## Project Final Presentation

Semi-supervised Sequence Learning

Sabiha Sultana<sup>1</sup>, Piyakorn Munegan<sup>2</sup>, Sanjay Kanakkot Viswanathan<sup>3</sup>, Mohammed Rizwan Amanullah<sup>4</sup>

> Department of Computing, Macquarie University

> > October 31, 2021

## Recap - Semi-supervised Sequence Learning Project

- Goal: Replicating project which aims on unlabelled data to improve sequence learning with Recurrent network
- (LM-LSTM) Language modelling. Predict what comes next in sequence.
- (SA-LSTM) Reads input sequence into vectors and predict the input sequence again.
- Requirements: AWS(EC2) with Python 3, tensorflow v 1.15.5
- Source: https://arxiv.org/pdf/1511.01432v1.pdf
- Source Repo: https://github.com/tensorflow/models/tree/master/research/adversarial
- Dataset:
  - IMDB movie review dataset
  - DBpedia dataset (DBpedia abstracts and its categories)

## Recap - Replication Process

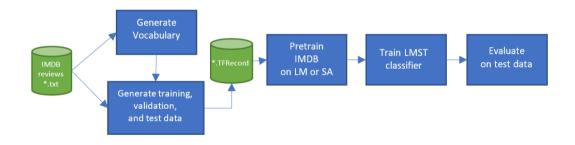


Figure 1: Explains the flow of replication for Sentiment Analysis.

#### IMDB - Data Generation

#### The Process

- 1. Created Python Script.
- 2. Scrapped the data in 7 categories.
- 3. BeautifulSoup python package.

Used python script to scrap the recent reviews from IMDB Website. Scrapped 15% of original data-set, latter manually annotated 2000 file, among them as positive and negative sentiment category and 1000 treated as unlabelled data.

Table 1: A summary of total data scrapped using Beautifulsoup from IMDB Website.

Dataset	Labelled	Unlabelled	
IMDB	2000	1000	

 Repository: https://github.com/sanjay-kv/Semi-supervised-sequence-learning-Project/tree/main/imdb-review-scrapping

#### Amazon New Data Generation

- Scrapped amazon.in website using selenium.
- We have extracted product listing and product reviews information for ten categories.
  - After extraction, we are manually tagging the product listing and reviews.
  - For product listing we are manually tagging the company names and for which having positive and negative reviews.

Table 2: A summary of total data scrapped using Beautifulsoup from IMDB Website.

Dataset	Scrapped Listings	
Amazon	3154	

 Repository: https://github.com/sanjay-kv/Semi-supervised-sequence-learning-Project/tree/main/amazon-scrapping

### Dataset for the sentiment classification task

#### **Dataset**

- 1. Original IMDB movie review dataset.
  - Training set: 25,000 labeled and 50,000 unlabeled reviews.
  - Test set: 25,000 labeled reviews.
- 2. Replication of 5% the size of the original IMDB dataset.
  - Training set: 1,111 labeled and 1,111 unlabeled reviews.
  - Test set: 1,111 labeled reviews.
- 3. Replication of new IMDB dataset.
  - Training set: 1,000 labeled and 1,000 unlabeled reviews.
  - Test set: 1,000 labeled reviews.

## Dataset the product classification task

#### **Dataset**

- 1. Dataset: DBpedia dataset.
  - Training set: 560,000 labeled examples.
  - Test set: 70,000 examples.
- 2. Replication of 2% the size of the original DBpedi dataset.
  - Training set: 11,200 labeled examples.
  - Test set: 1,400 examples.
- 3. Replication of new Amazon dataset
  - Training set: 2,795 examples.
  - Test set: 395 examples.

## Pre-training Process

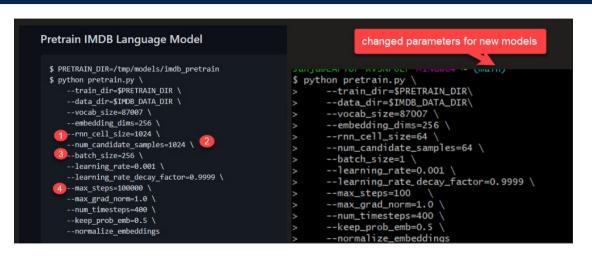


Figure 2: On left the Default parameter, On right Parameters we used for replication

### **Evaluation Results**

Table 3: Summary of the accuracy scores of models on the IMDB sentiment classification task

Dataset	LM-LSTM	SA-LSTM	LSTM
Original IMDB dataset (original work)	92.36	92.76	86.5
5 % of the size of the original IMDB dataset	57.6	56.8	_
New IMDB dataset	50	50	50

Table 4: Summary of the accuracy scores of models on the product classification task

Dataset	LM-LSTM	SA-LSTM	LSTM	SVM
Original DBpedia dataset (original work)	98.5	97.66	86.36	-
2 % of the size of the original DBpedia dataset	7.1	7.1	7.1	???
New Amazon dataset	9.4	9.4	9.4	4.54

## Findings/ Conclusion

#### The Process

- 1. In sentiment Analysis we were able to replicate and validate the accuracy.
- Impact of less data has been observed in the obtained accuracy of the classification task of amazon dataset

We Ensured for NLP task can use LSTM recurrent network, along with it the team were able to find and validate sequence auto-encoder can stabilize the LSTM learning process. LSTM Outperformed all other evaluation categories.

# The End