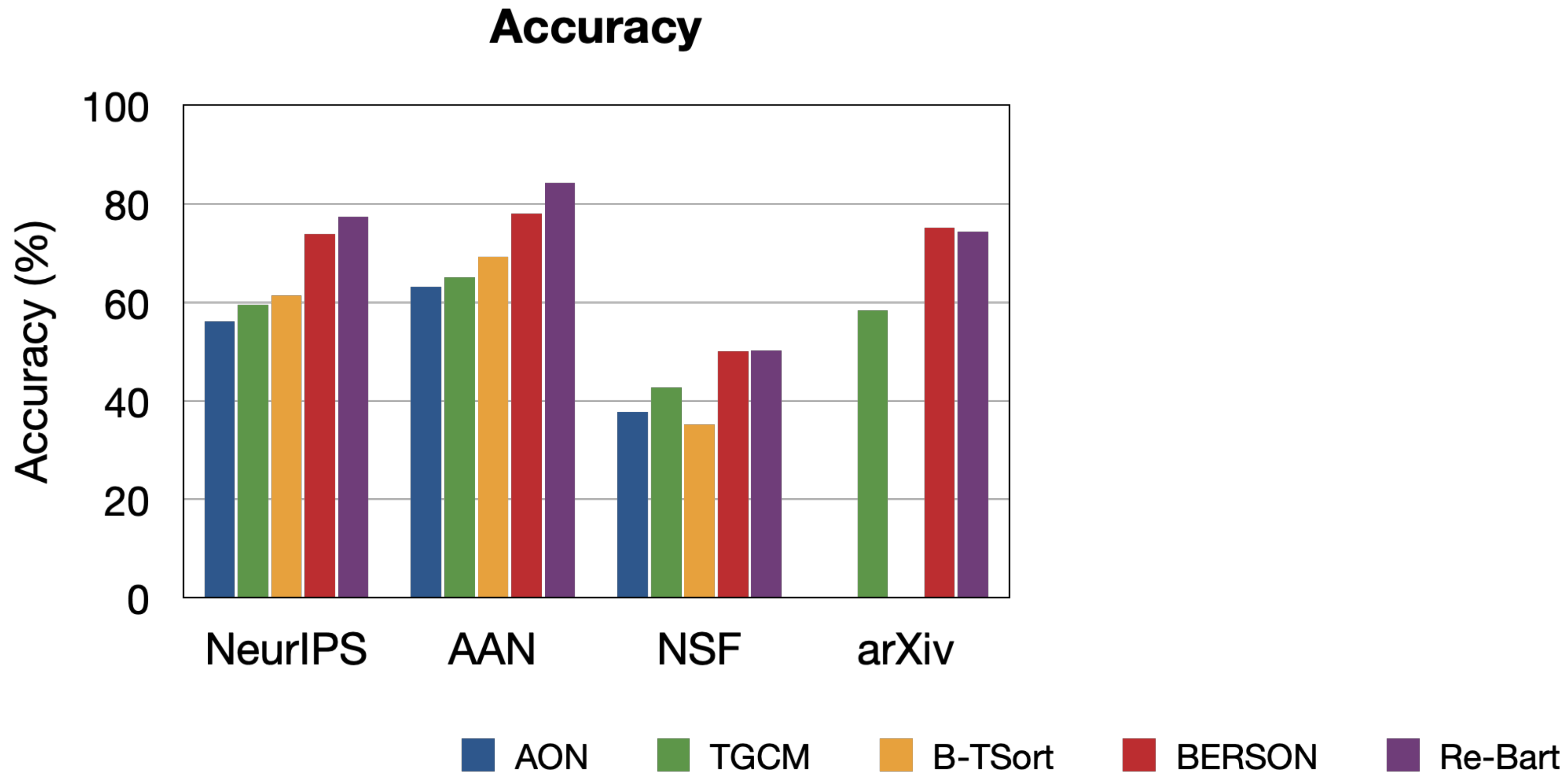
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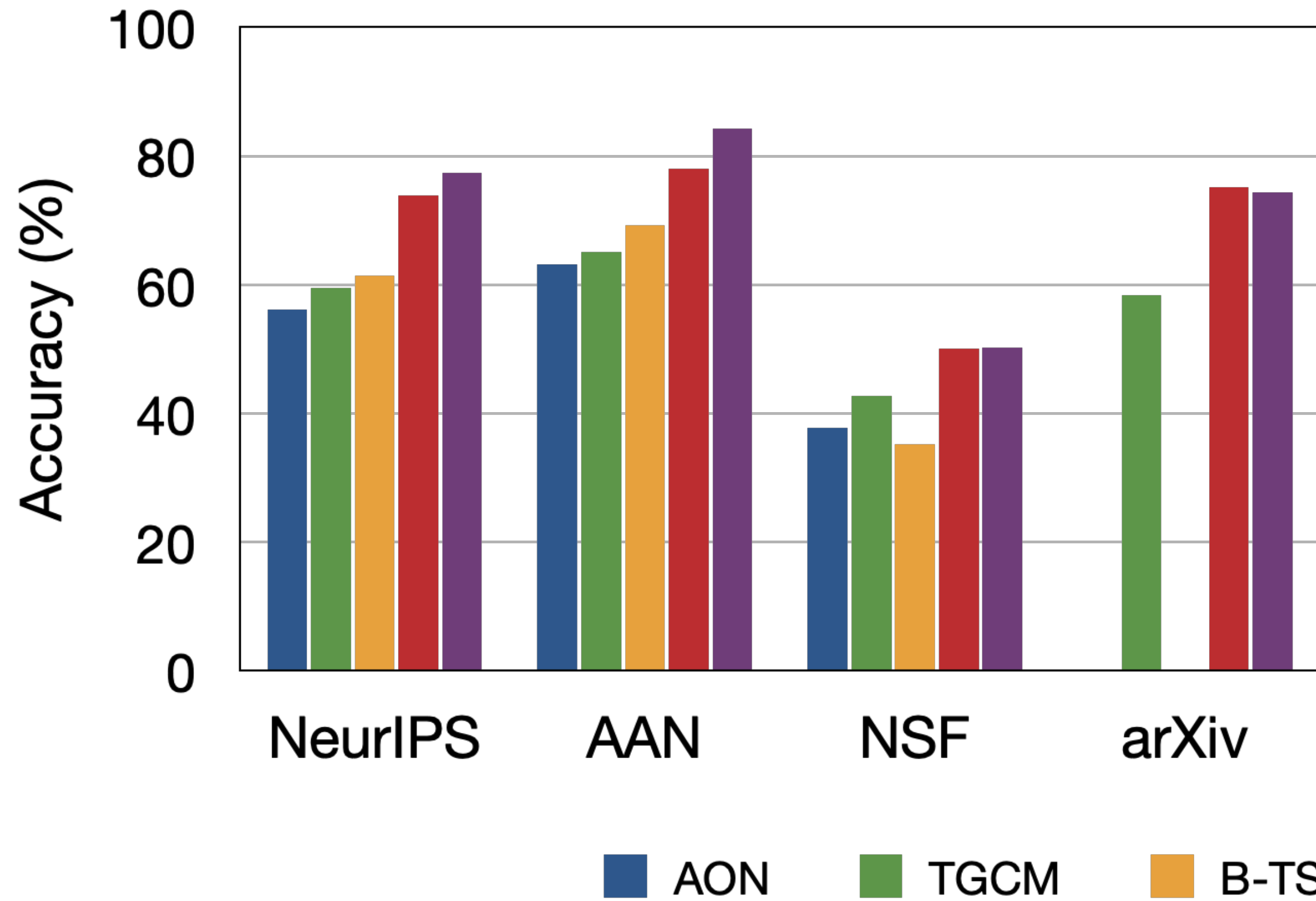
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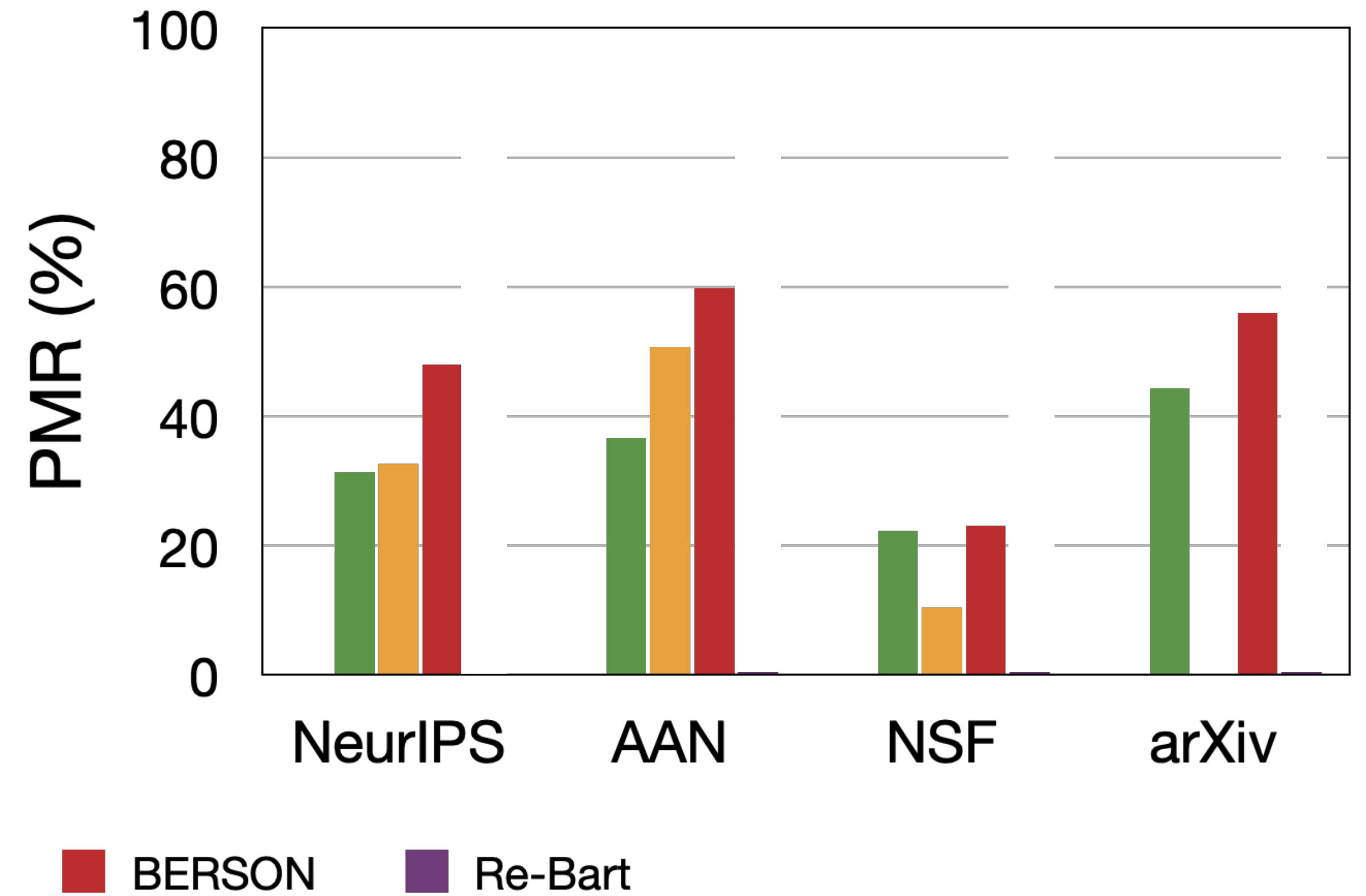
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Accuracy



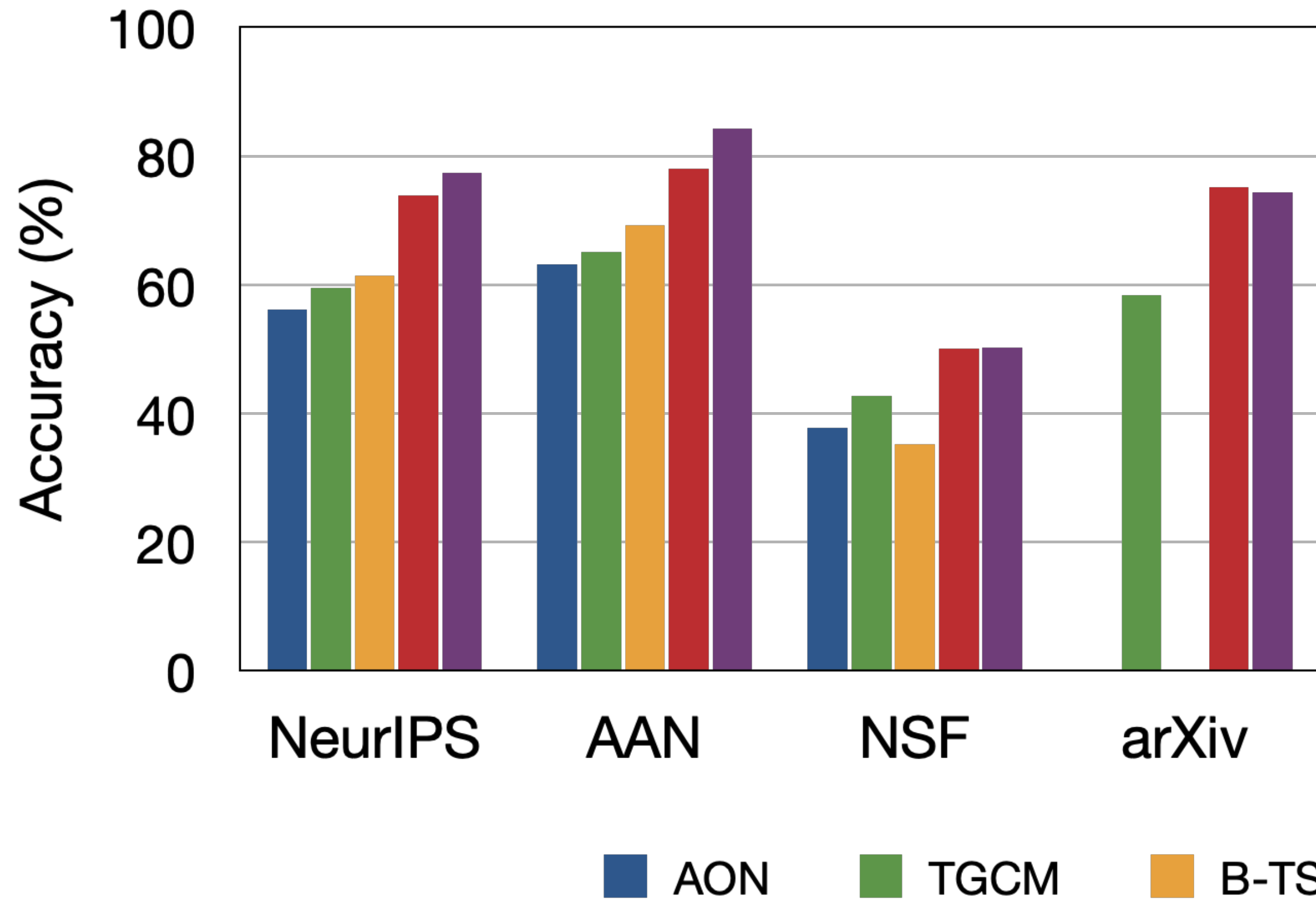
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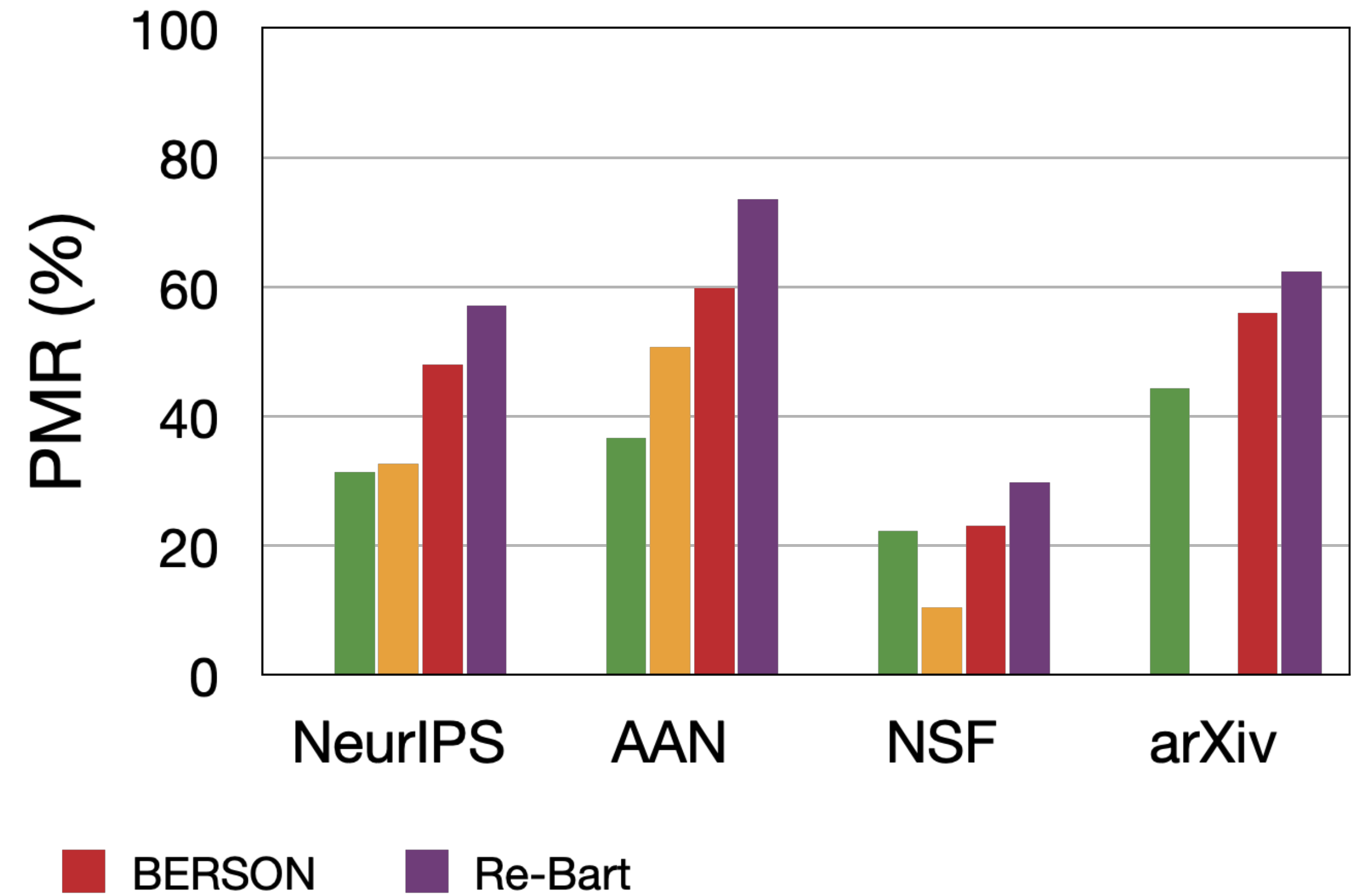
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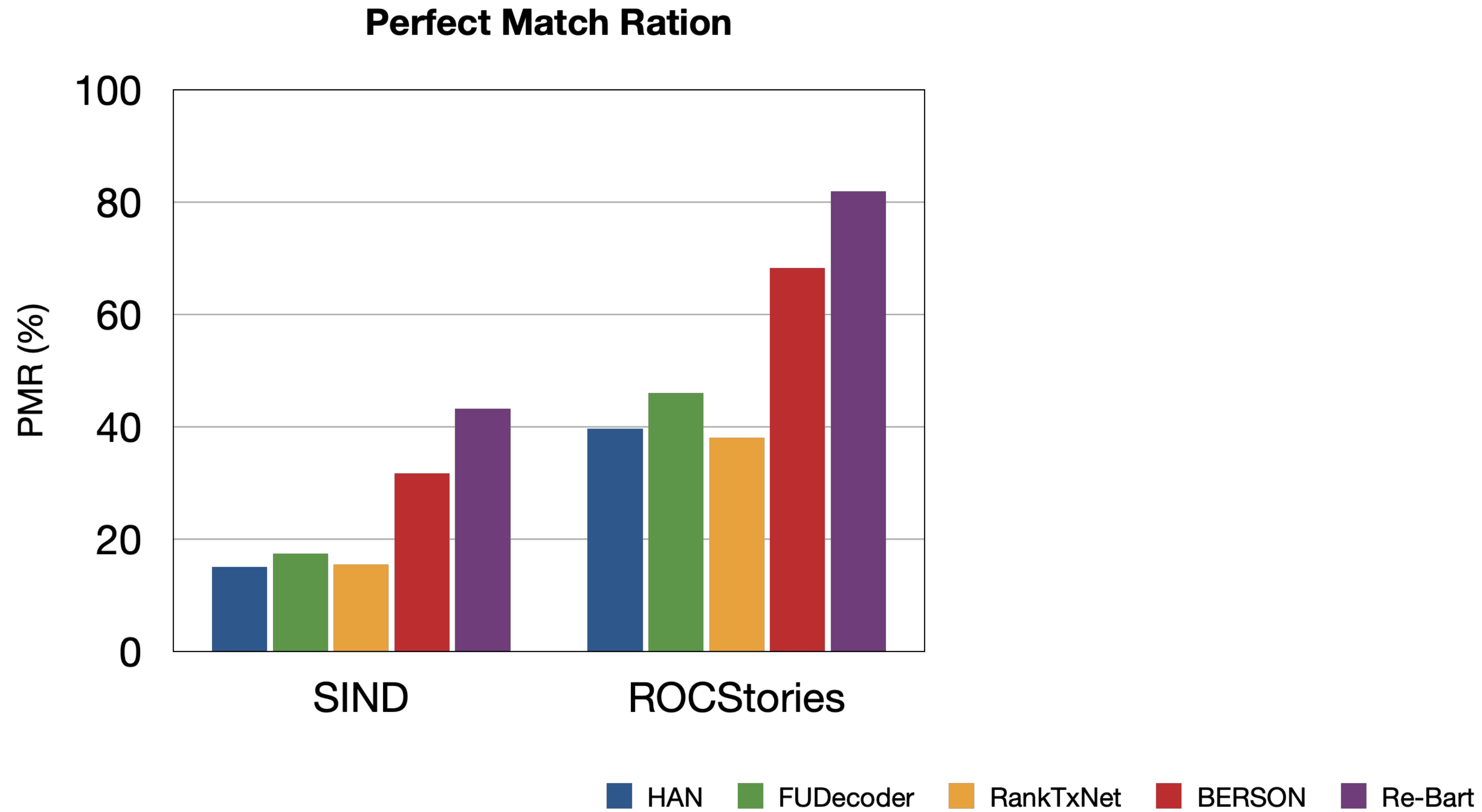


Perfect Match Ratio



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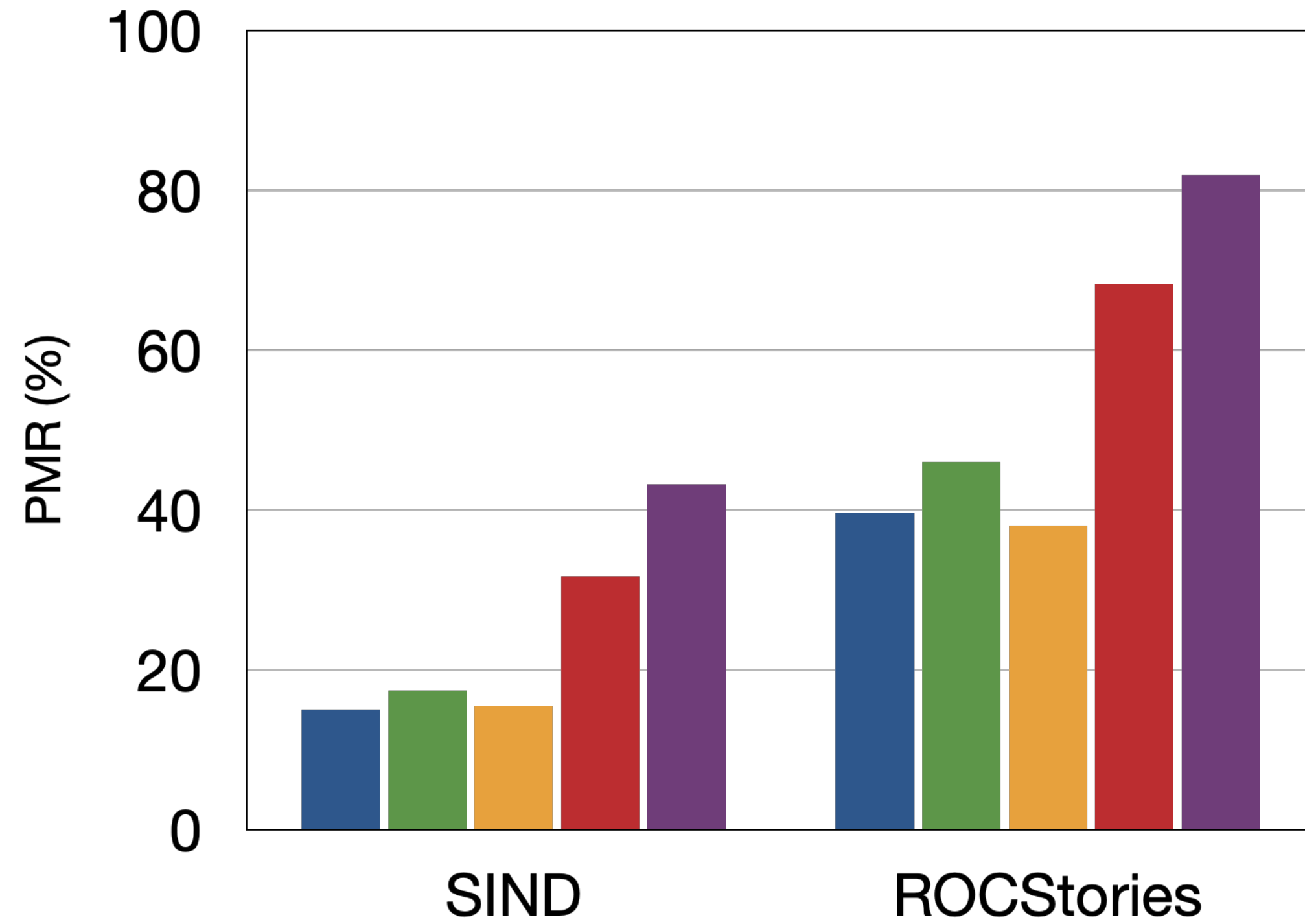
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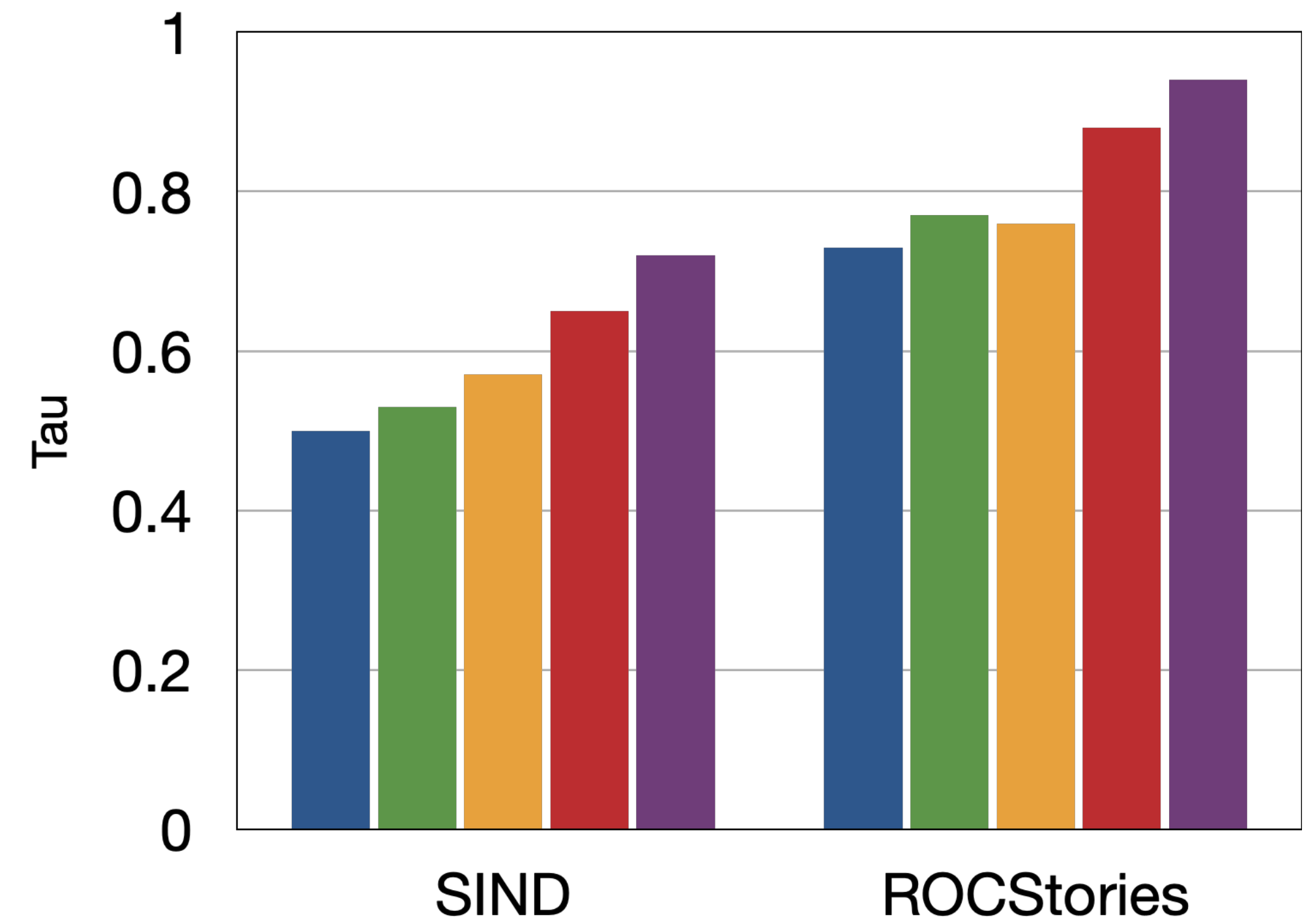
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Perfect Match Ration



Kendall's Tau



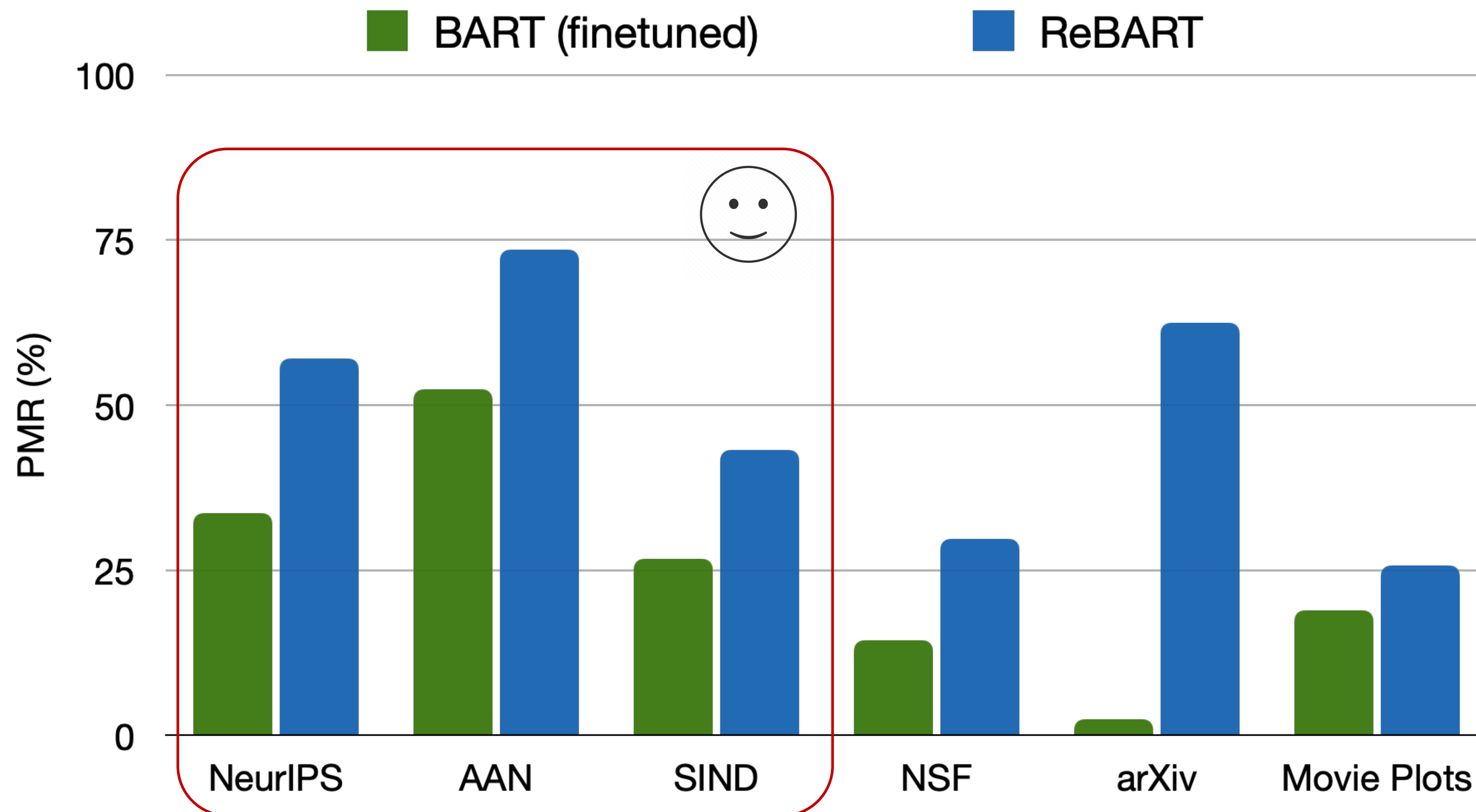
■ HAN ■ FUDecoder ■ RankTxNet ■ BERSON ■ Re-Bart

Results

How does text-to-text framework perform?

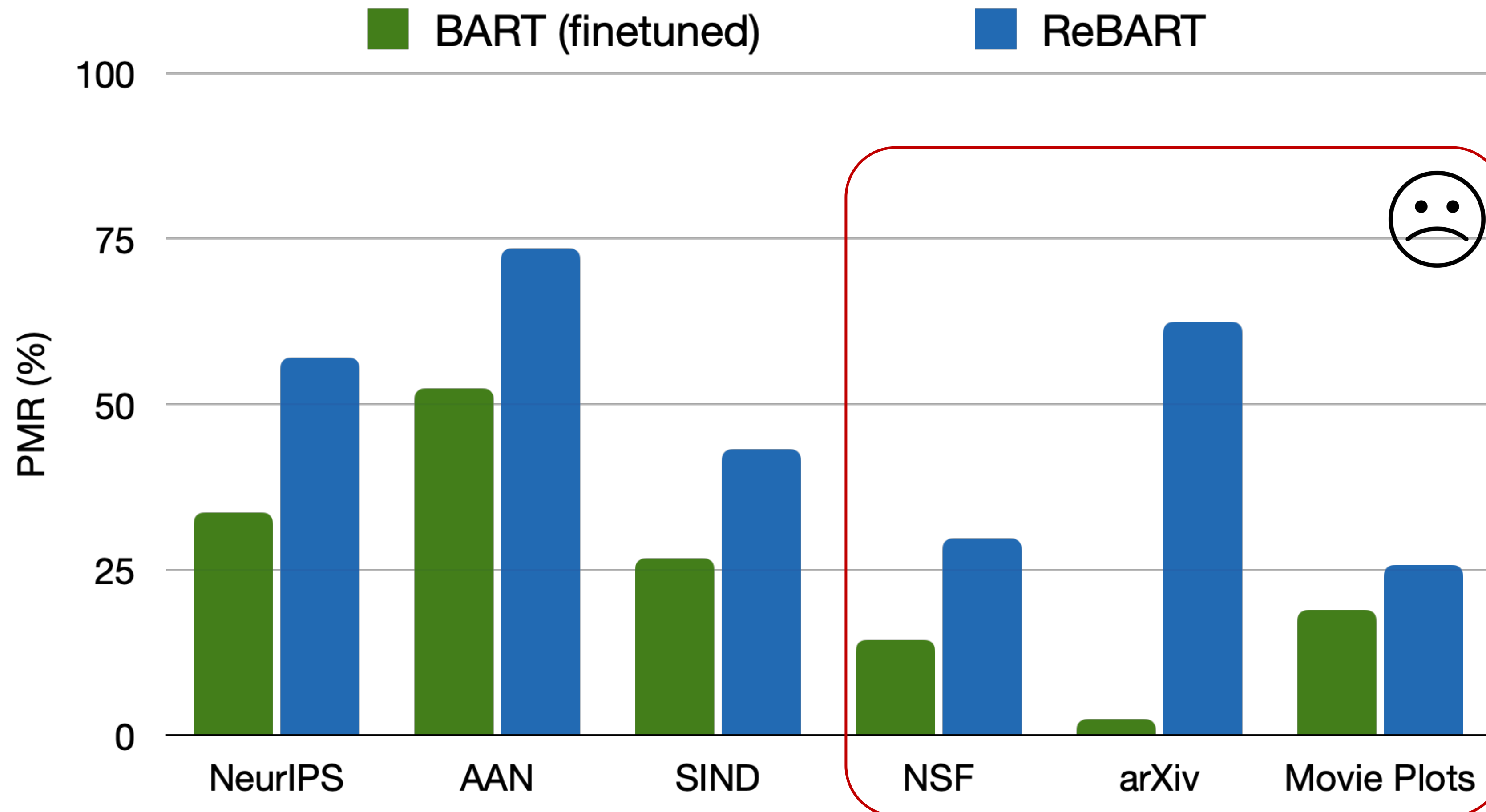
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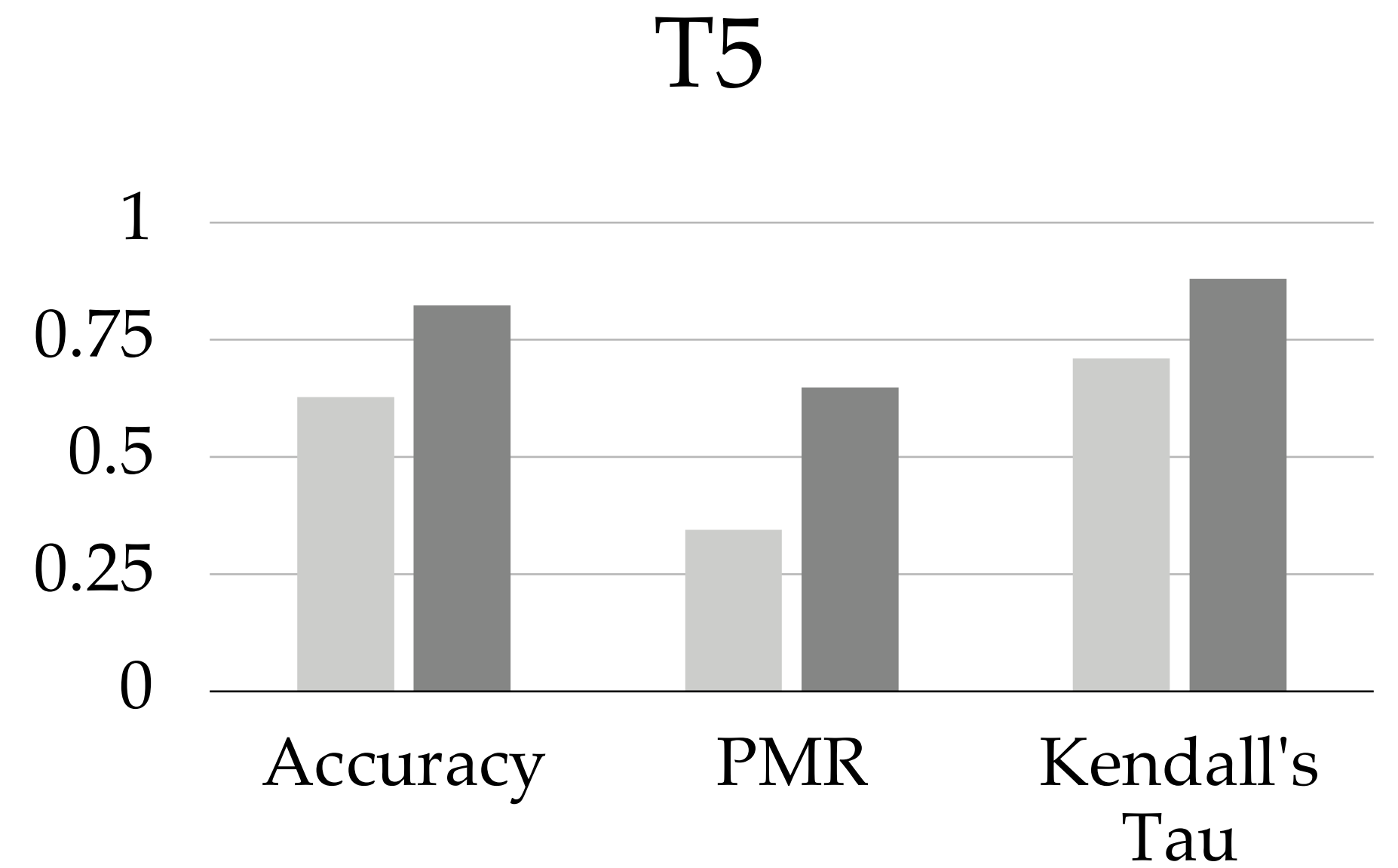
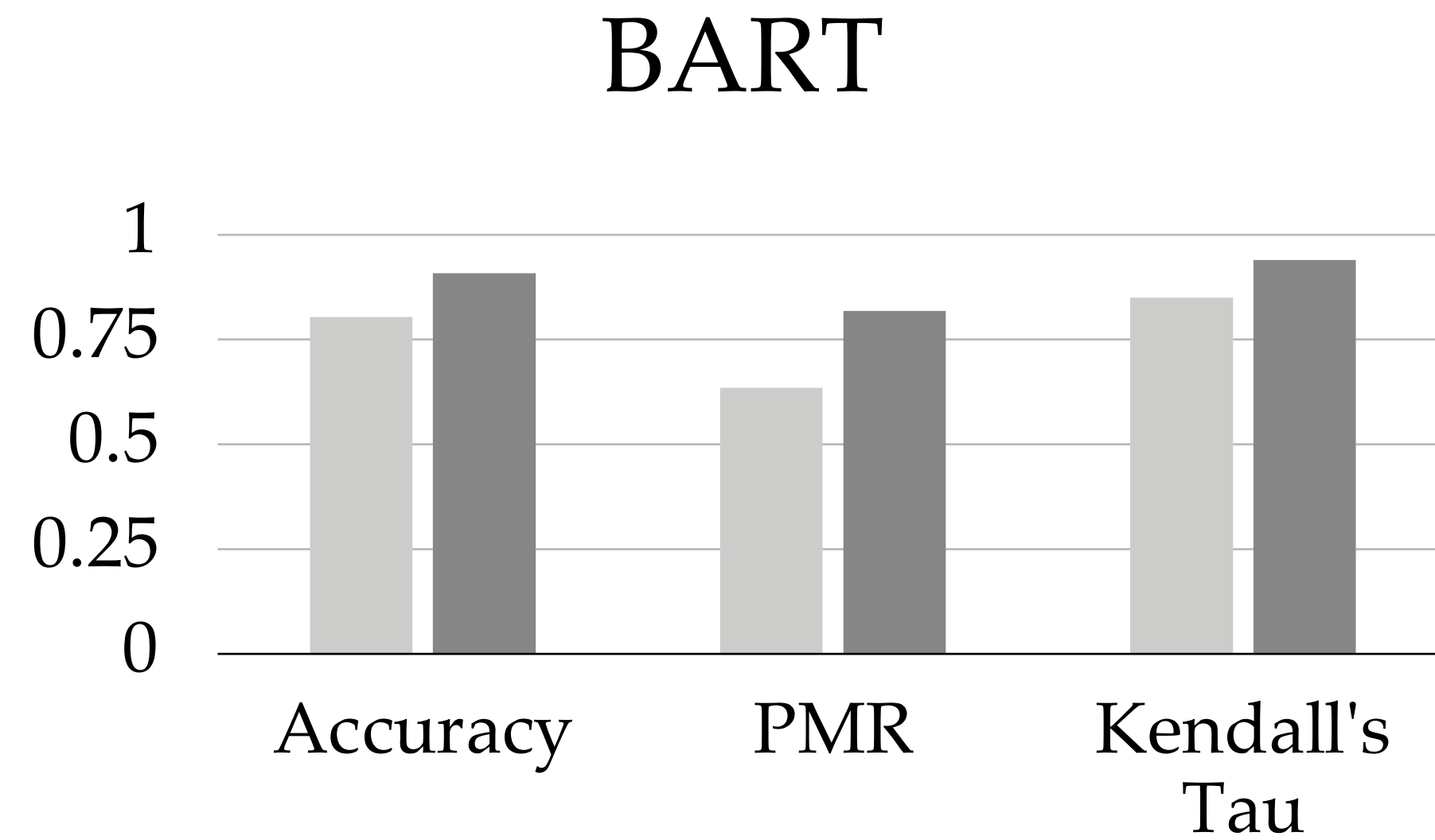


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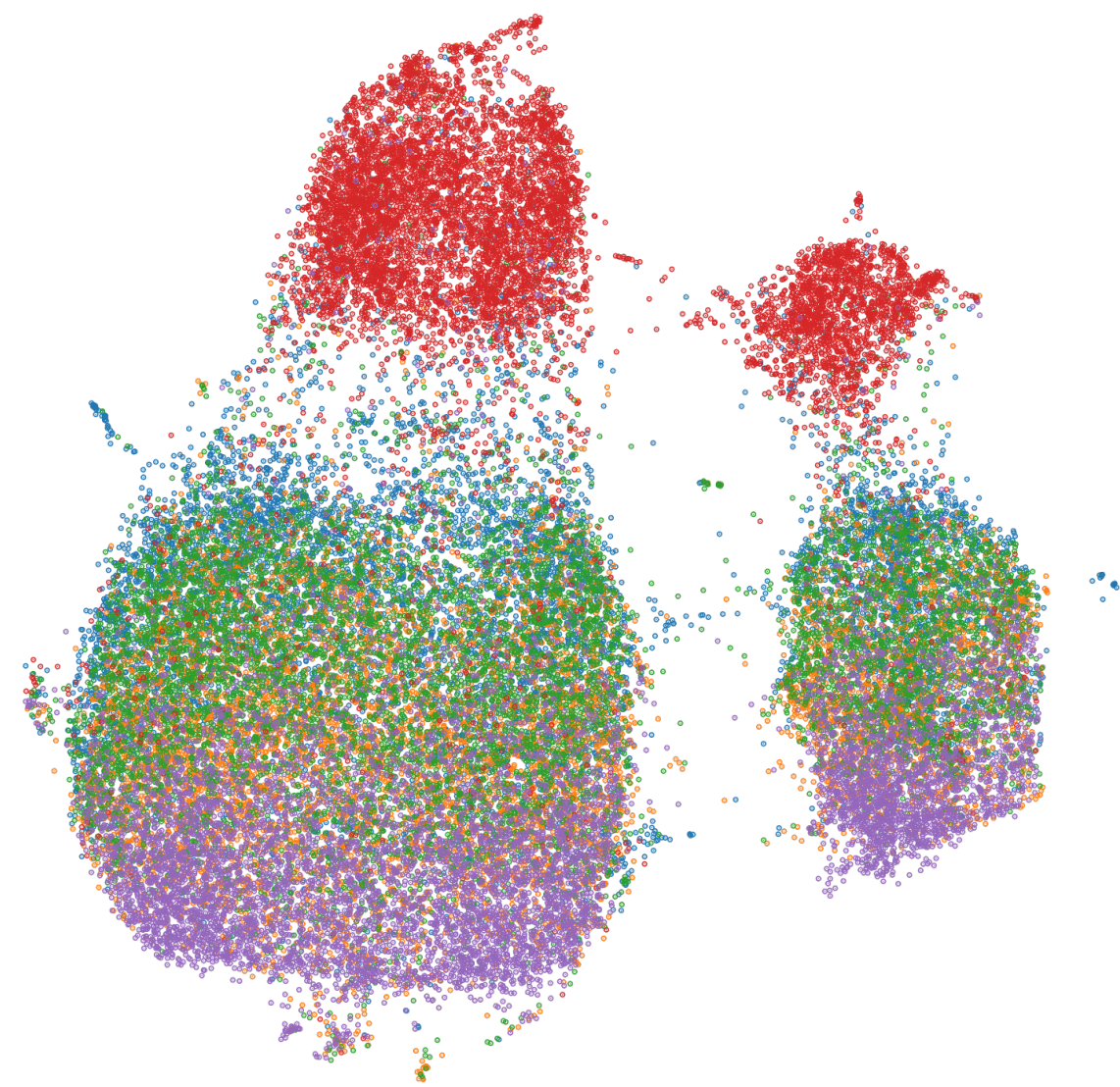


Results - BART vs T5

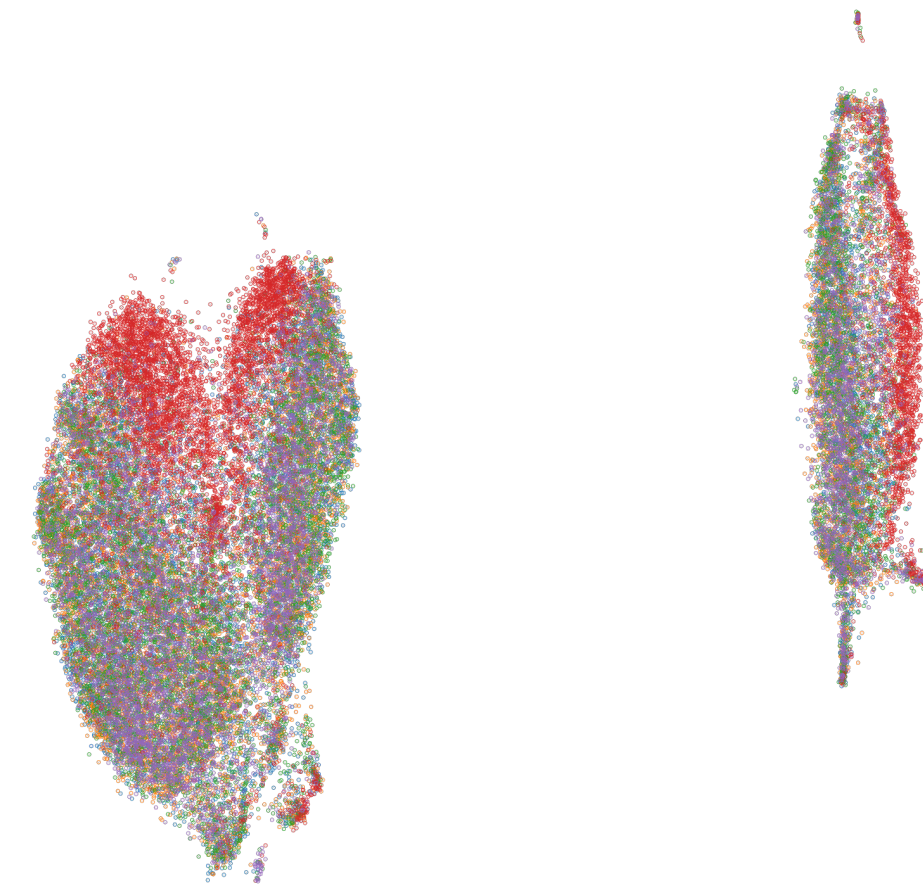


■ text-to-text ■ text-to-marker

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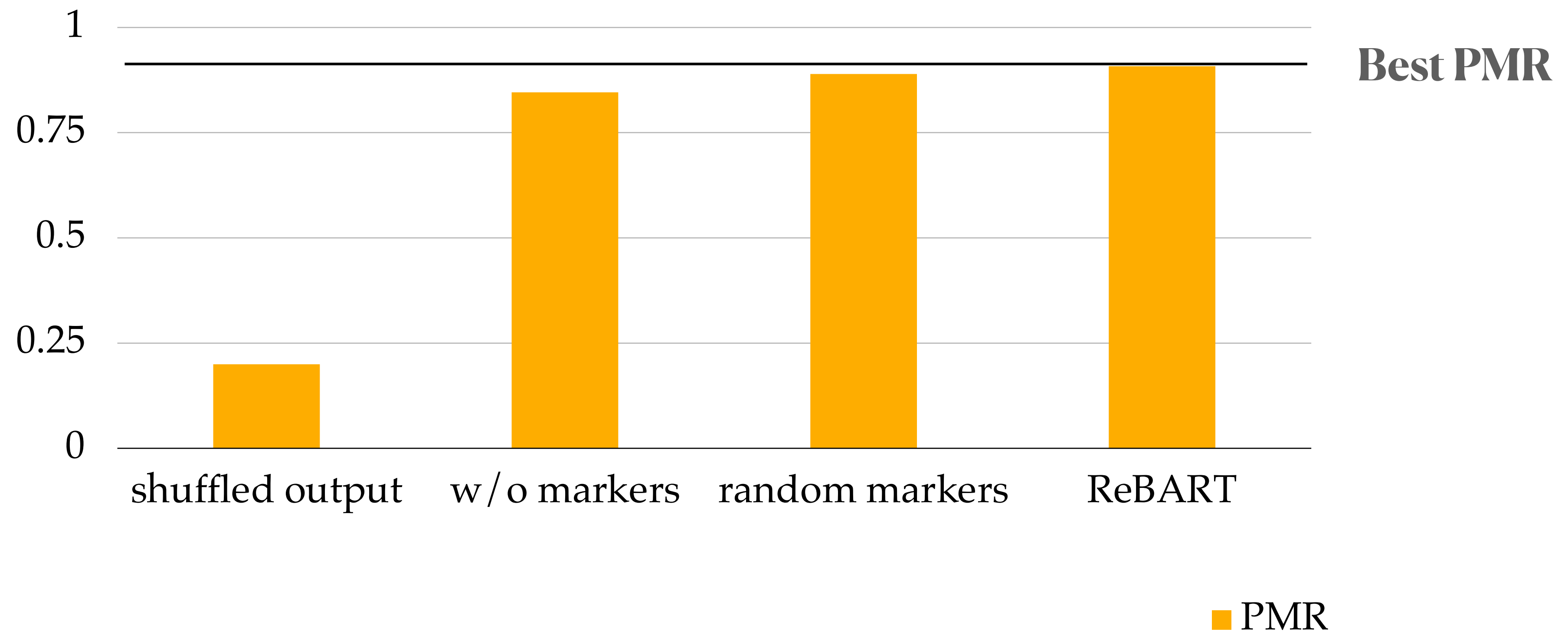


BART embeddings



T5 embeddings

Ablations



Analysis - Effect of Shuffling

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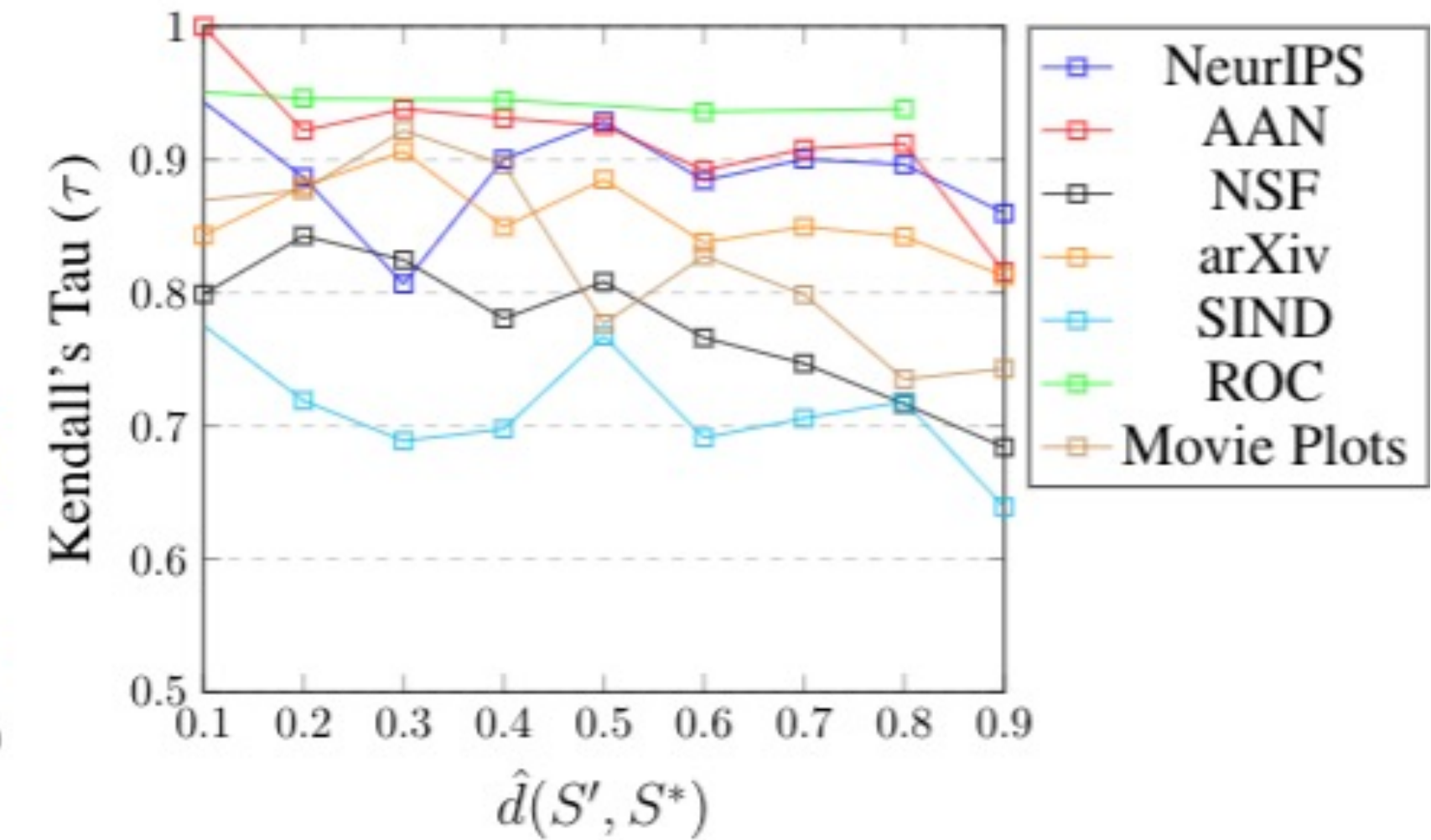
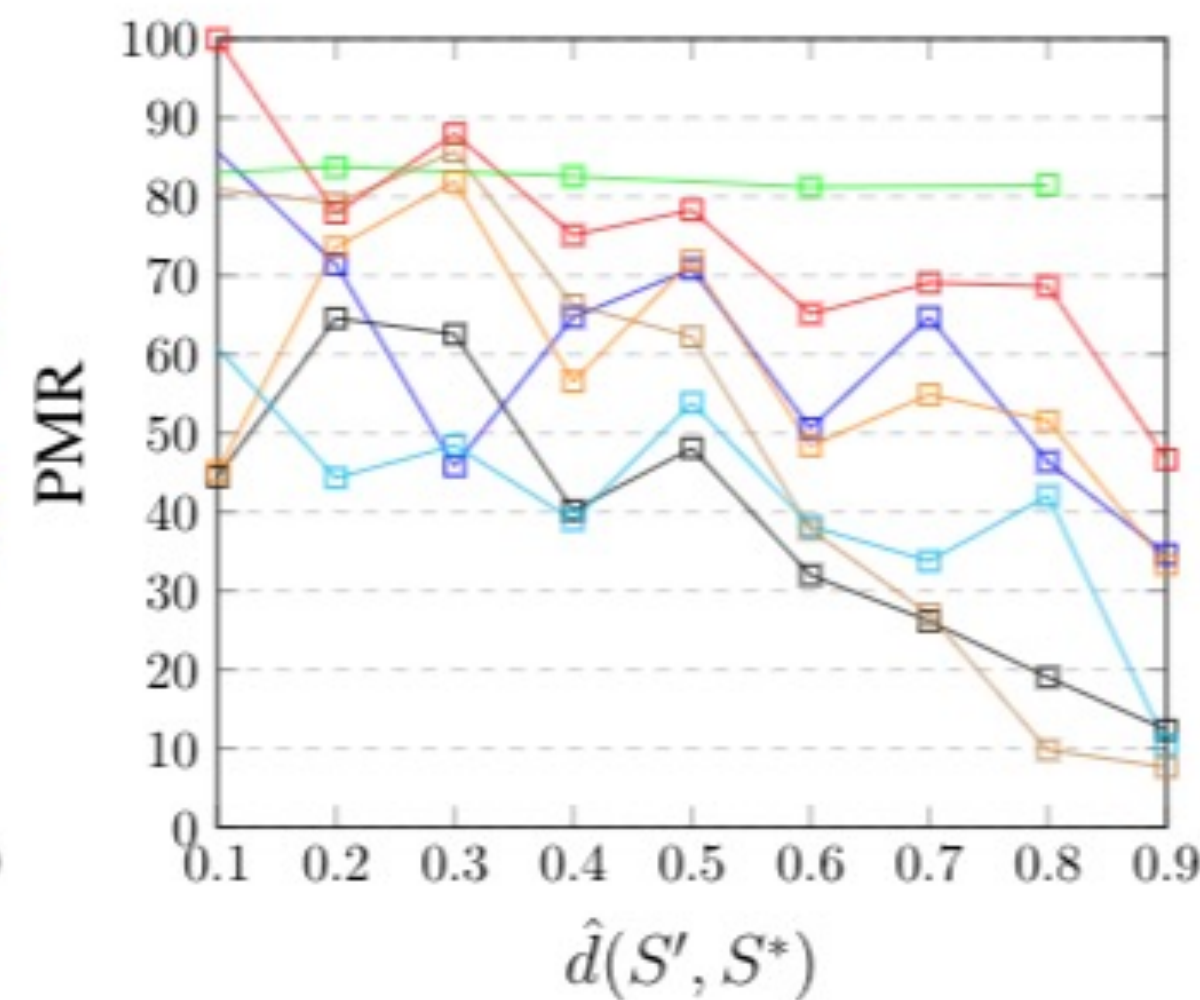
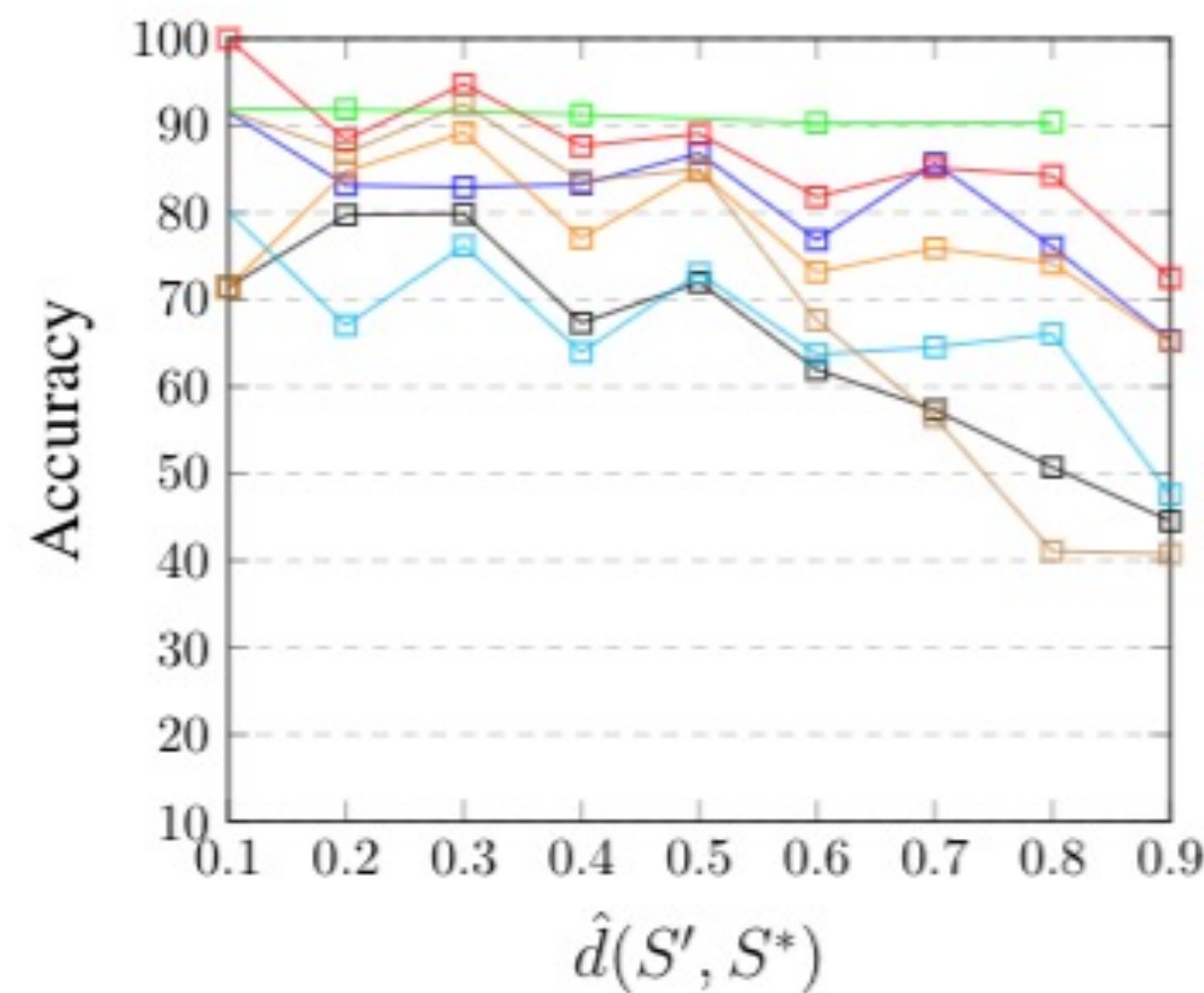
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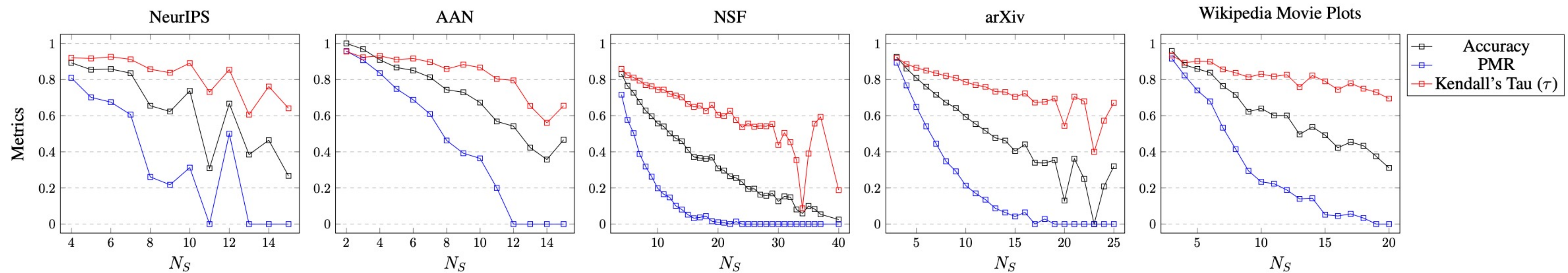
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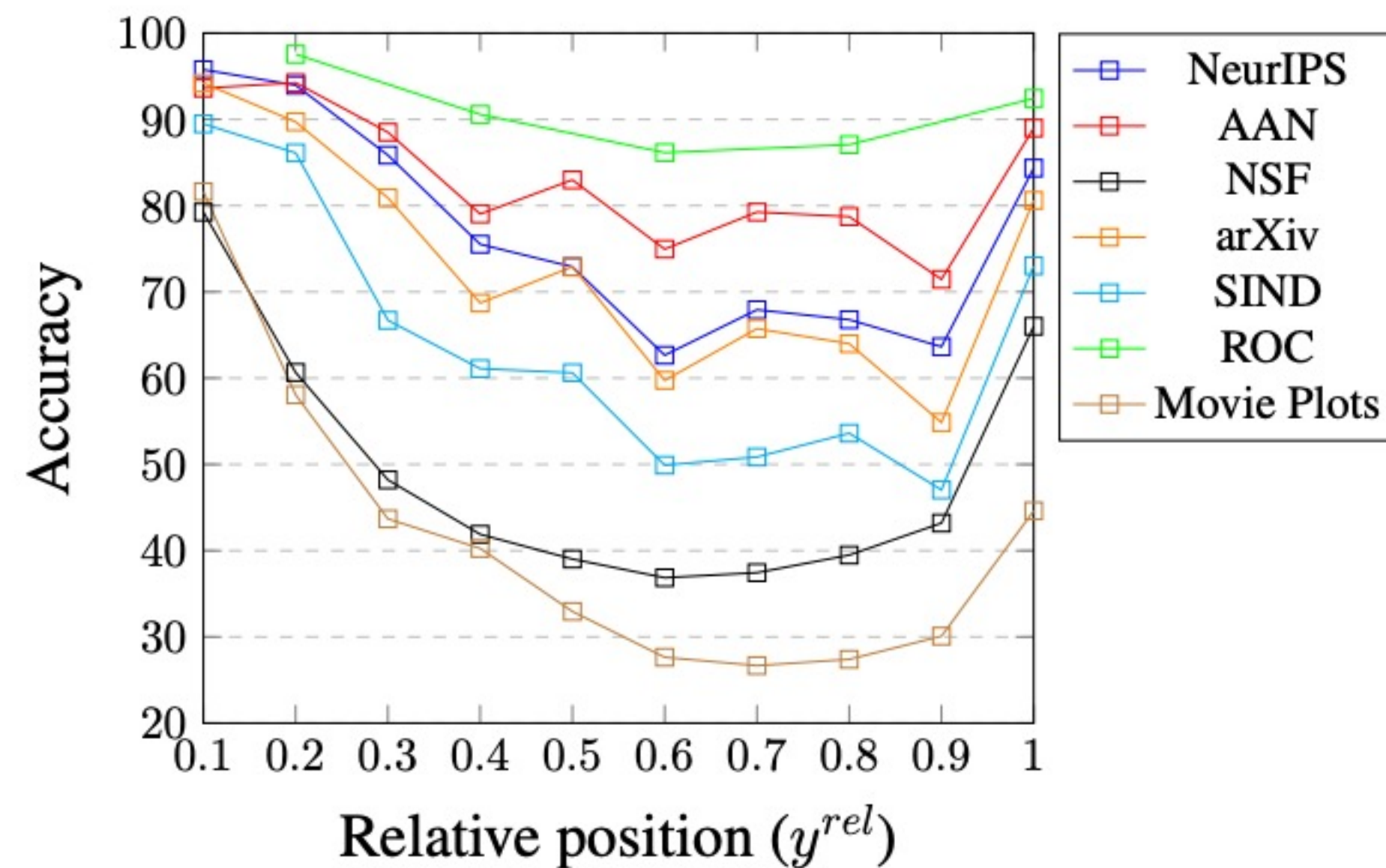
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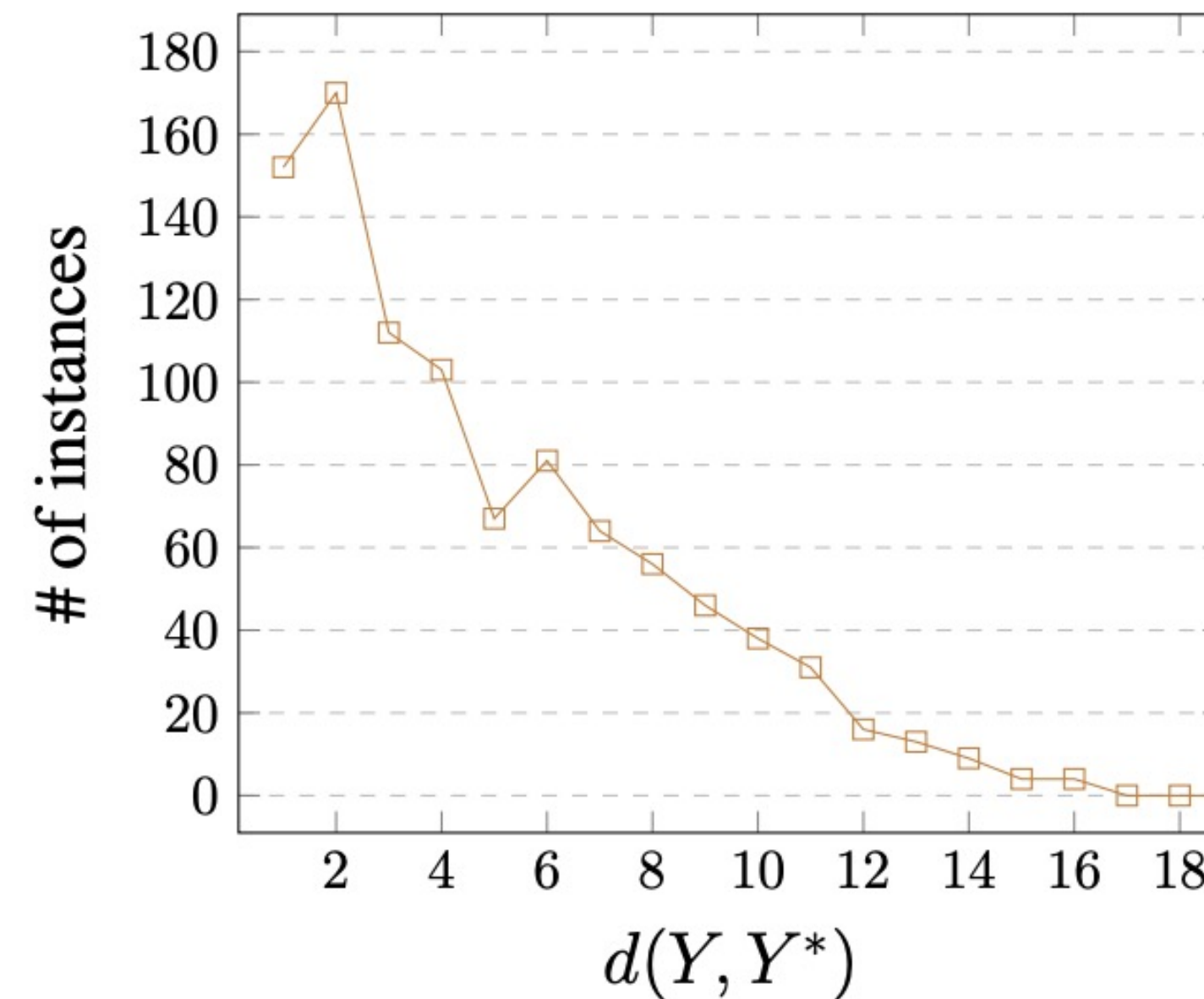
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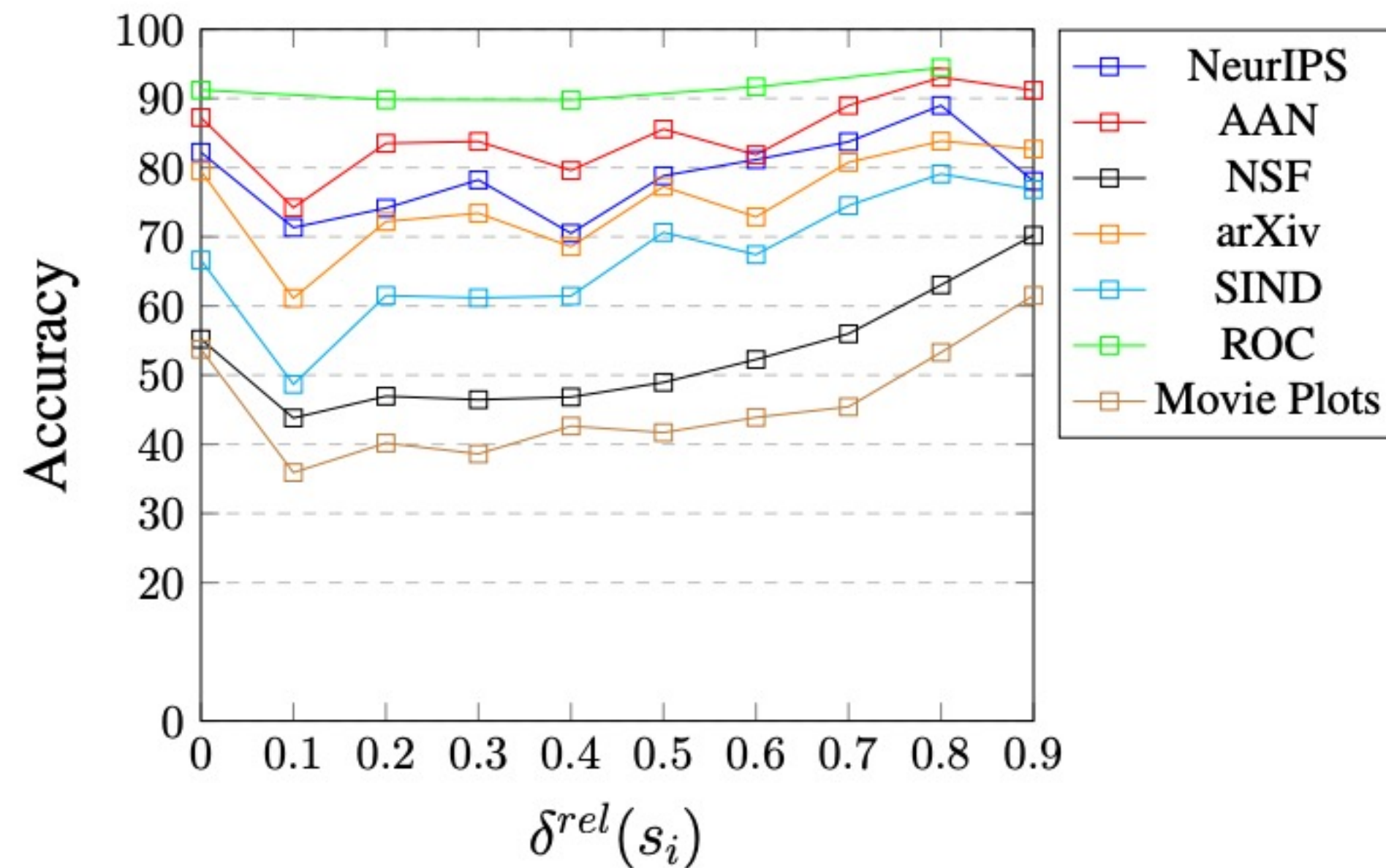
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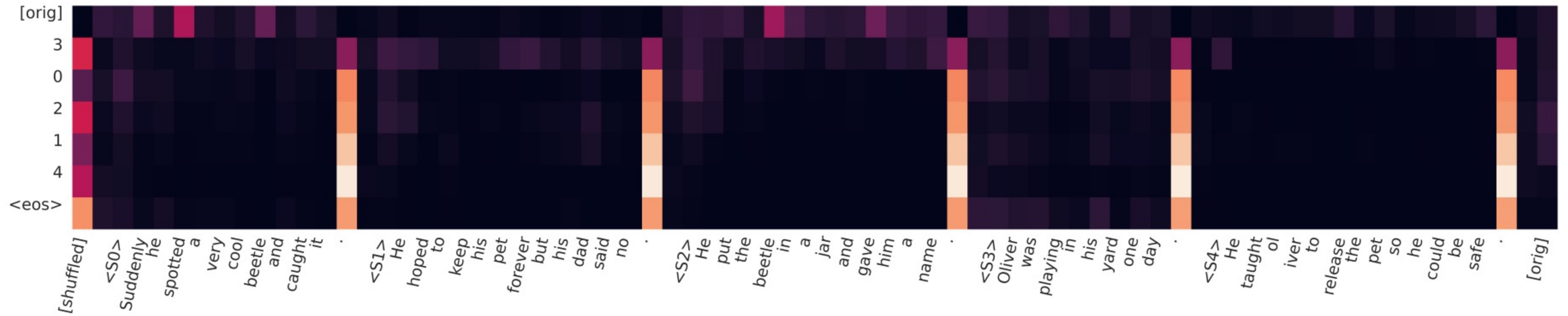
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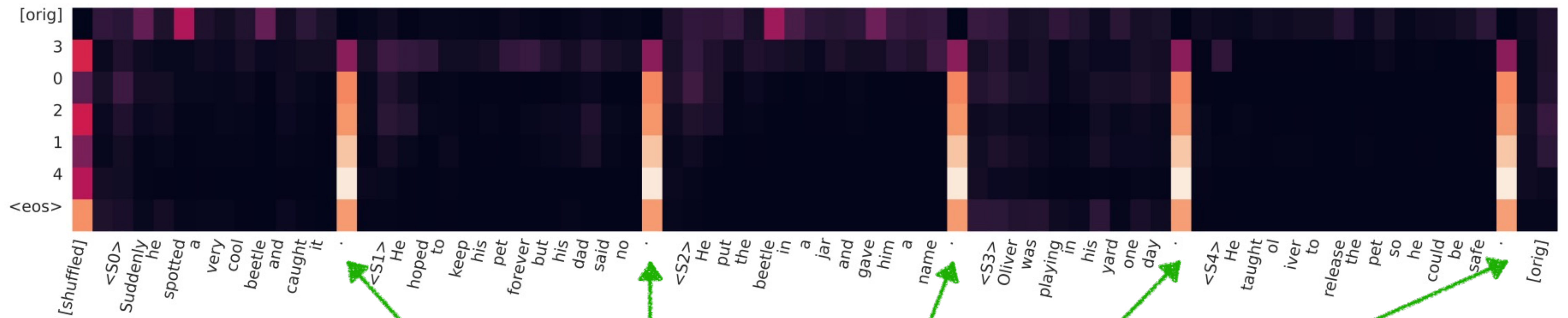


Cross-Attention Visualization

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Attention is concentrated near the sentence markers

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