

Viterbi Algo :- • most common decoding algo.

- Decoding task - determine the sequence of variables of the underlying source of some sequence of observation.
- It is a dynamic programming soln.

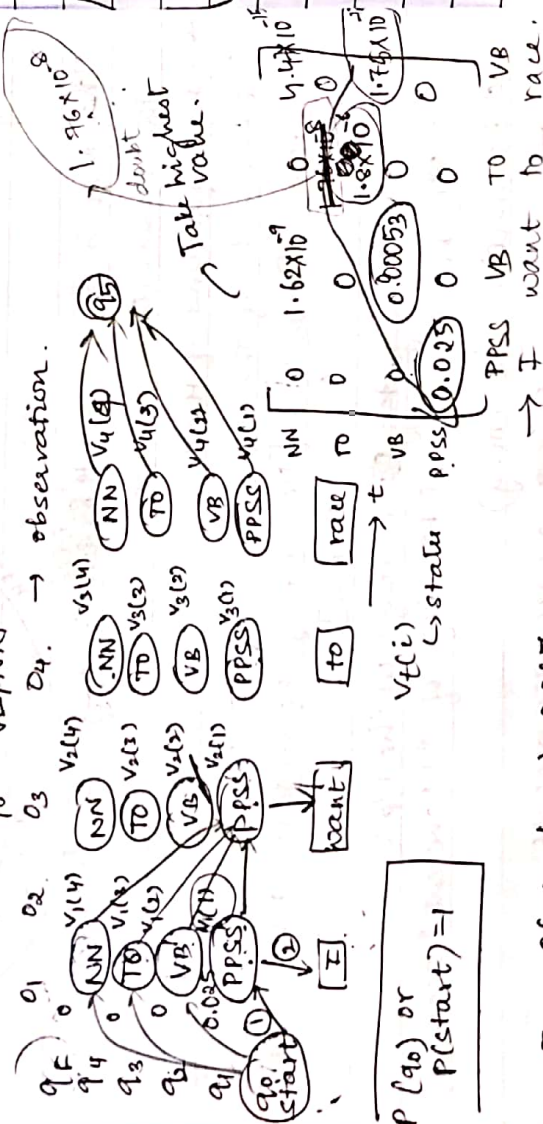
Viterbi algo.

Ifp: a single HMM and a set of observed words $O = (o_1 o_2 o_3 \dots o_T)$

Returns: the most probable state / tag sequence $Q = (q_1 q_2 \dots q_T)$

together with its probability.

Eg: I want to race \rightarrow i/p
VB NN PPSS



$$\textcircled{1} \Rightarrow P(\text{PPSS} | \text{<S>}) = 0.067$$

$$\textcircled{2} \Rightarrow P(I | \text{PPSS}) = 0.087$$

$$\therefore V_1(1) = P(\text{PPSS} | \text{<S>}) \cdot P(I | \text{PPSS}) = 0.067 \times 0.087 = 0.0058$$

$$V_1(2) = P(\text{VB} | \text{<S>}) \cdot P(I | \text{VB}) = 0$$

$$V_1(3) = P(\text{TO} | \text{<S>}) \cdot P(I | \text{TO}) = 0$$

$$V_1(4) = P(\text{NN} | \text{<S>}) \cdot P(I | \text{NN}) = 0$$

$$V_1(5) = P(\text{PPSS} | \text{<S>}) \cdot P(I | \text{PPSS}) = 0$$

$$V_1(6) = P(\text{VB} | \text{<S>}) \cdot P(I | \text{VB}) = 0$$

$$V_1(7) = P(\text{NN} | \text{<S>}) \cdot P(I | \text{NN}) = 0$$

$$V_1(8) = P(\text{PPSS} | \text{<S>}) \cdot P(I | \text{PPSS}) = 0$$

Tag Transition prob.

	VB	TO	NN	PPSS
<S>	0.019	0.0043	0.041	0.067
VB	0.0038	0.035	0.047	0.0070
TO	0.83	0	0.00047	0
NN	0.0040	0.016	0.087	0.0045
PPSS	0.23	0.00079	0.0012	0.00014

Observation likelihood

	I	want	to	race
VB	0	0.0093	0	0.00012
TO	0	0	0.99	0
NN	0	0.00054	0	0.00057
PPSS	0.37	0	0	0

Compute $v_2(1) = \max \{ \text{tag} \cdot P(\text{want} | \text{PPSS}) \}$

$$= \max \{ P(\text{PPSS}) + P(\text{PPSS} | \text{PPSS}) \cdot P(\text{VB}) * P(\text{PPSS} | \text{VB}), \\ V_3(1) \cdot P(\text{TO}) * P(\text{PPSS} | \text{TO}), P(\text{NN}) * P(\text{PPSS} | \text{NN}), P(\text{want} | \text{PPSS}) \}$$

$$= 0.025 * 0.00014 * 0 = 0 //$$

$$v_2(2) = \max \{ v_1(2) \cdot P(\text{VB} | \text{VB}) + v_1(1) \cdot P(\text{VB} | \text{PPSS}), \\ v_1(3) \cdot P(\text{VB} | \text{TO}), v_1(4) \cdot P(\text{VB} | \text{NN}) * P(\text{want} | \text{VB}) \}$$

$$= 0.23 * 0.025 + 0.0073 = 0.000053$$

$$= 0.000053$$

$$v_2(3) = \max \{ v_1(1) \cdot P(\text{TO} | \text{PPSS}), v_1(2) \cdot P(\text{TO} | \text{VB}), v_1(3) \cdot P(\text{TO} | \text{TO}), \\ v_1(4) \cdot P(\text{TO} | \text{NN}) * P(\text{want} | \text{TO}) \}$$

$$= \{ 0.025 * 0.00079, 0, 0, 0 \} * 0 = 0 //$$

HW :-

Eg 2: learning changes thoroughly

	NN	VB	Adv	qF
q_0	2×10^{-1}	3×10^{-1}	0	0
NN	1×10^{-1}	3×10^{-1}	1×10^{-1}	0
VB	4×10^{-1}	1×10^{-1}	4×10^{-1}	0
Adv	0	0	0	1×10^{-1}

	learning	changes	thoroughly
NN	1×10^{-3}	3×10^{-3}	0
VB	3×10^{-3}	4×10^{-3}	0
Adv	0	0	2×10^{-3}