# **Documentation**

This project is an extended work of Mooc which attempts to improve lecture segmentation and summarization of each segment based on subtitles of the lectures.

### **Overview**

The task of this project is reduced to sub-tasks: paragraph segmentation and keywords extraction. The approach for paragraph segmentation task uses features of each sentence in the text, and compares their similarity. The keywords extraction task summarizes a paragraph into one vector embedding and finds closest phrases to the embedded paragraph.

## **Implementation**

The data preprocessing is done in corpus.py and text\_parser.py. The keyword\_extraction handles keyword extraction task and needed models. The topic\_modeling.py has models for building bag-of-word topic models and sentence embedding models. And these models are used in papragraph\_segmentation to do the papragraph segmentation task.

See tables in the later sections for all used models. All details are documented in the comments of each file.

### **Environment Requirements**

The project is developped and tested only on python 3.8.12.

Here is all packages used in this project:

- torch=1.10.0
- sentence-transformers=2.1.0
- spacy=2.3.5
  - Installed model en\_core\_web\_sm
- pysrt=1.1.2
- webvtt-py=0.4.6

(I didn't use a requirement.txt mainly because there is an additional model to install...)

# **Usage**

#### **Paragraph Segmentation**

Use paragraph\_segmentation.baseline\_segmentation() to do text segmentation. The specific usege is documented in the comment.

#### **Keywords Extraction**

Use keyword\_extraction.extract\_keywords() Or keyword\_extraction.extract\_keywords\_all() to extract one or multiple paragraphs. The specific usege is documented in the comment.

#### **Evaluation**

Use evaluate\_evaluate\_segmentation() and evaluate\_evaluate\_keyword\_extraction() to evaluate both models. The specific usege is documented in the comment.

## **Test Result**

The entire algorithm is unsupervised, so there is no need to present training and testing error separately. And keep in mind, the testing data size is too small to yield any truly trustable result. Below is the performance of each model with untweaked parameters.

#### **Keyword Extraction**

Model	Precision	Recall
SciBERT-NIi	0.666766	0.506096
SciBERT	0.770466	0.542953
all-MiniLM-L6-v2	0.794343	0.562764

Precision is calculated by the maximum similarity between the predicted keyword and all true keywords.

And recall is calculated by the maximum similarity between the true keyword and all predicted keywords.

### **Paragraph Segmentation**

Model	Segmentation Score
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Model	Segmentation Score
LDA	3.6675
NMF	3.9311
LSI	4.0234
all-MiniLM-L6-v2	3.8637
SciBERT	4.1515

The paragraph segmentation is considered as a partition problem here. The score is the average of completeness score and adjusted MI score between predicted labels and true labels.

## Contribution

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Here is a presentation video link.