

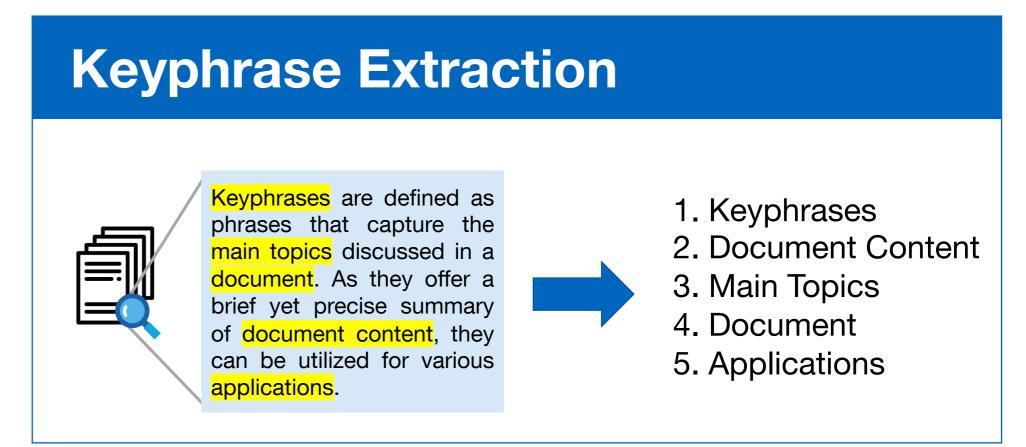
PatternRank: Leveraging Pretrained Language Models and Part of Speech for Unsupervised Keyphrase Extraction

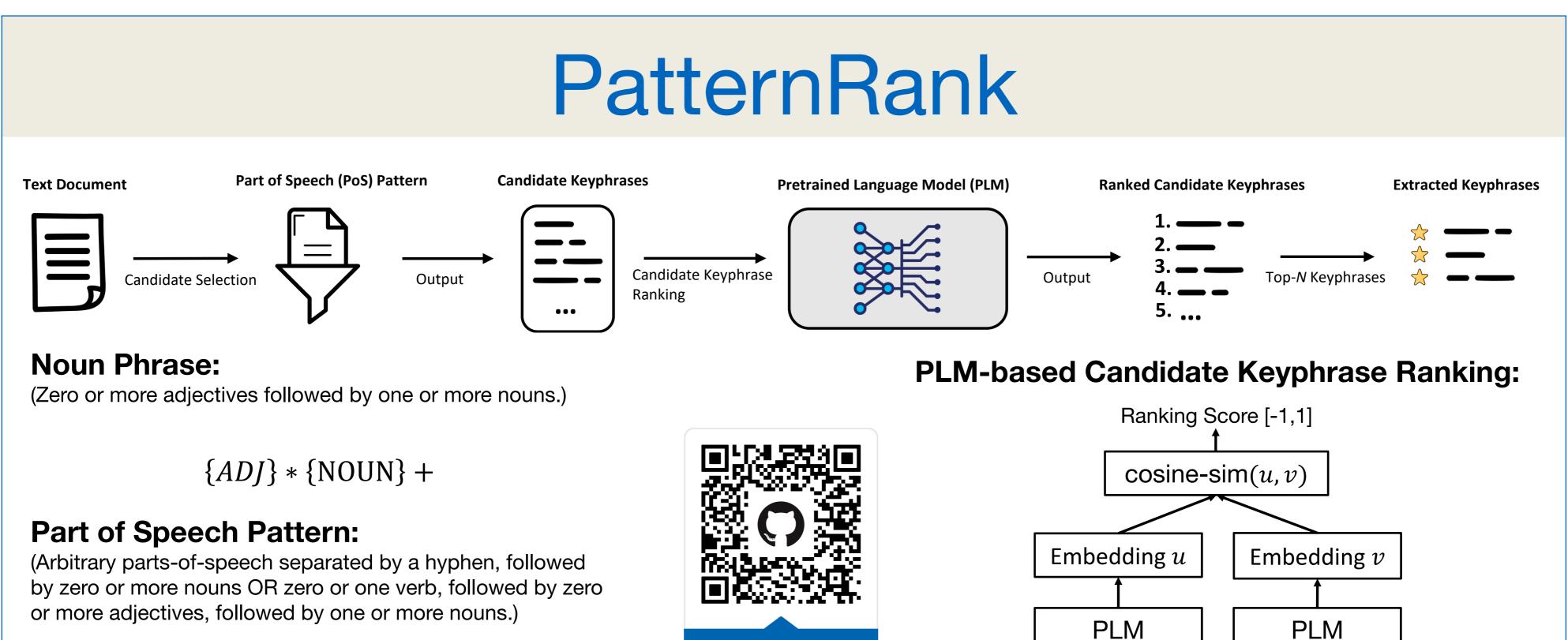
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Motivation

- Provide a quick overview of the content of a text.
- Keyphrases consist of several compound words and can concisely reflect the semantic context of a text.
- Unsupervised keyphrase extraction approaches do not require labeled training data and are mostly domain independent.





SCAN ME

Approach

1. The input consists of a single text document which is being word tokenized.

 $((\{.*\}\{HYPH\}\{.*\})\{NOUN\}*)$

 $((\{VBG\}|\{VBN\})?\{ADJ\}*\{NOUN\}+)$

- 2. The word tokens are then tagged with part-of-speech tags.
- 3. Tokens whose tags match a previously defined part-of-speech pattern are selected as candidate keyphrases.
- 4. Then, a pretrained language model embeds the entire text document as well as all candidate keyphrases as semantic vector representations.
- 5. Subsequently, the cosine similarities between the document representation and the candidate keyphrase representations are computed and the candidate keyphrases are ranked in descending order based on the computed similarity scores.
- 6. Finally, the top-N ranked keyphrases, which are most representative of the input document, are extracted.

Evaluation F₁@5 **Method** F₁@10 **F**₁@20 **YAKE** 15.37 18.50 19.65 21.97 SingleRank 30.80 28.55 KeyBERT 10.30 11.76 7.82 PatternRank NP 23.92 29.19 29.66 PatternRank Pos 24.35 30.99 31.37

Text

Document

Candidate

Keyphrase

- Inspec dataset consisting of 2,000 English computer science abstracts. Each abstract has assigned a set of gold keyphrases.
- Evaluation based on exact match of extracted keyphrases and gold keyphrases.

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