# <u>Capstone Project</u>: **Text Classification: Comparative Analysis of Different Deep Neural Network Architectures**

### **Description:**

In this project, you are expected to develop and compare three different deep neural network (DNN) architectures for the task of text classification. The following three architectures should be explored:

- 1. CNN: Convolutional Neural Networks
- 2. RNN: Recurrent Neural Networks
- 3. HAN: Hierarchical Attention Networks

You are also expected to use Word vectors generated by Google's GloVe as an underlying data model:

- https://nlp.stanford.edu/projects/glove/
- "GloVe is an unsupervised learning algorithm for obtaining vector representations for words (similar to Word2Vec). Training is performed on aggregated global word-word cooccurrence statistics from a corpus, and the resulting representations showcase interesting linear substructures of the word vector space."

#### Why text classification?

- Text classification is one of the most important Natural Language Processing & Supervised Machine Learning tasks in different business problems.
- Example business applications include but not limited to:
  - Understanding audience sentiment from social media
  - Detection of spam & non-spam emails
  - o Auto tagging of customer queries
  - o Categorization of news articles into predefined topics

## **Submission Requirements:**

Your submission must include:

- Jupyter notebooks (ipynb) and the corresponding html files for each of the DNN architectures: CNN, RNN, and HAN
  - You may choose Keras or PyTorch for implementation
  - o Implementation must be Python-based
- Power point slides and the corresponding PDF files illustrating the final DNN architecture
- Data sets or the links pointing to where download the data sets from
- Report with the results supporting your comparative analysis: Include the following information and metrics at a minimum:
  - o Description of the specific text classification problem of your choosing
  - Description of the data sets
  - Summary table of the data set sizes: train, validation, test
  - o Architectures: visual graphs

- o Architecture hyperparameters: Table
- o Training and Validation Accuracy and Loss over Epochs: line graphs
- o Time/Epoch (min.) bar graphs
- Hyperparameter Tuning: Choices, rationale, observed impact on the model performance
- If model training takes more than 10 minutes, then you should save your models and include the saved models with the submission and provide the code to load your models

#### **Useful Resources:**

- CNN: <a href="http://www.wildml.com/2015/11/understanding-convolutional-neural-networks-for-nlp/">http://www.wildml.com/2015/11/understanding-convolutional-neural-networks-for-nlp/</a>
- RNN: http://colah.github.io/posts/2015-08-Understanding-LSTMs/
- HAN for Text Paragraphs and Documents: https://arxiv.org/pdf/1506.01057v2.pdf
- HAN for Text Classification: https://www.cs.cmu.edu/~diyiy/docs/naacl16.pdf
- <a href="https://towardsdatascience.com/understanding-neural-networks-from-neuron-to-rnn-cnn-and-deep-learning-cd88e90e0a90">https://towardsdatascience.com/understanding-neural-networks-from-neuron-to-rnn-cnn-and-deep-learning-cd88e90e0a90</a>
- https://machinelearningmastery.com/cnn-long-short-term-memory-networks/