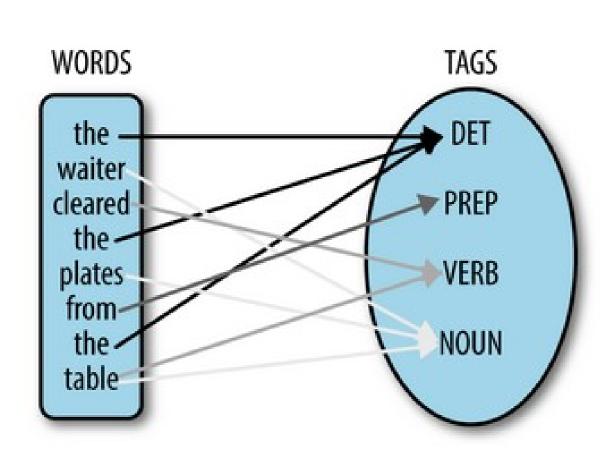
# Part-of-speech(POS) tagging on the Hidden Markov Model

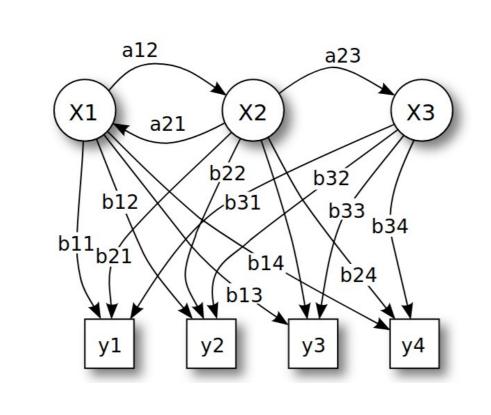
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#### Backgrounds

- Part of Speech(POS) is a category of words which have similar grammatical properties.
- POS tagging in sentences is very important for Natural Language Processing(NLP).
- Hidden Markov Model(HMM) is a statistical Markov model in which the system being modeled is assumed to be a Markov process with unobservable (hidden) states.
- HMM used extensively in tasks processing sequential data, like NLP

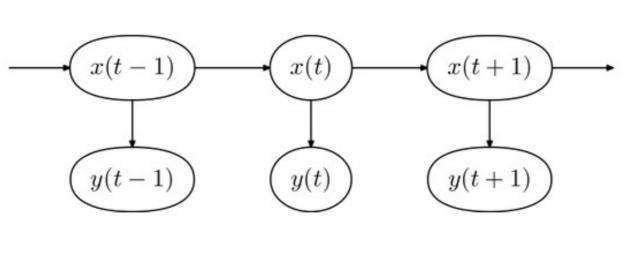




## **Project Goal**

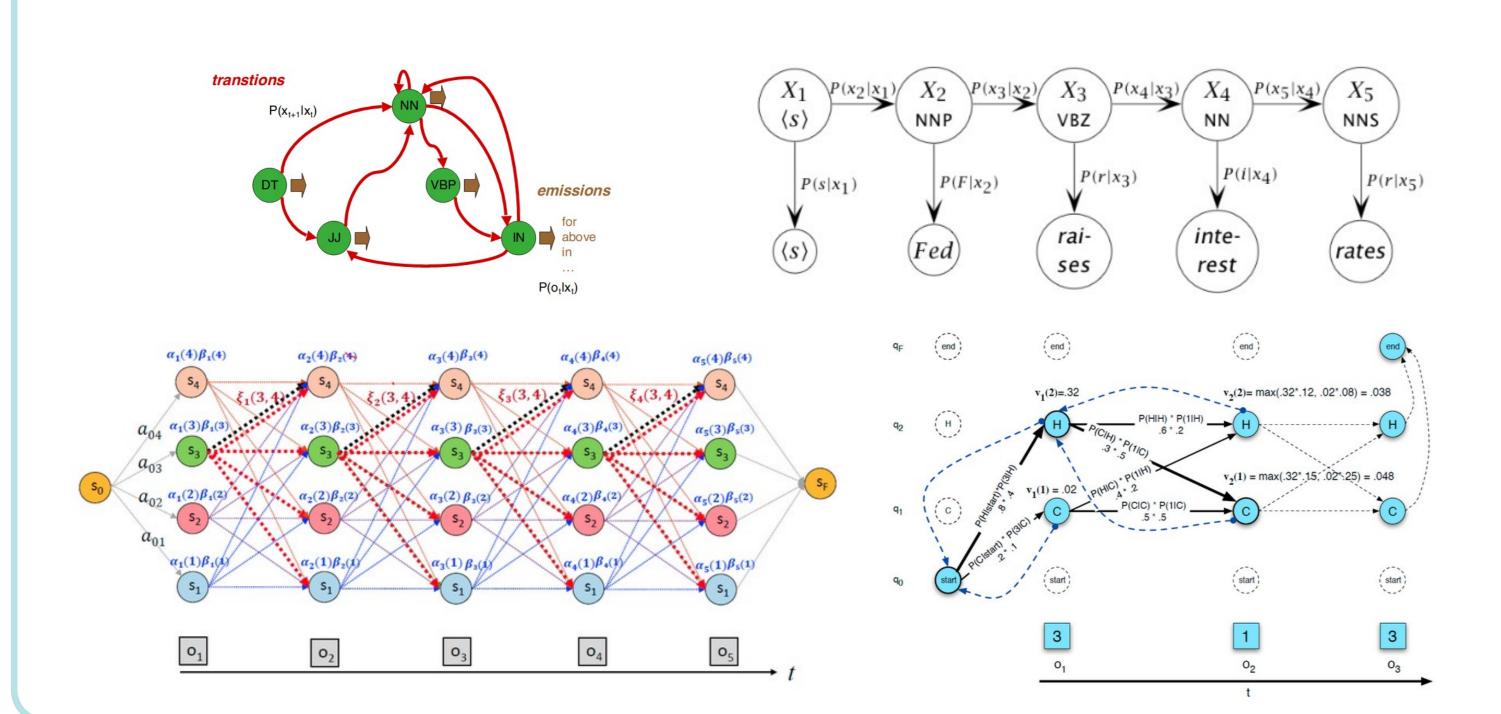
- To learn how HMM works in real world problem solving
  - Use HMM to solve POS-tagging problem in NLP
  - Learn structure of HMM and how it works





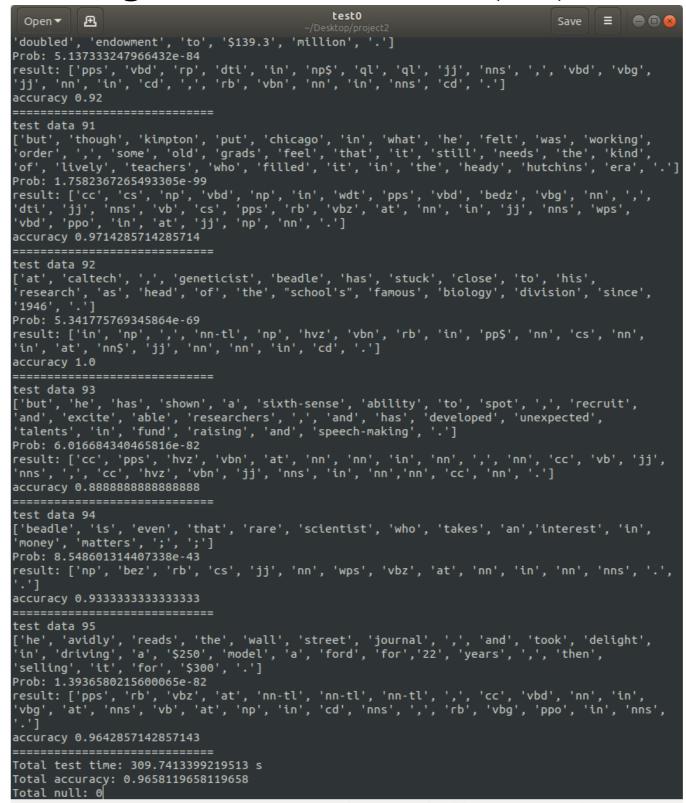
## Methodology

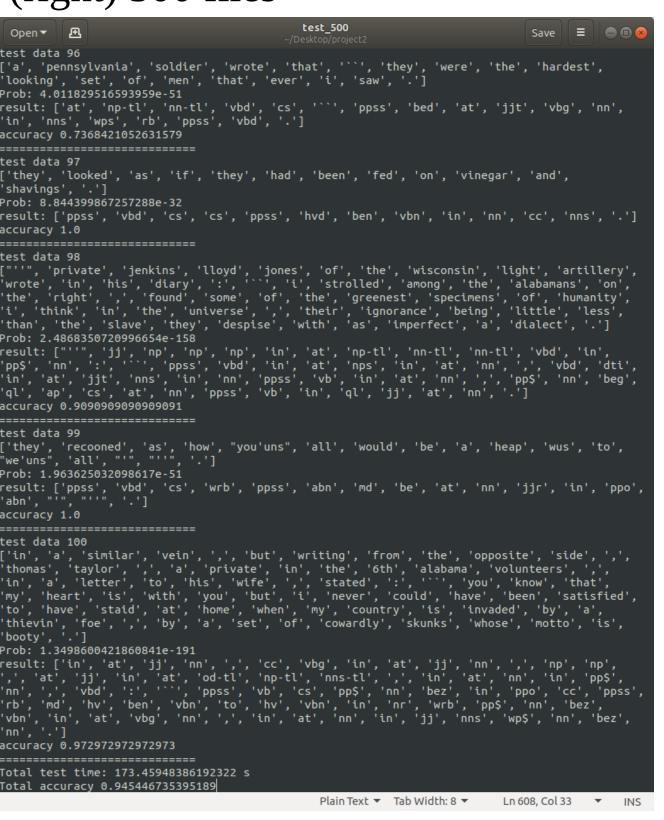
- Use Brown Corpus for training and testing
- English corpus from Brown University used extensively in NLP
- 57,340 sentences with pre-tagged POS in 500 files
- Extract statistical data to determine prior probability
- Calculate start/transition/emission probability by counting POSes and words
- Training HMM using EM algorithm
- Combine forward probability and backward probability
- Iterate until there is no meaningful difference on log-likelihood before and after training
- POS tagging using Viterbi algorithm
- Calculate Viterbi probability by dynamic programming method
- Backtrace hidden states from HMM for decoding most likely sequence of states



### Results & Analysis

- Do training & testing
- Training time is much longer than expected (above a day)
- So, I used prior probability on model instead of running EM algorithm
- Only used 100 among 500 files for training, about 11,000 sentences total
- The training taked about 2.5 hours, the model consists of total 289 states, 20272 symbols(words).
- Do POS tagging on test dataset
- Test dataset consisted of 703 sentences on 5 files which was not included in training dataset
- It taked about 5 minutes to test one file
- Result on about a half of sentences were nearly 90%~100% accuracy, on average.
- But decoding on the other half of sentences did not work, with probability 0.0.
- Why?
- Disclosing why it does not work on some sentences
- Tried a test on a file in training dataset (ca40) to find a clue
- There was no sentences with probability 0.0
- It was not because floating point precision problem
- It was because not-working sentences used words that did not appeared in training dataset
- Analysis for accuracy drop
- Each decoding accuracy of tests on 4 files is about 95% on average
- But one file shows about 81% accuracy on average. Why?
- It turned out that multi-word proper nouns drops the accuracy
- Careful concern of upper/lowercase of letter may resolve the problem
- Below are results of POS tagging
- Using model trained with (left) 100 files, (right) 500 files





#### Lessons learned

- Learned about how HMM works, in detail
- By solving real world problem (POS tagging)
- Training Takes a lot of time
- Maybe it would have been much better to start early
- It feels great to see POS tagging works!

