Neural Syntactic Preordering for Controlled Paraphrase Generation



Tanya Goyal and Greg Durrett ACL 2020



Input Sentence

If the Mumbai Indians win this match, they will reach the final.



seq2seq paraphrase model

Generated Output

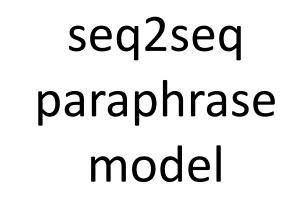
If Mumbai Indians win, they will proceed to the final.

If Mumbai Indians win, they will *reach* the finals.



Input Sentence

If the Mumbai Indians win this match, they will reach the final.



Generated Output

If Mumbai Indians win, they will **proceed** to the final.

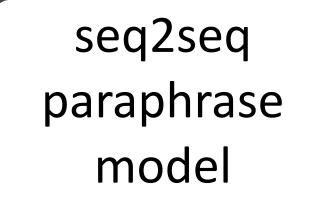
Small variations of the sentence!

If Mumbai Indians win, they will reach the finals.



Input Sentence

If the Mumbai Indians win this match, they will reach the final.



Paraphrase models should be able to achieve big structural reorderings!

Generated Output

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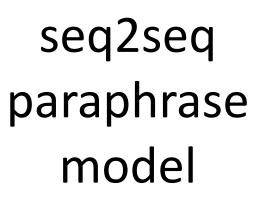


Input Sentence

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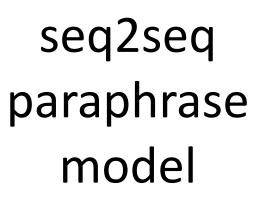


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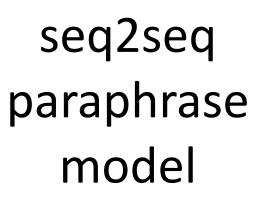
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Clausal reordering *if* — *they*



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Small variations of the sentence!

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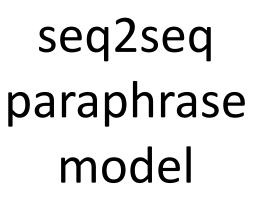
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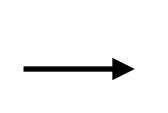
Small variations of the sentence!

Generated Output

If Mumbai Indians win, they will proceed to the final.

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If Mumbai Indians win this game, they will reach finals.



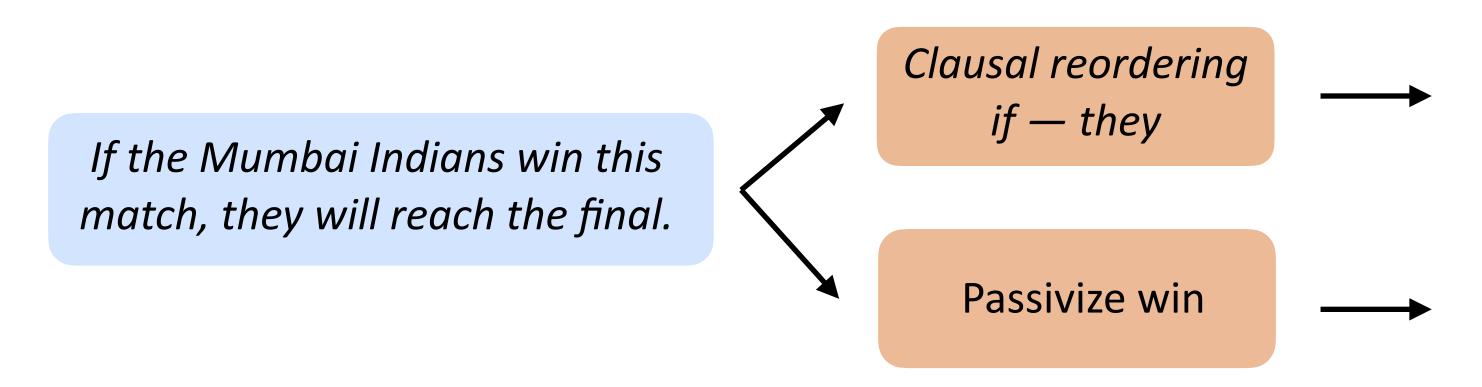
Mumbai Indians will reach the finals if they win this match.

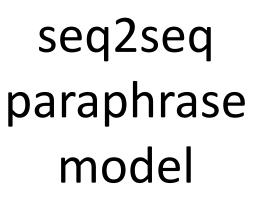


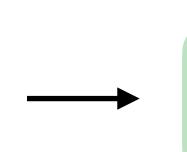
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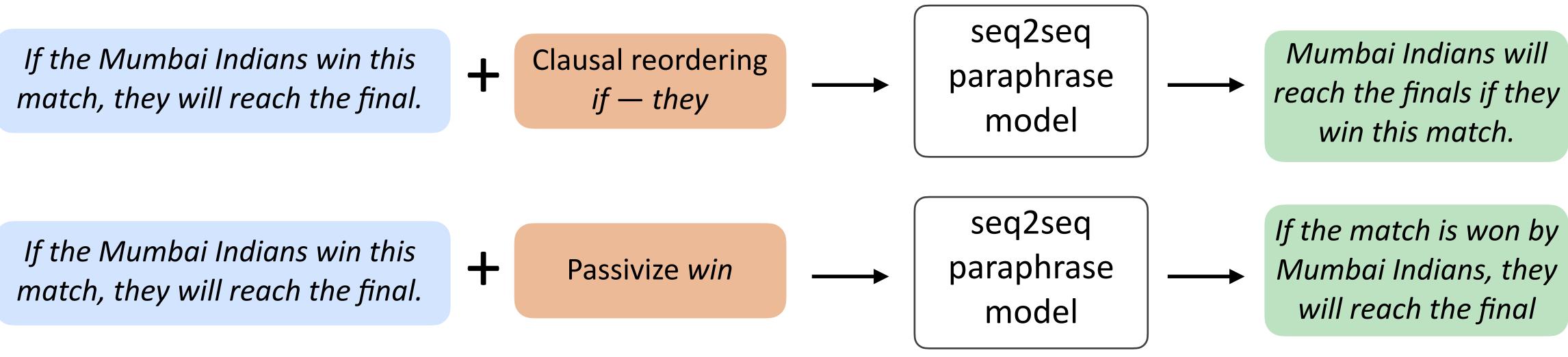
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Controllable Paraphrase Generation

Can we explicitly tell our seq2seq model to follow these desired structures? By controlling its behavior, we can achieve diversity!

. . .







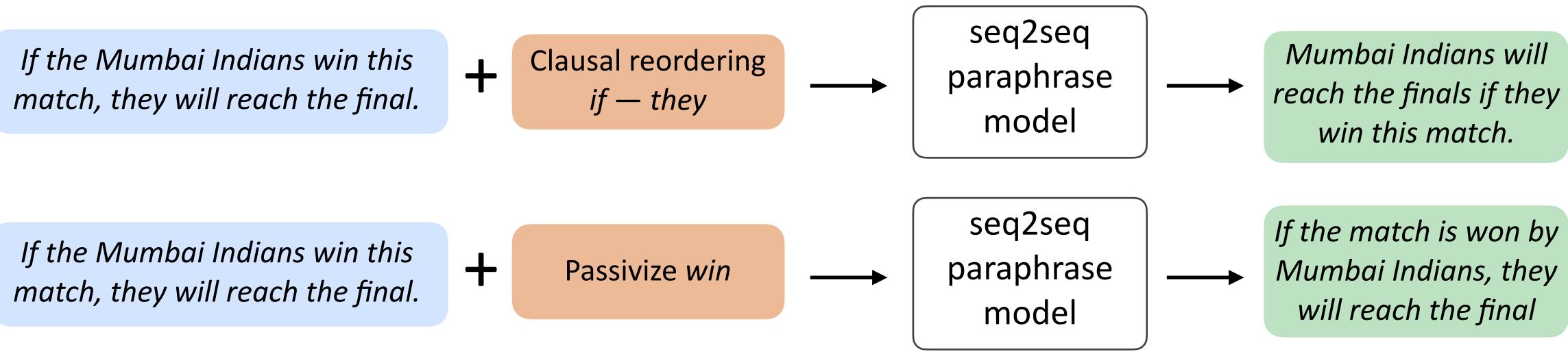




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How do we specify structure that the generation conditions on?





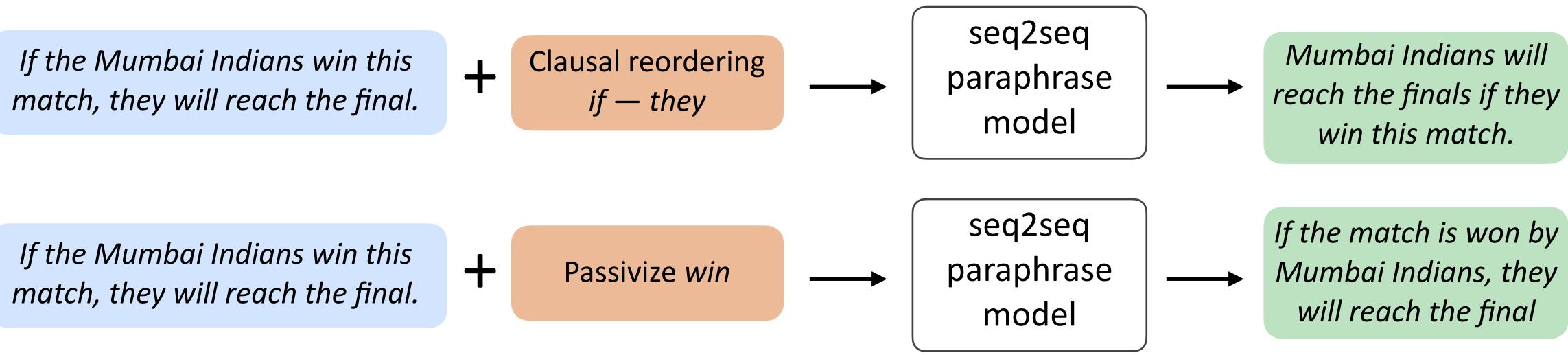




Controllable Paraphrase Generation

Can we explicitly tell our seq2seq model to follow these desired structures? By controlling its behavior, we can achieve diversity!

. . .



How do we specify structure that the generation conditions on?

How do we enumerate the different paraphrasing possibilities?









Exemplar-based (Chen et al. 2019)

Prior Work



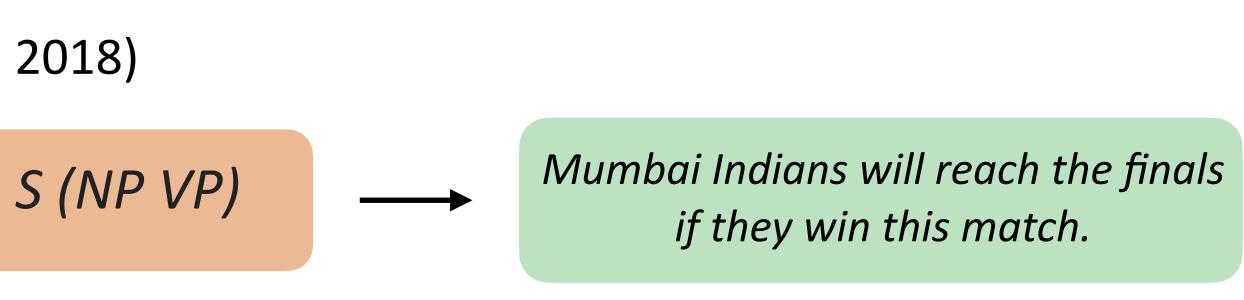
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Syntactic Template-based (lyyer et al. 2018)

If the Mumbai Indians win this match, they will reach the final.

Exemplar-based (Chen et al. 2019)

Prior Work



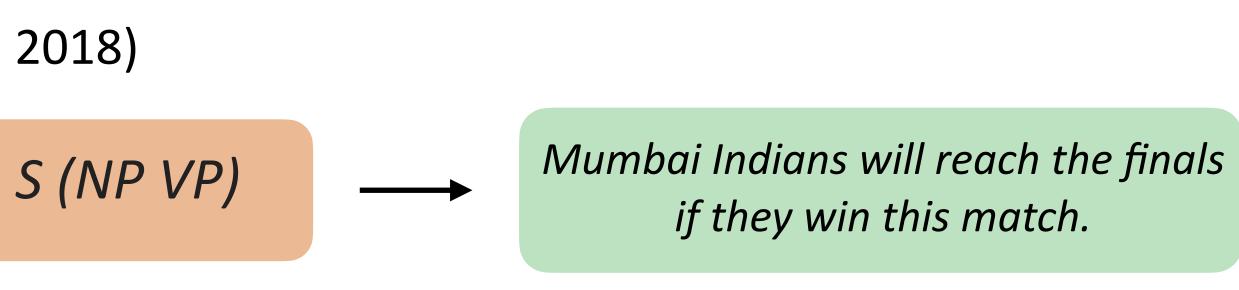


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Exemplar-based (Chen et al. 2019)

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Limitation: Uses the same set of templates for ALL inputs, cannot identify input-appropriate templates.



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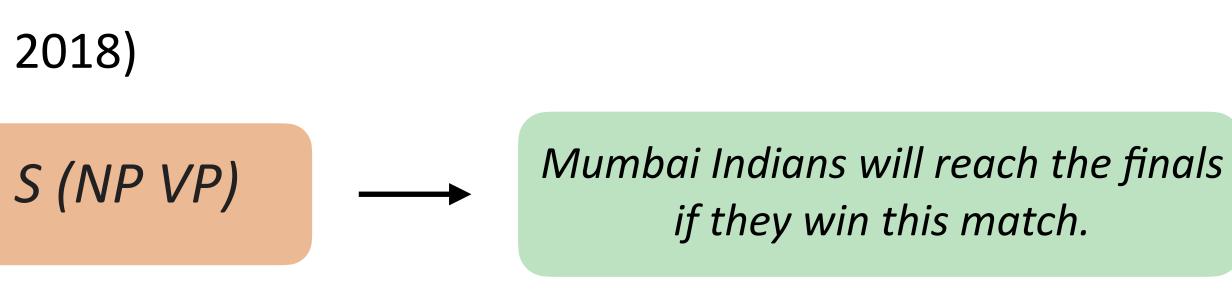
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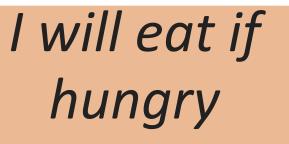
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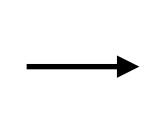
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Prior Work







Mumbai Indians will reach the finals *if they win this match.*



If the Mumbai Indians win this match, they will reach the final.

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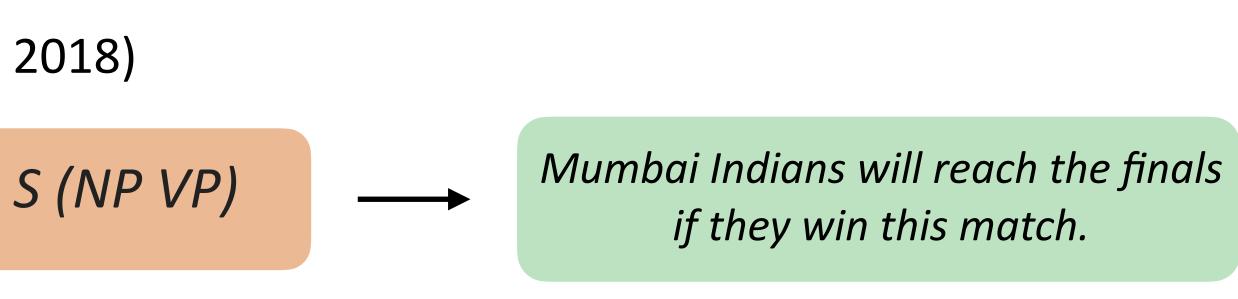
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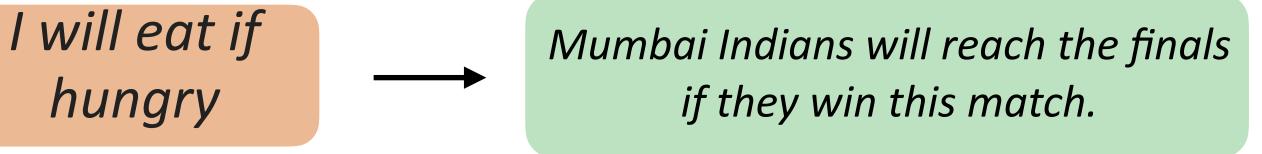
If the Mumbai Indians win this match, they will reach the final. ╋

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Limitation: Requires appropriate exemplar as input, cannot generate these automatically.

Prior Work





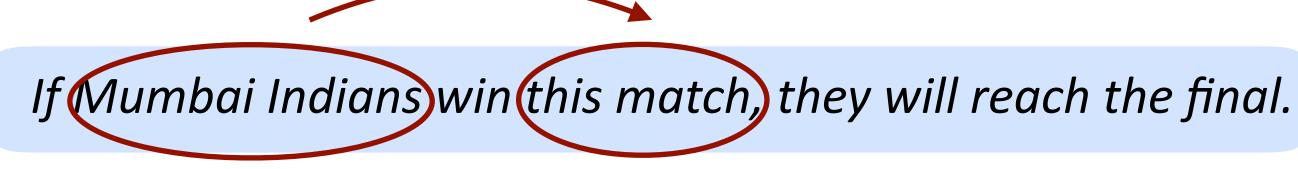


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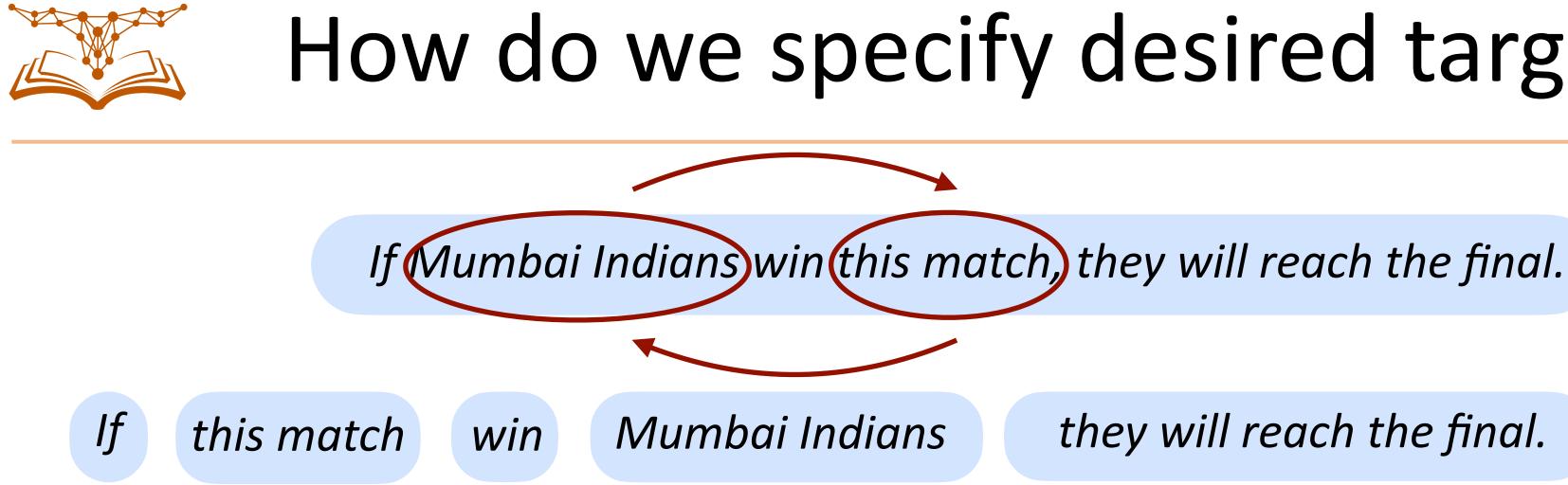




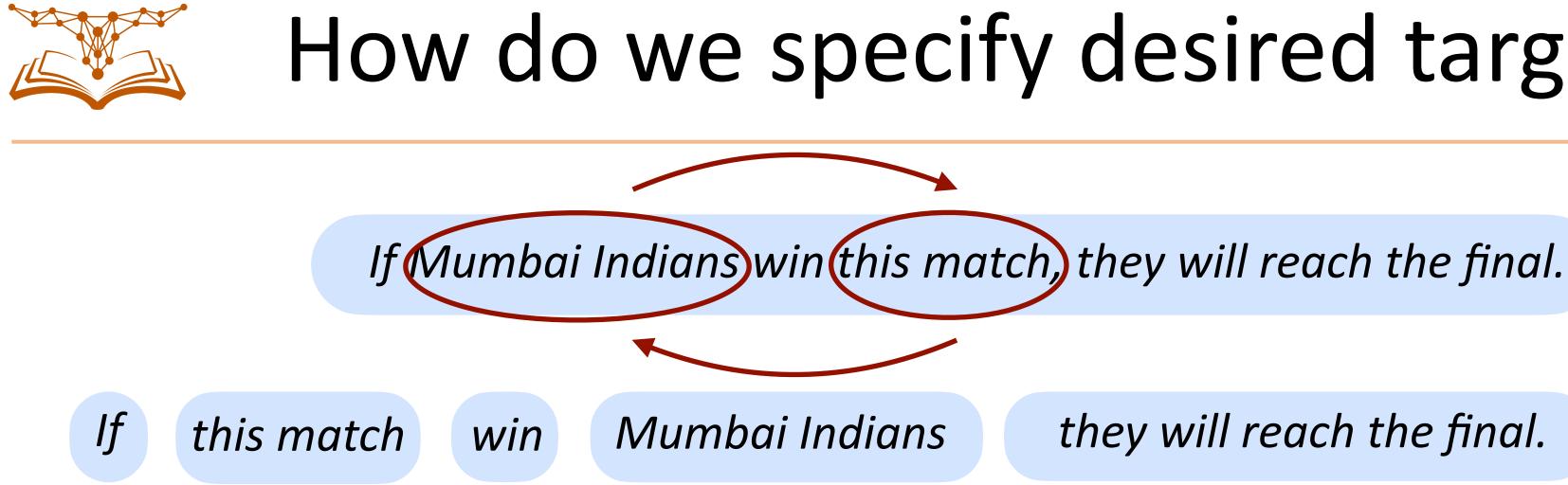








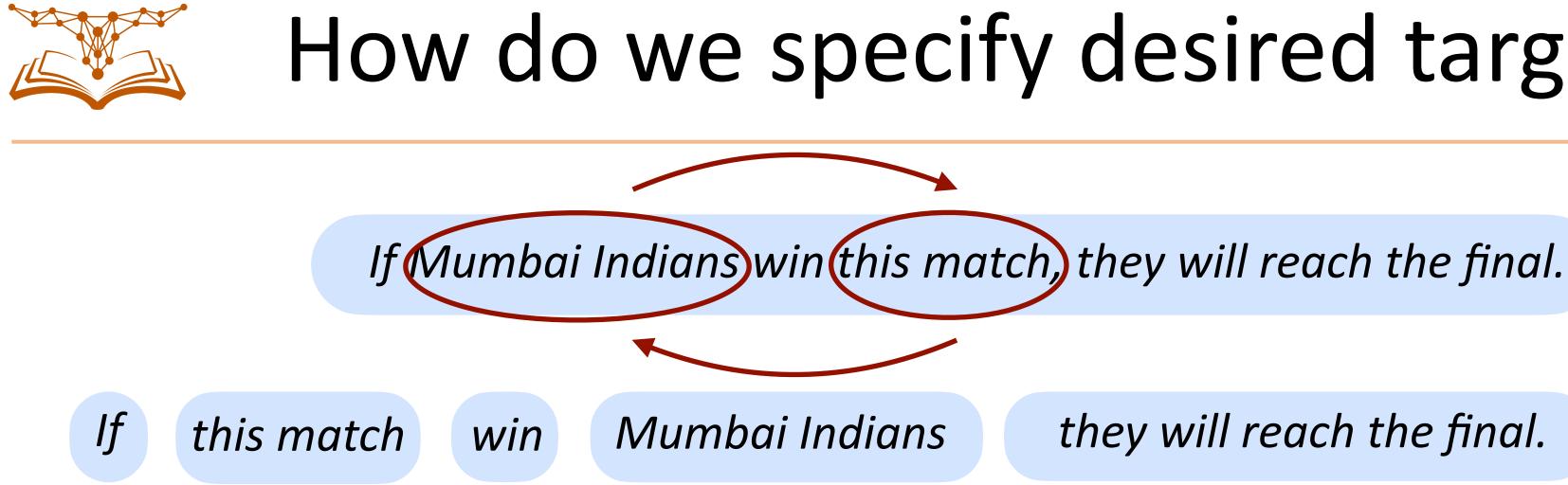
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Similar to preordering in MT and denotes the right order of content

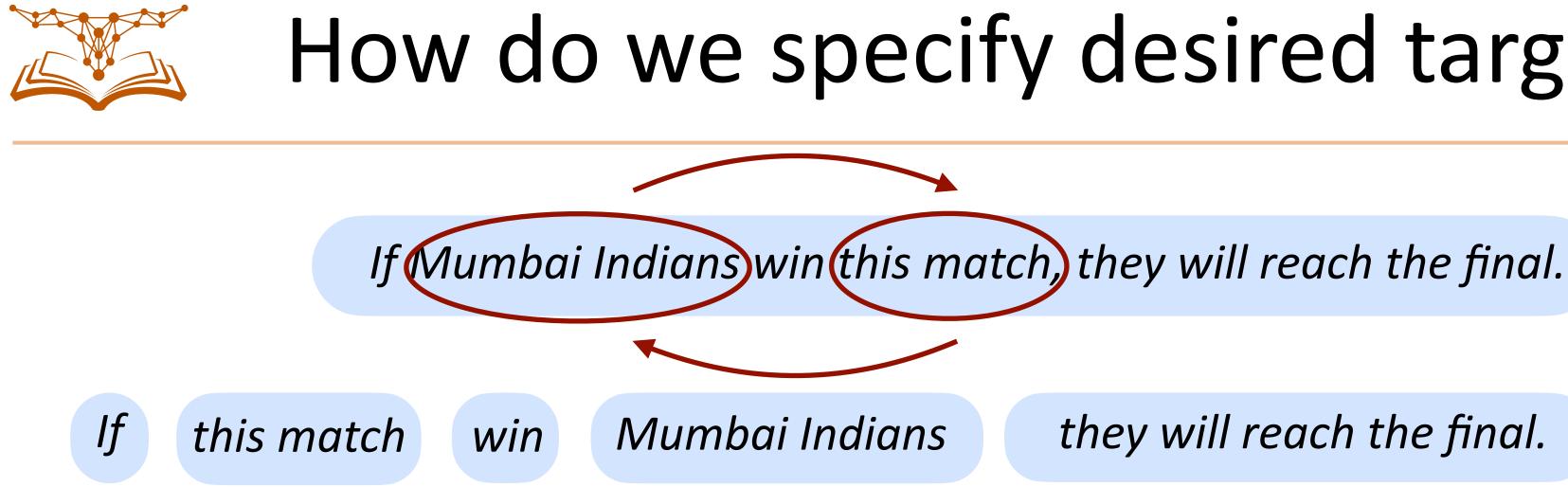




they will reach the final.

Similar to preordering in MT and denotes the right order of content Does not retain semantics.





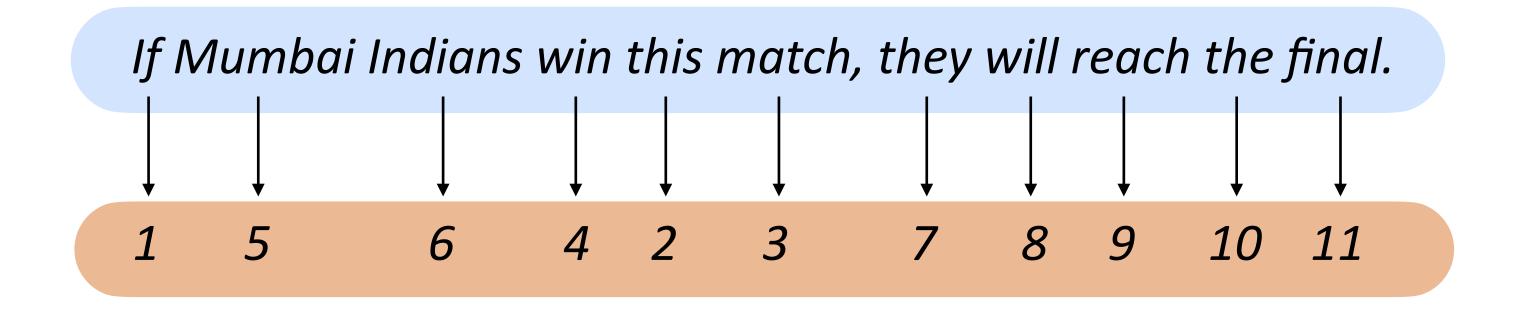
If the match is won by Mumbai Indians, they will reach the final.

How do we specify desired target structure?

they will reach the final.





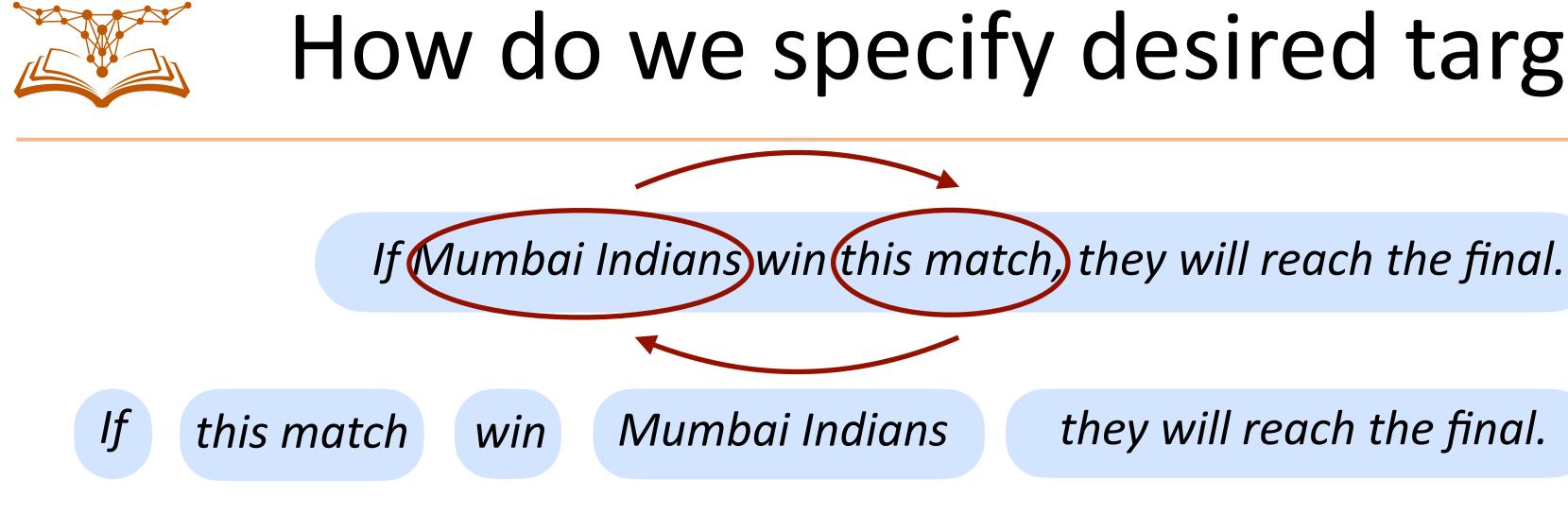


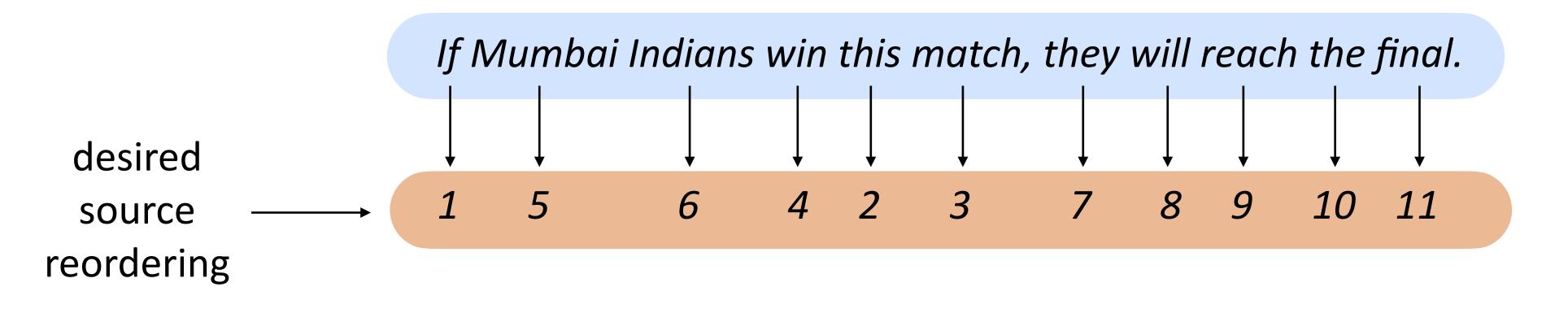
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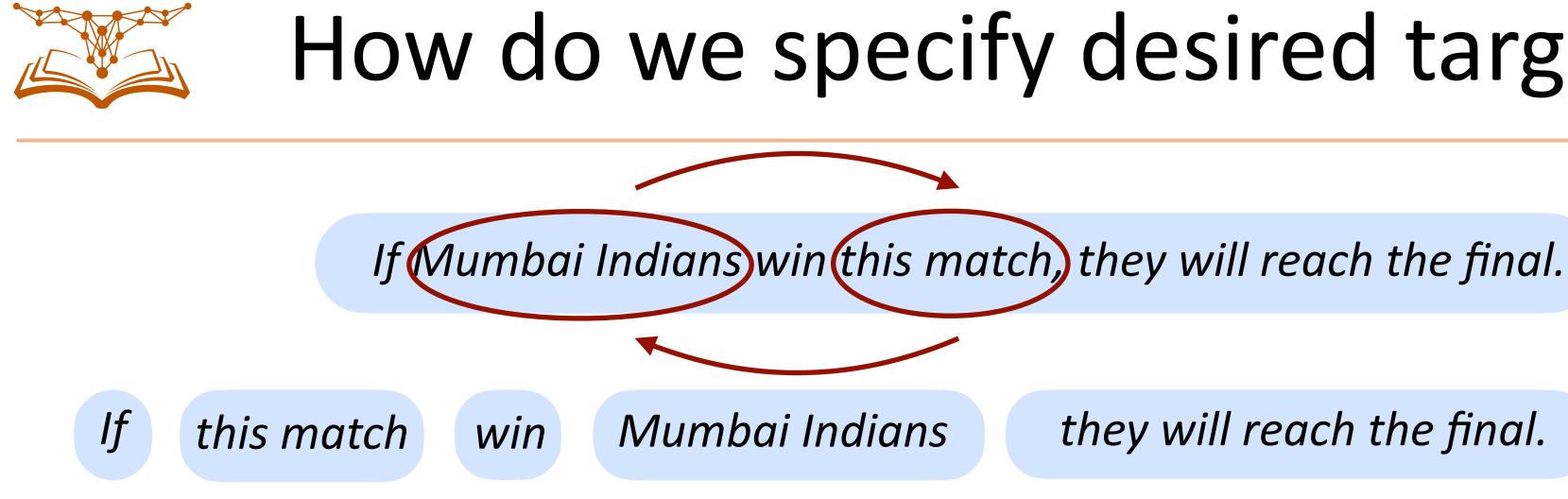


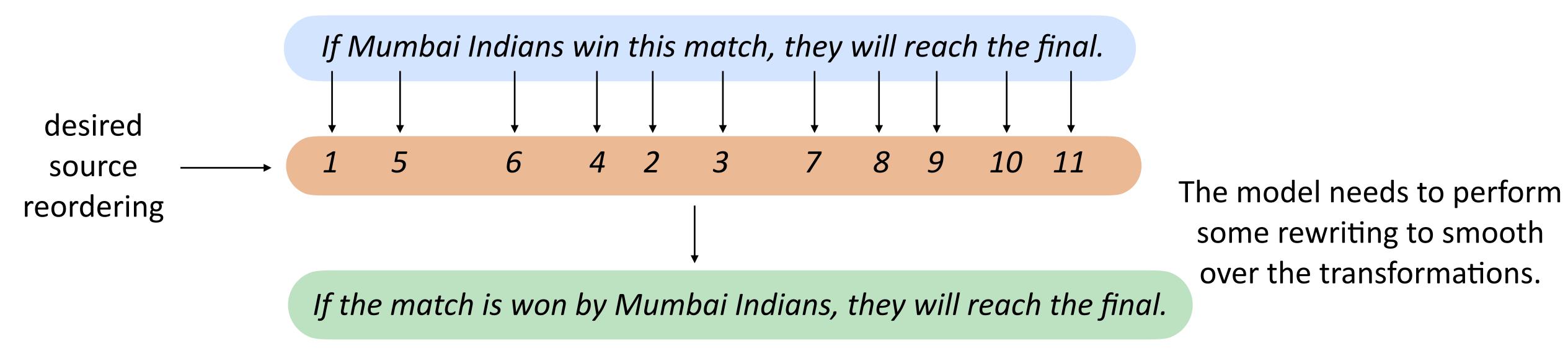
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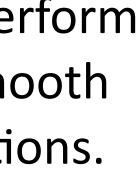






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Learn to reorder abstracted phrases instead of the whole sentence



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By considering multiple phrase abstractions, we can enumerate many structures



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(captures reordering like the grammar rules in syntactic MT)





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NP₁

Now use this to derive the input reordering:

How do we generate the source reordering?

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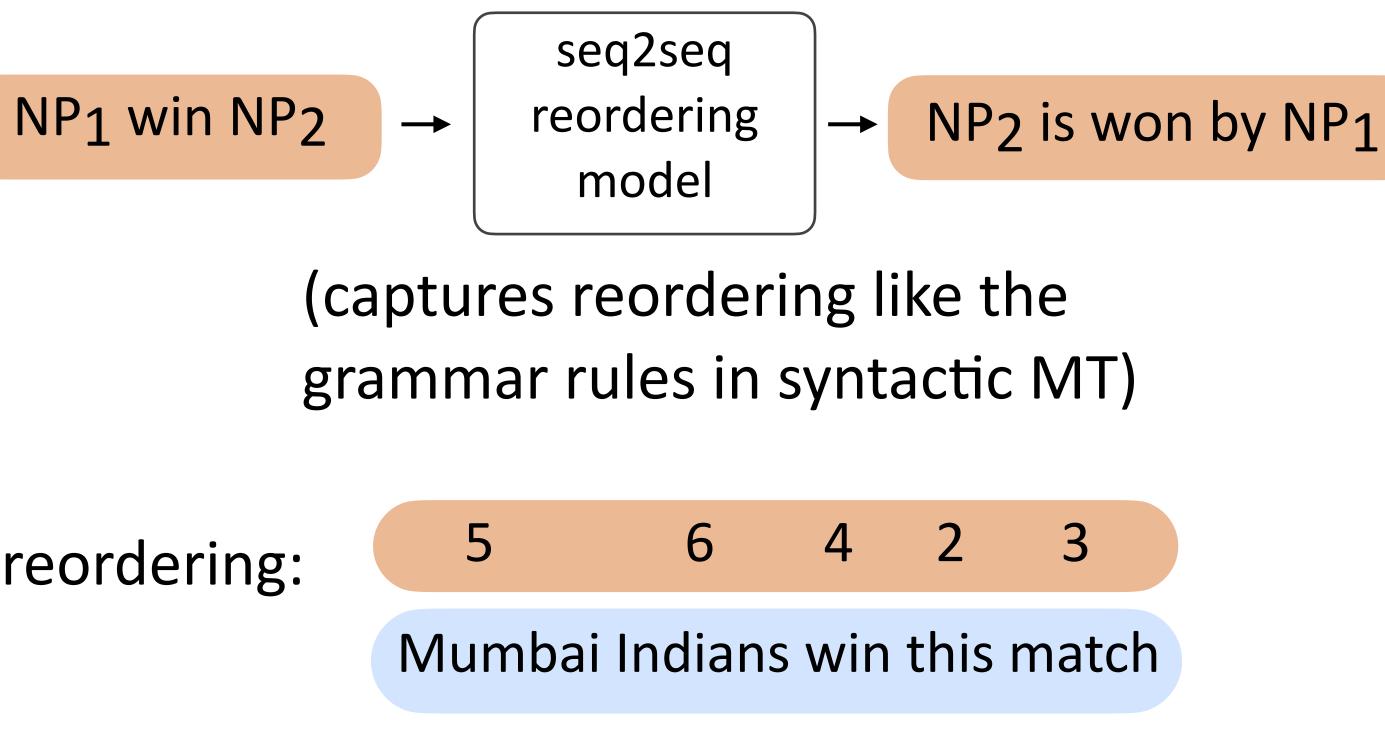




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Overview

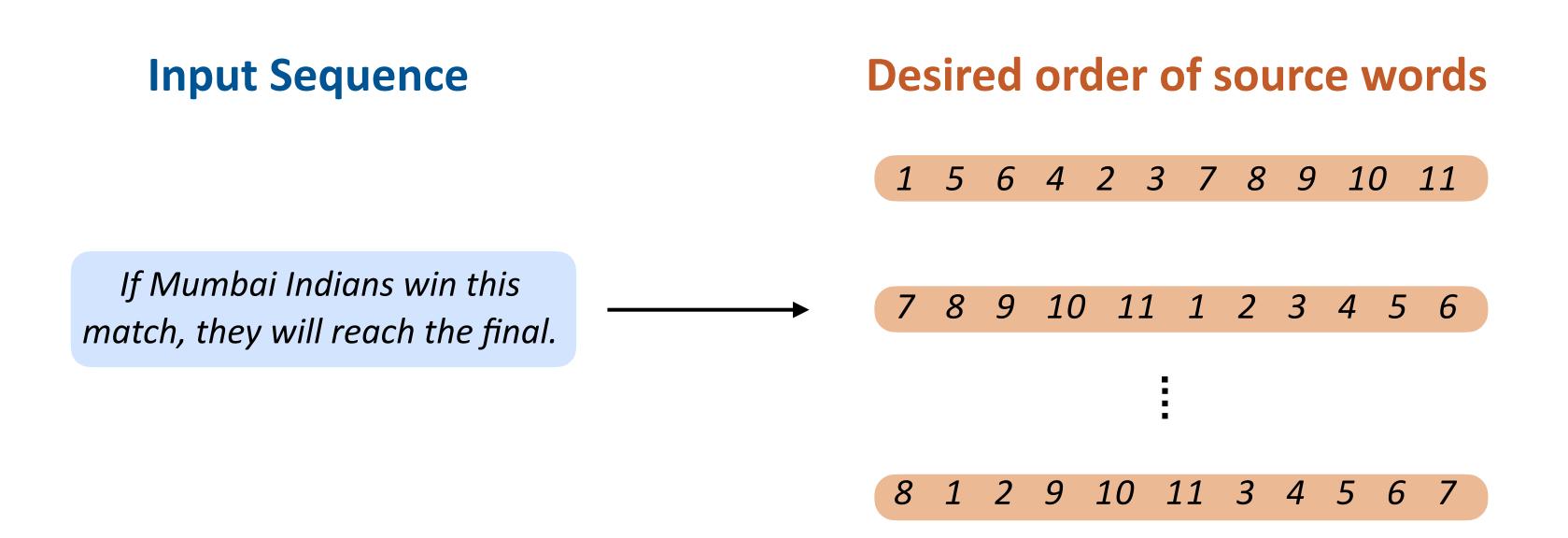


Input Sequence

If Mumbai Indians win this match, they will reach the final.

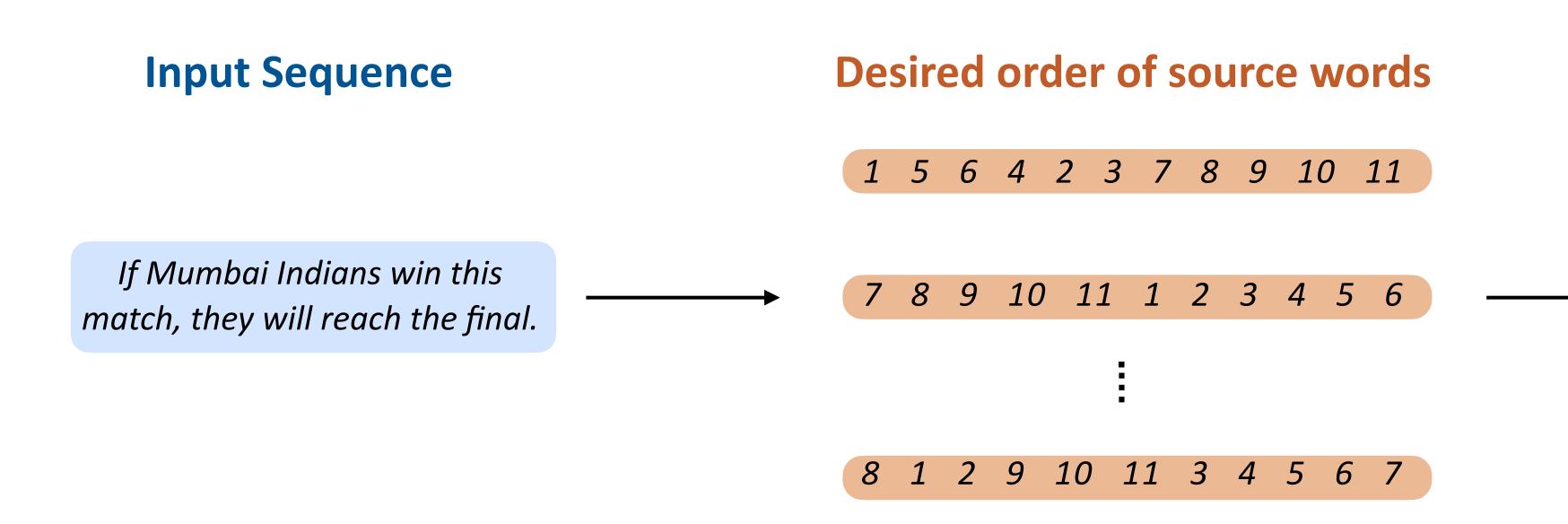
Overview





Overview





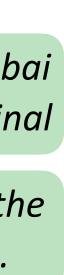
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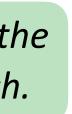
Generated Output

If the match is won by Mumbai Indians they will reach the final

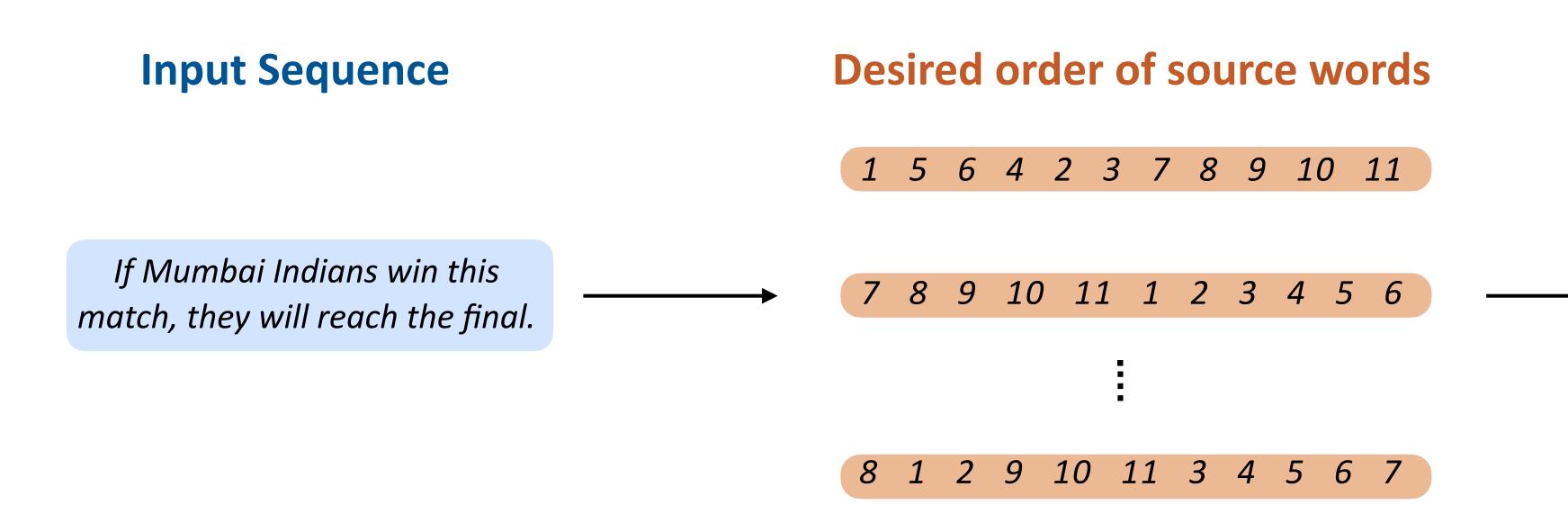
Mumbai Indians will reach the finals if this match is won.

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Sow: Source Order reWriting

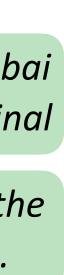
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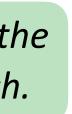
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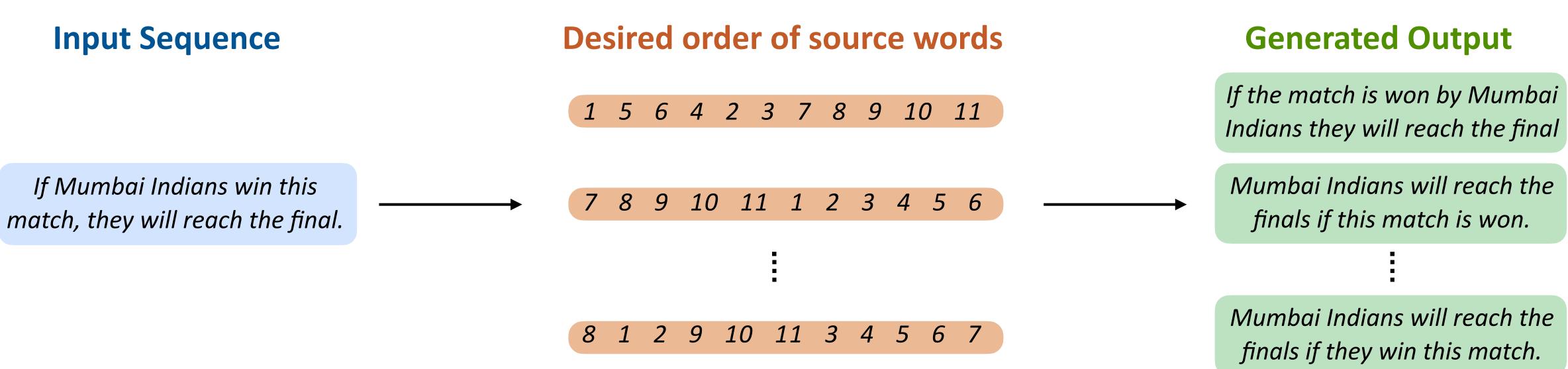
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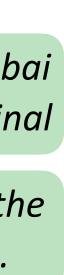




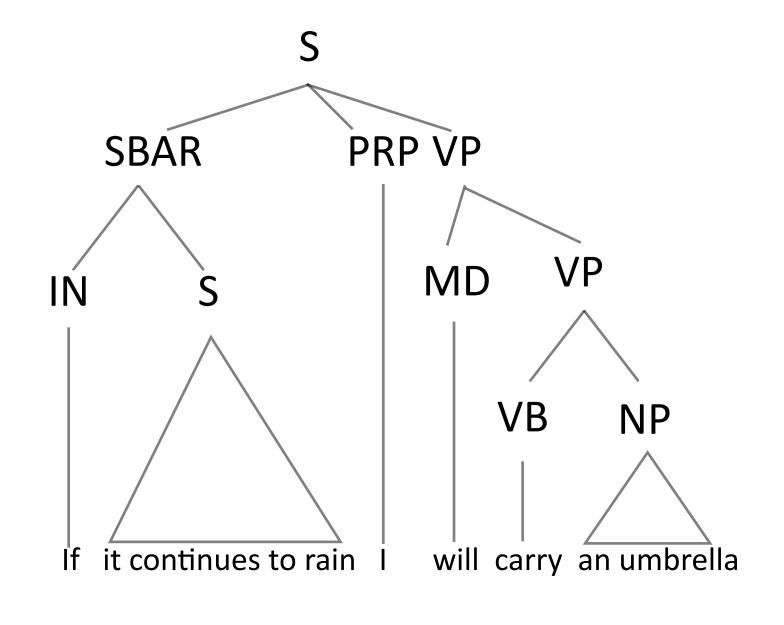


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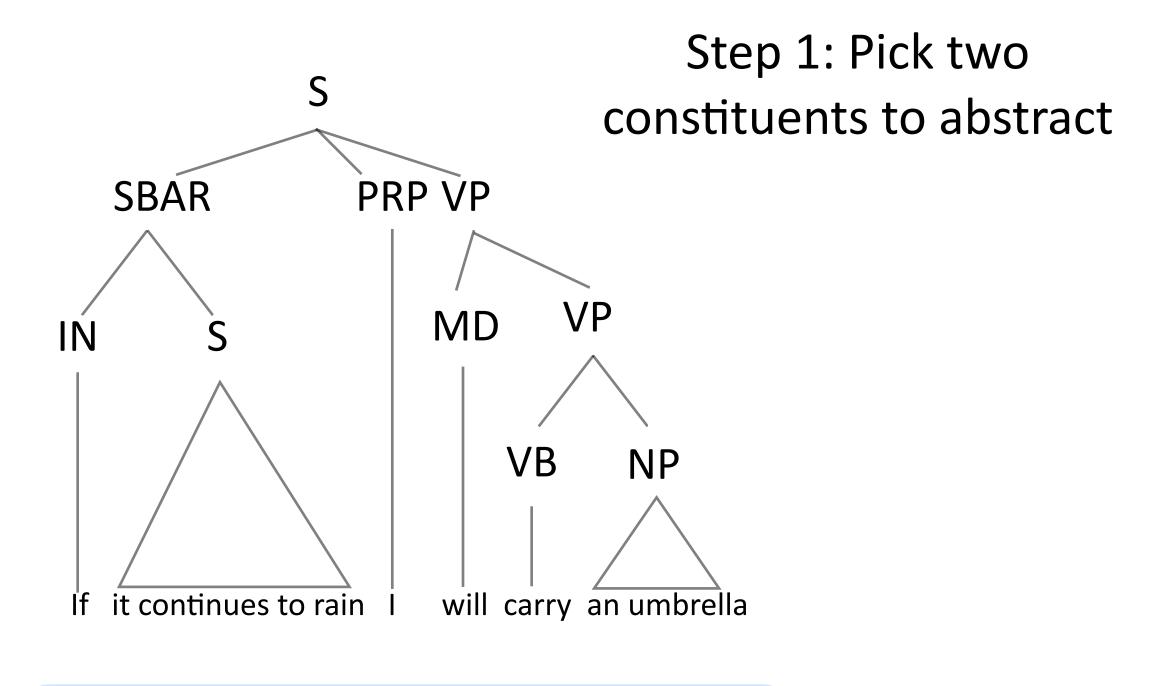






SOW model: Source Order Rewriting



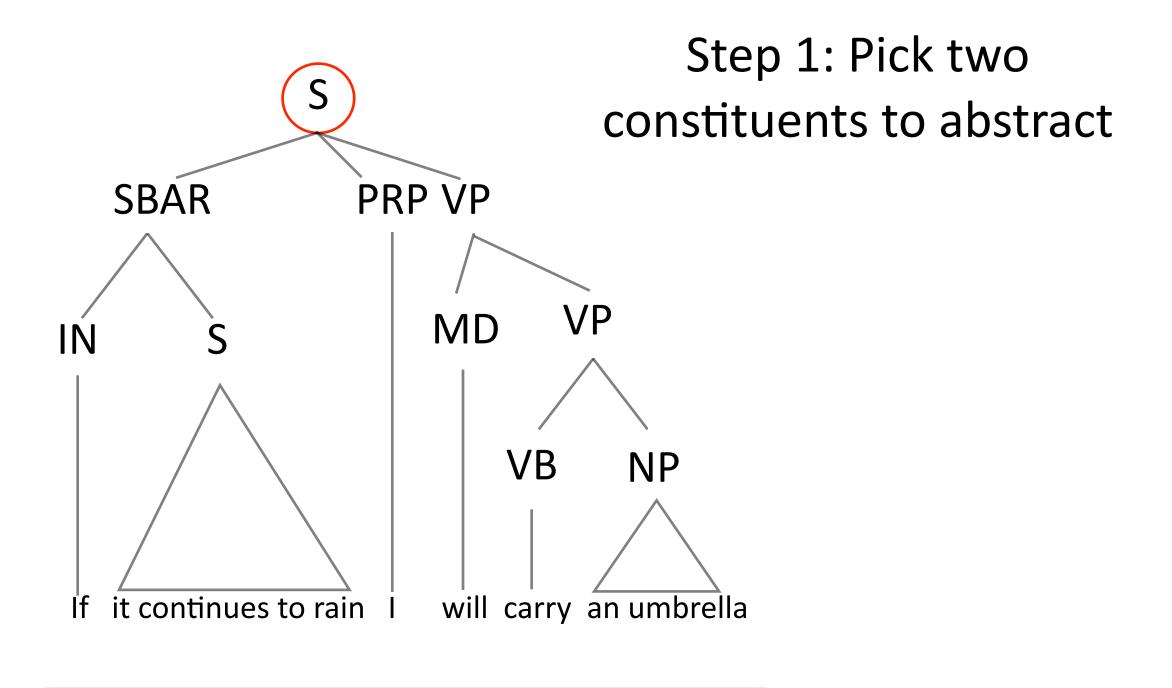


SOW model: Source Order Rewriting

Goal: Given an input sentence, generate a set of appropriate reorderings.

Step 2: Reorder with a seq2seq model



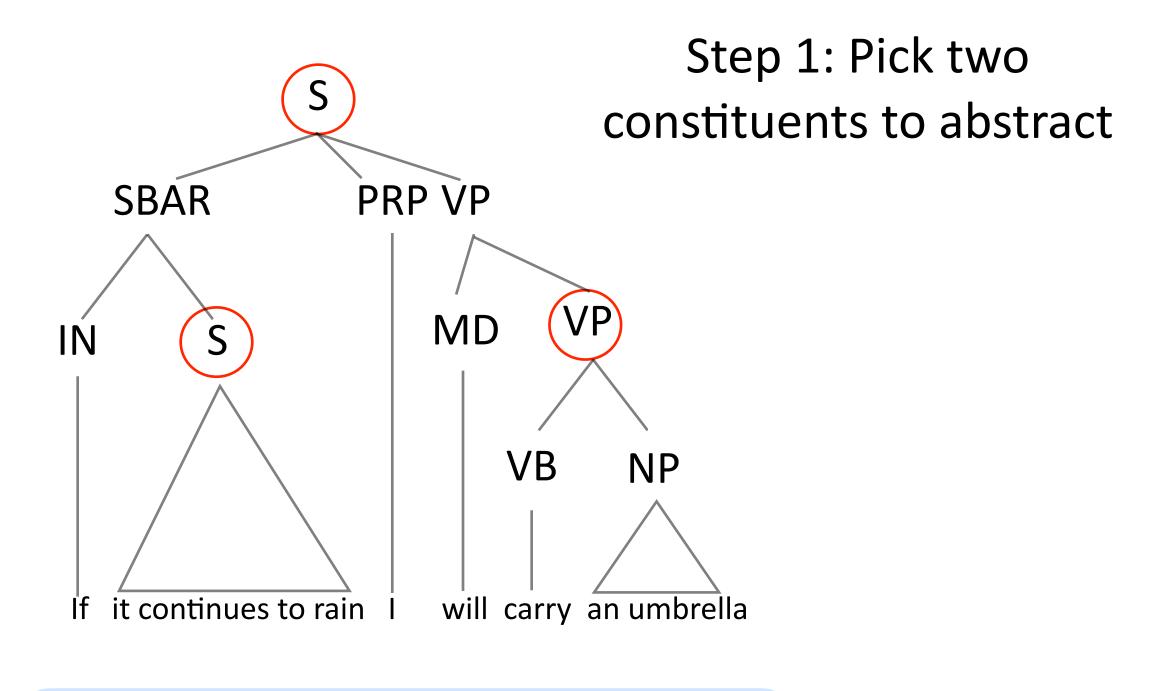


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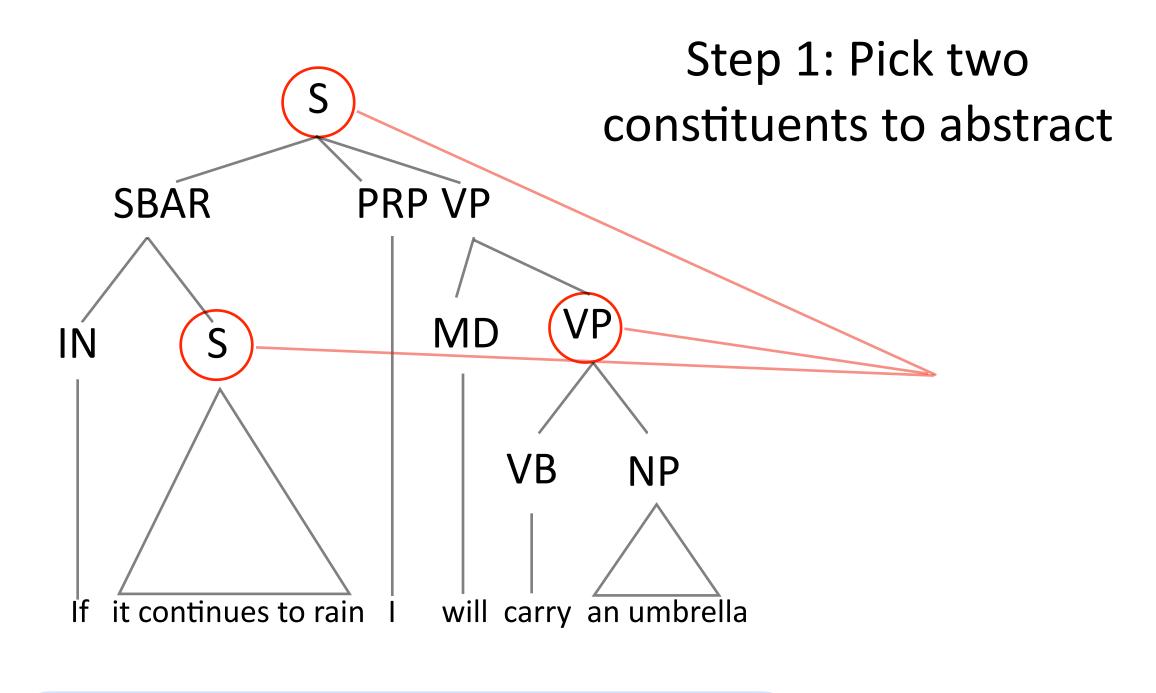


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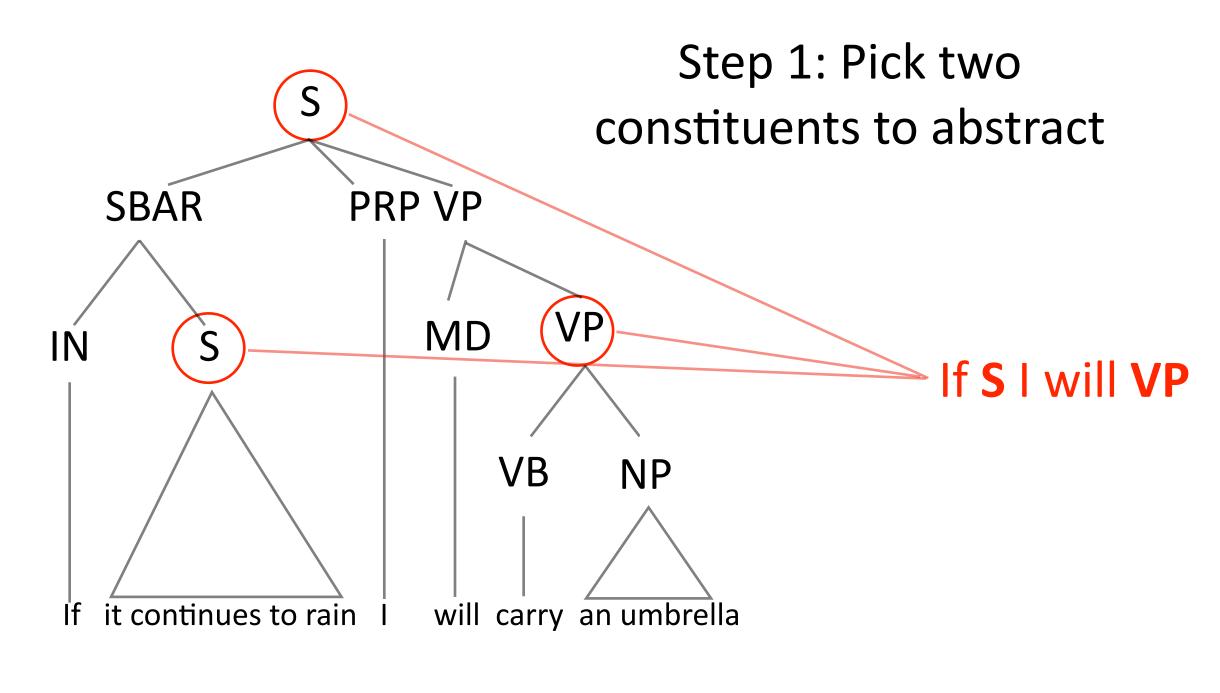


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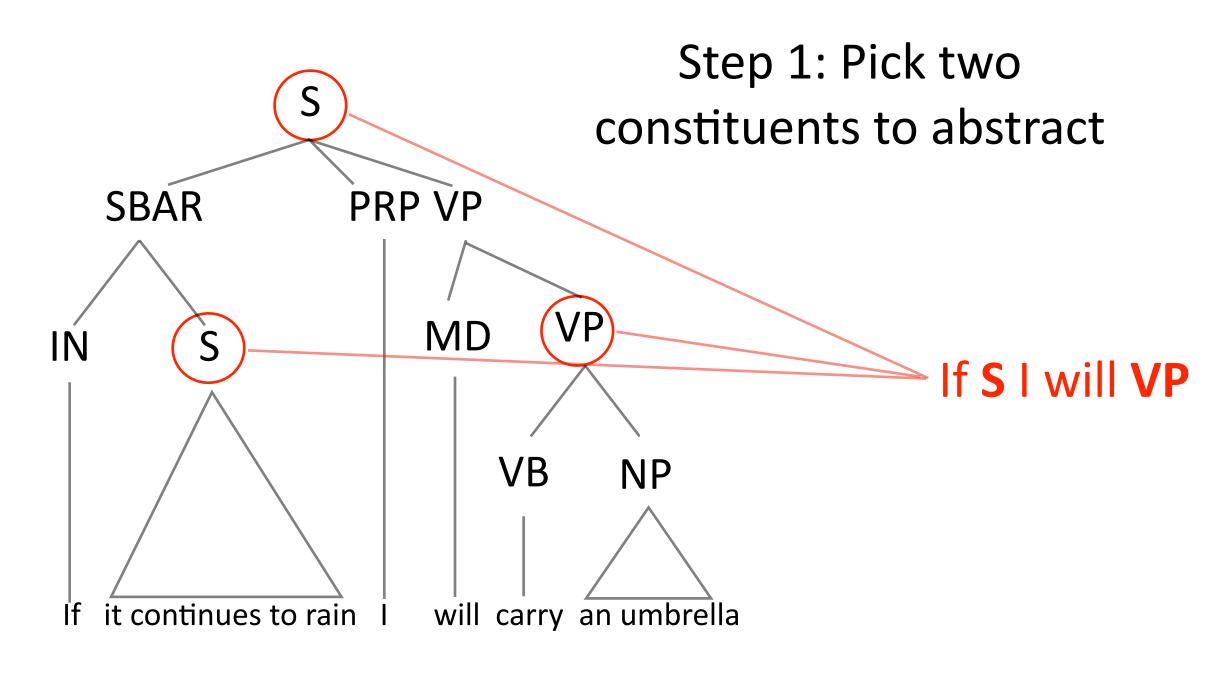


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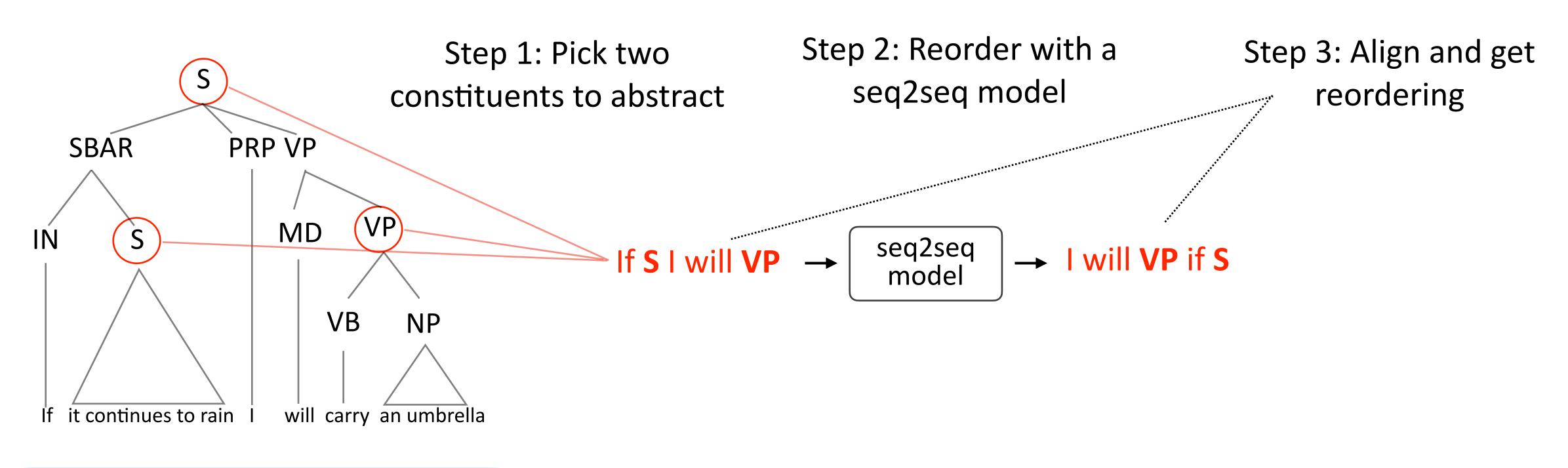
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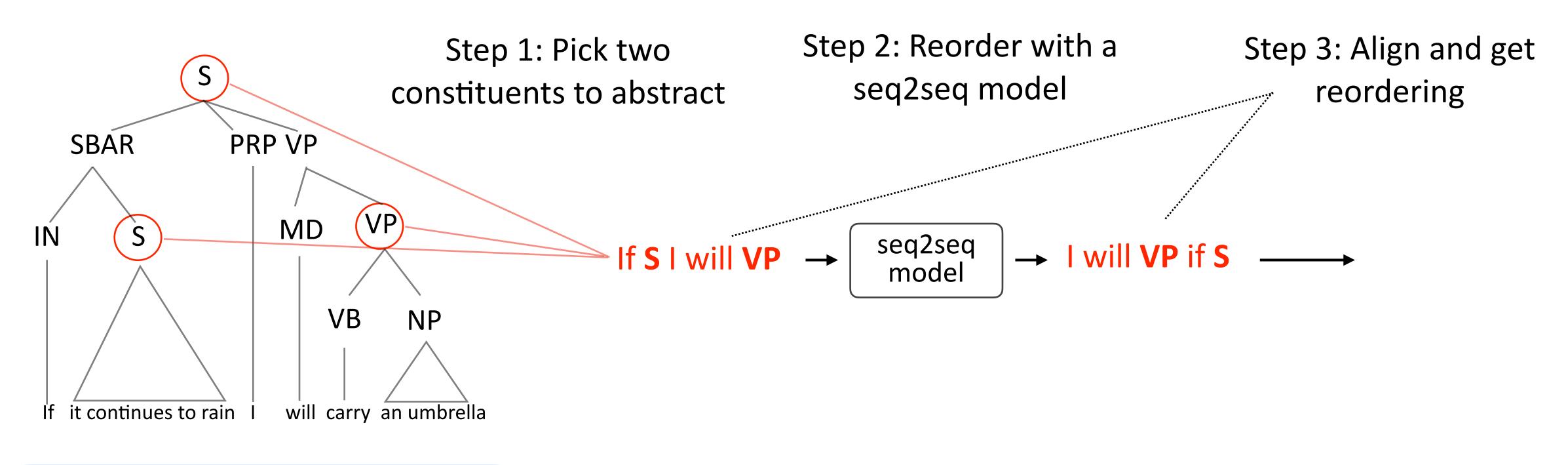
$$\rightarrow \boxed{\begin{array}{c} \text{seq2seq} \\ \text{model} \end{array}} \rightarrow 1 \text{ will VP if S}$$





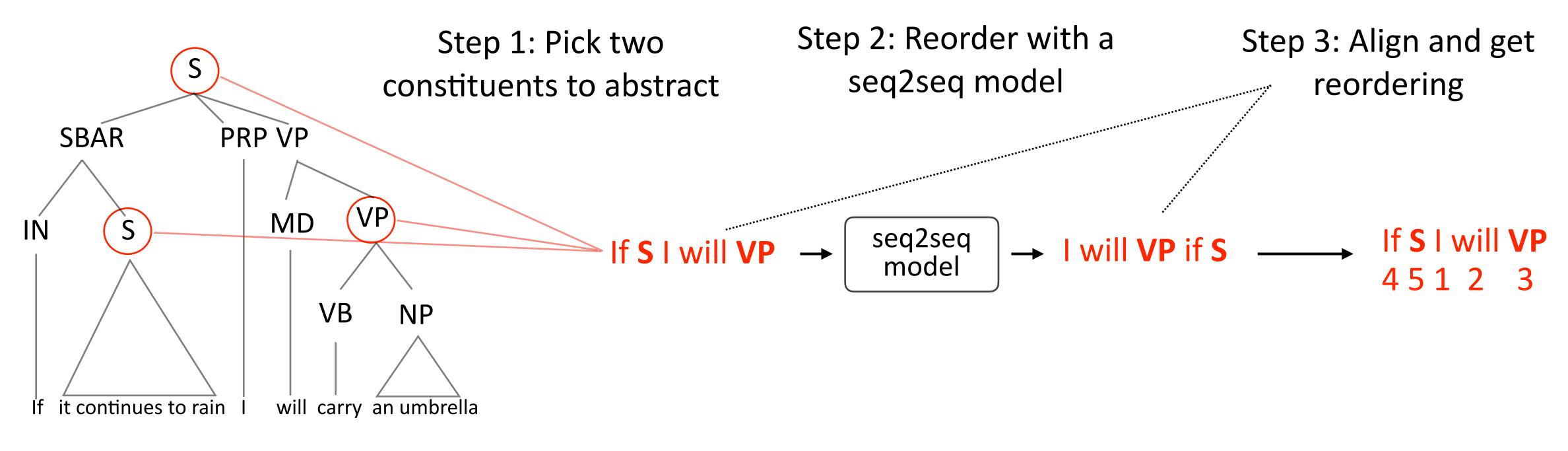
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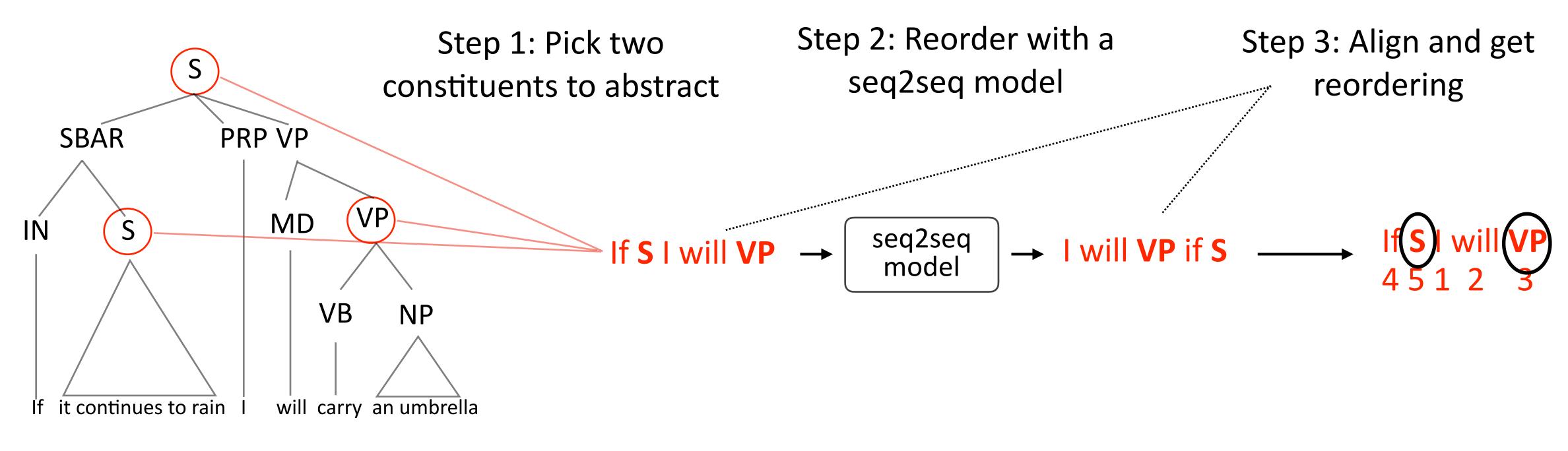
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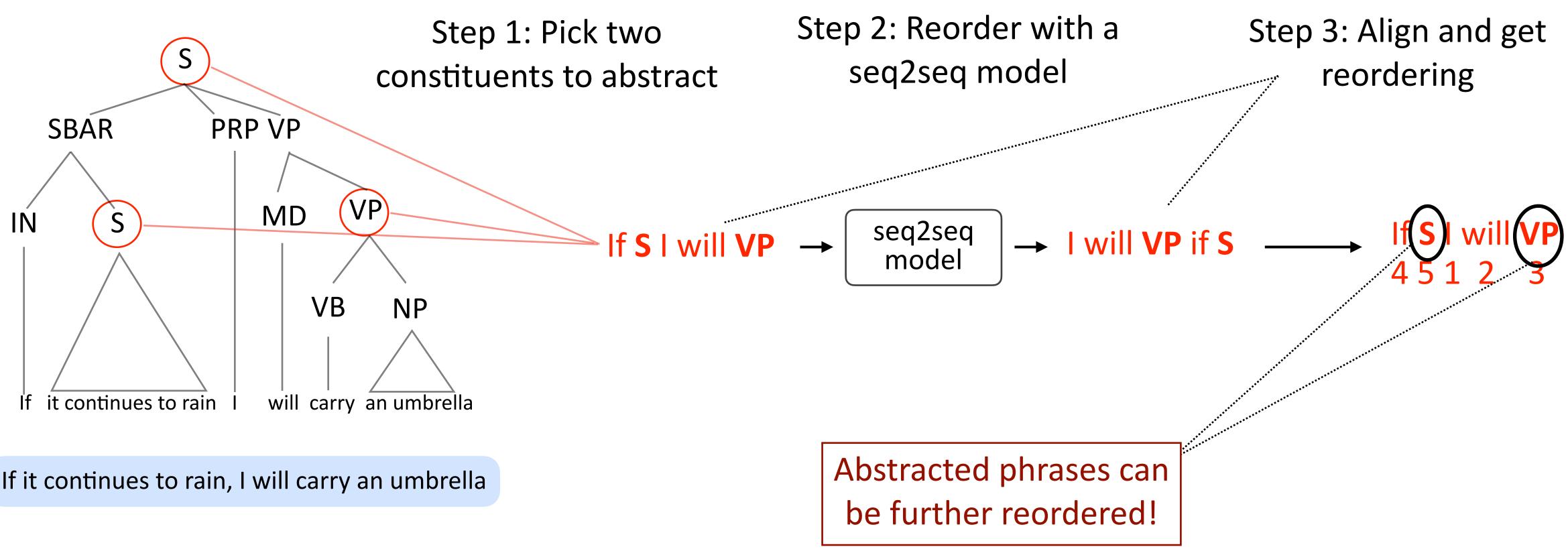
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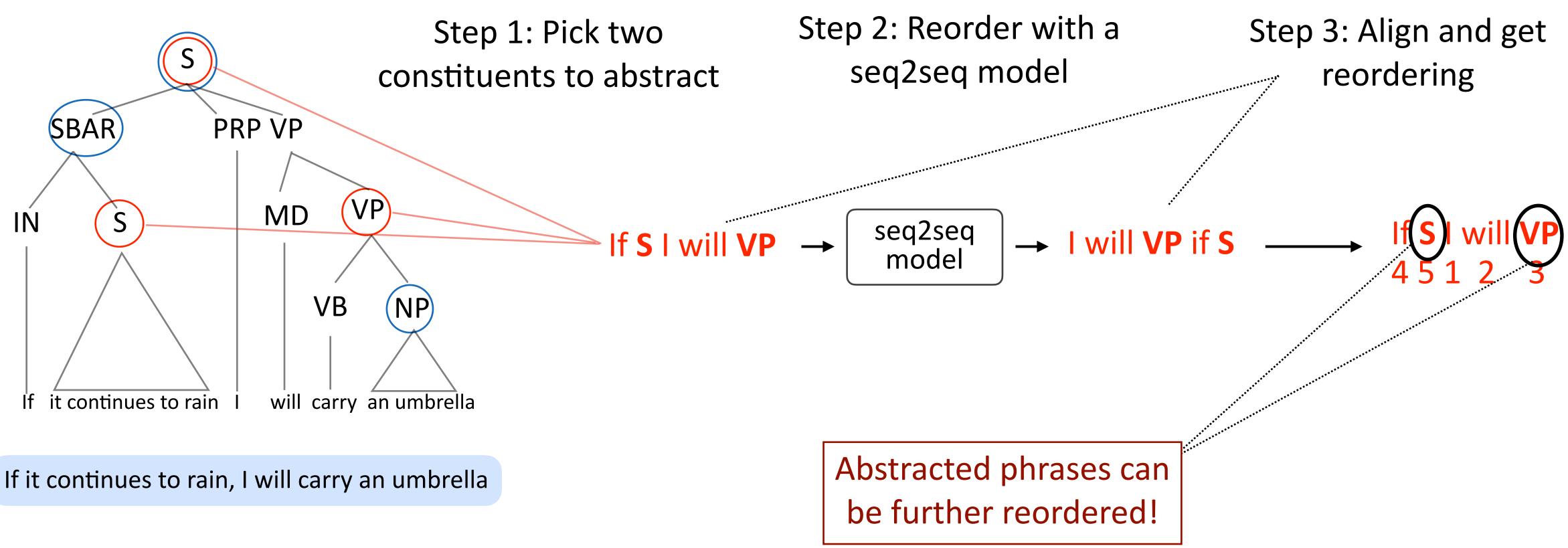
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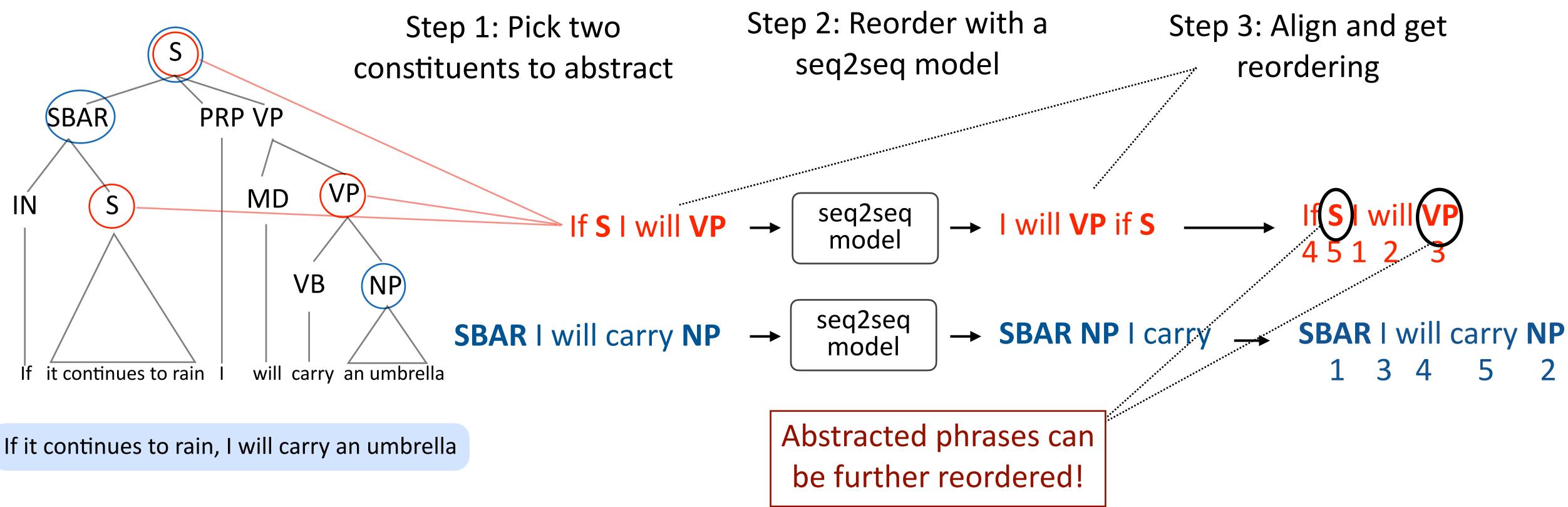
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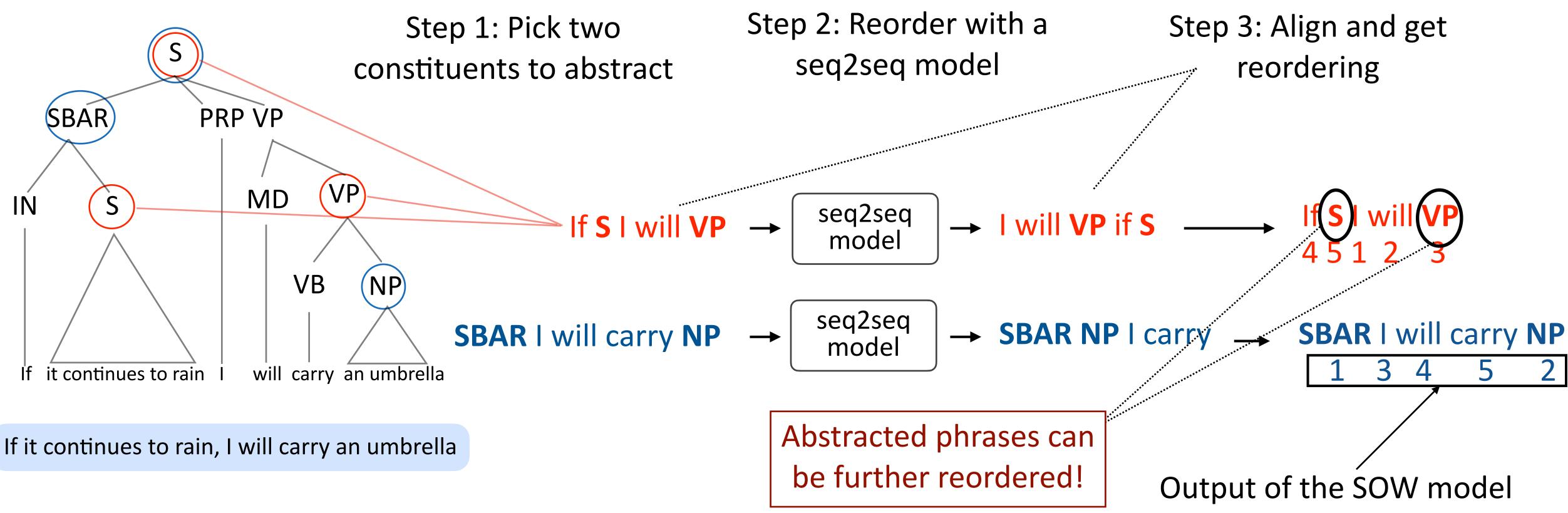
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SOW model: Source Order Rewriting

Training Data: SOW



- We want phrase pairs of the following kind:
 - If $S | will VP \longrightarrow | will VP$ if S
 - removing the NN from NP NP was excluded from NN

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- If it continues to rain I will carry an umbrella
 - I will carry an umbrella if rain continues



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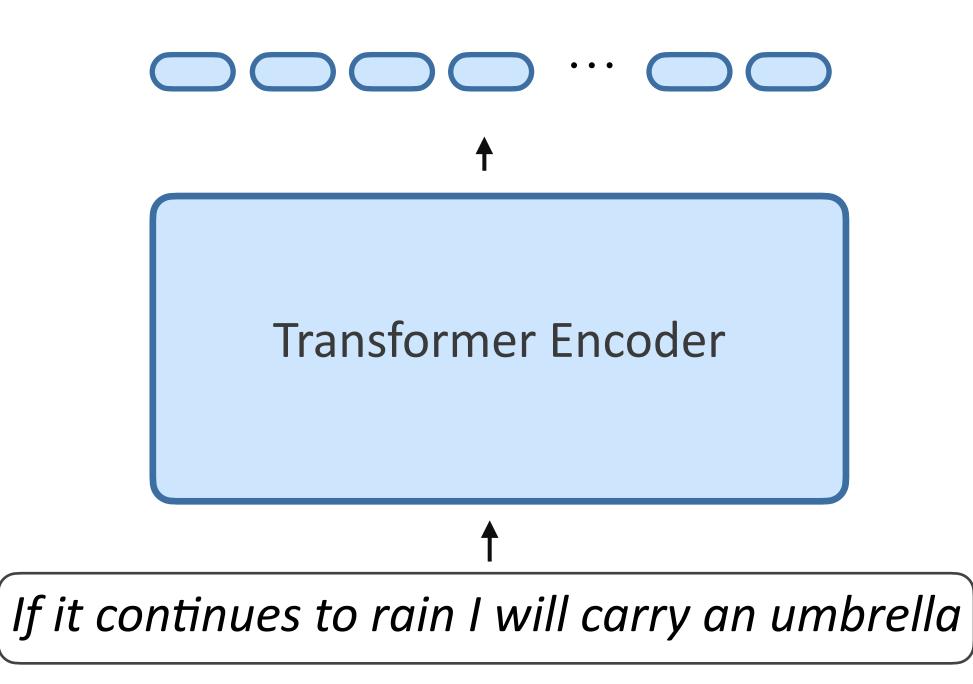




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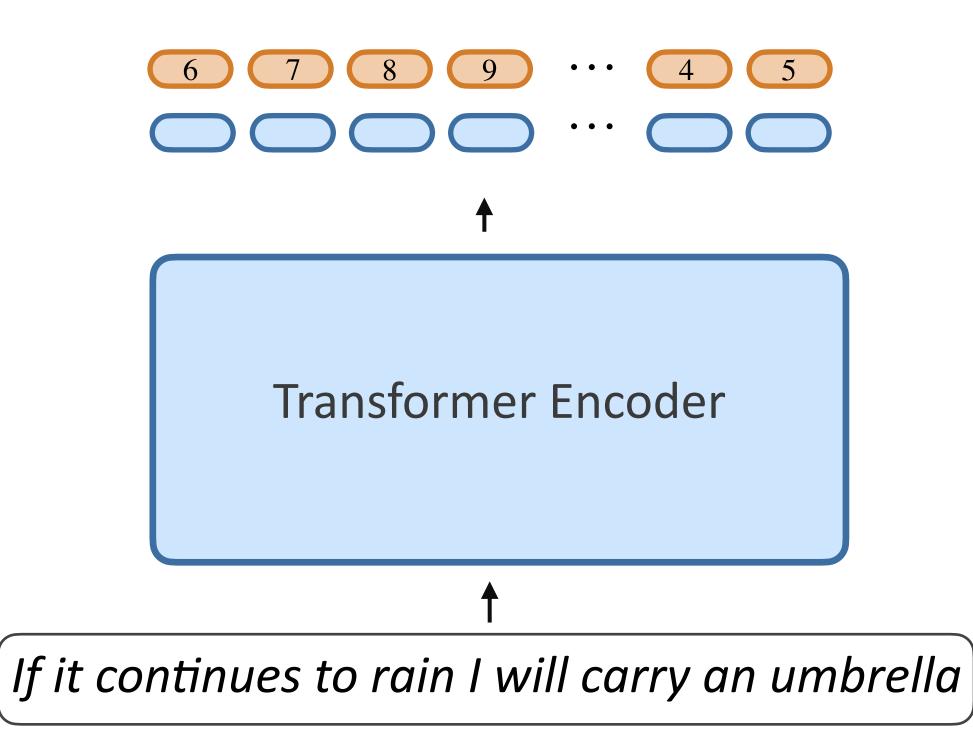


REAP model: Rearrangement Aware Paraphrasing

Goal: Given an input sentence x and source reordering r, generate a paraphrase **y** roughly following the order specified by **r**.





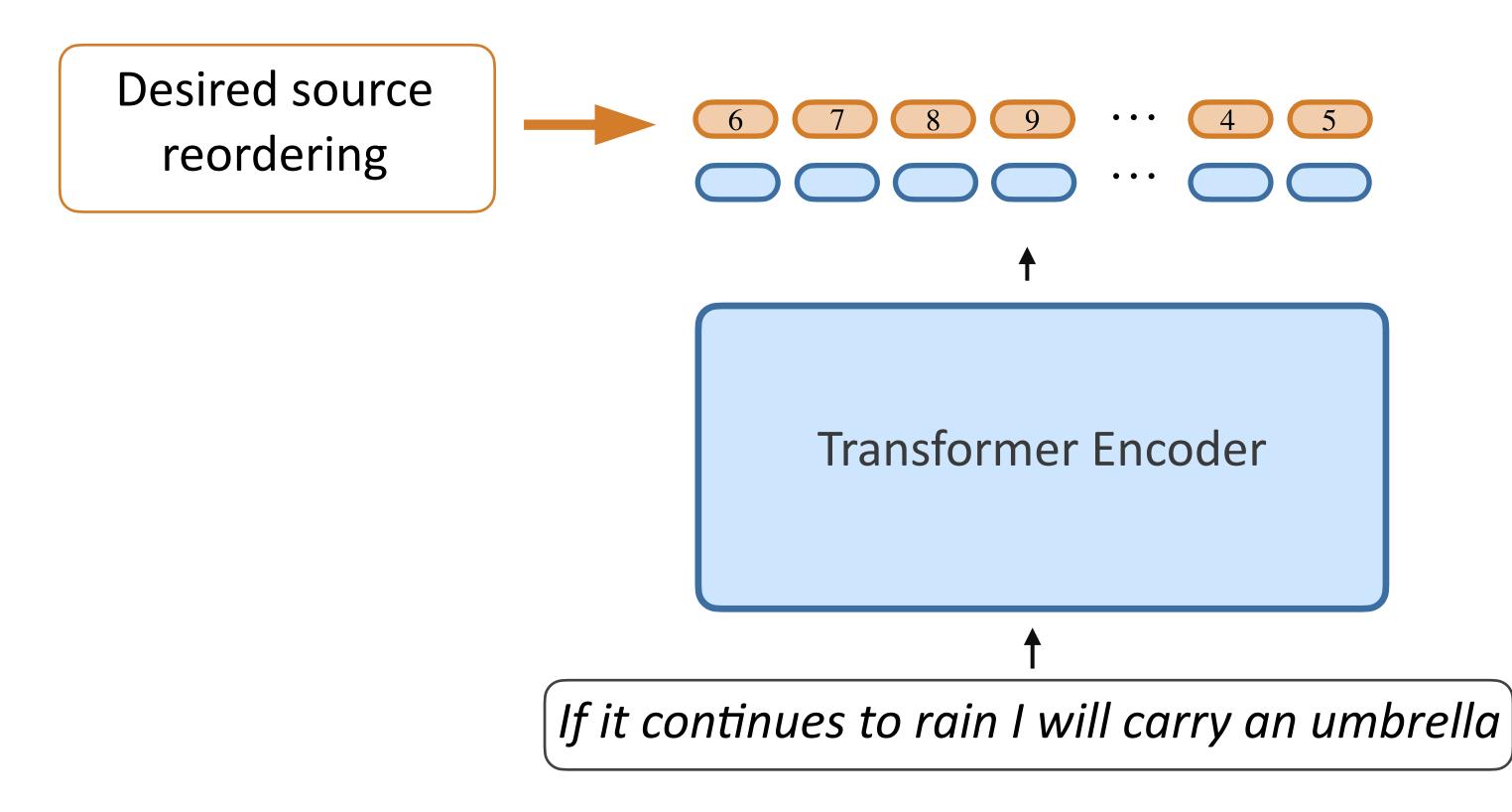


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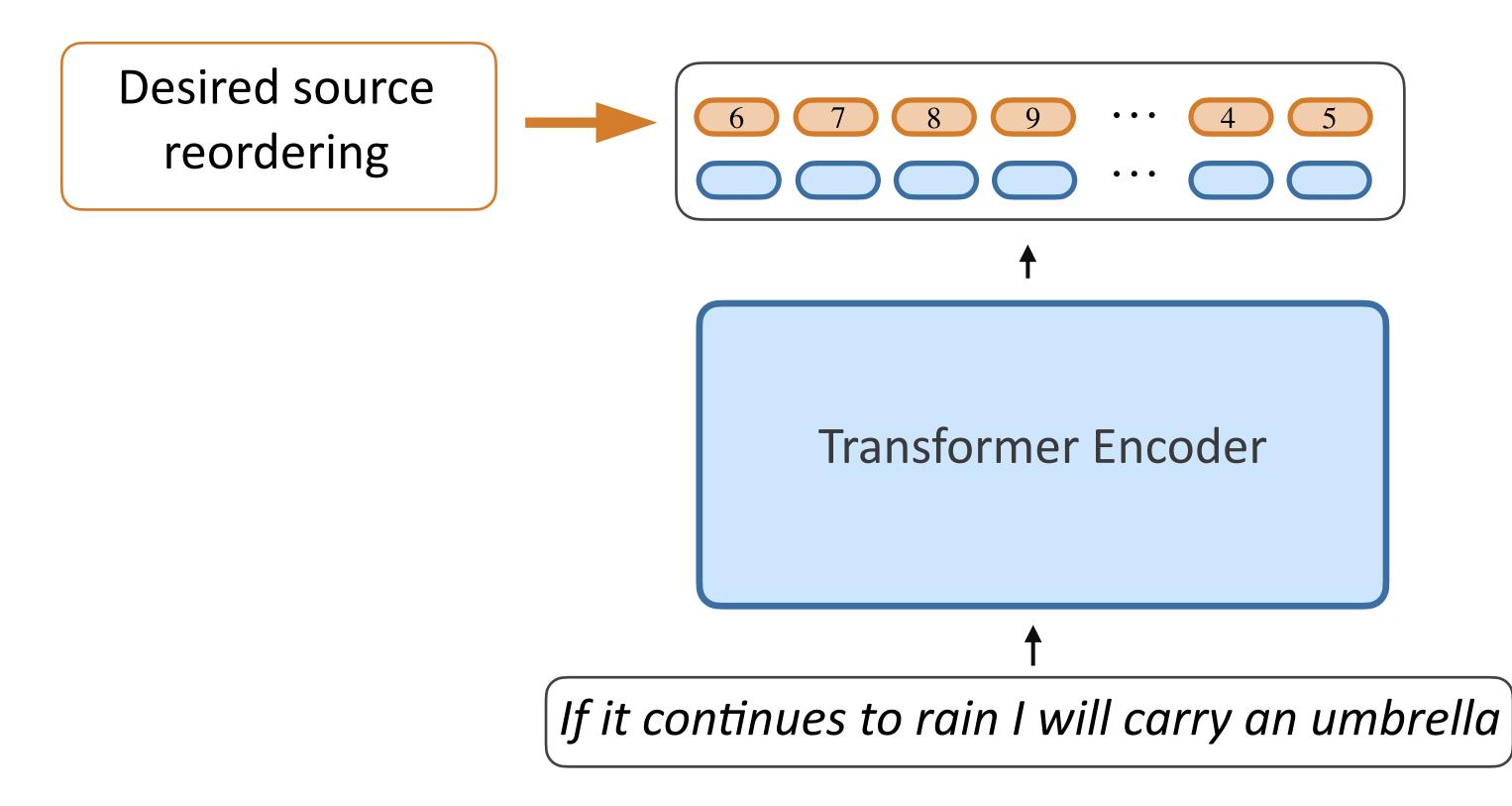






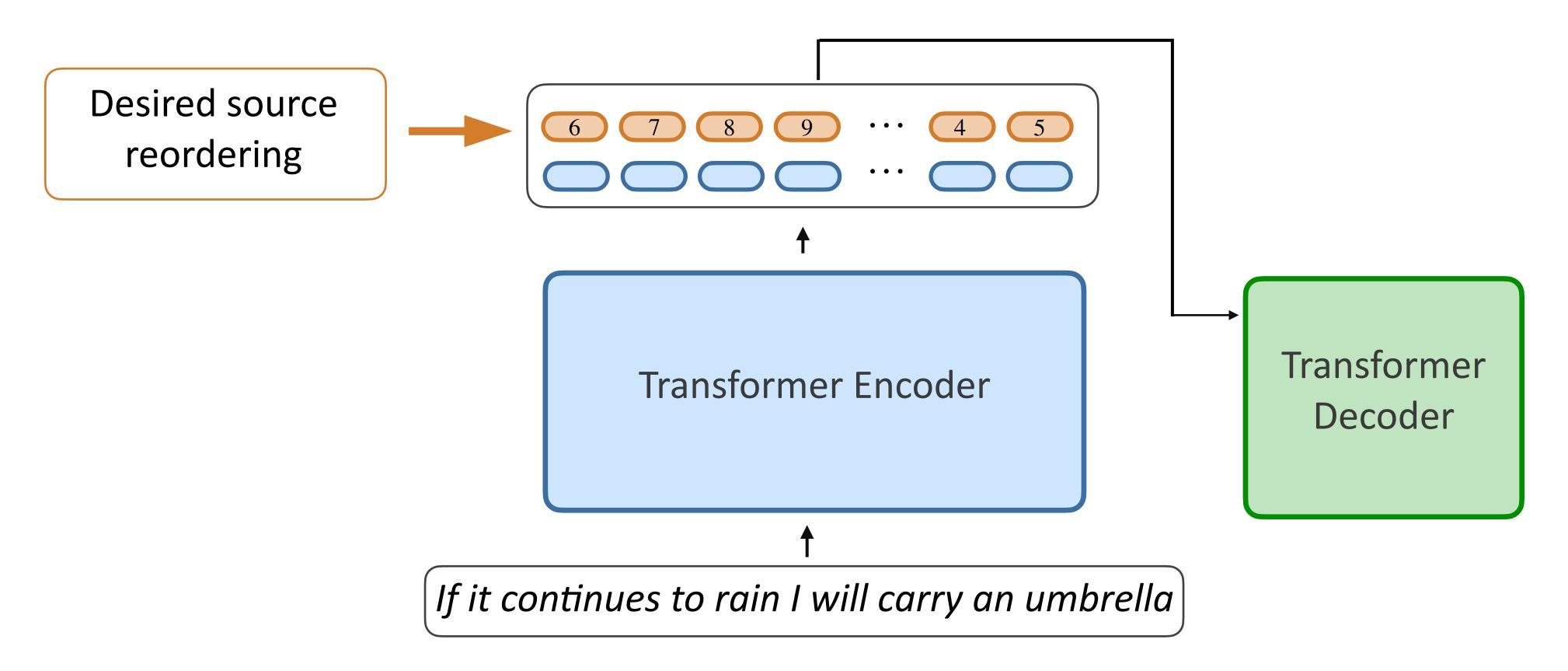






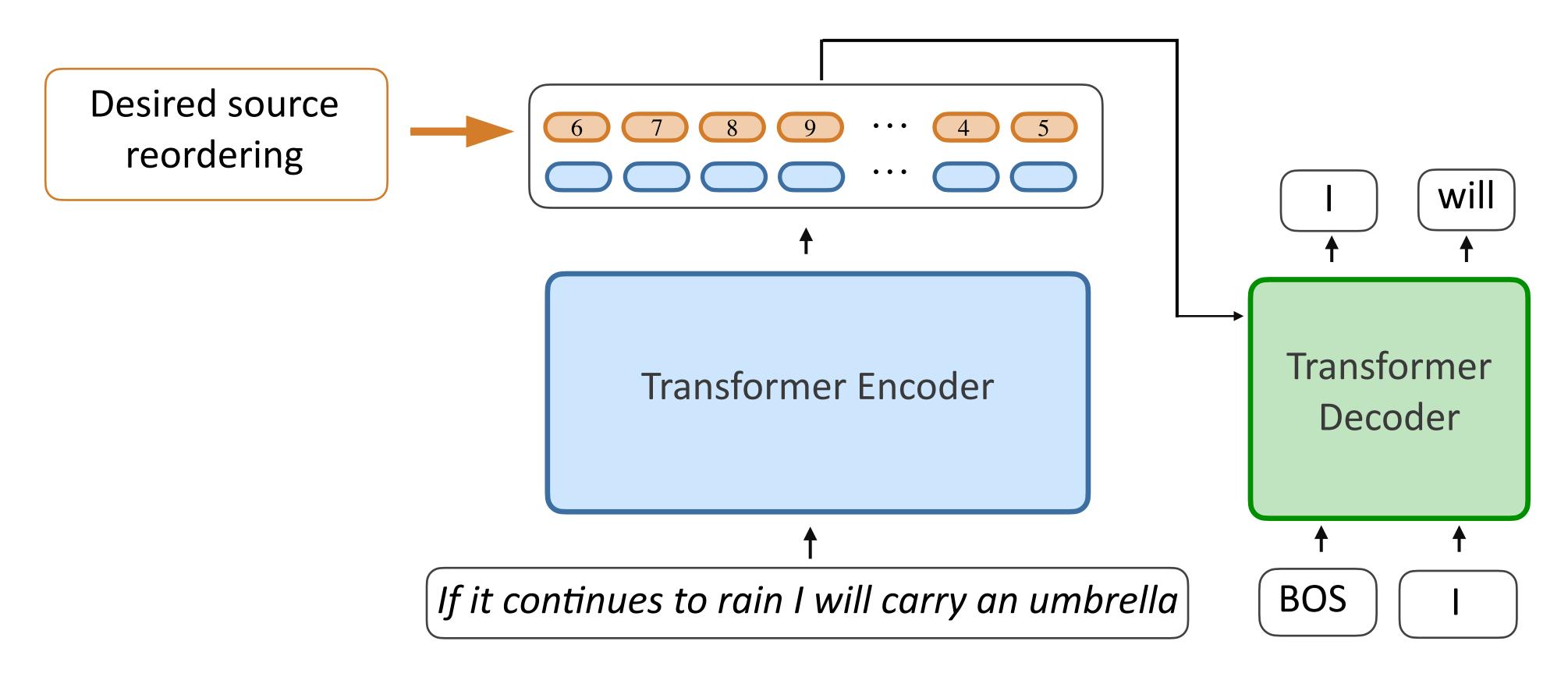
















Clippers won the game —— The game was won by Clippers 3 1 2 4

The sales went up in May —— In May, the sales went up 3 4 5 6 1 2

Training Data: REAP

- We want sentence pairs with reordering information about the source sentence.



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Extract training data from any paraphrase dataset (ParaNMT-50M) using word alignment to gold paraphrases:

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- Clippers won the game
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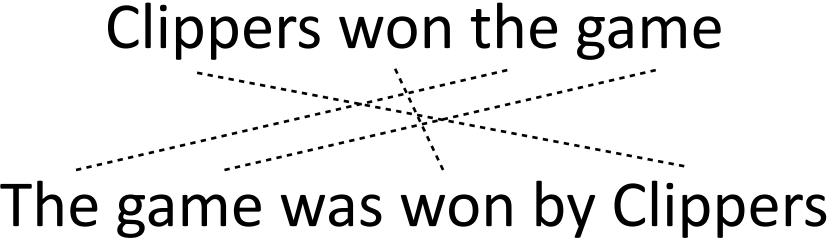


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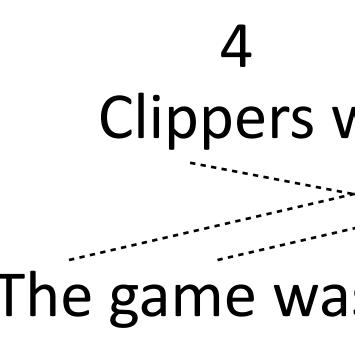
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ParaNMT dataset: English paraphrase pairs constructed using backtranslation.



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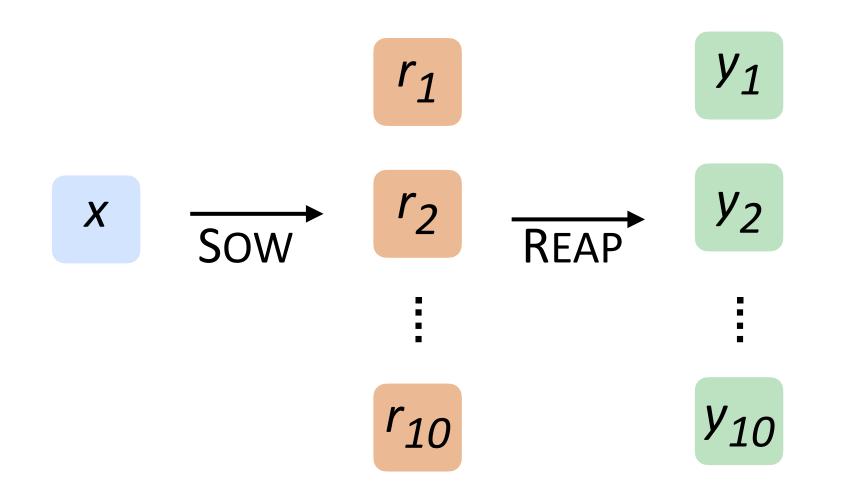
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Testing (10k paraphrase pairs): for each of 10 reorderings, produce a paraphrase.



Despite recognition in European legislation, it has proved difficult to implement the right of consumer associations to representation.





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SOW





Despite recognition PP, it VP. \rightarrow it VP despite recognition PP.

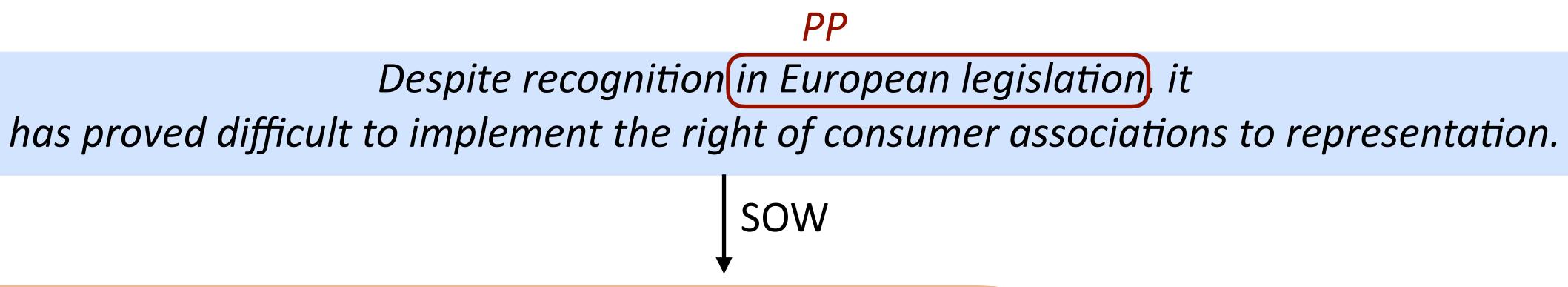
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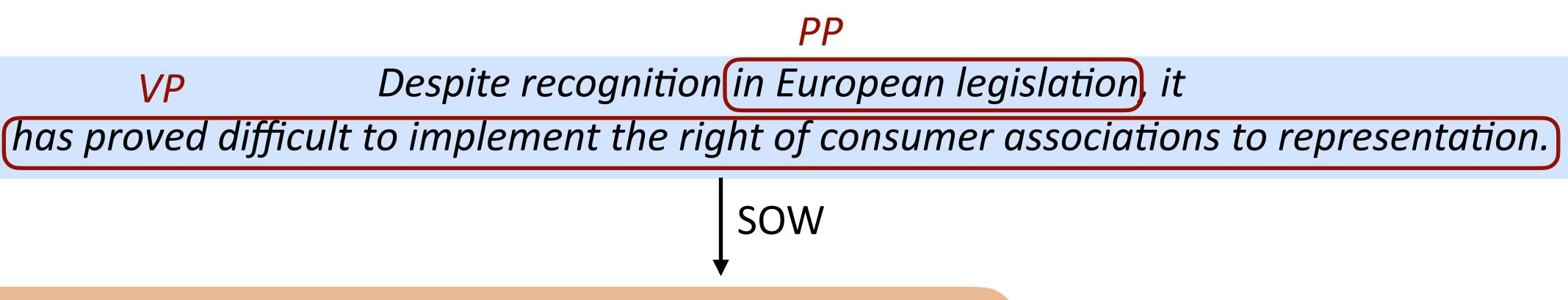






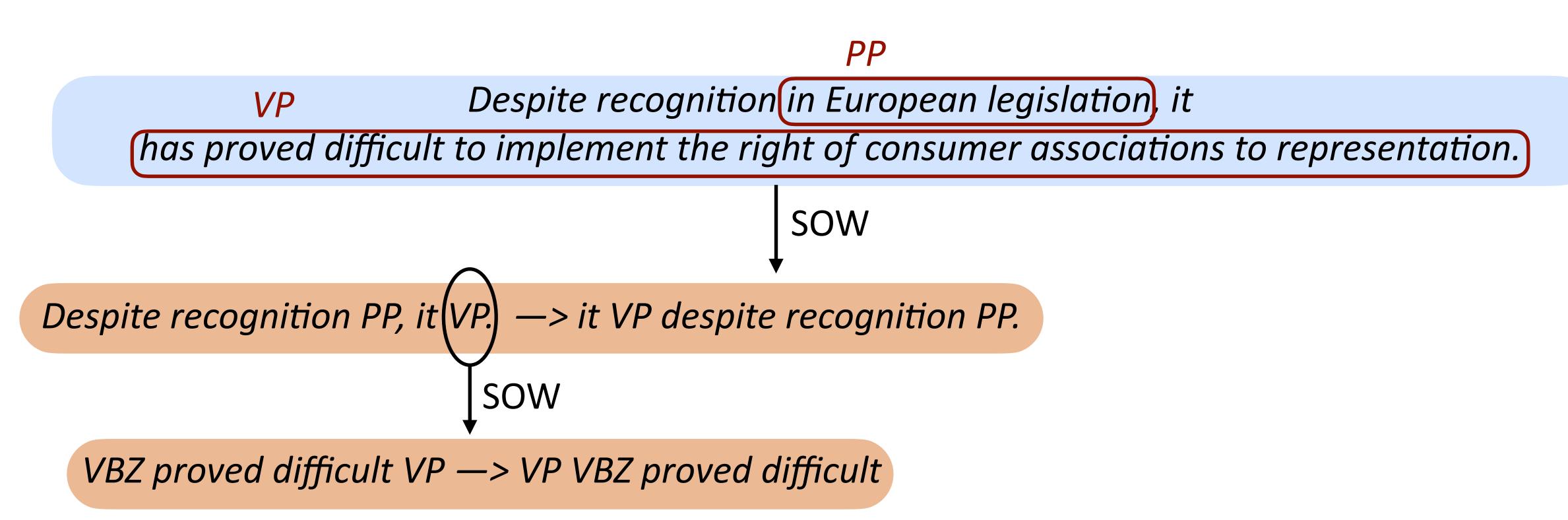
VP

Despite recognition PP, it VP. \rightarrow it VP despite recognition PP.



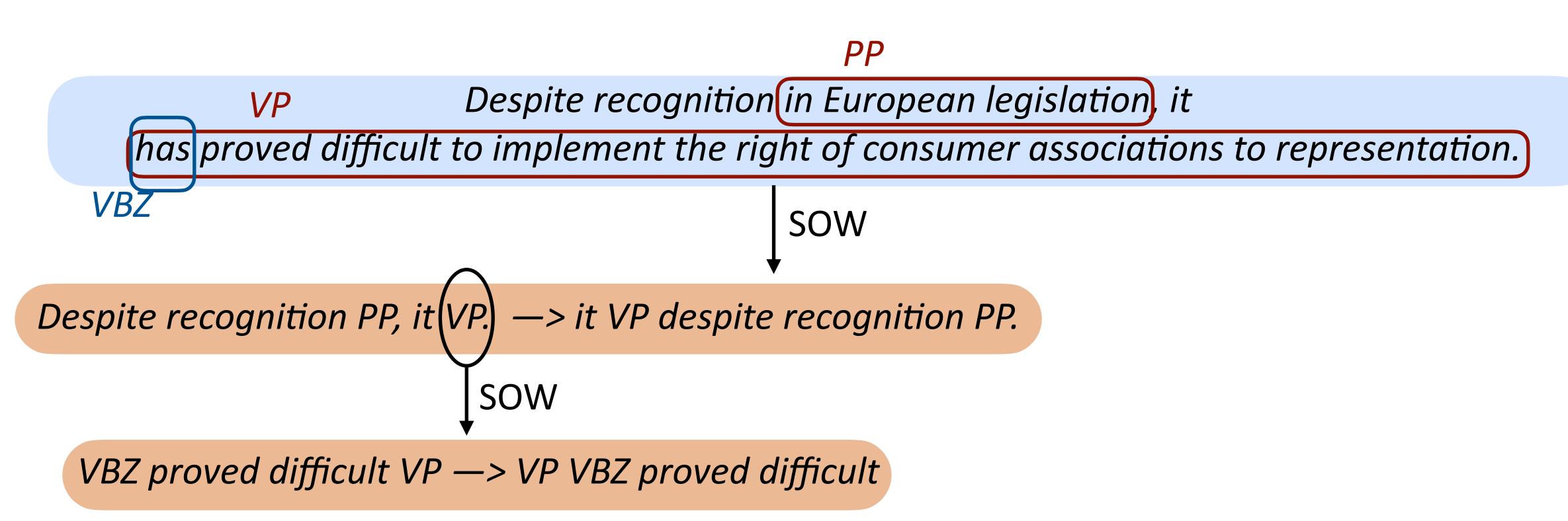






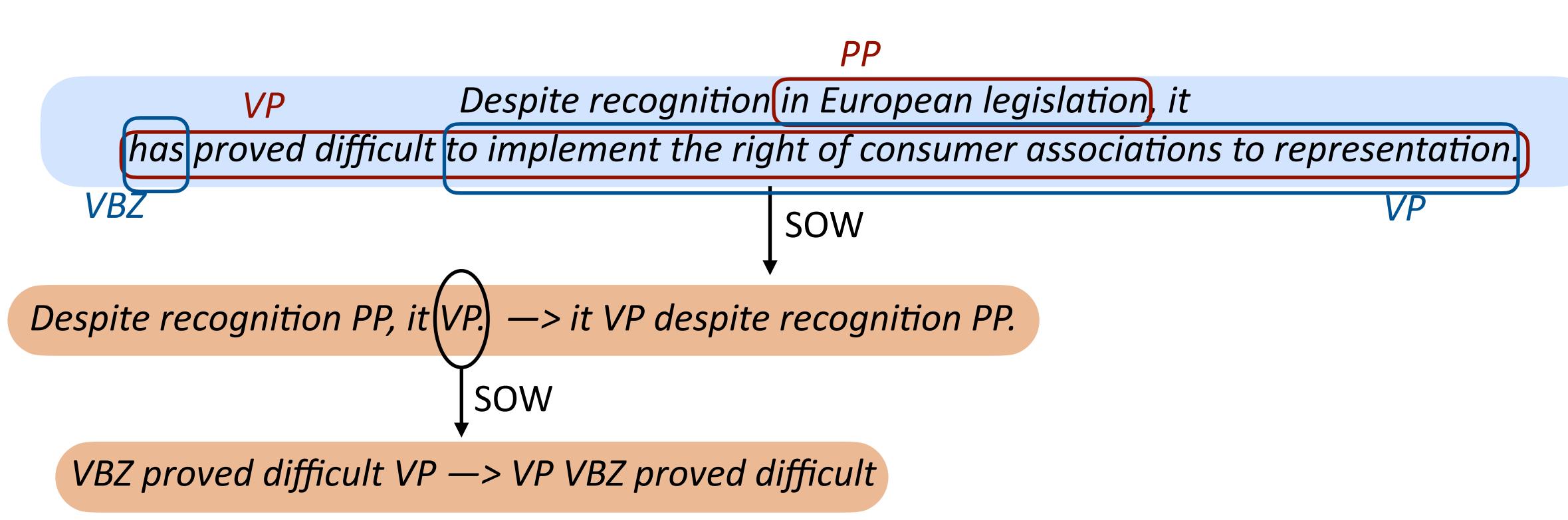






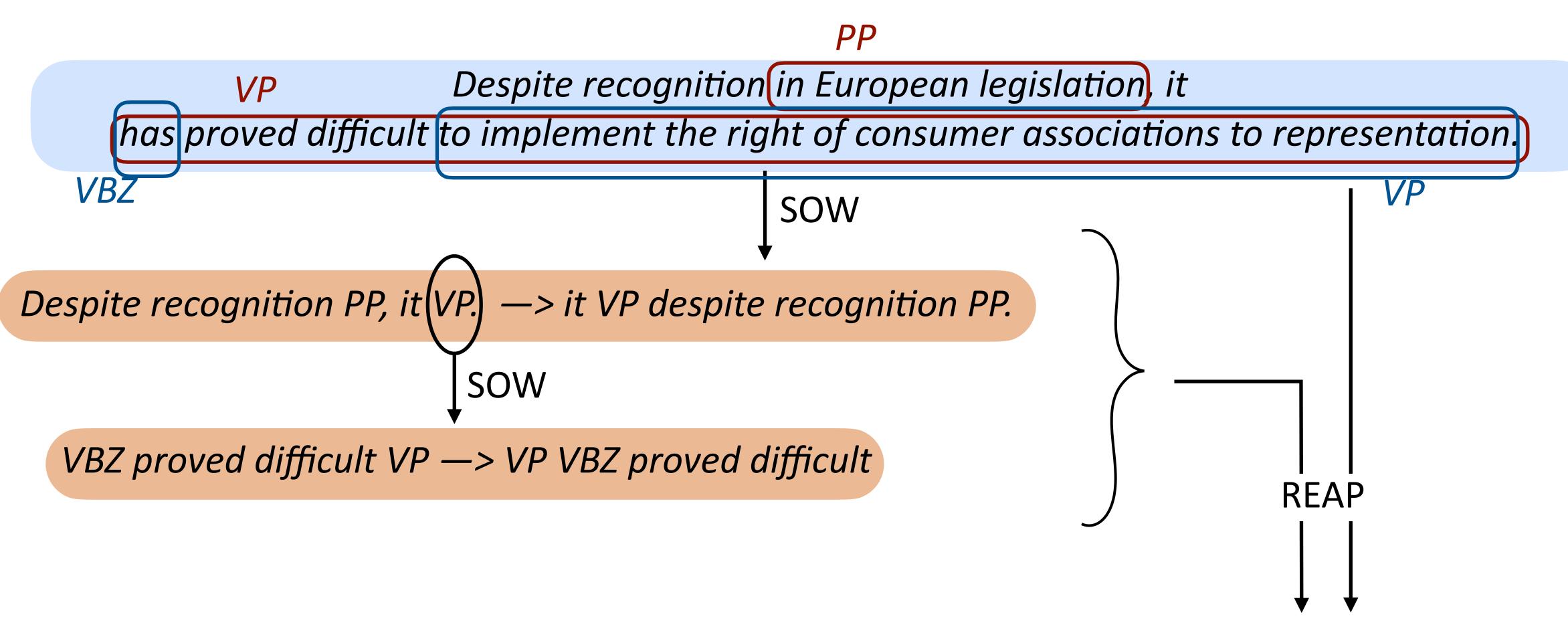






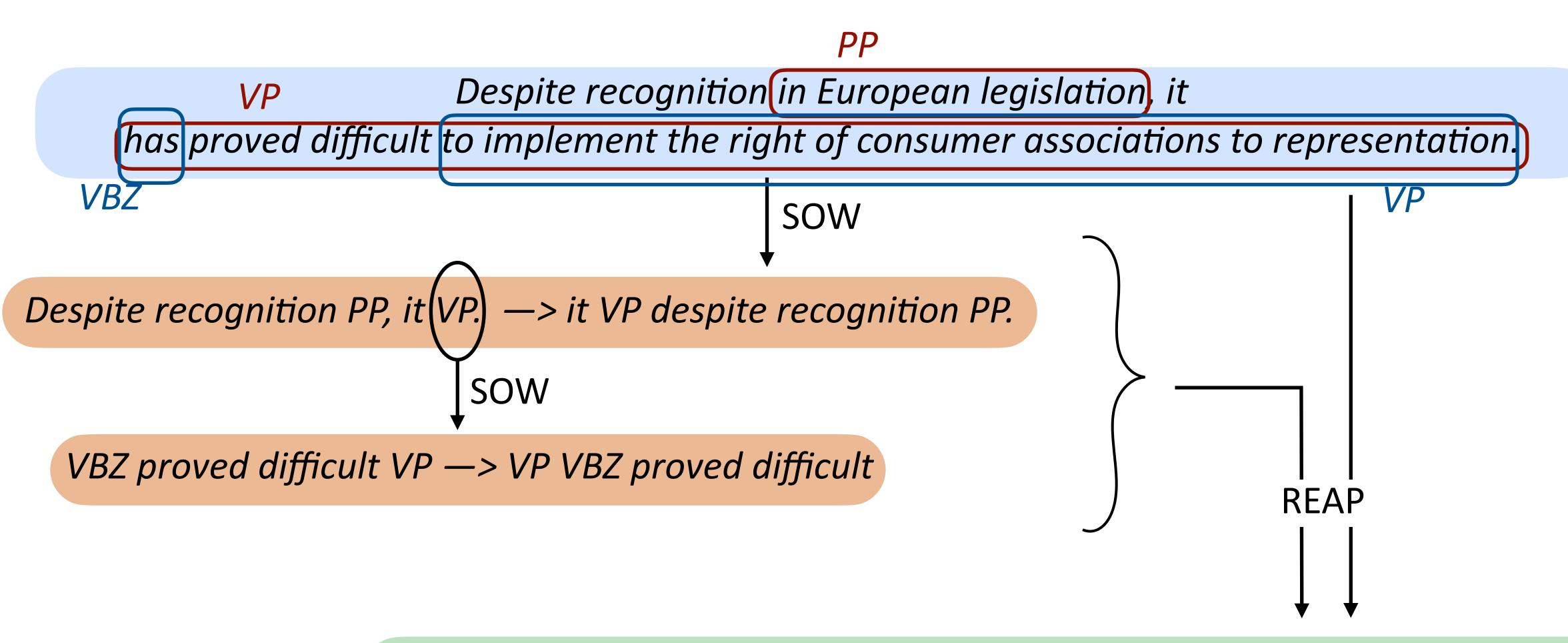






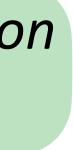






The implementation of the right of consumer associations to representation has proved difficult despite recognition in European legislation.









Transformer seq2seq

Top-k decoding

Diverse-decoding (Kumar et al. NAACL'19)



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SCPN

(lyyer et al. NAACL'18)



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SOW-REAP



<u>Quality (higher is better)</u>

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<u>Homogeneity (lower is better)</u>





<u>Quality (higher is better)</u>

oracle-BLEU best BLEU score w.r.t. to the target among the 10 paraphrases

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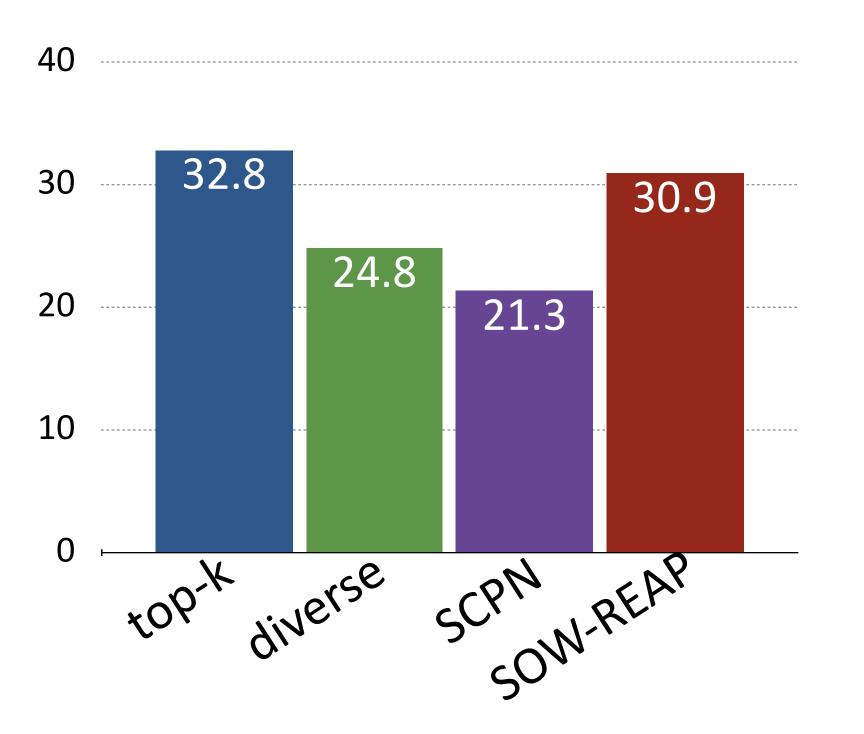
self-BLEU pair-wise BLEU between generated paraphrases for the same input





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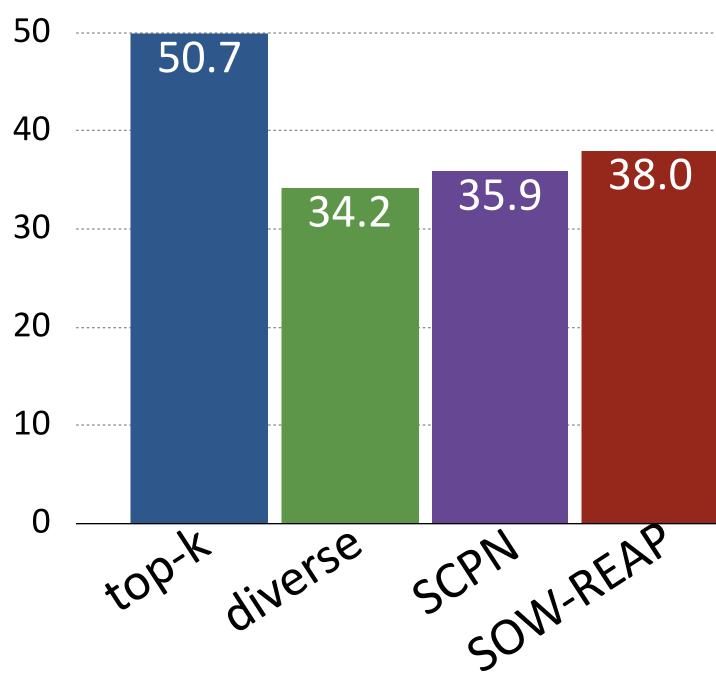
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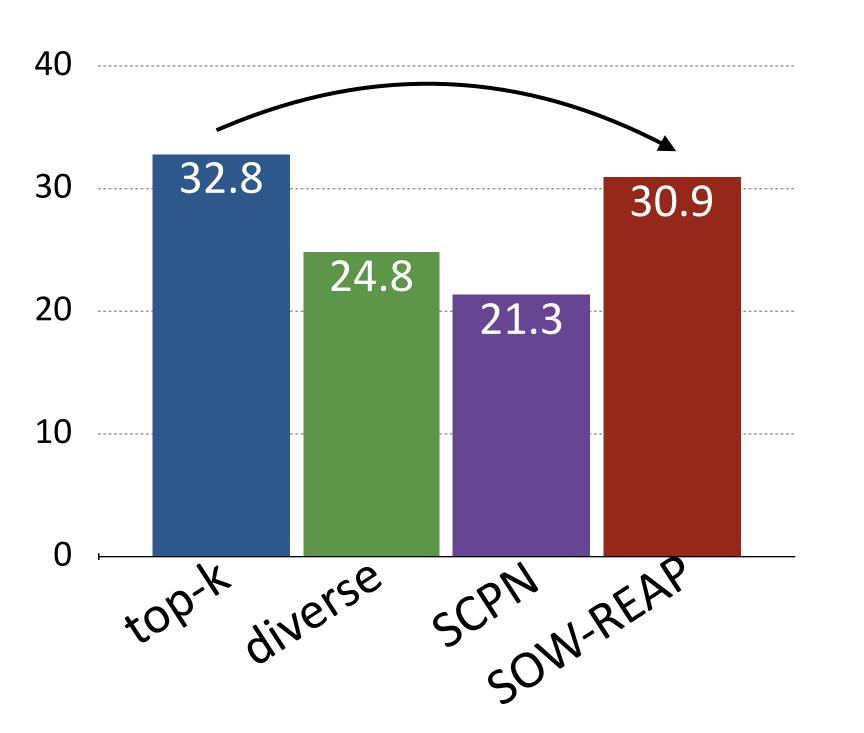
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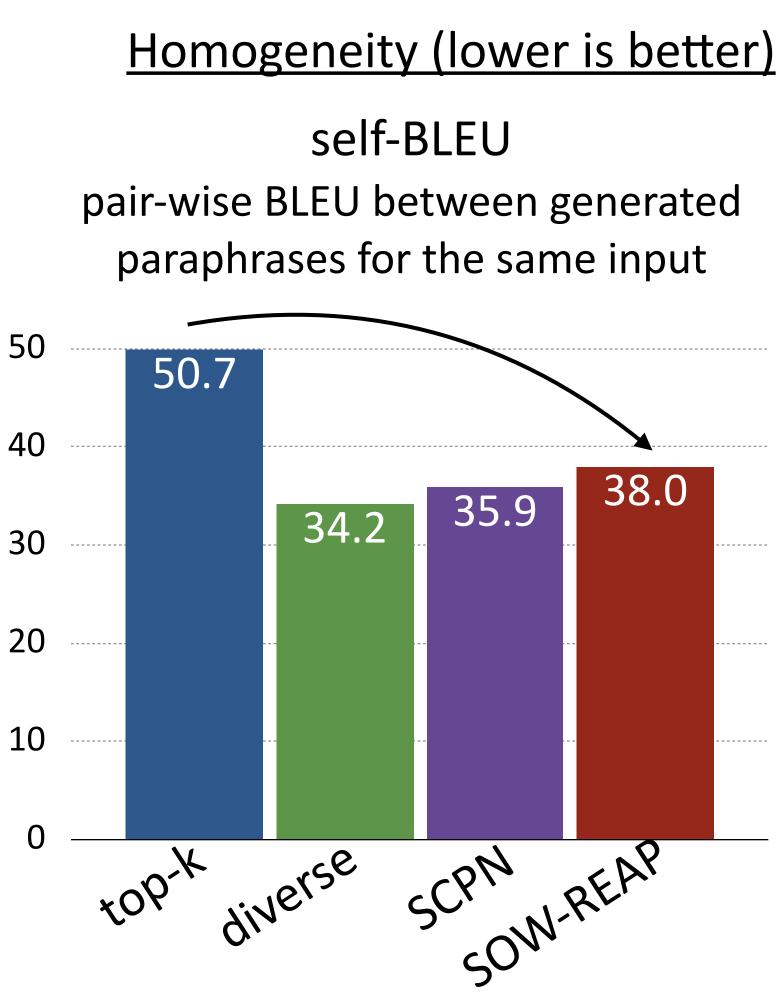
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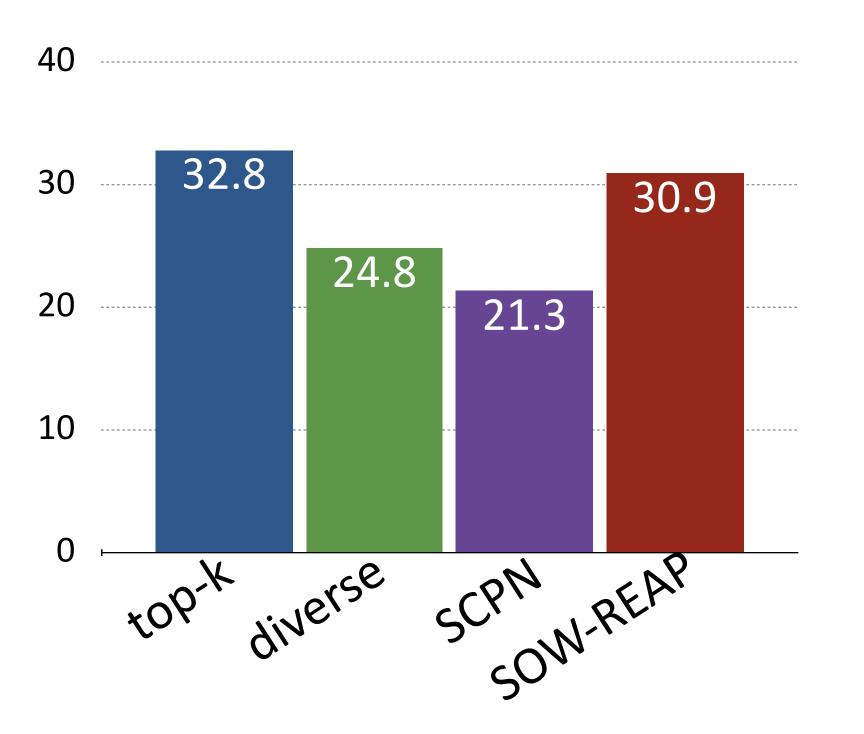
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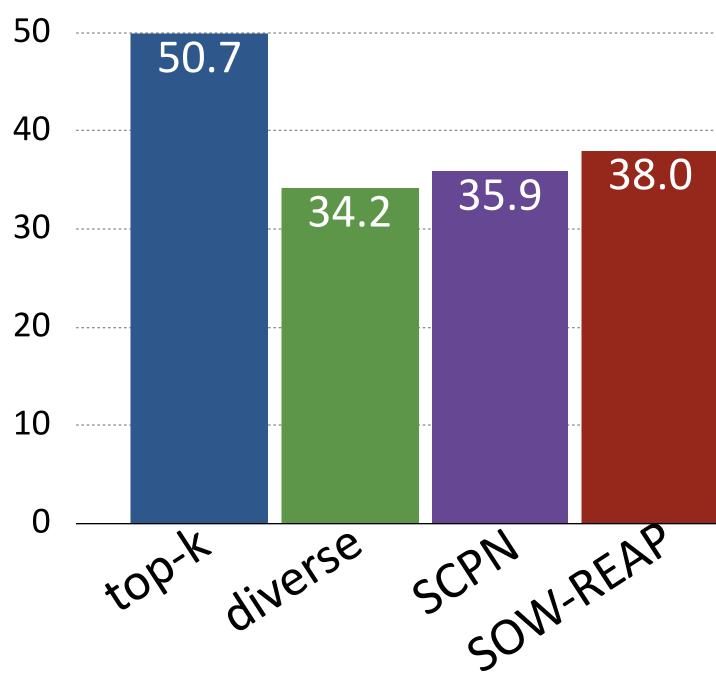
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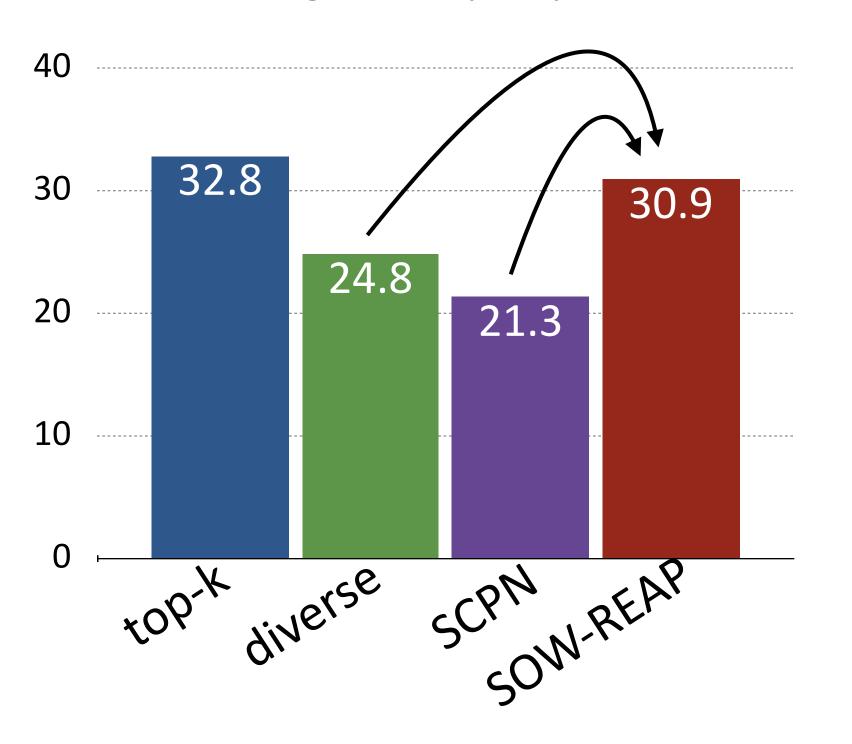
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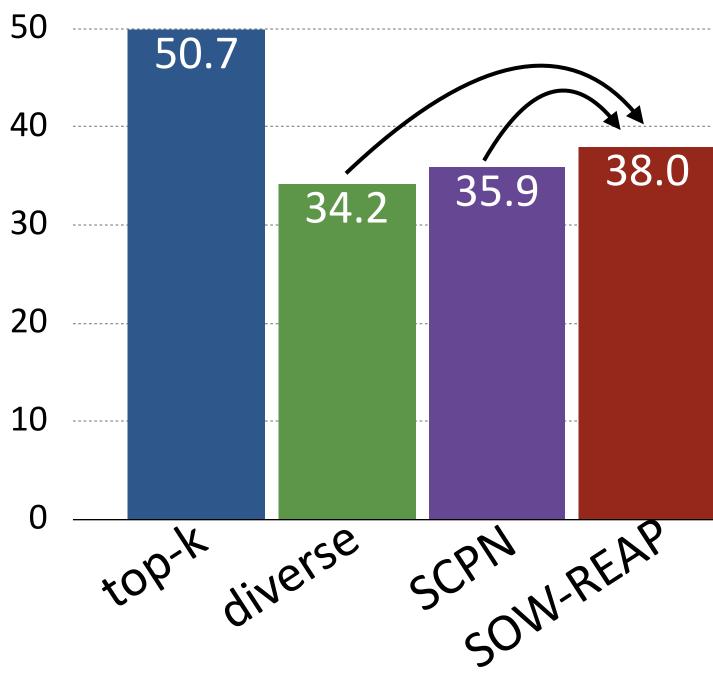
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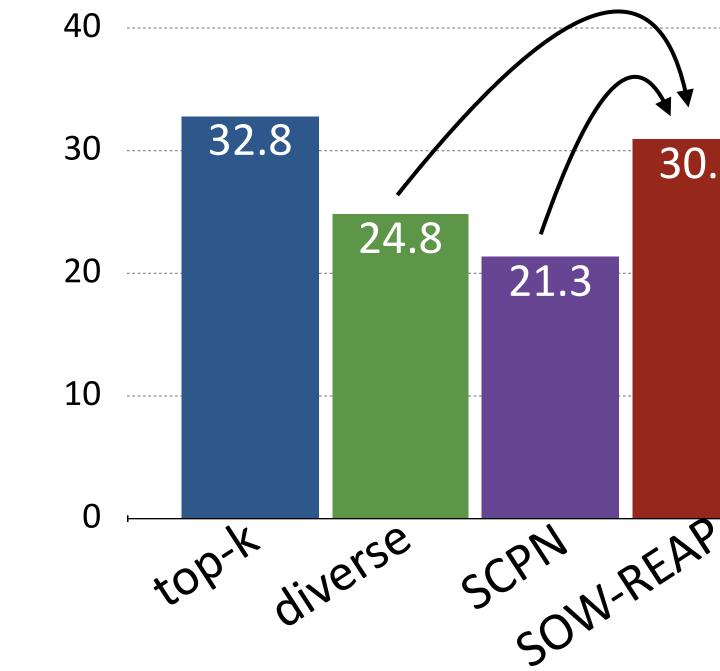
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SOW-REAP achieves a better quality-diversity tradeoff compared to baselines!

Transformer seq2seq

SCPN

SOW-REAP

Top-k decoding

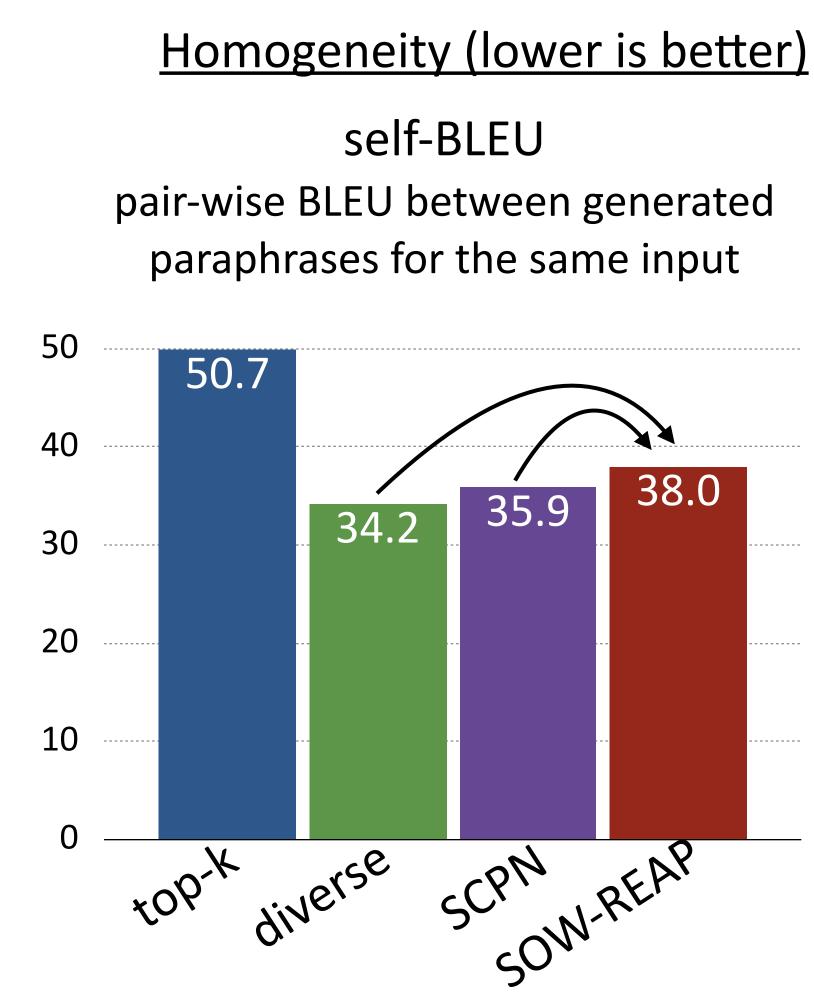
Diverse-decoding

(lyyer et al. NAACL'18)

(Kumar et al. NAACL'19)

Comparison with prior work

30.9





Human Evaluation



Human Evaluation

Rate 100 sampled paraphrases from each model on a 3-point quality scale with MTurk

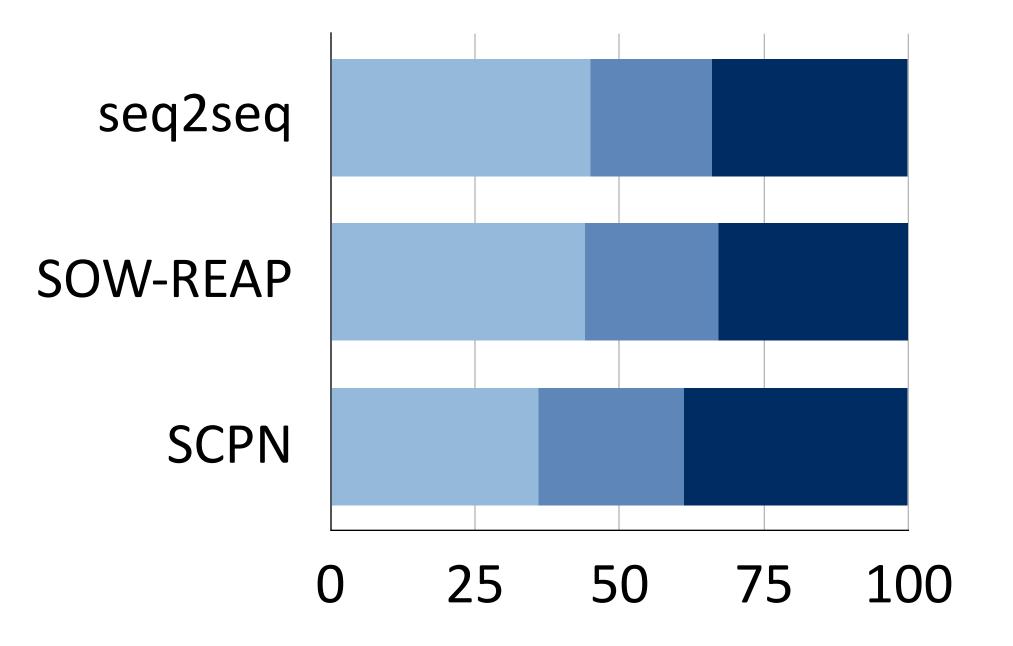




Grammatical Paraphrase Ungrammatical paraphrase Not a paraphrase

Human Evaluation

Rate 100 sampled paraphrases from each model on a 3-point quality scale with MTurk





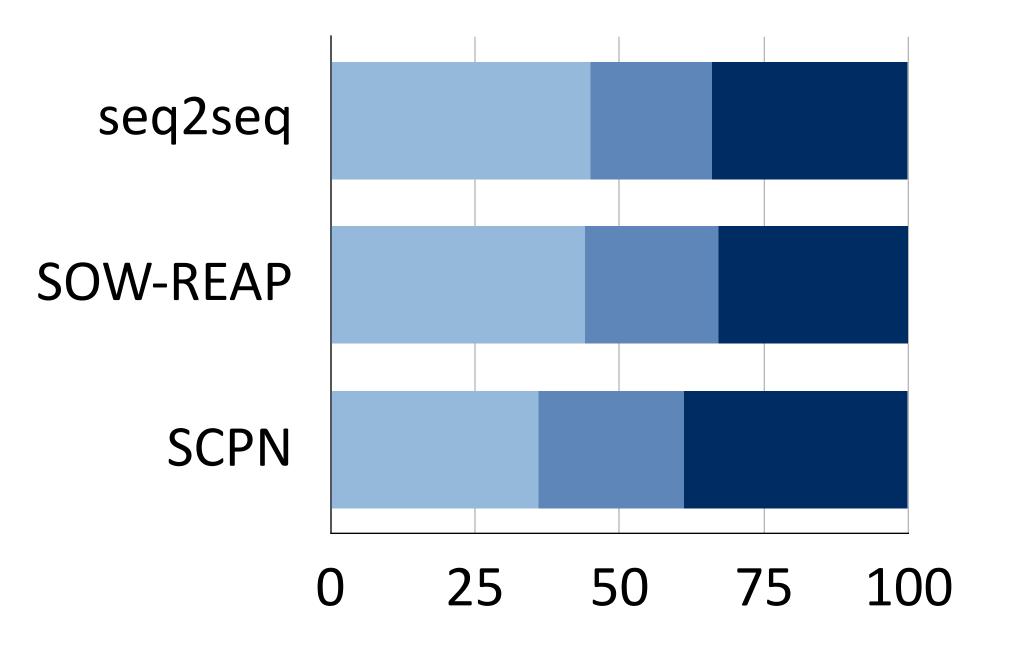


Grammatical Paraphrase Ungrammatical paraphrase Not a paraphrase

Quality of ours is **as high as** the basic seq2seq model

Human Evaluation

Rate 100 sampled paraphrases from each model on a 3-point quality scale with MTurk







Takeaways





paraphrasing phenomenon.

Takeaways

Learning syntactic transformations provide a flexible way to enumerate different



- paraphrasing phenomenon.
- generations.



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Neural models can be trained to follow these syntactic guides and produce diverse



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Thanks!