

Zero-Shot Information Extraction as a Unified Text-to-Triple Translation

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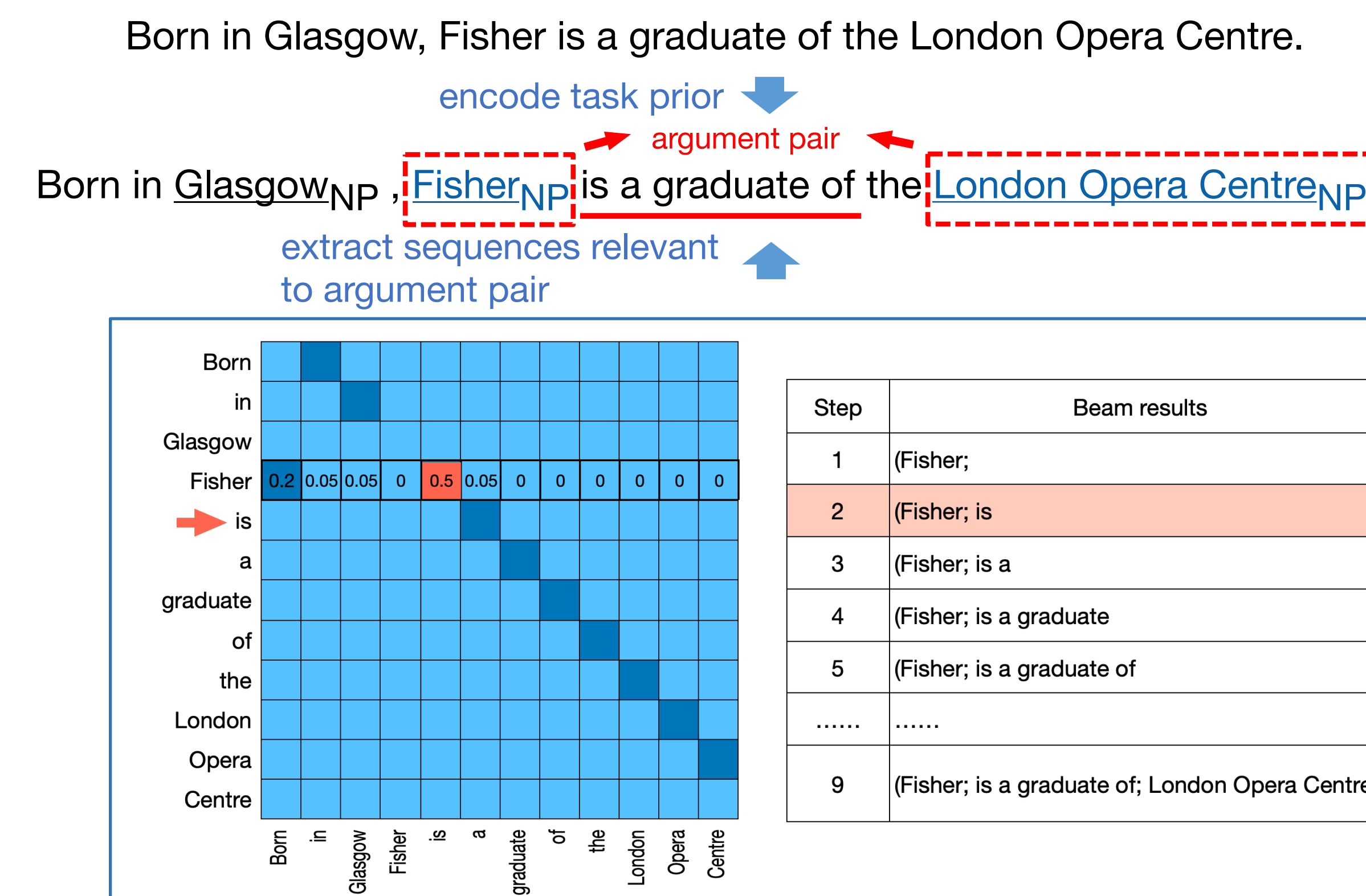


Overview

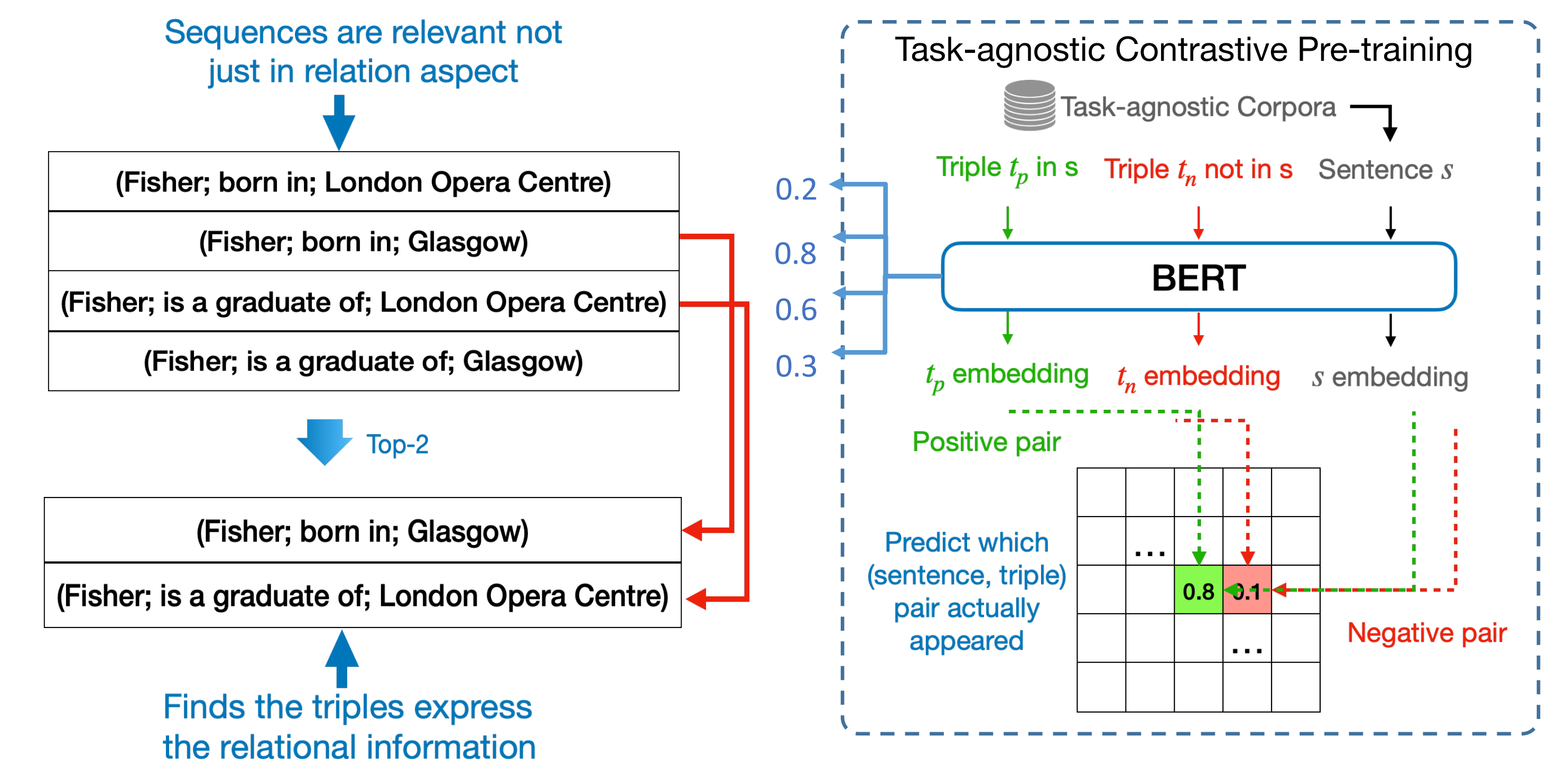
- Information extraction (IE) is crucial to many NLP applications
 - Limited transferability: there are many IE tasks with different task-specific pipelines
 - We need a unified IE approach
- Our approach: **Text-to-Triple Translation**
 - Unified framework that solves IE tasks
 - Zero-shot IE without the need of any task-specific training set
 - Generalization by transferring the latent knowledge that language models have
 - Better interpretability through the enhanced model transparency
 - State-of-the-art or competitive performance compared to fully supervised methods

Text-to-Triple Translation

- Generating: produces general information about the task via pre-trained language models



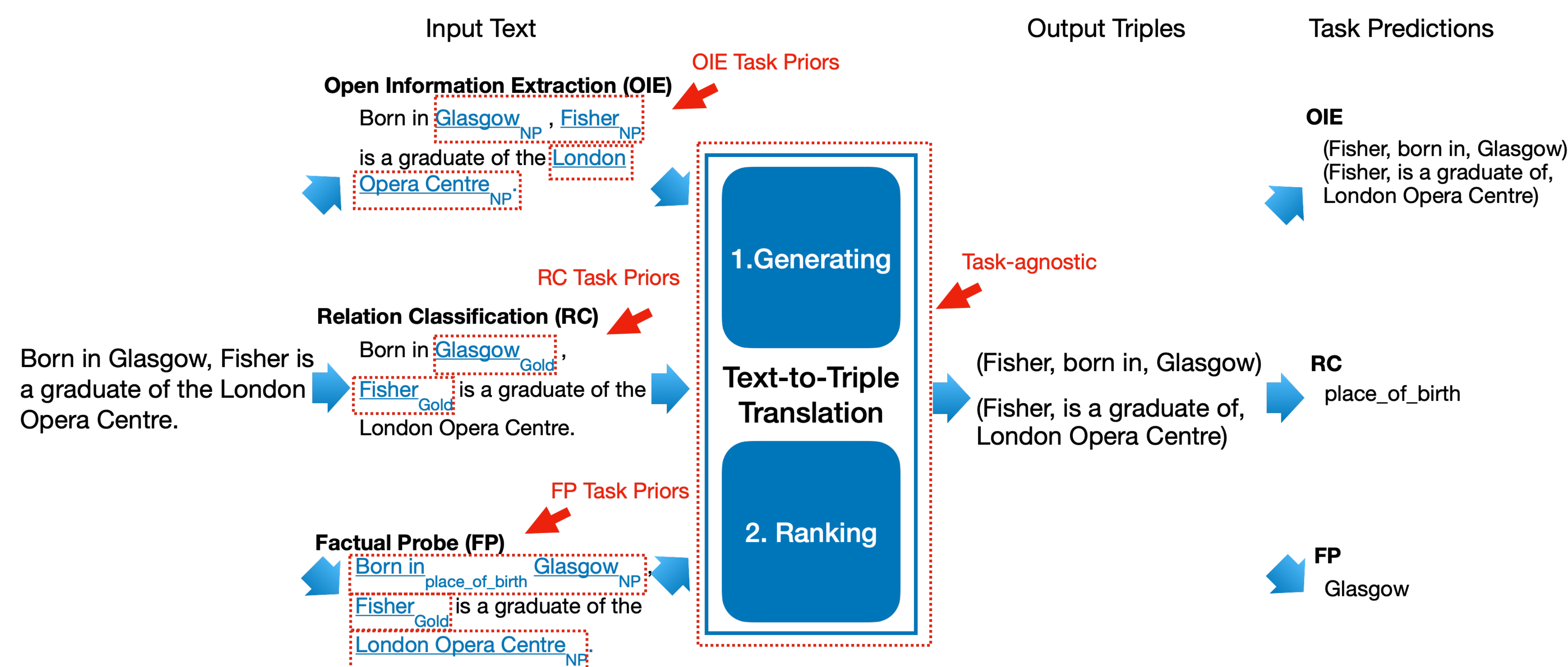
- Ranking: finds triples that are of interest to the task via a ranking model pre-trained on a task-agnostic relational corpus



By leveraging the task priors encoded in the input, we enable the zero-shot transfer of the general knowledge that a pre-trained language model has about the task

DeepEx

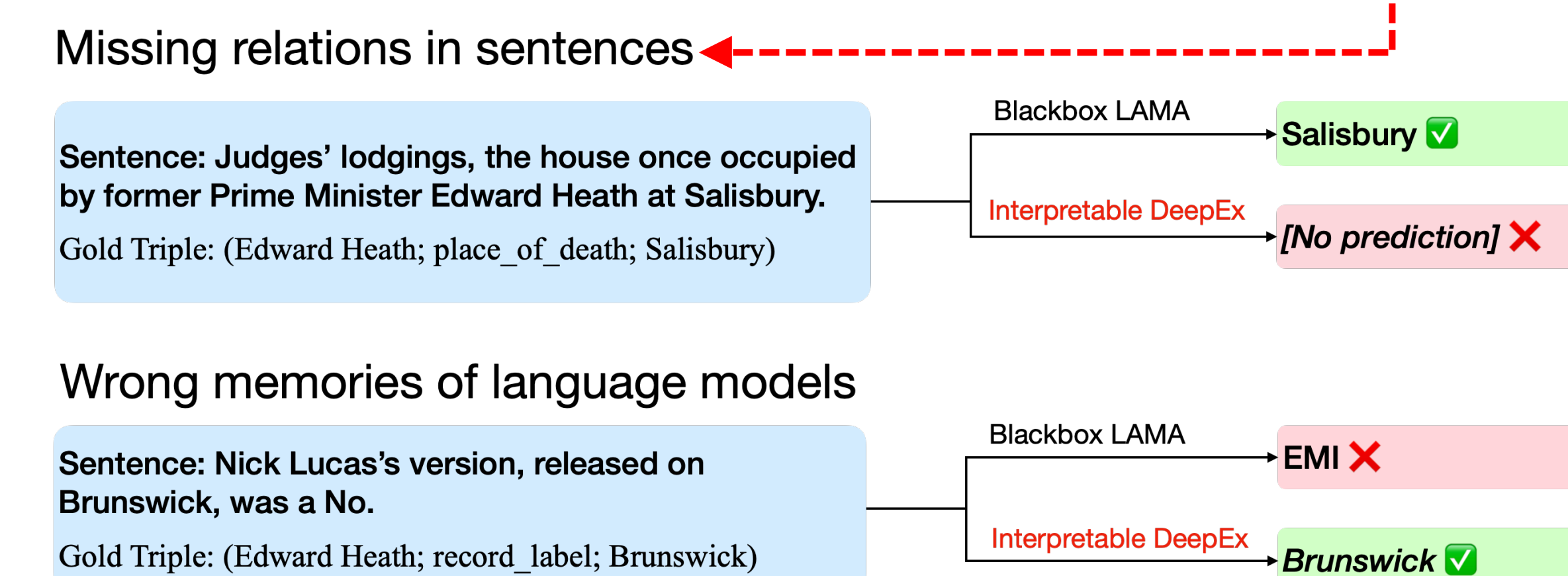
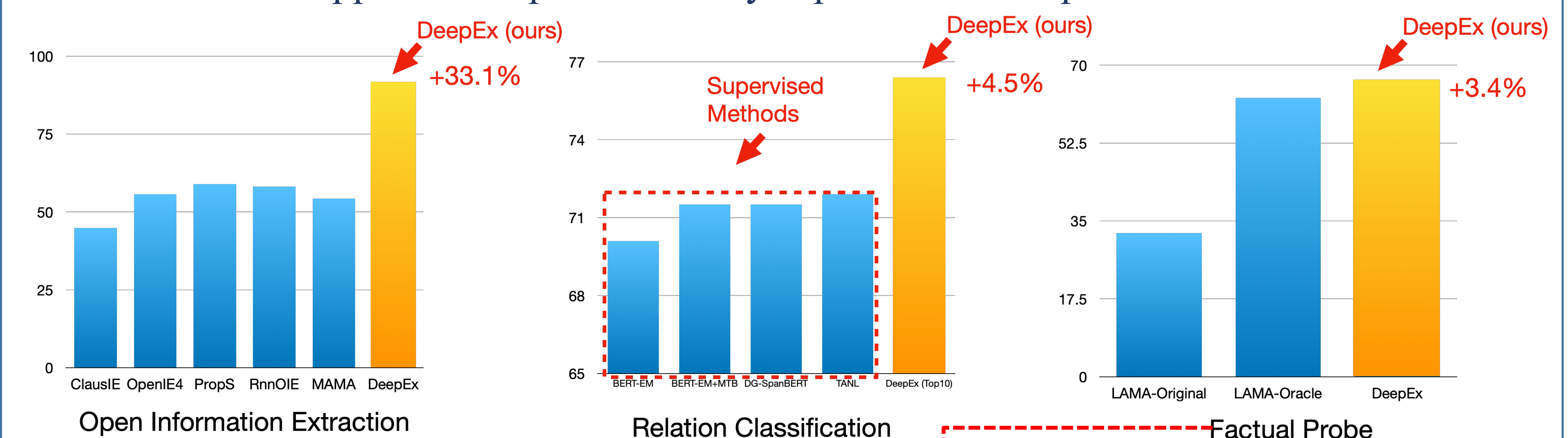
- Same text-to-triple translation is shared across tasks, the only difference is the input encoding



The basic idea: treat every information extraction problem as a “text-to-triple” problem, i.e., translating input text to output triples, then decoding into task predictions

Results

- Our unified approach achieves state-of-the-art or competitive results on all three tasks
- Our zero-shot approach outperforms fully supervised task-specific models



Get our code!

- Our method delivers more interpretable results thanks to the enhanced transparency