# Automatic summarization

**Automatic summarization:** Automatic summarization is the process of shortening a set of data computationally, to create a subset (a summary) that represents the most important or relevant information within the original content.

**Types of text summarization:** There are 2 types of text summarization,

- 1. Extractive Summarization: content is extracted from the original data, but the extracted content is not modified in any way.
- 2. Abstractive summarization: Abstractive methods build an internal semantic representation of the original content, and then use this representation to create a summary that is closer to what a human might express.

**Extractive Summarization:** In my sample project, I focused in extractive summarization. Where the summary is a set of important sentences from original document, which might represent the whole document briefly.

**Solving Approach:** The approach I follow has 3 steps,

**Step 1,** First we tokenize the sentences from given document. Then we need the numerical representation of the sentences. For that I use BERT from transfer learning. Basically I used hugging's pre-trained bert-base-uncased model [1]. It will give us encoded representation of the sentences.

**Step 2,** when I got the encoded representation of the document. I can feed my data to any machine learning model. Here, I used K-Means Clustering to cluster similar sentences. But here is a problem, for K-Means Clustering I need the value of K.

### How could I find the appropriate value for K?

For that, I use elbow and silhouette method. It will give me optimize K value for K-Means Clustering.

**Step 3,** when I got the cluster the next procedure is to extract the most important sentences for the summary.

#### How did I decide that which sentences to extract?

Here I extract one sentence from the cluster as they all represent the similar context.

## But, which one?

Sentence, which is closest to the centroid of the clusters.

To improve readability, I sorted the sentences and marge them. Finally we got our expected summary. Below, there is a diagram to represent the whole approach.

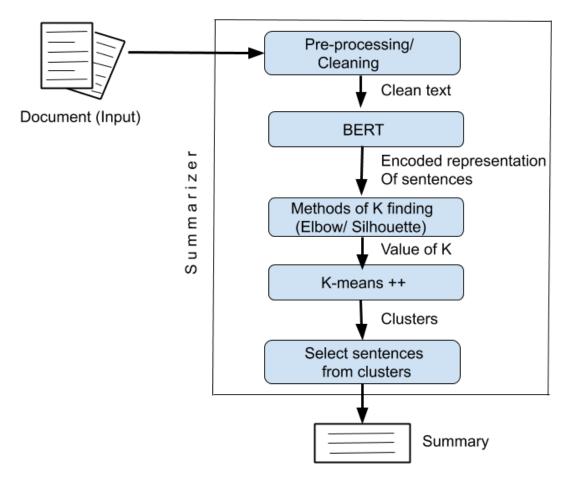


Fig: The flowchart of Solving Approach.

**Conclusion:** This a simple implementation of extractive text summarization. There is a lot of ways to improve it. To evaluate generated summary we can use ROUGE matrix. Which can tell us how good or bad the generated summary is.

**Project Link:** <a href="https://github.com/nowshad7/SampleProject\_Infolytx">https://github.com/nowshad7/SampleProject\_Infolytx</a>

# **References:**

1. https://huggingface.co/bert-base-uncased