

# HMM Inference: The Viterbi Algorithm

HMMs: model of  $P(\bar{y}, \bar{x}) = P(y_1) P(x_1 | y_1) P(y_2 | y_1) \dots$

$$\text{Inference: } \operatorname{argmax}_{\bar{y}} P(\bar{y} | \bar{x}) = \operatorname{argmax}_{\bar{y}} \frac{P(\bar{y}, \bar{x})}{P(\bar{x})} \leftarrow \begin{array}{l} \text{constant} \\ \text{w.r.t. } \bar{y} \end{array}$$

$$= \operatorname{argmax}_{\bar{y}} P(\bar{y}, \bar{x}) = \operatorname{argmax}_{\bar{y}} \log P(\bar{y}, \bar{x})$$

$$= \operatorname{argmax}_{\tilde{y}_1, \dots, \tilde{y}_n} \log P(\tilde{y}_1) + \log P(x_1 | \tilde{y}_1) + \log P(\tilde{y}_2 | \tilde{y}_1) + \dots$$

# Viterbi Dynamic Program

Define  $v_i(\tilde{y}) = n \times |\mathcal{T}|$   <sup>$n$  sent len</sup>  $|\mathcal{T}|$  number of tags  
score of the best path ending in  $\tilde{y}$  at time  $i$

$$\text{Base: } v_1(\tilde{y}) = \log P(x_1 | \tilde{y}) + \log P(\tilde{y})$$

$$\text{Recurrence: } v_i(\tilde{y}) = \log P(x_i | \tilde{y}) + \max_{\tilde{y}_{\text{prev}}} \log P(\tilde{y} | \tilde{y}_{\text{prev}}) + v_{i-1}(\tilde{y}_{\text{prev}})$$

Viterbi for  $i=1 \dots n$

for  $\tilde{y}$  in  $\mathcal{T}$ :

compute  $v_i(\tilde{y})$

Compute  $v_{n+1}(\text{STOP})$ , this =  $\max_{\tilde{y}} \log P(\bar{x}, \tilde{y})$

Track "backpointers"

Example

log probs

$$S = \begin{matrix} N \\ V \end{matrix} \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$T = \begin{matrix} N & V & \text{STOP} \\ N & -2 & -1 & -1 \\ V & -1 & -1 & -2 \end{matrix}$$

they can fish

$$E = \begin{matrix} N \\ V \end{matrix} \begin{bmatrix} -1 & -3 & -1 \\ -3 & -1 & -1 \end{bmatrix}$$

$V_i(\bar{y})$

they can can fish STOP

N

	they	can	can	fish	STOP
N	-2	$\star$ $-2 - 2 - 3 = -7$ <del>prev tr en</del> <del><math>-4 - 1 - 3 = -8</math></del>			
V	-4	$-2 - 1 - 1 = -4$			

STOP