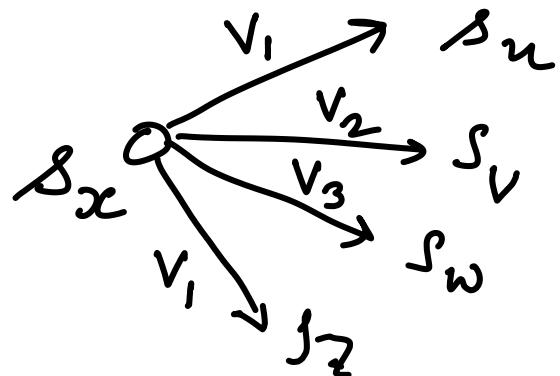


In class quiz on Fri (Sept 27)  
syllabus - dynamic Prog

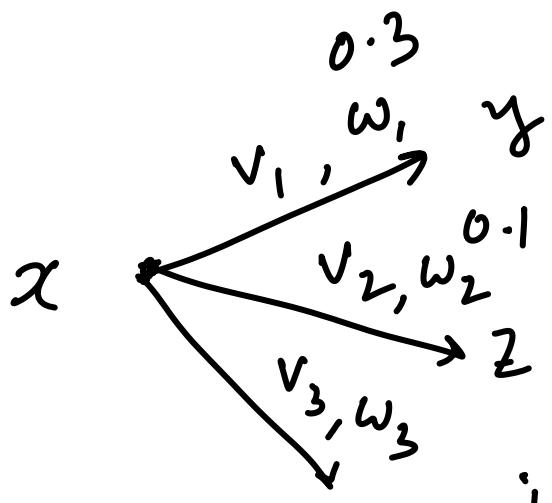
Often for many real life applications we form hypotheses on the basis of observations and we would like to form a hypothesis - that is most likely.

We have a weighted directed graph, say  $G = (V, E, w)$   
 $w$  is a weight function

Vertices represent the "states" corresponding to some partial deduction about our hypotheses



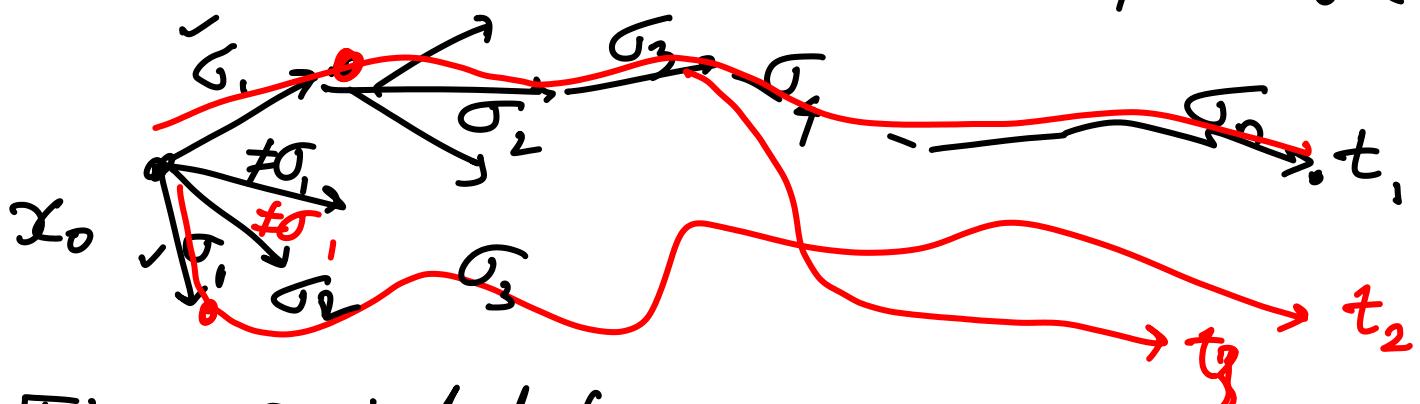
depending on the observation, say  $O$  - that can have several values  $v_1, v_2 \dots$



$w_i$  corresponds to our confidence in moving to the next state. In fact if we normalize the weights, we can make them correspond to probabilities.

Given observations  $\overbrace{O_1 O_2 O_3 \dots O_n}^{\sigma}$

$O_i$  is an observation, starting from an initial state (some starting vertex), we want to look at all the possible paths in the graph corresponding to  $\sigma$  and choose the one that is most probable



The probability of a specific path

$$P = e_1, e_2, e_3, \dots, e_n = \prod_i w(e_i)$$

Take the logarithm of the prod  
of probabilities

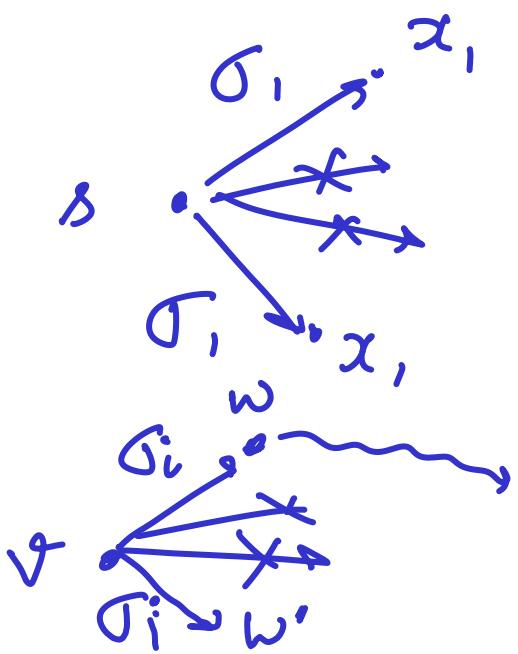
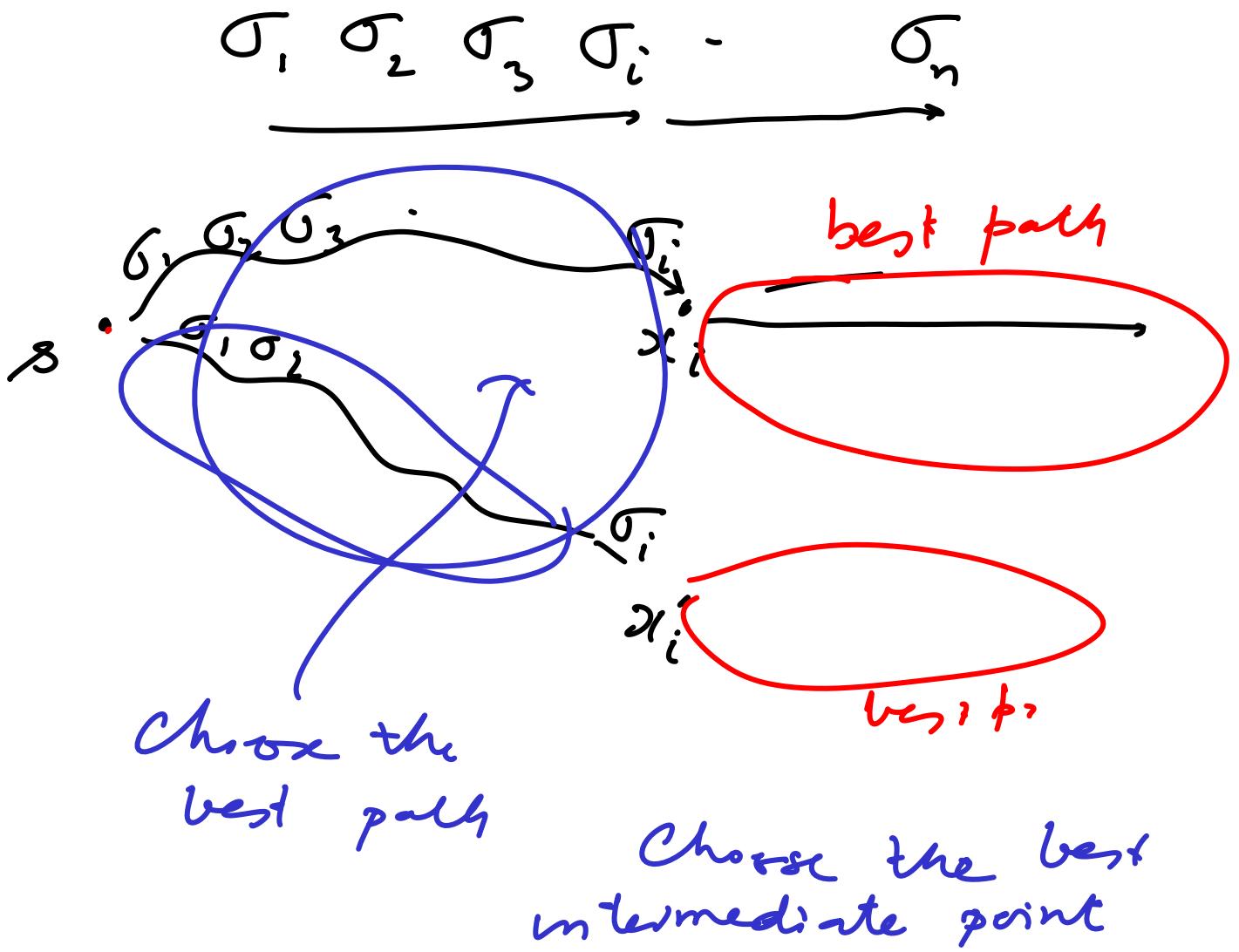
$$\log \cdot (w(e_1) \cdot w(e_2) \cdot w(e_3) \dots w(e_n))$$

↑      ↑      ↑  
correspond to prob

$$\Rightarrow \sum \underbrace{\log(w(e_i))}_{w'} \dots$$

Since logarithms of prob are negative  
we will actually minimise the  
sum of the weights (that are logs.)

Objective. Given a labelled,  
weighted graph  $G = (V, E, w)$ ; ...  
and a sequence  $\sigma$  of labels  
 $\sigma_1, \sigma_2, \sigma_3, \dots, \sigma_n$ , we want to find  
the path  $P^* = e_1^* e_2^* e_3^* \dots e_n^*$   
such that  $\sum w(e_i^*)$  is minimised  
where  $e_i^*$  has <sup>i</sup>-th label  $\sigma_i^*$



Let  $P_i(v)$  denote the best path from vertex  $v$  with labels  $\sigma_{i+1}, \sigma_{i+2}, \dots, \sigma_n$

$$P_{i-1}(v) = \min_{w \in V} [P_i(w) + W(v, w)]$$

label( $v, w$ ) =  $\sigma_i$ .

Must be calculate for all vertices  $v \in V$

Eventually or finally, we need

$P_0(s)$ . Base case paths of length 1, namely  
 $P_{n-1}(v) \quad \forall v \in V$

Running time? : We need to compute  
 $P_i(v)$  for  $0 \leq i \leq n$ ,  $v \in V$

$\Rightarrow O(n \cdot |V|)$  terms

How much time for each? : degree of  
a vertex which is  $O(|V|)$

$\Rightarrow$  Total time  $O(n|V|^2)$

Space: If all terms have to  
be stored, then  $O(n|V|)$

Viterbi's algorithm