# 21CSE426T – Financial Machine Learning Unit 5 Assessment Group Assignment

## Set 1 HOTS:

- 1. Explain how NLP can enhance the process of financial fraud detection. What are the limitations of using NLP in this context?
- 2. Describe how word embeddings like Word2Vec differ from traditional bag-of-words models in representing textual data. How might this affect sentiment analysis results?

# MCQs:

- 1. What is the main purpose of Natural Language Processing?
  - A. Data encryption
  - B. Analyzing numerical data
  - C. Enabling computers to understand human language
  - D. Developing operating systems
- 2. Named Entity Recognition (NER) primarily identifies:
  - A. Synonyms in a text
  - B. Named entities such as names, places, and organizations
  - C. Part-of-speech categories
  - D. Document structure
- 3. In sentiment analysis, a word embedding technique helps to:
  - A. Count word frequencies
  - B. Convert text into vector format
  - C. Recognize entities in text
  - D. Analyze document structure
- 4. Tokenization is used to:
  - A. Encrypt data
  - B. Split text into smaller parts
  - C. Organize documents
  - D. Filter irrelevant words
- 5. Financial sentiment analysis is challenging because:
  - A. Financial texts are often too short
  - B. Financial terms are context-sensitive
  - C. News articles are all unbiased
  - D. Financial data lacks numerical values

## Set 2 HOTS:

- 1. Compare and contrast the use of Naive Bayes and LDA for sentiment analysis in financial news data. How does each model impact the interpretation of results?
- 2. Explain the significance of event return in sentiment analysis-based trading strategies. How does it differ from simple return calculations?

#### MCQs:

- 1. Which of the following is a common step in text preprocessing for NLP?
  - A. Text encryption
  - B. Tokenization
  - C. Data sorting
  - D. Error correction
- 2. Word2Vec is a type of:
  - A. Text classification algorithm

- B. Word embedding model
- C. Language translation tool
- D. Syntax parser
- 3. NLP can be used in finance to analyze:
  - A. Loan repayments
  - B. Market sentiment from news
  - C. Database management
  - D. File compression
- 4. Which Python package is widely used for building NLP applications and has modules for tokenization, stemming, and more?
  - A. NumPy
  - B. pandas
  - C. NLTK
  - D. Matplotlib
- 5. What is the primary advantage of using LSTM for financial sentiment analysis?
  - A. Lower memory requirement
  - B. Handles long-term dependencies in text
  - C. Provides real-time analysis
  - D. Uses fewer data points

## Set 3

## **HOTS:**

- 1. Discuss how an NLP-powered chatbot could transform customer service in financial institutions. What ethical considerations should be addressed?
- 2. Propose a pipeline using spaCy for preprocessing financial news data, including tokenization, stop-word removal, and entity recognition. How would you evaluate the effectiveness of each step?

#### MCOs:

- 1. In a trading strategy, what is one role of NLP-based sentiment analysis?
  - A. Predicting transaction fees
  - B. Determining asset liquidity
  - C. Informing buy/sell decisions
  - D. Managing account balances
- 2. Lemmatization differs from stemming by:
  - A. Generating unique non-existent words
  - B. Returning root forms that are actual words
  - C. Ignoring suffixes and prefixes
  - D. Applying only to English text
- 3. In sentiment analysis, what does TextBlob primarily use to determine sentiment polarity?
  - A. Deep learning models
  - B. Naive Bayes classifiers
  - C. Regular expressions
  - D. Rule-based systems
- 4. Which model is commonly used for topic modeling in unsupervised sentiment analysis?
  - A. Naive Bayes
  - B. Latent Dirichlet Allocation (LDA)
  - C. LSTM

- D. Decision Tree
- 5. In a chatbot application, NLP is used primarily to:
  - A. Display web content
  - B. Generate predefined responses
  - C. Interpret and respond to user queries
  - D. Edit database records

#### Set 4

#### HOTS:

- 1. Analyze the challenges in using traditional sentiment lexicons for financial sentiment analysis. How could these lexicons be adapted for financial contexts?
- 2. Develop an argument for or against the use of LSTM-based sentiment analysis in realtime trading systems. Consider computational complexity and potential impacts on trading decisions.

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# Set 5 HOTS:

- 1. Construct a use case for NLP in assessing market volatility through news sentiment. What features would be essential for accurate predictions?
- 2. Evaluate the benefits and limitations of using TextBlob for initial sentiment analysis in financial contexts. What alternative models could provide more accurate sentiment scores?

## MCQs:

- 1. The main limitation of bag-of-words models in NLP is:
  - A. High computational cost
  - B. Loss of word context and order
  - C. Incompatibility with Python
  - D. Complexity of implementation
- 2. Why is a deep learning model like LSTM preferred for certain NLP tasks?
  - A. Less training data is required
  - B. It captures sequential dependencies in text
  - C. It has lower computational demands
  - D. It is suitable for rule-based tasks
- 3. For which purpose would Latent Dirichlet Allocation (LDA) be most suitable?
  - A. Sentiment analysis
  - B. Text classification
  - C. Topic modeling
  - D. Syntax parsing
- 4. Which sentiment analysis method is most suitable for analyzing stock market conversations?
  - A. TextBlob
  - B. Financial Lexicon-based method
  - C. Basic Naive Bayes
  - D. Movie-based sentiment analysis
- 5. Which is a characteristic of unsupervised learning in NLP?
  - A. Requires labeled data
  - B. Uses known categories for training
  - C. Discovers patterns without labeled output
  - D. Utilizes reinforcement techniques

# Set 6

#### **HOTS:**

- 1. Explain the impact of NLP in automating low-value tasks in financial institutions. How does it improve efficiency and objectivity?
- 2. Discuss the potential improvements that a financial sentiment analysis model could gain from using BERT or other advanced pretrained models.

# MCQs:

- 1. Which algorithm is frequently used for initial sentiment classification in NLP?
  - A. Decision trees
  - B. K-Nearest Neighbors
  - C. Naive Bayes
  - D. Reinforcement learning

- 2. What is an advantage of using pre-trained models for NLP tasks?
  - A. Reduced training time and computational cost
  - B. Better grammar correction
  - C. Enhanced visualization of text data
  - D. Improved encryption of data
- 3. The CountVectorizer is commonly used for which type of NLP model?
  - A. Bag-of-words
  - B. Word embedding
  - C. PoS tagging
  - D. Named Entity Recognition
- 4. A rule-based chatbot:
  - A. Learns from previous interactions
  - B. Uses a fixed set of responses
  - C. Generates new phrases
  - D. Trains itself over time
- 5. Dependency parsing is primarily used for:
  - A. Tokenizing text
  - B. Removing stop words
  - C. Calculating sentiment scores
  - D. Understanding relationships between words in a sentence

# Set 7

## **HOTS:**

- 1. Analyze how topic modeling can help uncover hidden themes in financial documents. What challenges might arise in applying topic modeling in this context?
- 2. Discuss the significance of supervised and unsupervised models in NLP for analyzing large volumes of news data. How does each type of model contribute to insights?

## MCQs:

- 1. Which of the following can help in representing the sentiment score of words within a specific context?
  - A. CountVectorizer
  - B. Word embedding
  - C. Rule-based system
  - D. Syntax parsing
- 2. A self-learning chatbot can improve through:
  - A. Rule-based responses
  - B. Using predefined answers
  - C. Supervised training on labeled data
  - D. Error correction
- 3. Which step typically follows tokenization in NLP preprocessing?
  - A. Stop word removal
  - B. Data encryption
  - C. Syntax parsing
  - D. Entity extraction
- 4. Inference in NLP can best be described as:

- A. Tokenizing data
- B. Splitting text into paragraphs
- C. Calculating word frequencies
- D. Generating decisions from processed text
- 5. What does the TF-IDF model achieve in text processing?
  - A. Counts total words in a text
  - B. Orders sentences based on structure
  - C. Highlights words that are important within a document but not common across documents
  - D. Encrypts textual data