

So far in class: (2012-11-16)

▪ Viterbi Alg.  $\arg \max_{\bar{S}} P(\bar{S} | \bar{O})$   
• Inference

▪ Forward Alg.  $P(\bar{O}) = \sum_{\bar{S}} P(\bar{O}, \bar{S})$

In order to get to Learning,  
we also need the backward  
Formulation.

▪ Backward Alg.  $P(\bar{O} | S_t = i)$

▪ Posterior Decoding:  $P(S_t = i | \bar{O})$   
$$= \frac{P(\bar{O}_{1:t}, S_t = i) P(\bar{O}_{t+1:T} | S_t = i)}{P(\bar{O}_{1:T})}$$

Learning: Baum-welch (EM)

Restate:

$$\pi_i = P(S_t = i) \quad n \text{ possible states}$$

$$a_{ij} = P(S_{t+1} = j | S_t = i)$$

$$b_{ik} = P(O_t = k | S_t = i) \quad [b_i(k) = b_{ik}]$$

Forward:  $\alpha_{it} \equiv P(\bar{O}_{1:t}, S_t = i)$

$$\alpha_{j,t+1} = \sum_{i=1}^n \alpha_{it} a_{ij} b_j(O_{t+1})$$

$$\alpha_{i1} = \pi_i b_i(O_1) \quad (\text{Basecase})$$

Backward:

$$\beta_{it} \equiv P(O_{t+1:T} | S_t = i)$$

① Basecase:  $P(\text{nothing} | S_T = i) = ?$

$$\beta_{iT} = 1$$

② Backwards:

$$\beta_{it} = P(O_{t+1:T} | S_t = i)$$

$$= \sum_{j=1}^n P(O_{t+1:T}, S_{t+1} = j | S_t = i)$$

$$= \sum_{j=1}^n P(O_{t+1:T} | S_{t+1} = j, S_t = i) P(S_{t+1} = j | S_t = i)$$

$$= \sum_{j=1}^n P(O_{t+1:T} | S_{t+1} = j) P(S_{t+1} = j | S_t = i)$$

$$= \sum_{j=1}^n \underbrace{P(O_{t+2:T} | S_{t+1} = j)}_{\beta_{j,t+1}} \underbrace{P(O_{t+1} | S_{t+1} = j)}_{b_j(O_{t+1})} \underbrace{P(S_{t+1} = j | S_t = i)}_{a_{ij}}$$

$$\text{So: } P(S_t = i | \bar{O})$$

$$= \frac{P(\bar{O}_{1:t}, S_t = i) P(\bar{O}_{(t+1):T} | S_t = i)}{P(\bar{O}_{1:T})} = \frac{P(\bar{O}_{1:T}, S_t = i)}{P(\bar{O}_{1:T})}$$

$$= \frac{\alpha_{it} \beta_{it}}{\sum_i \alpha_{iT}}$$

Q: How do you compute  
 $P(S_t = i, S_{t+1} = j | \bar{O})$  ?

A:

$$P(S_t = i, S_{t+1} = j | \bar{O}) = \frac{P(\bar{O}, S_t = i, S_{t+1} = j)}{P(\bar{O})}$$

$$= \frac{P(\bar{O}_{1:t}, S_t = i) P(S_{t+1} = j | S_t = i) P(\bar{O}_{t+1} | S_{t+1} = j) P(\bar{O}_{(t+2):T} | S_{t+1} = j)}{P(\bar{O})}$$

Aside:

$$\sum_i \underbrace{\pi_i b_i(c_1)}_{\alpha_{i1}} \beta_{i1}$$

generally:  $\sum_i \alpha_{i1} \beta_{i1}$