

# CS671 : Assignment-3

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#### Algorithm:

- Convert <.conll> data to trainable form(transitions).
- Extract features from each transition.
- Feed the above features to a self-designed neural network or sklearn.SGDClassifier.
- Get the predicted transitions and calculate the accuracy.

*Steps of the above mentioned algorithm are discussed in detail below.*

#### Conversion of data from <.conll> format to trainable form:

- Read the file <train.conll> single line and time and converted the dataset to a corpus of the form <list of list>.
- Iterated through the whole corpus and processed one datapoint at a time.
- For every datapoint: started with sigma(stack) containing only <ROOT> and beta(stack) containing all the ords of the sentence. Then used the following rules to decide the transition:

$$o(c = (\sigma, \beta, E)) = \begin{cases} la(r) & \text{if } (\beta[0], r, \sigma[0]) \in DG \\ ra(r) & \text{if } (\sigma[0], r, \beta[0]) \in DG \text{ and } (\forall w_k, r' \text{ if } (\beta[0], r', w_k) \in DG \\ & \text{then } (\beta[0], r', w_k) \in E) \\ sh & \text{Else} \end{cases}$$

*ra(r) says  $(\sigma[0], r, \beta[0])$  added only if all outgoing arcs from  $\beta[0]$  are already in  $E$  - because  $\beta[0]$  will exit from  $\sigma$  and  $\beta$  and cannot be part of any more edges.*

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- Once a transition was decided, it was executed and the resulting sigma and beta was sent to the next step.

#### Feature extraction:

- Following features were extracted:

$\sigma[0]$  to  $\sigma[2]$ ,  $\beta[0]$  to  $\beta[2]$  - 6; the first two leftmost/ rightmost children of  $\sigma[0], \sigma[1]$  - 8; leftmost of leftmost and rightmost of rightmost of  $\sigma[0], \sigma[1]$  - 4. This gives a total of 18 words, so 18 PoS tags and relation labels corresponding to the 12(=8+4) elements in the DG.

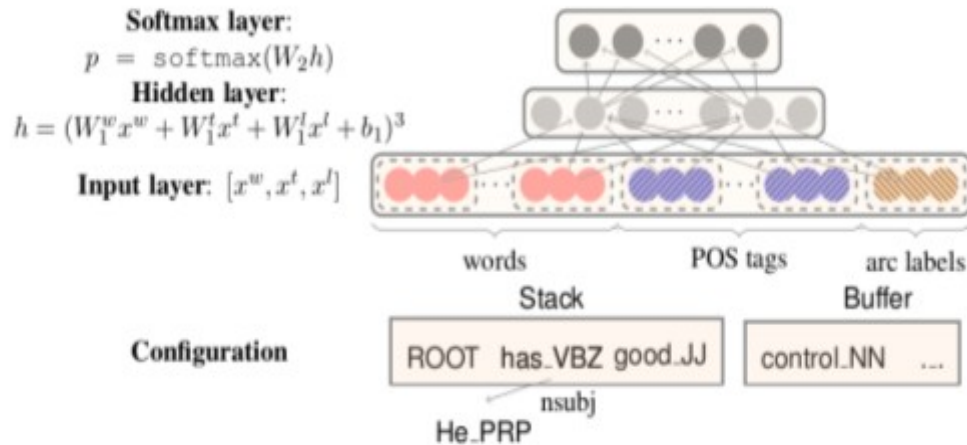
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- So, for every datapoint, our feature matrices were of shape:
  - word - #TRANSITIONS x 18 x VOCAB\_SIZE\_OF\_WORDS
  - PoS - #TRANSITIONS x 18 x #DISTINCT\_PoS
  - Relation - #TRANSITIONS x 18 x #DISTINCT\_RELATIONS

- Now, for every transition: all of the above mentioned feature matrices were concatenated and fed as input to `neural_network/sklearn.SGDClassifier`

## Neural Network:

- Following is the architecture of the neural network used:



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- It was implemented using `<pytorch>`.
- More information about weight tensors:
  - $W1_w$ : Initialed with vectors from pre-trained Word2vec.
  - $W1_t$ : Initialized with random values between -0.01 and 0.01.
  - $W1_l$ : Initialized with random values between -0.01 to 0.01
  - $W2$ : Initialized with random values

## `sklearn.linear_model.SGDClassifier`:

- This is the other model that was used. It was initialized with all the default params.

## Results:

The model was trained over 1000 datapoints and tested over 100 datapoints, both belonging to the train set. Number of training classes were 92.

- Average test accuracy with neural network: 80.53%
- Average test accuracy with `sklearn.linear_model.SGDClassifier`: 90.16%

## Files attached:

- `<preprocess_new_batch.py>`: Preprocessing with feature extraction + `sklearn.linear_model.SGDClassifier` + test(92 training classes)
- `<preprocess.py>`: Preprocessing with feature extraction + test(3 training classes)
- `<preprocess_new.py>`: Preprocessing with feature extraction + test(92 training classes)
- `<model.py>`: neural netork used by `<preprocess.py>` and `<preprocess_new.py>`