

# CSE 261 & DSC 253

## Adv. Data-Driven Text Mining

### Homework 1: Basic Text Classification Introduction

Deadline: Jan. 27<sup>th</sup> (Tuesday of 4<sup>th</sup> Week)

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In this assignment, you will train a text classifier using different feature representation techniques. You should submit the final writeup and code for this assignment. The writeup should be a pdf that includes experimental findings of the programming part. Submissions should be made on **Gradescope** (course code: N2G2NV). Please complete homework **individually**. Please include the code of your solutions in the submission with a write-up describing how to run the code.

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**Resource:** Please use two datasets for this homework that are included in the HW1 download files on Piazza: **New York Times (NYT) news** and **AG News**. **NYT dataset** contains a text column consisting of news articles and a label column indicating the category to which this article belongs. **AG News** has just the text column. For questions 1 and 2, use the logistic regression classifier.

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**Classifier:** It should be trained and tested on the NYT dataset. Shuffle the NYT data with random seed 42, and split it into training, validation, and test splits, with an 80%/10%/10% ratio.

*(Grading will be only based on correct implementation, trivial performance difference caused by running random won't affect your grade.)*

## Task 1: Bag Of Words (30')

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Train a text classifier using 3 types of bag of words techniques and report accuracy & macro-f1 score on the NYT test set.

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- Each document is represented as a **binary-valued** vector of dimension equal to the size of the vocabulary. The value at an index is 1 if the word corresponding to that index is present in the document, else 0. (Consider using `“from nltk import word_tokenize”`)
- Each document is represented by a vector of dimension equal to the size of the vocabulary where the value corresponding to each word is its **frequency** in the document.
- Each document is represented by a vector of dimension equal to the size of the vocabulary where the value corresponding to each word is its **tf-idf value**.

## Task 2: Word2Vec (30')

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Train a text classifier using 3 types of word2vec representation techniques using **100-dimensional** word vectors and report accuracy & macro-f1 score on the NYT test set.

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- Using publicly available pre-trained **Glove embeddings**

(<http://nlp.stanford.edu/data/glove.6B.zip>) as word vectors, a document vector is represented as an average of word vectors of its constituent words.

- Train Word2Vec (e.g., by **gensim** package) on AGNews text data and use them as word vectors to compute document vectors by averaging word vectors of its constituent words.

- Train Word2Vec on NYT text data and use them as word vectors to compute document vectors by averaging word vectors of its constituent words.

## Task 3: Pre-trained Neural Models (40')

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Train a text classifier using 2 types of pre-trained neural models and report accuracy & macro-f1 score on the NYT test set.

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- Fine-tune the **BERT** (<https://huggingface.co/google-bert/bert-base-uncased>) for text classification. While tokenizing, set the maximum length to 64 and fine-tune for 3 epochs.
- Fine-tune the **ModernBERT** (<https://huggingface.co/answerdotai/ModernBERT-base>) for text classification. While tokenizing, set the maximum length to 64 (*ignore if you can train with full-length input*) and fine-tune for 3 epochs.