

HMMs: Parameter Estimation

Labeled data $(\bar{x}^{(i)}, \bar{y}^{(i)})_{i=1}^D$

Maximize $\sum_i \log P(\bar{y}^{(i)}, \bar{x}^{(i)})$ generative (joint) likelihood

$$= \sum_i \log P(y_i^{(i)}) + \sum_{\substack{\text{training} \\ \text{data}}} \sum_j \log P(x_j^{(i)} | y_j^{(i)}) + \sum_i \sum_j \log P(y_j^{(i)} | y_{j-1}^{(i)})$$

training data sent index

MLE with frequency counts: Biased coin w/prob p of H
HHHT $3/4 = \arg \max_p 3 \log p + \log(1-p)$

HMM param estimation: Count + normalizing

Example

$$\mathcal{T} = \{N, V, STOP\} \quad \mathcal{V} = \{they, can, fish\}$$

Data

N	V	STOP
they	can	
N	V	STOP
they	fish	

$$S = \begin{matrix} N \\ \downarrow \\ V \end{matrix} \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

$$T = \begin{matrix} & N & V & STOP \\ N & 0 & 2 & 0 \\ V & 0 & 0 & 2 \end{matrix}$$

↓

0	1	0
0	0	1

$$E = \begin{matrix} & N & V \\ V & 2 & 1 \end{matrix}$$

they can fish

t ↓ c f

1	0	0
0	1/2	1/2

Smoothing add counts to avoid 0s

$$T \rightarrow \begin{bmatrix} 1 & 3 & 1 \\ 1 & 1 & 3 \end{bmatrix} \xrightarrow{+1 \text{ everywhere}} \begin{matrix} 1/5 & 3/5 & 1/5 \\ 1/5 & 1/5 & 3/5 \end{matrix}$$

NOT $P(V|can)$

$P(can|V)$

$$V-V \downarrow \begin{matrix} 3/5 & 1/5 & 3/5 \\ 1/2 & 1/2 & \end{matrix}$$

Ex

N	V	V
they	can	fish

$$\Rightarrow P(\bar{y}, \bar{x}) =$$