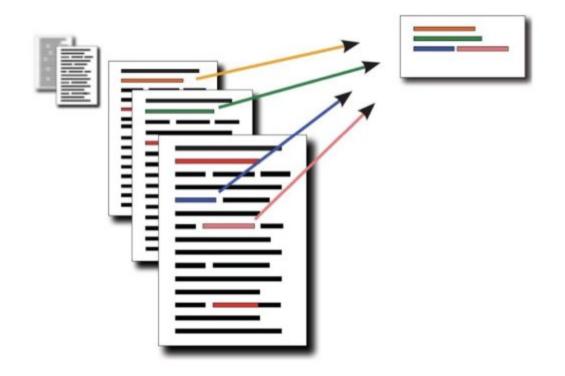
# **Text Summarization**

Long text, big problem?

What about TL;DR?

## What is it?

Task: produce an abridged version of a text while retaining the key, relevant information



## What is it?

### Useful for creating:

- · outlines or abstracts of any document, article, etc
- summaries of chat and email
- action items from a meeting
- simplifying text by compressing sentences

### Single-doc summarization

#### Document

Cambodian leader Hun Sen on Friday rejected opposition parties 'demands for talks outside the country, accusing them of trying to 'internationalize "the political crisis.

Government and opposition parties have asked King Norodom Sihanouk to host a summit meeting after a series of post-election negotiations between the two opposition groups and Hun Sen's party to form a new government failed.

Opposition leaders Prince Norodom Ranariddh and Sam Rainsy, citing Hun Sen's threats to arrest opposition figures after two alleged attempts on his life, said they could not negotiate freely in Cambodia and called for talks at Sihanouk's residence in Beijing. Hun Sen, however, rejected that."

I would like to make it clear that all meetings related to Cambodian affairs must be conducted in the Kingdom of Cambodia, "Hun Sen told reporters after a Cabinet meeting on Friday." No-one should internationalize Cambodian affairs.

It is detrimental to the sovereignty of Cambodia, "he said .Hun Sen's Cambodian People's Party won 64 of the 122 parliamentary seats in July's elections, short of the two-thirds majority needed to form a government on its own .Ranariddh and Sam Rainsy have charged that Hun Sen's victory in the elections was achieved through widespread fraud .They have demanded a thorough investigation into their election complaints as a precondition for their cooperation in getting the national assembly moving and a new government formed ......

#### Summary

Cambodian government rejects opposition's call for talks abroad

### Output

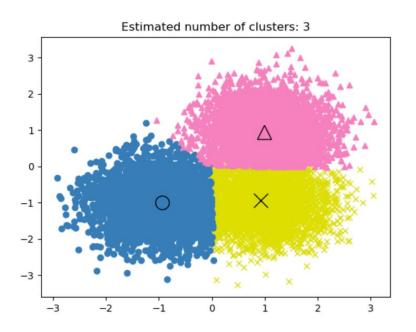
#### Extractive summarization

Select from the source text spans that capture the key information

#### Abstractive summarization

Generate new text that encapsulates the key information from the source text

For abstractive, we generally use transformers and GPT based architecture



#### Brute force:

Sentence embeddings from BERT or USE

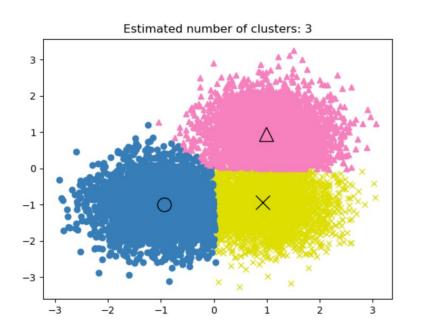
K-means clustering to find common themes

Convert into lower dimensional space

Take the sentence at or the closest to centroid

Sometimes this approach works

But do you see any problems? Let's discuss!



Brute force:

Sentence embeddings from BERT or USE

K-means clustering to find major themes

Convert into lower dimensional space

Take the sentence at or closest to the centroid

Problems still exist!

Centroid may not be the best choice.

Corset could be a solution

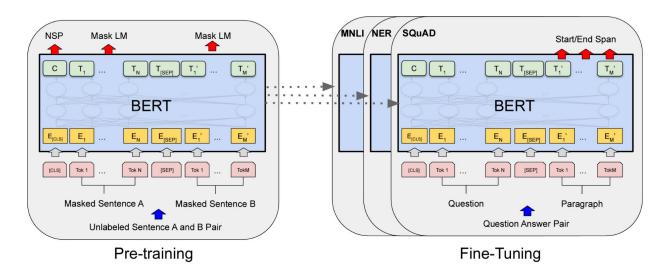


Figure 1: Overall pre-training and fine-tuning procedures for BERT. Apart from output layers, the same architectures are used in both pre-training and fine-tuning. The same pre-trained model parameters are used to initialize models for different down-stream tasks. During fine-tuning, all parameters are fine-tuned. [CLS] is a special symbol added in front of every input example, and [SEP] is a special separator token (e.g. separating questions/answers).

Let's discuss! How can you fine-tune BERT for summarization?

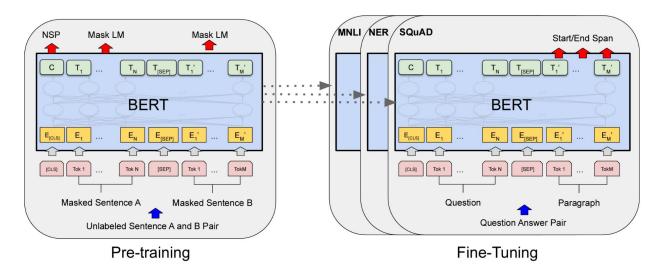


Figure 1: Overall pre-training and fine-tuning procedures for BERT. Apart from output layers, the same architectures are used in both pre-training and fine-tuning. The same pre-trained model parameters are used to initialize models for different down-stream tasks. During fine-tuning, all parameters are fine-tuned. [CLS] is a special symbol added in front of every input example, and [SEP] is a special separator token (e.g. separating questions/answers).

In transfer learning, what are we exactly transferring?



Machine generated summary

6 common words



Human reference summary



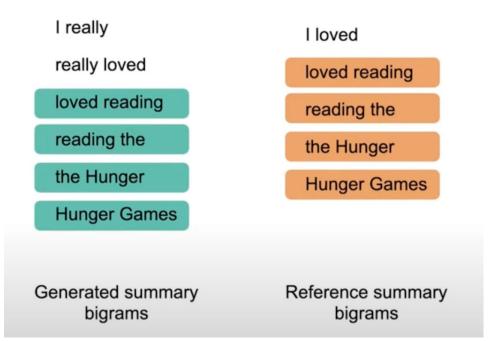
Machine generated summary

Human reference summary

$$\frac{\text{ROUGE-1}}{\text{recall}} = \frac{\text{Num word matches}}{\text{Num words in reference}} = \frac{6}{6}$$

$$\frac{\text{ROUGE-1}}{\text{precision}} = \frac{\text{Num word matches}}{\text{Num words in summary}} = \frac{6}{7}$$

$$\frac{\text{ROUGE-1}}{\text{F1-score}} = 2 \left( \frac{\text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}} \right)$$



$$\frac{\text{ROUGE-2}}{\text{recall}} = \frac{\text{Num bigram matches}}{\text{Num bigrams in reference}} = \frac{4}{5}$$

$$\frac{\text{ROUGE-2}}{\text{precision}} = \frac{\text{Num bigram matches}}{\text{Num bigram in summary}} = \frac{4}{6}$$



Machine generated summary

Human reference summary

$$\frac{\text{ROUGE-L}}{\text{recall}} = \frac{\text{LCS(gen, ref)}}{\text{Num words in reference}} = \frac{6}{6}$$

$$\frac{\text{ROUGE-L}}{\text{precision}} = \frac{\text{LCS(gen, ref)}}{\text{Num words in summary}} = \frac{6}{5}$$

Longest Common Subsequence!

## Acknowledgements

Dr. Chris Tanner at Harvard

Hugging Face YouTube Lectures