

IBM Model 1

$$\bar{a} = (a_1, \dots, a_n) \quad \bar{T} = (t_1, \dots, t_n) \quad \bar{s} = (s_1, \dots, s_m, \text{NULL})$$

each word in \bar{T}
aligns to one word
in \bar{s}

placeholder for
unaligned words

$$P(\bar{T}, \bar{a} | \bar{s}) = P(\bar{a}) P(\bar{T} | \bar{a}, \bar{s}) = \prod_{i=1}^n P(a_i) P(t_i | s_{a_i})$$

Model params: translation prob. matrix

includes $\rightarrow |V_s| \times |V_t|$ $P(\text{target} | \text{source})$
NULL

$P(t_i | s_{a_i})$: look up the prob of t_i given
source word s_{a_i} a_i is a "pointer"

$$P(a_i) = \text{uniform over } \{1, \dots, m, \text{NULL}\} \frac{1}{m+1}$$

Inference in Model 1

$$P(\bar{a} | \bar{s}, \bar{t}) = \frac{P(\bar{a}, \bar{t} | \bar{s})}{P(\bar{t} | \bar{s})} =$$

$$\frac{\prod_{i=1}^n P(a_i) P(t_i | s_{a_i})}{\cancel{P(\bar{t} | \bar{s})} \text{ ignore}}$$

$\xrightarrow{\frac{1}{m+1}}$
 $\xrightarrow{\frac{1}{m+1}}$

$$P(\bar{a} | \bar{s}, \bar{t}) \propto \prod_i P(t_i | s_{a_i})$$

		I	like	eat	\bar{t}
\bar{s}	\bar{t}_e	0.8	0.1	0.1	
	\bar{t}'	0.8	0.1	0.1	
	mange	0	0	1.0	
	aime	0	1.0	0	
	NULL	0.4	0.3	0.4	

$$P(a_i | \bar{s}, \bar{t}) \propto P(t_i | s_{a_i})$$

\bar{t}'	aime	NULL
I	like	

$$P(a_i | \bar{s}, \bar{t})$$

$$\propto \begin{cases} 0.8 & \bar{t}' & \frac{2}{3} \bar{t}' \\ 0 & \text{aime} \rightarrow & 0 \text{ aime} \\ 0.4 & \text{NULL} & \frac{1}{3} \text{NULL} \end{cases}$$

HMM Model (Vogel, 1996)

$$P(\bar{a}) = \prod_{i=1}^n P(a_i | a_{i-1})$$

"Categorical ($a_i - a_{i-1}$)"



"moving the alignment pointer by +1 is most likely"

Learning No labeled alignment data

$$\text{Optimize : } \sum_{i=1}^D \log \underbrace{P(\bar{F} | \bar{S})}_{\text{model: } P(\bar{F}, \bar{a} | \bar{S})} = \sum_{i=1}^D \log \overbrace{\sum_{\bar{a}} P(\bar{F}, \bar{a} | \bar{S})}^{\text{marginal log likelihood}}$$

Expectation maximization (EM)