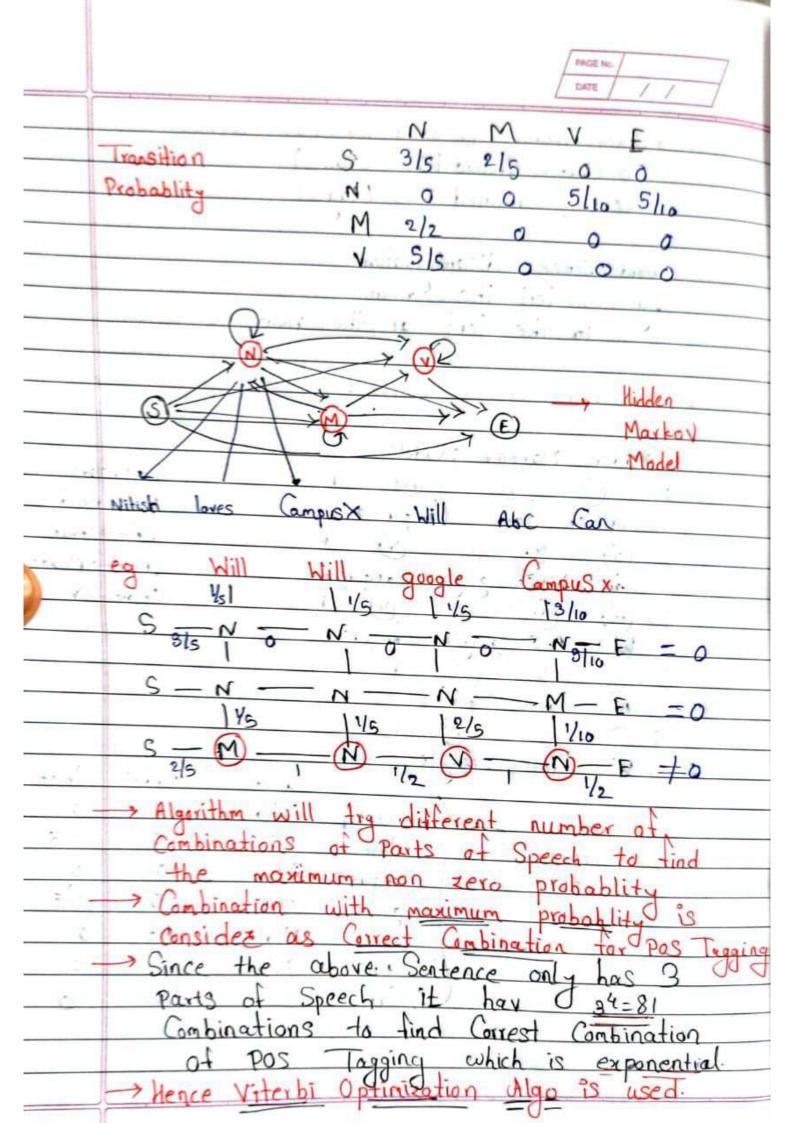
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	Use in chatbots, and advanced NLP for preprocessing
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	Pos:  Pos tagging is a task of labelling each word in a sentence with its appropriate part of Speech.
	in a sentence with its appropriate part of Speech.
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	In traditional grammar, part of speech or part- of-speech is a category of words that have similar grammatical properties
	similar grammatical properties
tories	e.governor
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<b>→</b>	Optimization Technique -> Varterbi Algo
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	* Virterbi Op' Algorithm is just an Optimization * On each node chooses and select the best path
	* On each node chooses and select the best path
	on basis of maximum probablity.  ** Transition probablity is used for finding best  path between two nodes on basis of naximum
	* Transition probablity is used for tinding best
	path between two nodes on basis of naximum
	probablity + Transition probablity with zero is not considered as it may cause overall product of probablity to zero.
	Transition probability with zero is not considered as It
	may cause overall product of probability