

# **Assignment-Discussion**

## **POS tagging using (a)**

## **EnCo-DeCo, (b) FFNN-BP**

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10-03-2023

# Problem Statement: Part 1

- Objective: Given a sequence of words, produce the POS tag sequence
- Dataset: Universal Tag Set  
NOUN, VERB, ADJ, ADV, PRON, DET, ADP, NUM, CONJ, PRT, . , X
- Technique used: RNN and Encoder-Decoder LSTM
- Results: RNN gives a better accuracy of around 99%

# Data Processing Info (Pre-processing)

- Corpus has universal tagset and consists of treebank, brown and conll corpora
  - Total number of tagged sentences: 72202
  - Vocabulary size: 59448
  - Total number of tags: 12
- Lower cased all the sentences
- Tokenized the words and tags
- Set max sequence length to 100

# Experimental Setup

- Library used: keras for all models
- Embedding dimension=300, Epochs=10

## RNN Network:

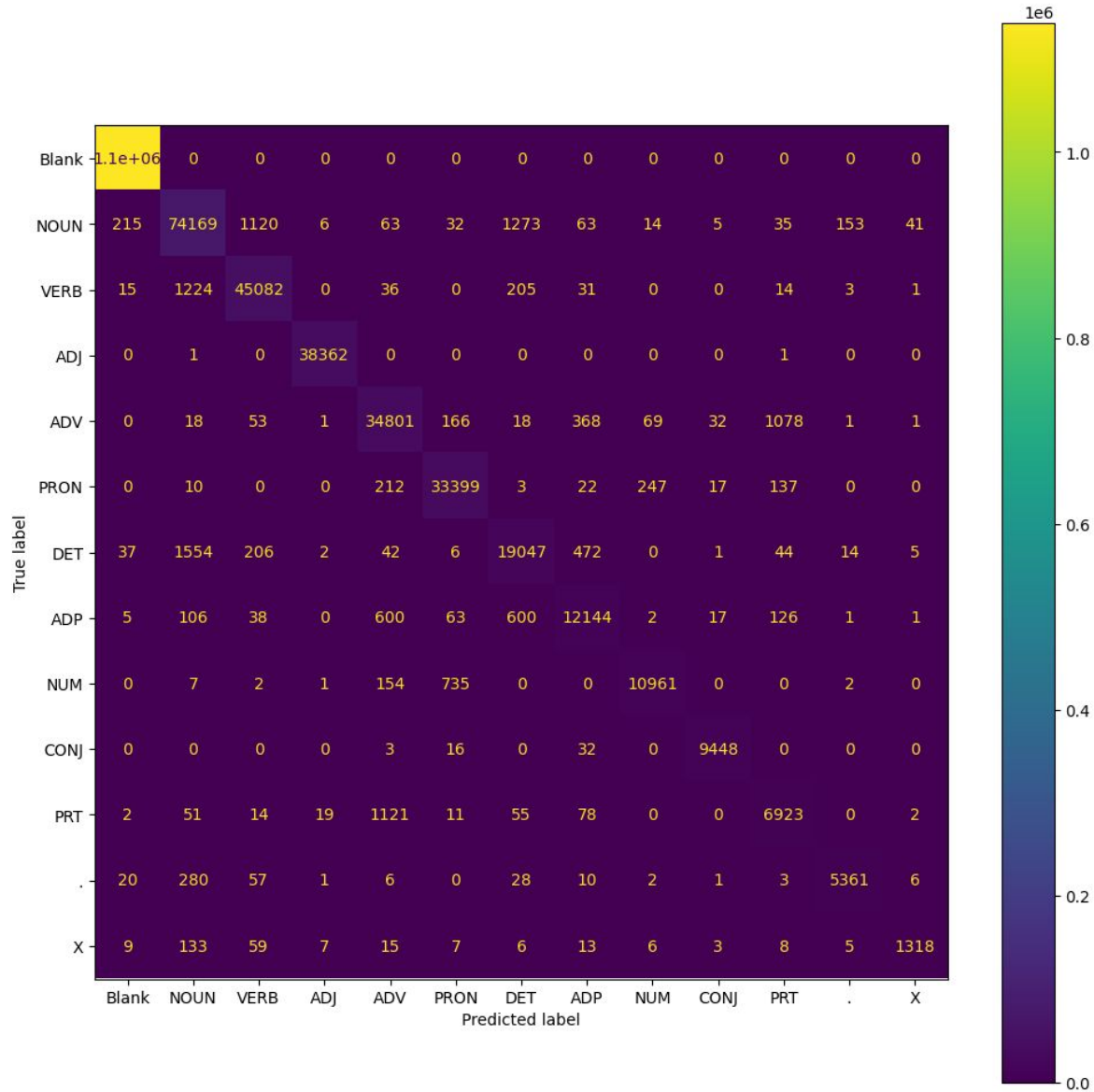
- ```
Model: "sequential"
```

| Layer (type)                       | Output Shape     | Param #  |
|------------------------------------|------------------|----------|
| embedding (Embedding)              | (None, 100, 300) | 17834700 |
| simple_rnn (SimpleRNN)             | (None, 100, 64)  | 23360    |
| time_distributed (TimeDistributed) | (None, 100, 13)  | 845      |
- Initialized weights of the RNN network with word2vec embeddings (gensim)

# Overall performance (Part 1)

|              | Count   | P        | R        | F1       | F0.5     | F2       |
|--------------|---------|----------|----------|----------|----------|----------|
| <b>Blank</b> | 1139149 | 0.999734 | 1.000000 | 0.999867 | 0.999787 | 0.999947 |
| <b>NOUN</b>  | 77189   | 0.956365 | 0.960875 | 0.958615 | 0.957264 | 0.959970 |
| <b>VERB</b>  | 46611   | 0.966782 | 0.967197 | 0.966989 | 0.966865 | 0.967114 |
| <b>ADJ</b>   | 38364   | 0.999036 | 0.999948 | 0.999492 | 0.999219 | 0.999765 |
| <b>ADV</b>   | 36606   | 0.939222 | 0.950691 | 0.944922 | 0.941494 | 0.948375 |
| <b>PRON</b>  | 34047   | 0.969914 | 0.980967 | 0.975410 | 0.972105 | 0.978737 |
| <b>DET</b>   | 21430   | 0.896963 | 0.888801 | 0.892863 | 0.895318 | 0.890421 |
| <b>ADP</b>   | 13703   | 0.917706 | 0.886229 | 0.901693 | 0.911233 | 0.892351 |
| <b>NUM</b>   | 11862   | 0.969914 | 0.924043 | 0.946423 | 0.960379 | 0.932867 |
| <b>CONJ</b>  | 9499    | 0.992020 | 0.994631 | 0.993324 | 0.992541 | 0.994108 |
| <b>PRT</b>   | 8276    | 0.827220 | 0.836515 | 0.831841 | 0.829062 | 0.834639 |
| <b>.</b>     | 5775    | 0.967690 | 0.928312 | 0.947592 | 0.959549 | 0.935929 |
| <b>X</b>     | 1589    | 0.958545 | 0.829452 | 0.889339 | 0.929609 | 0.852412 |

# Confusion Matrix (Part 1)



# Interpretation of confusion (error analysis)

- Maximal confusions:
  - ADV with PRT
  - DET with NOUN
  - PRT with ADV