



Natural Language Processing

Anoop Sarkar

anoopsarkar.github.io/nlp-class

Simon Fraser University

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Part 1: Generative Models for Word Alignment

Statistical Machine Translation

Generative Model of Word Alignment

Word Alignments: IBM Model 3

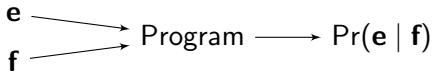
Word Alignments: IBM Model 1

Statistical Machine Translation

Noisy Channel Model

$$\mathbf{e}^* = \arg \max_{\mathbf{e}} \underbrace{\Pr(\mathbf{e})}_{\text{Language Model}} \cdot \underbrace{\Pr(\mathbf{f} | \mathbf{e})}_{\text{Alignment Model}}$$

Alignment Task



Training Data

- **Alignment Model:** learn a mapping between **f** and **e**.
Training data: lots of translation pairs between **f** and **e**.

Statistical Machine Translation

The IBM Models

- ▶ The first statistical machine translation models were developed at IBM Research (Yorktown Heights, NY) in the 1980s
- ▶ The models were published in 1993:
Brown et. al. The Mathematics of Statistical Machine Translation. *Computational Linguistics*. 1993.
<http://aclweb.org/anthology/J/J93/J93-2003.pdf>
- ▶ These models are the basic SMT models, called:
IBM Model 1, IBM Model 2, IBM Model 3, IBM Model 4, IBM Model 5
as they were called in the 1993 paper.
- ▶ We use **e** and **f** in the equations in honor of their system which translated from French to English.
Trained on the Canadian Hansards (Parliament Proceedings)

Statistical Machine Translation

Generative Model of Word Alignment

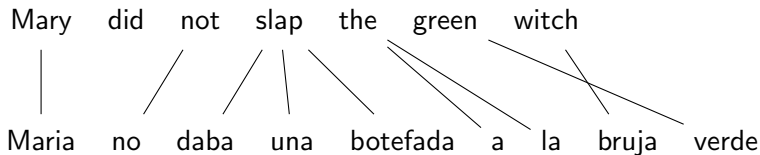
Word Alignments: IBM Model 3

Word Alignments: IBM Model 1

Generative Model of Word Alignment

- ▶ English **e**: Mary did not slap the green witch
- ▶ “French” **f**: Maria no daba una botefada a la bruja verde
- ▶ Alignment **a**: $\{1, 3, 4, 4, 4, 5, 5, 7, 6\}$
e.g. $(f_8, e_{a_8}) = (f_8, e_7) = (\text{bruja}, \text{witch})$

Visualizing alignment **a**



Generative Model of Word Alignment

Data Set

- ▶ Data set \mathcal{D} of N sentences:

$$\mathcal{D} = \{(\mathbf{f}^{(1)}, \mathbf{e}^{(1)}), \dots, (\mathbf{f}^{(N)}, \mathbf{e}^{(N)})\}$$

- ▶ French \mathbf{f} : (f_1, f_2, \dots, f_I)
- ▶ English \mathbf{e} : (e_1, e_2, \dots, e_J)
- ▶ Alignment \mathbf{a} : (a_1, a_2, \dots, a_I)

Generative Model of Word Alignment

Find the best alignment for each translation pair

$$\mathbf{a}^* = \arg \max_{\mathbf{a}} \Pr(\mathbf{a} \mid \mathbf{f}, \mathbf{e})$$

Chain rule revisited

$$\begin{aligned}\Pr(\mathbf{f}, \mathbf{a}) &= \Pr(\mathbf{f}) \Pr(\mathbf{a} \mid \mathbf{f}) \\ \Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e}) &= \Pr(\mathbf{f} \mid \mathbf{e}) \Pr(\mathbf{a} \mid \mathbf{f}, \mathbf{e})\end{aligned}$$

Alignment probability

$$\Pr(\mathbf{a} \mid \mathbf{f}, \mathbf{e}) = \frac{\Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e})}{\Pr(\mathbf{f} \mid \mathbf{e})} = \frac{\Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e})}{\sum_{\mathbf{a}} \Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e})}$$

Statistical Machine Translation

Generative Model of Word Alignment

Word Alignments: IBM Model 3

Word Alignments: IBM Model 1

Word Alignments: IBM Model 3

Generative “story” for $P(\mathbf{a} \mid \mathbf{f}, \mathbf{e})$



Word Alignments: IBM Model 3

Fertility parameter

$$n(\phi_j \mid e_j) : n(3 \mid \textit{slap})$$

Translation parameter

$$t(f_i \mid e_j) : t(\textit{bruja} \mid \textit{witch})$$

Distortion parameter

$$d(f_{pos} \mid e_{pos}, I, J) : d(8 \mid 7, 8, 6)$$

Statistical Machine Translation

Generative Model of Word Alignment

Word Alignments: IBM Model 3

Word Alignments: IBM Model 1

Word Alignments: IBM Model 1

Alignment probability

$$\Pr(\mathbf{a} \mid \mathbf{f}, \mathbf{e}) = \frac{\Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e})}{\sum_{\mathbf{a}} \Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e})}$$

Example alignment

1	2	3	4
das	Haus	ist	klein
1	2	3	4
the	house	is	small

$$\Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e}) = \prod_{i=1}^I t(f_i \mid e_{a_i})$$

$$\begin{aligned} \Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e}) = & t(\text{das} \mid \text{the}) \times \\ & t(\text{Haus} \mid \text{house}) \times \\ & t(\text{ist} \mid \text{is}) \times \\ & t(\text{klein} \mid \text{small}) \end{aligned}$$

Word Alignments: IBM Model 1

Sum over all alignments

$$\sum_{\mathbf{a}} \Pr(\mathbf{f}, \mathbf{a} \mid \mathbf{e}) = \sum_{a_1=1}^J \sum_{a_2=1}^J \dots \sum_{a_I=1}^J \prod_{i=1}^I t(f_i \mid e_{a_i})$$

Assume (f_1, f_2, f_3) and (e_1, e_2)

$$\sum_{a_1=1}^2 \sum_{a_2=1}^2 \sum_{a_3=1}^2 t(f_1 \mid e_{a_1}) \times t(f_2 \mid e_{a_2}) \times t(f_3 \mid e_{a_3})$$

Word Alignments: IBM Model 1

Assume (f_1, f_2, f_3) and (e_1, e_2) : $I = 3$ and $J = 2$

$$\sum_{a_1=1}^2 \sum_{a_2=1}^2 \sum_{a_3=1}^2 t(f_1 | e_{a_1}) \times t(f_2 | e_{a_2}) \times t(f_3 | e_{a_3})$$

$J' = 2^3$ terms to be added:

$t(f_1 e_1)$	\times	$t(f_2 e_1)$	\times	$t(f_3 e_1)$	$+$
$t(f_1 e_1)$	\times	$t(f_2 e_1)$	\times	$t(f_3 e_2)$	$+$
$t(f_1 e_1)$	\times	$t(f_2 e_2)$	\times	$t(f_3 e_1)$	$+$
$t(f_1 e_1)$	\times	$t(f_2 e_2)$	\times	$t(f_3 e_2)$	$+$
$t(f_1 e_2)$	\times	$t(f_2 e_1)$	\times	$t(f_3 e_1)$	$+$
$t(f_1 e_2)$	\times	$t(f_2 e_1)$	\times	$t(f_3 e_2)$	$+$
$t(f_1 e_2)$	\times	$t(f_2 e_2)$	\times	$t(f_3 e_1)$	$+$
$t(f_1 e_2)$	\times	$t(f_2 e_2)$	\times	$t(f_3 e_2)$	$+$

Word Alignments: IBM Model 1

Factor the terms:

$$\begin{array}{l} (t(f_1 | e_1) \times t(f_2 | e_1)) \times (t(f_3 | e_1) + t(f_3 | e_2)) + \\ (t(f_1 | e_1) \times t(f_2 | e_2)) \times (t(f_3 | e_1) + t(f_3 | e_2)) + \\ (t(f_1 | e_2) \times t(f_2 | e_1)) \times (t(f_3 | e_1) + t(f_3 | e_2)) + \\ (t(f_1 | e_2) \times t(f_2 | e_2)) \times (t(f_3 | e_1) + t(f_3 | e_2)) \end{array}$$

$$(t(f_3 | e_1) + t(f_3 | e_2)) \left(\begin{array}{l} t(f_1 | e_1) \times t(f_2 | e_1) + \\ t(f_1 | e_1) \times t(f_2 | e_2) + \\ t(f_1 | e_2) \times t(f_2 | e_1) + \\ t(f_1 | e_2) \times t(f_2 | e_2) \end{array} \right)$$

$$(t(f_3 | e_1) + t(f_3 | e_2)) \left(\begin{array}{l} t(f_1 | e_1) \times (t(f_2 | e_1) + t(f_2 | e_2)) + \\ t(f_1 | e_2) \times (t(f_2 | e_1) + t(f_2 | e_2)) \end{array} \right)$$

Word Alignments: IBM Model 1

Assume (f_1, f_2, f_3) and (e_1, e_2) : $I = 3$ and $J = 2$

$$\prod_{i=1}^3 \sum_{a_i=1}^2 t(f_i | e_{a_i})$$

$I \times J = 2 \times 3$ terms to be added:

$(t(f_1 e_1) + t(f_1 e_2))$	\times
$(t(f_2 e_1) + t(f_2 e_2))$	\times
$(t(f_3 e_1) + t(f_3 e_2))$	

Acknowledgements

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