

→ Simplified Formula :

$$\hat{T} = \arg\max_T \prod_i P(w_i | t_i) \times P(t_i | t_{i-1})$$

→ by Transition Probabilities : $P(t_i | t_{i-1})$

$$P(t_i | t_{i-1}) = \frac{c(t_{i-1}, t_i)}{c(t_{i-1})}$$

$$\text{Eg : } P(NN | DT) = \frac{c(DT, NN)}{c(DT)}$$

→ Word Likelihood Probabilities : $P(w_i | t_i)$

$$P(w_i | t_i) = \frac{c(t_i, w_i)}{c(t_i)}$$

$$\text{Eg : } P(is | VBZ) = \frac{c(VBZ, is)}{c(VBZ)}$$

→ Disambiguiting (Brown corpus) :

→ Thief is expected to race tomorrow.

→ Hicf is expected to race tomorrow.

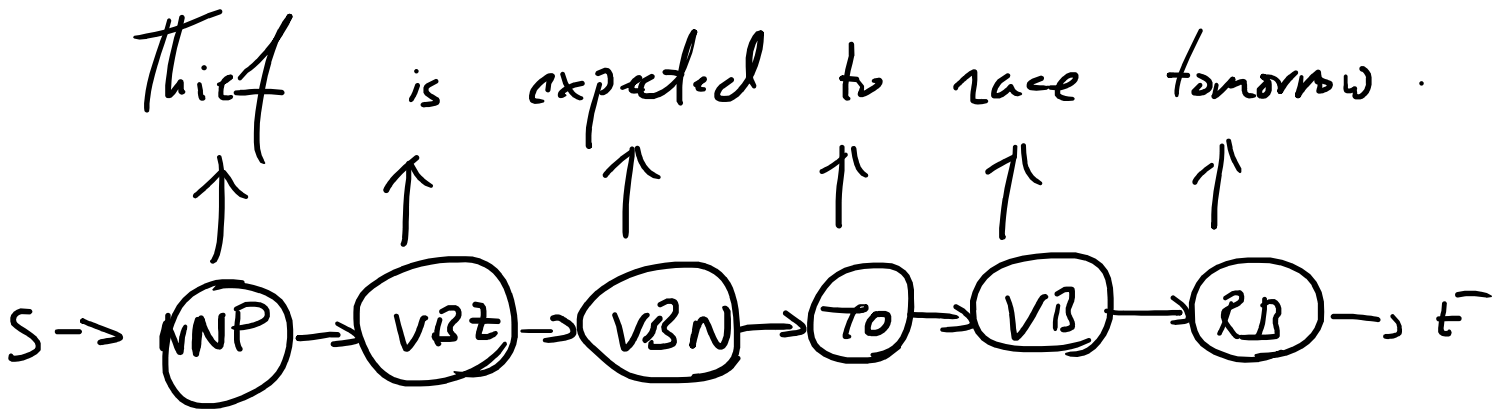
→ Sentence formation:

a) Thief | NNP → is | VBZ → expected | VBN → to | To →
race | VB → tomorrow | RB.

↓) Thief / NNP → is / VBZ → expected / VBN → to / To →
race / NN → tomorrow / RB.

$$a) P(VB | T0) \times P(RB | VB) \times P(Race | VB)$$
$$b) P(NN|TO) \times P(RD|NN) \times P(Rac|NN)$$

Graph Style System:



→ Hidden Markov Model:

→ Tag Transitioning Probabilities : $P(t_i | t_{i-1})$

→ Word likelihood Probabilities: (Emissions)
 $: P(w_i | t_i)$

→ First-Order Markov Model:

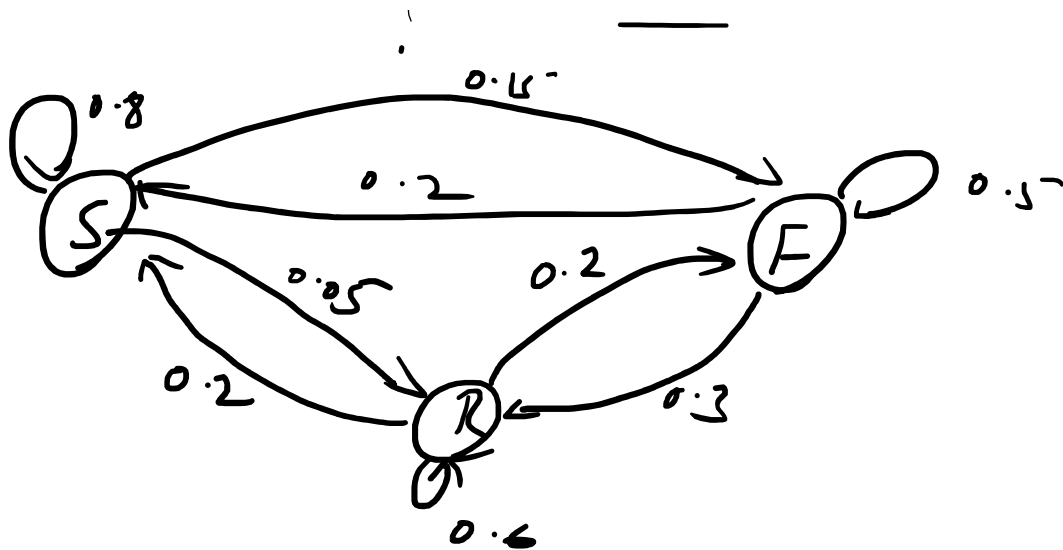
→ Weather Example: Sunny, Rainy, Foggy.

q_n : Variable denoting the weather
 on the n th day.

$$\therefore P(q_n | q_{n-1}, q_{n-2}, \dots, q_1)$$

$$\Rightarrow \text{F-O-M-V} = P(q_n | q_{n-1})$$

	Tomorrow		
Today	S	R	F
<u>S</u>	0.8	<u>0.05</u>	0.15
R	<u>0.2</u>	0.6	0.2
F	0.2	<u>0.3</u>	0.5
		<u>0.4</u>	



→ Given that today the weather is sunny, what is the probability that tomorrow is sunny & day after tomorrow is raining.

$$= P(q_{n+1} = \text{Sunny} \mid q_n = \text{Sunny}) \times P(q_{n+2} = \text{Raining} \mid q_n = \text{Sunny}, q_{n+1} = \text{Sunny})$$

$$= P(q_{n+1} = \text{Sunny} \mid q_n = \text{Sunny}) \times P(q_{n+2} = \text{Raining} \mid q_{n+1} = \text{Sunny})$$

$$= 0.8 \times 0.05$$

$$= 0.04$$

→ the probability of raining day after tomorrow

is 4% iff tomorrow is sunny.

