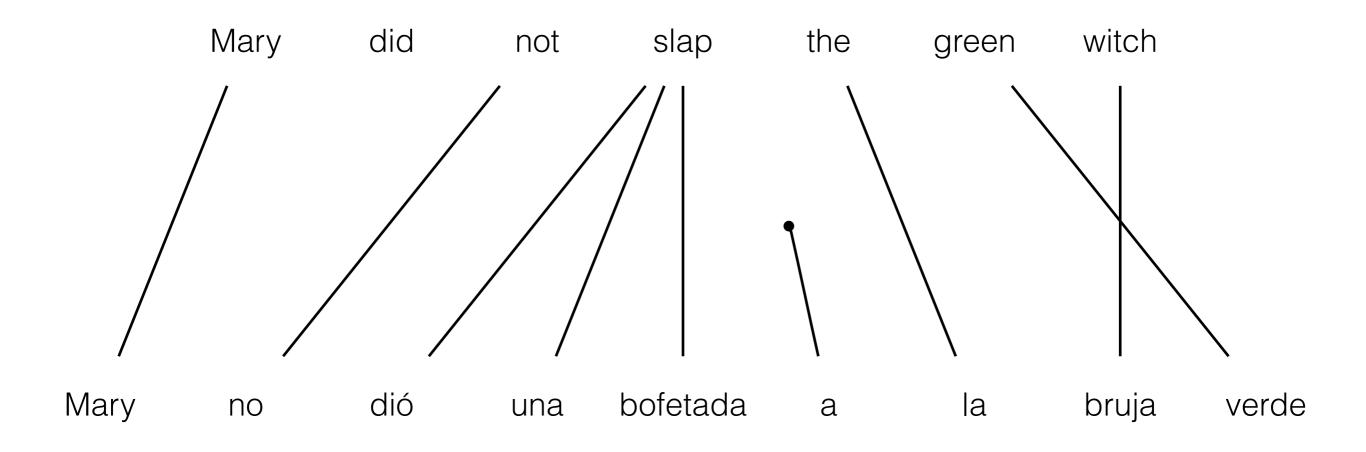
# Inversion Transduction Grammars

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# Word-based Translation



Every French word is generated by an English word (or null)

#### Generative Story IBM≥3: Given E

| Mary | did | not | slap | the | green | witch |
|------|-----|-----|------|-----|-------|-------|
|      |     |     |      |     |       |       |

#### Generative Story IBM≥3: Fertility

| Mary | did            | not |      | slap |      | the | green | witch |
|------|----------------|-----|------|------|------|-----|-------|-------|
| Mary | <del>did</del> | not | slap | slap | slap | the | green | witch |

### Generative Story IBM≥3: NULL insertion

| Mary | did            | not |      | slap |      |      | the | green | witch |
|------|----------------|-----|------|------|------|------|-----|-------|-------|
| Mary | <del>did</del> | not | slap | slap | slap |      | the | green | witch |
|      |                |     |      |      |      | NULL |     |       |       |

#### Generative Story IBM≥3: Translation

| Mary | did            | not | slap |      |          |      | the | green | witch |
|------|----------------|-----|------|------|----------|------|-----|-------|-------|
| Mary | <del>did</del> | not | slap | slap | slap     |      | the | green | witch |
|      |                |     |      |      |          | NULL |     |       |       |
| Mary |                | no  | dió  | una  | bofetada | а    | la  | verde | bruja |

### Generative Story IBM≥3: Distortion

| Mary | did            | not |      | slap |          |      | the | green | witch |
|------|----------------|-----|------|------|----------|------|-----|-------|-------|
| Mary | <del>did</del> | not | slap | slap | slap     |      | the | green | witch |
|      |                |     |      |      |          | NULL |     |       |       |
| Mary |                | no  | dió  | una  | bofetada | а    | la  | verde | bruja |
|      |                |     |      |      |          |      |     |       |       |
| Mary |                | no  | dió  | una  | bofetada | а    | la  | bruja | verde |

# Discussion

- IBM models do not constrain divergence with respect to word order
- Distortion step must consider

#### all the m! permutations

of m French words

### All permutations: sensible or not?

If we do not impose structural constraints (yet they do exist)

- the model will have to learn (rather *implicitly*)
  how not to violate them
- which ought to require more data

### Practical consequences

#### **Estimation**

 modelling outcomes that even though possible are not plausible (unlikely to be observed)

#### Generation

NP-completeness!

# NP-completeness

NP-complete problem

Generalised TSP

[Knight, 1999; Zaslavskiy et al, 2009]

Perfect matching

[DeNero and Klein, 2008]

All permutations

[Asveld, 2006; 2008]

# All permutations

Let 
$$\Sigma_n = \{a_1, ..., a_n\}$$

- $S \rightarrow A_{\Sigma_n}$
- $A_X \rightarrow a A_{X-\{a\}}$  for  $X \subseteq \Sigma_n$ ,  $\#X \ge 2$ ,  $a \in X$
- $A_{\{a\}} \rightarrow a$

Regular grammar (there is an equivalent FSA)

# Complexity

Note that nonterminals are indexed by subsets of  $\Sigma_n$ 

#### i.e. power set of $\Sigma$

- 2<sup>n</sup> nonterminals (states)
- $n \times 2^n$  productions (transitions)
- *n*! strings (paths)

# Example: 3 elements

$$S \rightarrow A_{123}$$
 $A_{123} \rightarrow a_1 A_{23} | a_2 A_{13} | a_3 A_{23}$ 
 $A_{12} \rightarrow a_1 A_2 | a_2 A_1$ 
 $A_{13} \rightarrow a_1 A_3 | a_3 A_1$ 
 $A_{23} \rightarrow a_2 A_3 | a_3 A_2$ 
 $A_1 \rightarrow a_1$ 
 $A_2 \rightarrow a_2$ 
 $A_3 \rightarrow a_3$ 

# "IBM constraint"

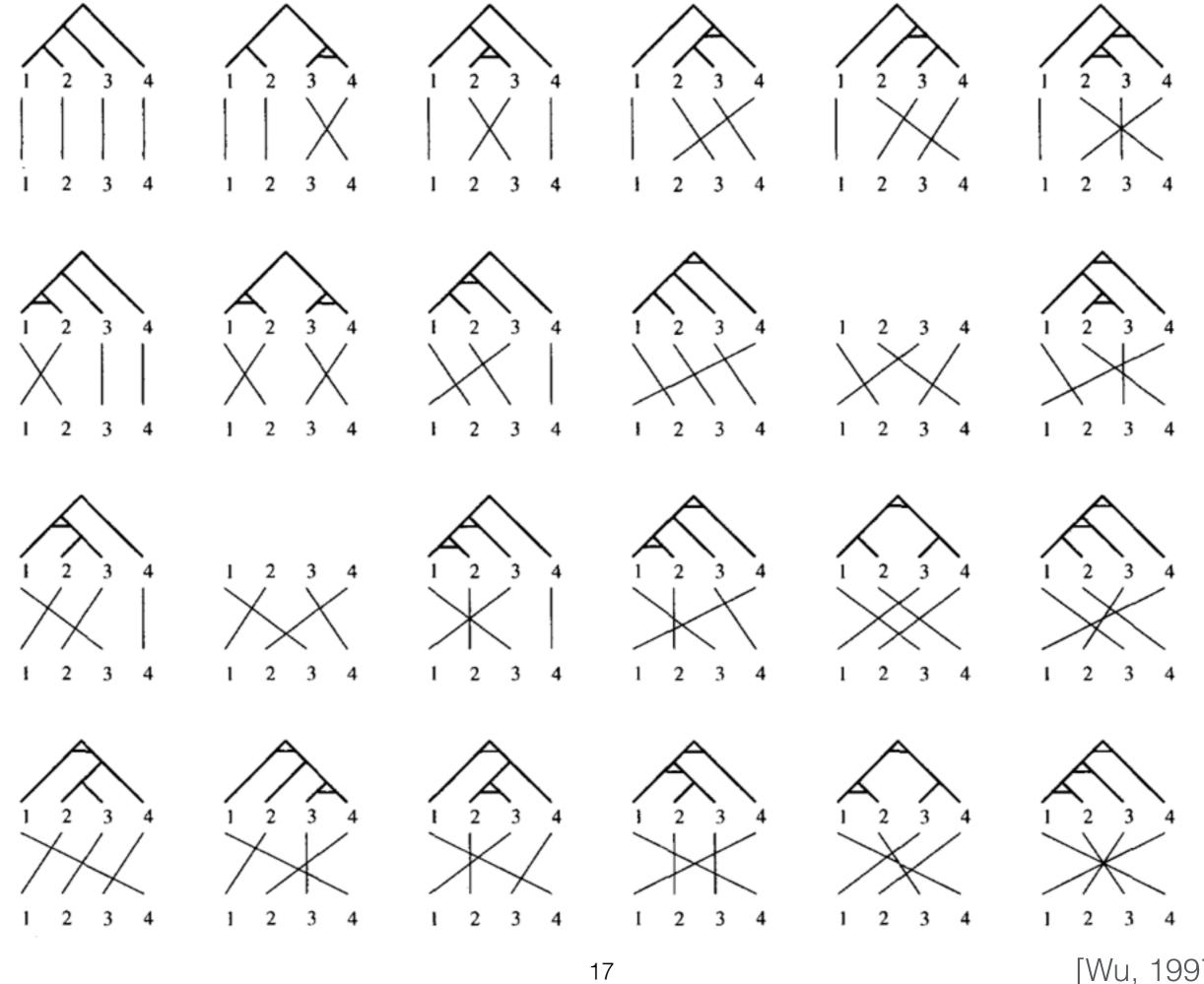
Distortion limit in generation but not in estimation

any reasons why that may be unsatisfactory?

# Constraining permutations without a distortion limit

Inversion Transduction Grammars (ITGs) [Wu, 1995; 1997]

- Binarizable permutations
  - two streams are simultaneously generated
  - context-free backbone



[Wu, 1997]

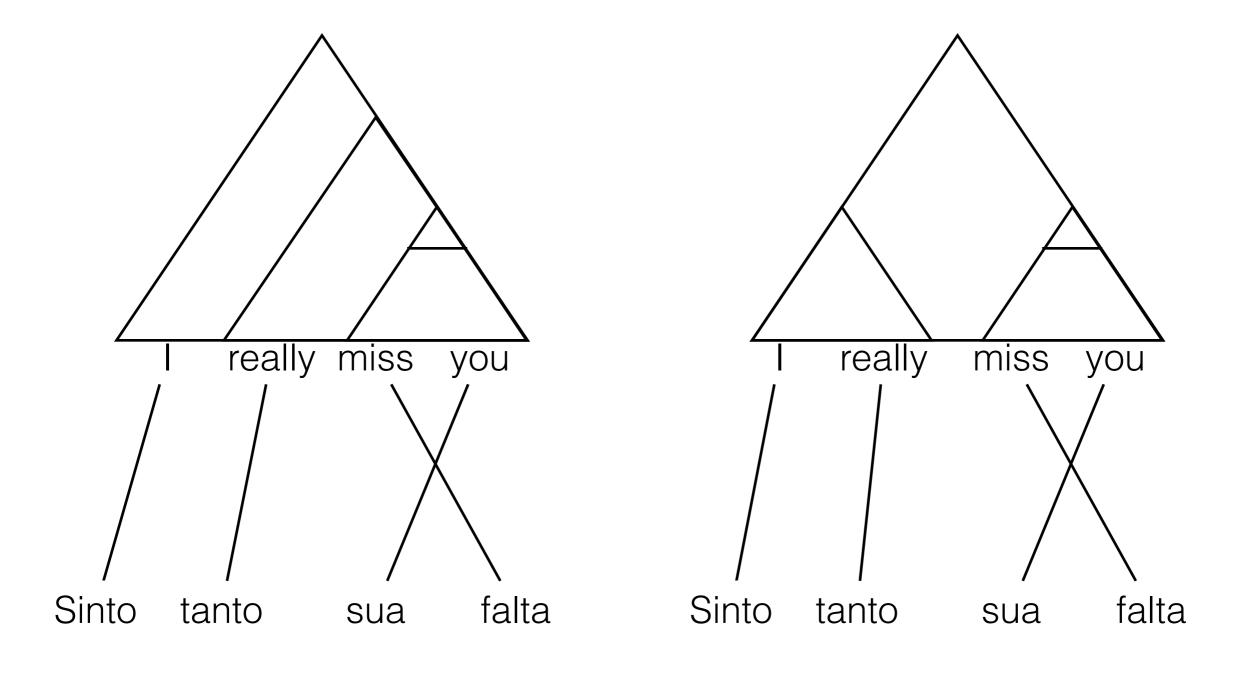
# Number of Permutations

|               | *****   |   |
|---------------|---|---|
| ITG           | all matchings   | ratio   |
| 1             | 1   | 1.000   |
| 1             | 1   | 1.000   |
| 2             | 2   | 1.000   |
| 6             | 6   | 1.000   |
| 22            | 24  | 0.917   |
| 90            | 120   | 0.750   |
| 394           | 720   | 0.547   |
| 1,806         | 5,040   | 0.358   |
| 8,558         | 40,320  | 0.212   |
| 41,586        | 362,880   | 0.115   |
| 206,098       | 3,628,800   | 0.057   |
| 1,037,718     | 39,916,800  | 0.026   |
| 5,293,446     | 479,001,600   | 0.011   |
| 27,297,738    | 6,227,020,800   | 0.004   |
| 142,078,746   | 87,178,291,200  | 0.002   |
| 745,387,038   | 1,307,674,368,000   | 0.001   |
| 3,937,603,038 | 20,922,789,888,000  | 0.000   |
|               | 6<br>22<br>90<br>394<br>1,806<br>8,558<br>41,586<br>206,098<br>1,037,718<br>5,293,446<br>27,297,738<br>142,078,746<br>745,387,038 | 1 1 1 1 1 1 2 2 2 2 4 6 6 6 6 6 22 2 24 90 120 394 720 1,806 5,040 8,558 40,320 41,586 362,880 206,098 3,628,800 1,037,718 39,916,800 5,293,446 479,001,600 27,297,738 6,227,020,800 142,078,746 87,178,291,200 745,387,038 1,307,674,368,000 |

# ITG

|                 | English   | French    |           |
|-----------------|-----------|-----------|-----------|
| $S \rightarrow$ | X         | X         | сору      |
| $X \rightarrow$ | $X_1 X_2$ | $X_1 X_2$ | сору      |
|                 |           | $X_2 X_1$ | invert    |
| $X \rightarrow$ | е         | f         | transduce |
| $X \rightarrow$ | е         | 3         | delete    |
| $X \rightarrow$ | 3         | f         | insert    |

# ITG Trees



## Model

Joint probability model P(T, E, F)

$$t = \langle r_1, \dots, r_n \rangle$$
  
 $e = \text{yield}_1(t)$   
 $f = \text{yield}_2(t)$ 

$$P(T = t, E = e, F = f) = \prod_{i=1}^{N} \theta_{r_i}$$

## Parametrisation

Multinomial: one parameter per rule

- $\theta_{[]}$  one parameter for **monotone**
- $\theta_{<>}$  one parameter for **swap**
- $\theta_{e/f}$  one parameter per word pair
- $\theta_{e/\epsilon}$  one parameter per deleted **English** word
- $\theta_{\epsilon/f}$  one parameter per inserted **French** word

# MLE

We do not typically construct treebanks of ITG trees

potential counts instead of observed counts

$$\theta_{X \to \alpha} = \frac{\langle n(X \to \alpha) \rangle_{P(T|F,E)}}{\sum_{\alpha'} \langle n(X \to \alpha') \rangle_{P(T|F,E)}}$$

Expectations from parse forests

Inside-Outside

[Baker, 1979; Lari and Young, 1990; Goodman, 1999]

Typically initialised with IBM1

# Difficulties

Inference: complexity O(I3m3)

Model: too few reordering parameters

Decisions: ambiguity

Disambiguation problem is NP-complete [Sima'an, 1996]

$$\arg \max_{A} P(A|F, E) = \arg \max_{A} \sum_{B} P(A, B|F, E)$$

$$\approx \arg \max_{A, B} P(A, B|F, E)$$

$$A, B$$
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# Bibliography

- Knight, Kevin. 1999. Decoding complexity in word-replacement translation models. In Computational Linguistics. MIT Press.
- Zaslavskiy, Mikhail and Dymetman, Marc and Cancedda, Nicola. 2009. Phrase-based statistical machine translation as a traveling salesman problem. In *Proceedings of the Joint* Conference of the 47th Annual Meeting of the ACL and the 4th International Joint Conference on Natural Language Processing of the AFNLP: Volume 1 - Volume 1.
- DeNero, John and Klein, Dan. 2008. The Complexity of Phrase Alignment Problems. In Proceedings of ACL-08: HLT.
- Asveld, Peter R. J. 2006. Generating All Permutations by Context-free Grammars in Chomsky Normal Form. In *Theoretical Computer Science*. Elsevier Science Publishers Ltd.
- Asveld, Peter R. J. 2008. Generating All Permutations by Context-free Grammars in Greibach Normal Form. In *Theoretical Computer Science*. Elsevier Science Publishers Ltd.
- Wu, D. 1995. An Algorithm for Simultaneously Bracketing Parallel Texts by Aligning Words. In Proceedings of the 33rd Annual Meeting of the Association for Computational Linguistics. ACL.

# Bibliography

- Wu, D. 1997. Stochastic Inversion Transduction Grammars and Bilingual Parsing of Parallel Corpora. In Computational Linguistics. MIT Press.
- James K. Baker. 1979. Trainable grammars for speech recognition. In Proceedings of the Spring Conference of the Acoustical Society of America.
- Karim Lari and Steve J. Young. 1990. The estimation of stochastic contextfree grammars using the inside--outside algorithm. In *Computer Speech* and Language.
- Goodman, Joshua. 1999. Semiring parsing. In Computational Linguistics.
- Sima'an, Khalil. 1996. Computational complexity of probabilistic disambiguation by means of tree-grammars. In *Proceedings of the 16th* conference on Computational linguistics - Volume 2.