

Hidden Markov models

- In an acoustic model computes the probability of feature vector sequences under the assumption that particular word sequence had produced these vectors.

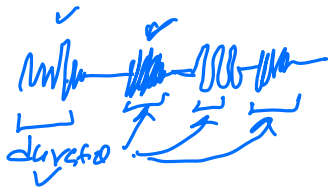
- Variation in a word or phone's pronunciation manifests in:
1. duration 2. spectral content (freq.)
1+2 = acoustic observation.

$$p(w/x) = \frac{p(x/w) * p(w)}{p(x)}$$

← language model

← acoustic model

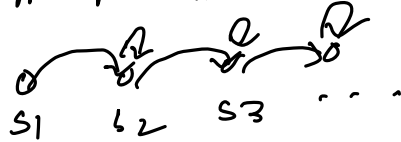
x dropped, being common denominator



Exercise: Today in your phone, record the voice & see wave forms.

So in speech recognition (acoustic model) there are two processes:

X - time varies (temporal variability)



no so easy

y - spectral variability (freq.) - can be easily detected

$x \rightarrow$ is chain of sequences $s_1, s_2, \dots \rightarrow$ in process is called

Markov chain, it is hidden.

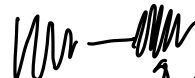
y - is observable process.

in x : $p(s_{i+1} | s_0, s_1, \dots, s_i) = p(\underline{s}_{i+1} | \underline{s}_i)$

y also depends on x . ?

θ $\underline{z}_t$ = ~~x~~

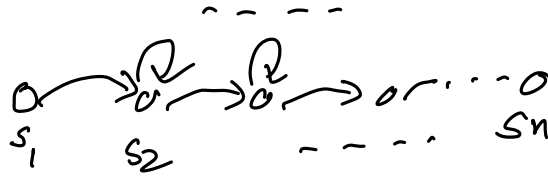
By freq. and we can estimate $\theta, \underline{z}_t$, etc. (i.e. x)


freq = y

X & Y both are probabilistic, $\therefore$

this model is double probabilistic i.e. double stochastic model

& HMM is chosen since it is double stochastic model



$P(s_{i+1}|s_i)$, or $p(s_j|s_i) \dots i, j \text{ prob.}$

$\checkmark A[i,j] = \begin{matrix} \begin{matrix} \uparrow \\ \downarrow \end{matrix} \begin{matrix} \leftarrow \begin{matrix} i \\ j \end{matrix} \rightarrow \end{matrix} \end{matrix} a_{ij} = \text{prob. of transition from } s_i \text{ to } s_j$

$\checkmark A = \{a_{ij} | i, j \in X\}$, $S = \{s_1, \dots, s_n\}$

o/p. prob. of o/p $B = \{b_{ij} | i, j \in X\}$ o/p distributed

prob. of initial state s_i , $\therefore p(s_1), p(s_2), \dots, p(s_n)$

$$\underline{\Pi} = \{ \pi_i \mid i \in X \} \text{ initial prob. of states}$$

$$\therefore \text{HMM: } \lambda(\tilde{S}, \tilde{M}, \tilde{A}, \tilde{B}, \tilde{\Pi}). \quad \underline{P(S_{i+1}|S_i^V)}$$

m = total no. of unique o/p symbols per state

$$a_{ij} = P(\underline{T_{i+1} = s_j} \mid \underline{T_i = s_i})$$

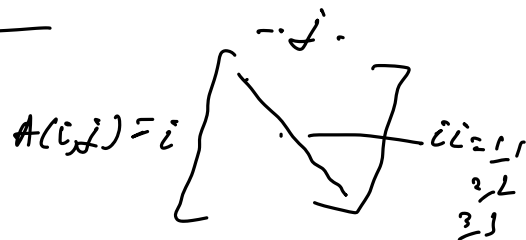
$\sim V$ = set of all possible o/p symbols = $\{o_1, o_2, \dots, o_m\}$
(phonemes)

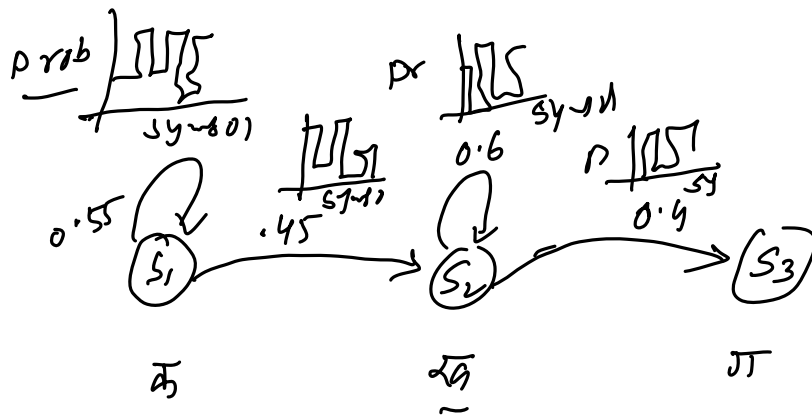
$O = (o_1, o_2, \dots, o_T)$ at different times

$$\checkmark \underline{P_L(O \mid \tilde{\Pi}, \tilde{A}, \tilde{B})} = \sum_{i=1}^T \frac{a_{i-1} b(o_i)}{q_i} \checkmark$$

Expected duration of each

$$\text{state} = \boxed{1/(1-a_{ii})}$$





discrete HMM

... $\mathcal{S}$

N no. of symbols
 $\uparrow$
 (phonemes)
 are pre-decided
IPA

