



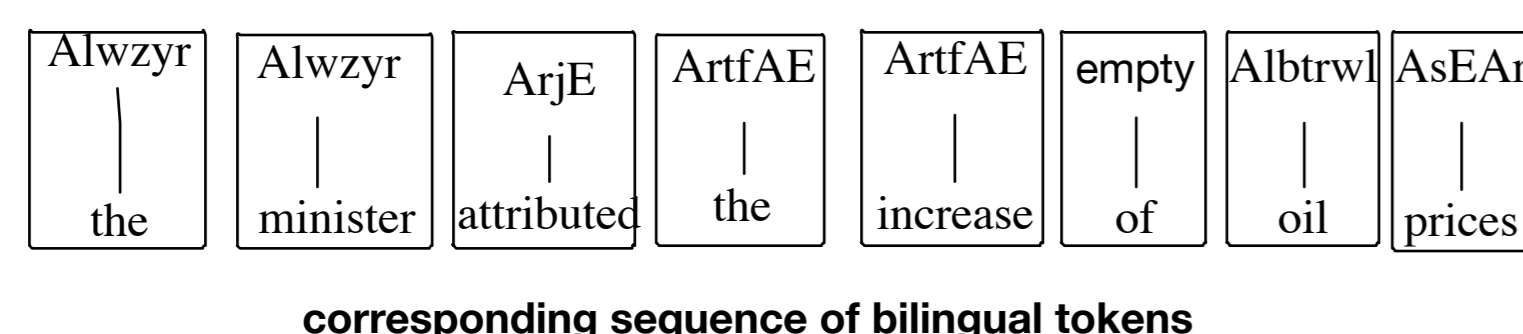
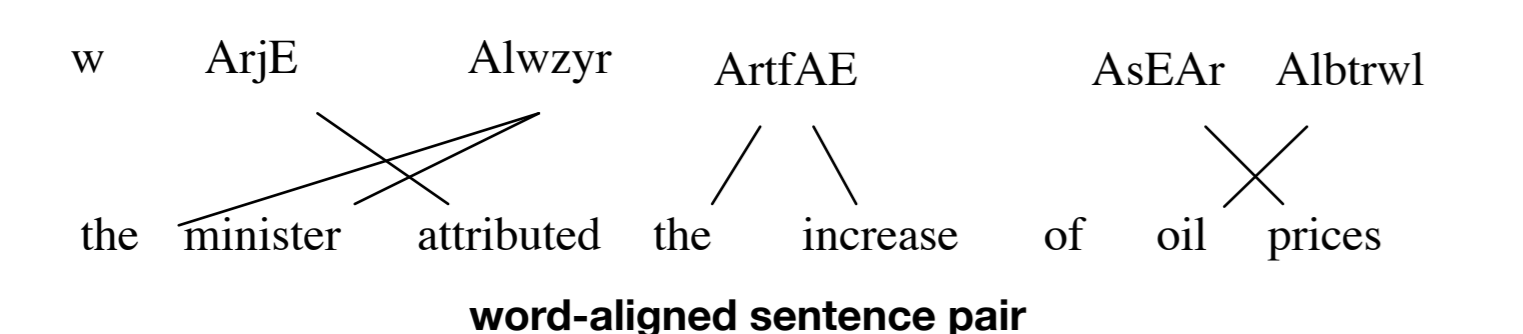
## Contributions

- capture reordering as an order of a sequence of translation events
- characterize translation events with their source and target syntactic features
- simple alternative to tree-based models
- up to +0.98 BLEU improvement for Chinese-English and +0.4 BLEU improvement for Arabic-English over lexicalized BiLM

## Motivation

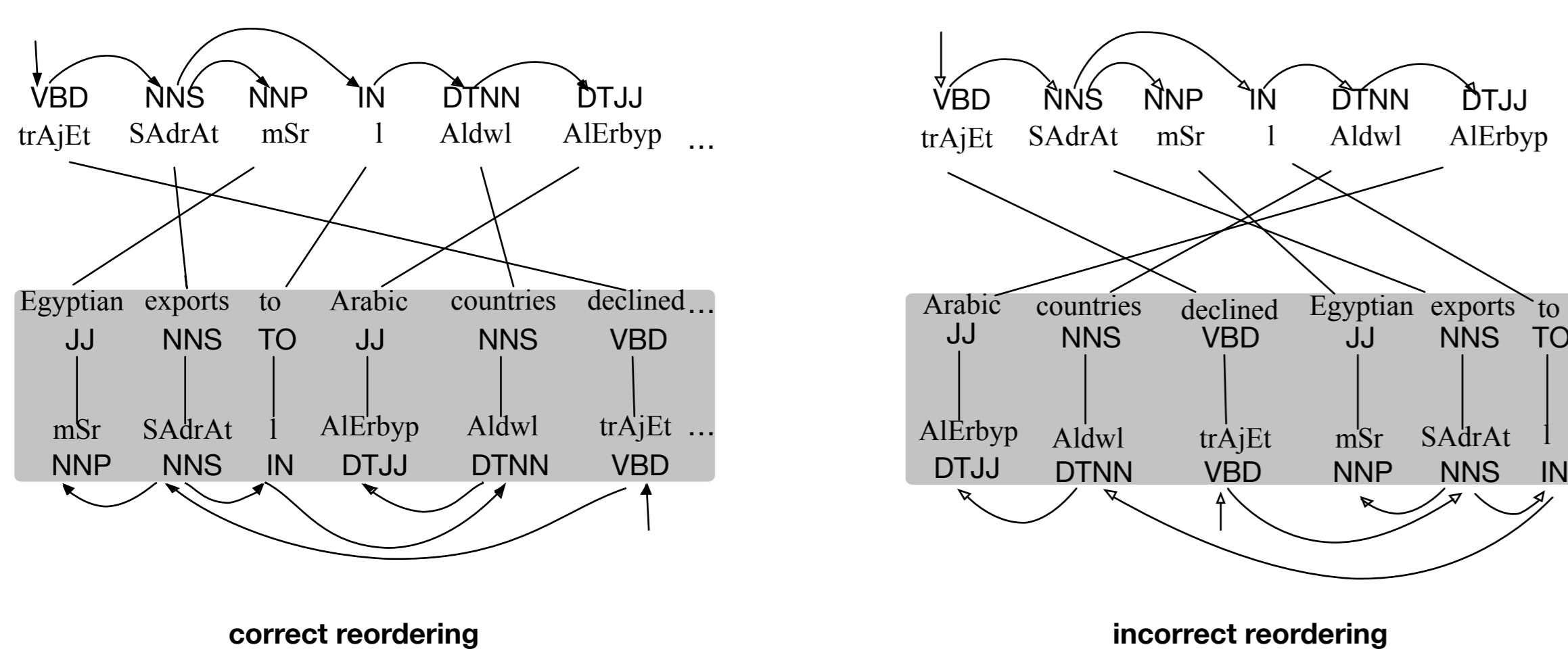
### Background: bilingual language models (BiLMs)

- n-gram model of sequences of elementary translation events
- elementary translation event — a pair of source and target words
- we adopt the definition of Niehues et al. (2011) of a **bilingual token**: (given word alignment) a target word and all the source words aligned to it



### Reordering with BiLMs

How well do various labelings differentiate between correct and incorrect reorderings?



### expressiveness

- lexical-based BiLMs are too sparse to capture reordering regularities

vs.

### generality

- Niehues et al. (2011): bilingual tokens substituted with the words' POS tags

## Approach

### Dependency-based BiLMs

- dependency grammar is commonly used in NLP to formalise role-based relations between words
- to label bilingual tokens, we try out combinations of different properties based on a source dependency parse
- generalized definition of a labeling for a bilingual token sequence  $t_1 \dots t_n$

$$t_i = \langle e_i, \{ContF(f) | f \in A(e_i)\} \rangle$$

where  $e_i$  is the  $i$ -th target word,  $A: E \rightarrow \mathcal{P}(F)$  is an alignment function,  $F$  and  $E$  — source and target sentences,  $ContF$  and  $ContE$  - **contextual functions**

### Contextual functions

Return a word's sentential context (source or target).

We focus on  $ContF$ 's, since they allow for a richer set of definitions in the MT setting (source side fully given) than  $ContE$ 's.

### Proposed contextual functions return:

- the word itself (designation: Lex)
- POS tag of the word (designation: Pos)
- POS tag of the word's parent (see below)
- POS tag of the word's grandparent (see below)

### Notation

We use the proposed contextual functions in combinations to define individual BiLMs.

“•” horizontally connects source (left) and target (right) contextual functions

Lex•Lex is a BiLM with lexicalized tokens

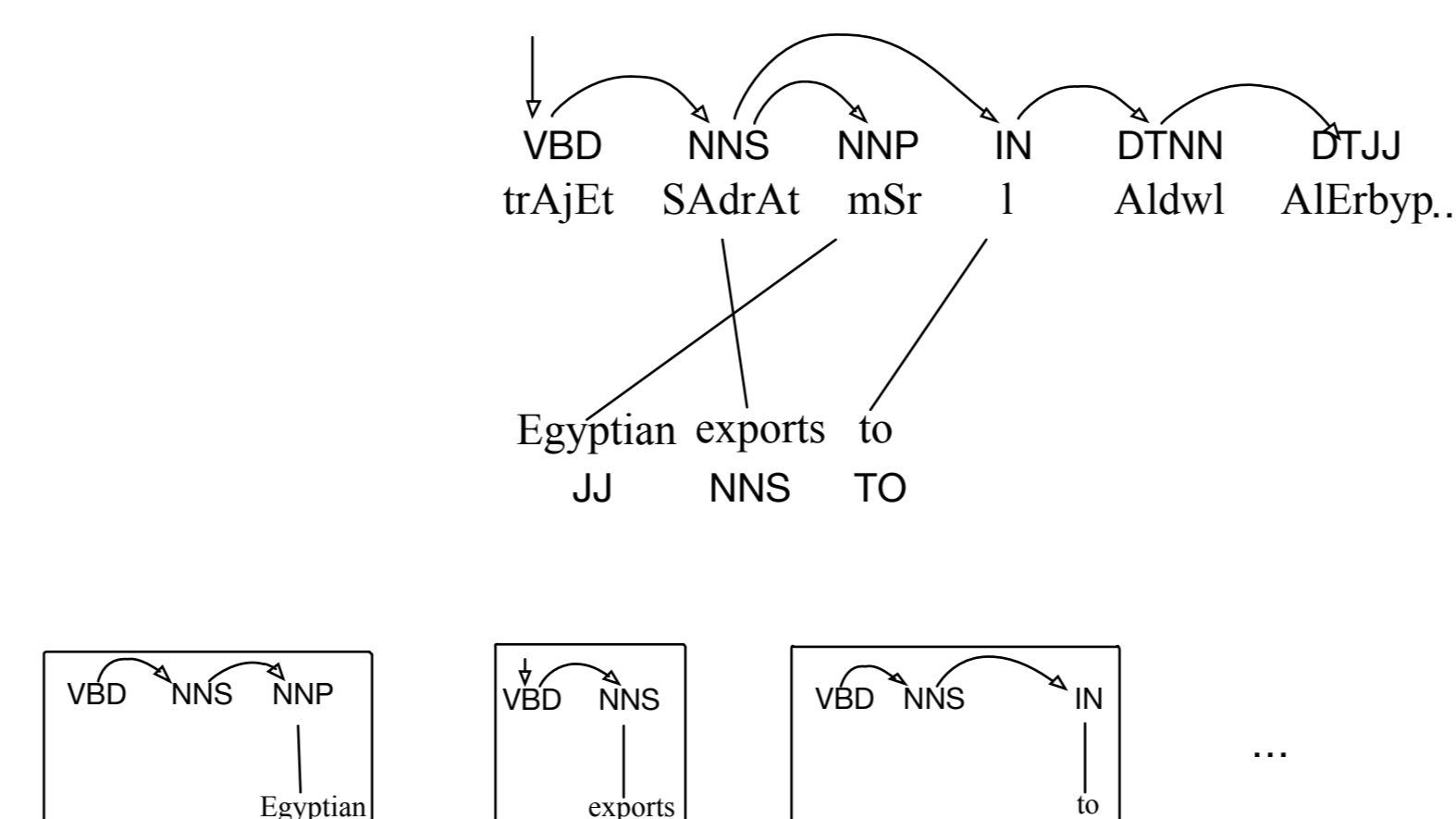
Pos•Pos is a BiLM with words in tokens substituted with their POS tags

“→” connects parent (left) and child (right) from a dependency tree

Pos→Pos→Pos is a combination of functions returning the word's POS tag, its parent's POS tag and its grandparent's POS tag

### Example

Sequence of bilingual tokens produced by a Pos→Pos→Pos•Lex after translating three words of the source sentence:



## Implementation and Experiments

### Implementation and integration into PBSMT decoder

- employed tools: Chinese Stanford dependency parser (Chang et al. 2009), Arabic Stanford constituency parser (Green and Manning, 2010) + extract dependencies based on Collins (1999), English Stanford POS-tagger (Toutanova et al., 2003)
- n-gram model training: 5-gram model with Kneser-Ney smoothing using SRILM (Stolke et al., 2011)
- dependency-based BiLMs are integrated as a feature in a log-linear model
- for each phrase pair, its most likely internal word alignment and target-side POS labelling is stored in the phrase table

### Basic experimental setup

- phrase-based decoder
- distortion limit: 5
- lexicalized distortion model included in the log-linear interpolation
- compare performance of the original BiLMs (Niehues et al. 2011) and the dependency-based BiLMs

### Statistical significance notation

- ▲ — significant improvement over Lex•Lex at  $p < .01$
- △ — significant improvement over Lex•Lex at  $p < .05$

### Arabic-English experiments

	MT08+MT09 test set	
	BLEU	TER
PBSMT baseline	46.57	45.60
Lex•Lex	46.98	45.96
Pos→Pos•Pos	47.25 <sup>▲</sup> +0.27	45.40 <sup>▲</sup>
Pos→Pos→Pos•Pos	47.30 <sup>▲</sup> +0.32	46.21 <sup>▲</sup>
Lex•Lex + Pos→Pos→Pos•Pos	<b>47.38<sup>▲</sup></b> +0.4	45.63 <sup>▲</sup>

### Chinese-English experiments

	MT06+MT08 test set	
	BLEU	TER
PBSMT baseline	28.99	59.14
Lex•Lex	29.69	58.72
Pos→Pos•Pos	29.78 +0.09	58.36 <sup>▲</sup>
Pos→Pos→Pos•Pos	30.05 <sup>▲</sup> +0.36	58.54
Lex•Lex + Pos→Pos→Pos•Pos	<b>30.28<sup>▲</sup></b> +0.59	58.30 <sup>▲</sup>

### Reordering potential of the model

Distortion limit extended to 10 words.

	Arabic-English MT08+MT09 test set		Chinese-English MT06+MT08 test set	
	BLEU	TER	BLEU	TER
Lex•Lex	46.72	45.97	29.79	58.38
Pos→Pos→Pos•Pos	<b>47.12<sup>▲</sup></b> +0.4	45.52 <sup>▲</sup>	<b>30.77<sup>▲</sup></b> +0.98	57.82 <sup>▲</sup>

