

**Institute of Business Administration**  
**Introduction to Text Analytics**  
**Assignment 03**  
**Due Date: 9<sup>th</sup> March 2025 (11:55 PM)**

---

**Note:** This is an individual assignment; hence, everyone must submit it separately.

**Assignment: Clustering News Headlines Using Word2Vec Averaging and Doc2Vec Embeddings**

**Objective:** The goal of this assignment is to explore how different text vectorization techniques impact clustering results. You will apply **K-Means clustering** on a dataset of news headlines using the following different embedding methods:

1. Word2Vec – Averaged word embeddings to obtain headlines embeddings
2. Doc2Vec

Compare how these vectorization techniques affect the clustering quality, as measured by **Within-Cluster Sum of Squares (WSS)** and **Silhouette Score**.

**Dataset:** You are provided with a dataset (news\_Feb\_14.csv) containing around 450 news headlines.

**Tasks:**

1. **Text Vectorization:**
  - **Word2Vec Averaging:**
    - Train a Word2Vec model on the dataset (alternatively, you may use pre-trained embeddings like Google's Word2Vec or GloVe or FastText).
    - Compute headline embeddings via vector averaging.
  - **Doc2Vec:**
    - Train a Doc2Vec model to obtain embeddings for each headline.
2. **Clustering using K-Means:**
  - Perform K-Means clustering using fixed values of  $k = 5, 9$ , and  $13$ . **Please set random\_state parameter to your ERP ID in K-Means initialization. For instance,  $km = KMeans(n\_clusters=4, random\_state=12345)$  if your ERP\_ID = 12345.**
  - Report the Within-Cluster Sum of Squares (WSS) (`kmeans.inertia_` in sklearn) and Silhouette Score.
  - Compare the results across different embeddings.
3. **Analysis & Interpretation:**
  - Identify which embedding technique resulted in the best clustering.
  - Compare the performance of word2vec and doc2vec embeddings with those used in previous assignment (Assignment 02).

**Evaluation Criteria:**

Your submission will be evaluated based on:

- Correct implementation of vectorization and clustering techniques
- Comparison and justification of different approaches
- Quality of analysis and interpretation of clustering results
- Proper use of evaluation metrics (WSS, Silhouette Score)
- Code clarity and documentation

**Deliverables:**

1. Python code notebooks that you used for experimentation.
2. Filled version of the attached document "A3\_Assessment.docx".