

AMPERE: AMR-Aware Prefix for Generation-Based Event Argument Extraction Model





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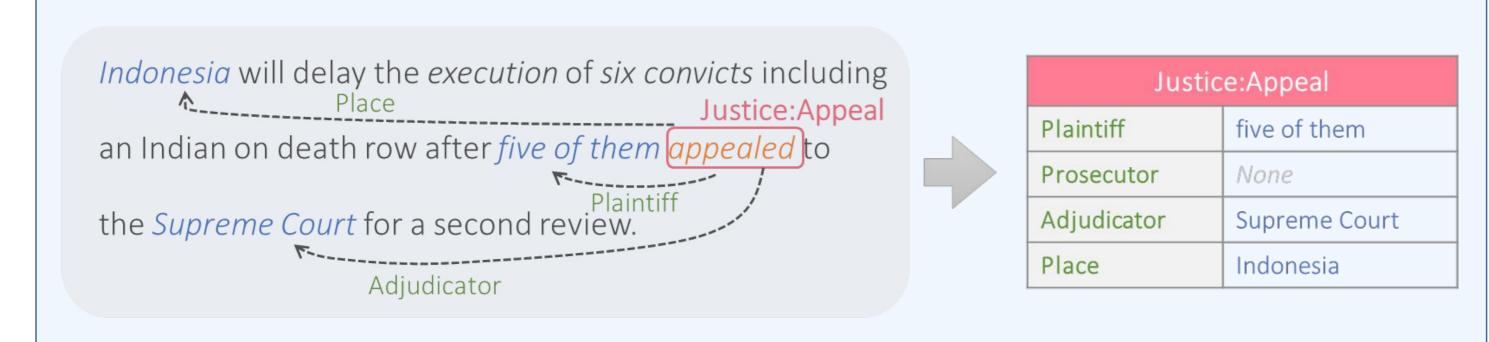
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Event argument extraction (EAE)

Goal: extract the participants (arguments) for an event described in the given sentence.

- Input: sentence, event trigger and the event type
- Output: (arguments, role) pairs

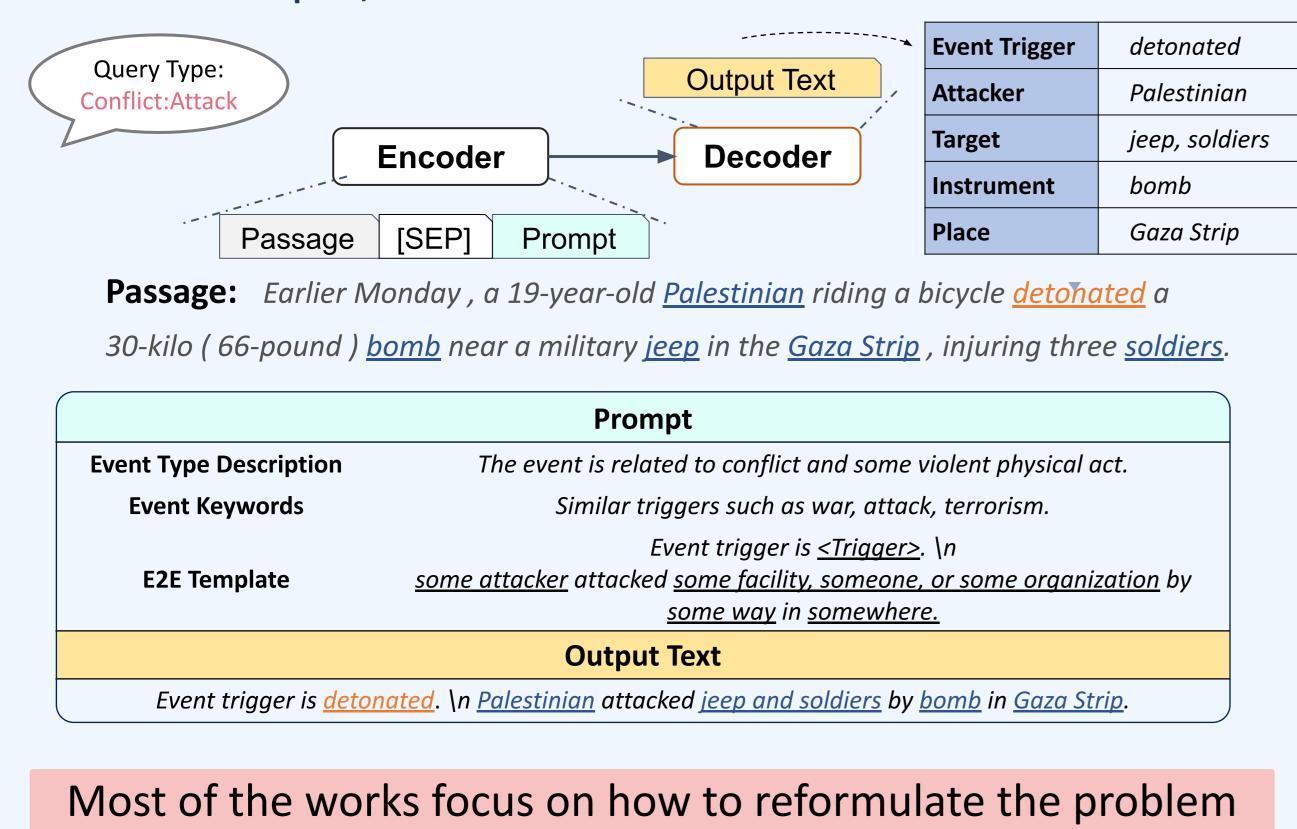


Generation-based EAE models

EAE is typically solved by classification-based models.

- Focus on improving token representations

 More and more generation-based models for EAE are proposed.
 - Leverage large language models' power
 - TANL [Paolini+ 2021], Text2Event[Li+ 2021], DEGREE [Hsu+ 2022], etc.
 - For example, DEGREE:



How to enhance generative models using auxiliary information?

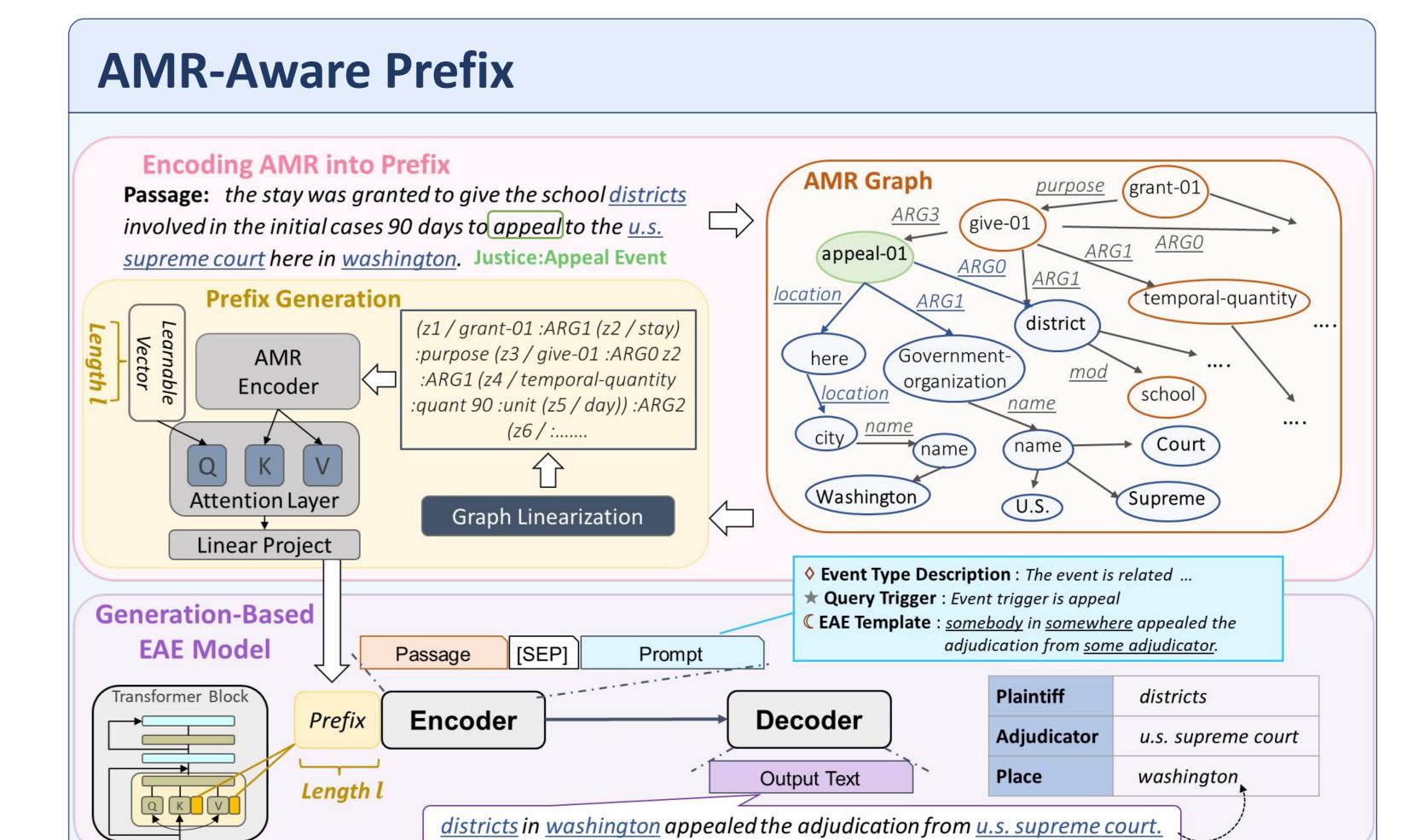
and how to prompt the generative models!

Several auxiliary information, such as POS taggings, dependency parsing, Abstract meaning representations (AMR), is helpful for EAE.

• Classification-based EAE methods incorporate such information by enriching the features.

Challenges -- Can we do better than prompting?

- Lengthy for linearized AMR
- AMR information is different from input passage in nature.
 - Containing words with special meaning
 - Graphical structure



More details

- 1. Adapt DEGREE as the generation-based EAE model.
- 2. Two variations of AMR Encoders
- AMPERE(AMRBART): Use the encoder of AMRBART, which is the current SOTA AMR-to-Text model.
- AMPERE(RoBERTa): RoBERTa-large as the AMR encoder. Add AMR-specific tokens as special tokens.
- 3. Training:
- Copy mechanism:

 $P(y_{i} = t | y_{<i}, x_{1}, ..., x_{m}) =$ $w_{gen}^{i} P_{gen}(y_{i} = t | y_{<i}, x_{1}, ..., x_{m})) +$ $(1 - w_{gen}^{i}) (\sum_{j=0}^{m} P_{copy}^{i}(j | y_{<i}, x_{1}, ..., x_{m}) \times \mathbb{1}(x_{j} = t))$

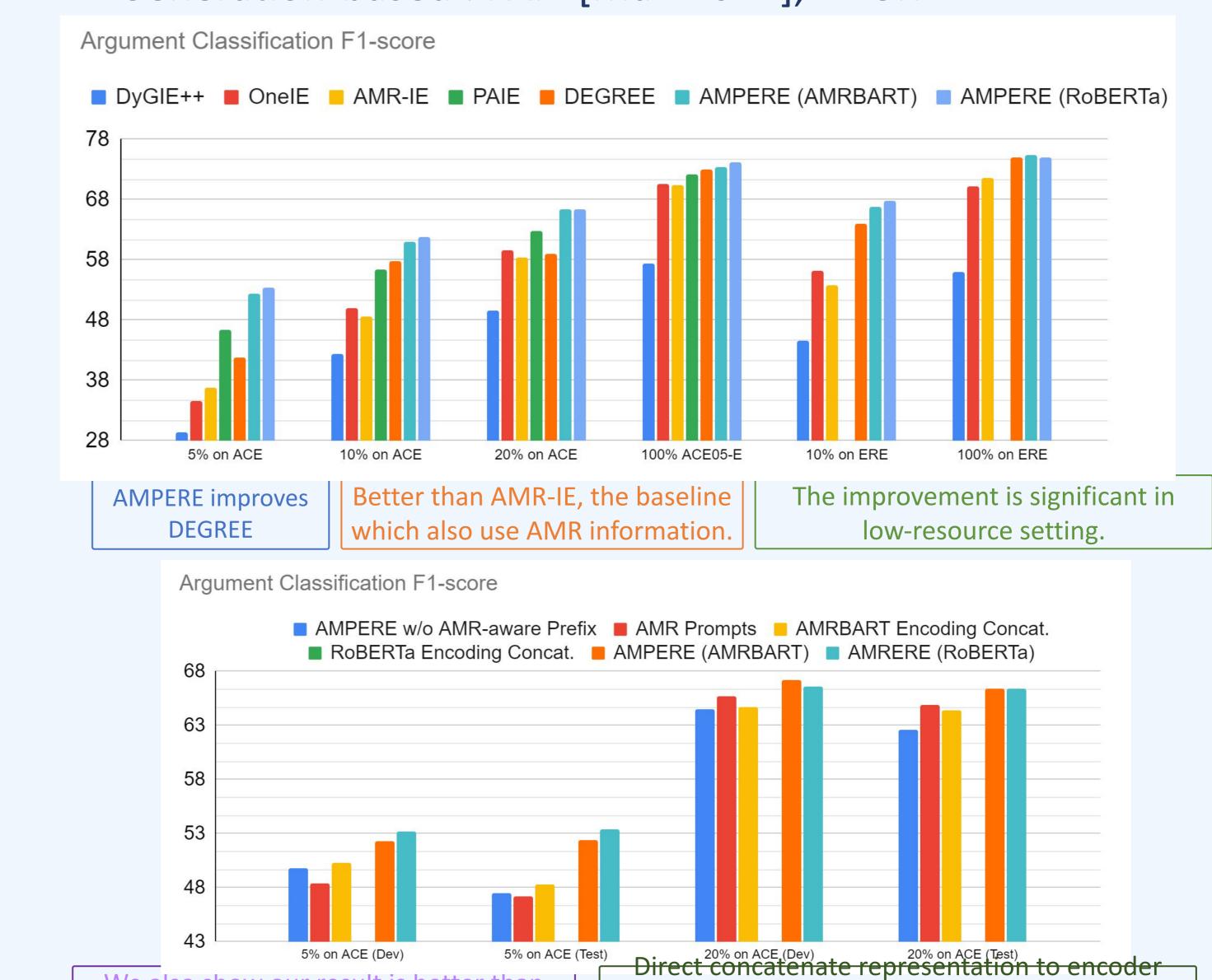
To encourage copy:

 $Loss_{AMPERE} = -\log(\sum_{i} P(y_i|y_{< i}, x_1, ..., x_m)) + \lambda \sum_{i} w_{gen}^{i}$

Experimental results on two benchmarks

Baselines

- Classification-based: DyGIE++ [Wadden+ 2019], OneIE[Lin+ 2020]
- Classification-based with AMR: AMR-IE [Zhang+ 2021]
- Generation-based : PAIE [Ma+ 2022], DEGREE



won't work stably, like RoBERTa Encdoding

Concat

We also show our result is better than

different ways for AMR incorporation.