# Assignment - 05

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#### **PROBLEM STATEMENT**

- To get a more comprehensive understanding of chunking and chunk types see the posguidelines.pdf attached under this topic
- Using the algorithms learnt in the classes before, develop a chunker.
- Use all other linguistic resources we have dealt with previously : POS, Morph etc.
- Data—English: import this data from nltk: http://www.nltk.org/howto/corpus.html#chunked-corpora
  here is the document related to the data: http://www.cnts.ua.ac.be/conll2000/pdf/12732tjo.pdf
- Develop a POS LM
- From the chunk-annotated corpora develop a chunk-level LM.
- -First make an LM for chunk tags.
- -Generate some chunk-sequences from that LM. are they grammatical?.
- -Make chunk-tag specific LMs i.e. P(wordi|wordi-1,chunk tag) for each word.
- -Now, given a chunk-tag, you can generate words in a chunk.
- -Combine all of this and you get a chunk-based LM generation
- Compare POS LM and Chunk LM .

#### PART A:

## Step1:

Brown corpus downloaded from below nltk link:

http://www.nltk.org/howto/corpus.html#chunked-corpora

### Step2:

Divided Conell corpus in training and testing dataset.

75%- training (8936 sentences)

25% - testing (2012)

#### Step3:

## PART1: Training

- 1. Read tagged corpus
- 2. Tokenize the tagged corpus
- 3. Find all bi-grams in training dataset.
- 4. Find list of all words, tags and chunks in training corpus.
- 5. Compute  $P(t_i|t_{i-1})$  (transition probability)
- 6. Compute  $P(w_i|t_i)$  (emission probability)
- 7. Compute **P(ch<sub>i</sub>|ch<sub>i-1</sub>)** (probability of chunk given previous chunk)
- 8. Compute **P(ch<sub>i</sub>|t<sub>i</sub>)** (probability if chunk given tag)
- 9. Find probability of every tag being a starting tag (Applied Add-1 smoothing while computing this field)
- 10. Find probability of every chunk being starting chunk (Applied Add-1 smoothing while computing this field)

### PART2: Testing

- 1. Read test data
- 2. For each sentence

- $\circ$  Using emission  $P(w_i|t_i)$  and transition probabilities  $P(t_i|t_{i-1})$ , list of all possible tags and words computed on training data using viterbi algorithm find max probability path of tags
- Using emission P(ch<sub>i</sub>|t<sub>i</sub>) and transition probabilities P(ch<sub>i</sub>|ch<sub>i-1</sub>), list of all possible chunks and tags computed on training data using viterbi algorithm find max probability path of tags
- 3. Compute efficiency of code.

For each sentence:

For every word:

If predicted\_chunk==actual\_chunk:

correct+=1

Total +=1

efficiency=(correct/total)\*100

#### **EXPERIMENTAL RESULTS:**

A pack of CoNELL corpus was selected to compute the efficiency of viterbi algorithm.

Total Number of training sentences: 8936

Total Number of testing sentences: 2012

Efficiency:

Count of total number of chunks predicted: 47377

Total tags chunks correctly = 13903

efficiency= 29.6%

PART B: