



# Vidyavardhini's College of Engineering and Technology

Department of Artificial Intelligence & Data Science

AY: 2025-26

<b>Class:</b>	<b>BE-AI&amp;DS</b>	<b>Semester:</b>	<b>VII</b>
<b>Course Code:</b>	<b>CSDOL7011</b>	<b>Course Name:</b>	<b>NLP Lab</b>

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<b>Roll No. :</b>	61
<b>Experiment No.:</b>	8
<b>Title of the Experiment:</b>	Measuring Semantic Similarity Between Sentences using Sentence Transformers
<b>Date of Performance:</b>	
<b>Date of Submission:</b>	

## Evaluation

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<b>Performance Indicator</b>	<b>Max. Marks</b>	<b>Marks Obtained</b>
Performance	5	
Understanding	5	
Journal work and timely submission	10	
Total	20	

<b>Performance Indicator</b>	<b>Exceed Expectations (EE)</b>	<b>Meet Expectations (ME)</b>	<b>Below Expectations (BE)</b>
Performance	4-5	2-3	1
Understanding	4-5	2-3	1
Journal work and timely submission	8-10	5-8	1-4

Checked by

**Name of Faculty** :  
**Signature** :  
**Date** :

**Aim:** To compute the semantic similarity between sentence pairs using pre-trained sentence embedding models from the Sentence Transformers library.

**Objective:** • To measure sentence-level semantic similarity using pretrained sentence transformer models

**Tools Required:**

1. Python (Jupyter Notebook or Google Colab)
2. sentence-transformers library
3. scikit-learn (for cosine similarity)
4. numpy
5. Install Sentence Transformers (if not already installed):
  - a. `pip install -U sentence-transformers`

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### Procedure:

1. Import required libraries:
  - a. `from sentence_transformers import SentenceTransformer`
  - b. `from sklearn.metrics.pairwise import cosine_similarity`
  - c. `import numpy as np`
2. Load a pre-trained model:
  - a. `model = SentenceTransformer('all-MiniLM-L6-v2')`
3. Define two or more sentences to compare:
  - a. `sentences = [`
  - b.  `"A man is playing a guitar.",`
  - c.  `"A person is playing a musical instrument."`
  - d. `]`
4. Generate embeddings:
  - a. `embeddings = model.encode(sentences)`
5. Compute cosine similarity:
  - a. `similarity = cosine_similarity([embeddings[0]], [embeddings[1]])`
  - b. `print(f'Semantic Similarity Score: {similarity[0][0]:.4f}')`
6. Experiment with unrelated sentence pairs and observe similarity values.

### Description of the Experiment:

In this experiment, students explore how sentence-level semantic similarity is measured using transformer-based sentence embeddings. By comparing similar and dissimilar sentences, they



gain an intuitive understanding of how meaning—not just surface words—affects similarity scores.

### Detailed Description of the NLP Technique:

#### 1. Sentence Embeddings:

Sentence embeddings are fixed-length dense vector representations of entire sentences. Unlike word embeddings (e.g., Word2Vec), these models capture the semantic meaning of full sentences.

#### 2. Sentence Transformers:

Built on top of BERT or RoBERTa, the Sentence Transformers framework fine-tunes models to produce high-quality sentence embeddings suitable for:

Semantic textual similarity

Clustering

Semantic search

Question-answer retrieval

The all-MiniLM-L6-v2 model used here is a compact and fast model ideal for educational use.

#### 3. Cosine Similarity:

Measures the cosine of the angle between two vectors. Closer to 1 means more semantically similar:

$$\text{Cosine Similarity} = \frac{A \cdot B}{\|A\| \|B\|}$$



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## Why Use Sentence Embeddings:

- Capture context and meaning rather than individual words.
- Robust to word order changes and synonyms.
- Highly effective in tasks requiring semantic understanding.

## Conclusion:

The results demonstrate that pre-trained sentence embedding models from the Sentence Transformers library effectively capture semantic meaning and context in sentences. The computed similarity scores accurately reflect the closeness in meaning between sentence pairs, validating the model's ability to understand and compare textual content.