

Efficient Parsing for Transducer Grammars



John DeNero, Mohit Bansal, Adam Pauls, and Dan Klein

Overview

- Syntactic decoding can decompose into two phases: parsing and language model integration
- The parsing phase alone is time consuming for large tree transducer grammars
- Grammar transformations, optimizations, and coarse-to-fine techniques increase parsing speed
- The accelerated parsing pass improves translation

Tree Transducer Grammars

No se olvide de subir un canto rodado en Colorado

Synchronous Grammar

Output

Tree Transducer Grammars

No se olvide de subir un canto rodado en Colorado

Synchronous Grammar

NNP → Colorado ; *Colorado*

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NNP → Colorado ; *Colorado*

NN → canto rodado ; *boulder*

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S → No se olvide de subir un ***NN*** en ***NNP*** ; *Don't forget to climb a ***NN*** in ***NNP****

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Tree Transducer Grammars

No se olvide de subir un can

Synchronous G

NNP → Colorado ; *Colorado*

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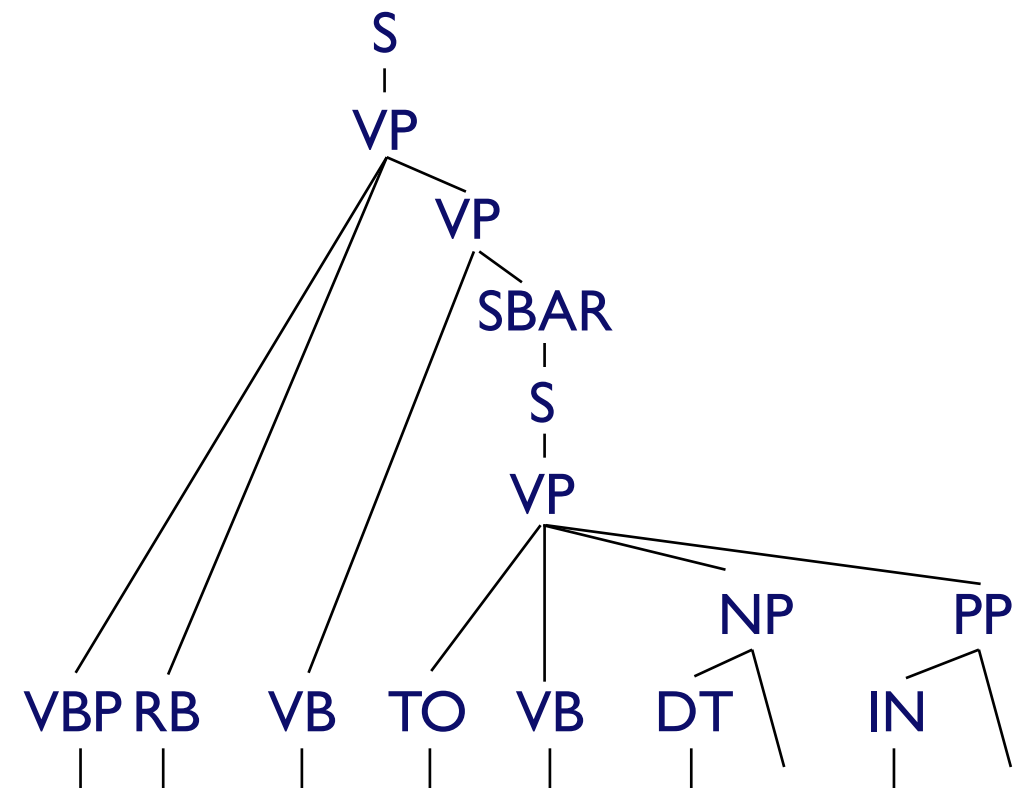
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NNP → Colorado ; *Colorado*

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Tree Transducer Grammars

No se olvide de subir un $\overline{\text{NN}}$ canto rodado en Colorado

Synchronous Grammar

NNP \rightarrow Colorado ; *Colorado*

NN \rightarrow canto rodado ; *boulder*

S \rightarrow No se olvide de subir un ***NN*** en ***NNP*** ; *Don't forget to climb a ***NN*** in ***NNP****

Output

$\overline{\text{NN}}$

boulder

Tree Transducer Grammars

No se olvide de subir un $\overline{\text{NN}}$ canto rodado en $\overline{\text{NNP}}$ Colorado

Synchronous Grammar

$\text{NNP} \rightarrow \text{Colorado} \quad ; \quad \text{Colorado}$

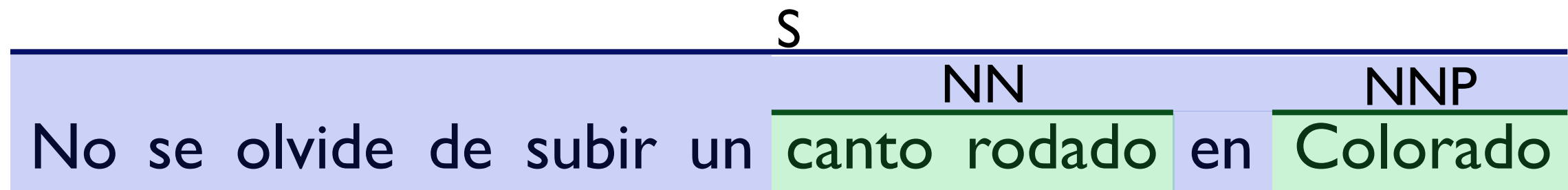
$\text{NN} \rightarrow \text{canto rodado} \quad ; \quad \text{boulder}$

$S \rightarrow \text{No se olvide de subir un } \text{NN} \text{ en } \text{NNP} \quad ; \quad \text{Don't forget to climb a } \text{NN} \text{ in } \text{NNP}$

Output

$\overline{\text{NN}}$ boulder $\overline{\text{NNP}}$ Colorado

Tree Transducer Grammars



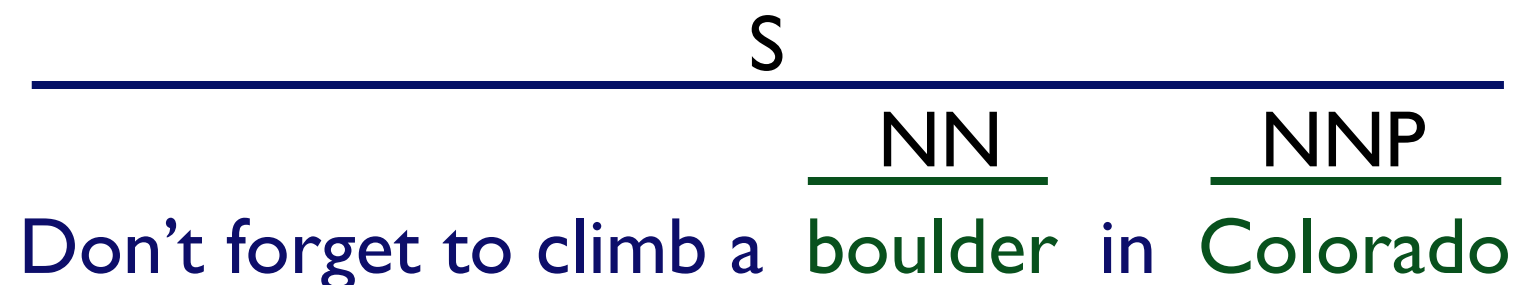
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Multi-Pass Syntactic Decoding

Parse input sentence
with source-side
grammar projection

Rerank derivations
rooted at each parse state
using a language model

Cube growing [Huang and Chiang '07]:
Lazy forest reranking algorithm

Two-pass SCFG decoding [Venugopal et al '07]:
Local search over derivations in a parse forest

Coarse-to-fine LMs [Zhang et al '08, Petrov et al '08]:
Multi-pass bottom-up LM integration over forests

The Size of Tree Transducer Grammars

Extracted a transducer grammar from a 220 million word bitext

Relativized the grammar to each test sentence

Kept all rules with at most 6 non-terminals

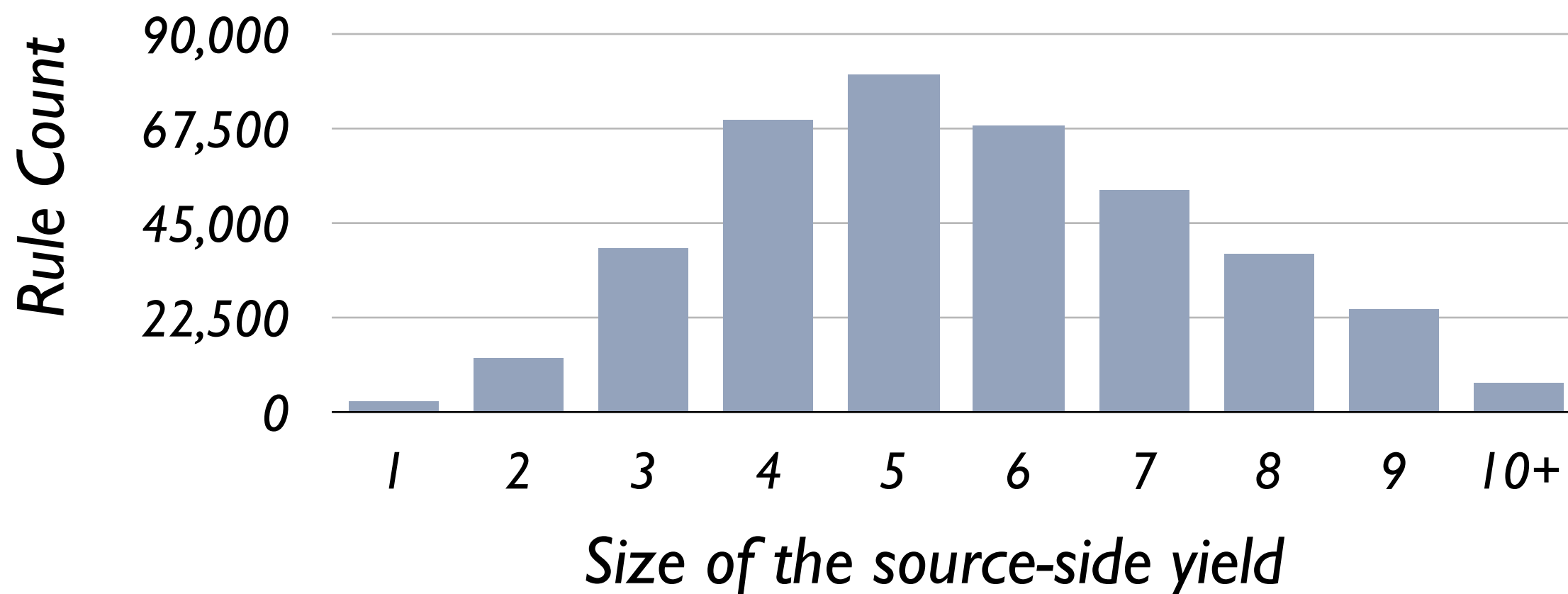
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Rules matching an example 40-word sentence



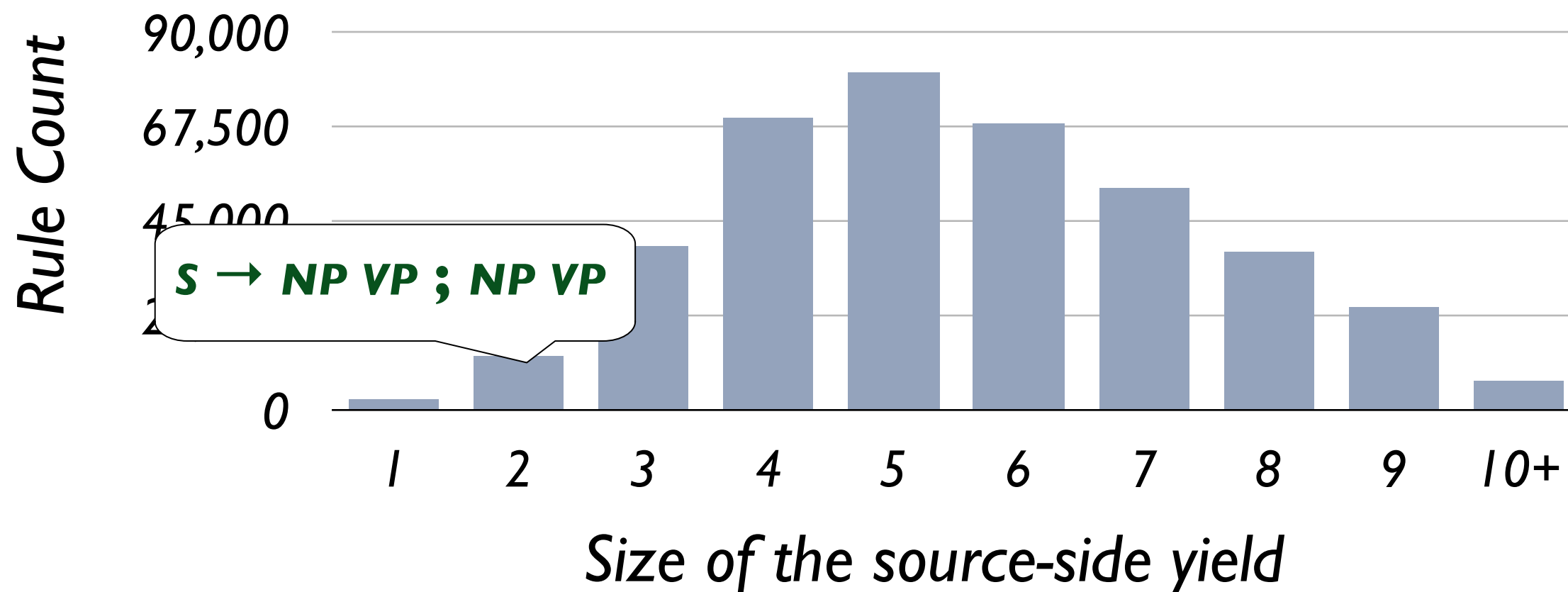
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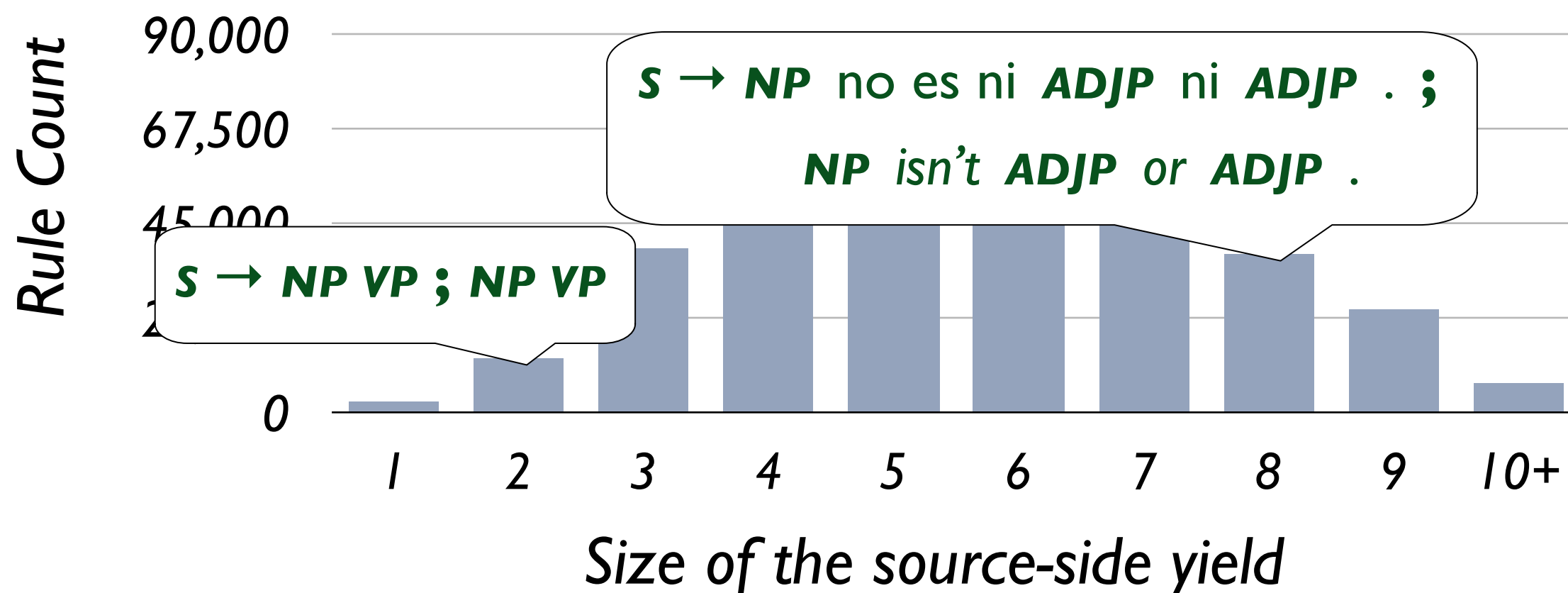
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CKY-style Bottom-up Parsing

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span length:

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Binary rule: $X \rightarrow Y Z$

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Binary rule: $X \rightarrow Y Z$

Split points: $i < k < j$

Operations: $O(j - i)$

Time scales with: Grammar constant

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$S \rightarrow$ No se **VB de subir un **NN** en **NNP****

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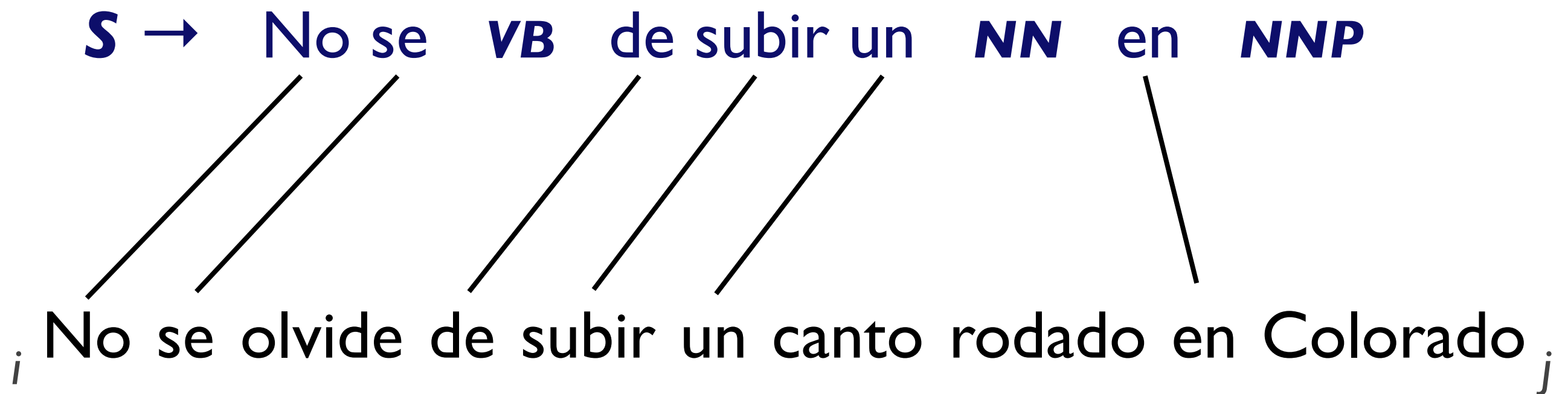
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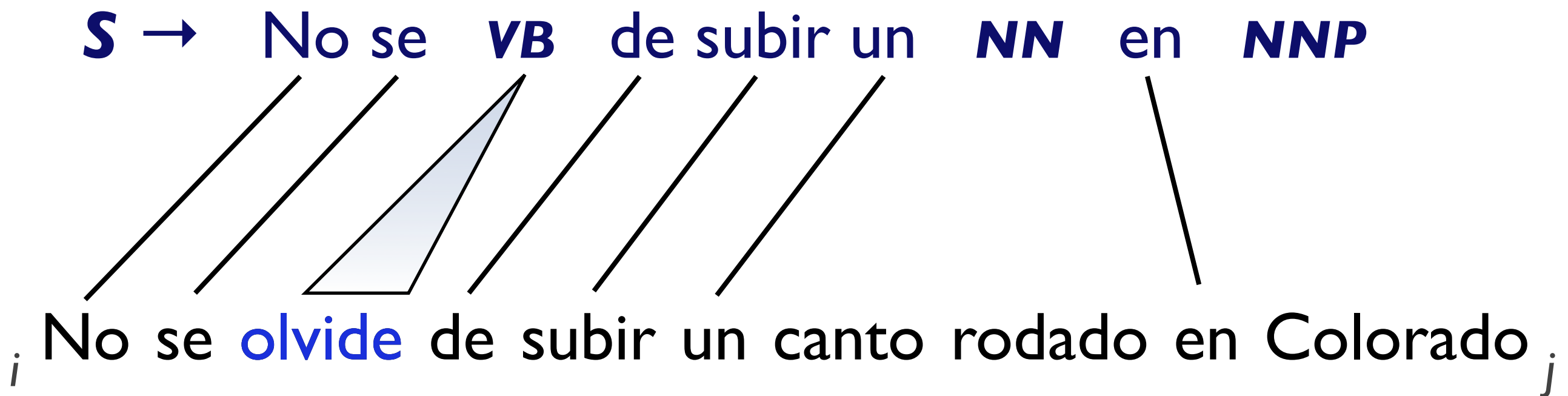


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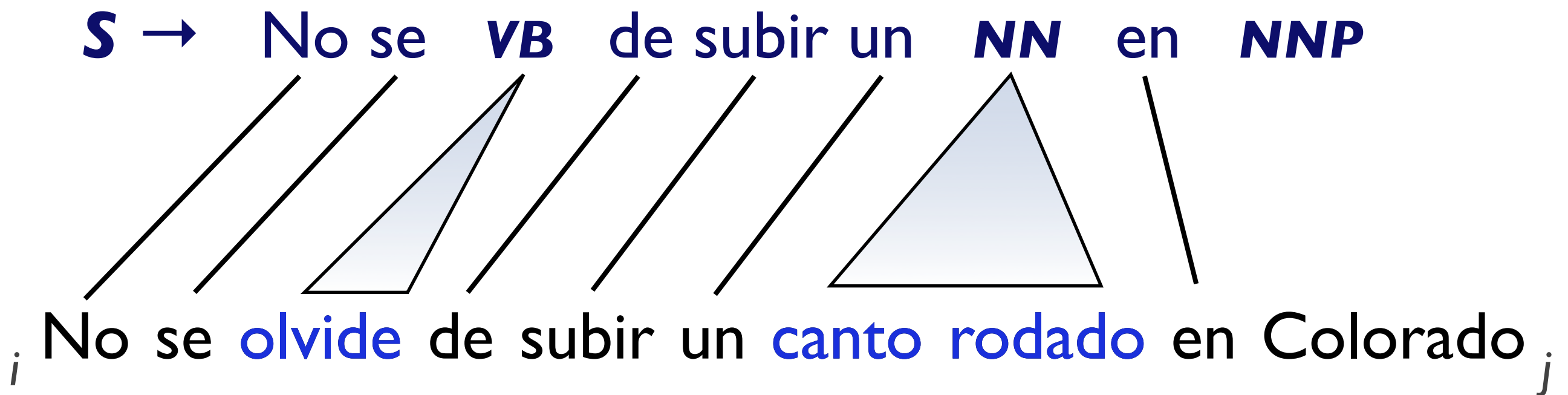


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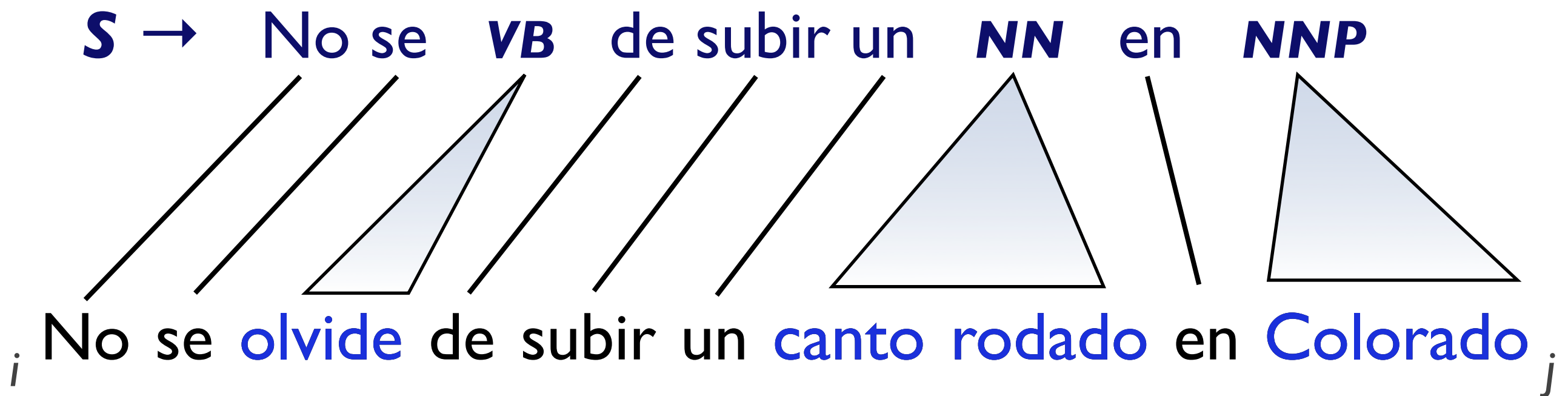


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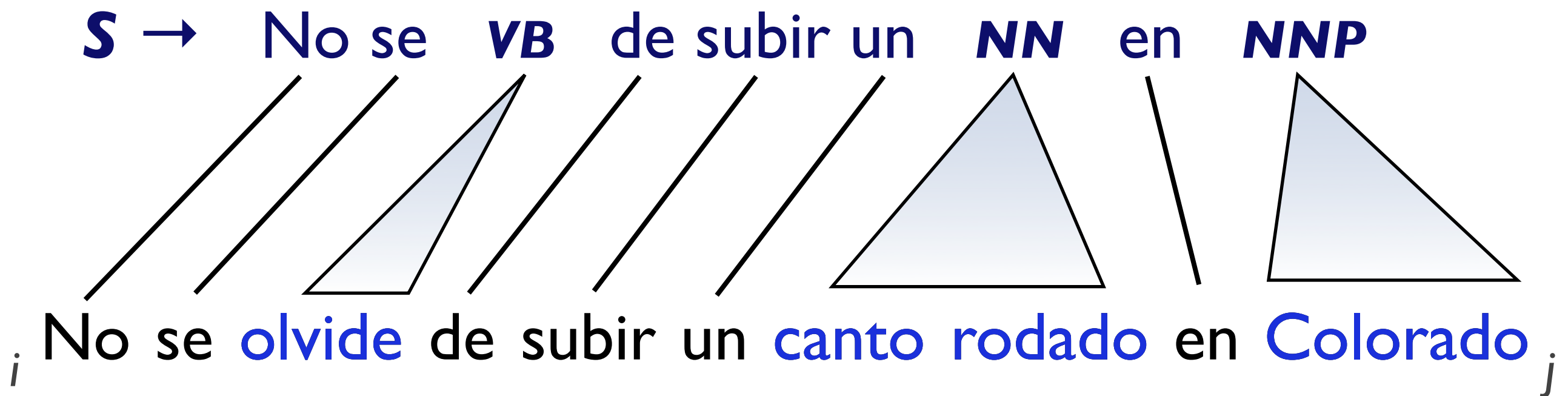


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Many untransformed lexical rules can be applied in linear time

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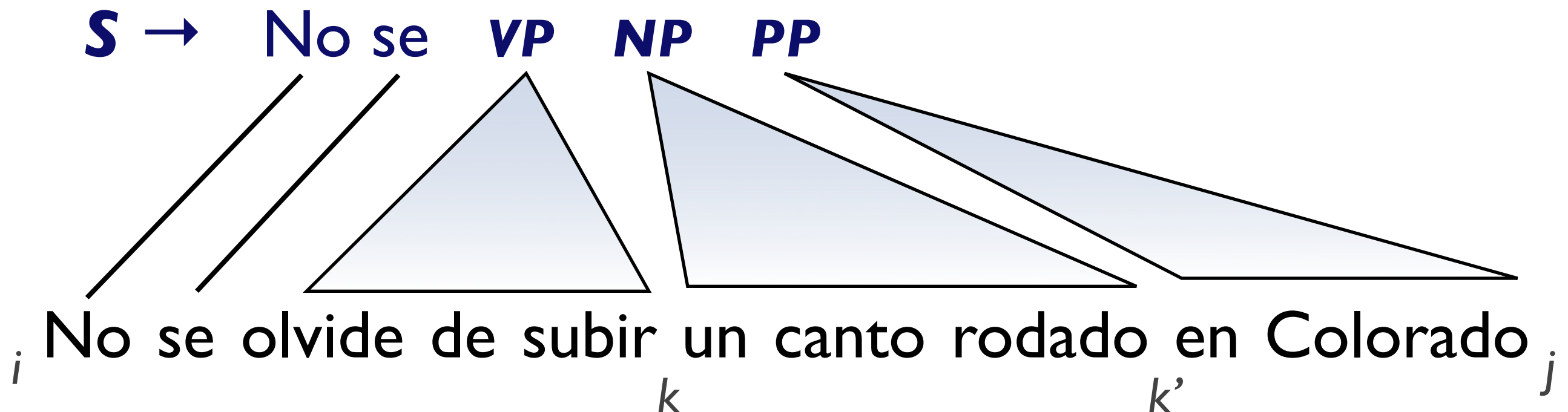
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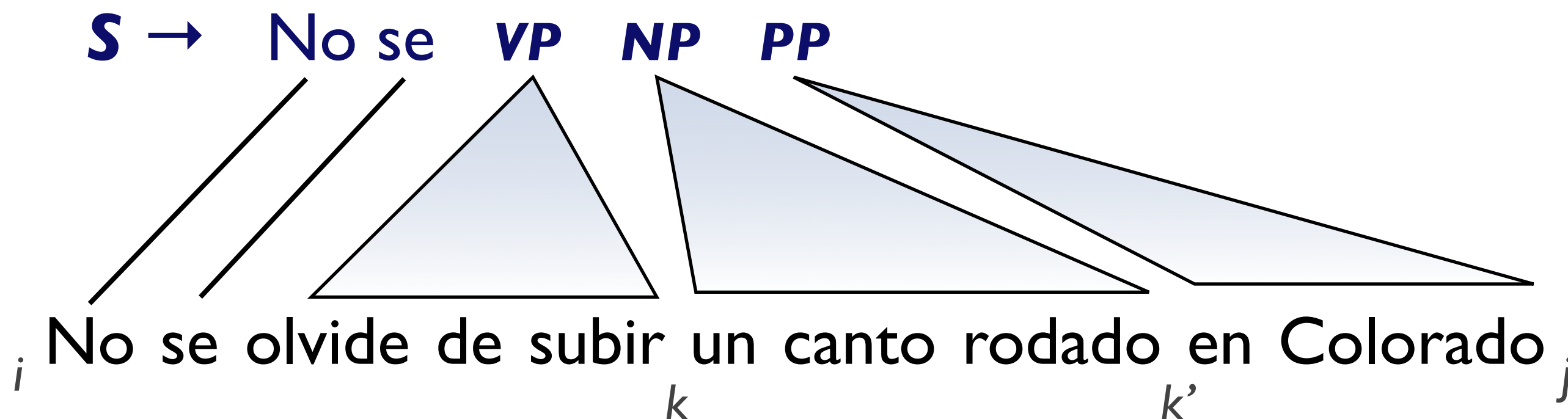


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Problem: Applying adjacent non-terminals is slow

Eliminating Non-terminal Sequences

Lexical Normal Form (LNF)

- (a) lexical rules have at most one adjacent non-terminal*
- (b) all unlexicalized rules are binary.*

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Transformed rules: $S \rightarrow \text{No se } VB \sim VB \text{ } \text{un } NN \sim PP$

$VB \sim VB \rightarrow VB \text{ } VB$

$NN \sim PP \rightarrow NN \text{ } PP$

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- Parsing stages:
- Lexical rules are applied by matching
 - Unlexicalized rules are applied by iterating over split points

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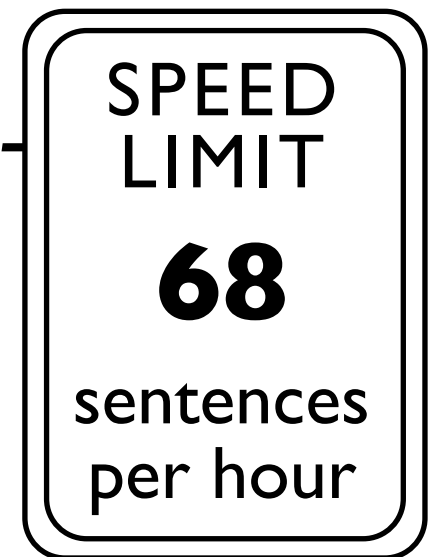
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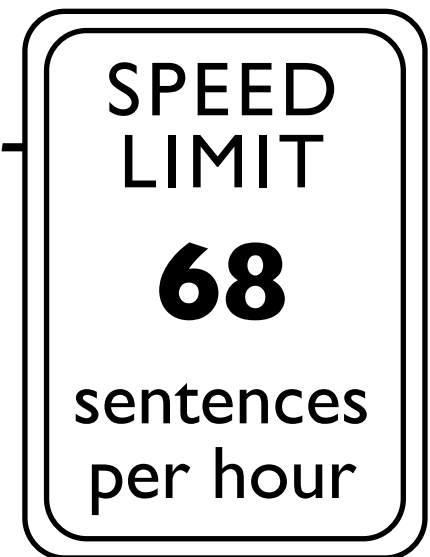
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$\text{VB} \sim \text{VB} \rightarrow \text{VB VB}$

$\text{NN} \sim \text{PP} \rightarrow \text{NN PP}$

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Bounding Split Points

Problem: *Applying binary rules performs wasted work*

$VP \sim VP \rightarrow \text{VP} \text{ VP}$

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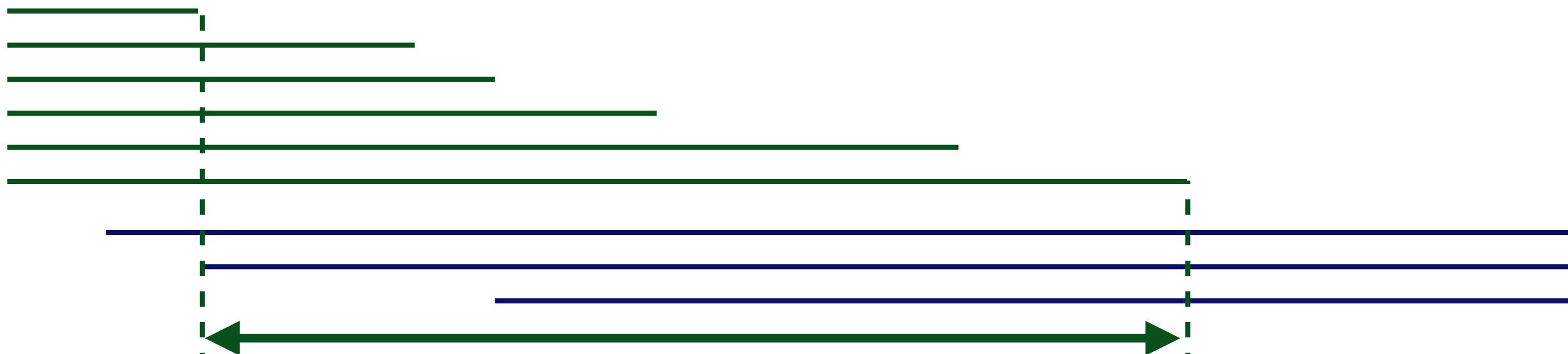
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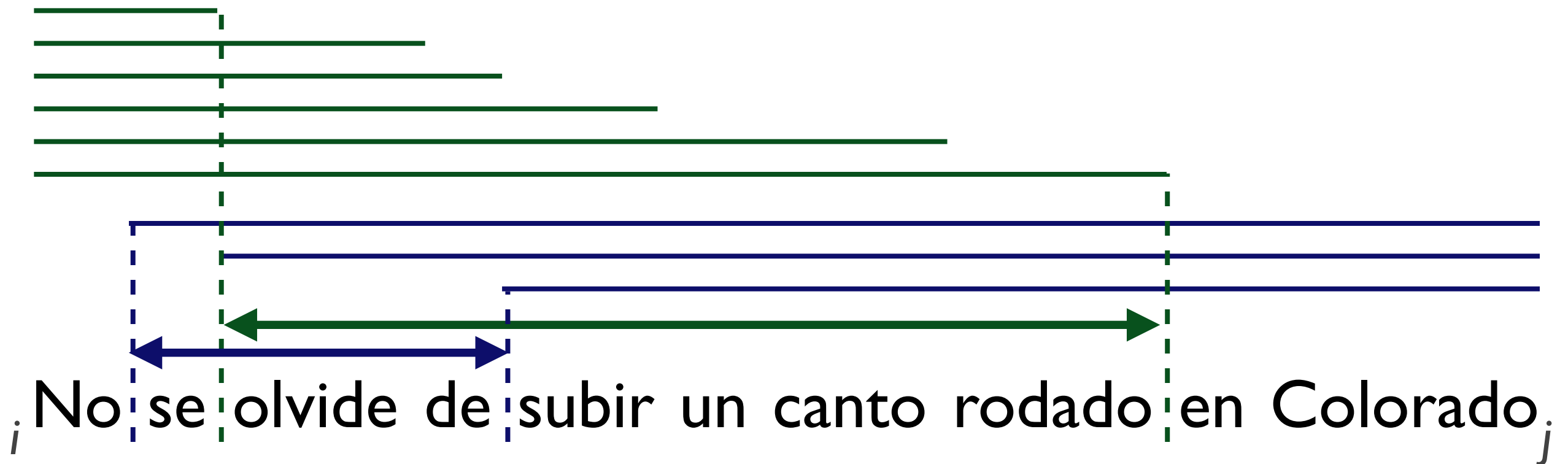


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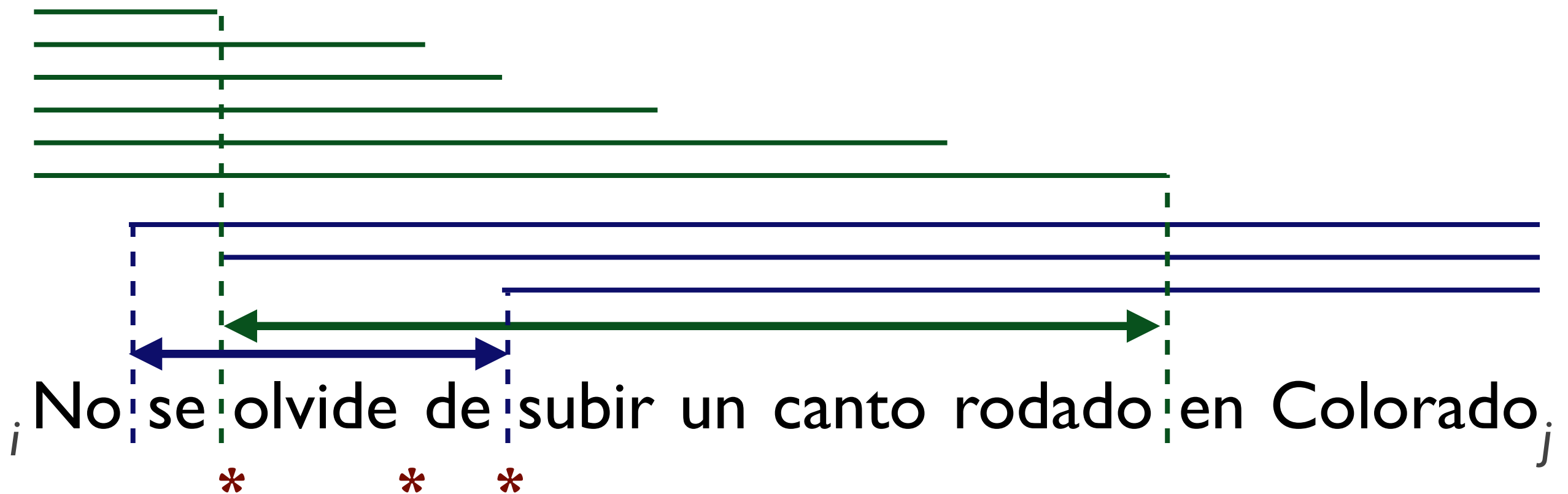
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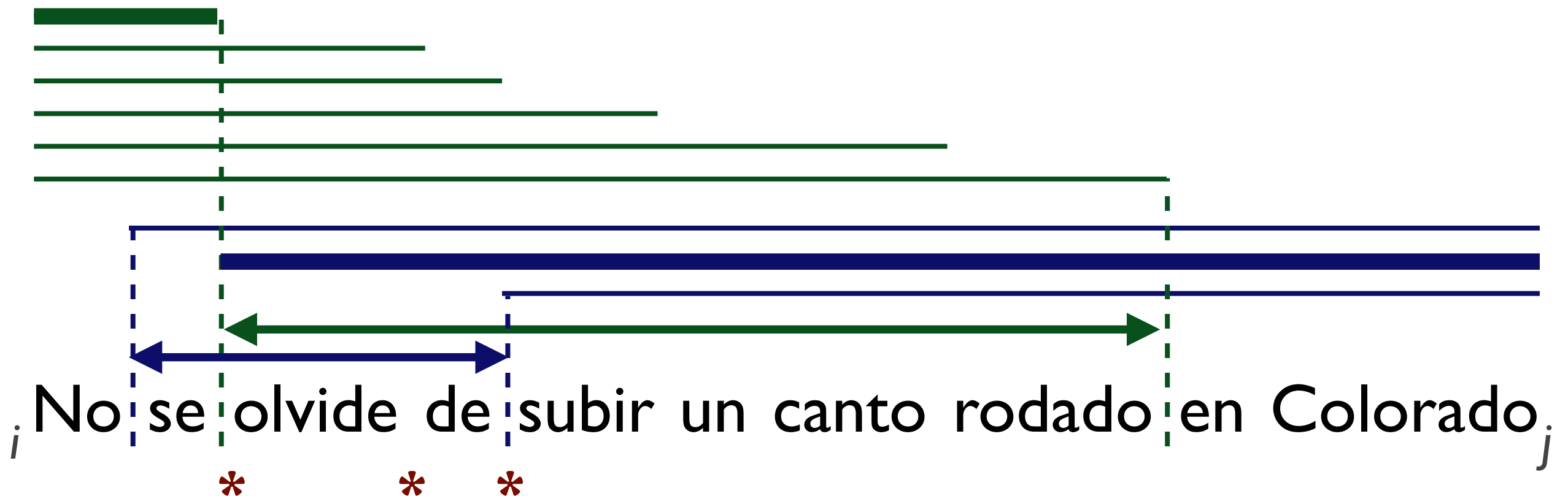


Only consider split points k that might result in a valid parse

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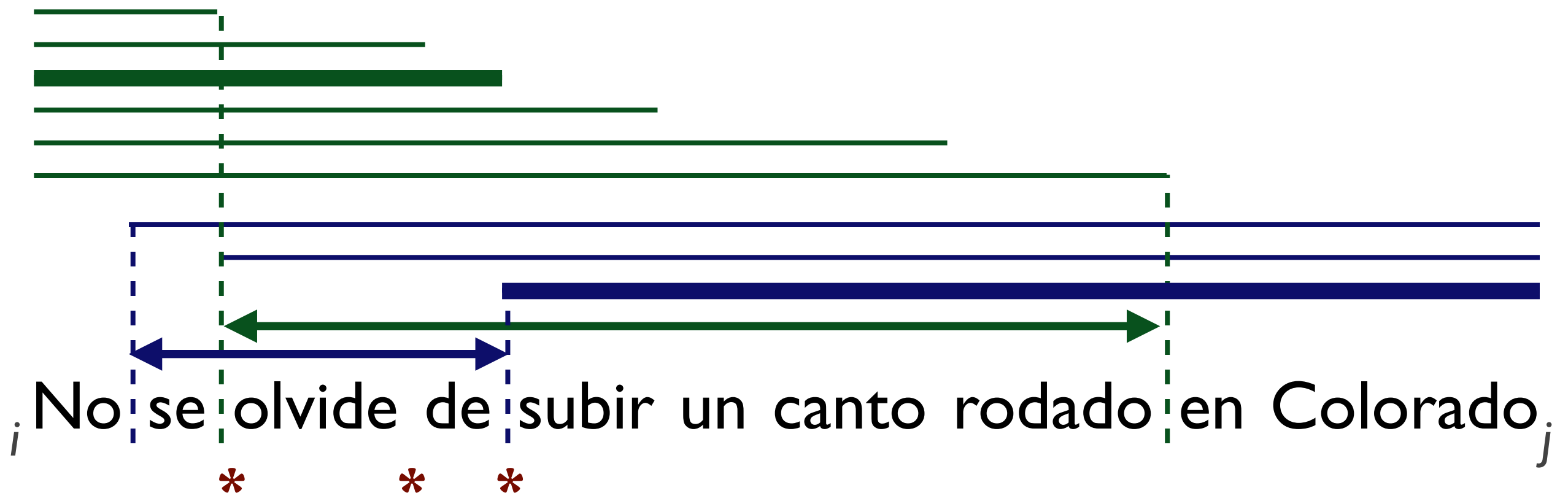


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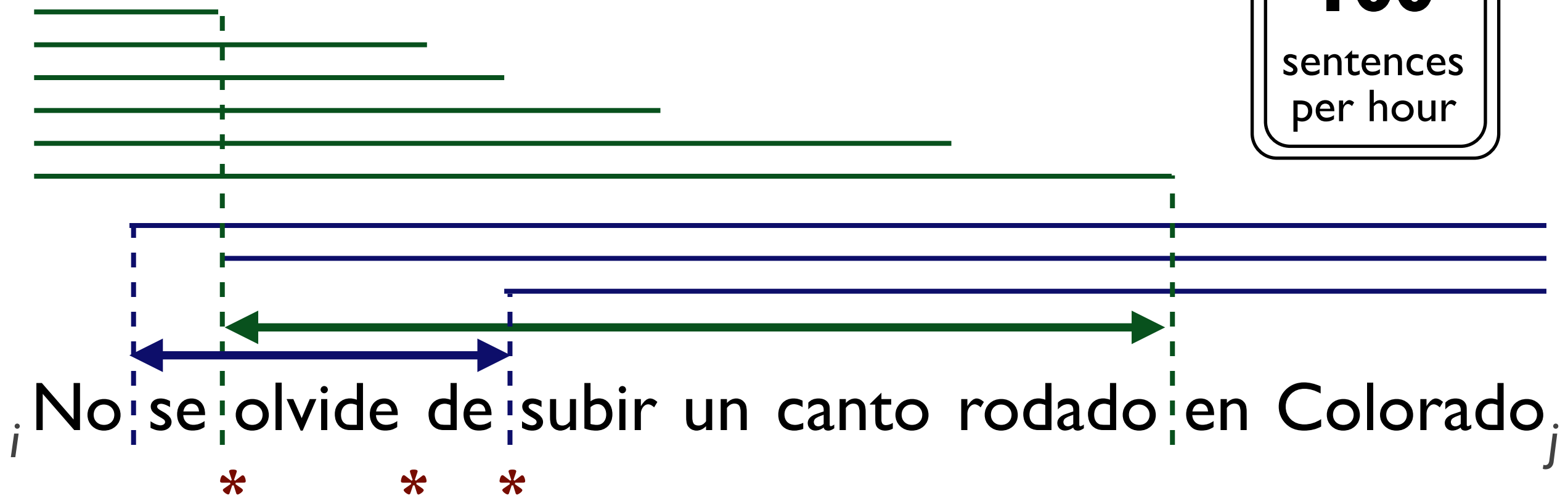
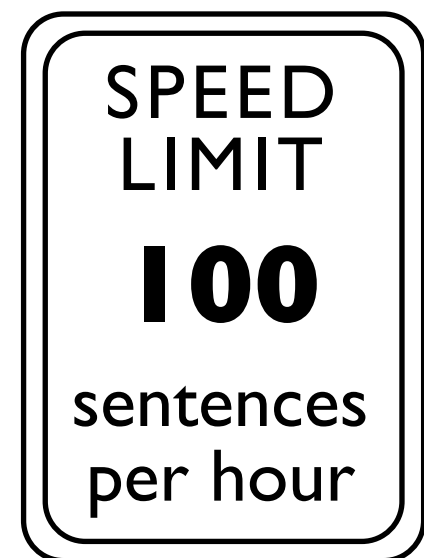


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Speeding up Lexical Rule Application

Problem: *Lexical rules can apply to many spans*

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$S \rightarrow$ No se olvide de subir **NP**

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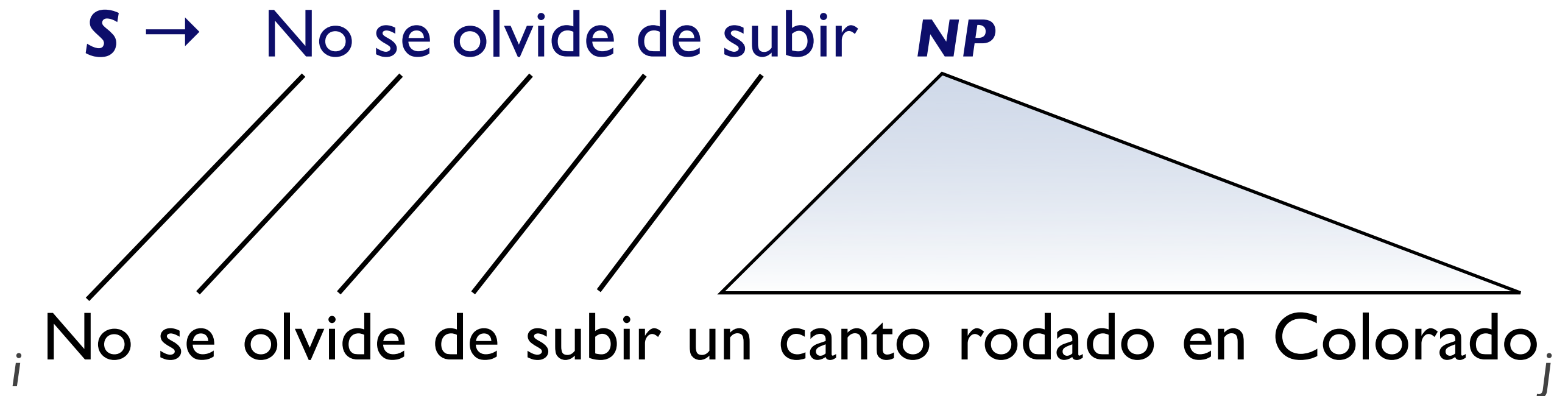
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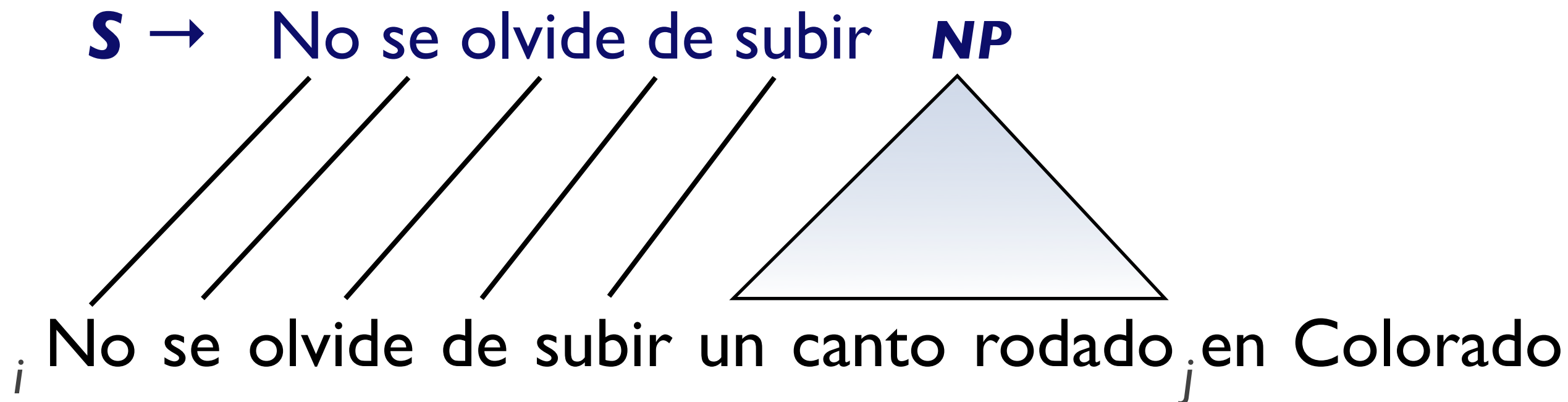
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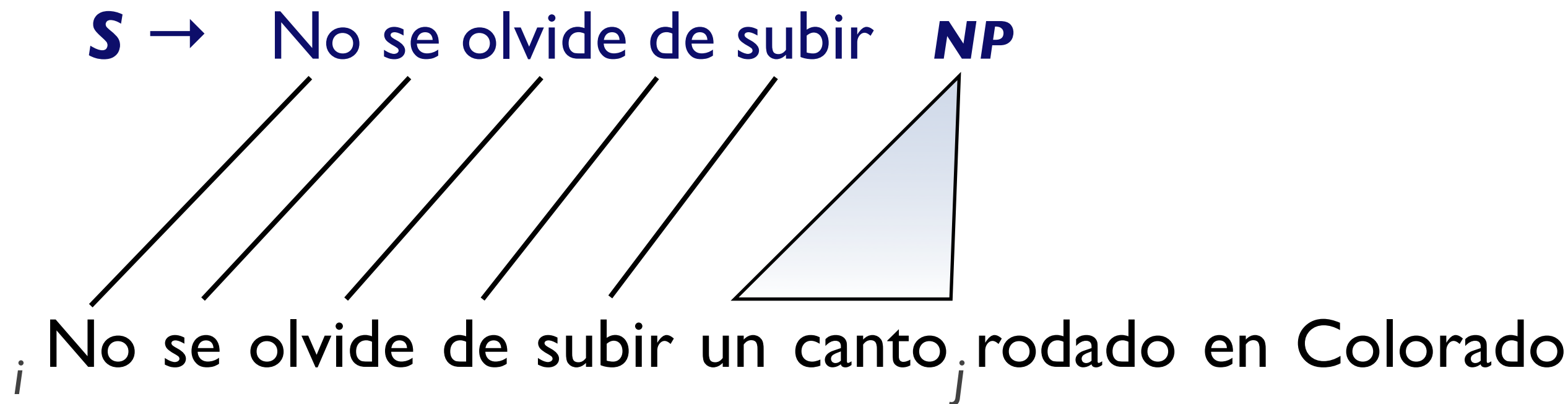
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Transformed rules: $S \rightarrow S/NN \sim PP \ NN \sim PP$

$S/NN \sim PP \rightarrow \text{No se } VB \sim VB \ \text{un}$

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All lexical rule yields begin and end with a lexical item

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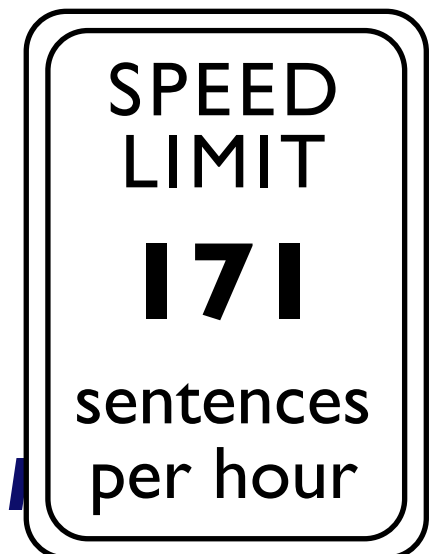
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Binarizing Sequences of Non-Terminals

We must select a binary derivation for each non-terminal sequence

Original:

$S \rightarrow VB \ NP \ NP \ PP$

Binarization options:

$S \rightarrow VB \sim NP \sim NP \ PP$

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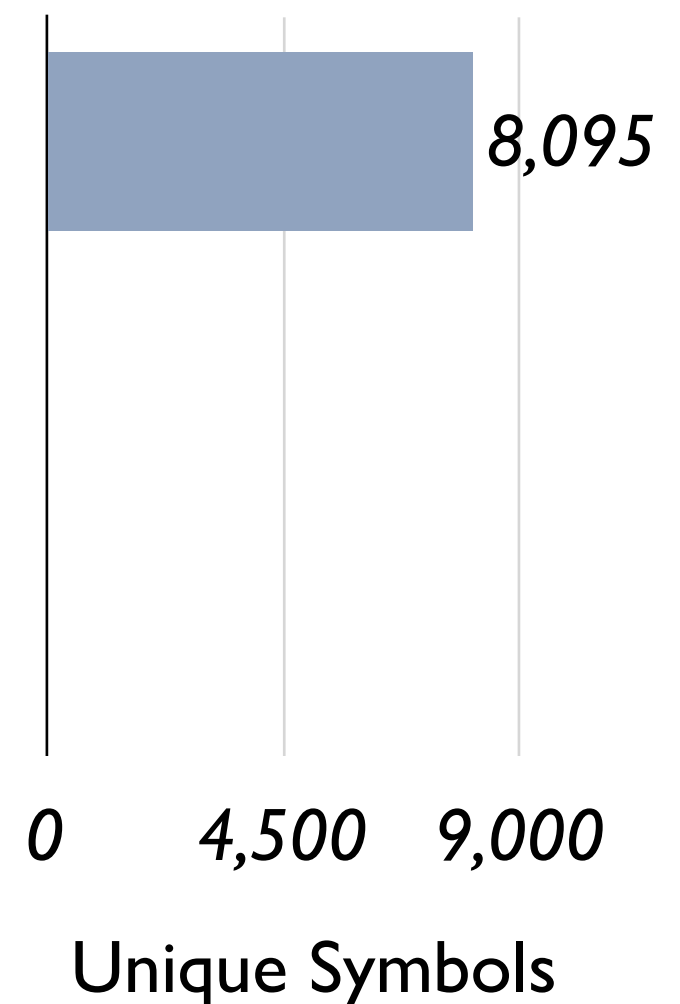
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Binarization of an example
sentence-specific grammar

Right-branching



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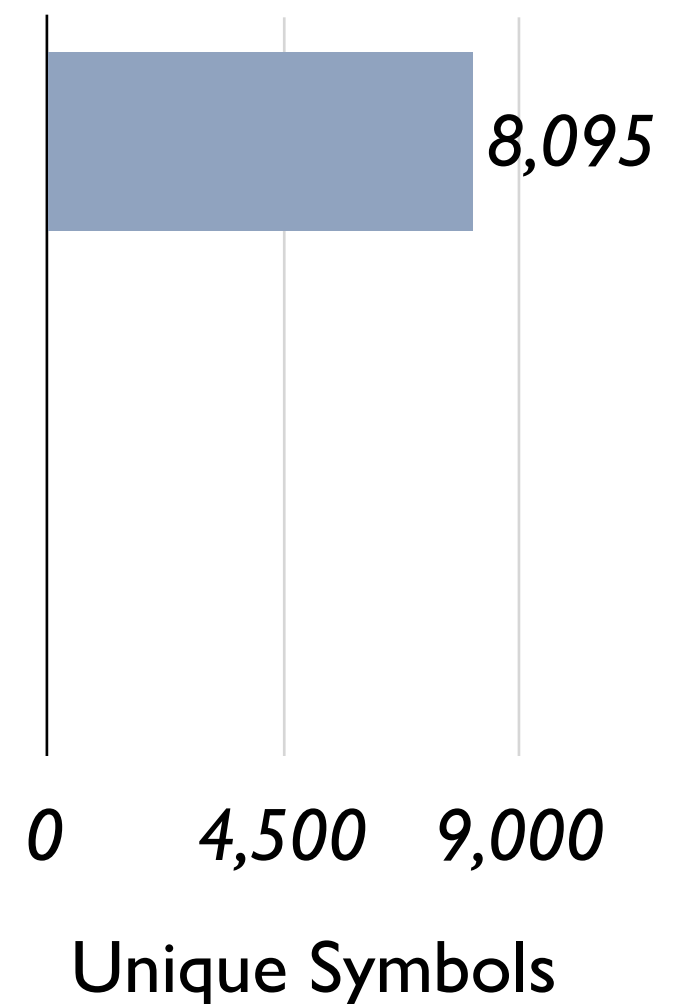
$S \rightarrow VB \sim NP \ NP \sim PP$

Objective function:

The minimum number of grammar symbols, such that all non-terminal sequences have binary derivations

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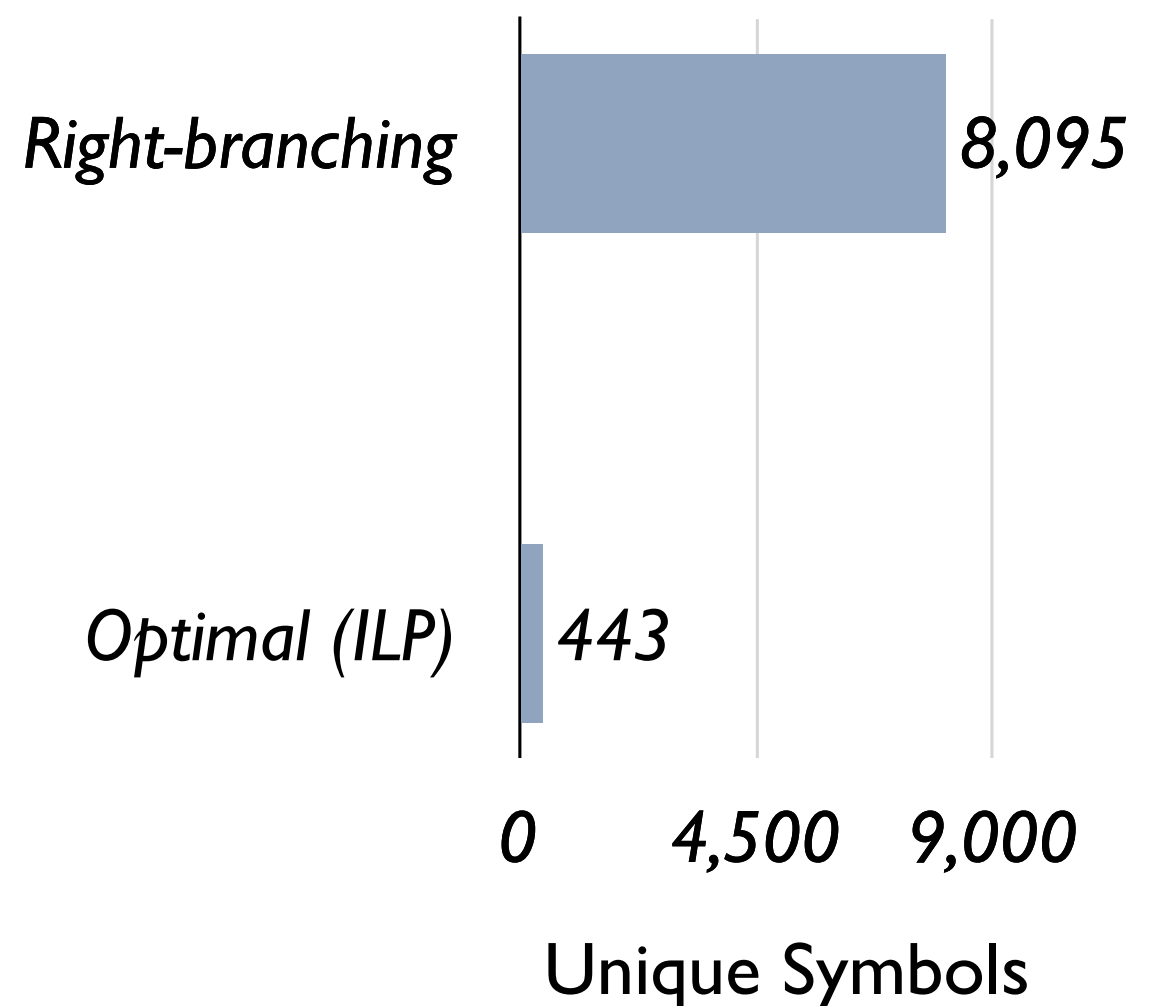
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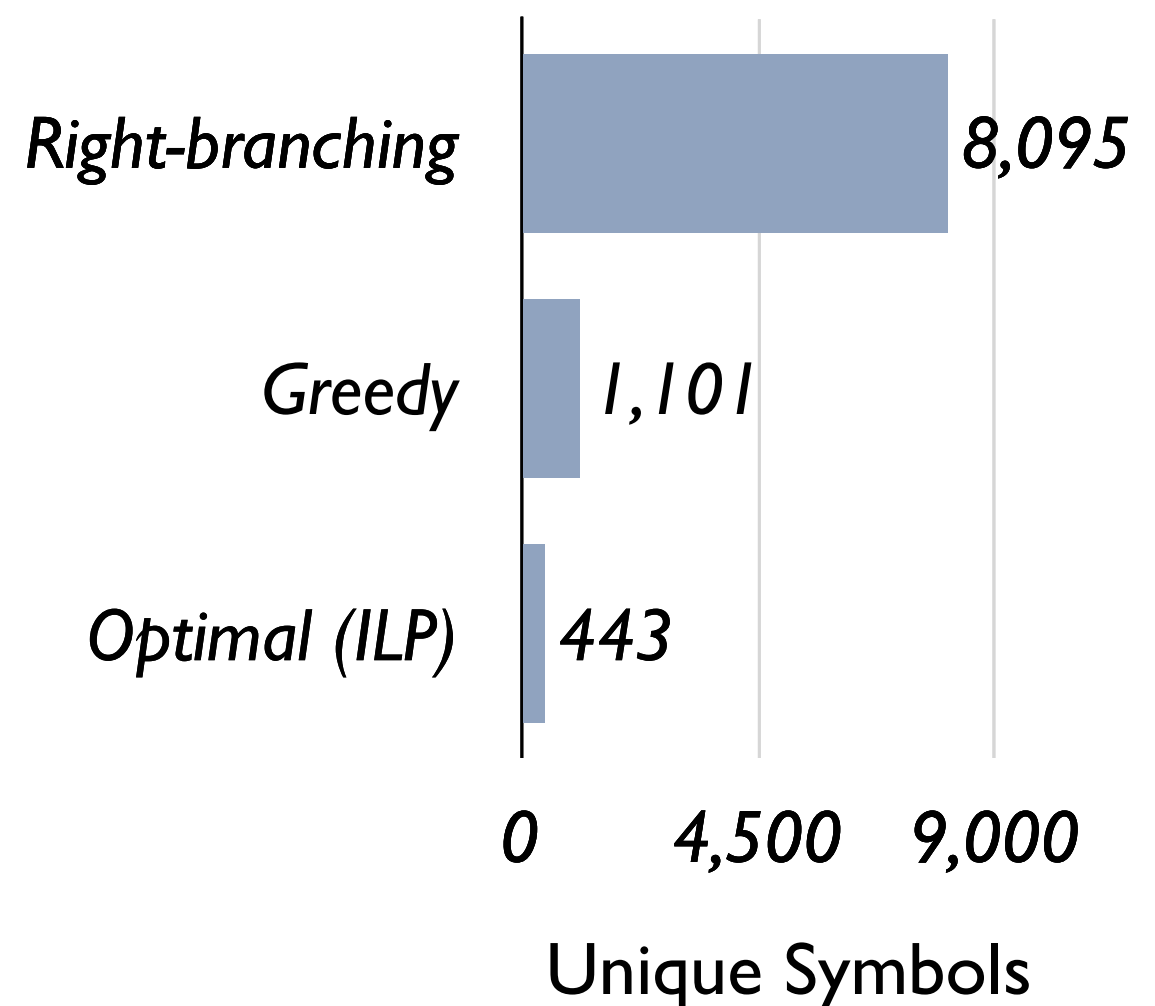
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Subsets of Grammar Symbols

Problem: *Certain large rules always introduce new symbols*

$S \rightarrow VBP \ RB \ VB \ TO \ \text{subir} \ NP \ VP$

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 $\dashrightarrow VBP \sim RB \sim VB \sim TO$

Subsets of Grammar Symbols

Problem: *Certain large rules always introduce new symbols*

$S \rightarrow$ **VBP RB VB TO** *subir* **NP VP**
 \rightarrow **VBP~RB~VB~TO**

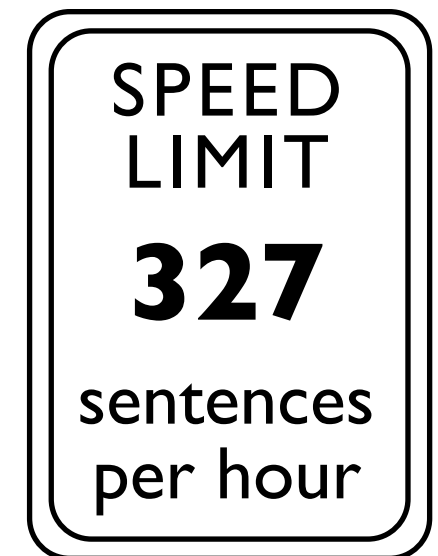
Coarse-to-fine parsing:

1. Choose a small subset of high-use symbols
2. Parse with the coarse grammar and prune unlikely states
3. Parse with the fine grammar in the pruned search space

Subsets of Grammar Symbols

Problem: *Certain large rules always introduce new symbols*

$S \rightarrow \boxed{VBP \ RB \ VB \ TO} \text{ subir } NP \ VP$
 $\rightarrow VBP \sim RB \sim VB \sim TO$



Coarse-to-fine parsing:

1. Choose a small subset of high-use symbols
2. Parse with the coarse grammar and prune unlikely states
3. Parse with the fine grammar in the pruned search space

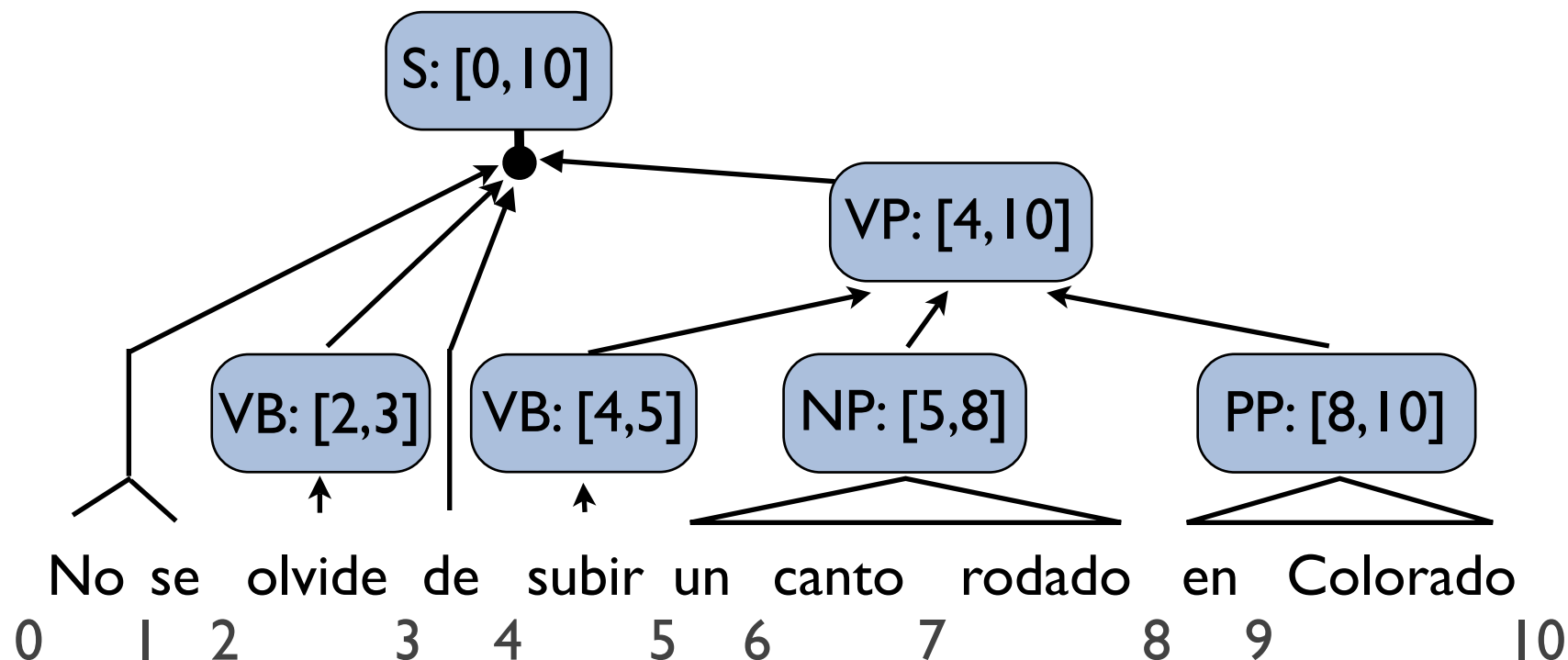


Integrating a Language Model

Approach: *Top-down lazy forest reranking with priority queues (a.k.a., cube growing)*

Integrating a Language Model

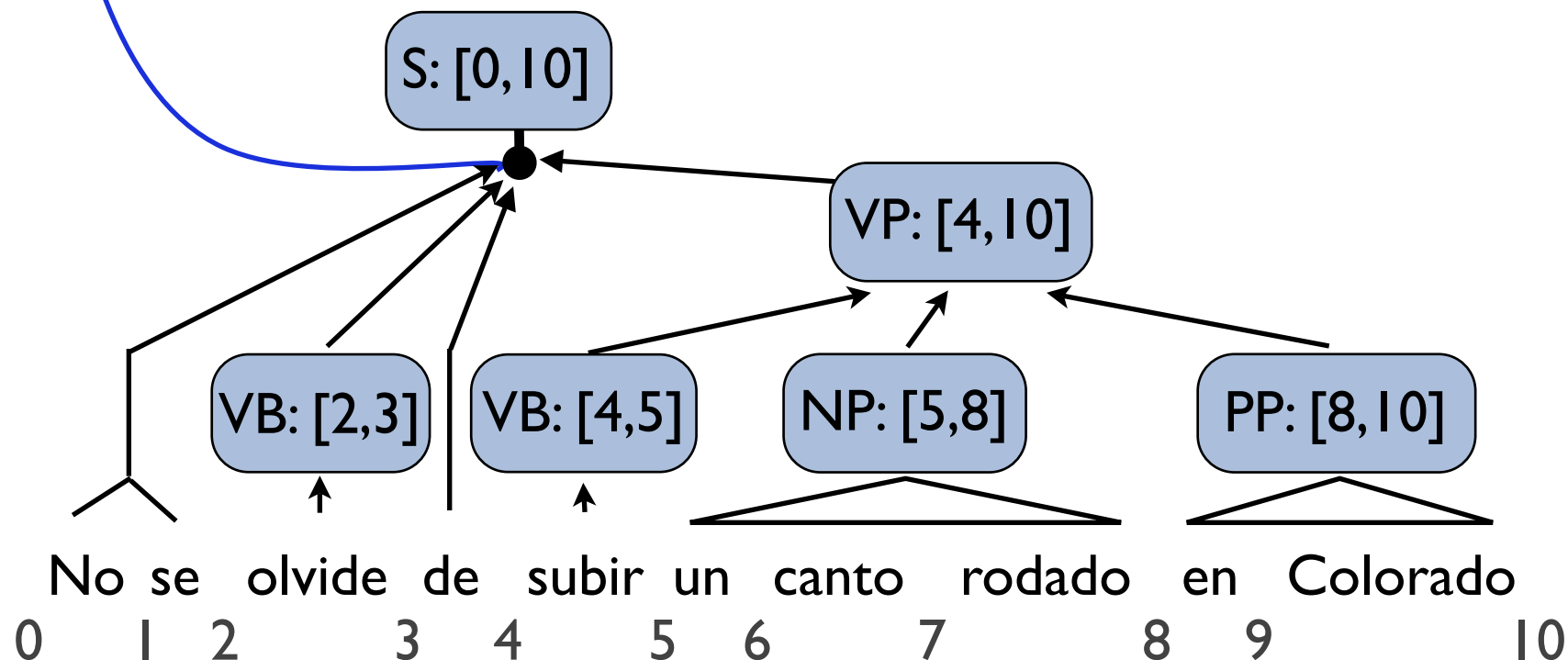
Approach: *Top-down lazy forest reranking with priority queues (a.k.a., cube growing)*



Integrating a Language Model

Approach: *Top-down lazy forest reranking with priority queues (a.k.a., cube growing)*

S → No se **VB** de **VP** ; *



Integrating a Language Model

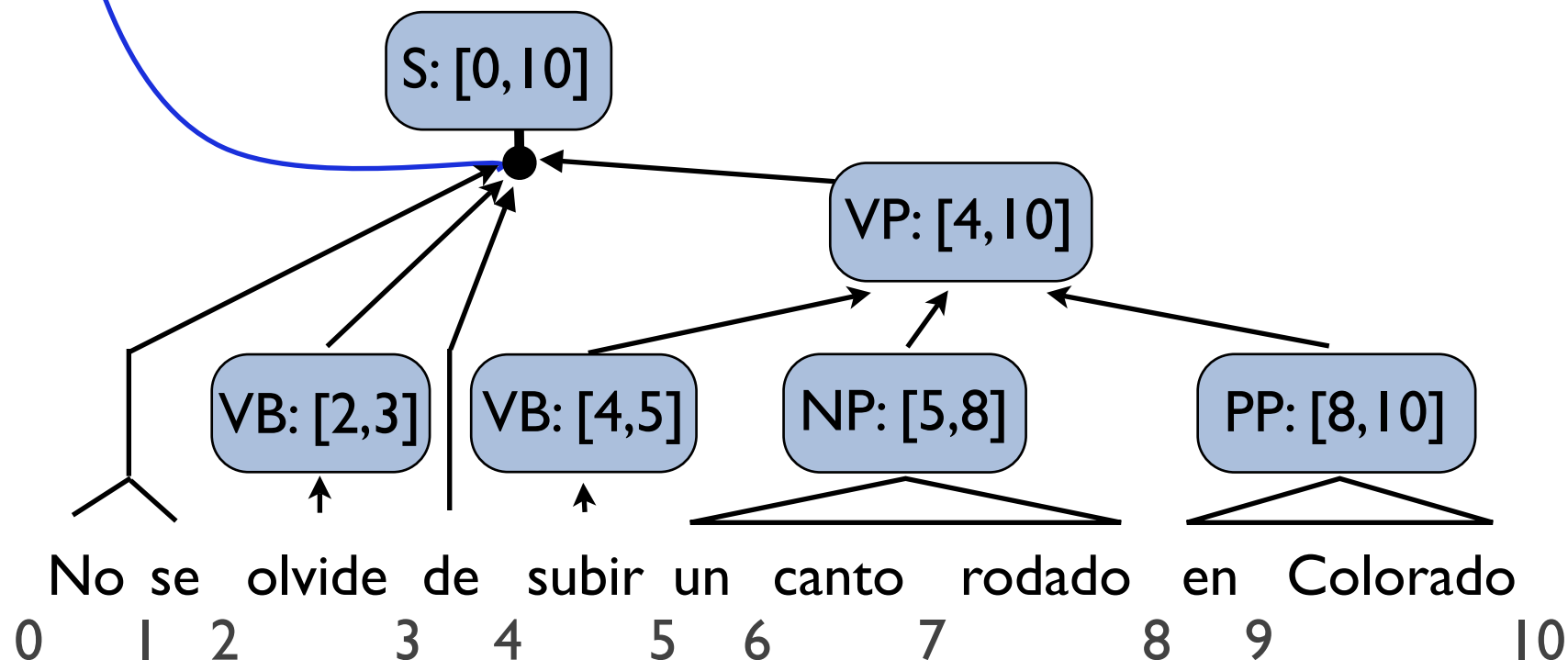
Approach: *Top-down lazy forest reranking with priority queues (a.k.a., cube growing)*

S → No se **VB** de **VP** ; *

S → Don't **VB** to **VP**

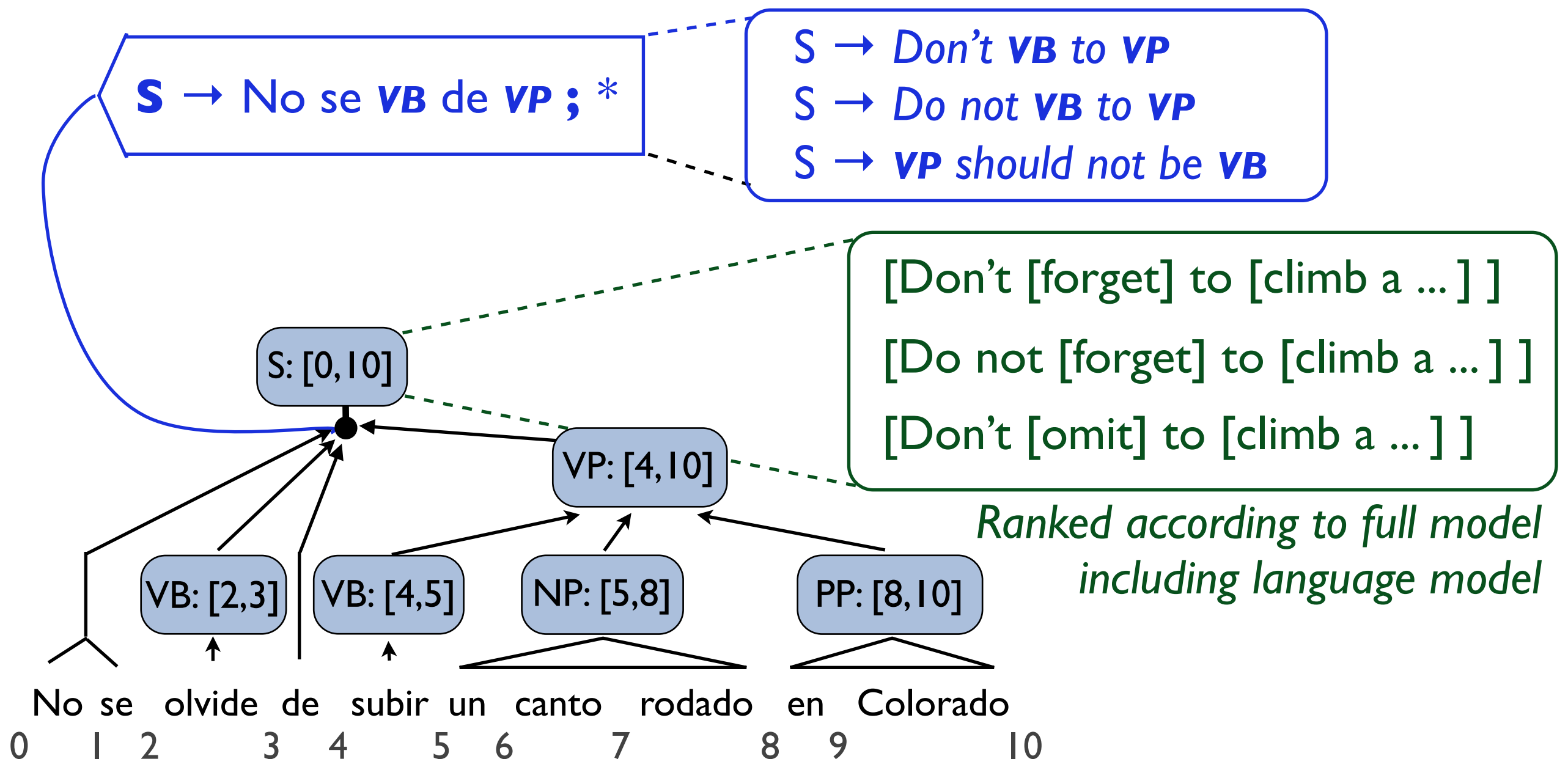
S → Do not **VB** to **VP**

S → **VP** should not be **VB**



Integrating a Language Model

Approach: *Top-down lazy forest reranking with priority queues (a.k.a., cube growing)*



Coarse-to-Fine LM Integration

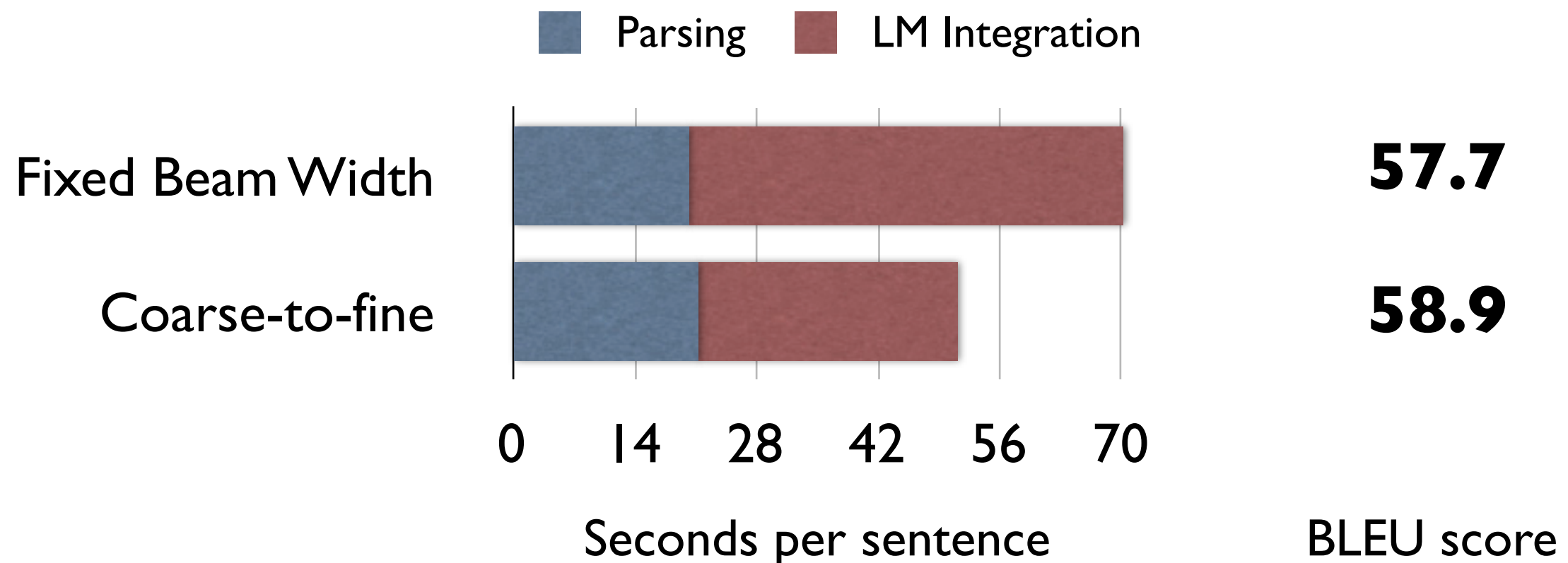
Observation: *The best translations almost always have a translation model score close to the Viterbi parse score*

Coarse-to-fine beaming: *A forest node's beam size is proportional to its posterior under the translation model*

Coarse-to-Fine LM Integration

Observation: *The best translations almost always have a translation model score close to the Viterbi parse score*

Coarse-to-fine beaming: *A forest node's beam size is proportional to its posterior under the translation model*



Summary

- Parsing with the projection of a tree transducer grammar is a non-trivial search problem
- Grammar transformations and algorithmic optimizations decrease parsing time
- Coarse-to-fine search speeds up parsing and makes language model integration more accurate

Thanks!

Questions?

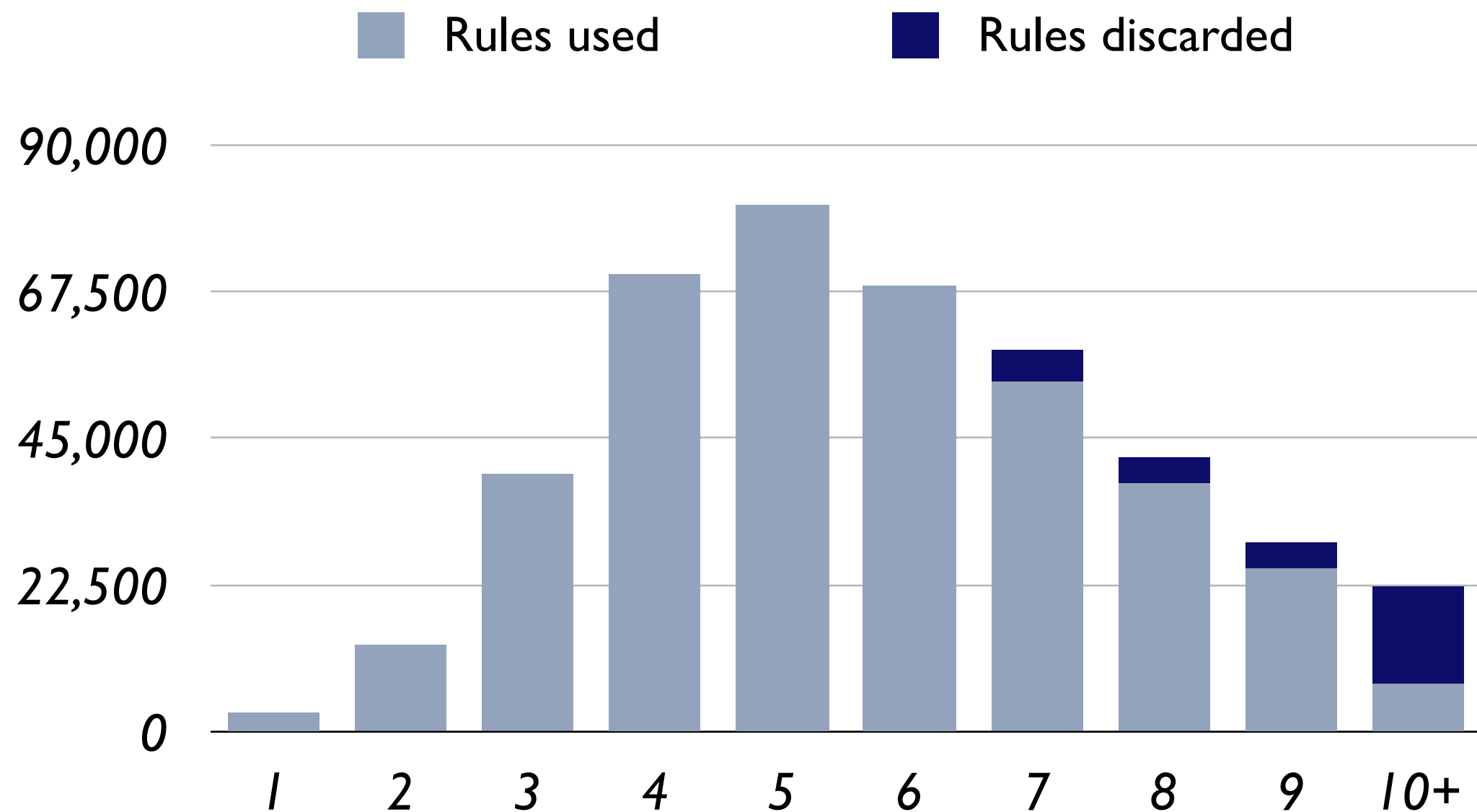


The Size of Tree Transducer Grammars

■ Rules used

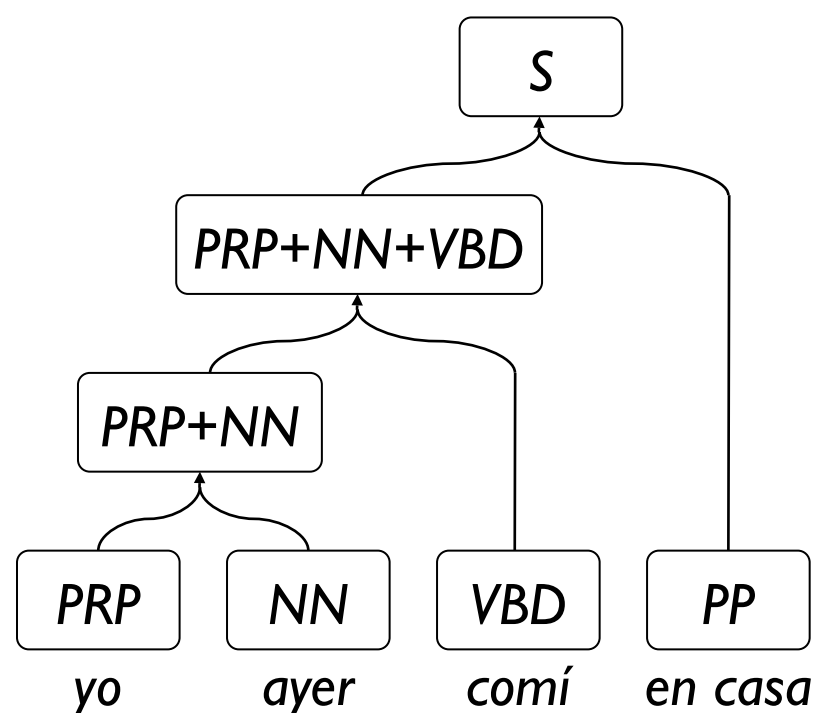
■ Rules discarded

The Size of Tree Transducer Grammars



Rebinarizing for LM Integration (ACL '09)

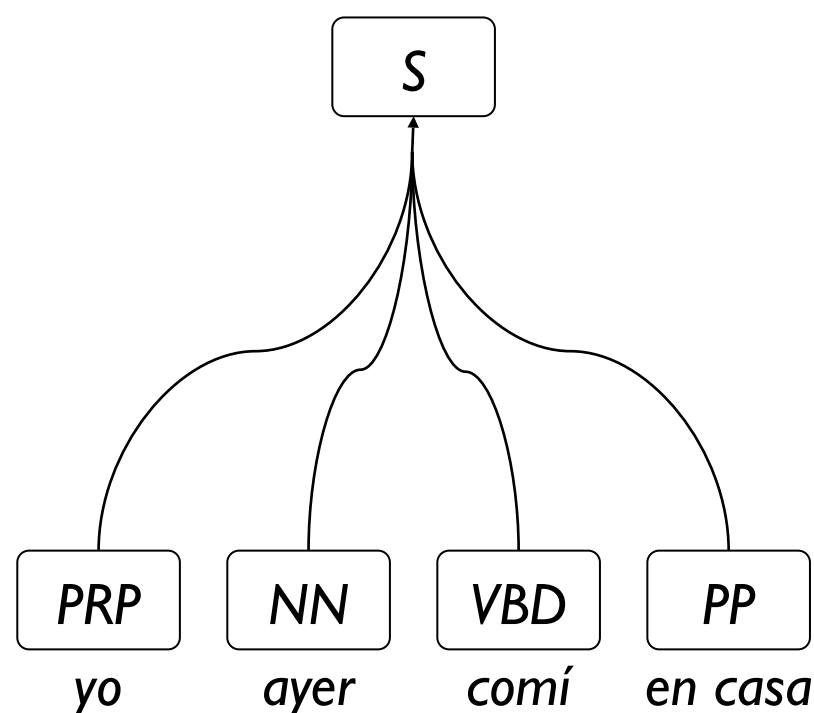
Parse with ALNF grammar



$S \rightarrow$ PRP_1 VBD_3 NN_2 PP_4
 $[[PRP_1$ $NN_2]$ $VBD_3]$ PP_4

$S \rightarrow$ PRP_1 VBD_3 PP_4 NN_2
 $[[PRP_1$ $NN_2]$ $VBD_3]$ PP_4

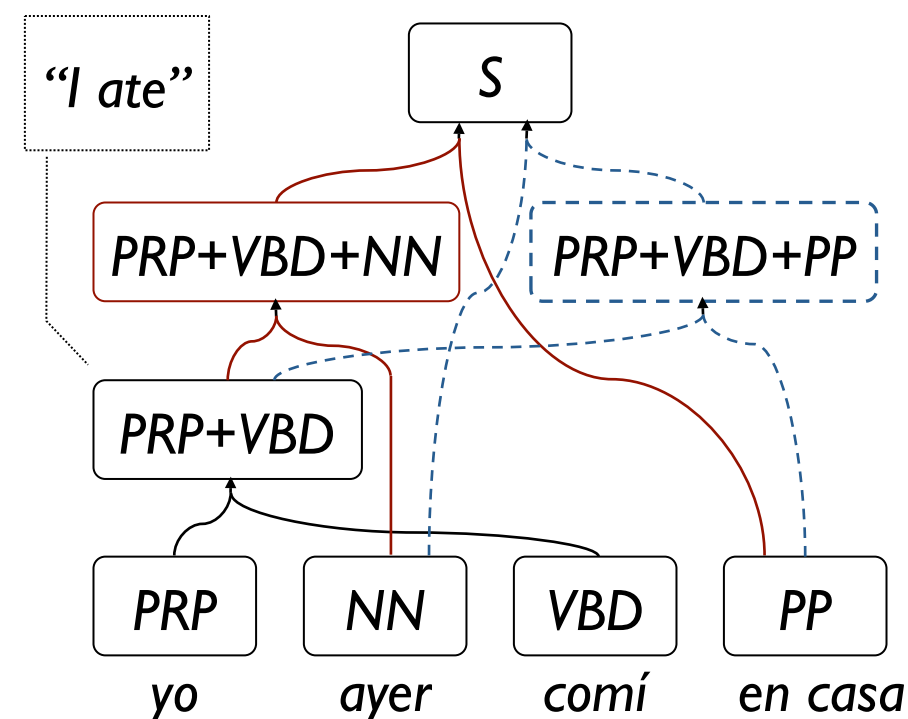
Collapse out binarization



$S \rightarrow$ PRP_1 VBD_3 NN_2 PP_4
 PRP_1 NN_2 VBD_3 PP_4

$S \rightarrow$ PRP_1 VBD_3 PP_4 NN_2
 PRP_1 NN_2 VBD_3 PP_4

Rebinarize for LM integration



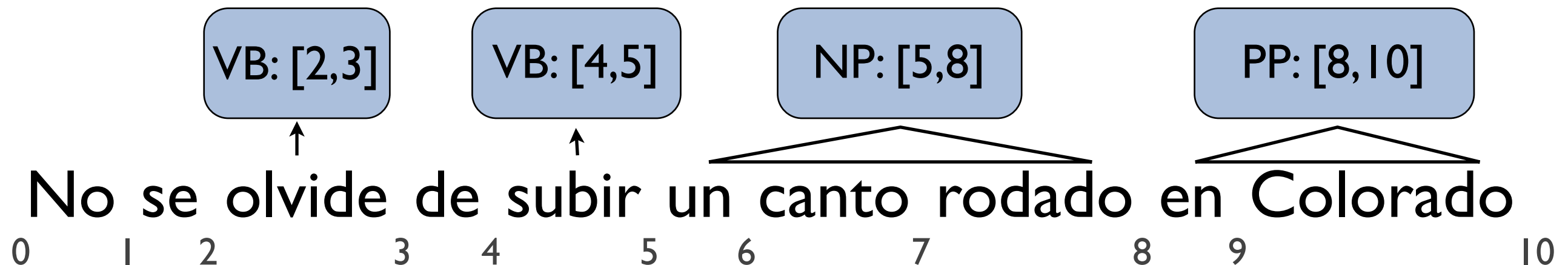
$S \rightarrow$ $[[PRP_1$ $VBD_3]$ $NN_2]$ PP_4
 PRP_1 NN_2 VBD_3 PP_4

$S \rightarrow$ $[[PRP_1$ $VBD_3]$ $PP_4]$ NN_2
 PRP_1 NN_2 VBD_3 PP_4

Multi-Pass Syntactic Decoding

Parse input sentence
with source-side
grammar projection

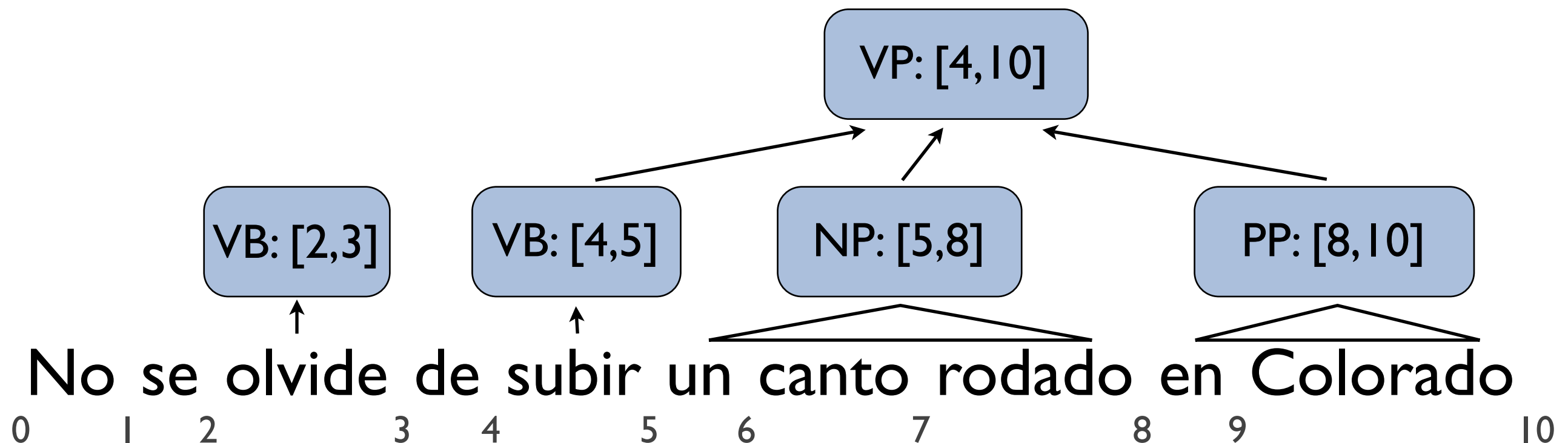
Rerank derivations
rooted at each state with
a language model



Multi-Pass Syntactic Decoding

Parse input sentence
with source-side
grammar projection

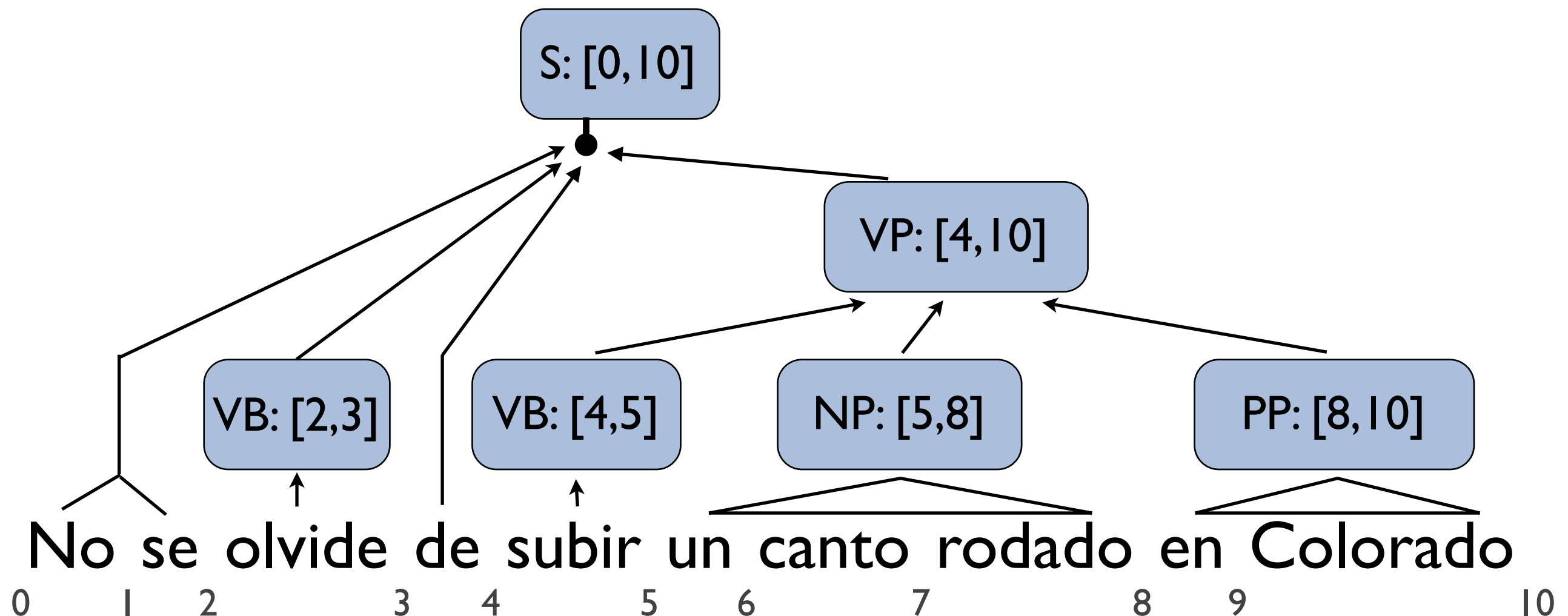
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Multi-Pass Syntactic Decoding

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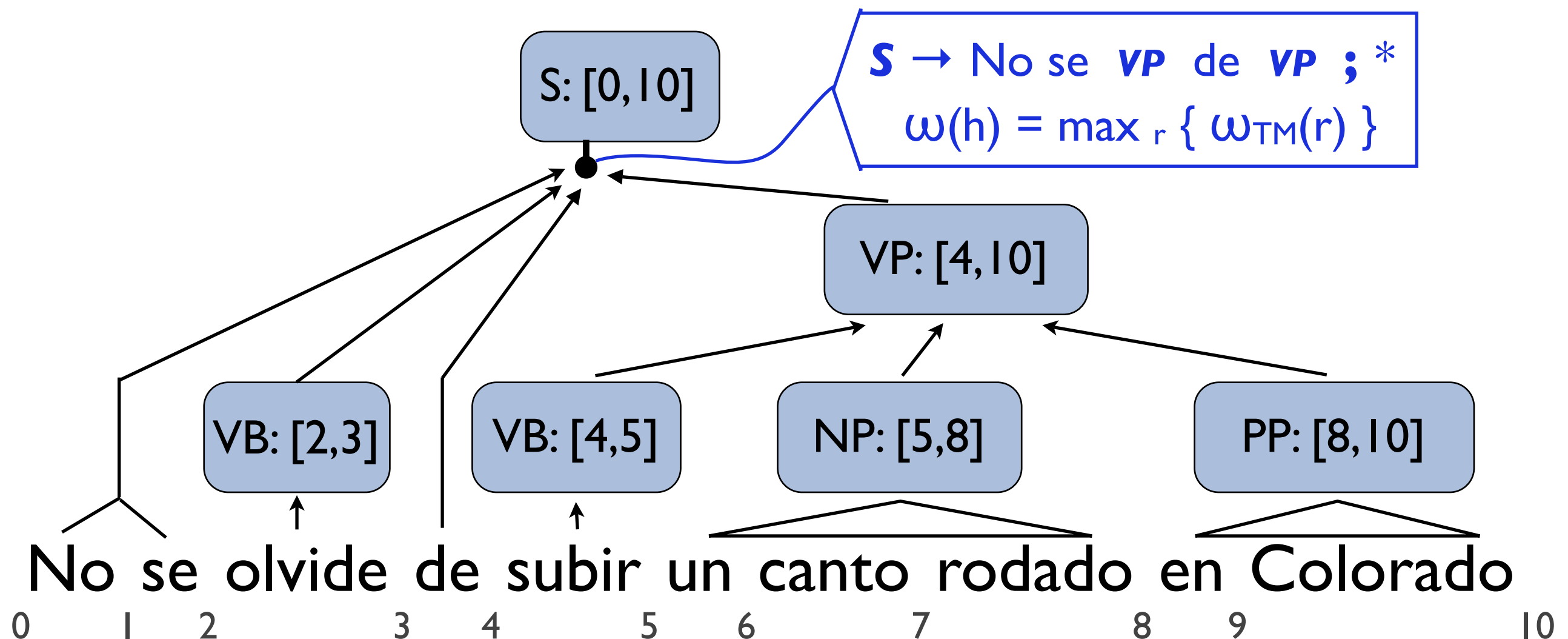
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Multi-Pass Syntactic Decoding

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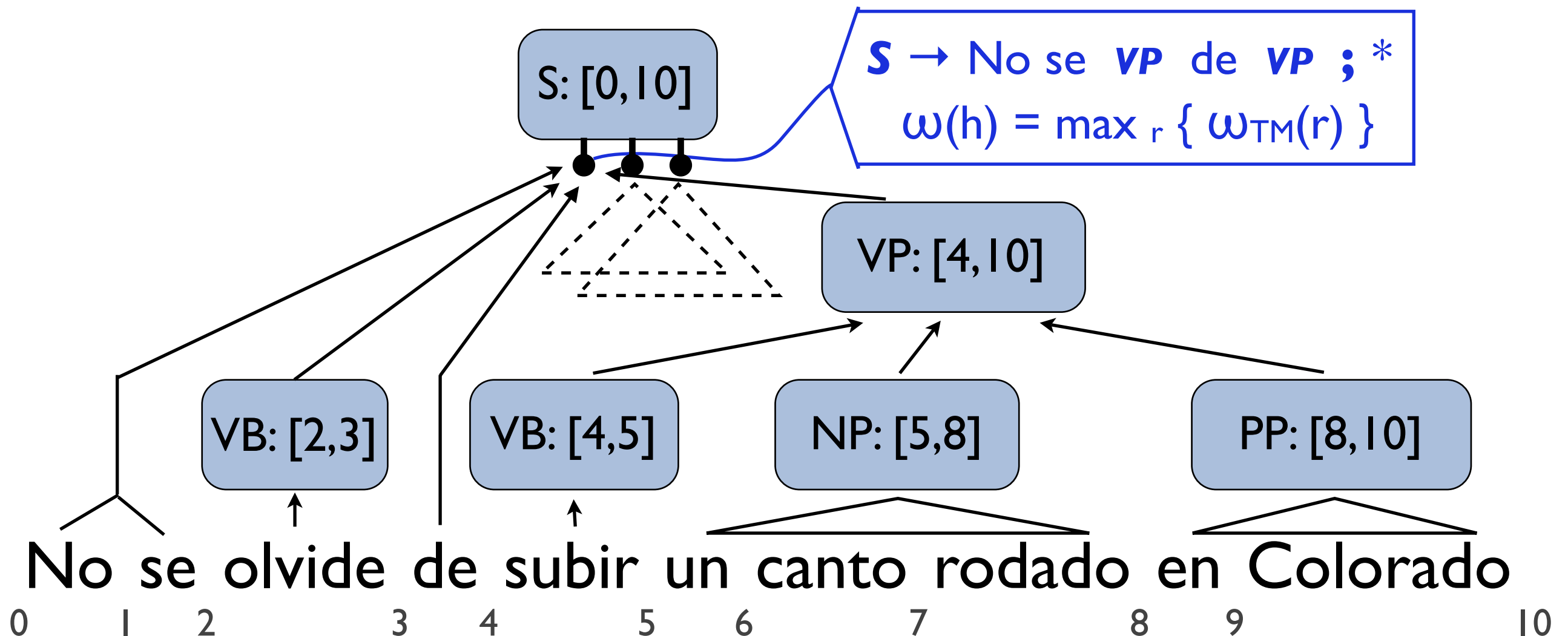
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Multi-Pass Syntactic Decoding

Parse input sentence
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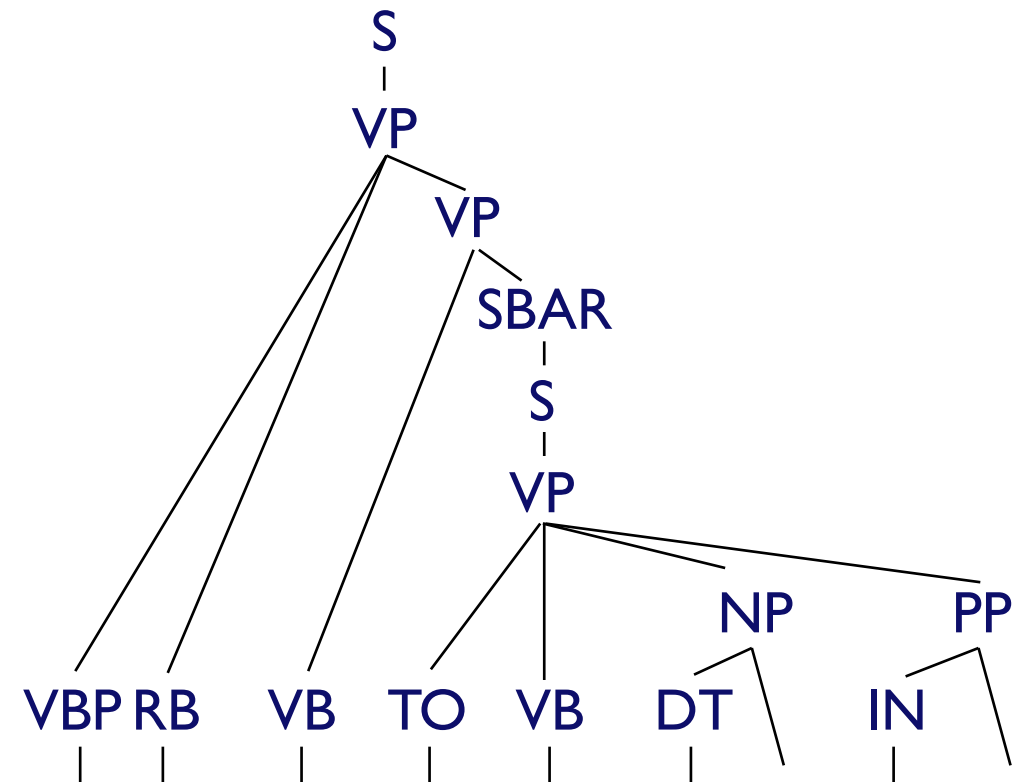
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Tree Transducer Grammars

No se olvide de subir un canto rodado en Colorado

Rules



No se olvide de subir un NN_1 en NNP_2 ; *Don't forget to climb a NN_1 in NNP_2*

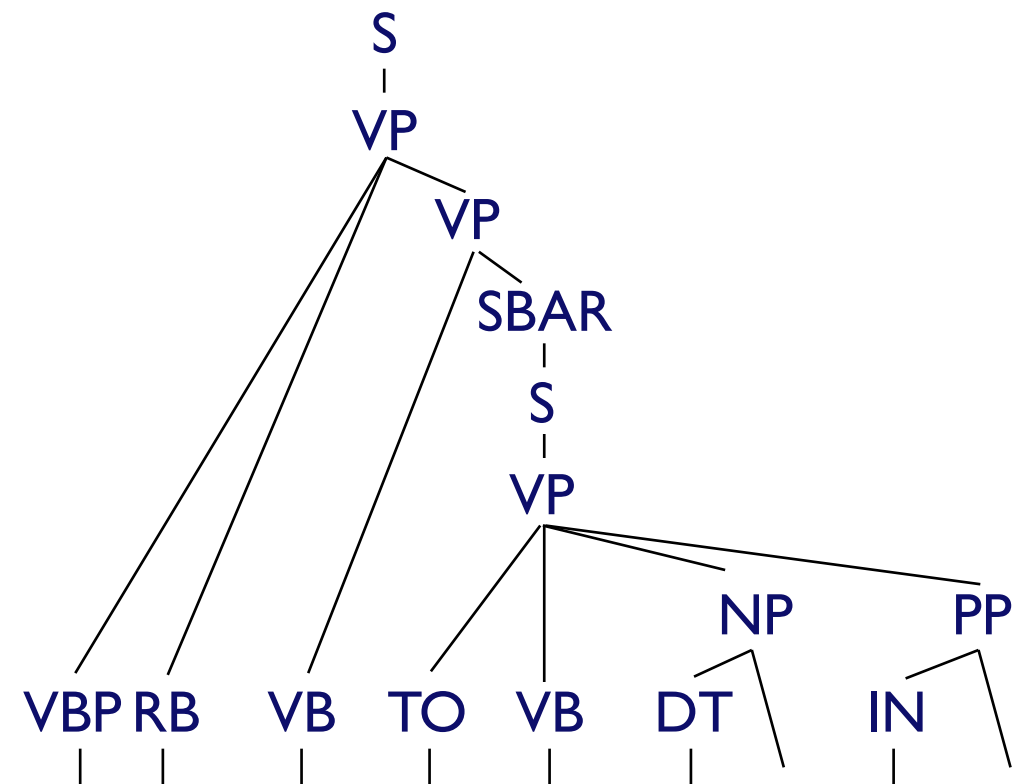
Tree Transducer Grammars

No se olvide de subir un canto rodado en Colorado

Rules

NN
|
canto rodado ; *boulder*

NNP
|
Colorado ; *Colorado*



No se olvide de subir un NN_1 en NNP_2 ; *Don't forget to climb a NN_1 in NNP_2*

Tree Transducer Grammars

No se olvide de subir un canto rodado en Colorado

Rules

