

* Part of Speech Tagging (POS)

- Use for machines to understand Syntactical meaning of sentence
- Use in chatbots, and advanced NLP for preprocessing

POS:

POS tagging is a task of labelling each word in a sentence with its appropriate part of Speech.

In traditional grammar, part of speech or part-of-speech is a category of words that have similar grammatical properties

e.g.

Why	not	tell	Someone	?
adverb	adverb	verb	noun	punctuation mark, sentence closer.

→ Applications

[POS Tagging is Preprocessing Step]

1. Named Entity Recognition
2. Question Answering System
3. Word Sense disambiguation → [Same word with different meaning]
4. chatbots.

* Spacy library is used for POS Tagging

1) Course Grained POS (POS_)

2) Fine Grained POS (tag_)

Word Sense disambiguation

e.g. I left the room.

To the left of the room.

* How POS Tagging Works?

[Hidden Markov Models]

S (N) (V) (N) E
Nitish love CampusX

S (M) (N) (V) (N) E
Can Nitish google CampusX

S (M) (N) (V) (N) E
Will-ABC google CampusX

S (N) (V) (N) E
ABC loves Will

S (N) (V) (N) E
Will loves Google

S → start

E → End

N → Noun

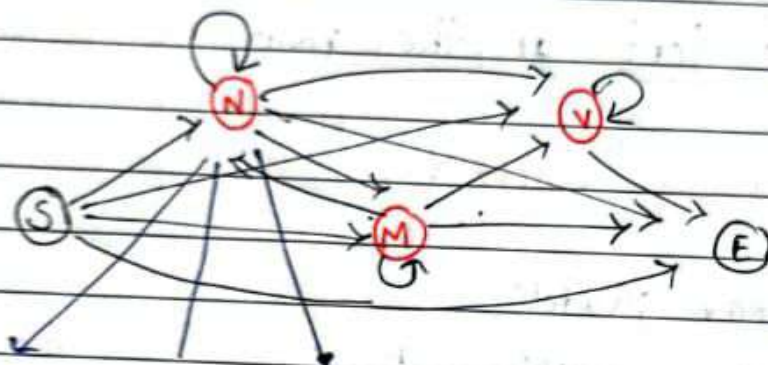
M → Model

Verb → V

Emission
Probability

	Noun	Model	Verb
Nitish	2/10	0	0
loves	0	0	3/5
CampusX	3/10	0	0
google	1/10	0	2/5
Will	2/10	1/2	0
ABC	2/10	0	0
Can	0	1/2	0

Transition Probability		N	M	V	E
S		$3/5$	$2/5$	0	0
N		0	0	$5/10$	$5/10$
M		$2/2$	0	0	0
V		$5/5$	0	0	0



→ Hidden Markov Model

Nitish loves CampusX Will Abc Car

eg

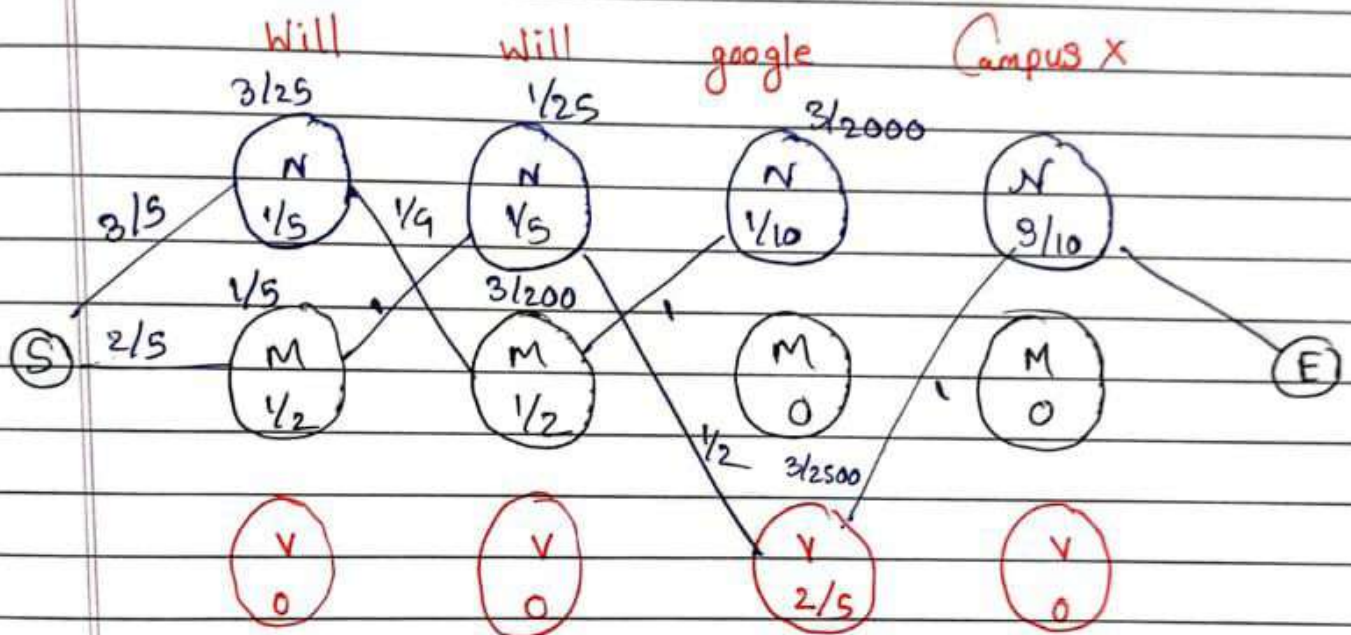
	Will	Will	google	CampusX	
	$1/5$	$1/5$	$1/5$	$1/10$	
S	$3/5$ N	0 N	0 N	0 N	$9/10$ E = 0
S	N	N	N	M	E = 0
	$1/5$	$1/5$	$2/5$	$1/10$	
S	$2/5$ M	1 N	$1/2$ V	1 N	$1/2$ E $\neq 0$

- Algorithm will try different number of combinations of Parts of Speech to find the maximum non zero probability
- Combination with maximum probability is considered as correct combination for POS Tagging
- Since the above sentence only has 3 Parts of Speech it has $3^4 = 81$ combinations to find correct combination of POS Tagging which is exponential.
- Hence Viterbi Optimization Algo is used.

→ Optimization Technique → Viterbi Algo

Transition Probability

	N	M	V	E
S	$\frac{3}{5}$	$\frac{2}{5}$	0	0
N	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{3}{10}$	$\frac{3}{10}$
M	$\frac{2}{2}$	0	0	0
V	$\frac{5}{5}$	0	0	0



$$\frac{3}{5} \times \frac{1}{5} = \frac{3}{25}$$

$$\frac{2}{5} \times \frac{1}{2} = \frac{2}{10} = \frac{1}{5}$$

$$\frac{3}{200} \times \frac{1}{10} = \frac{3}{2000}$$

$$\frac{1}{5} \times \frac{1}{5} = \frac{1}{25}$$

$$\frac{3}{25} \times \frac{1}{4} \times \frac{1}{2} = \frac{3}{200}$$

$$\frac{1}{25} \times \frac{1}{2} \times \frac{2}{5} = \frac{3}{2500}$$

- * Viterbi Algorithm is just an Optimization
- * On each node chooses and select the best path on basis of maximum probability.
- * Transition probability is used for finding best path between two nodes on basis of maximum probability
- * Transition probability with zero is not considered as it may cause overall product of probability to zero.