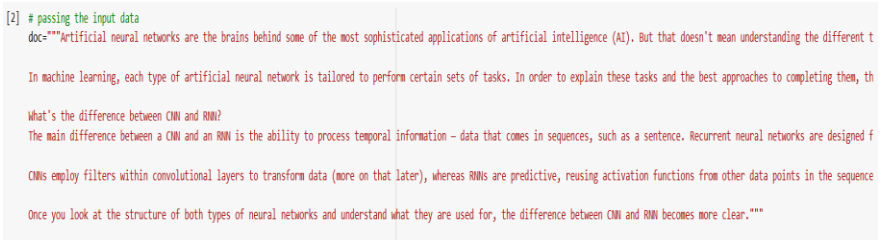
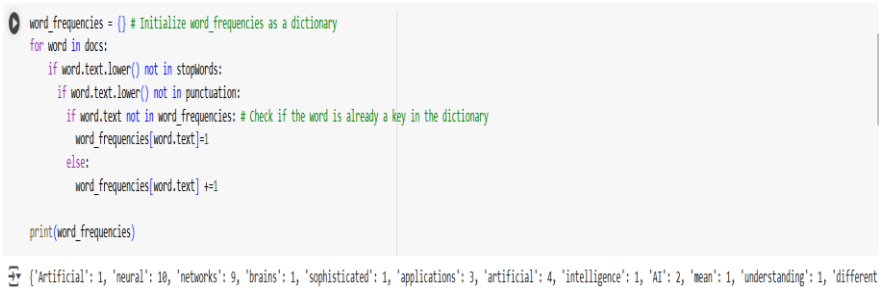
Data Collection and Preprocessing Phase

Date	11 March 2025
Team ID	740052
Project Title	AI-Based Intelligent Insight Extractor
Maximum Marks	6 Marks

Data Exploration and Preprocessing Template:

Data exploration and preprocessing for a text summarization project begin with thorough exploratory data analysis (EDA) to understand the structure, distribution, and relationships within the textual data. Handling missing or incomplete text entries is crucial, often addressed through techniques like data imputation or removal. Text data is cleaned and normalized through processes such as lowercasing, removing stop words, and stemming or lemmatization. Tokenization and vectorization (e.g., TF-IDF or word embeddings) convert text into numerical formats suitable for machine learning models. These steps ensure the dataset is well-prepared for training accurate summarization models.

Section	Description
Data Overview	
Word Frequencies	

Word Tokenization	**Word Tokenization**
	`[7] #Word tokenization is performed`
	`tokens = [i.text for i in docs]`
	`print(tokens)`
	`['Artificial', 'neural', 'networks', 'are', 'the', 'brains', 'behind', 'some', 'of', 'the', 'most', 'sophisticated', 'applications', 'of', 'artificial', 'intelligence', '(', 'AI', ')',`

Normalization	**Normalization**
	`[9] #taking max frequency for normalization`
	`maxFrequency = max(word_frequencies.values())`
	`maxFrequency`
	`10`
	`[10] #normalizing the data`
	`for i in word_frequencies.keys():`
	`word_frequencies[i] = word_frequencies[i]/maxFrequency`
	`print(word_frequencies)`
	`{'Artificial': 0.1, 'neural': 1.0, 'networks': 0.9, 'brains': 0.1, 'sophisticated': 0.1, 'applications': 0.3, 'artificial': 0.4, 'intelligence': 0.1, 'AI': 0.2, 'mean': 0.1, 'understand`

Sentence Tokenization	**Sentence Tokenization**
	`[11] sent_tokens = [sent for sent in docs.sents]`
	`print(sent_tokens)`
	`[Artificial neural networks are the brains behind some of the most sophisticated applications of artificial intelligence (AI)., But that doesn't mean understanding the different types n`
	`, In machine learning, each type of artificial neural network is tailored to perform certain sets of tasks., In order to explain these tasks and the best approaches to completing them,`
	`, What's the difference between CNN and RNN?`
	`, The main difference between a CNN and an RNN is the ability to process temporal information – data that comes in sequences, such as a sentence., Recurrent neural networks are designed`
	`, CNNs employ filters within convolutional layers to transform data (more on that later), whereas RNNs are predictive, reusing activation functions from other data points in the sequenc`
	`, Once you look at the structure of both types of neural networks and understand what they are used for, the difference between CNN and RNN becomes more clear.]`

Data Preprocessing Code Screenshots	

Loading Data	`Artificial neural networks are the brains behind some of the most sophisticated applications of artificial intelligence (AI). But that doesn't mean understanding the different types nee`
	`In machine learning, each type of artificial neural network is tailored to perform certain sets of tasks. In order to explain these tasks and the best approaches to completing them, thi`
	`What's the difference between CNN and RNN?`
	`The main difference between a CNN and an RNN is the ability to process temporal information – data that comes in sequences, such as a sentence. Recurrent neural networks are designed fo`
	`CNNs employ filters within convolutional layers to transform data (more on that later), whereas RNNs are predictive, reusing activation functions from other data points in the sequence`
	`Once you look at the structure of both types of neural networks and understand what they are used for, the difference between CNN and RNN becomes more clear.`

Load Spacy Pipeline	**Load Spacy Language Pipeline**
	`[5] #Loading english language.....(3 different packages are available --small--medium--large). We are loading small packages.`
	`nlp = spacy.load('en_core_web_sm')`
	`[6] #We are passing the input data to spacy`
	`docs = nlp(doc)`
	`print(doc)`

Text Transformation	`[13] from heapq import nlargest`
	`[14] select_len = int(len(sent_tokens)*0.3)`
	`select_len`
	`3`
	`[15] summary = nlargest(select_len, sentence_score, sentence_score.get)`
	`summary`

Feature Engineering	Attached the codes in final submission.
Save Processed Data	-