# Tailoring Word Alignments to Syntactic Machine Translation

John DeNero and Dan Klein



Presentation and paper: http://nlp.cs.berkeley.edu/pages/WordAligner.html



# Tailoring Word Alignments to Syntactic Machine Translation

Setting:

Syntactic MT with tree transducers

**Problem:** 

Alignment errors that contradict constituent structure impede the rule extraction process

Proposal:

Condition word alignment on syntactic structure



Source: Les emplois sont axés sur la carrière



Source: Les emplois sont axés sur la carrière

Gloss: The jobs are centered on the career



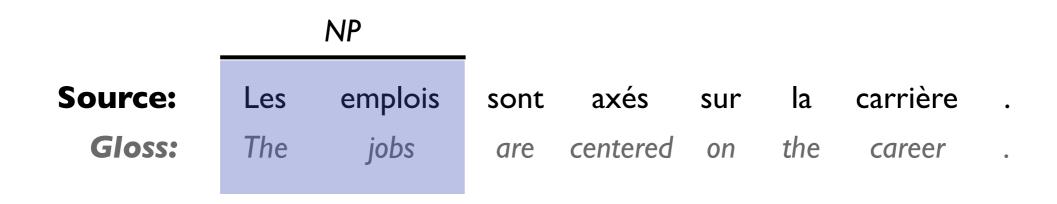
```
Source: Les emplois sont axés sur la carrière .

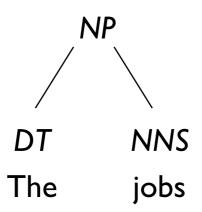
Gloss: The jobs are centered on the career .
```

**Transducer rule:** 

(NP (DT The) (NNS jobs)) => Les emplois



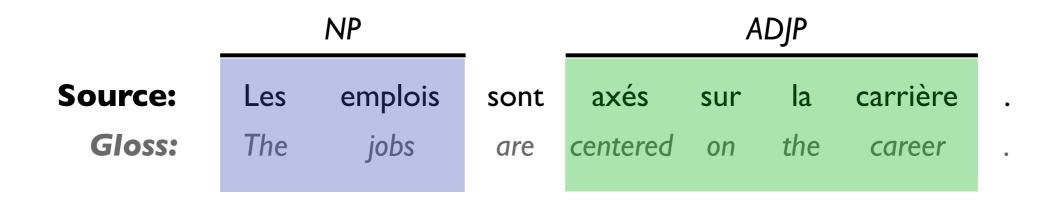




**Transducer rule:** 

(NP (DT The) (NNS jobs)) => Les emplois







Transducer rule: (ADJP (NN career) (VBN oriented)) => axés sure la carrière

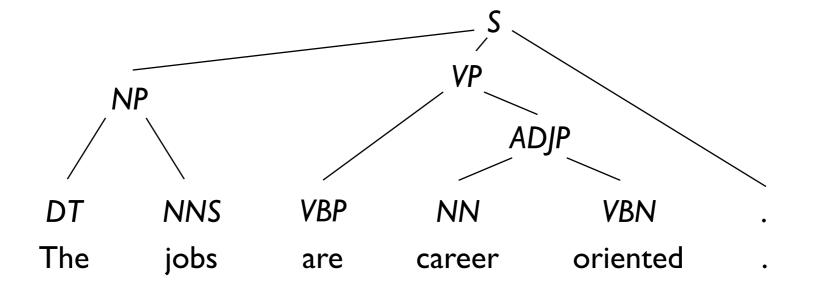
http://nlp.cs.berkeley.edu/pages/WordAligner.html



S

Source: Gloss:

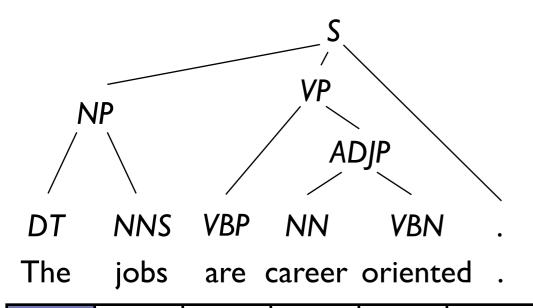
NP			ADJP				
Les	emplois	sont	axés	sur	la	carrière	
The	jobs	are	centered	on	the	career	

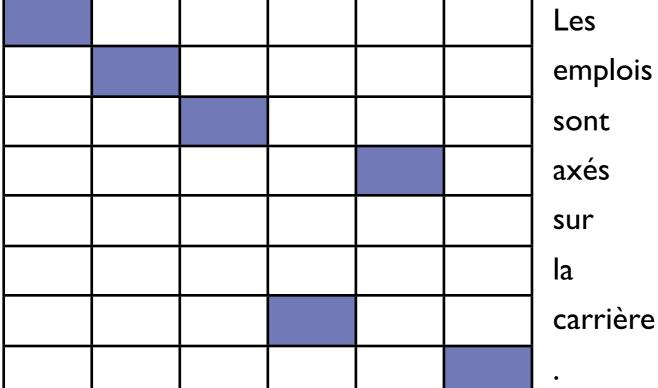


**Transducer rule:** 

(S  $NP_1$  (VP (VBP are)  $ADJP_2$ ) (. .)) =>  $NP_1$  sont  $ADJP_2$ .



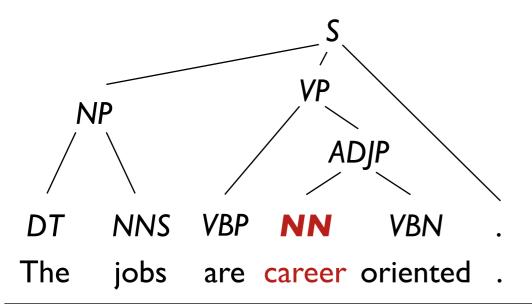


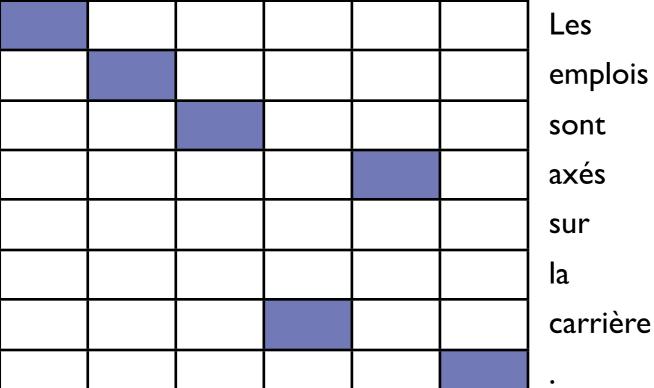


### Extraction Procedure (Galley et al., '04 & '06)

http://nlp.cs.berkeley.edu/pages/WordAligner.html





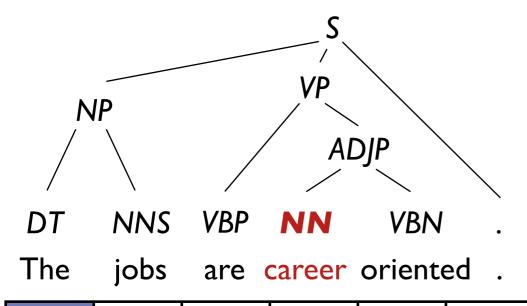


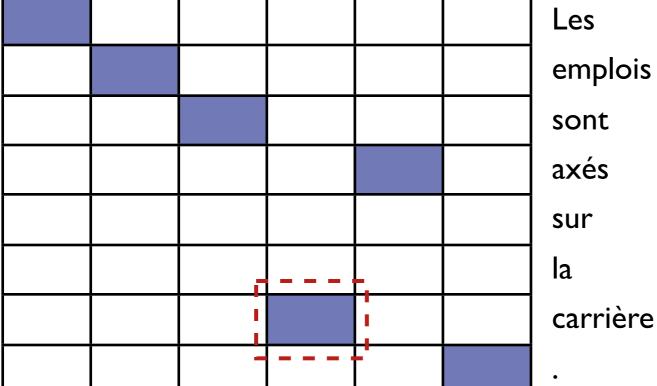
#### **Extraction Procedure**

(Galley et al., '04 & '06)

I. Choose a constituent



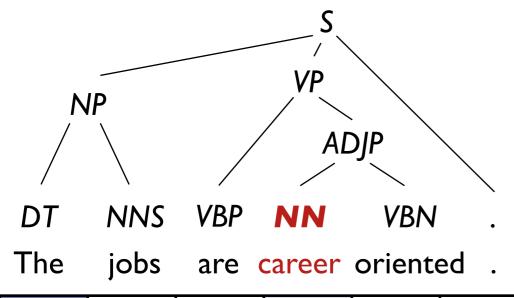


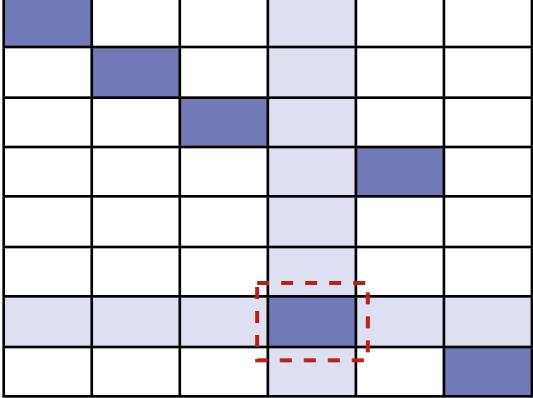


Extraction Procedure (Galley et al., '04 & '06)

- I. Choose a constituent
- 2. Choose a region around constituent alignments





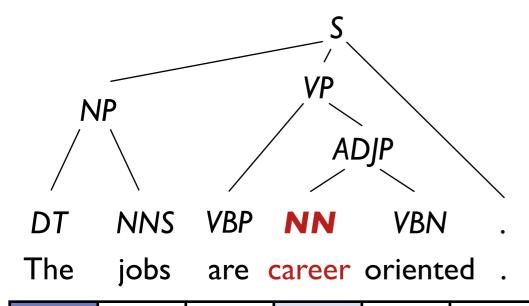


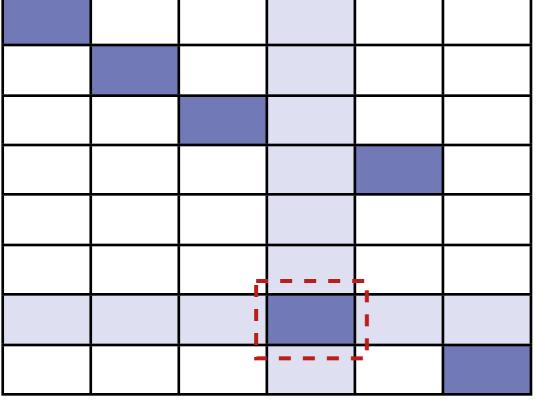
Les
emplois
sont
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### Extraction Procedure (Galley et al., '04 & '06)

- I. Choose a constituent
- 2. Choose a region around constituent alignments
- 3. Verify that alignment is consistent with region







Les
emplois
sont
axés
sur
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carrière

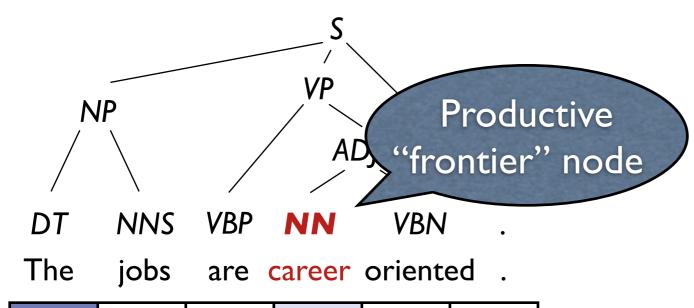
#### **Extraction Procedure**

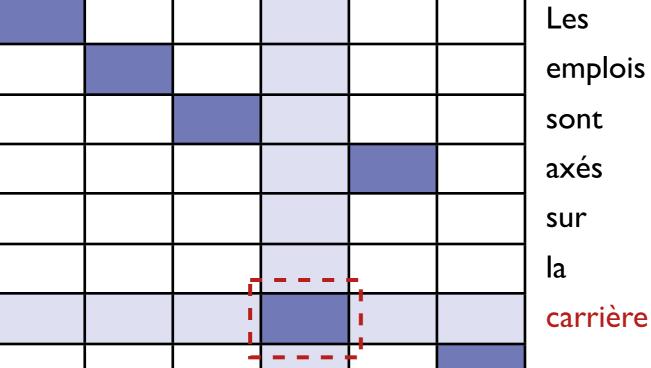
(Galley et al., '04 & '06)

- I. Choose a constituent
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- 3. Verify that alignment is consistent with region
- 4. Extract phrase:

(NN career) => carrière







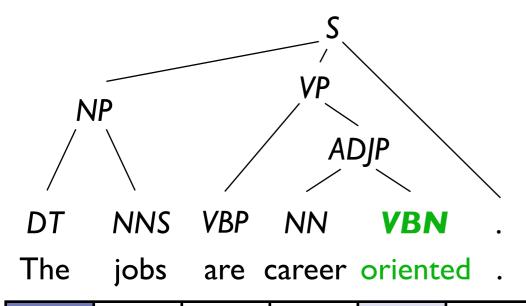
#### **Extraction Procedure**

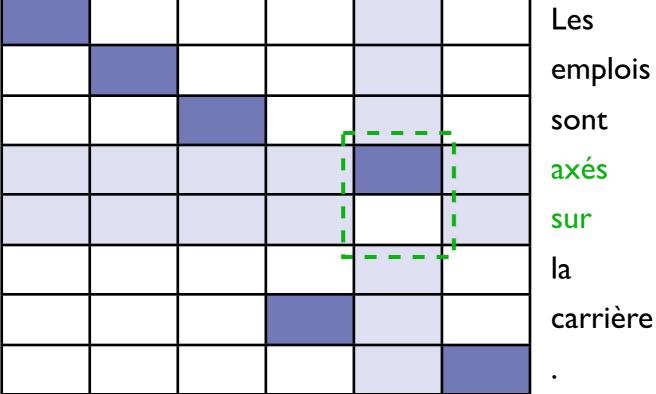
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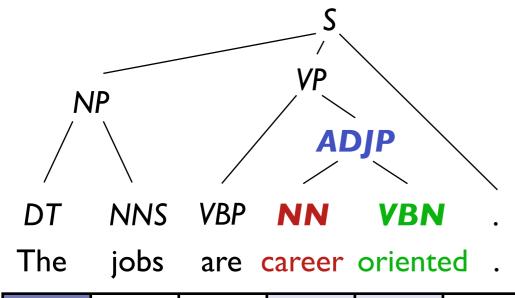


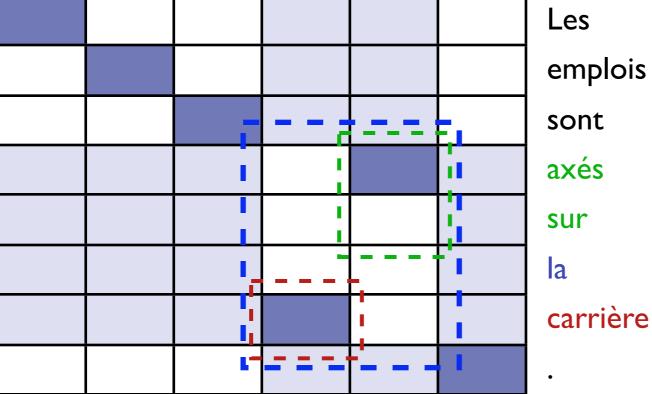
Extraction Procedure (Galley et al., '04 & '06)

- I. Choose a constituent
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(VBN oriented) => axés sur







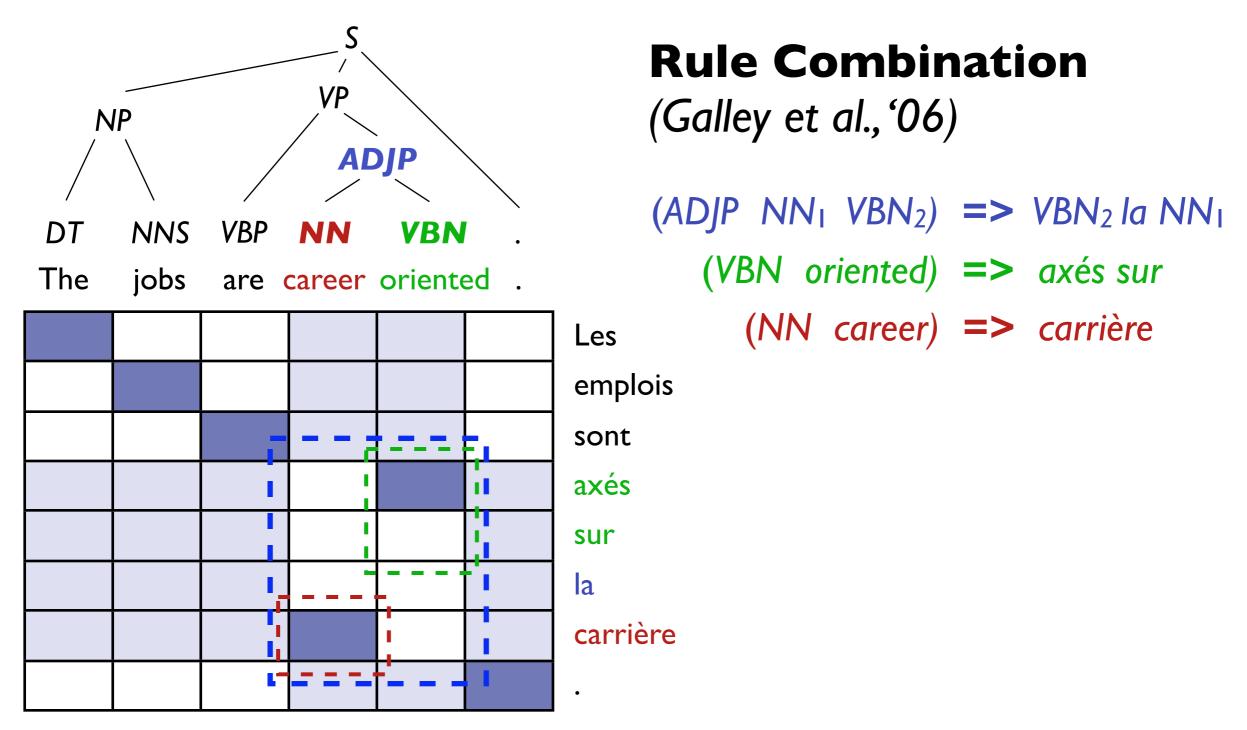
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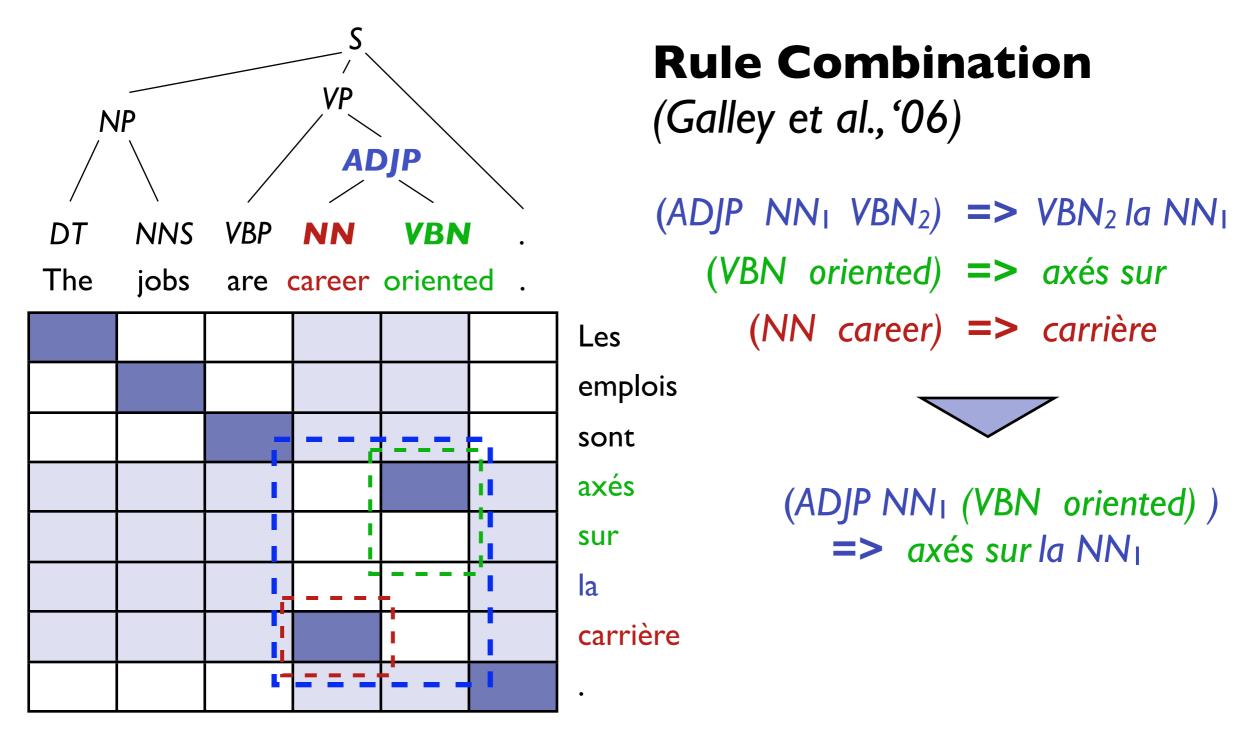
- I. Choose a constituent
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- 3. Verify that alignment is consistent with region
- 4. Extract phrase:

 $(ADJP NN_1 VBN_2) => VBN_2 Ia NN_1$ 

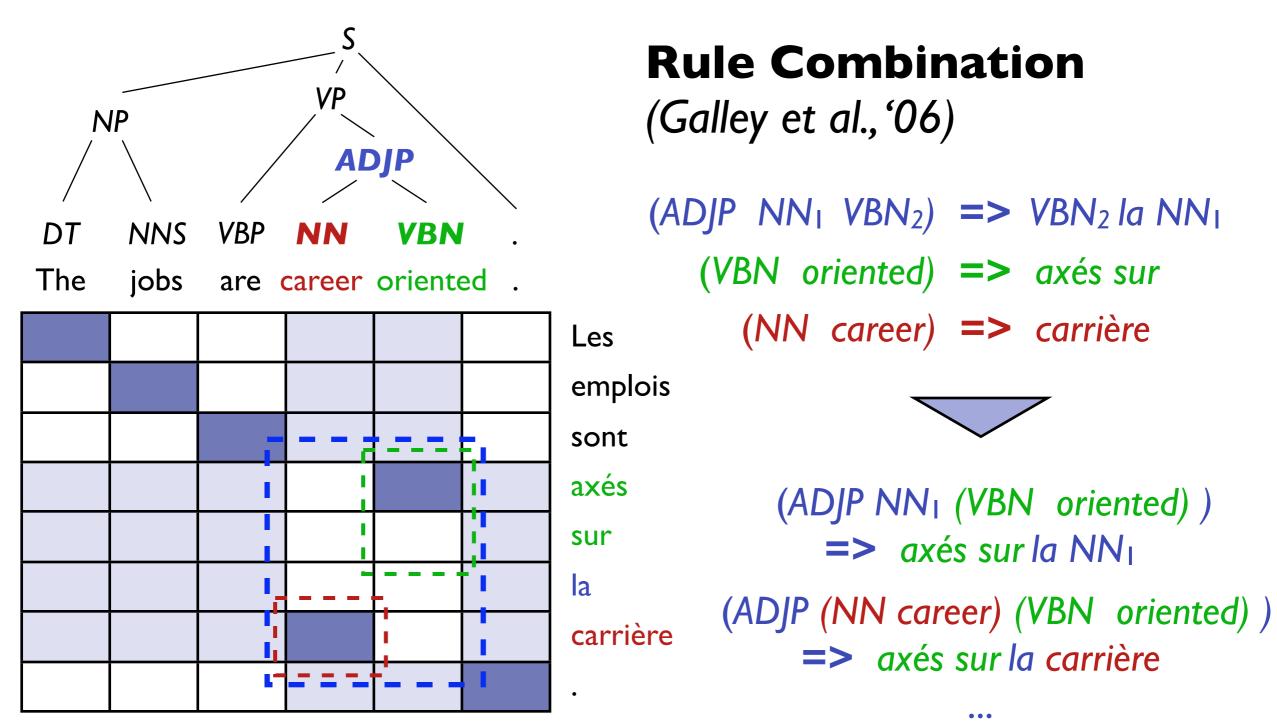




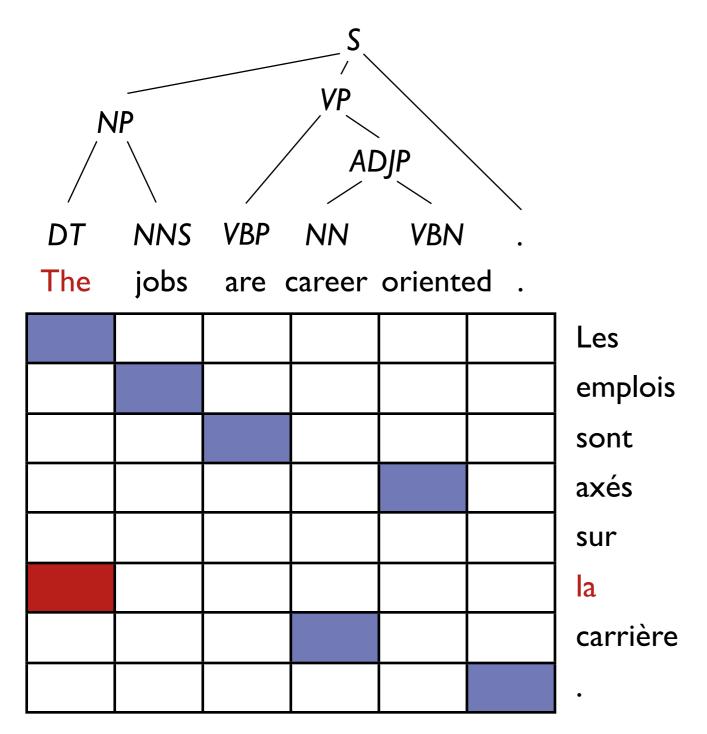




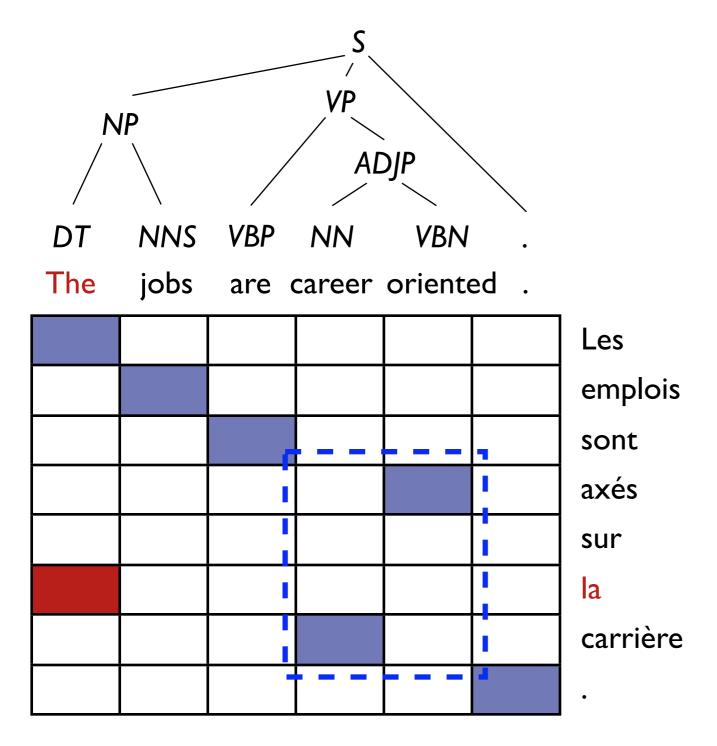




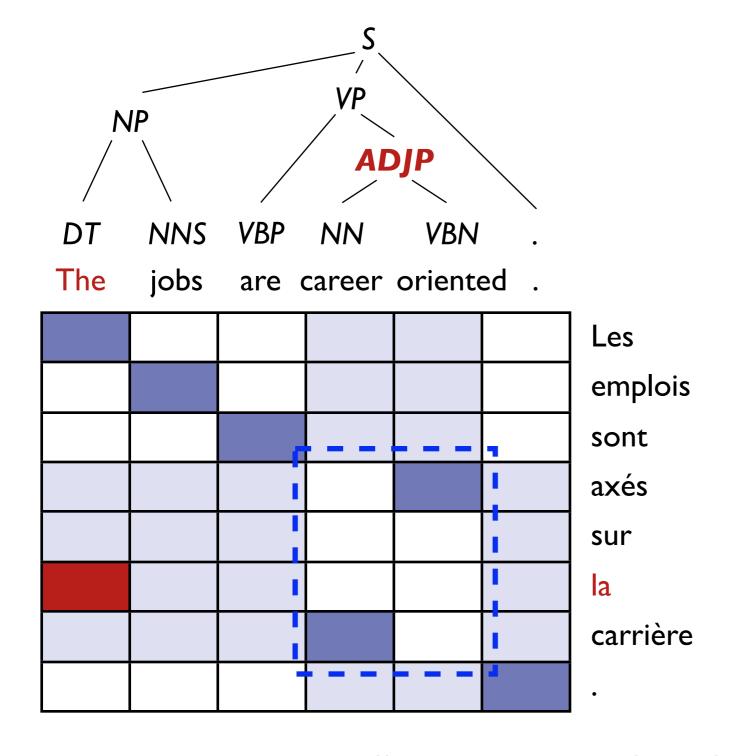




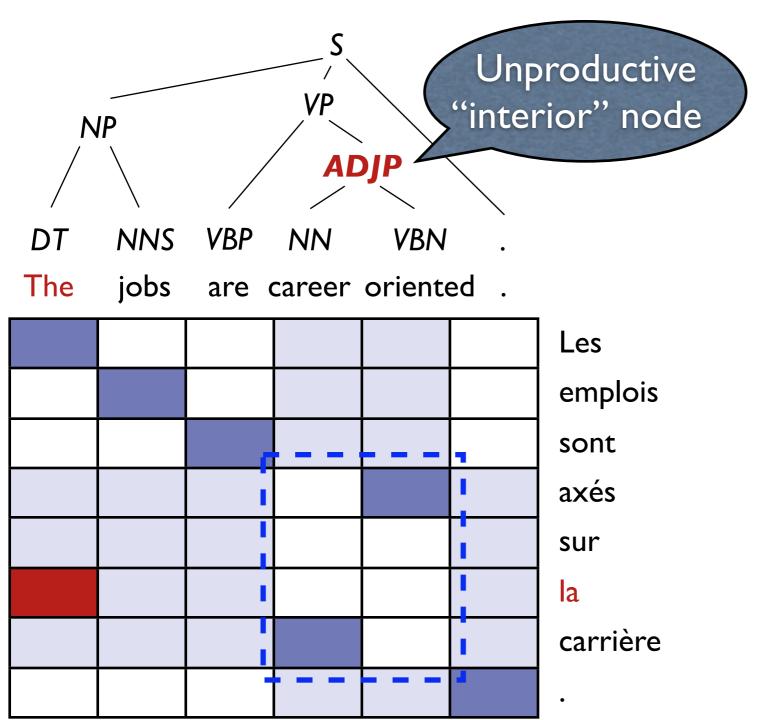




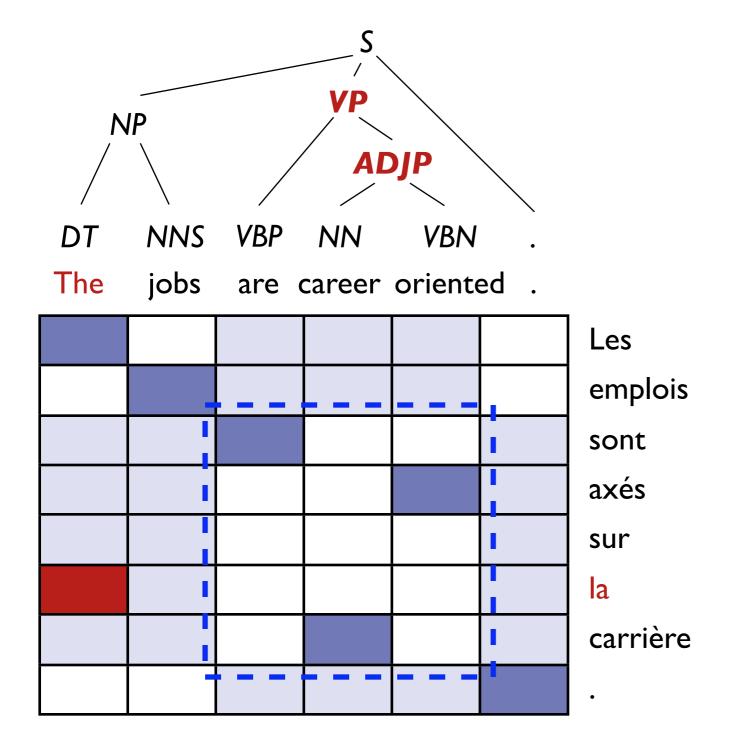




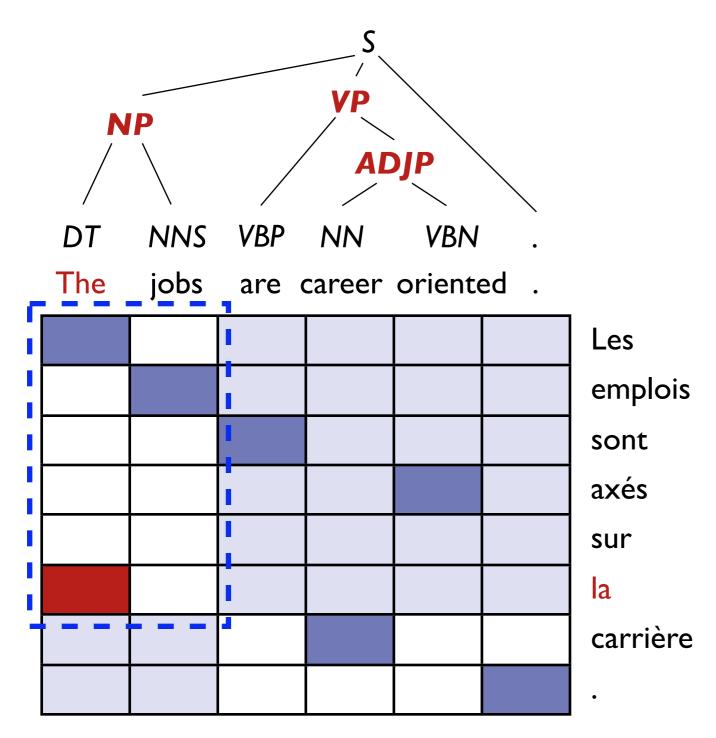




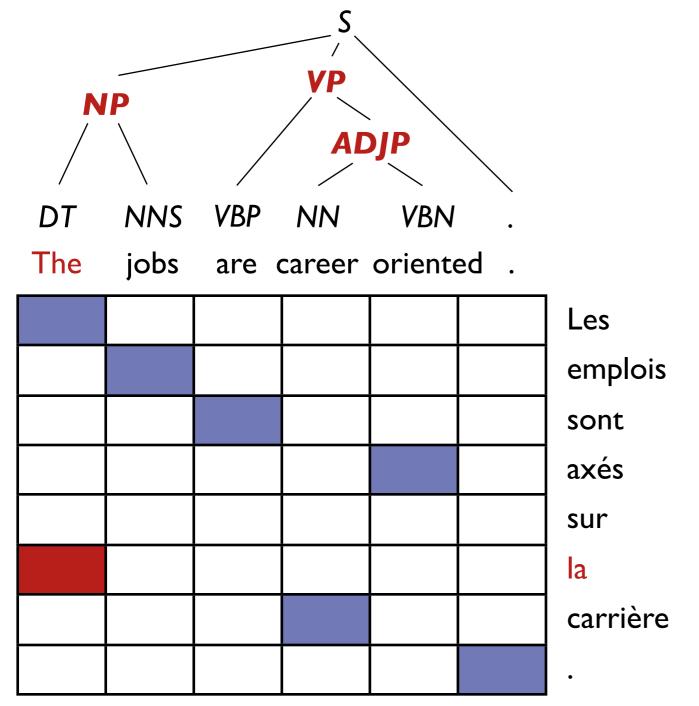






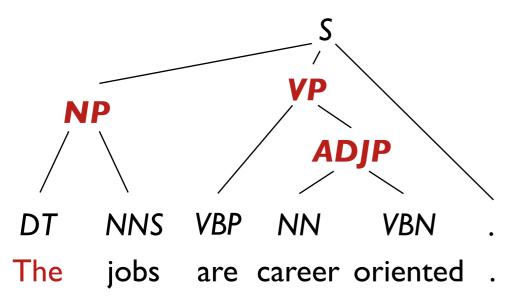


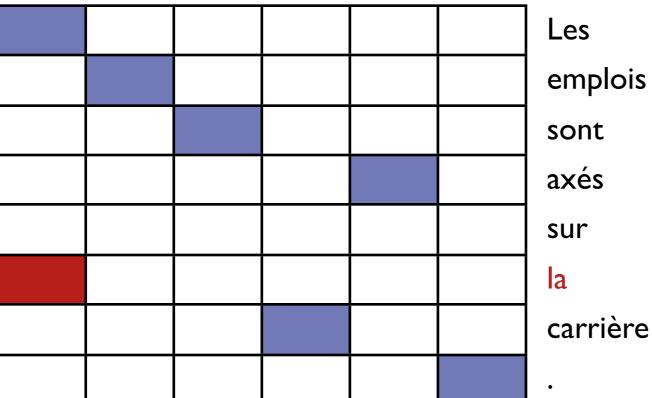




**Net effect on extraction:** 



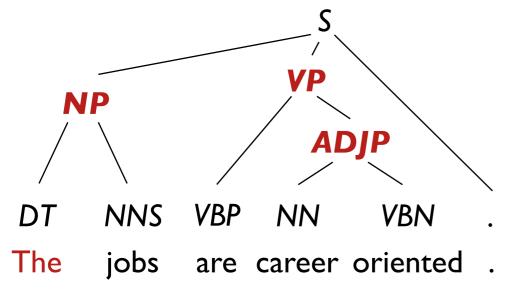


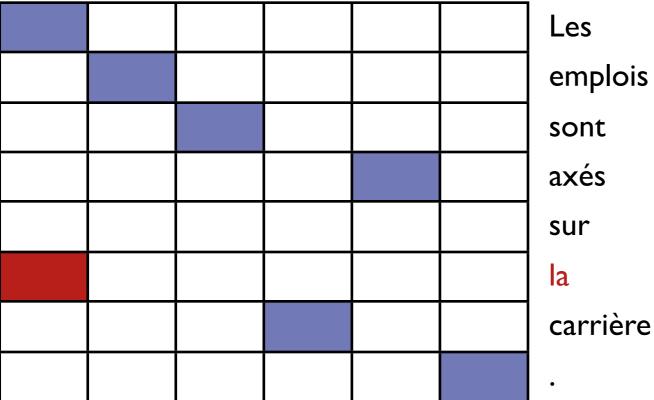


#### **Net effect on extraction:**

 2 instead of 7 recursive rules can be extracted



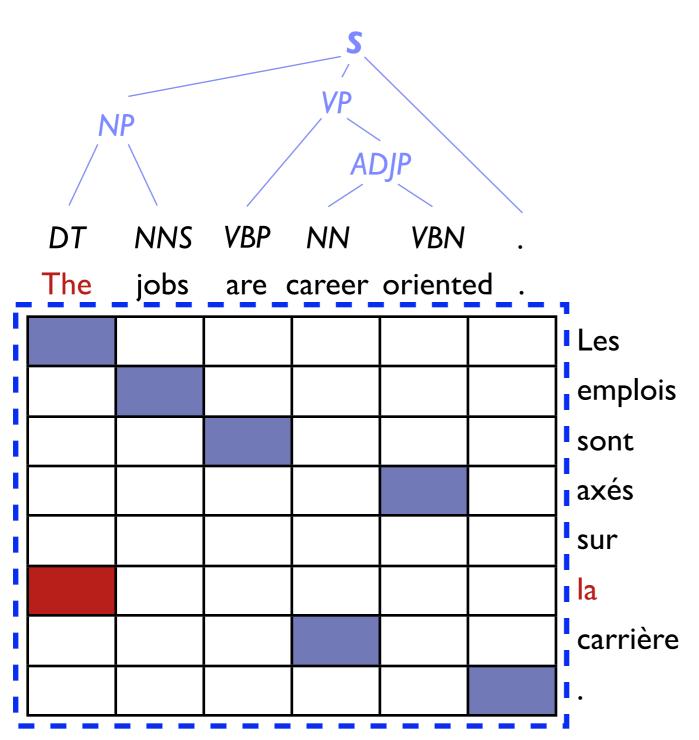




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- Smallest recursive rule that can be extracted:





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- 2 instead of 7 recursive rules can be extracted
- Smallest recursive rule that can be extracted:

```
(S (NP (DT The) NNS<sub>2</sub>)

(VP VBP<sub>3</sub>

(ADJP NN<sub>4</sub> VBN<sub>5</sub>))

.6)

=> Les NNS<sub>2</sub> VBP<sub>3</sub> VBN<sub>5</sub> NN<sub>4</sub> .6
```

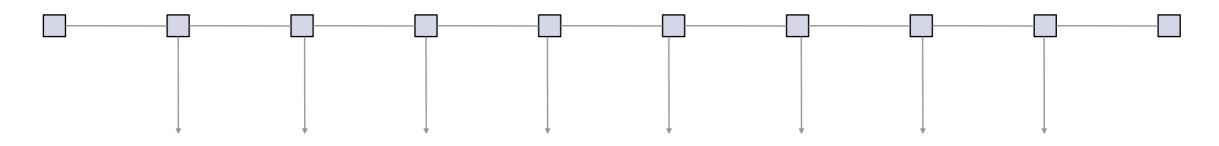


$$p(f, a|e) = \prod_{j} p(f_j|e_{a_j}) \cdot p(a_j|a_{j-1})$$



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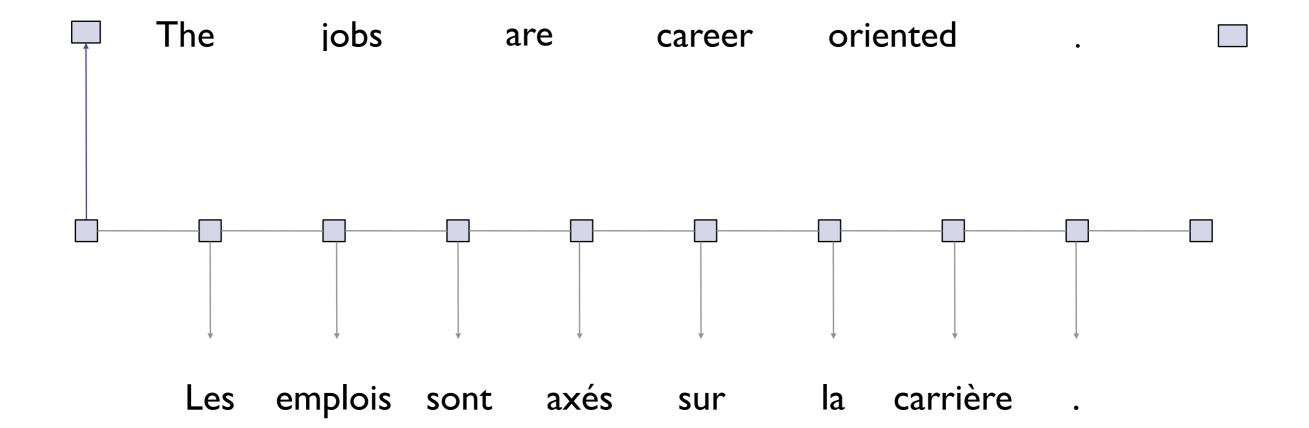




Les emplois sont axés sur la carrière

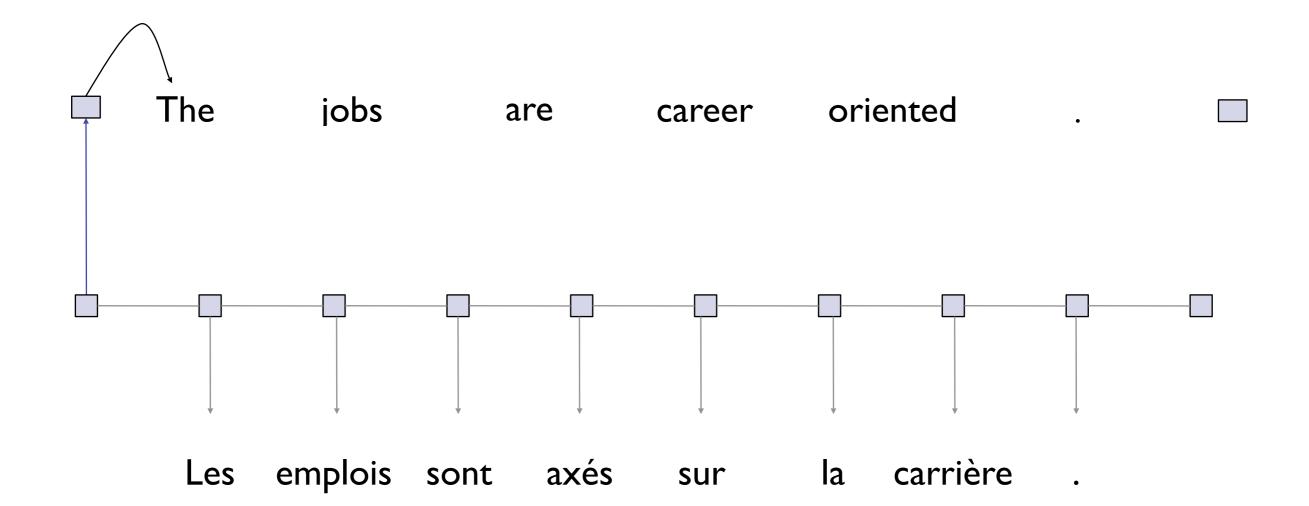


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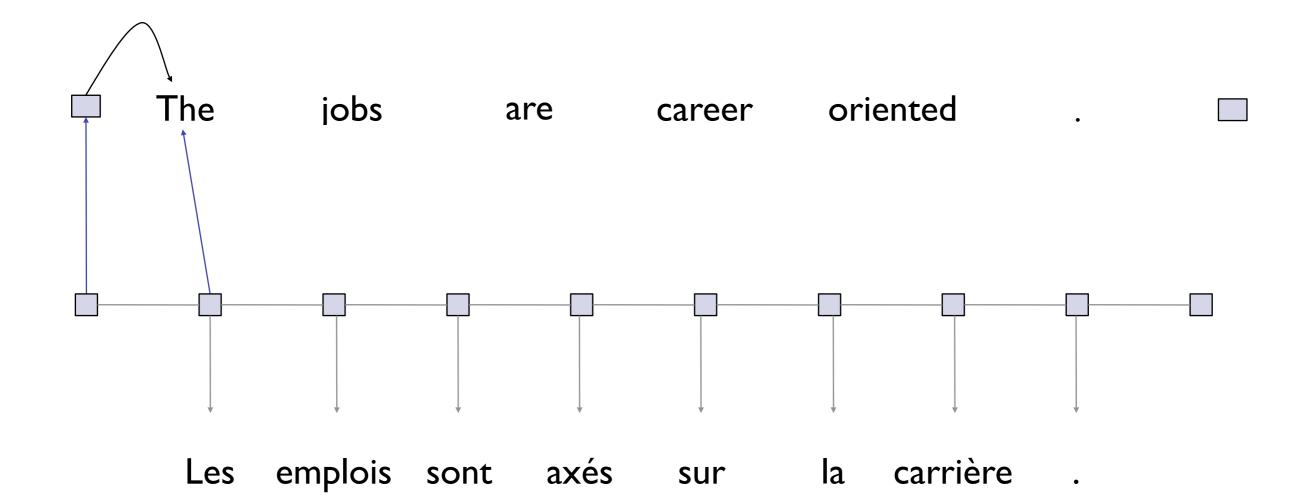


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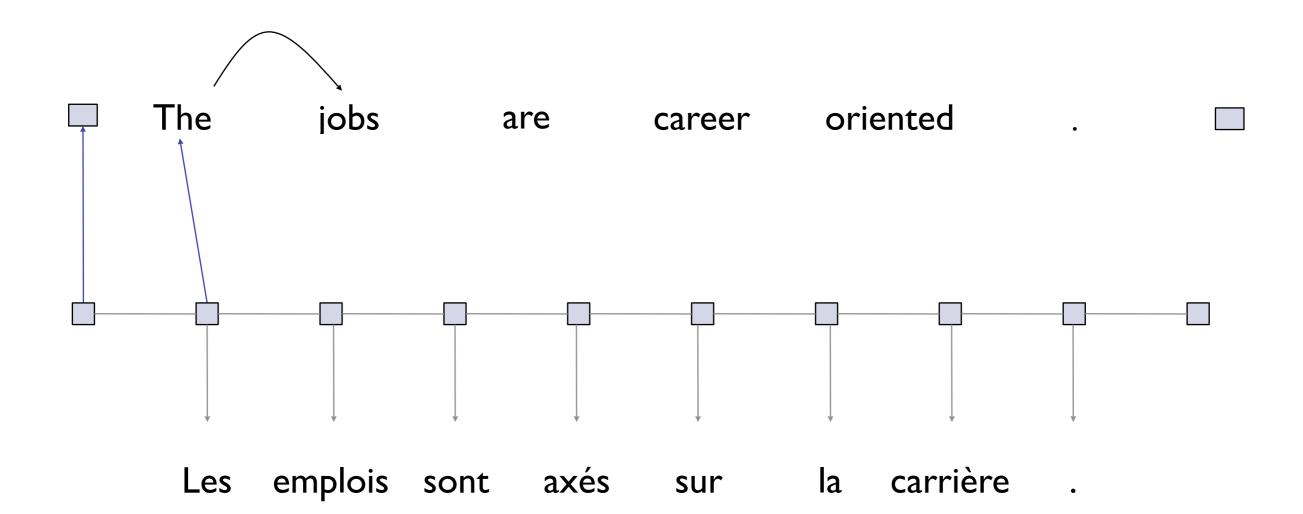


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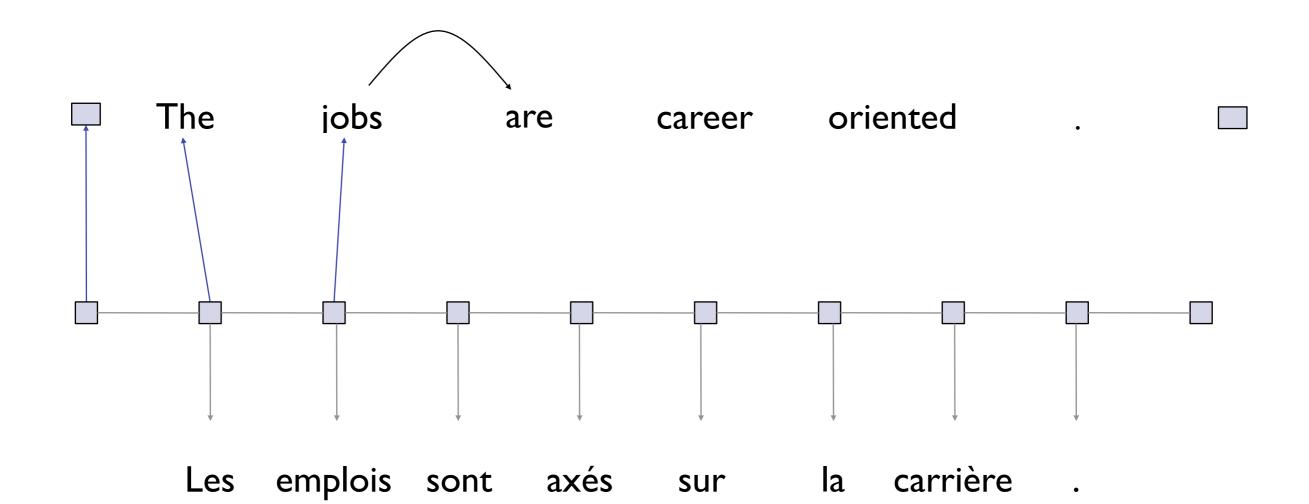


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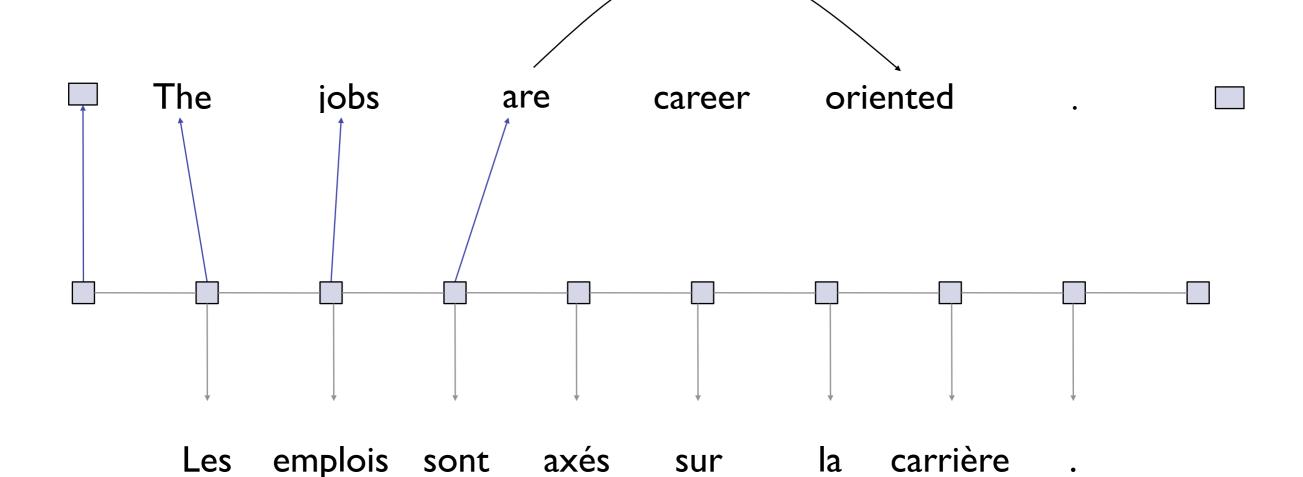


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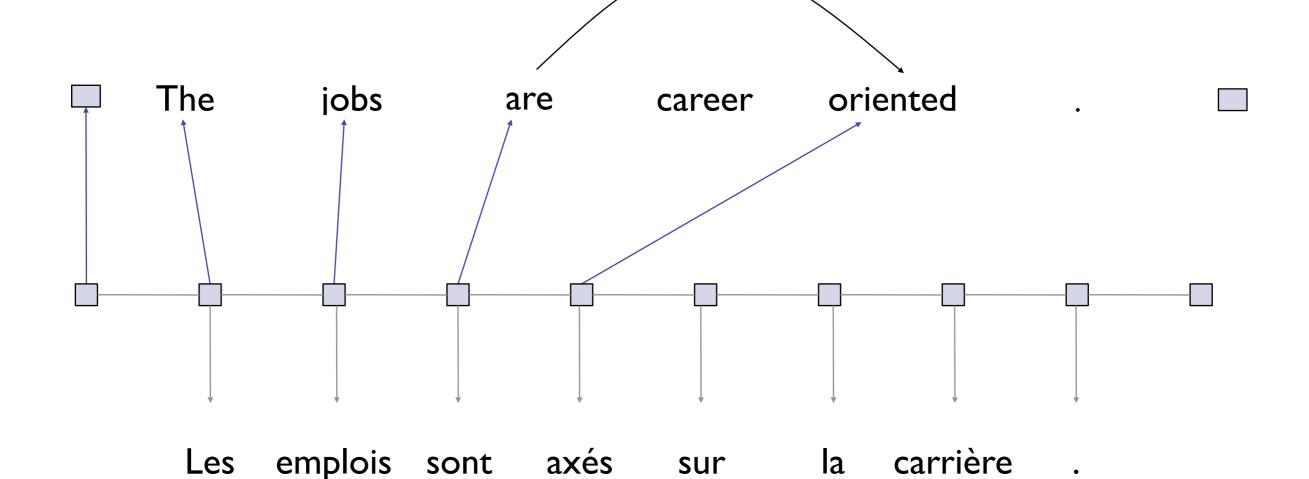


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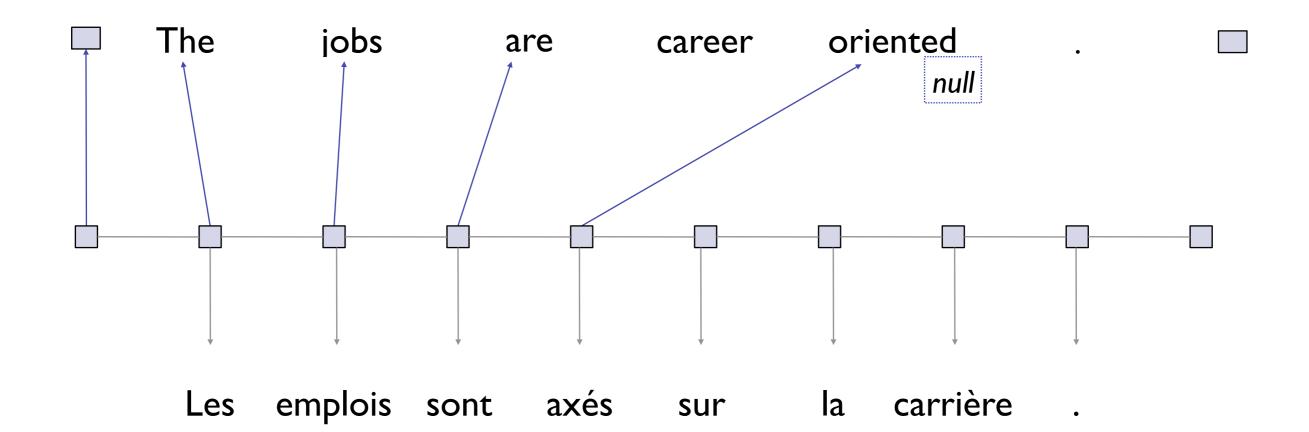


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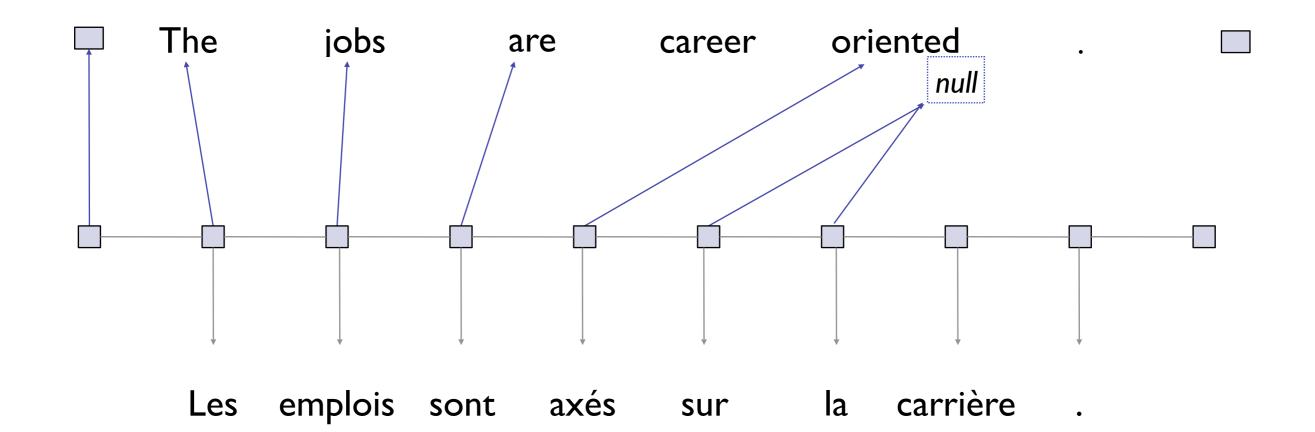


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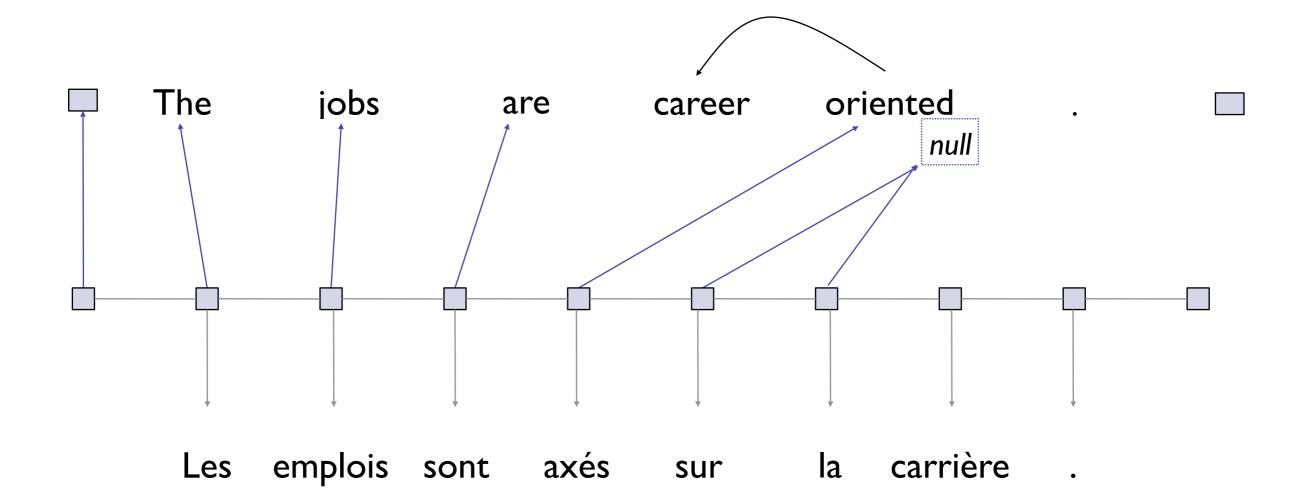


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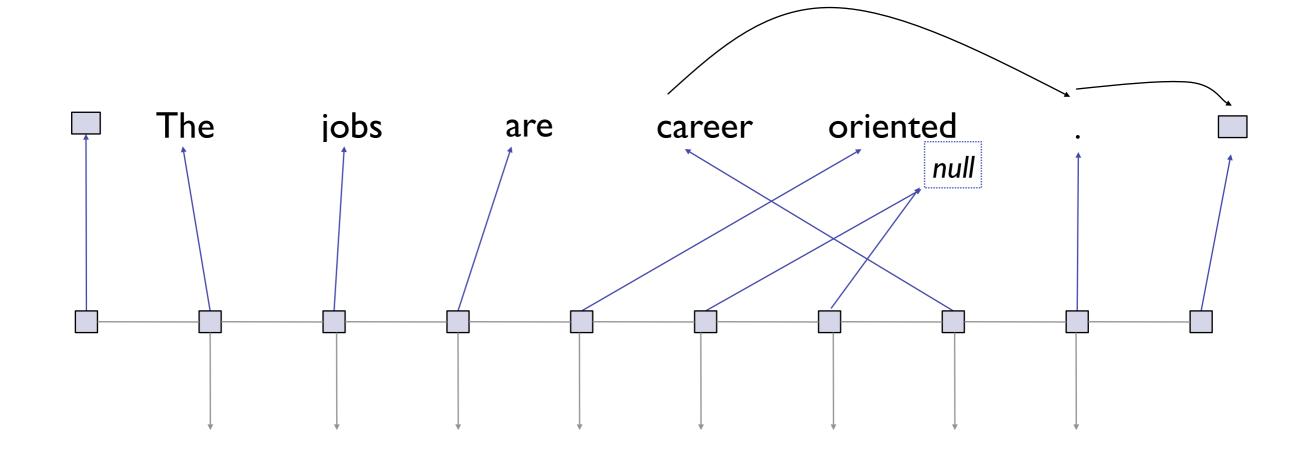


Les

emplois

### Alignment Errors under the HMM Alignment Model

$$p(f, a|e) = \prod_{j} p(f_j|e_{a_j}) \cdot p(a_j|a_{j-1})$$



sur

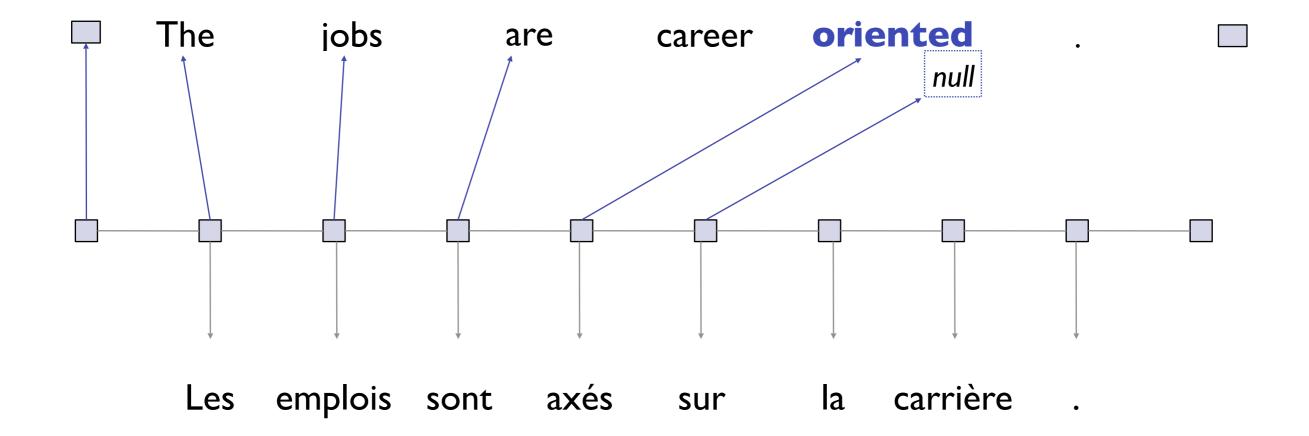
axés

sont

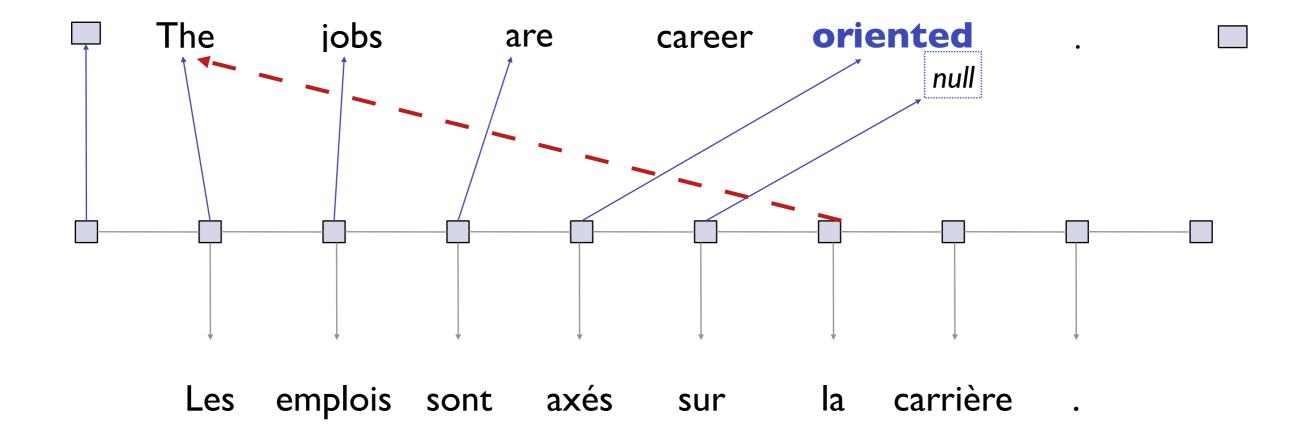
carrière

la

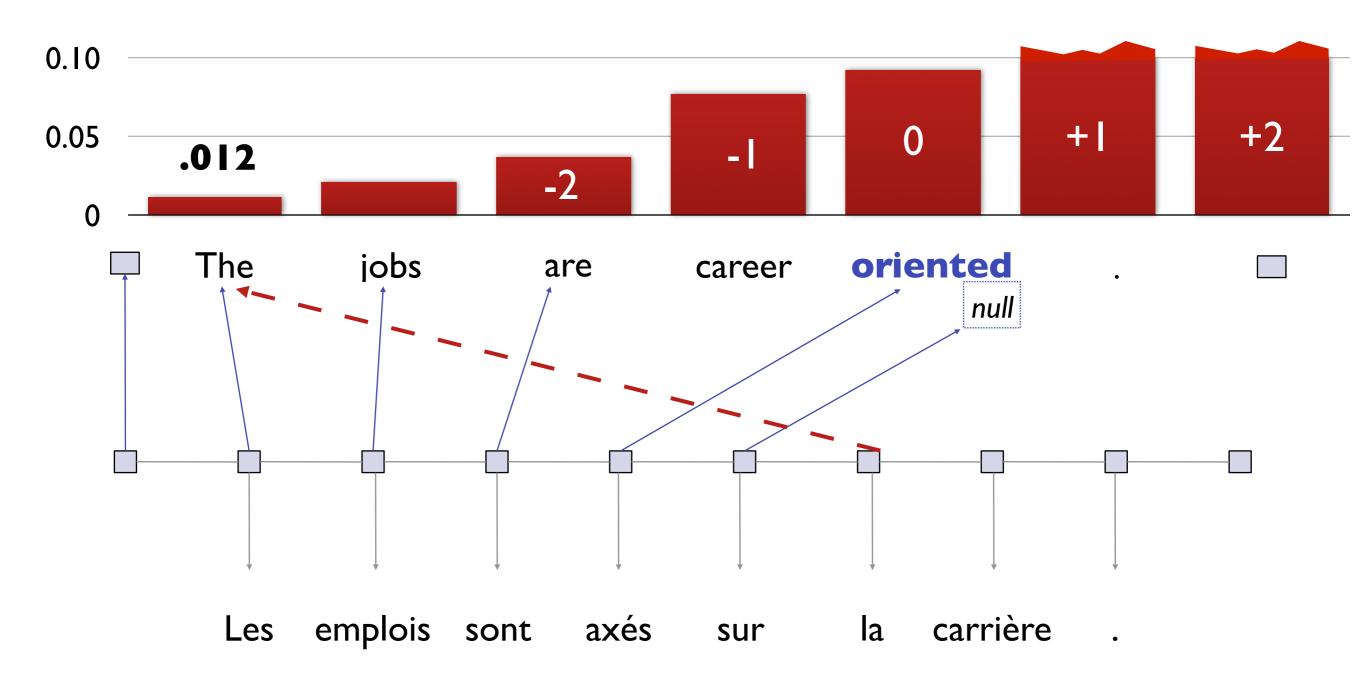




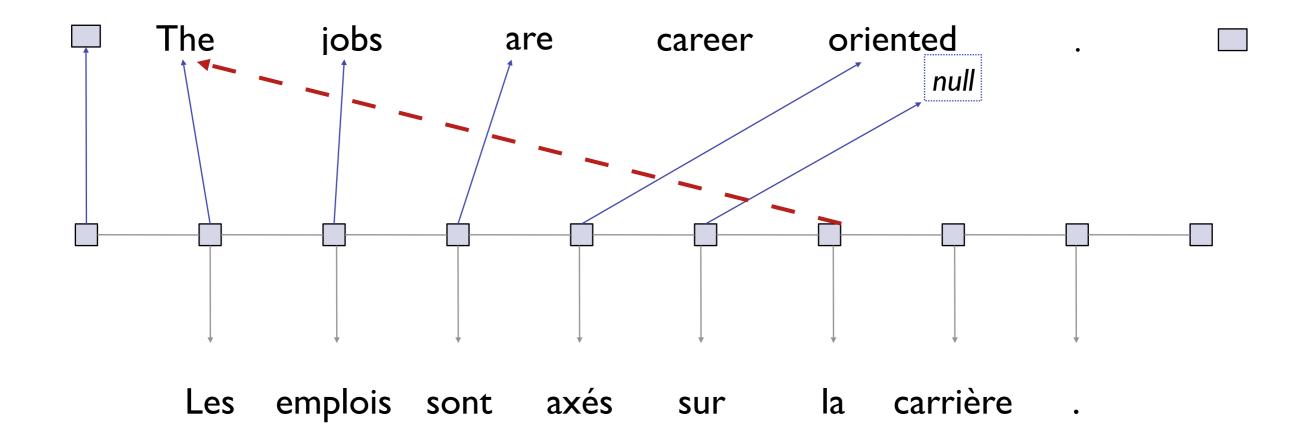




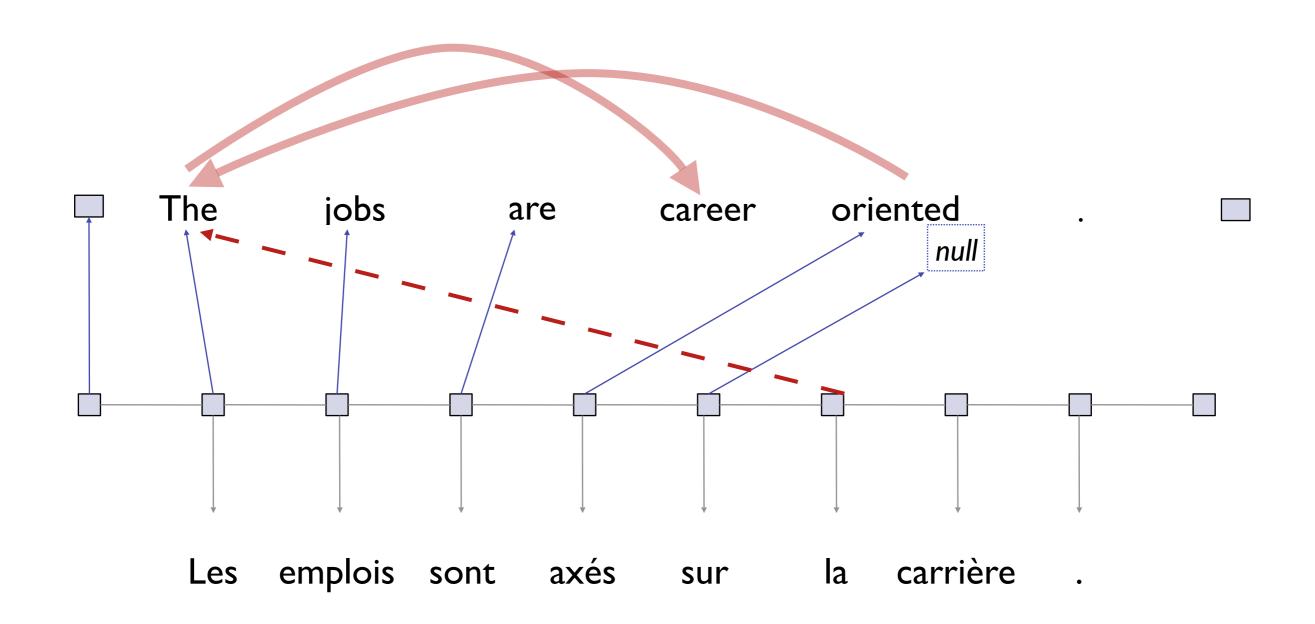




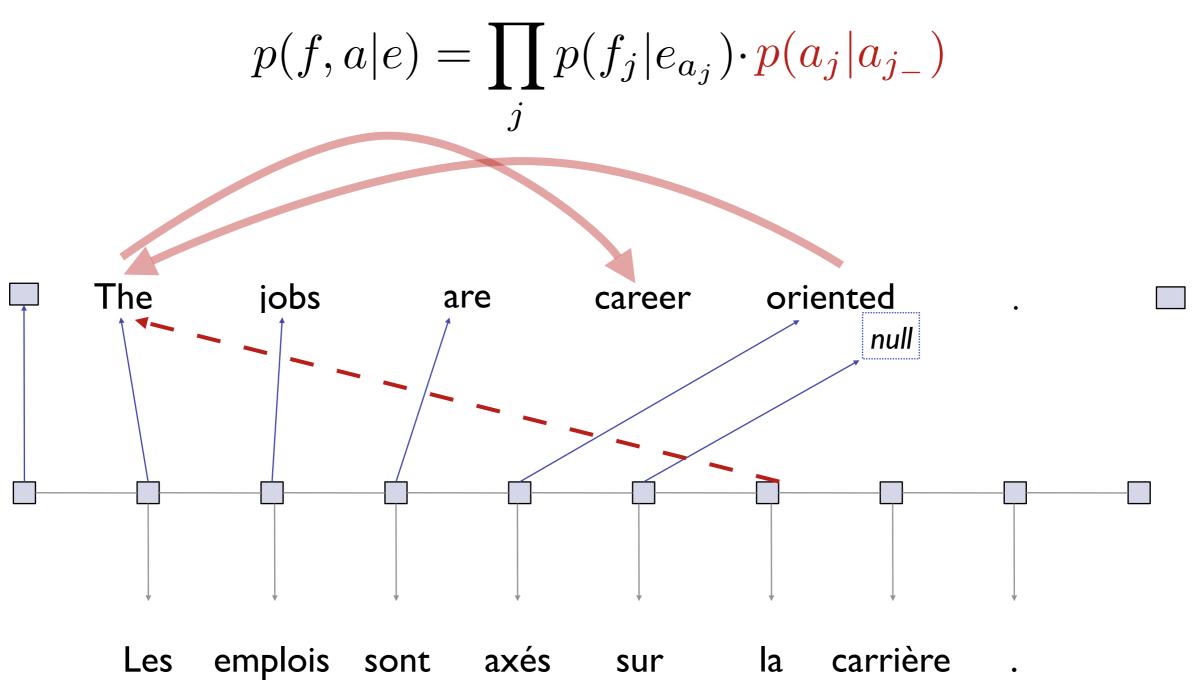




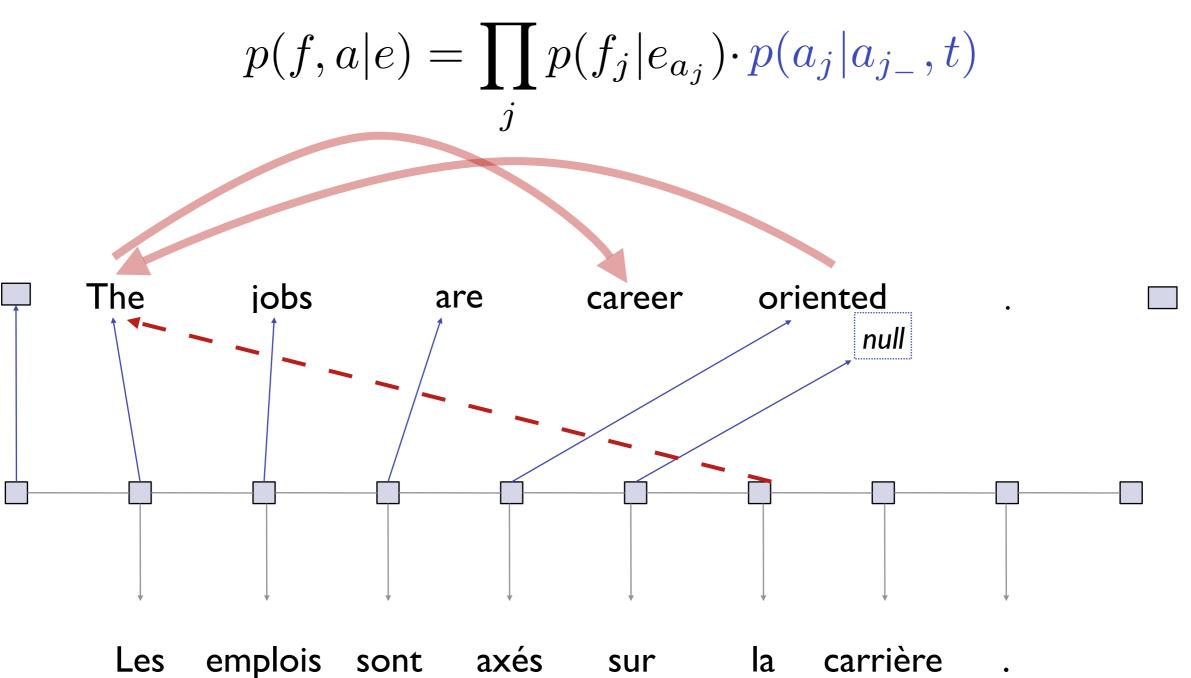




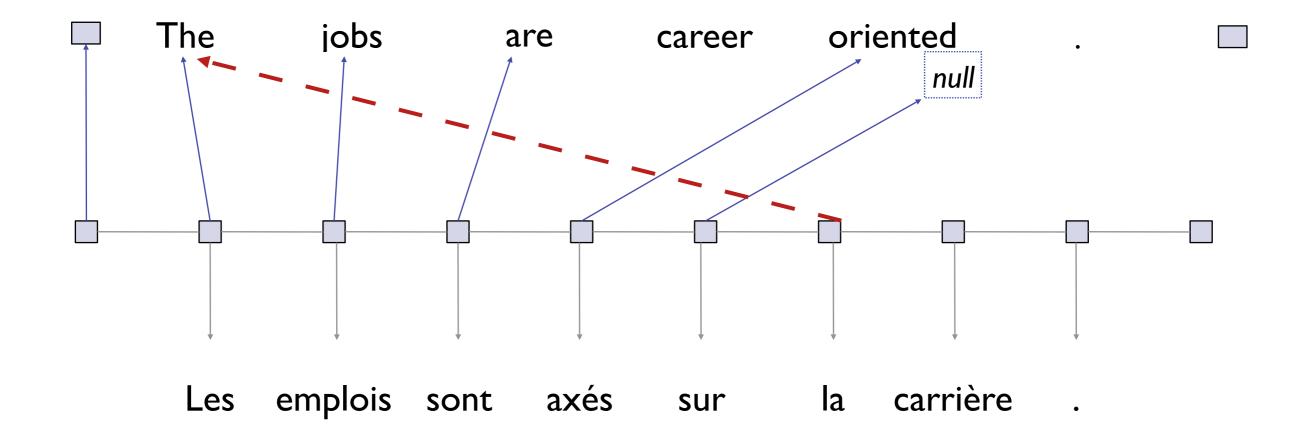




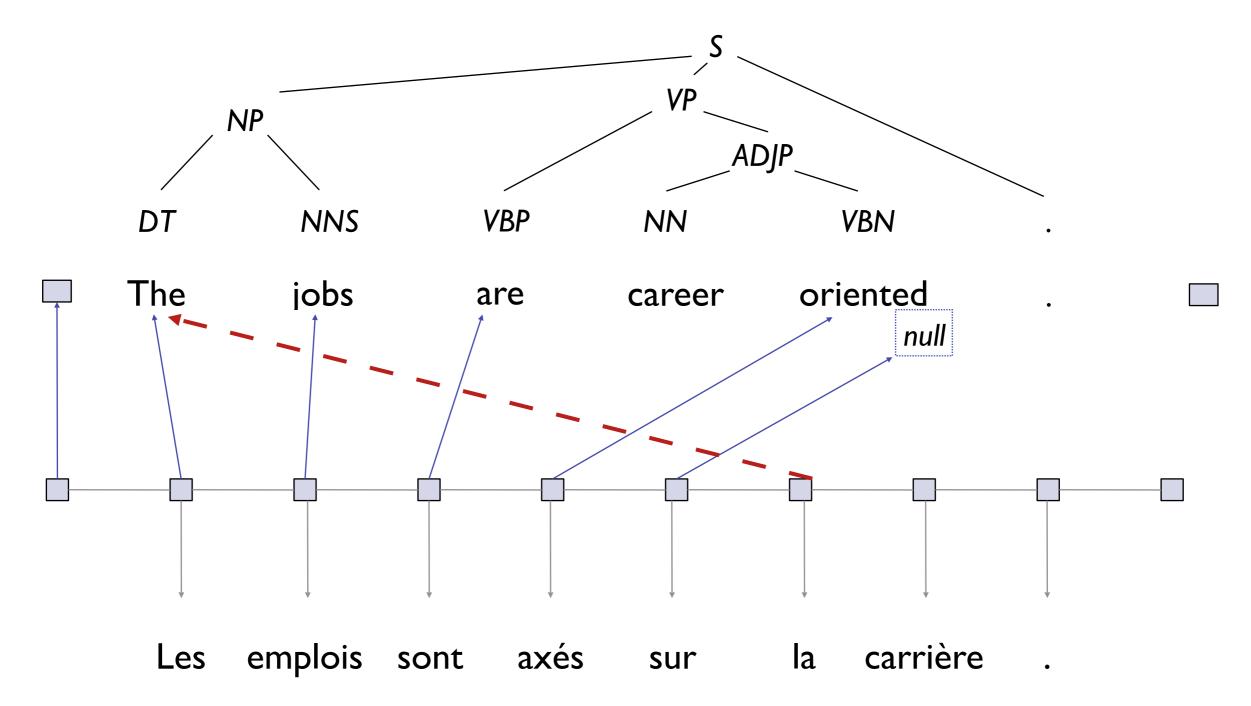




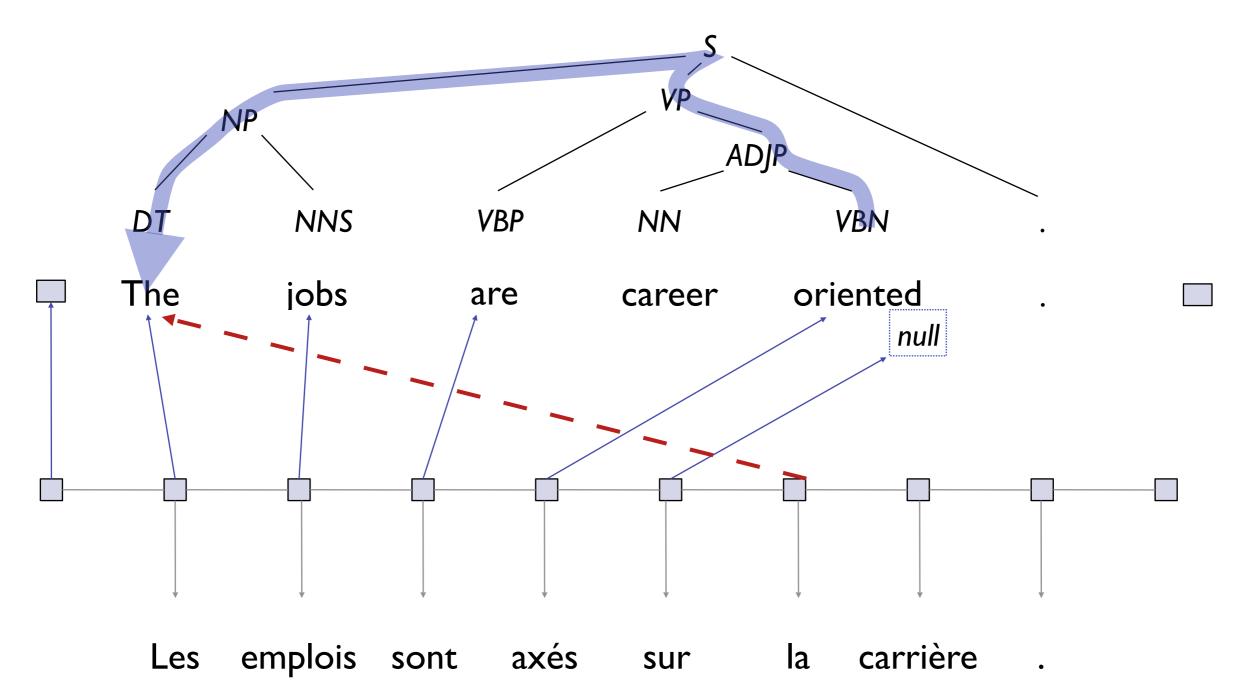




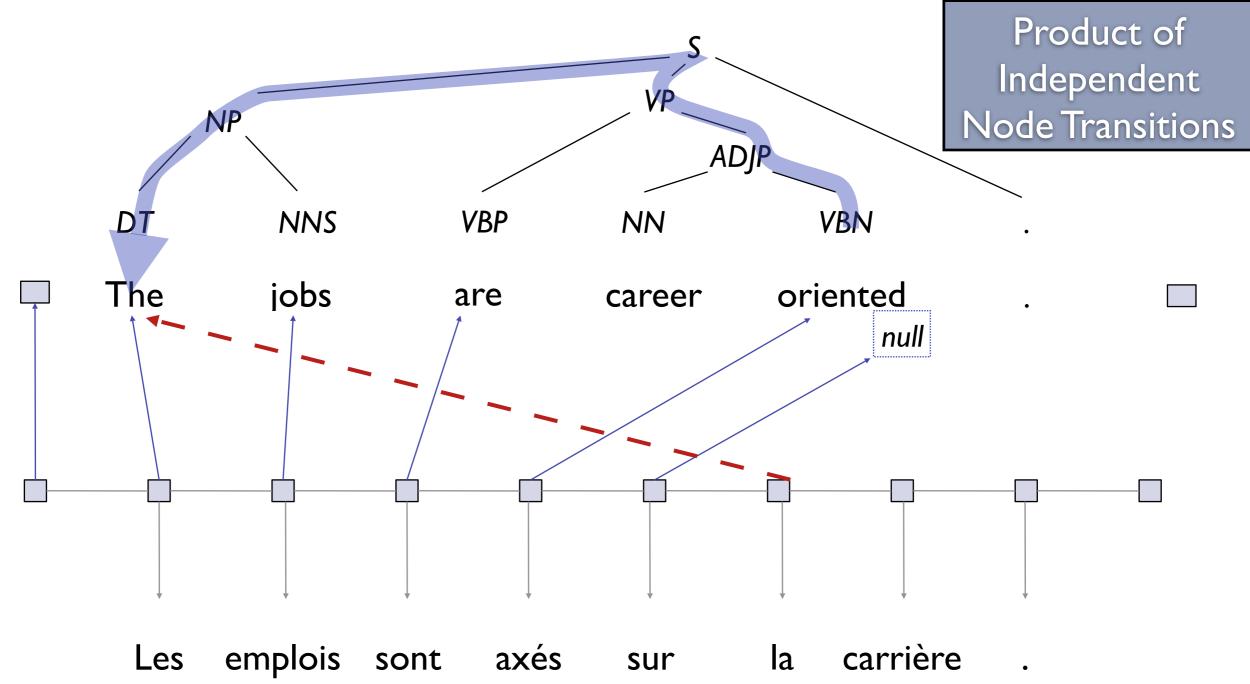




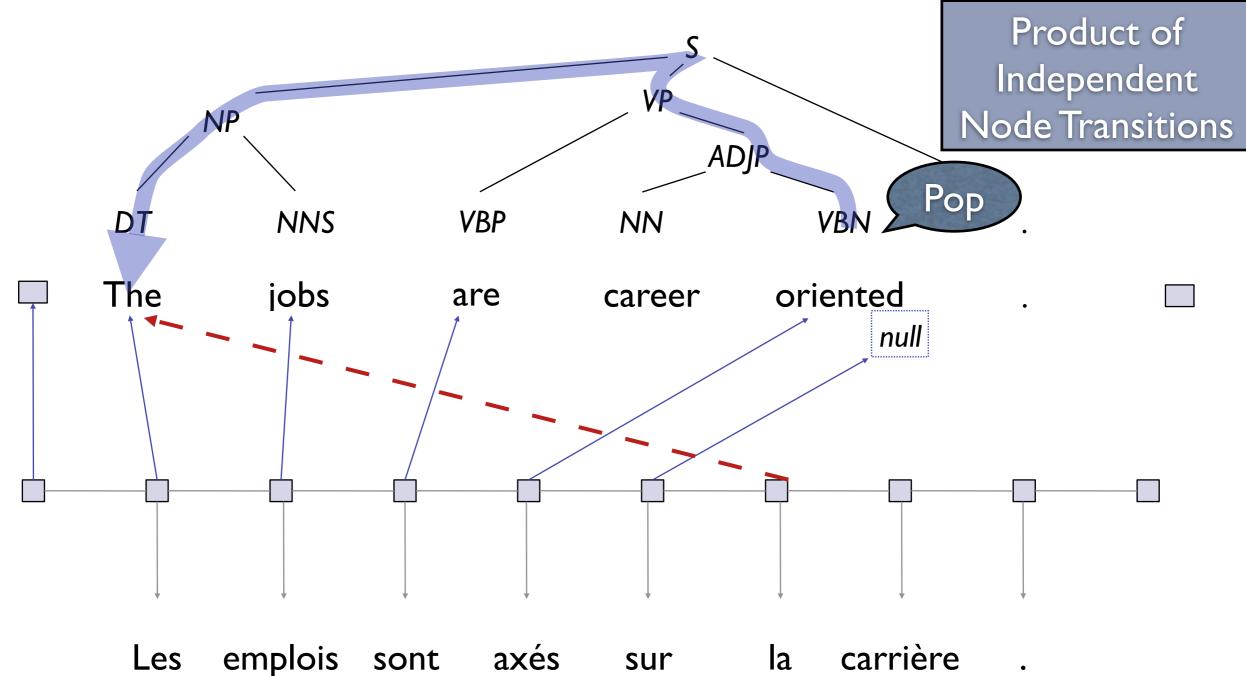




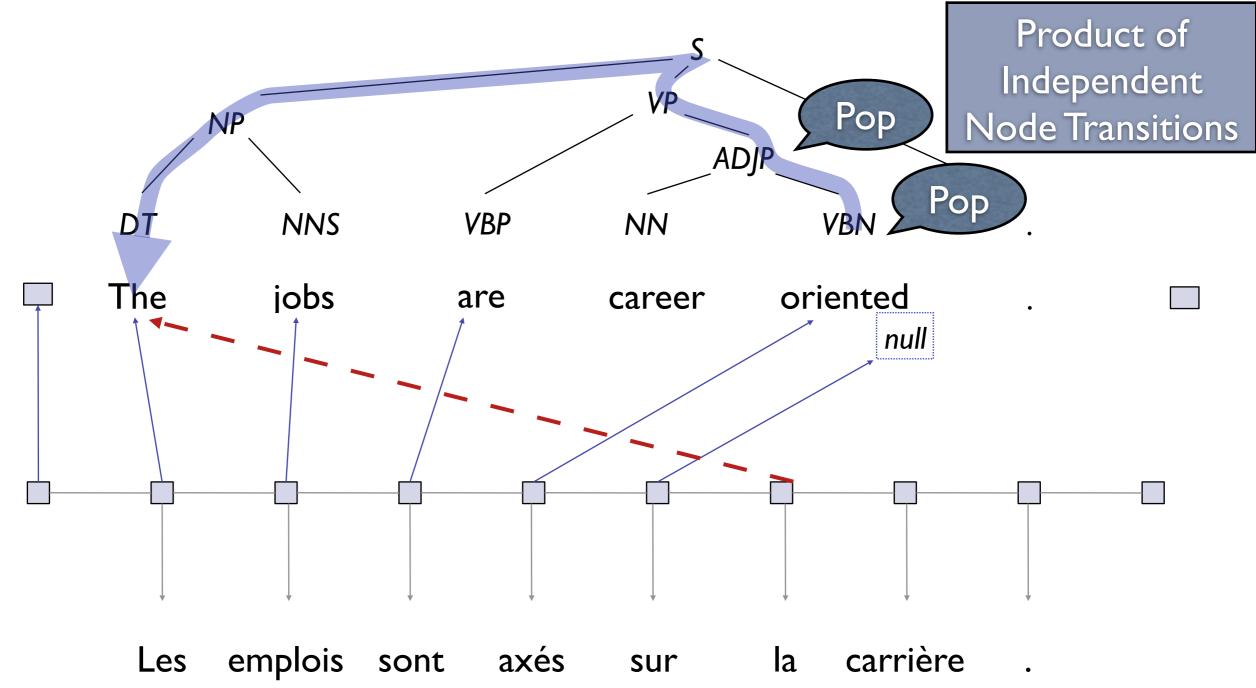




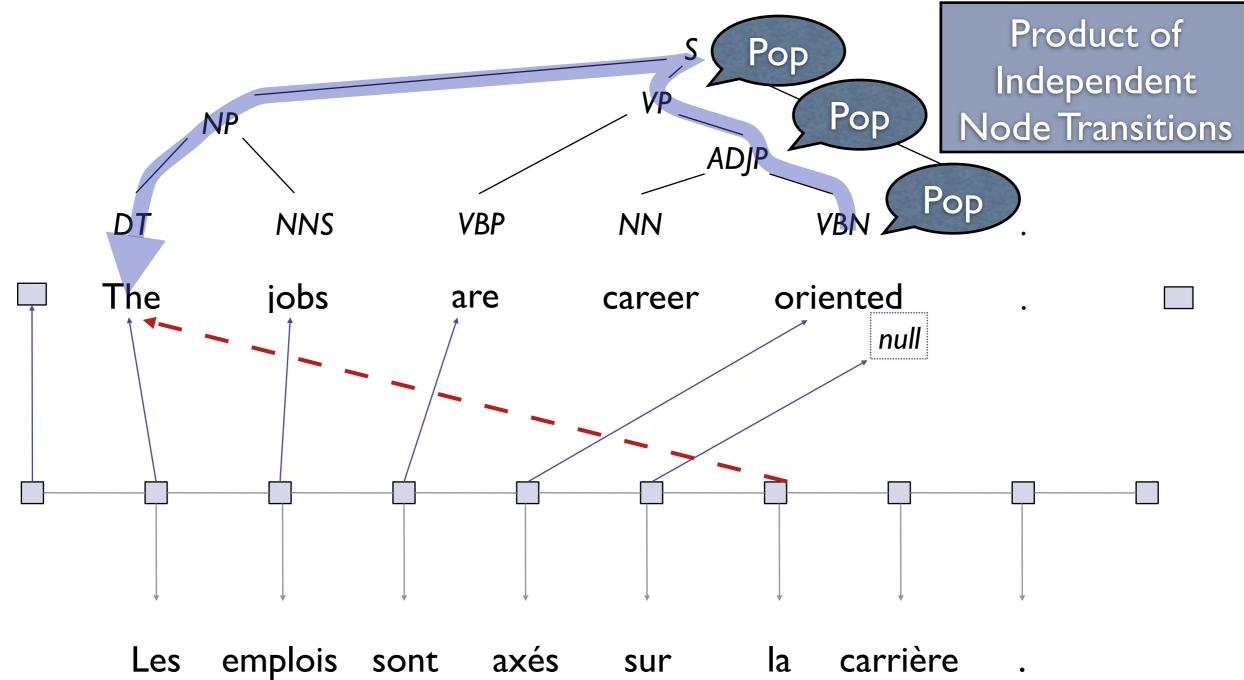




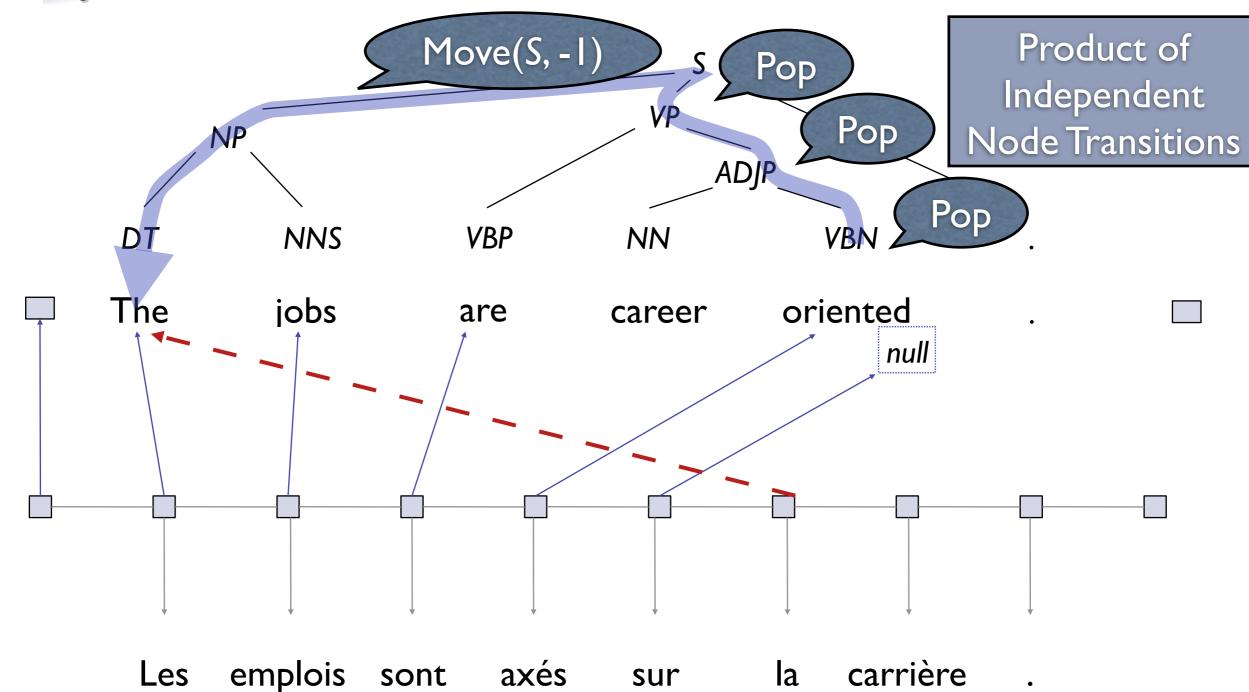




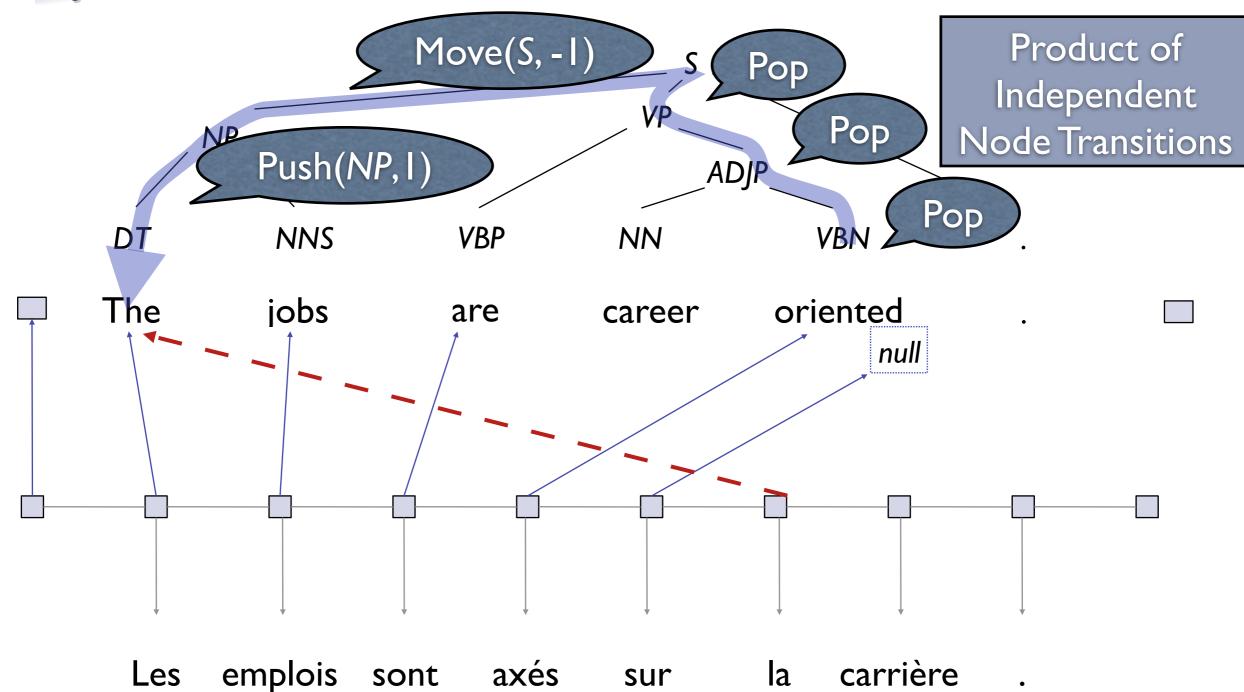




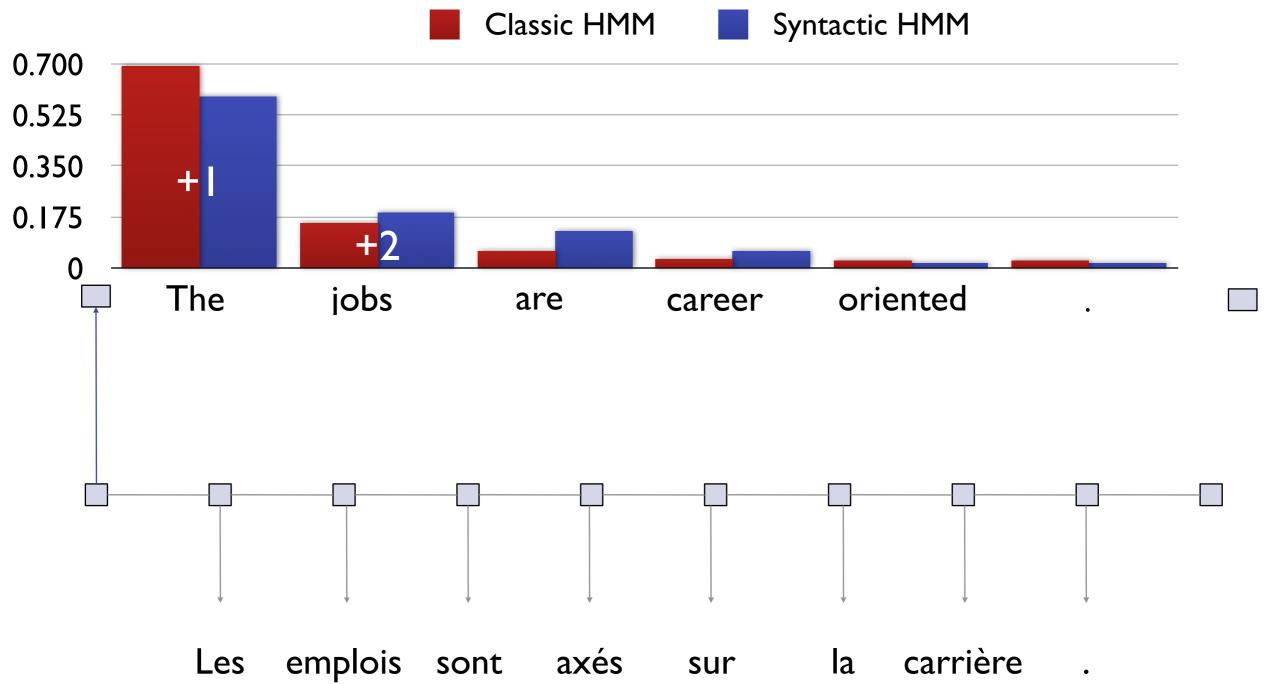




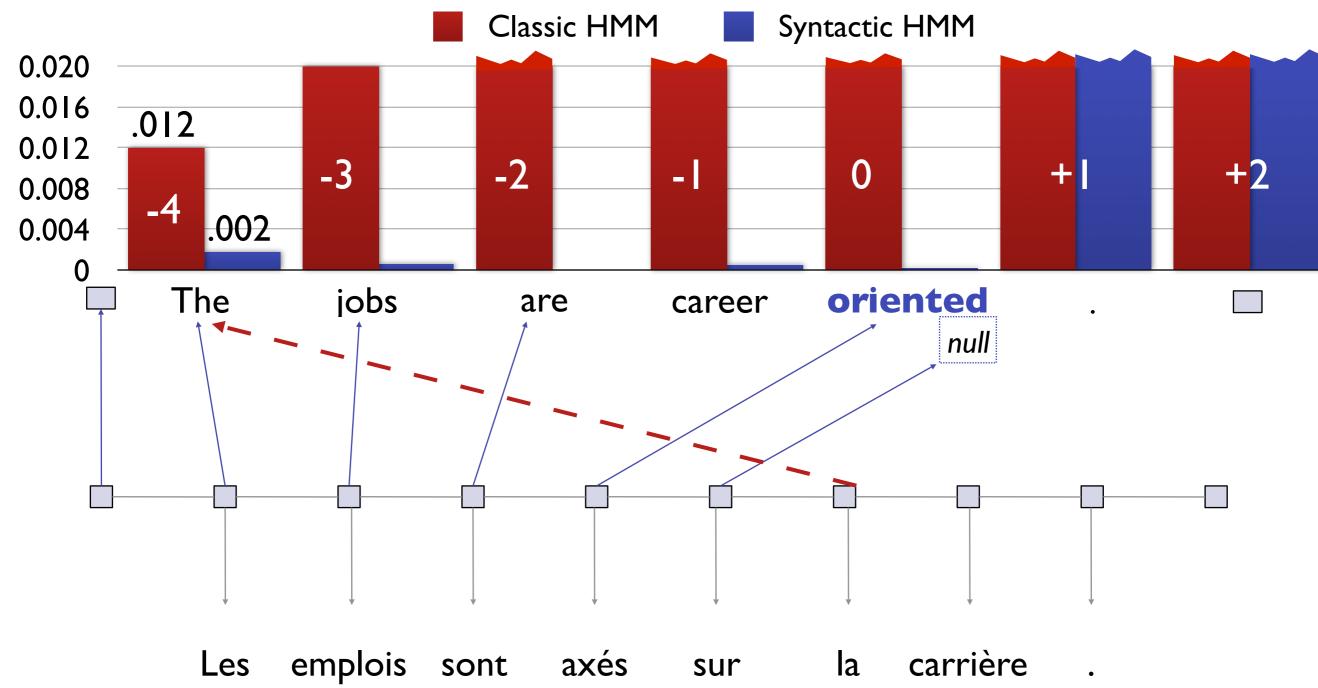




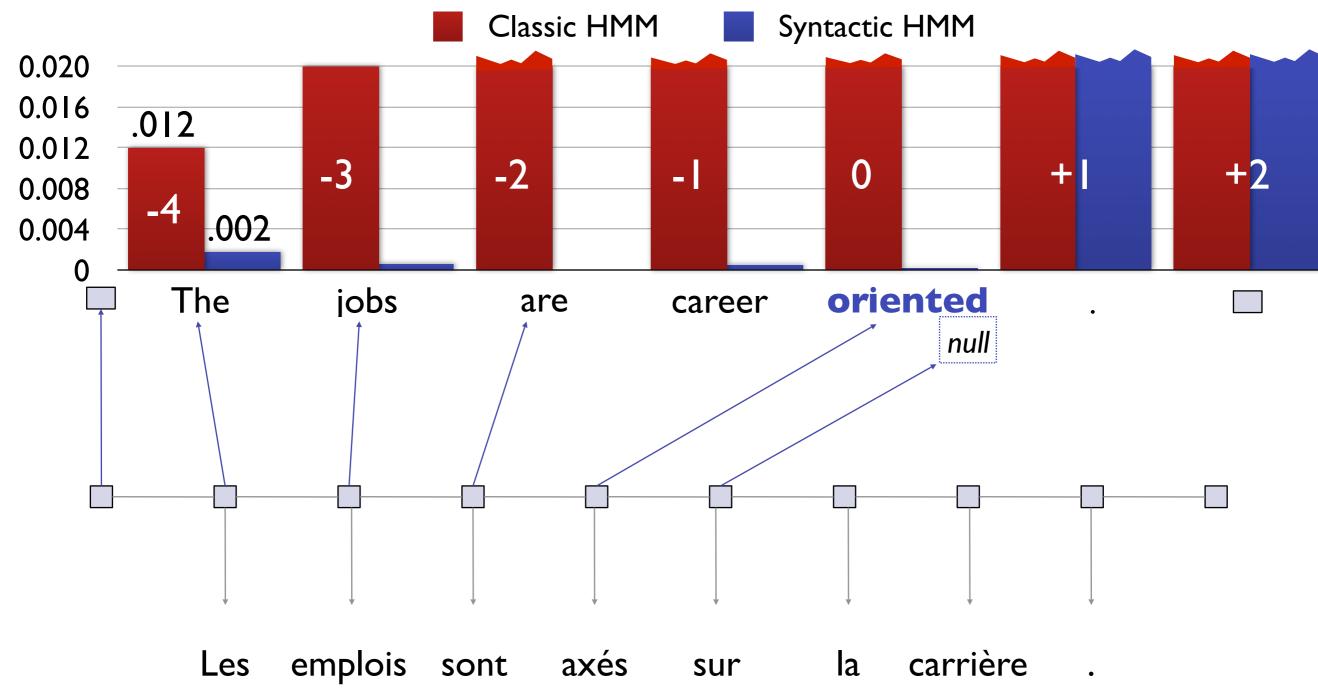














# Evaluation: Alignment Error Rate (AER)



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#### **Test Conditions**

- Chinese-English from MT-Eval 02 test set
- 100k training sentences from FBIS
- Initialized with agreement training for Model I (Liang et al., 06)



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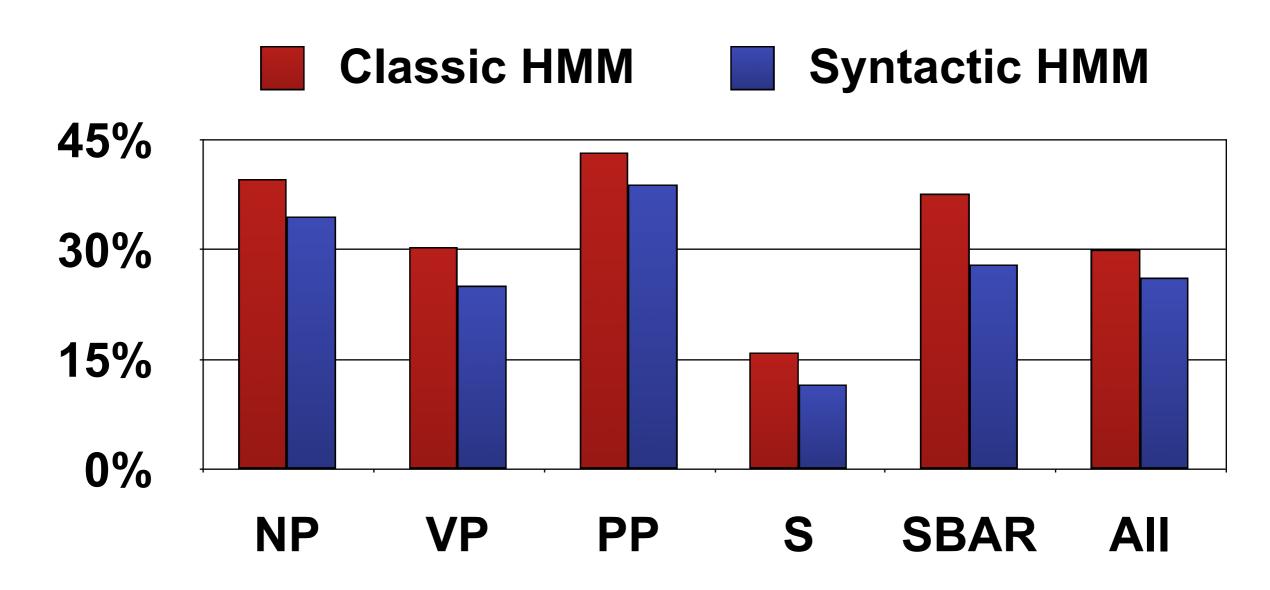
#### Results

	Precision	Recall	AER
Classic HMM	81.6	78.8	19.8
Syntactic HMM	82.2	76.8	20.5
GIZA++	61.9	82.6	29.7

http://nlp.cs.berkeley.edu/pages/WordAligner.html



### Evaluation: Unproductive Constituent Rates

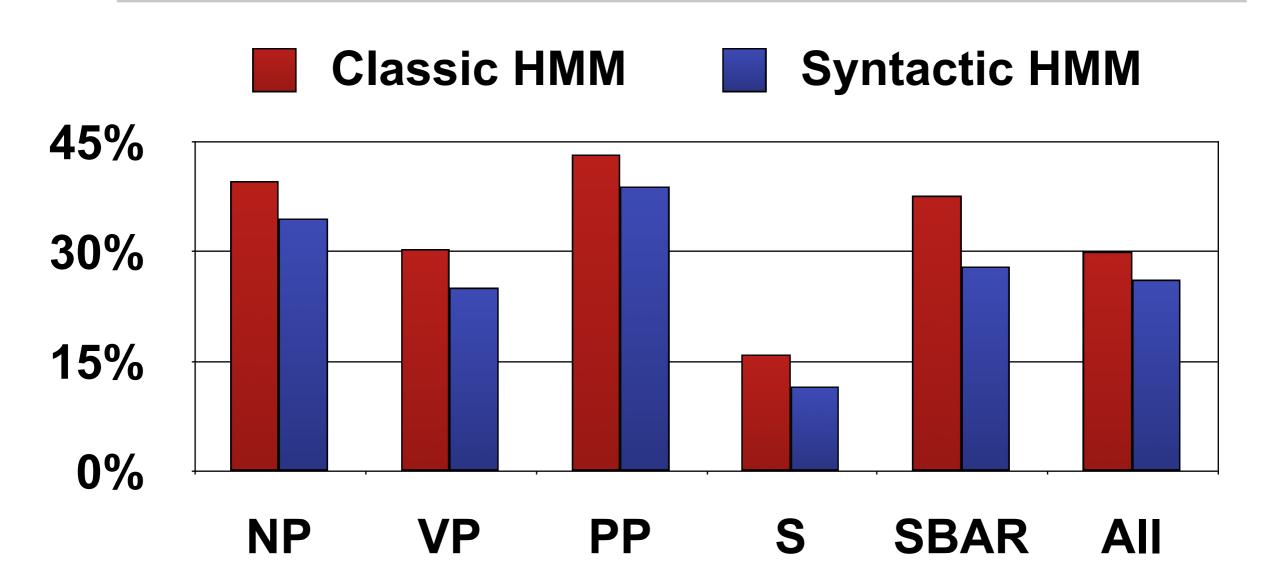


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### Evaluation: Unproductive Constituent Rates

The Syntactic HMM Reduces the Frequency of Unproductive *Interior* Nodes by 13%





## Decoding Heuristic: Competitive Thresholding

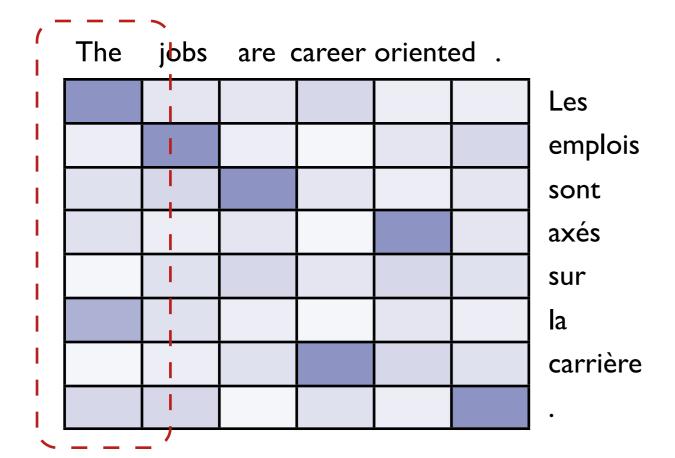
Only the maximum posterior in each row or column and its neighbors can be included in the alignment

The jobs are career oriented.

			Les
			emplois
			sont
			axés
			sur
			la
			carrière
			•

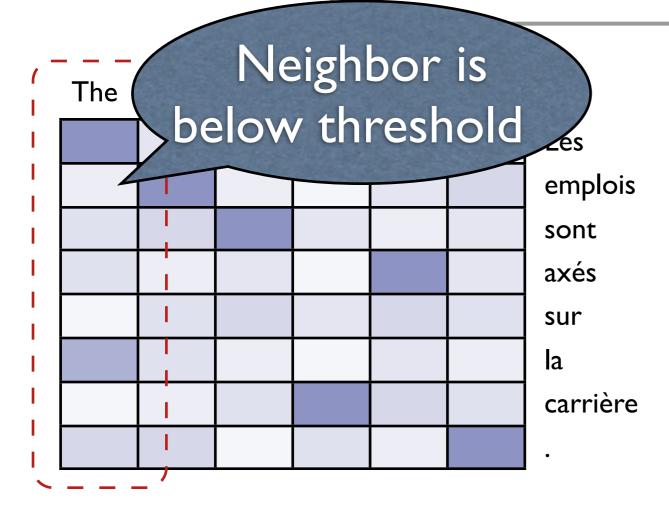


## Decoding Heuristic: Competitive Thresholding



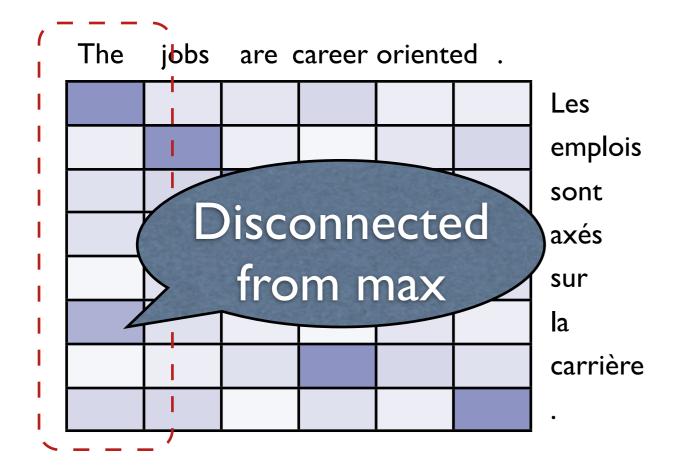


### Decoding Heuristic: Competitive Thresholding



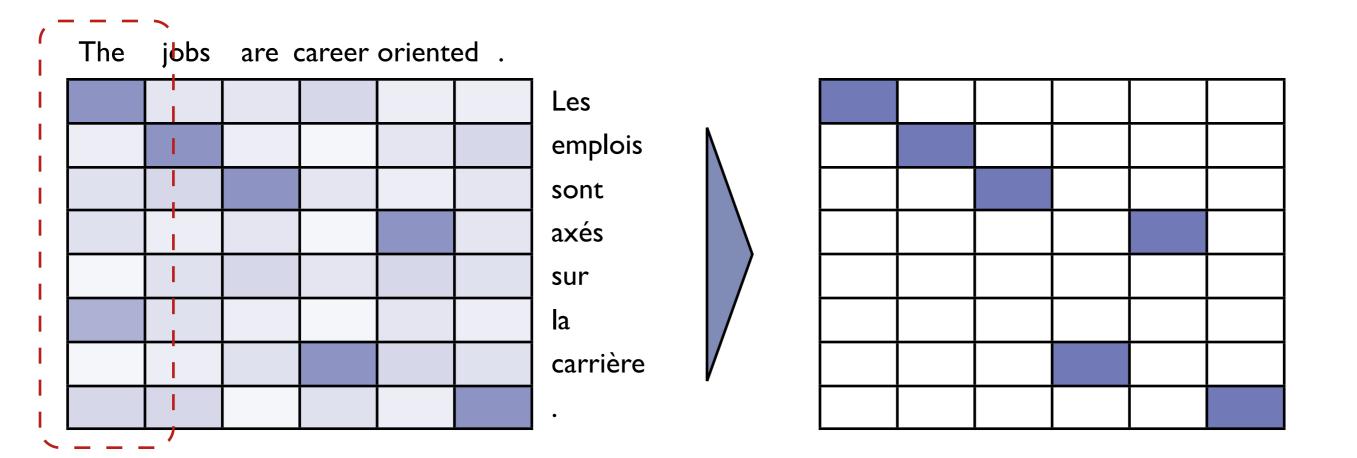


## Decoding Heuristic: Competitive Thresholding





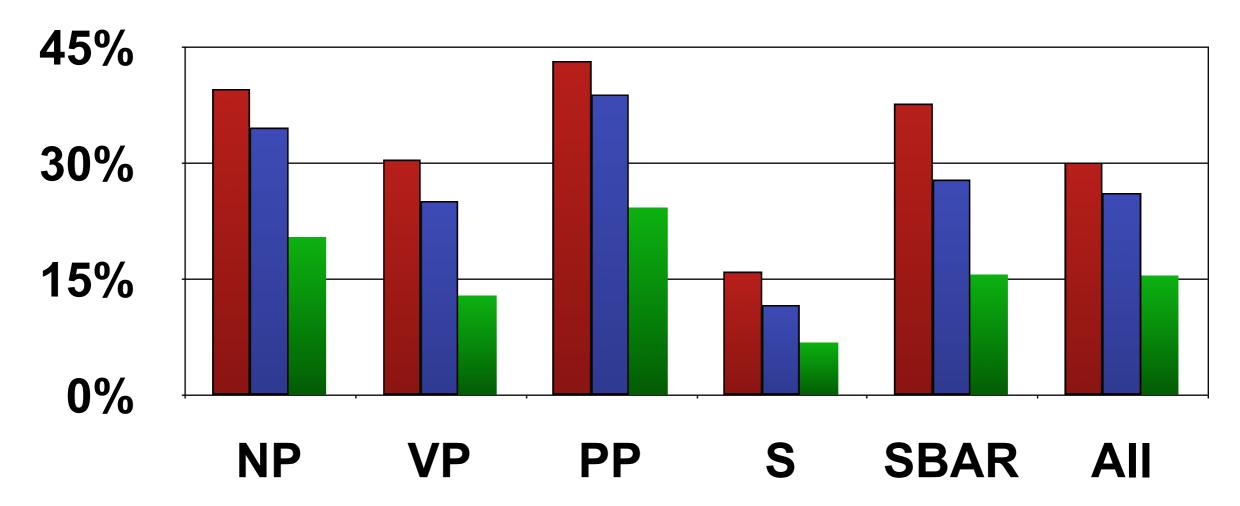
## Decoding Heuristic: Competitive Thresholding





### Evaluation: Unproductive Constituent Rates

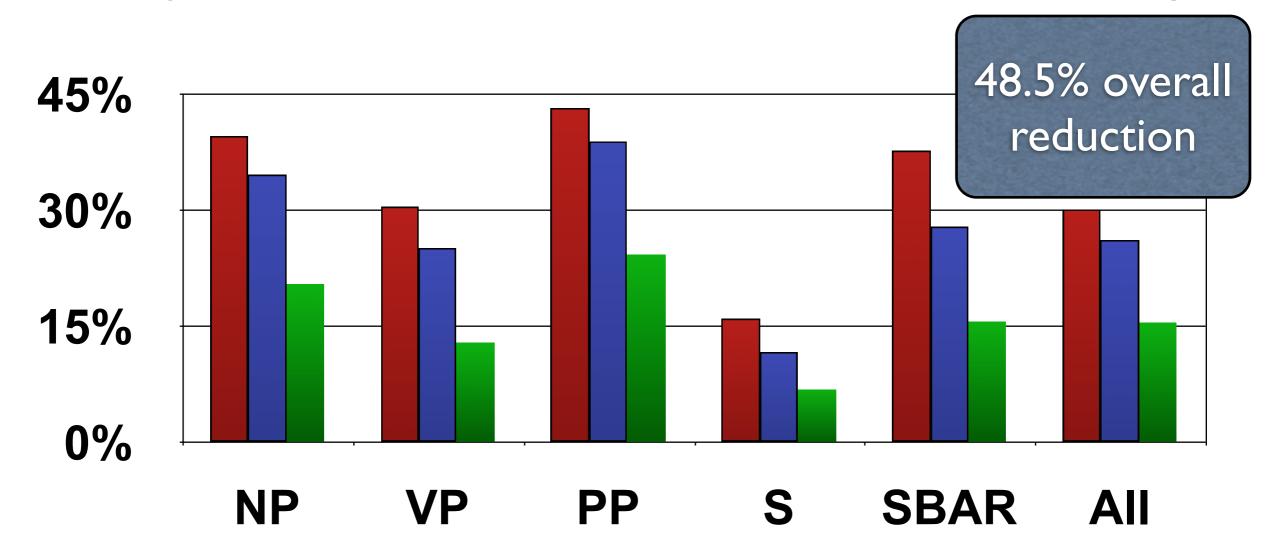
- Classic HMM
- Syntactic HMM
- Syntactic HMM with Competitive Thresholding





### Evaluation: Unproductive Constituent Rates

- Classic HMM
- Syntactic HMM
- Syntactic HMM with Competitive Thresholding







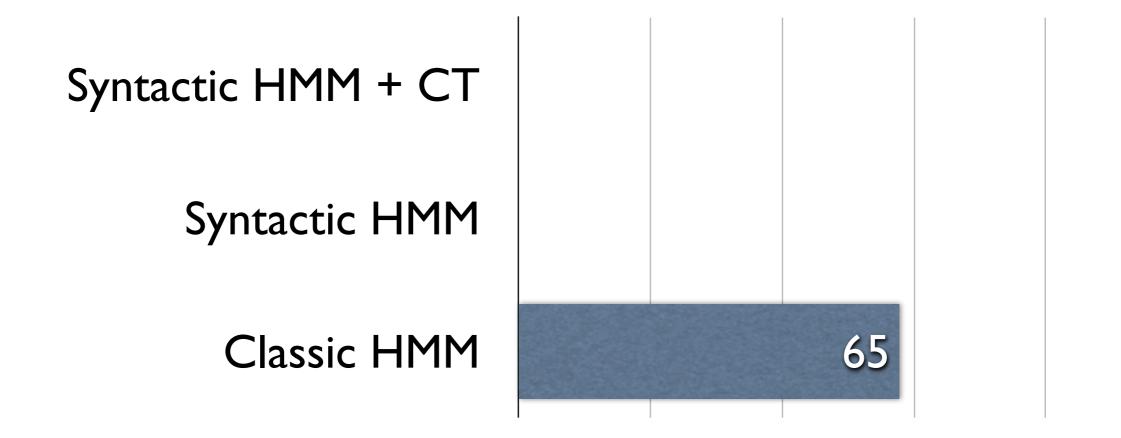
#### Rules extracted per sentence

Syntactic HMM + CT

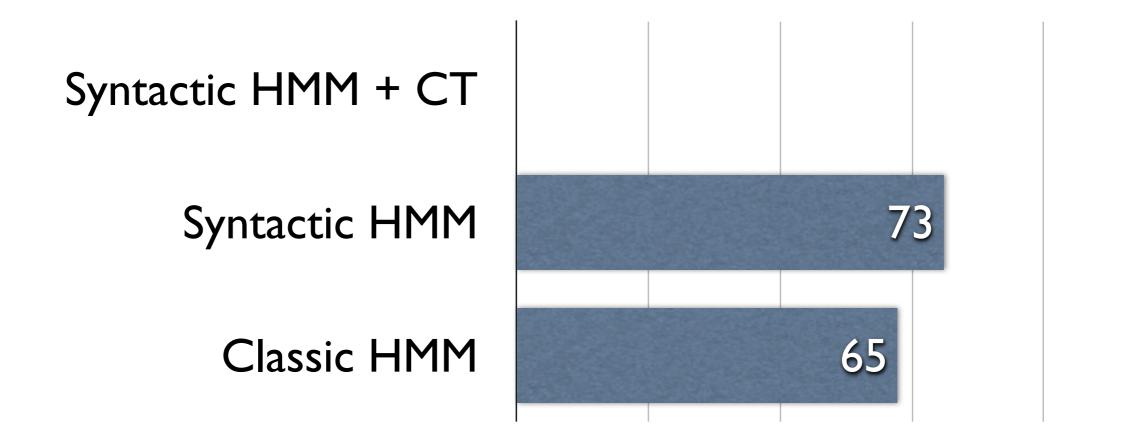
Syntactic HMM

Classic HMM

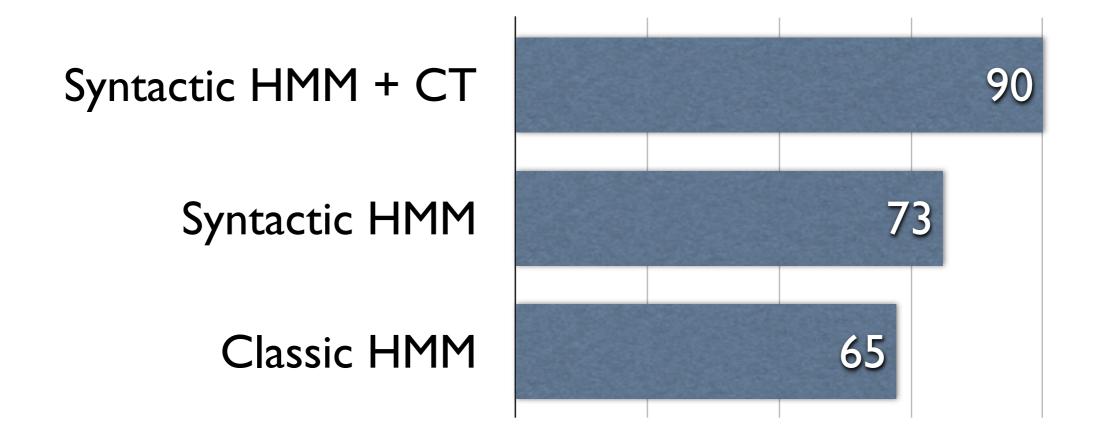














### Evaluation: Comparing Gold and Induced Rules

#### **Evaluation Metric Idea:**

Compare rules from gold alignments and induced alignments on both precision and recall.

Analog to the consistent phrase error rate (CPER) metric of Ayan & Dorr (06)



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Classic HMM
Syntactic HMM
Syntactic HMM + CT

Precision	Recall	FI
40.4	33.9	36.8
41.3	36.7	38.9
39.6	41.1	40.3



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#### **Evaluation Metric Idea:**

Compare rules from gold alignments and induced alignments on both precision and recall.

Analog to the consistent phrase error rate (CPER) metric of Ayan & Dorr (06)

	Precision	Recall	<u> </u>
Classic HMM	40.4	33.9	36.8
Syntactic HMM	41.3	36.7	38.9
Syntactic HMM + CT	39.6	41.1	40.3

FI Increase: 9.5% in Chinese; 18.7% in French



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- Future work: end-to-end translation (BLEU)



# Coming 07/07: BerkeleyAligner Software Package

- Agreement training of IBM models, which reduces AER 32% relative to GIZA++ (Liang et al., 06)
- Syntactic distortion model (this paper)
- Posterior decoding heuristics (this paper)
- Evaluation code: searches for posterior thresholds, compares decoding methods, & tracks AER during training
- Easy integration with the Berkeley Parser
- Pure Java 1.5 will run on any platform

#### **Check it out:**

http://nlp.cs.berkeley.edu/pages/WordAligner.html

### Thank You



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