

# CKY Parsing & CNF Conversion

LING 571 — Deep Processing Techniques for NLP  
Shane Steinert-Threlkeld

# Announcements

- HW #1 due tonight at 11:59pm.
- Use full paths (for python binary and for files)!
  - Condor broken symlink:
    - /mnt/dropbox instead of /dropbox will always work
    - Updated example.sh and hw1 spec to reflect this

# Type Hinting in Python



\*always\* type-annotate your Python

the cost to you is minimal (you have to type a few extra characters)

the benefits to you are great (documentation + help from your IDE / editor) \*even if you never run a static type checker\*

it's such a no-brainer

```
from typing import List

def process(xs: List[int]) -> None:
    xs.
        append
        clear
        copy
        count
        extend
        index
        insert
        pop
        remove
        reverse
        sort
        __add__
```

# Type Hinting in Python

- Supported in  $\geq 3.6$

```
from nltk.grammar import Production
```

```
def fix_hybrid_production(hybrid_prod: Production) -> list[Production]:  
    ...
```

Joel Grus ❤️  
@joelgrus

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- [https://mypy.readthedocs.io/en/stable/cheat\\_sheet\\_py3.html](https://mypy.readthedocs.io/en/stable/cheat_sheet_py3.html)

A screenshot of a Python IDE interface. On the left, there is a code editor window containing the following Python code:from typing import List  
  
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A tooltip or callout box is overlaid on the screen, pointing to the 'xs.' part of the code. The tooltip contains the following text:

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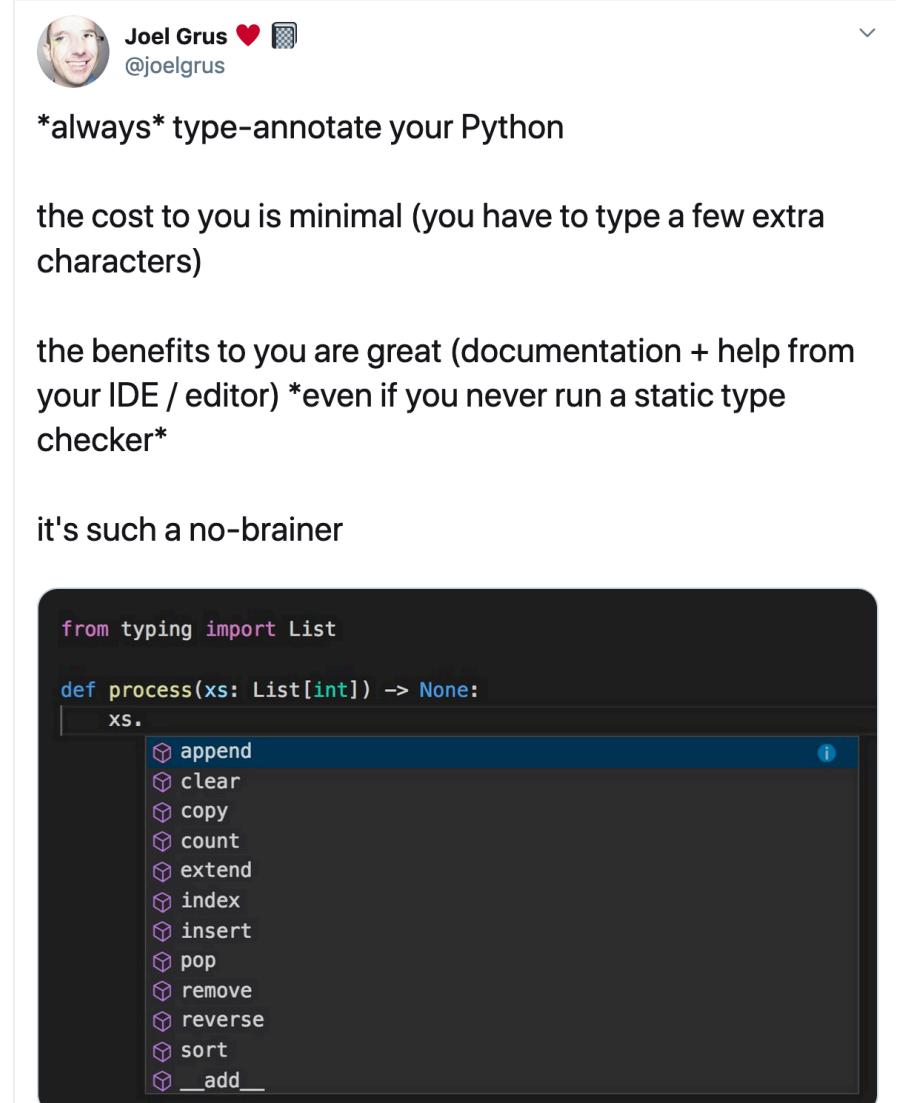
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- <https://peps.python.org/pep-0483/>



A screenshot of a Python IDE showing a code editor with the following code:

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        __add__
```

The code editor shows a dropdown menu with various methods for the `List[int]` type. The `append` method is highlighted.

# Recursion in the Wild



Stephen Boisvert @srboisvert.bsky.social · 5d  
They sometimes surprise with their diets.

ALT

3 3 48 ...

Ed Yong · 5d  
@edyong209.bsky.social

A snapping-turtle-snapping heron.

And you: a snapping-turtle-snapping-heron-snapping human.

Sep 28, 2023 at 1:31 PM

2 reposts 78 likes

# Roadmap

- Parsing-as-Search
- Parsing Challenges
- Strategy: Dynamic Programming
- Grammar Equivalence
- CKY parsing algorithm

# Computational Parsing

- Given a body of (annotated) text, how can we derive the grammar rules of a language, and employ them in automatic parsing?
  - Treebanks & PCFGs
- Given a grammar, how can we derive the analysis of an input sentence?
  - Parsing as search
  - CKY parsing
  - Conversion to CNF

# What is Parsing?

- CFG parsing is the task of assigning trees to input strings
  - For any input  $A$  and grammar  $G$ 
    - ...assign  $\geq 0$  parse trees  $T$  that represent its syntactic structure, and...
    - Cover all and only the elements of  $A$
    - Have, as root, the start symbol  $S$  of  $G$
    - ...do not necessarily pick one single (or correct) analysis
  - Subtask: Recognition
    - Given input  $A, G$  – is  $A$  in language defined by  $G$  or not?

# Motivation

- Is this sentence in the language — i.e. is it “grammatical?”
  - *\* I prefer United has the earliest flight.*
  - FSAs accept regular languages defined by finite-state automata.
  - Our parsers accept languages defined by CFG (equiv. pushdown automata).

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- Is this sentence in the language — i.e. is it “grammatical?”
  - *\* I prefer United has the earliest flight.*
  - FSAs accept regular languages defined by finite-state automata.
  - Our parsers accept languages defined by CFG (equiv. pushdown automata).
- What is the syntactic structure of this sentence?
  - *What airline has the cheapest flight?*
  - *What airport does Southwest fly from near Boston?*
  - Syntactic parse provides framework for semantic analysis
    - What is the subject? Direct object?

# Parsing as Search

- Syntactic parsing searches through possible trees to find one or more trees that derive input

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- Syntactic parsing searches through possible trees to find one or more trees that derive input
- Formally, search problems are defined by:
  - Start state  $S$
  - Goal state  $G$  (with a test)
  - Set of actions that transition from one state to another
    - “Successor function”
  - A path cost function

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- Path cost:
  - ...ignored for now.

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- Search start node (initial state):
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  - Start symbol of CFG
- Goal node:
  - Full parse tree: covering all of, and only the input, rooted at  $S$

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- Depth First
  - Keep expanding nonterminals until they reach words
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  - Keep expanding nonterminals until they reach words
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- Breadth First
  - Consider all parses that expand a single nonterminal...
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- Other alternatives, if have associated path costs.

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- Two constraints on parsing:
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  - Must cover exactly the input string
- Correspond to main parsing search strategies
  - Top-down search (Goal-directed)
  - Bottom-up search (Data-driven search)

# A Grammar

Grammar	Lexicon
$S \rightarrow NP\ VP$	$Det \rightarrow that \mid this \mid a$
$S \rightarrow Aux\ NP\ VP$	$Noun \rightarrow book \mid flight \mid meal \mid money$
$S \rightarrow VP$	$Verb \rightarrow book \mid include \mid prefer$

Jurafsky & Martin, *Speech and Language Processing*, p.390

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$NP \rightarrow Pronoun$	$Pronoun \rightarrow I \mid she \mid me$
$NP \rightarrow Proper-Noun$	$Proper-Noun \rightarrow Houston \mid NWA$
$NP \rightarrow Det\ Nominal$	$Aux \rightarrow does$
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$VP \rightarrow Verb$	

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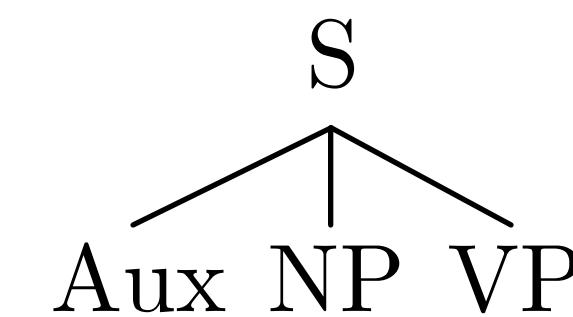
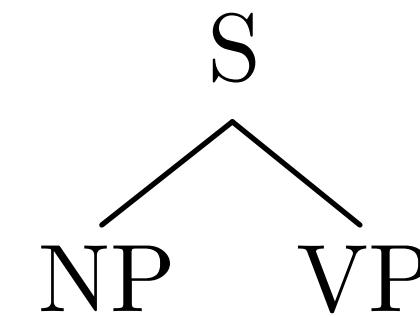
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  - e.g.  $S \rightarrow NP \ VP$
- Successively expand nonterminals
  - e.g.  $NP \rightarrow Det \ Nominal; \ VP \rightarrow V \ NP$
- Terminate when all leaves are terminals

# Depth-First Search

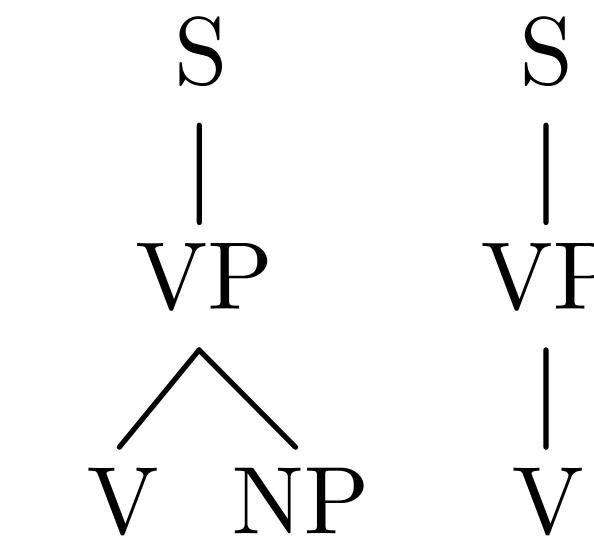
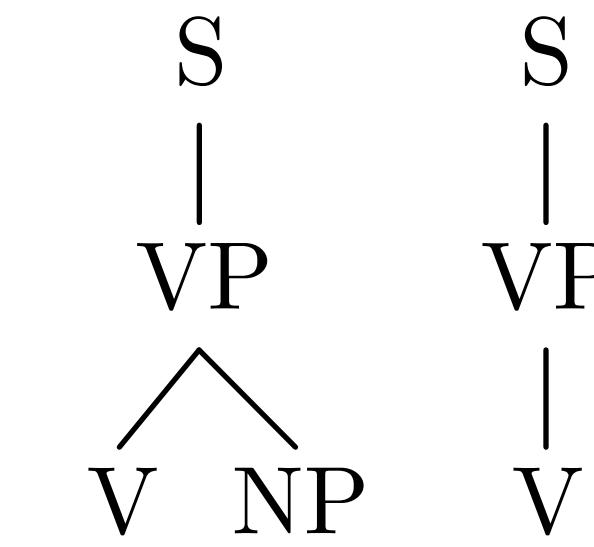
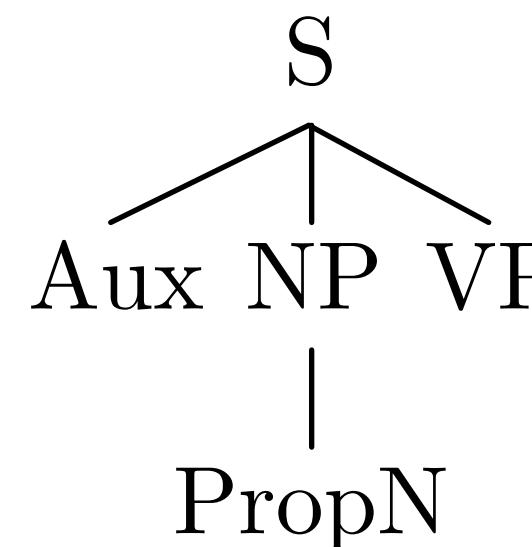
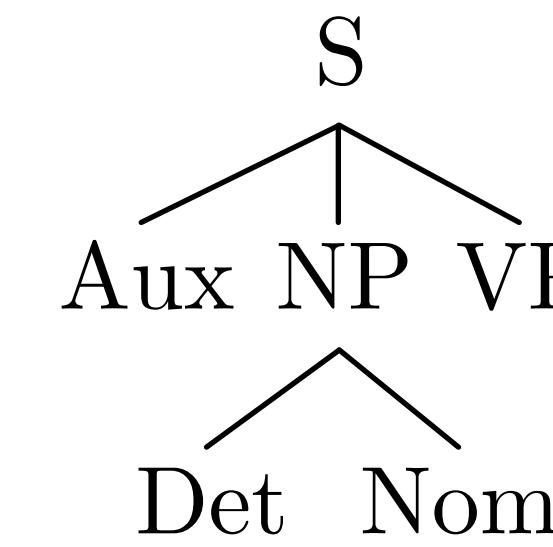
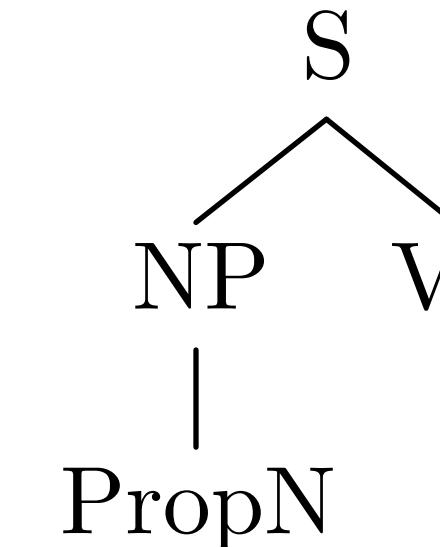
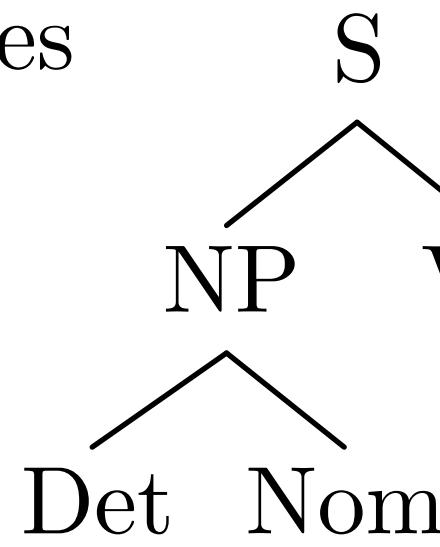
Start State

S

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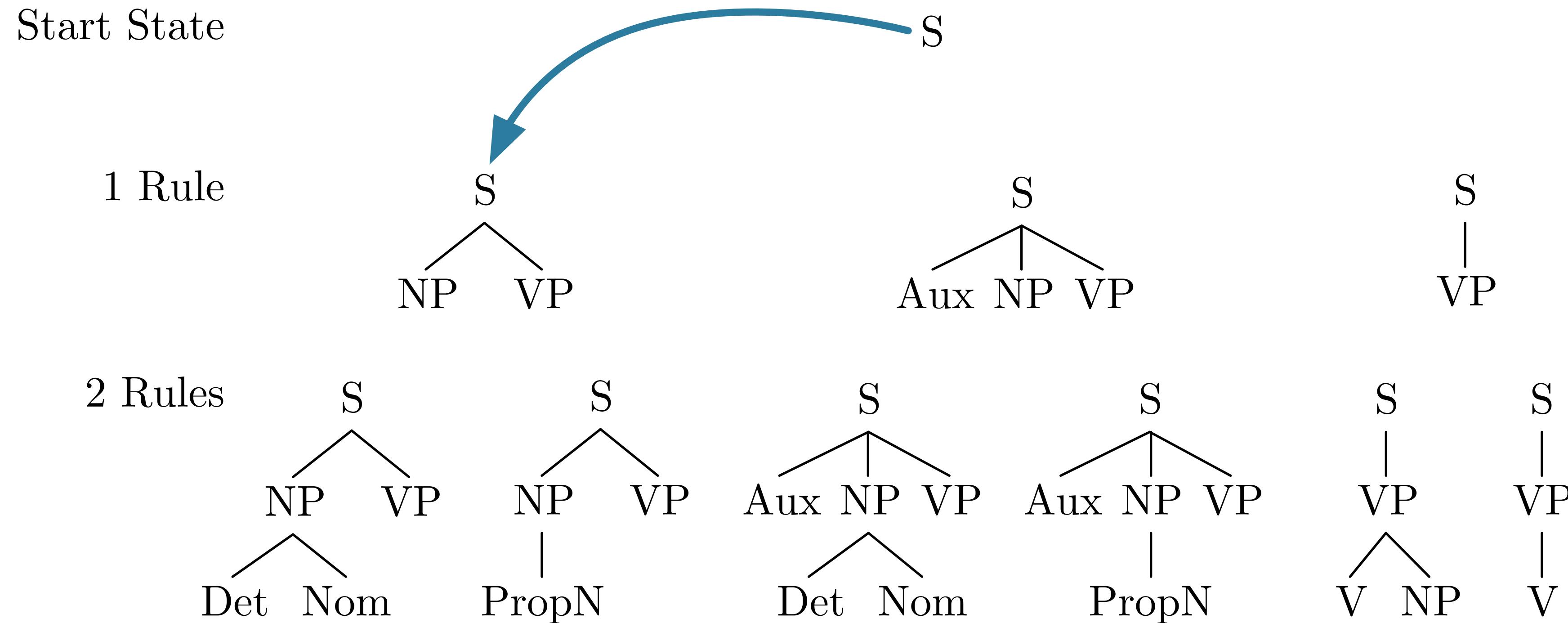


2 Rules



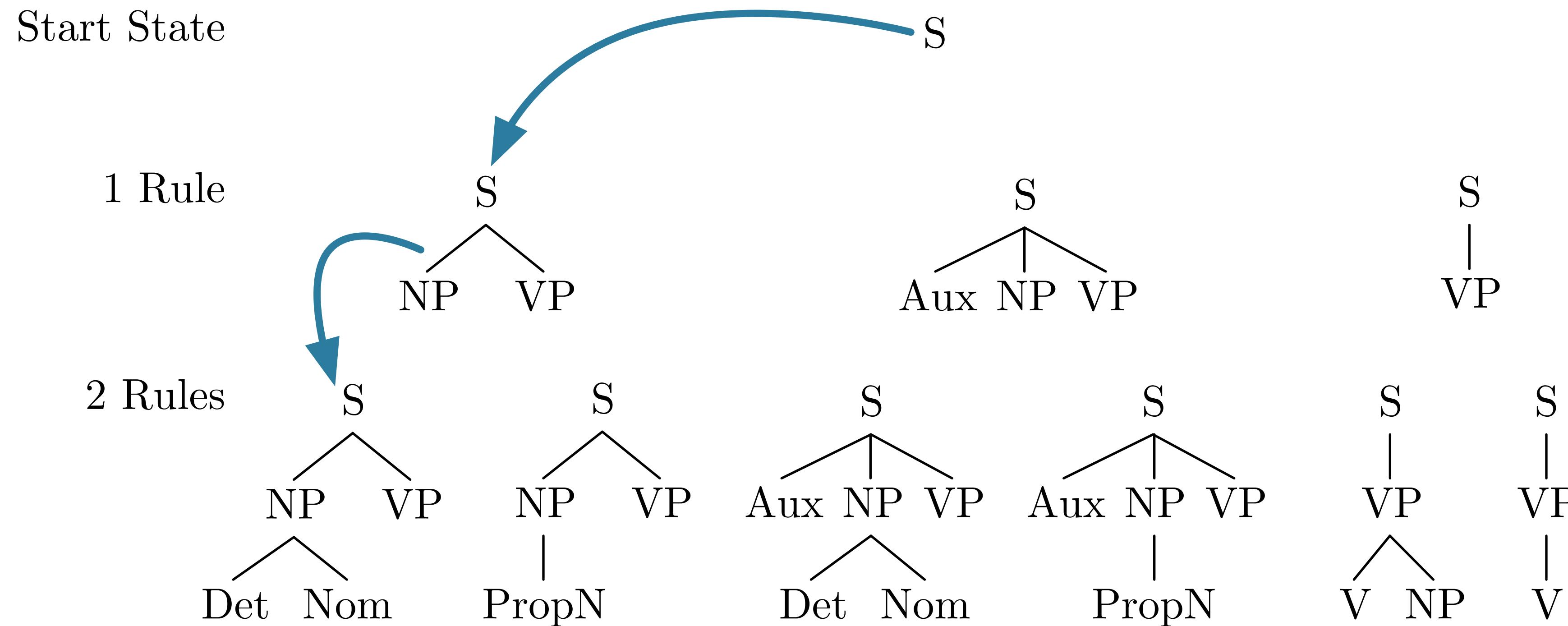
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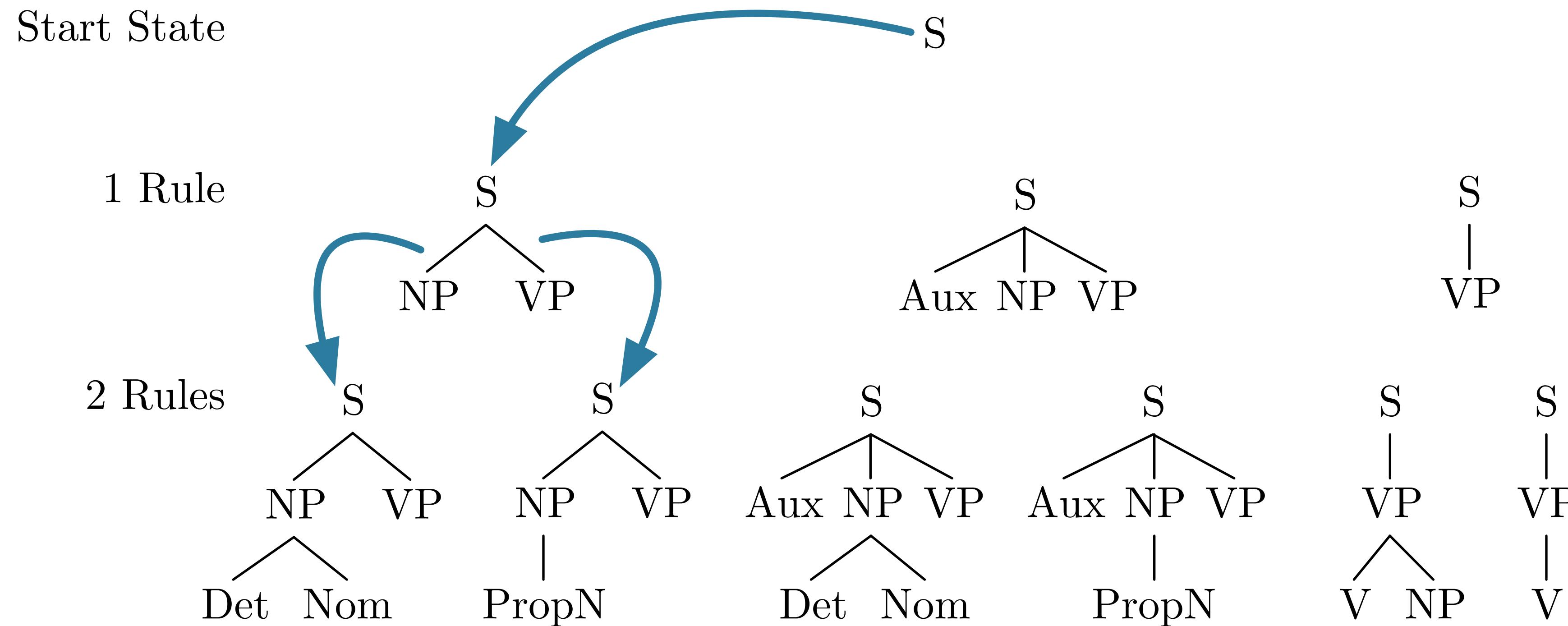
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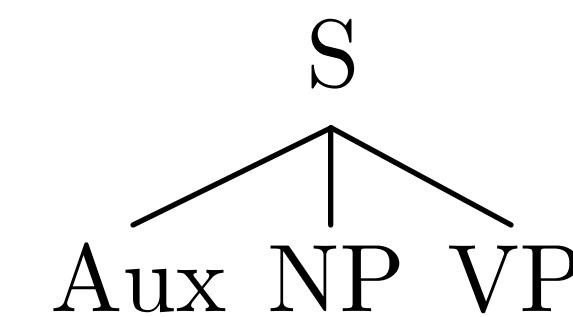
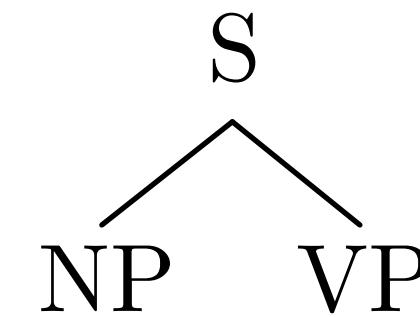
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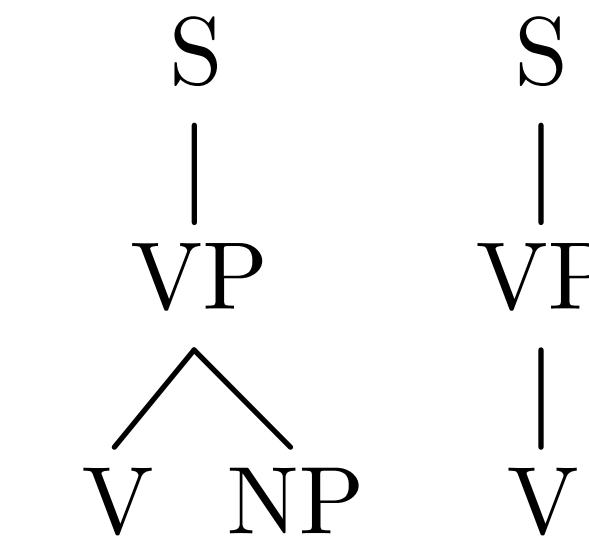
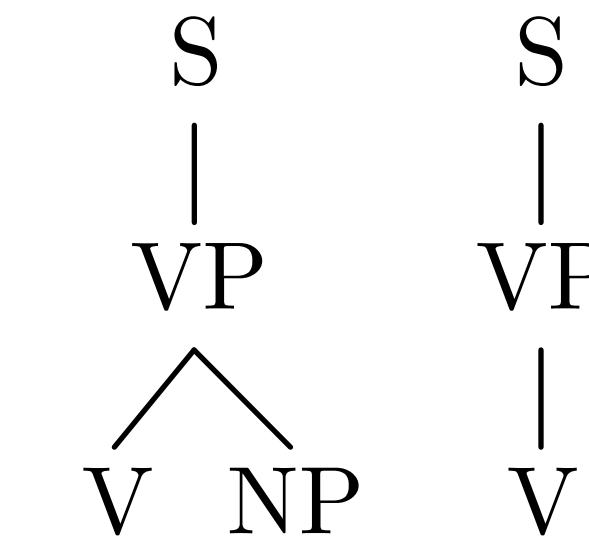
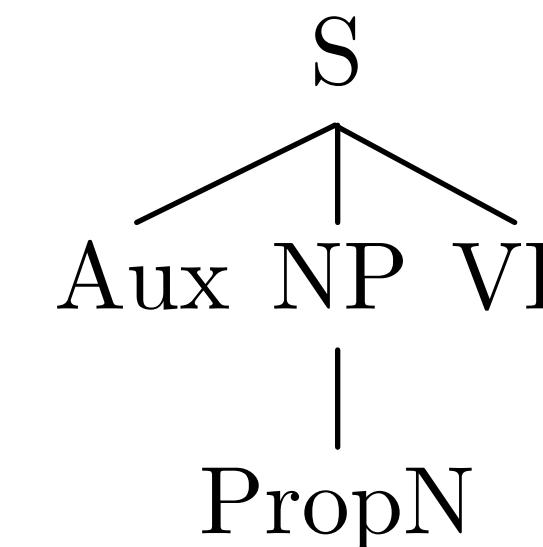
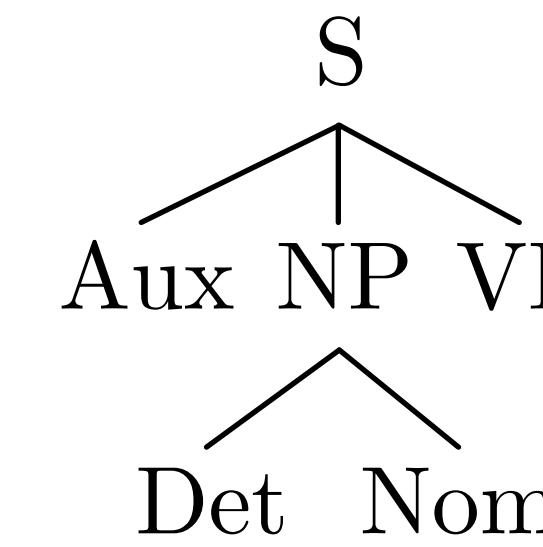
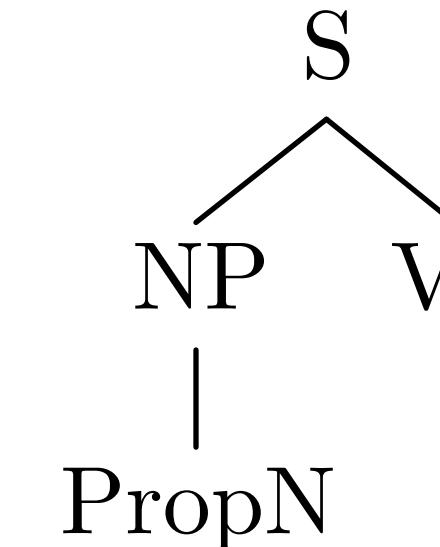
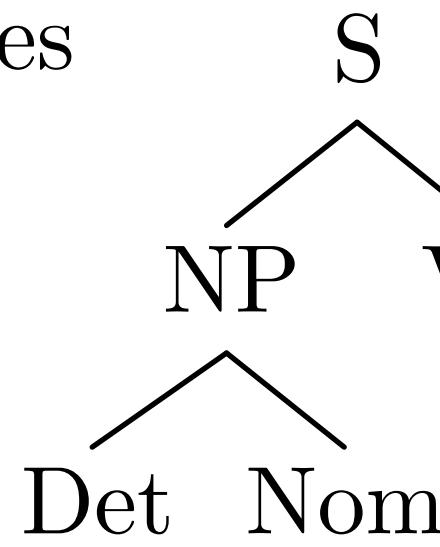
Start State

S

1 Rule



2 Rules

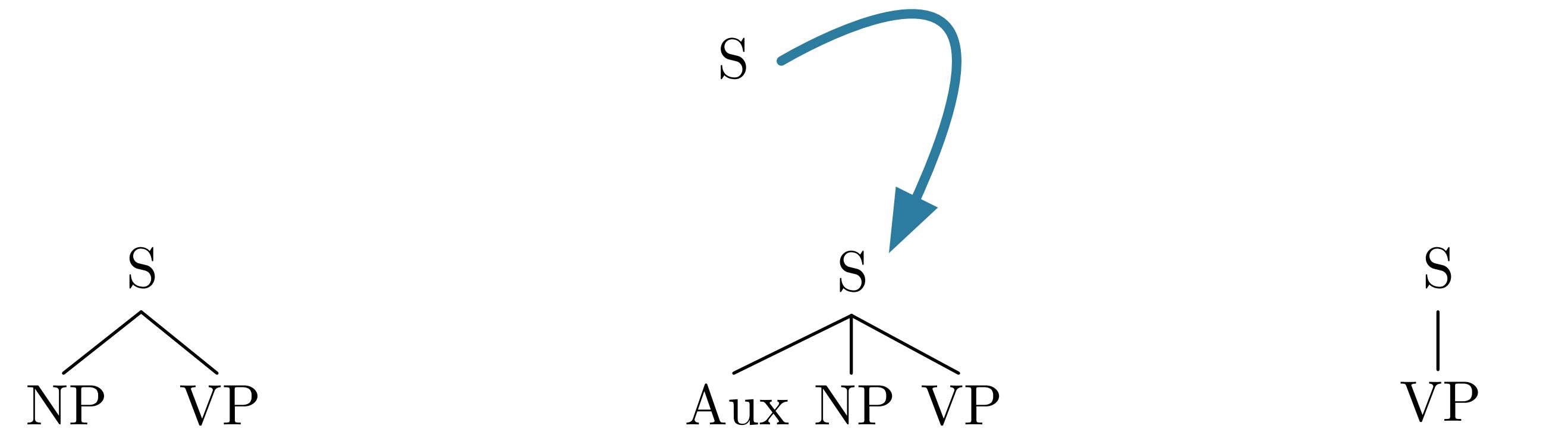


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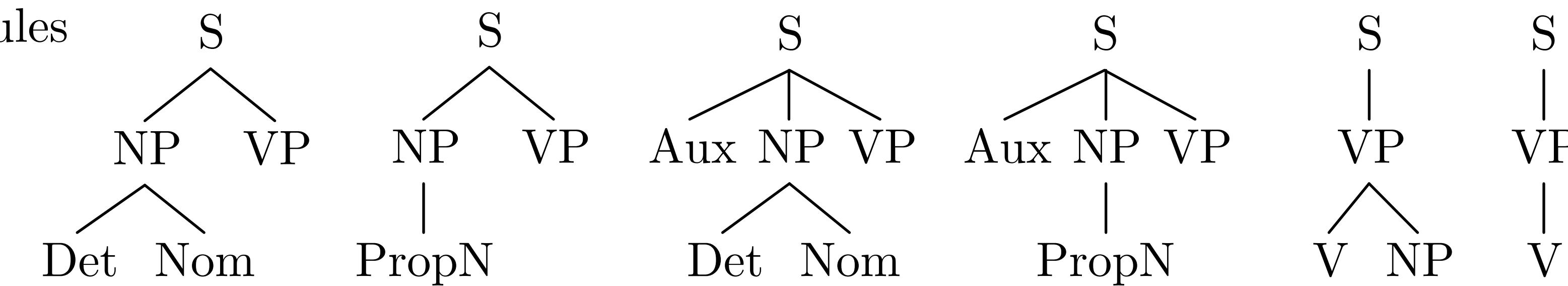
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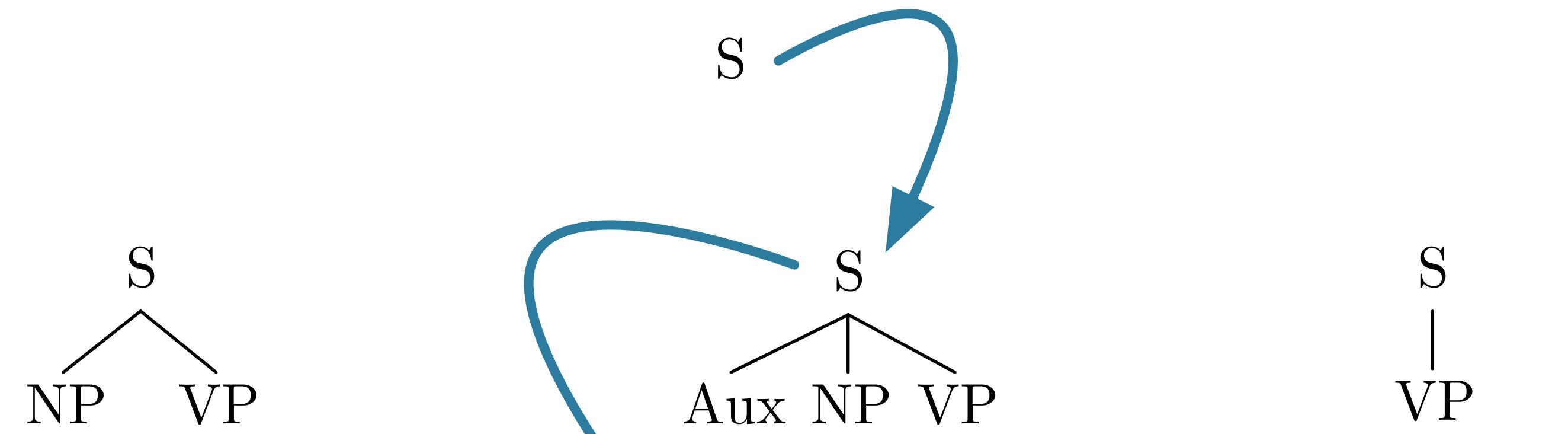


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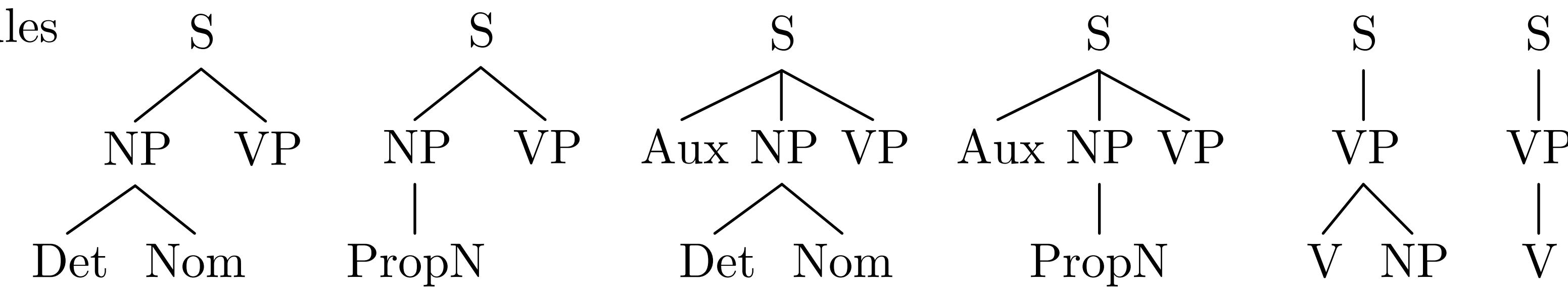
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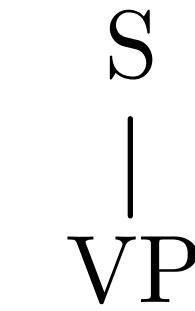
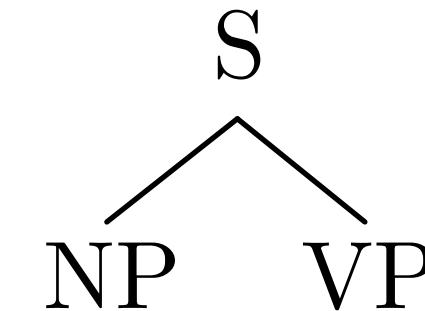


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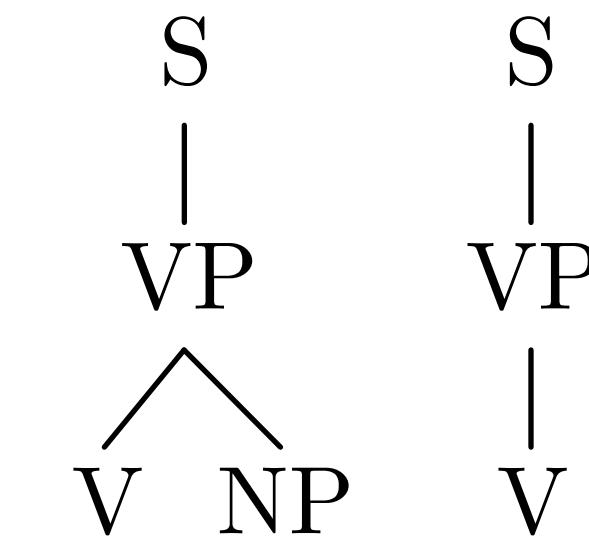
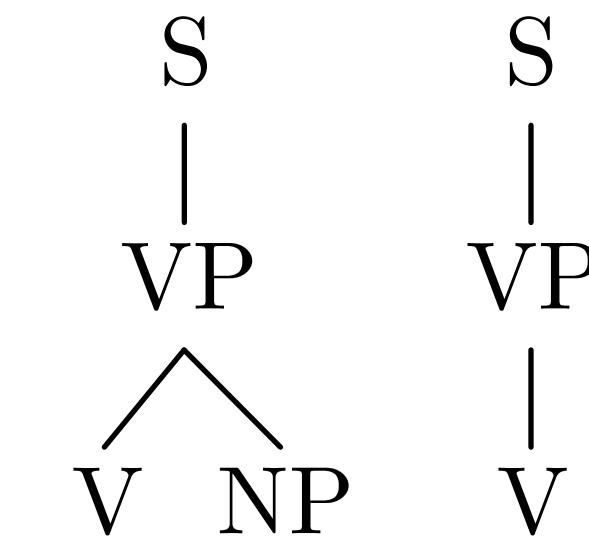
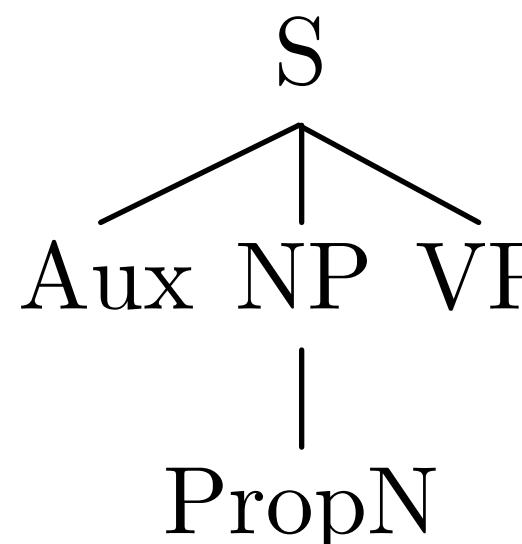
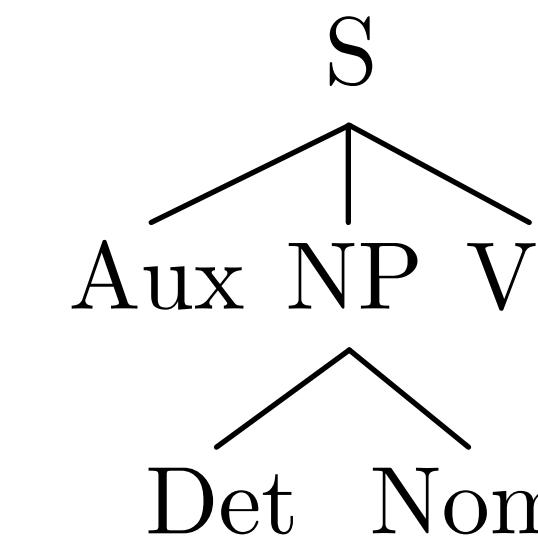
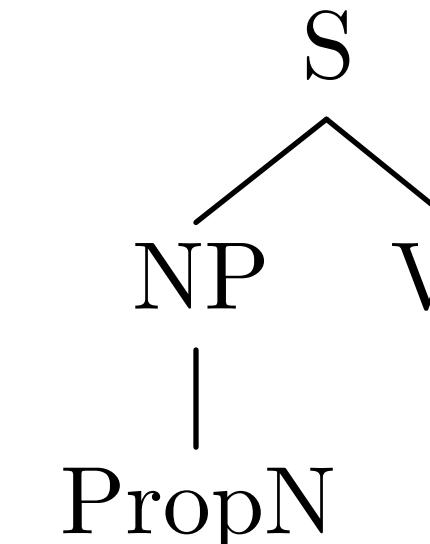
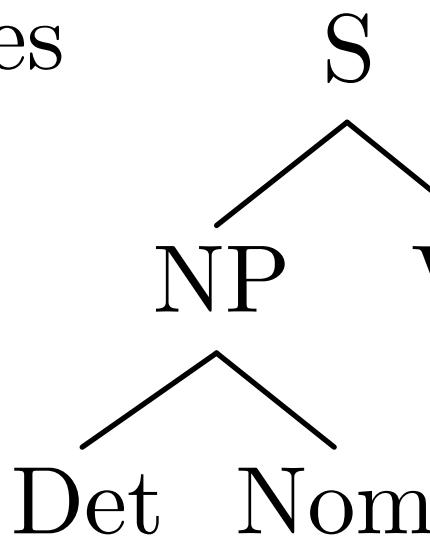
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2 Rules



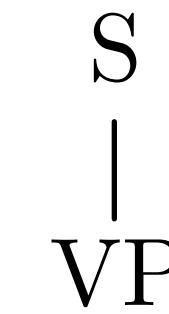
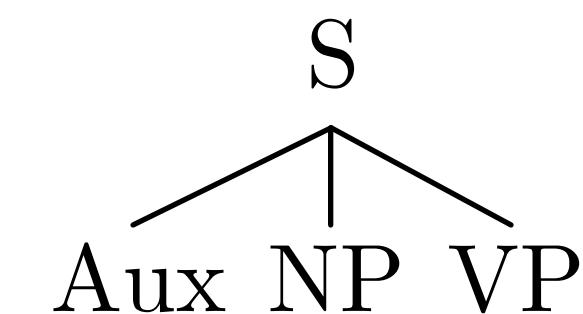
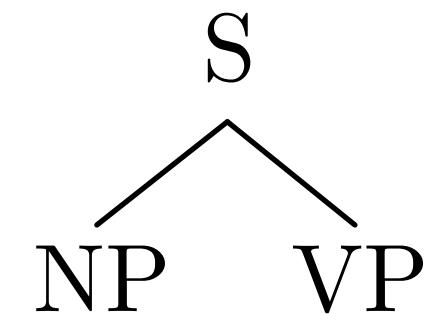
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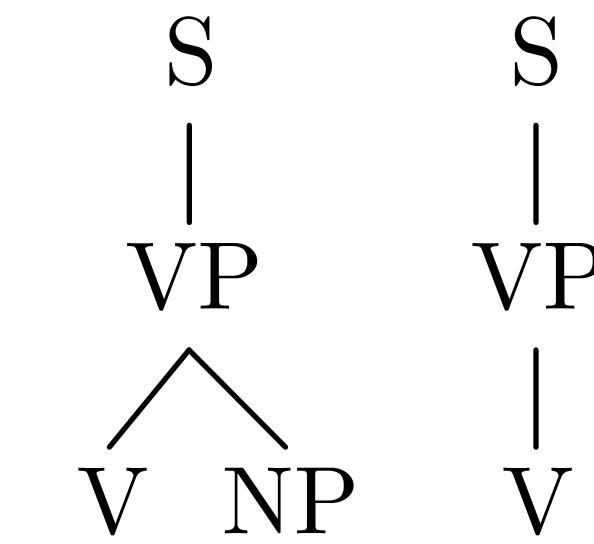
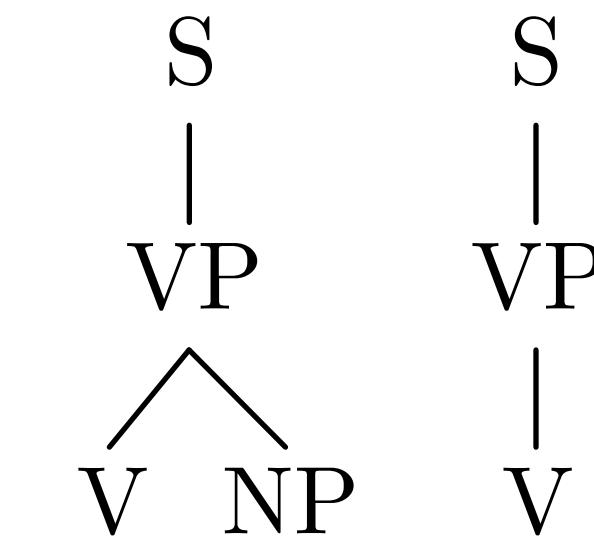
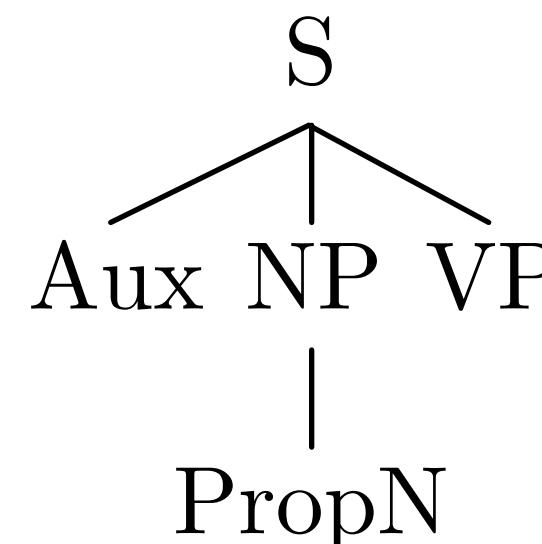
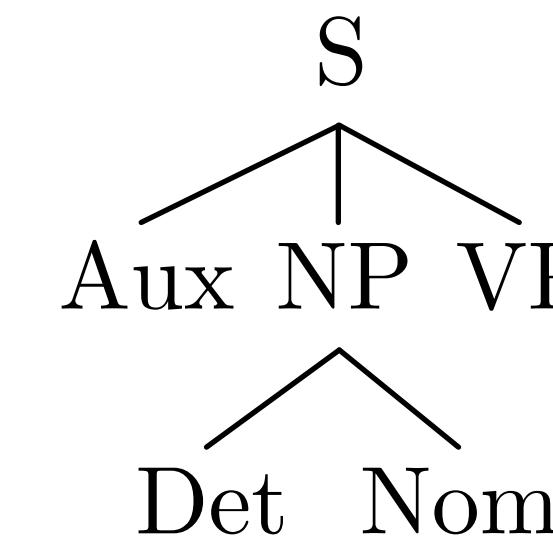
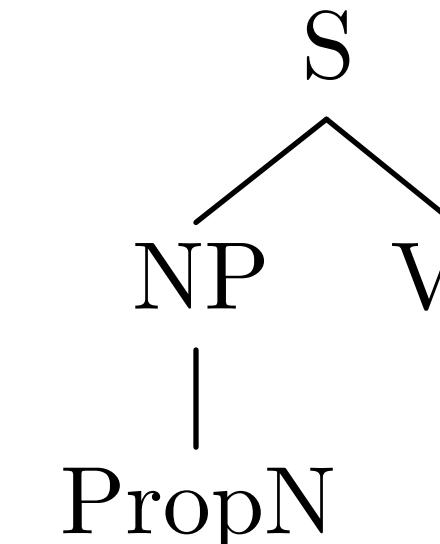
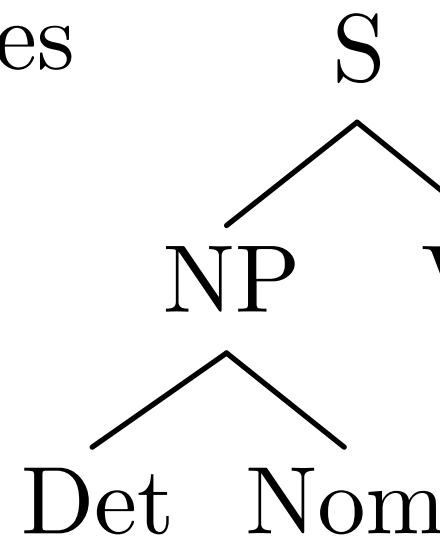
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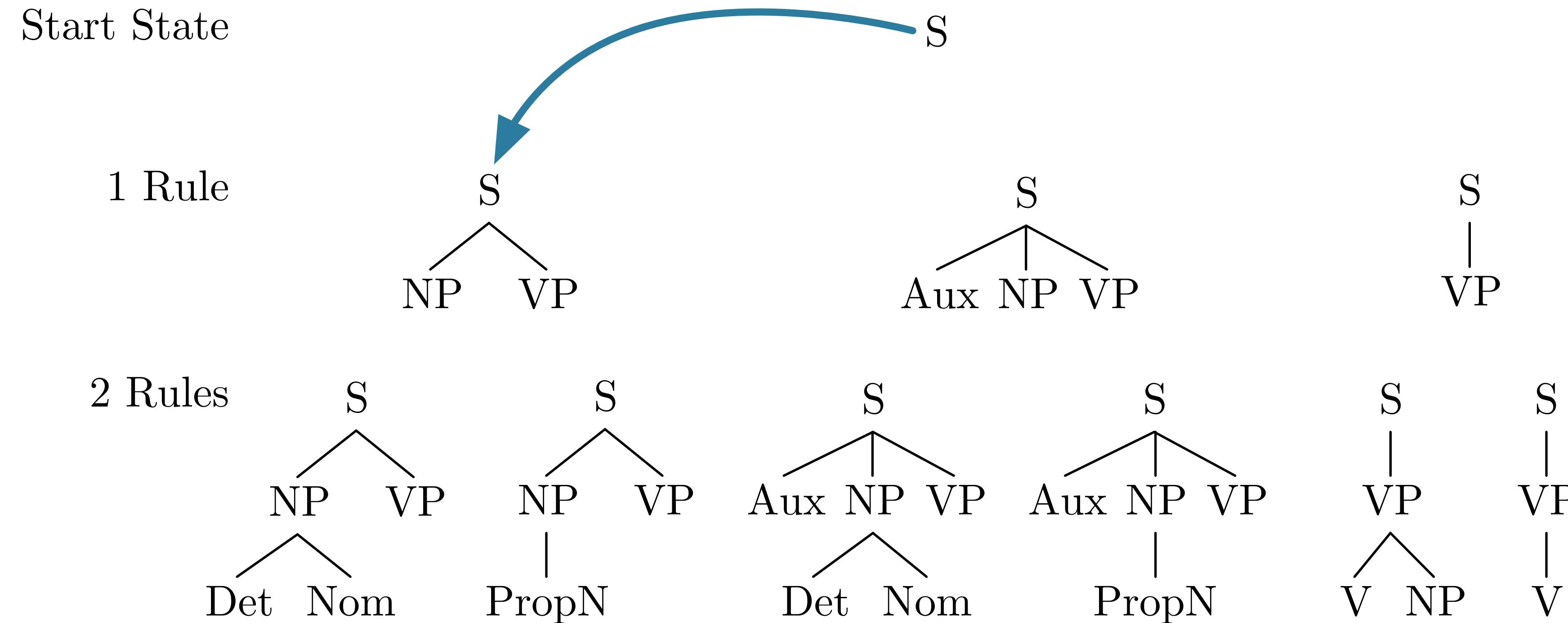


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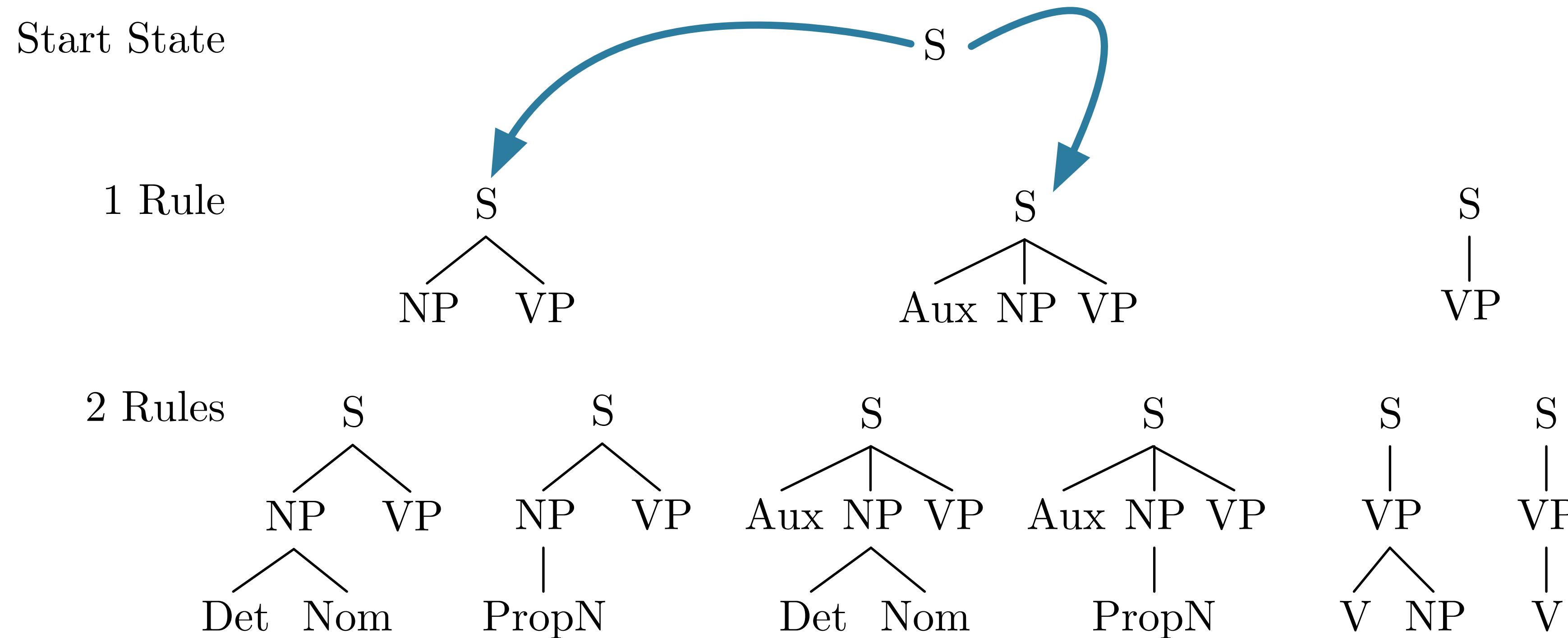
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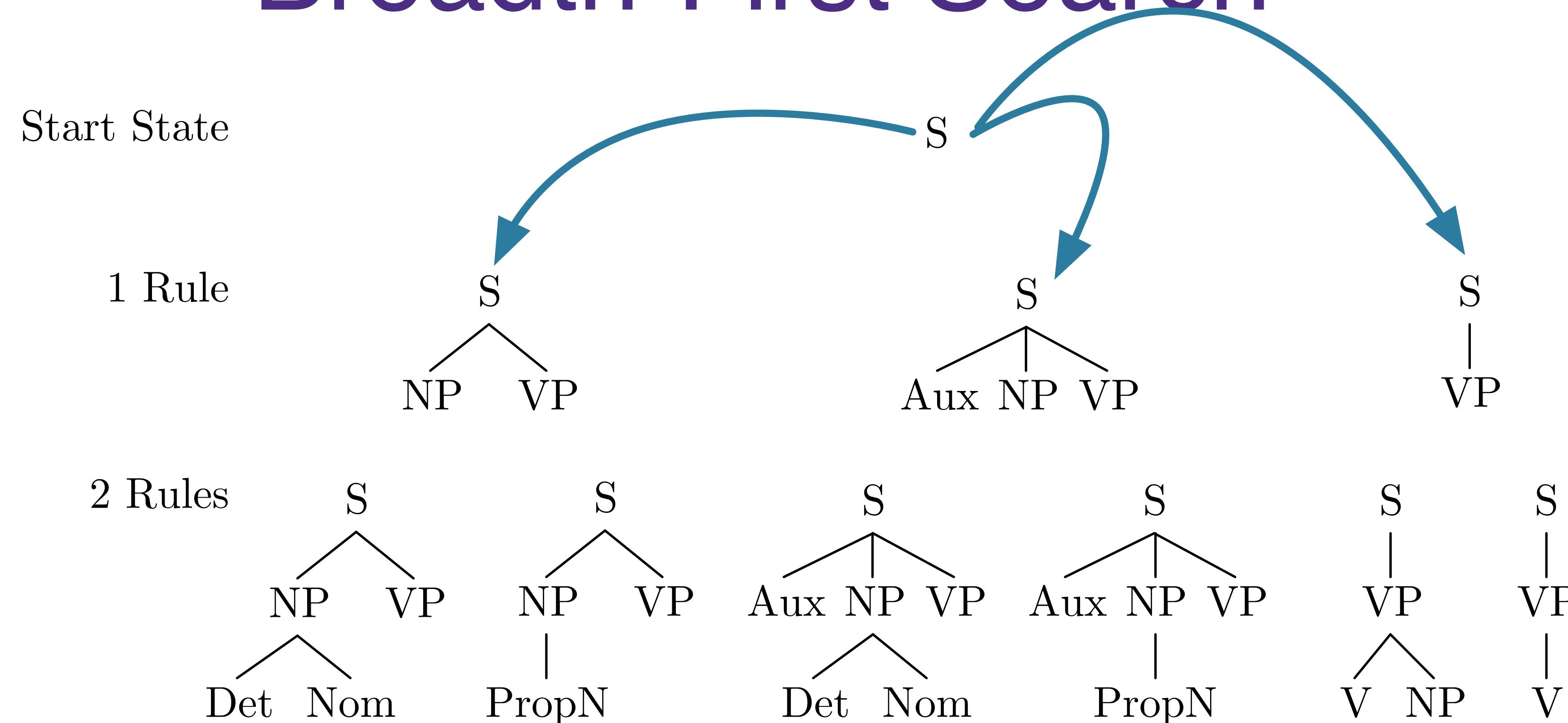


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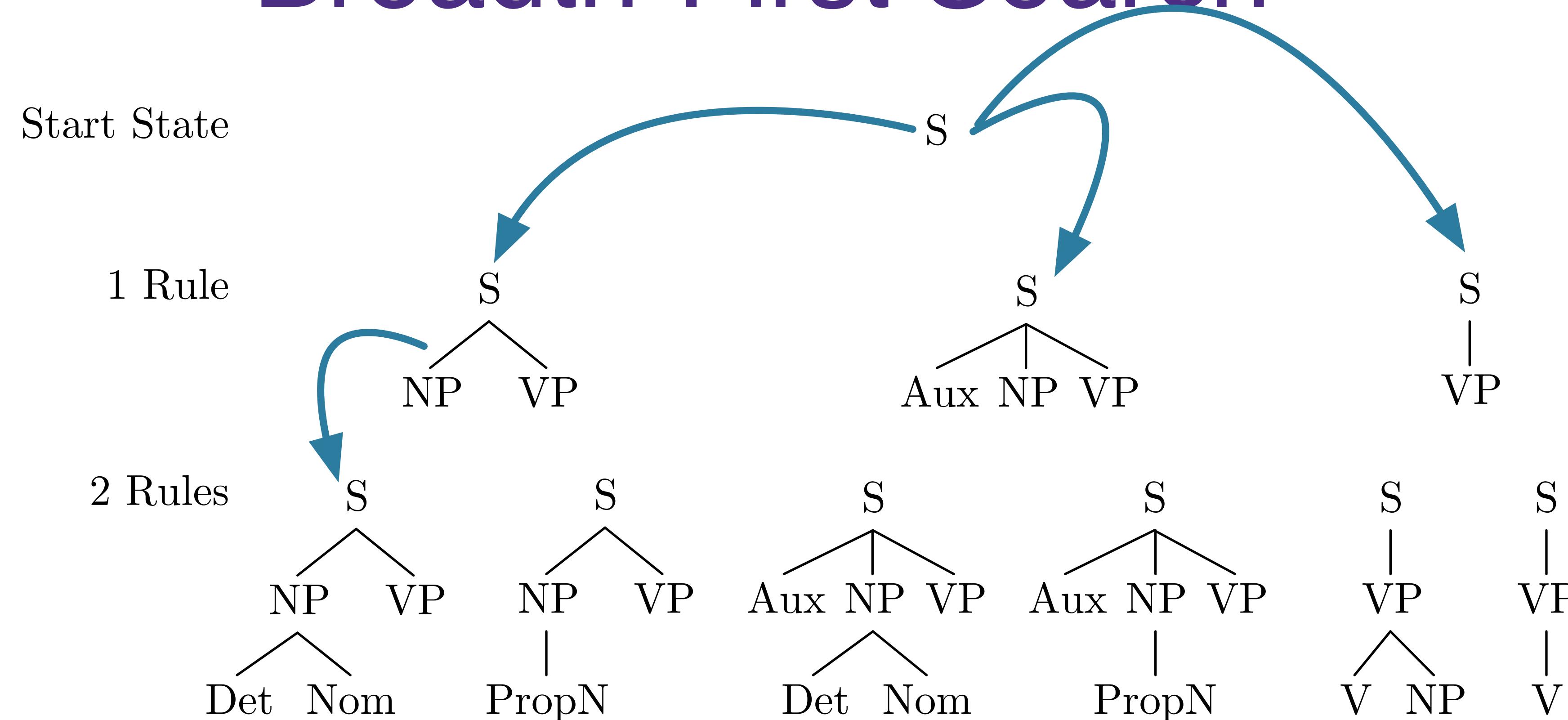


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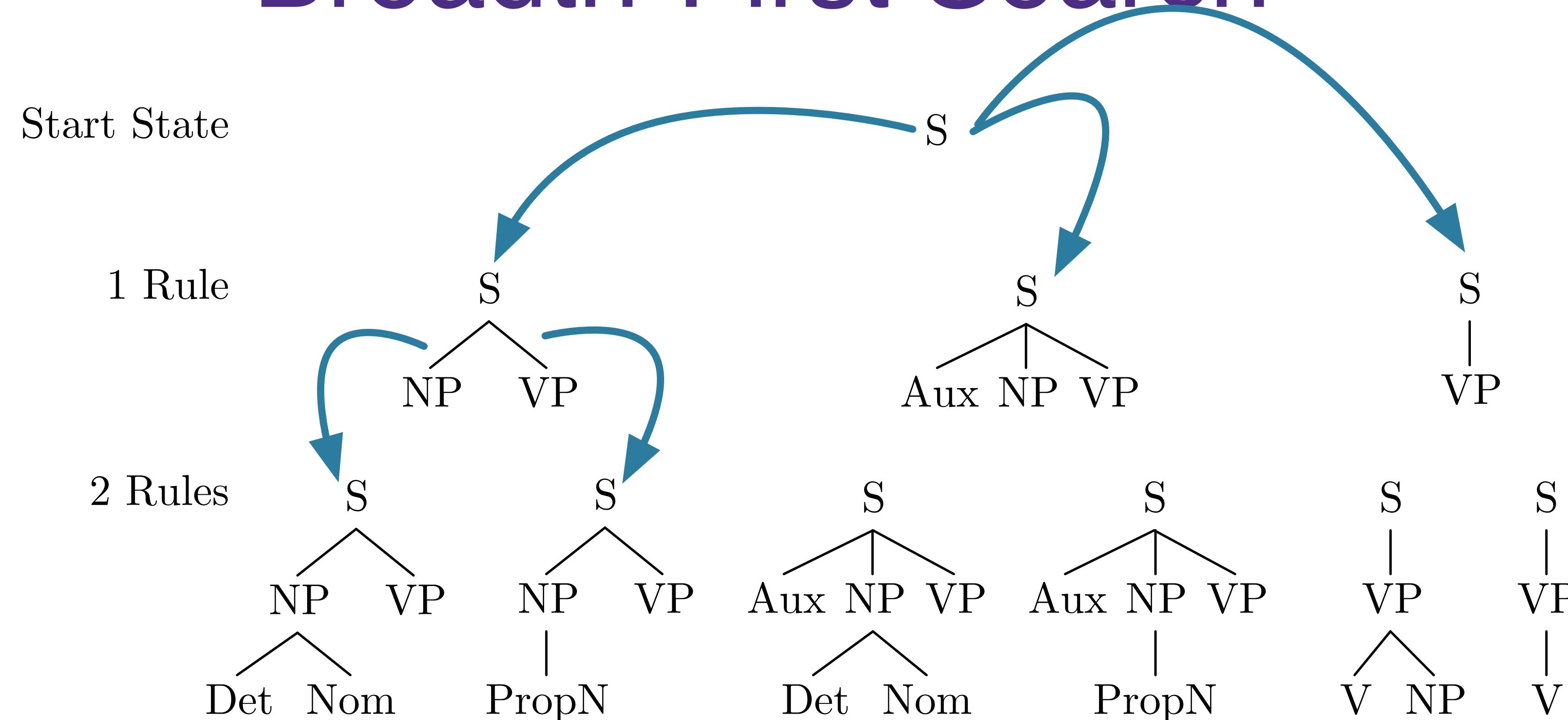
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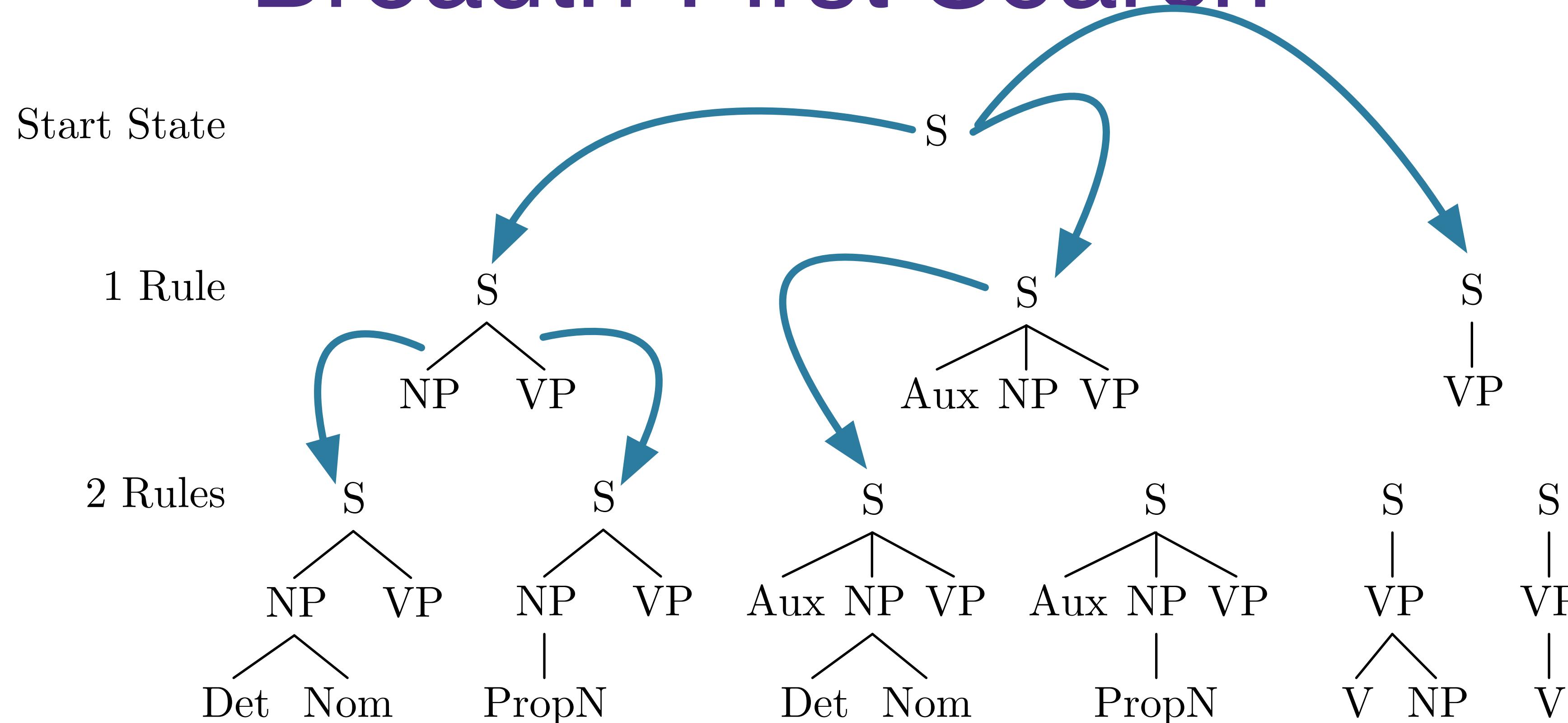
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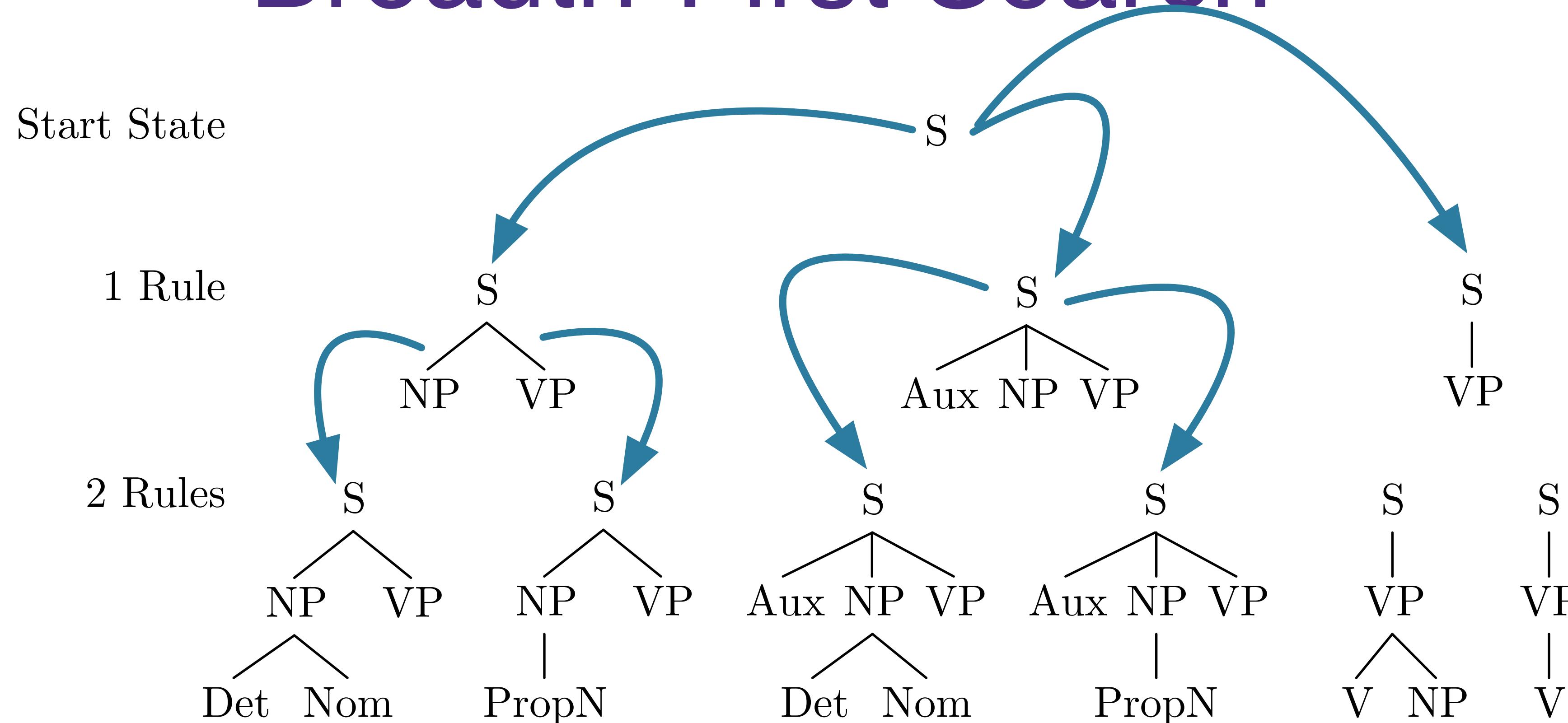
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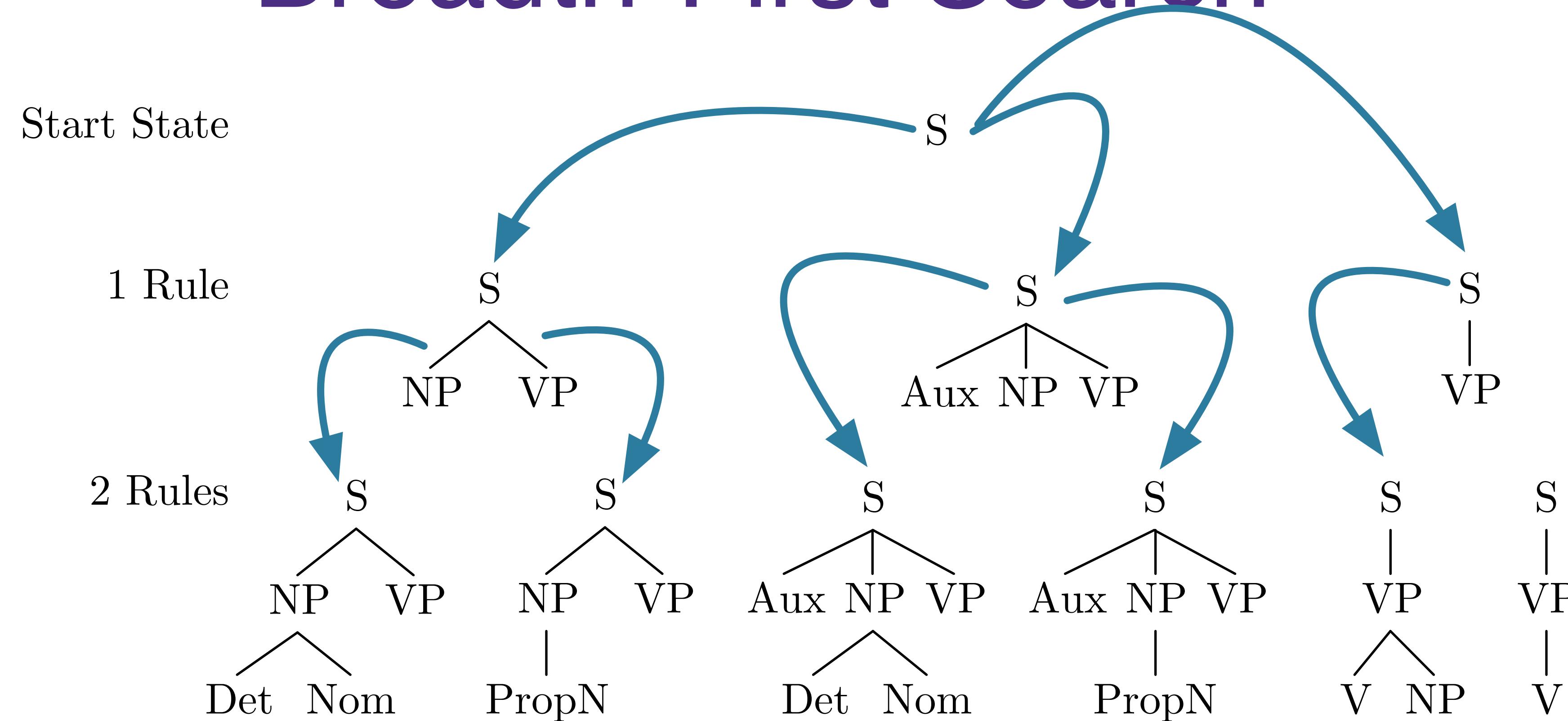
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# Breadth-First Search



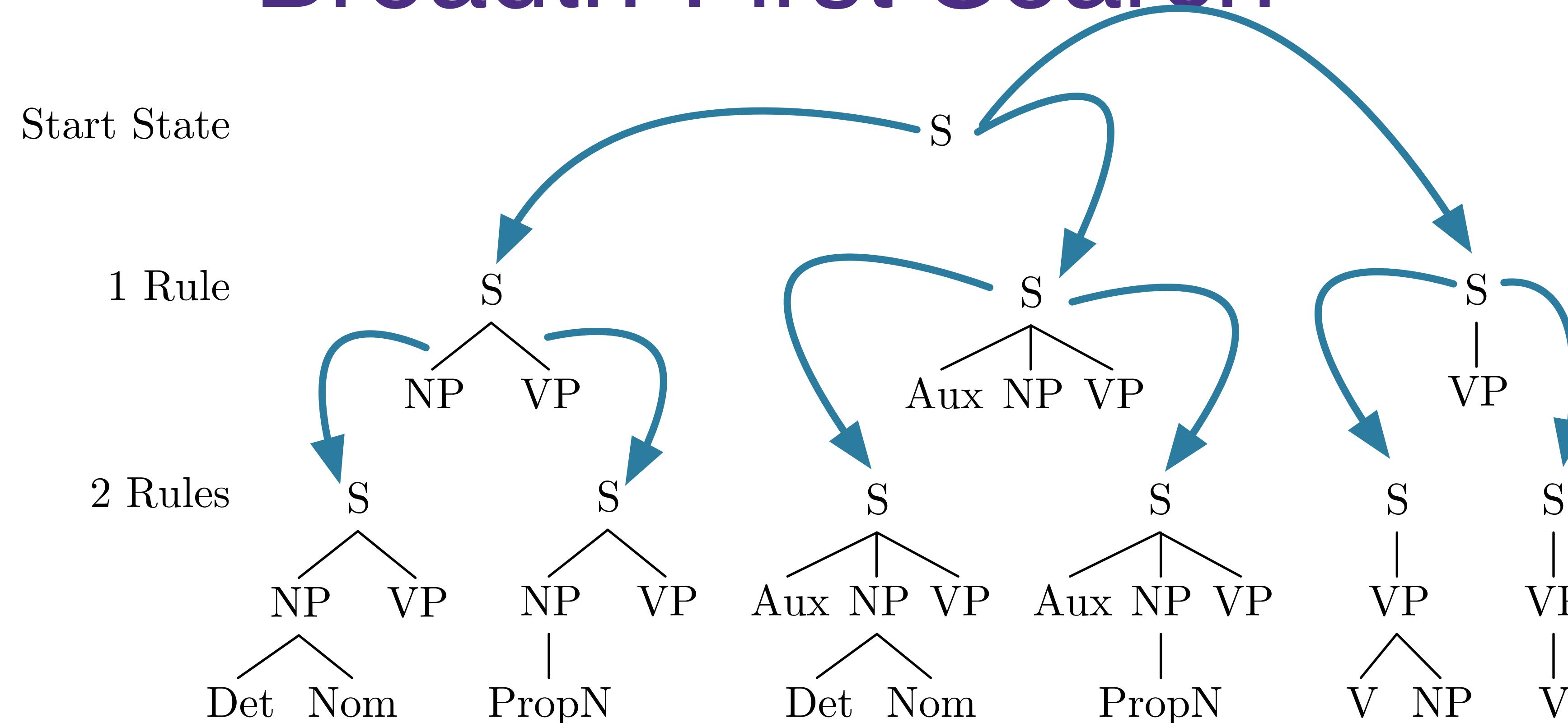
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# Breadth-First Search



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# Breadth-First Search



Book that flight.

# Pros and Cons of Top-down Parsing

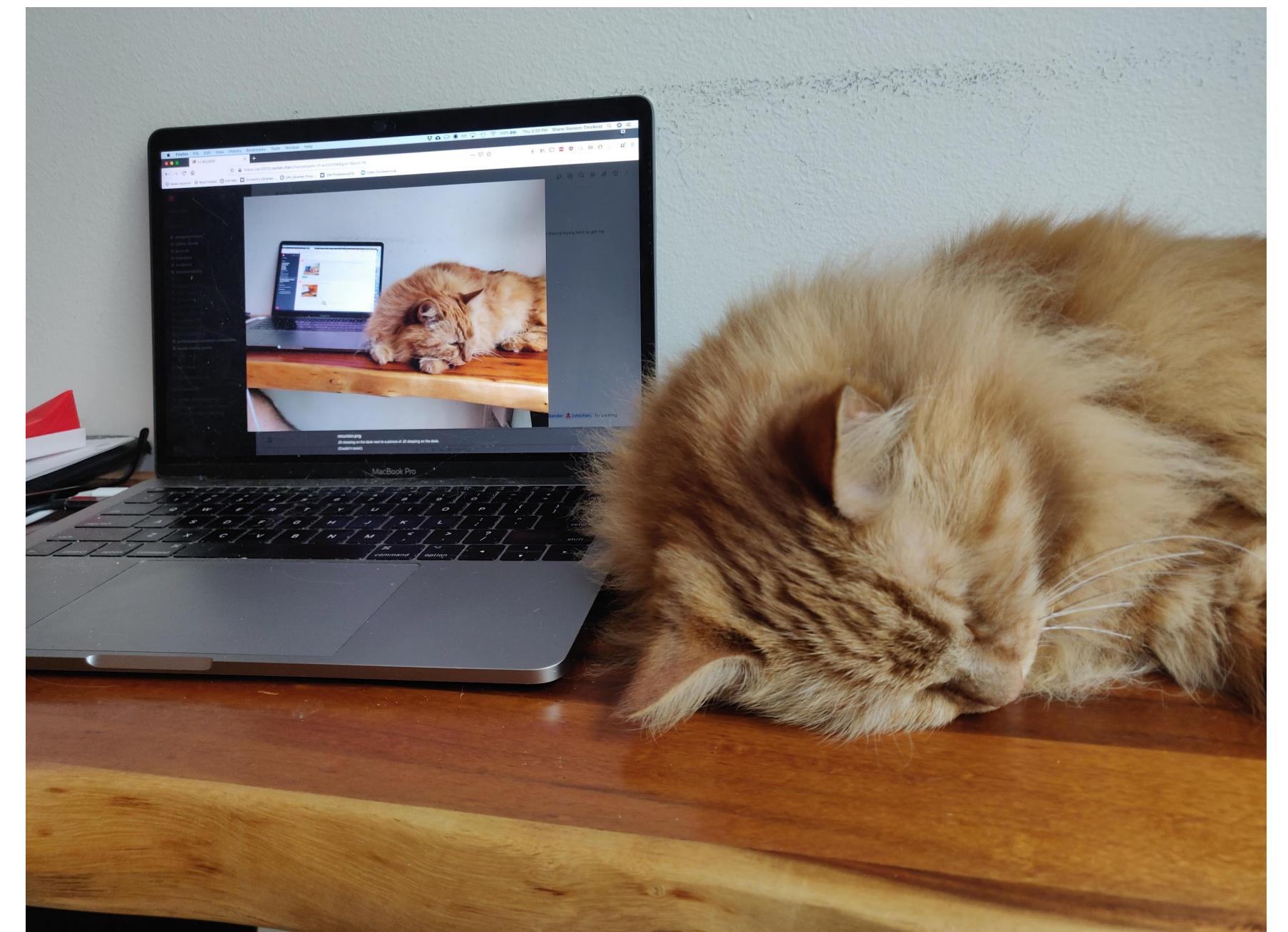
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# Bottom-Up Parsing

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- Try to find all trees that span the input
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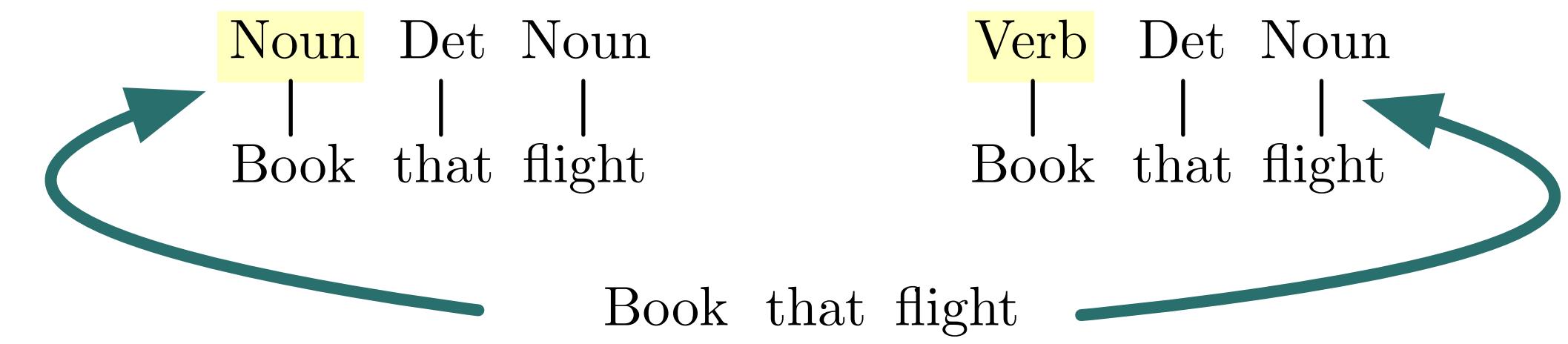
# Bottom-Up Parsing

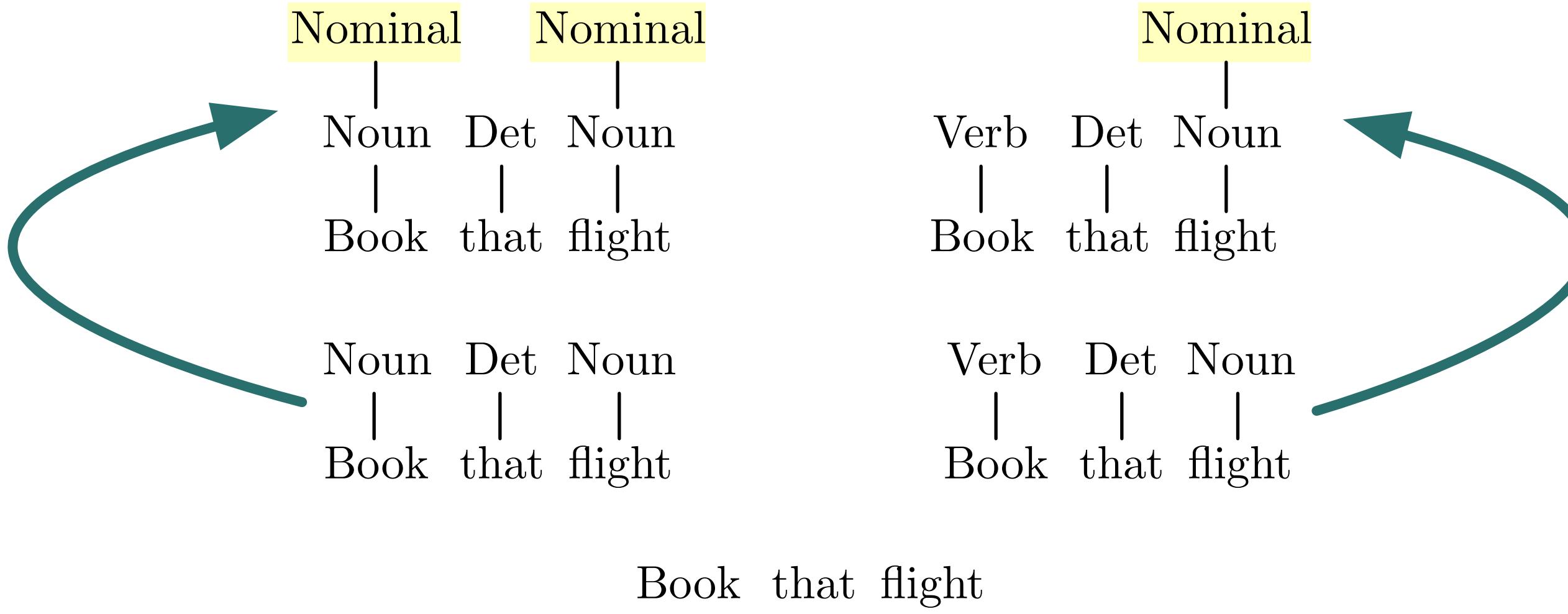
- Try to find all trees that span the input
  - Start with input string
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  - Use all productions with current subtree(s) on RHS
    - e.g.  $N \rightarrow \text{Book}$ ;  $V \rightarrow \text{Book}$

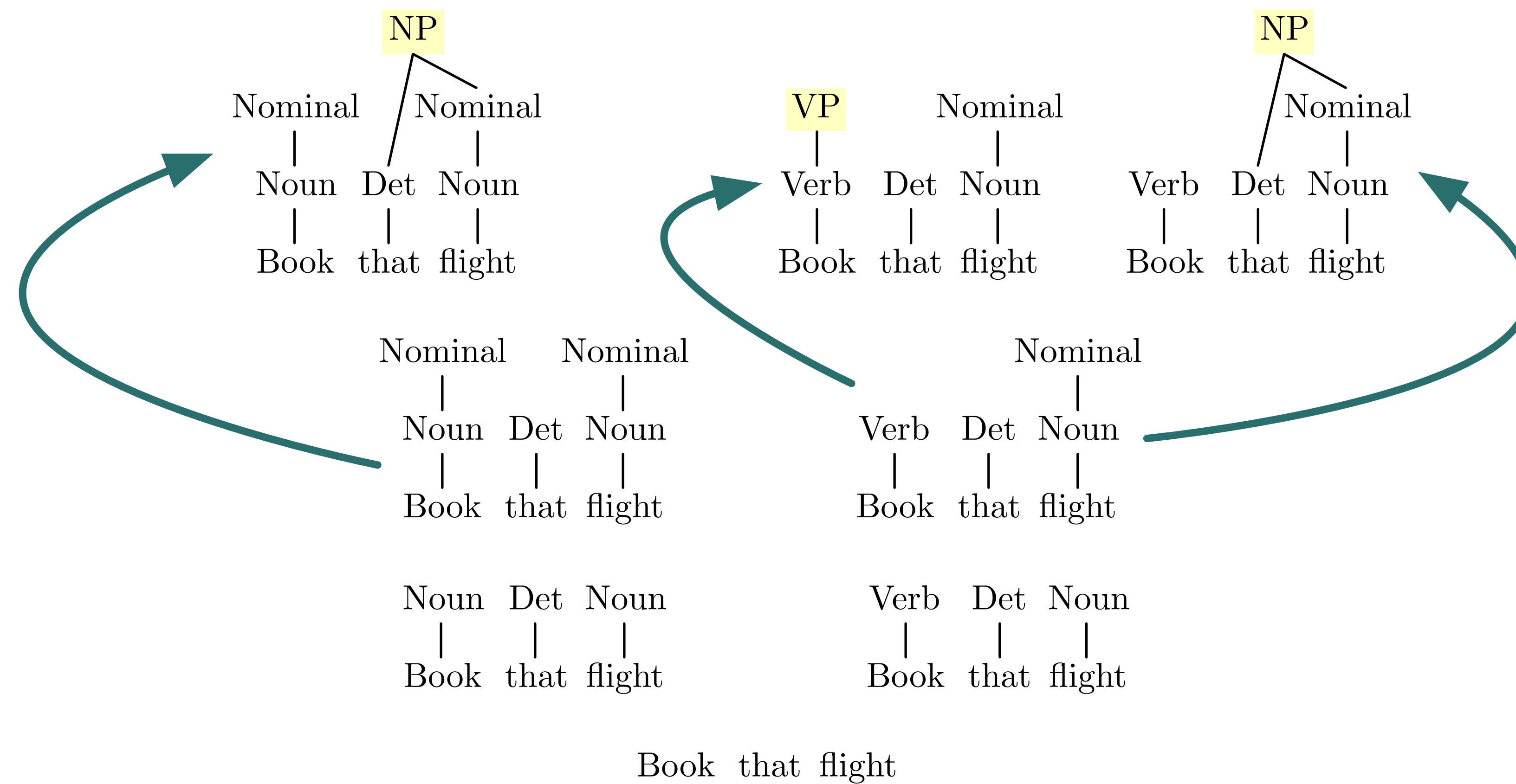
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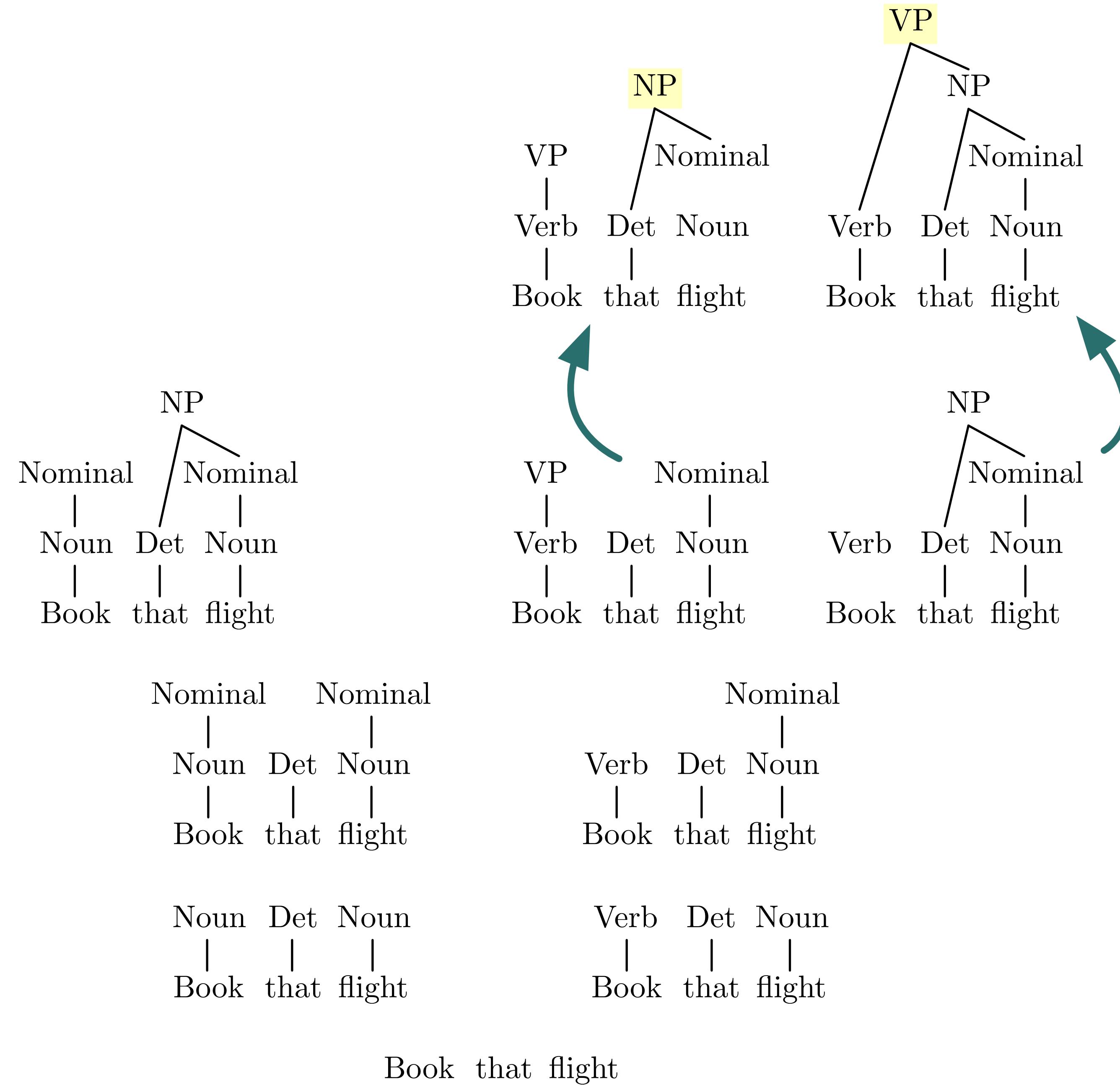
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  - Use all productions with current subtree(s) on RHS
    - e.g.  $N \rightarrow \text{Book}$ ;  $V \rightarrow \text{Book}$
  - Stop when spanned by S, or no more rules apply

Book that flight









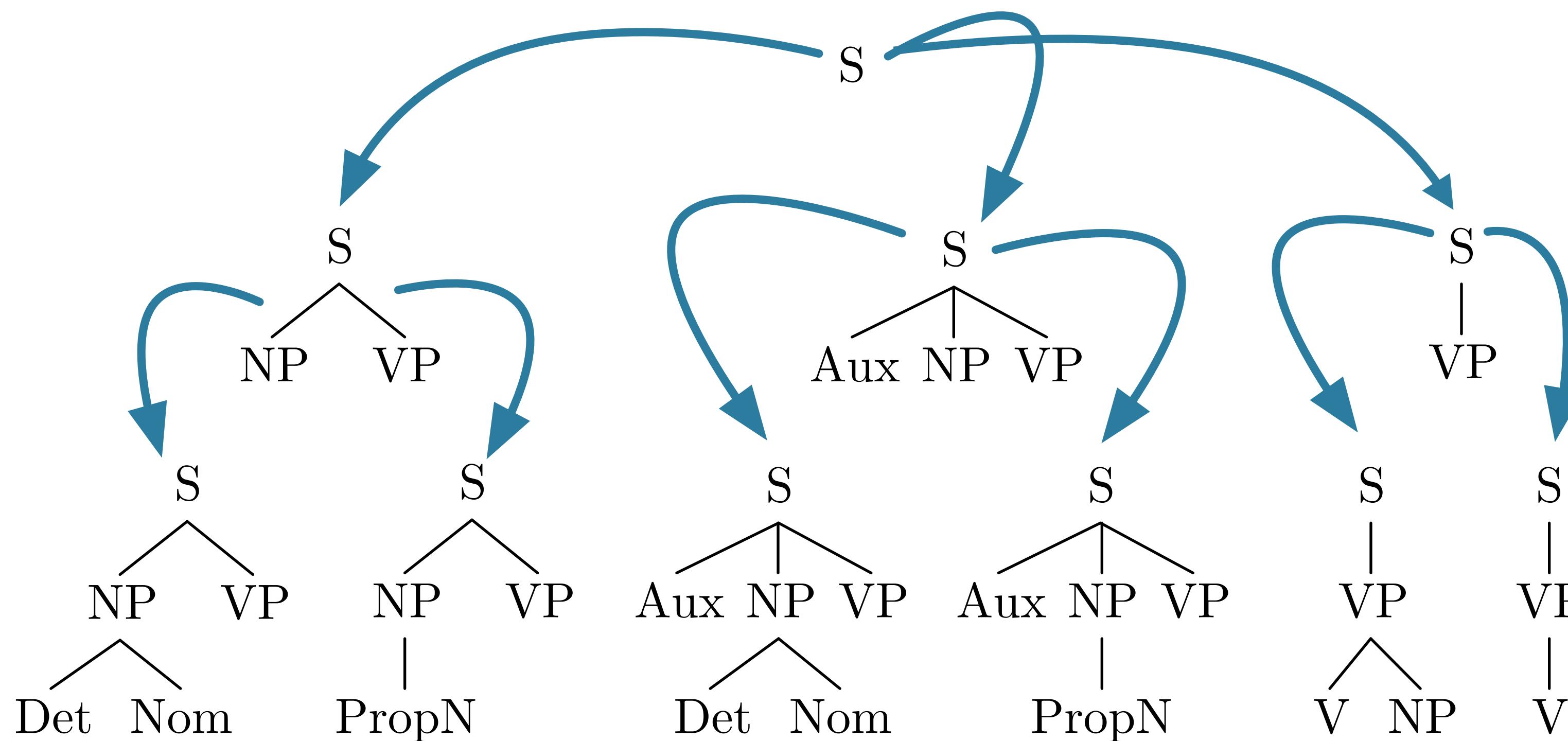
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- Pros:
  - Will not explore trees that don't match input
  - Recursive rules less problematic
  - Useful for incremental/fragment parsing

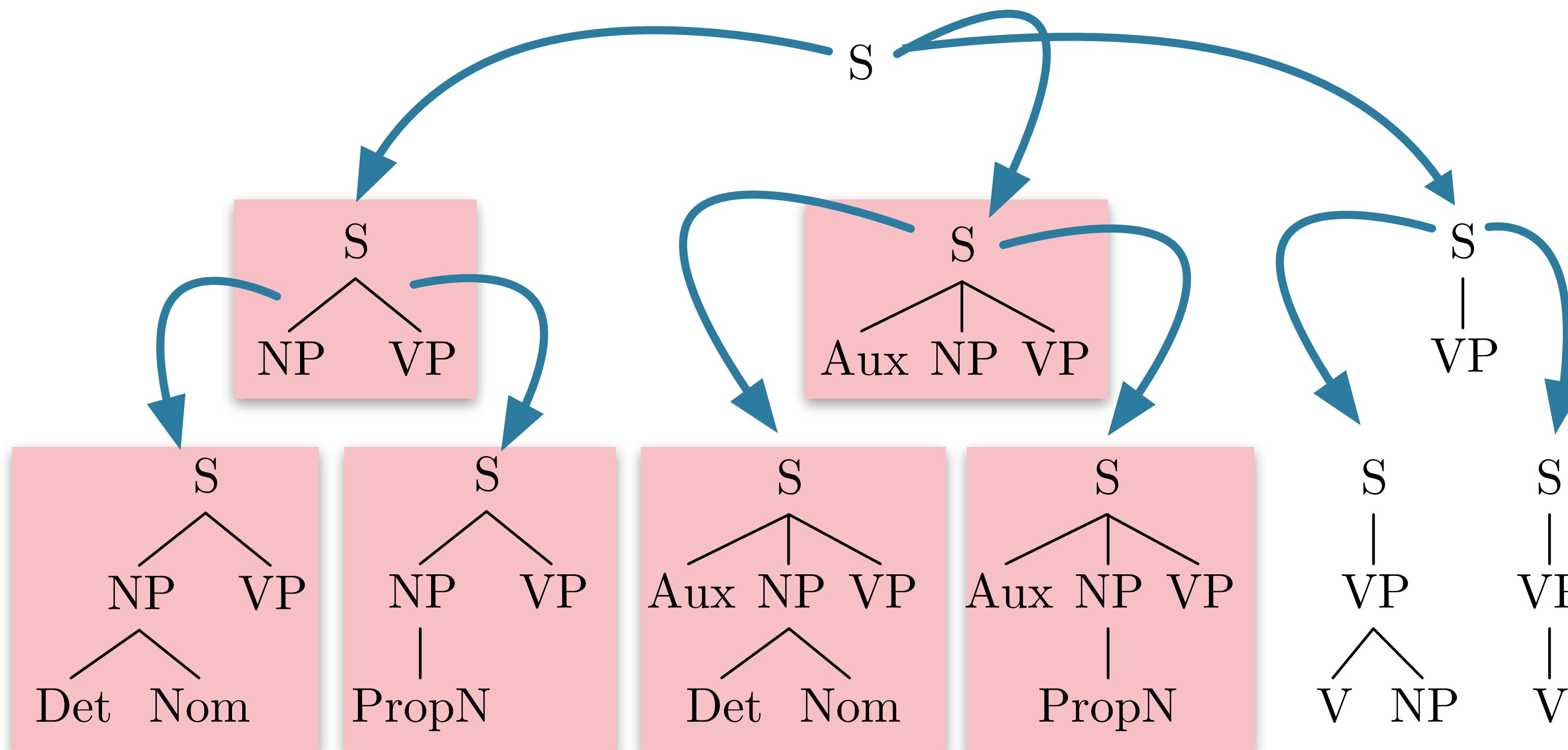
# Pros and Cons of Bottom-Up Search

- Pros:
  - Will not explore trees that don't match input
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  - Useful for incremental/fragment parsing
- Cons:
  - Explore subtrees that will not fit full input

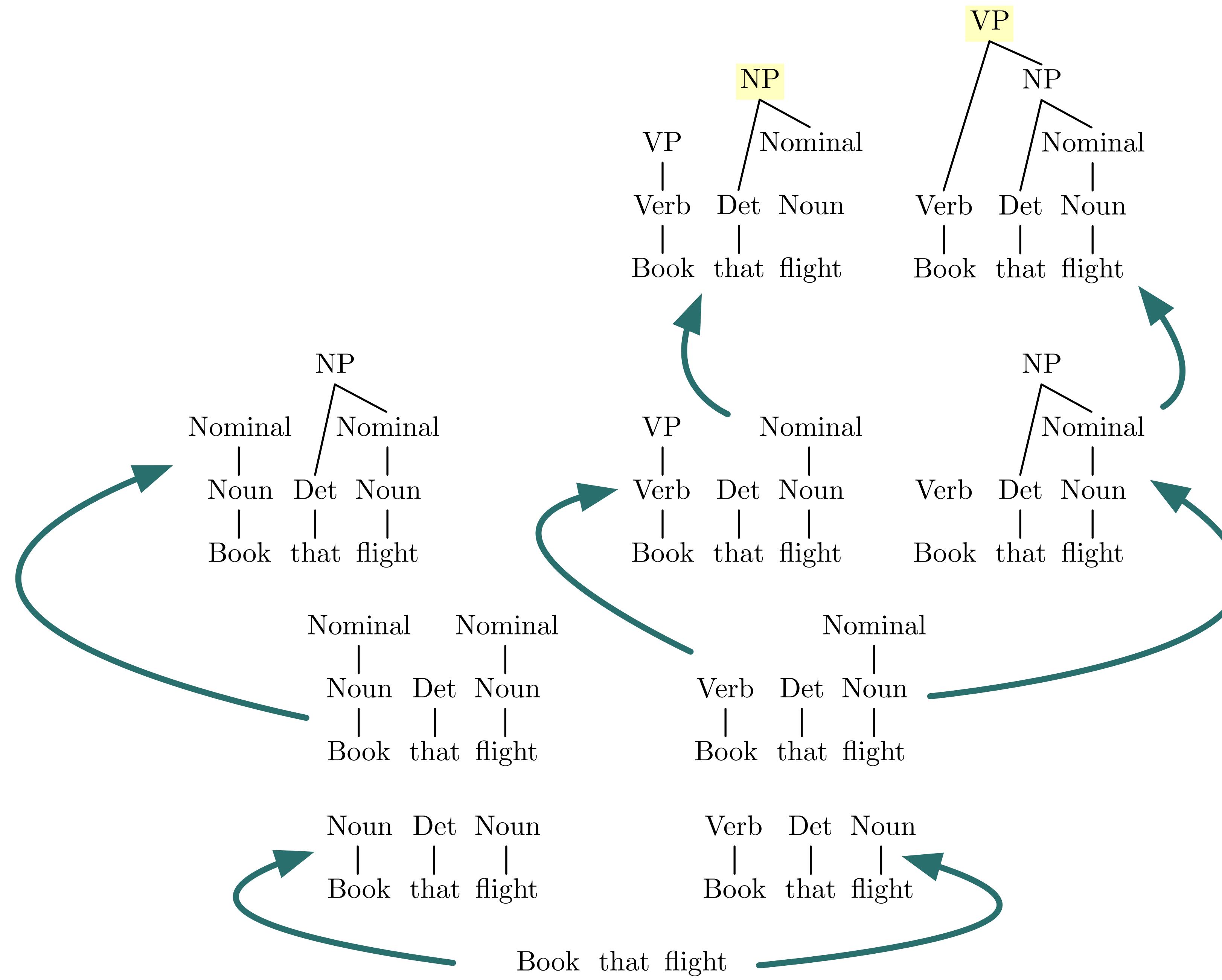
# Recap: Parsing as Search



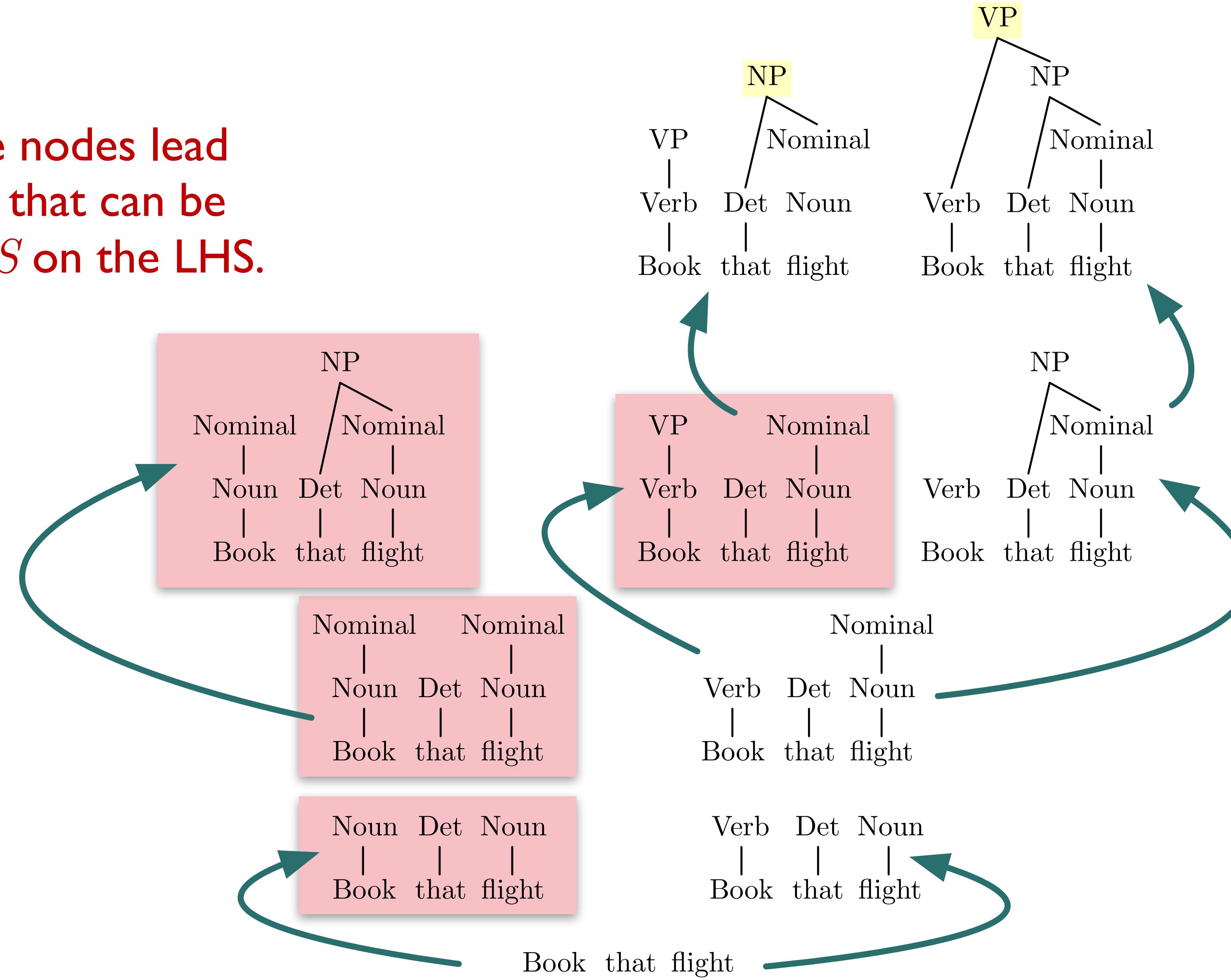
# Recap: Parsing as Search



None of these nodes can produce *book* as first terminal



None of these nodes lead  
lead to a RHS that can be  
combined with  $S$  on the LHS.



# Parsing Challenges

- Parsing-as-Search
- **Parsing Challenges**
  - Ambiguity
  - Repeated Substructure
  - Recursion
- Strategy: Dynamic Programming
- Grammar Equivalence
- CKY parsing algorithm

# Parsing Ambiguity

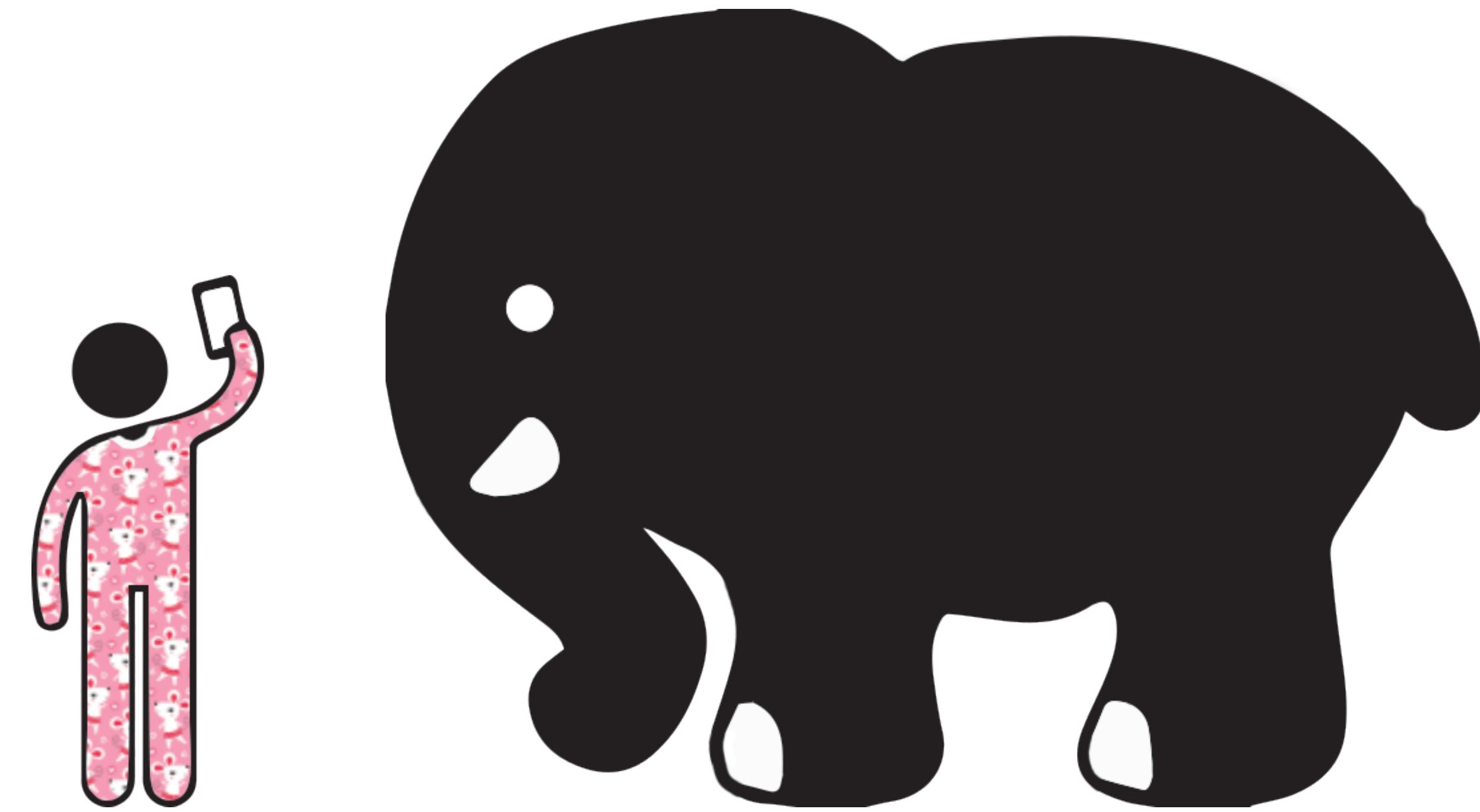
- **Lexical Ambiguity:**

- Book/NN → *I left a book on the table.*
- Book/VB → *Book that flight.*

- Structural Ambiguity

# Attachment Ambiguity

“One morning, I shot an elephant in my pajamas.

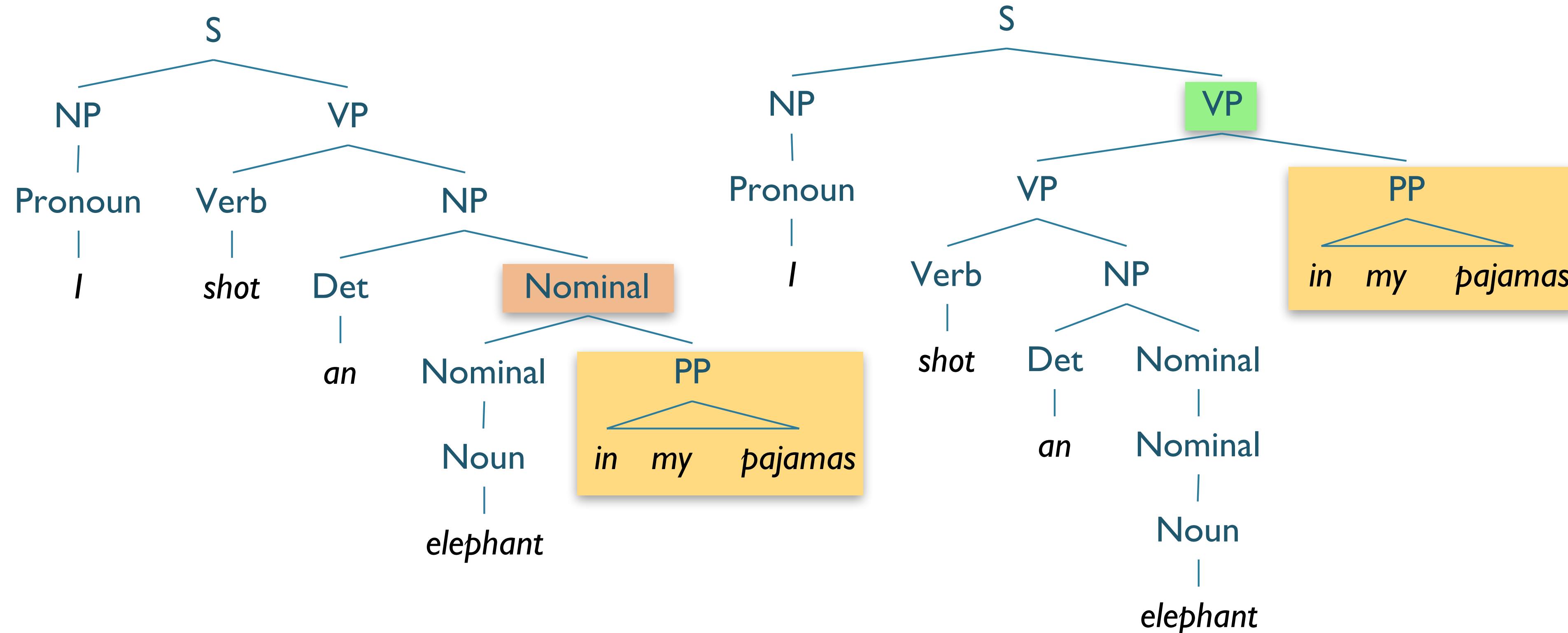


# Attachment Ambiguity

“One morning, I shot an elephant in my pajamas.  
How he got into my pajamas, I’ll never know.” — *Groucho Marx*



# Attachment Ambiguity



*“We saw the Eiffel Tower flying to Paris”*



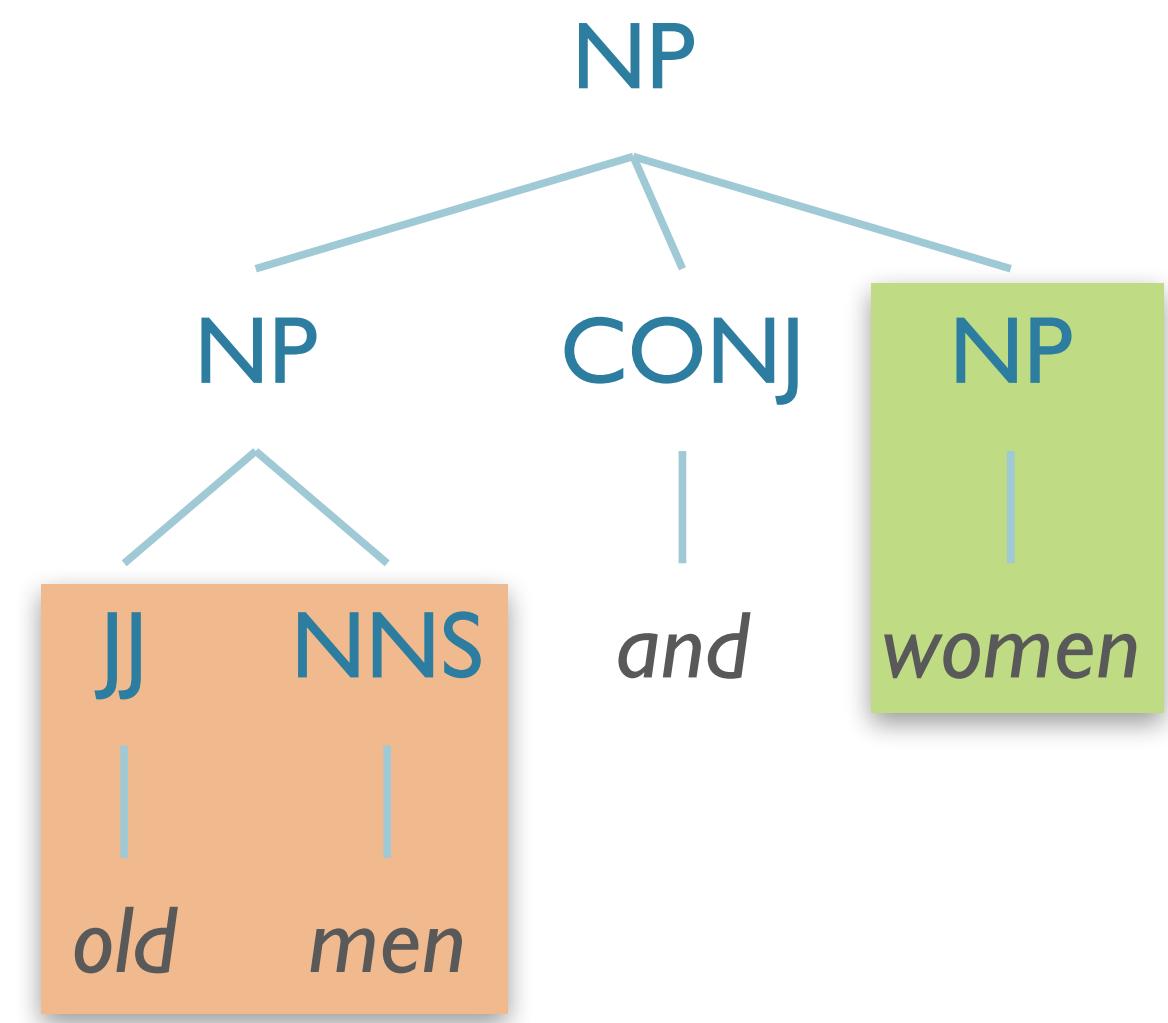
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# Coordination Ambiguity:

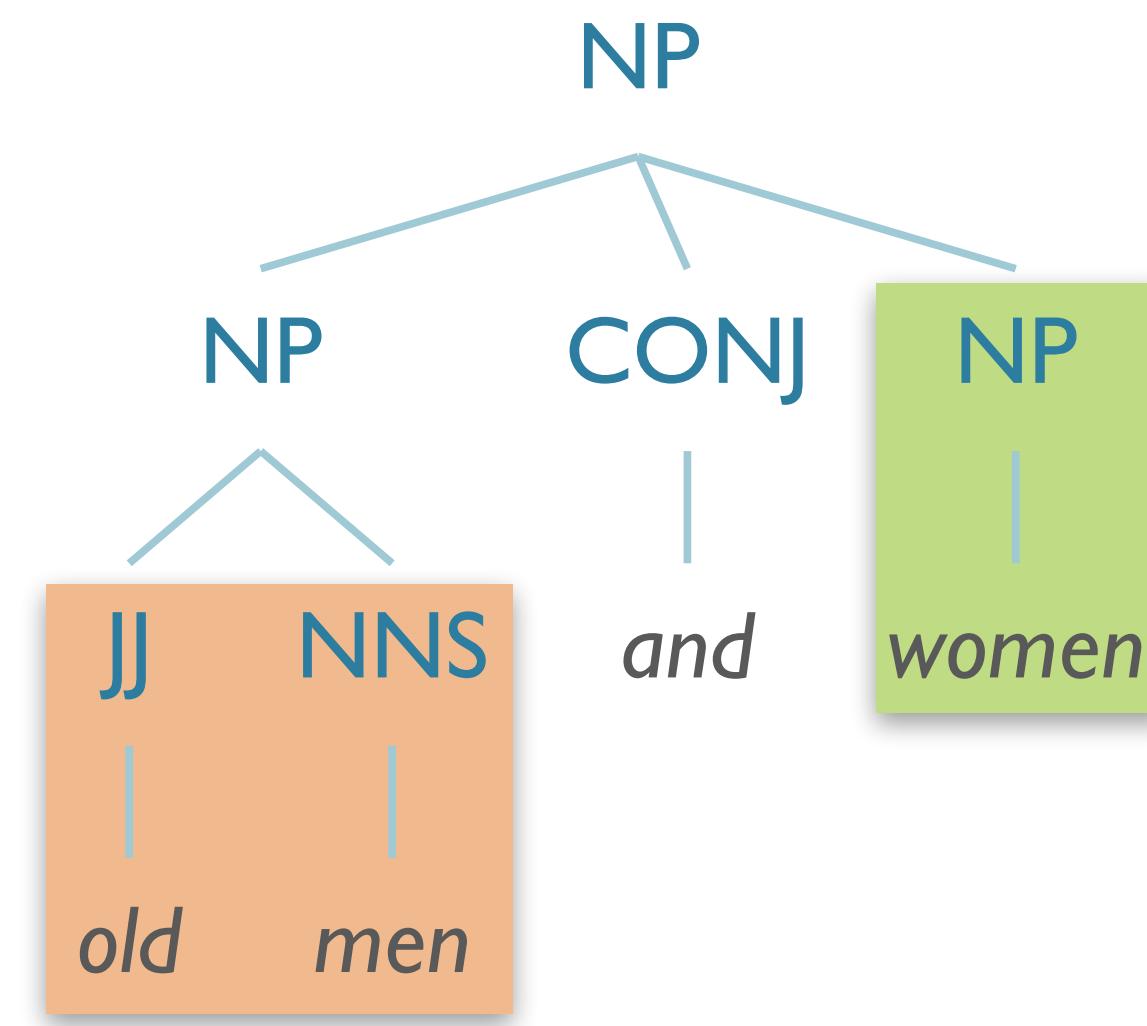
# Coordination Ambiguity:

**[old men] and [women]**

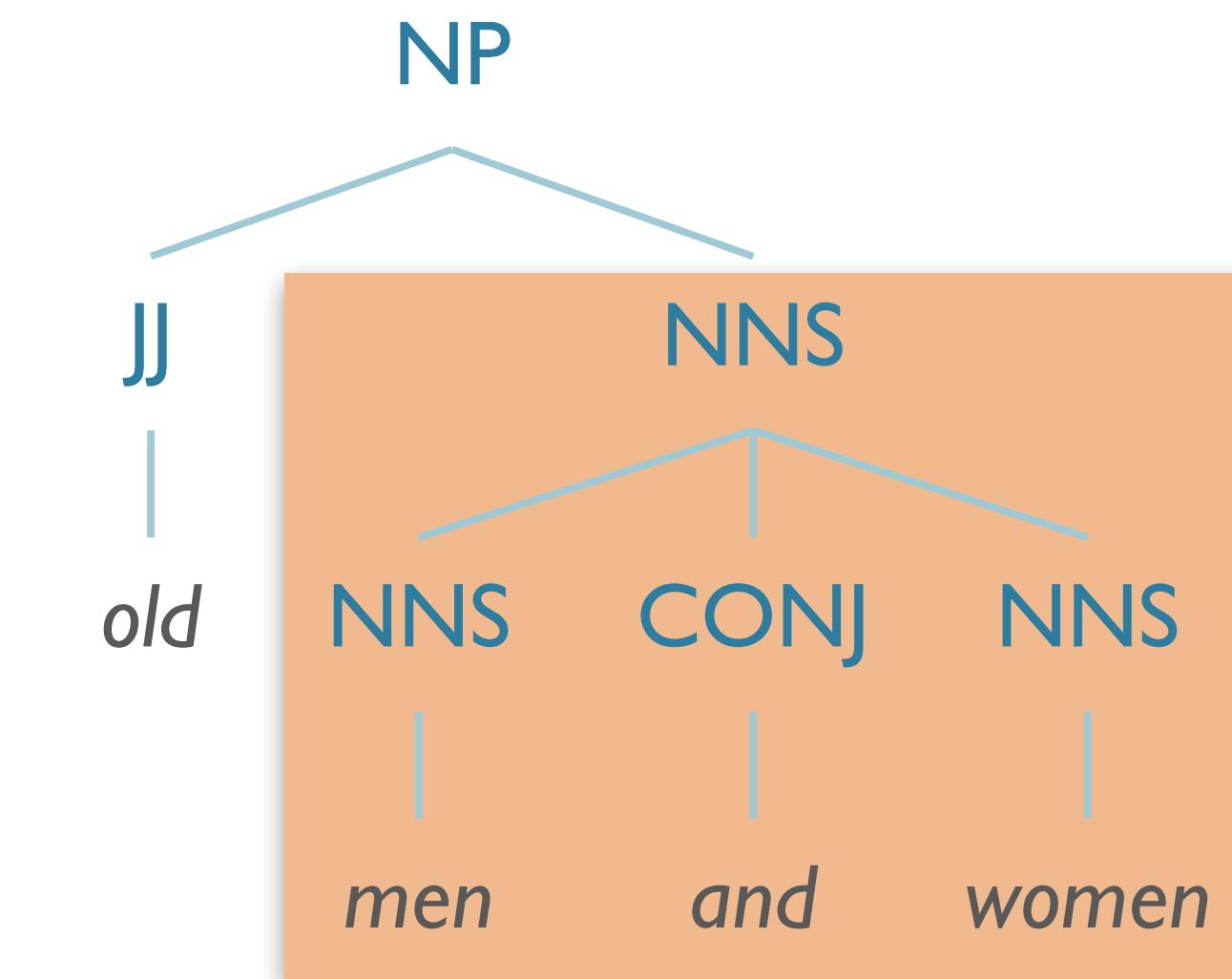


# Coordination Ambiguity:

**[old men] and [women]**



**[old [men and women]]**



# Local vs. Global Ambiguity

- *Local* ambiguity:
  - Ambiguity that cannot contribute to a full, valid parse
  - e.g. *Book/NN* in “*Book that flight*”

# Local vs. Global Ambiguity

- *Local* ambiguity:
  - Ambiguity that cannot contribute to a full, valid parse
  - e.g. *Book/NN* in “*Book that flight*”
- *Global* ambiguity
  - Multiple valid parses

# Why is Ambiguity a Problem?

- *Local* ambiguity:
  - increased processing time
- *Global* ambiguity:
  - Would like to yield only “reasonable” parses
  - Ideally, the one that was intended\*

# Solution to Ambiguity?

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- *Disambiguation!*

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- *Disambiguation!*
- Different possible strategies to select correct interpretation:

# Disambiguation Strategy: Statistical

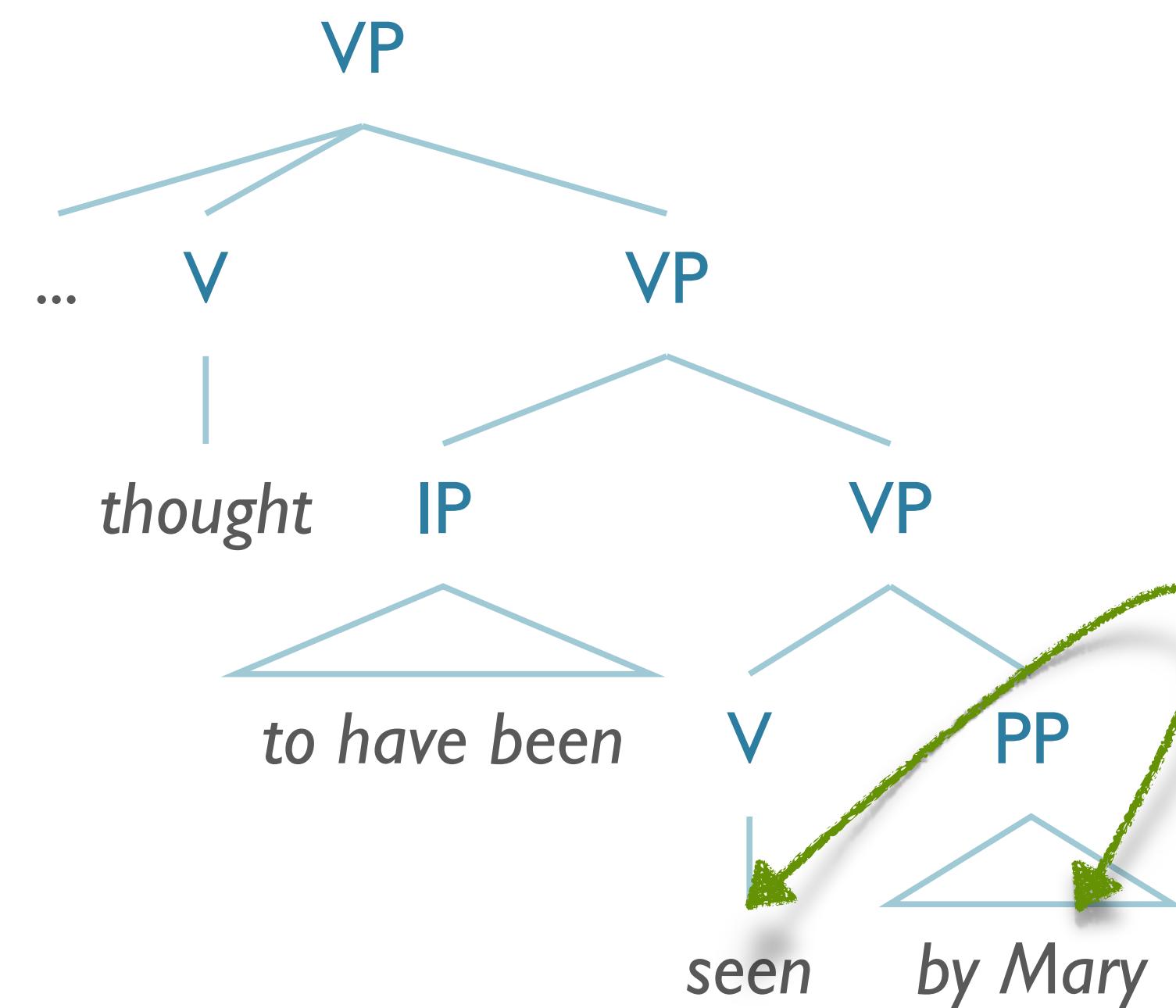
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  - *John was thought to have been seen by Mary*
  - Mary could be doing the seeing or thinking — seeing more likely

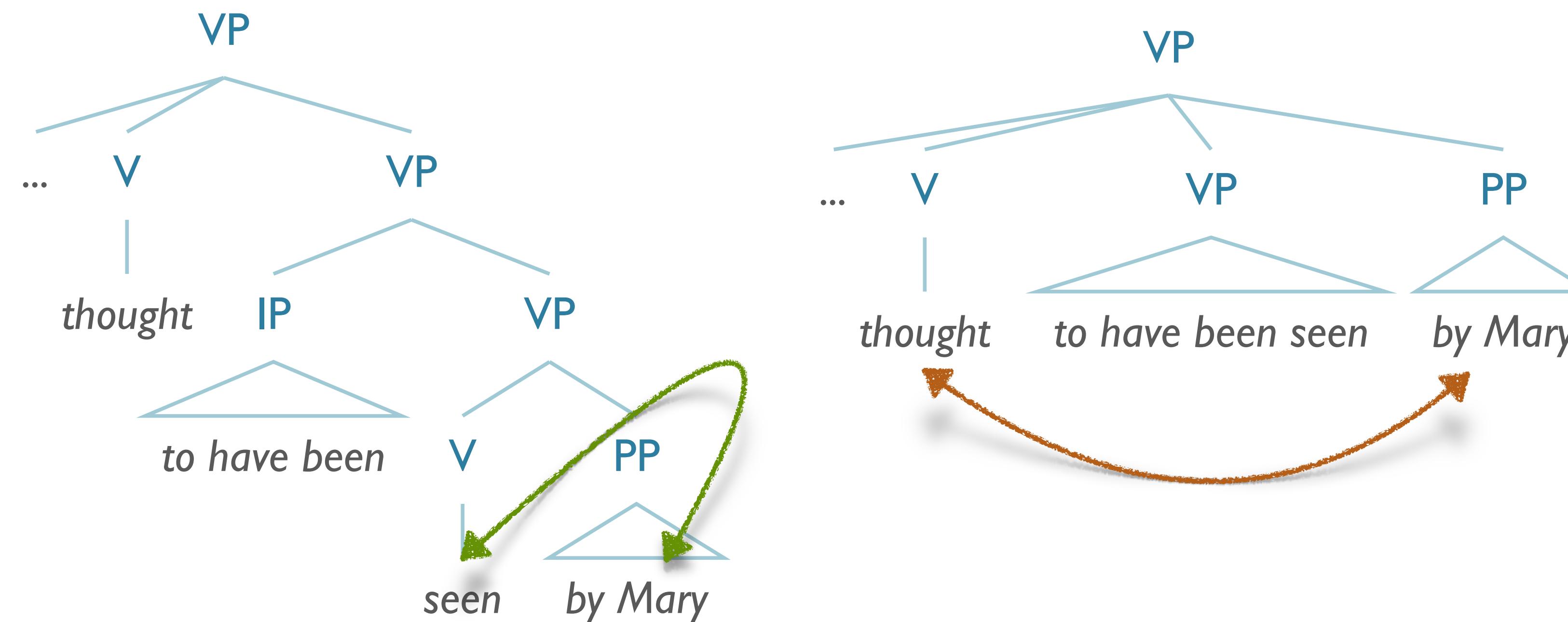
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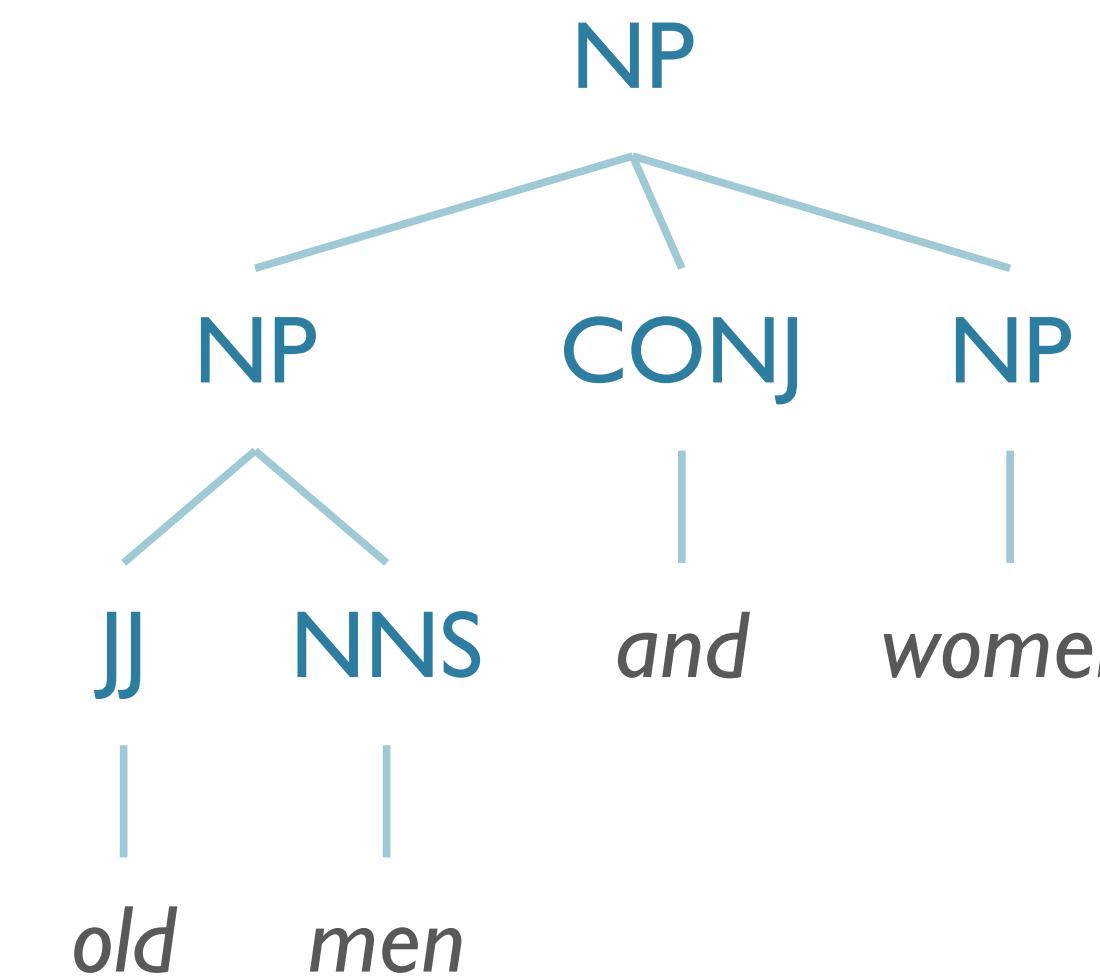
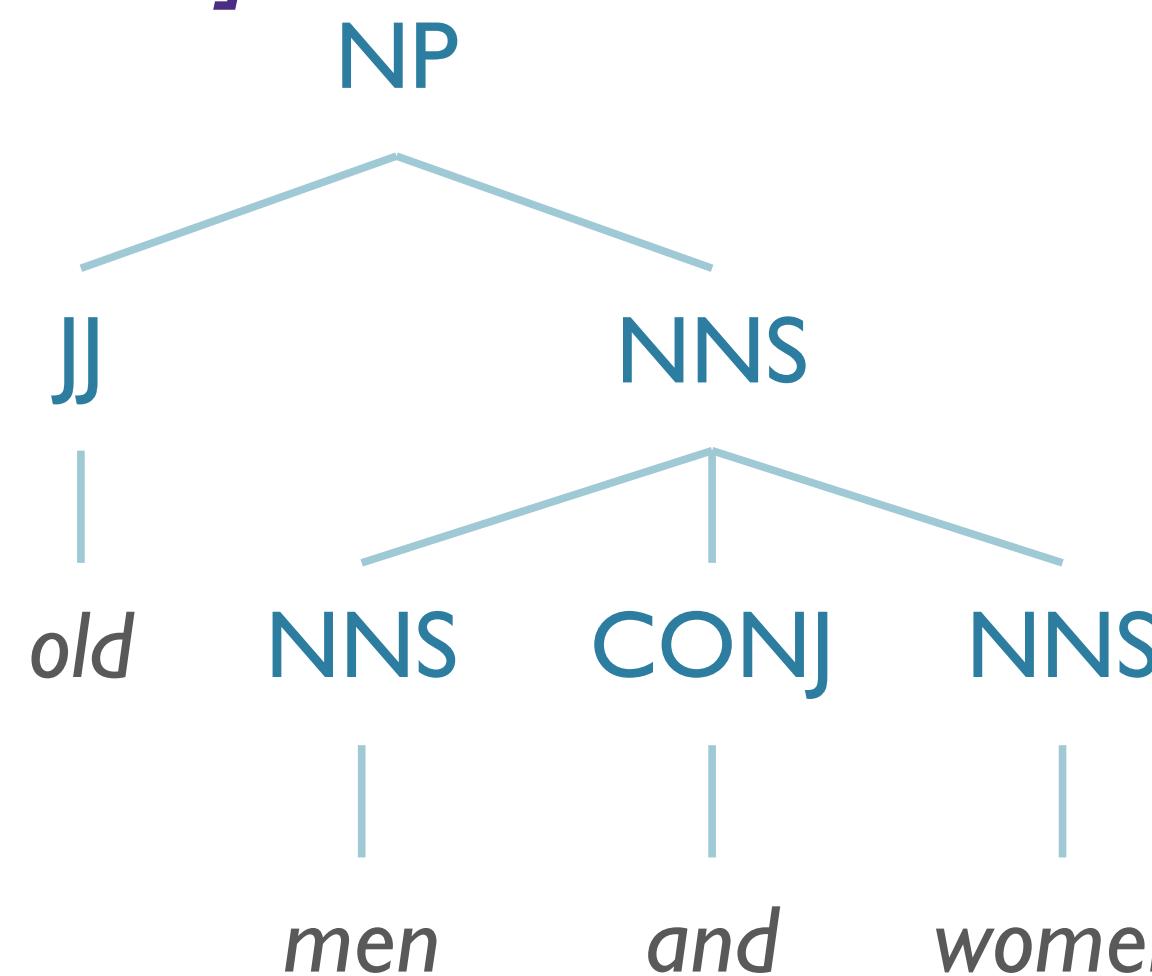


# Disambiguation Strategy: Statistical

- Some phrases more likely overall

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- Some phrases more likely overall
- *[old [men and women]] is a more common construction than [old men] and [women]*



# Disambiguation Strategy: Semantic

- Some interpretations we know to be semantically impossible

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- *Eiffel tower* as subject of *fly*

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- Some interpretations are possible, unlikely given world knowledge
  - e.g. elephants and pajamas

# Incremental Parsing and Garden Paths

- Idea: model *left-to-right* nature of (English) text
- Problem: “garden path” sentences

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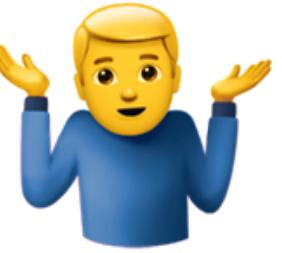
Business Markets World Politics TV More

SPORTS NEWS SEPTEMBER 30, 2019 / 9:17 AM / A DAY AGO

## California to let college athletes be paid in blow to NCAA rules

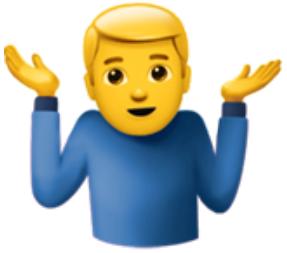
<https://www.reuters.com/article/us-sport-california-education/california-to-let-college-athletes-be-paid-in-blow-to-ncaa-rules-idUSKBN1WF1SR>

## Disambiguation Strategy:



- Alternatively, keep all parses

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- Alternatively, keep all parses
  - (*Might even be the appropriate action for some jokes*)

# Parsing Challenges

- Parsing-as-Search
- **Parsing Challenges**
  - Ambiguity
  - **Repeated Substructure**
  - Recursion
- Strategy: Dynamic Programming
- Grammar Equivalence
- CKY parsing algorithm

# Repeated Work

- Search (top-down/bottom-up) both lead to repeated substructures
  - Globally bad parses can construct good subtrees
  - ...will reconstruct along another branch
  - No static backtracking can avoid

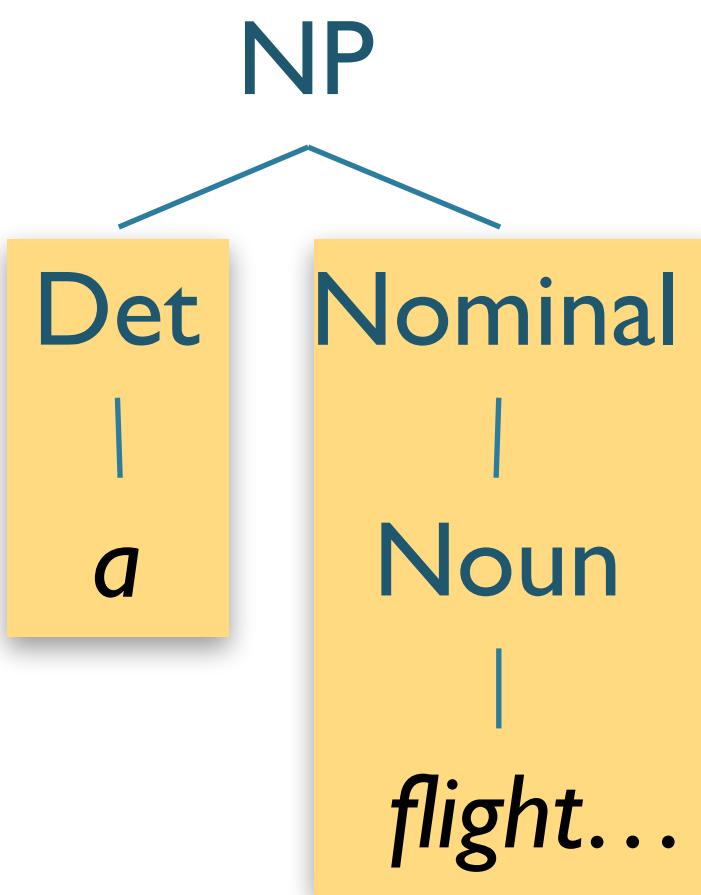
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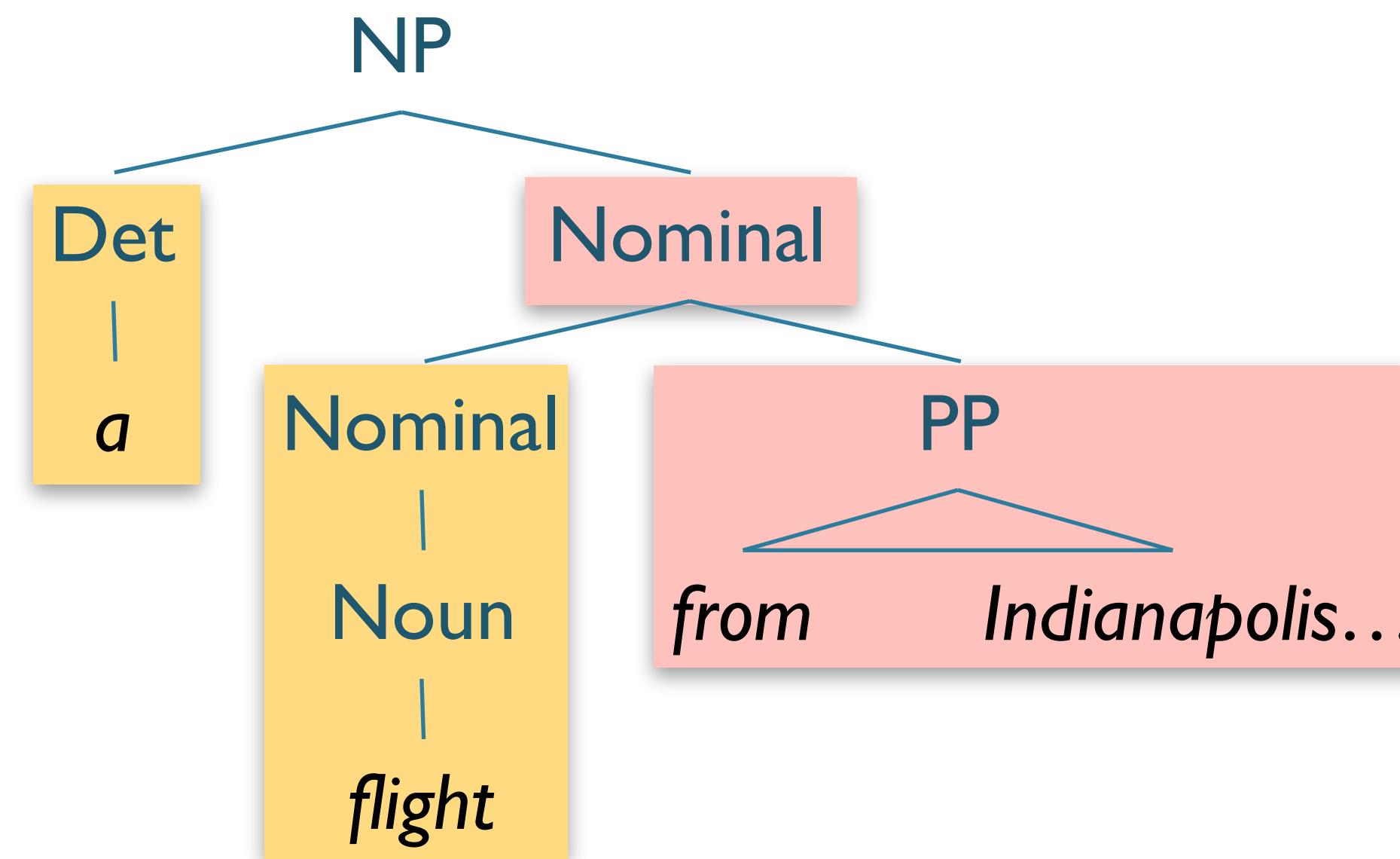
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- Efficient parsing techniques require storage of partial solutions
- Example: *a flight from Indianapolis to Houston on TWA*

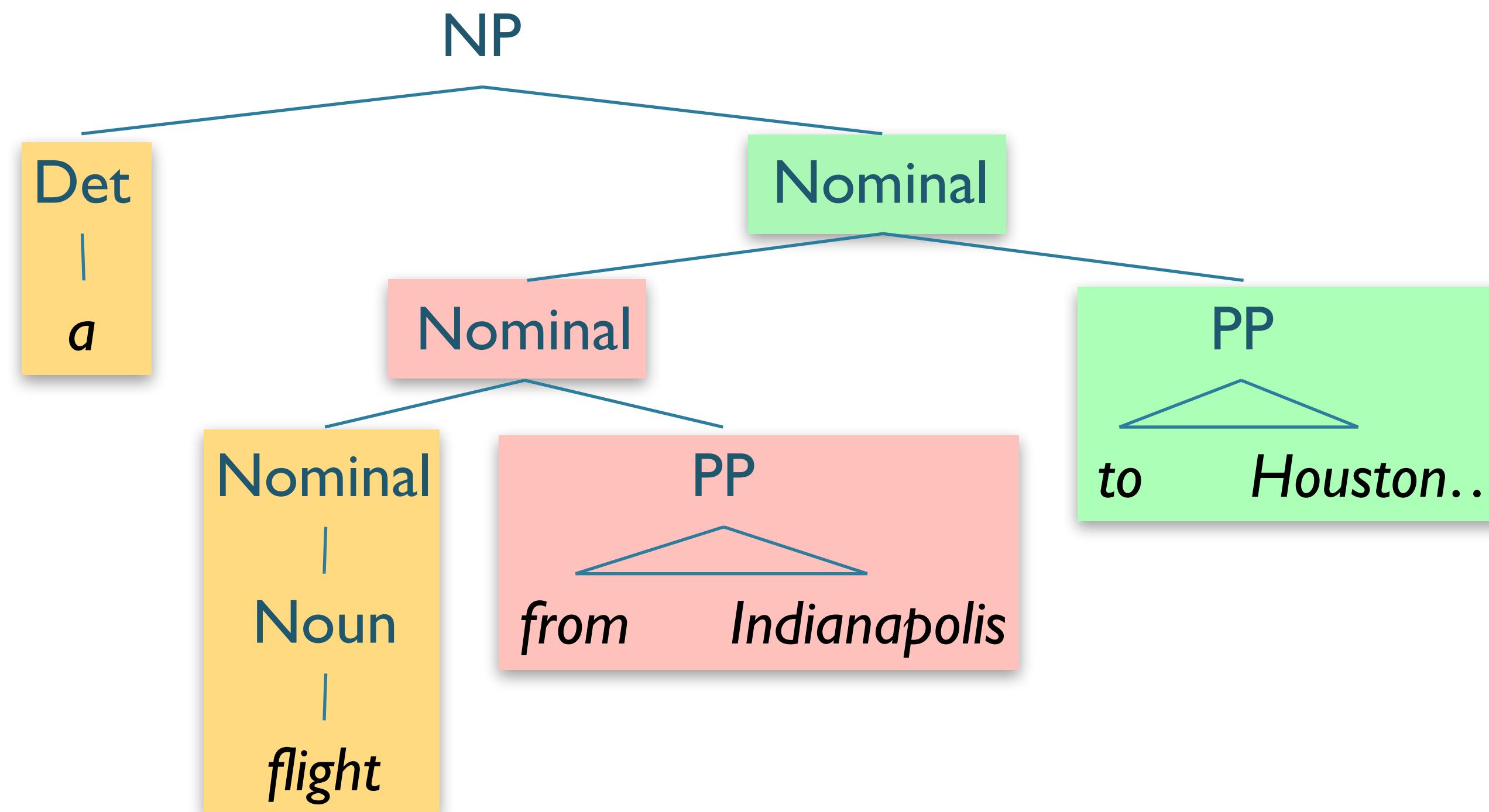
# Shared Sub-Problems



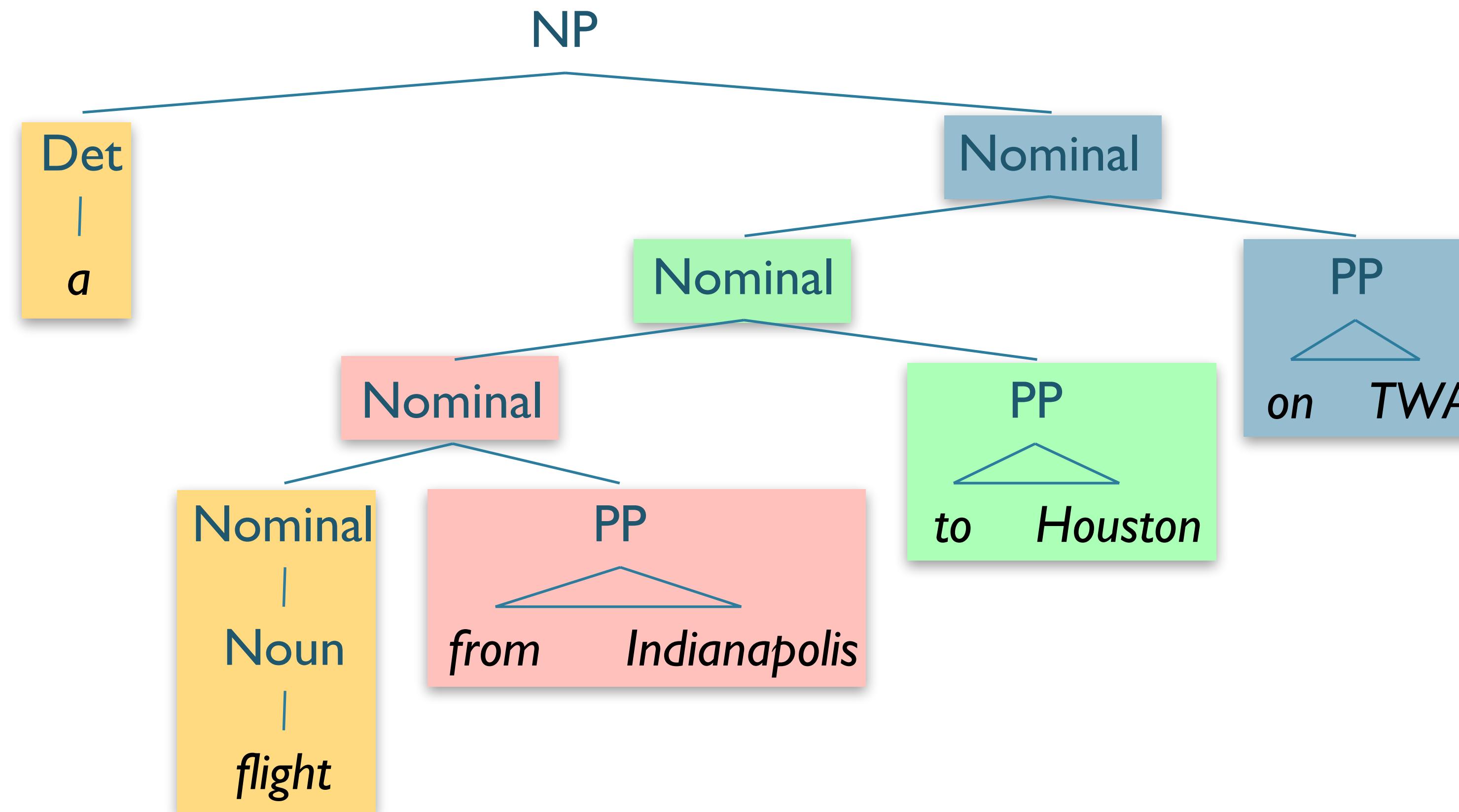
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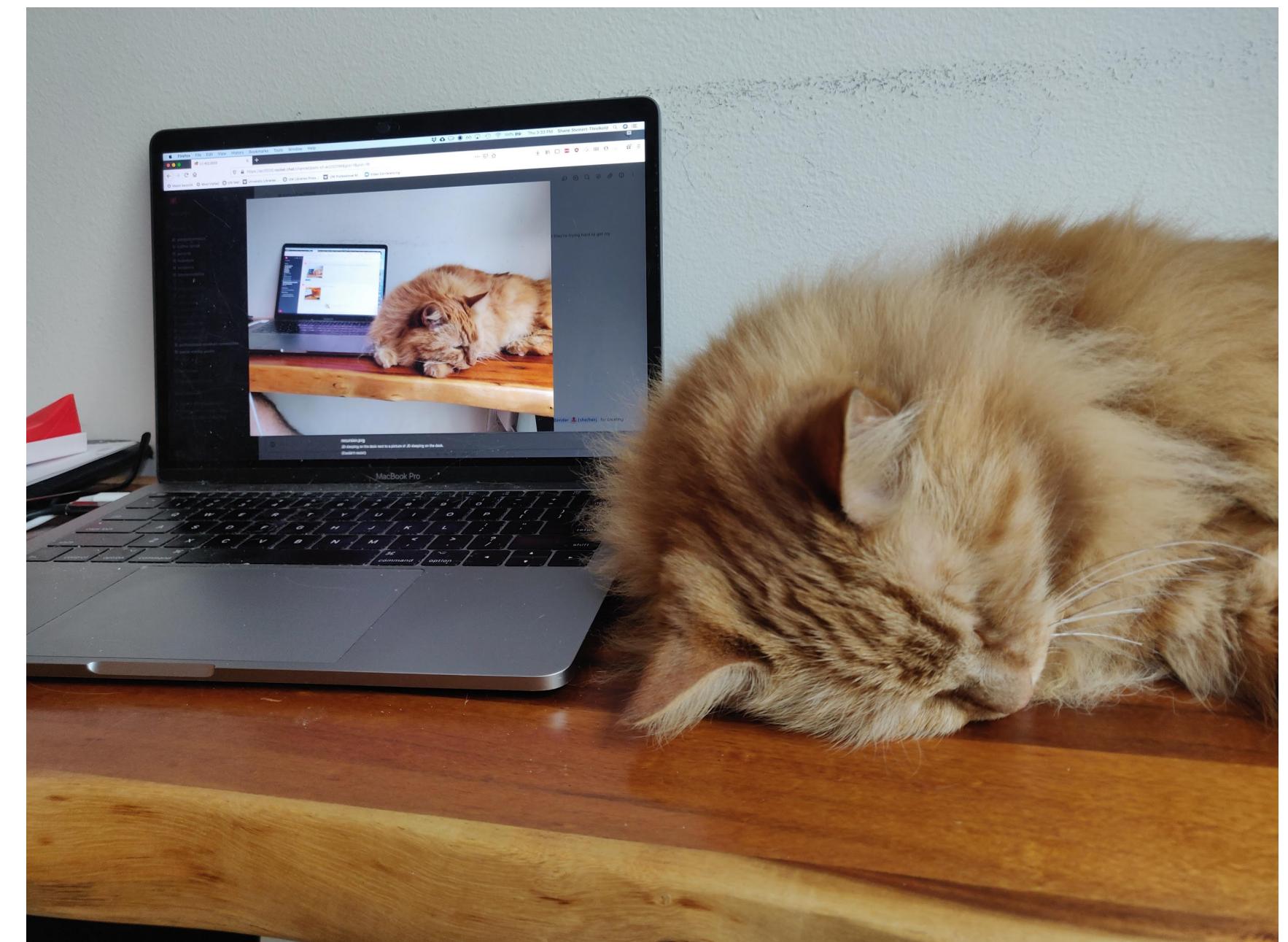


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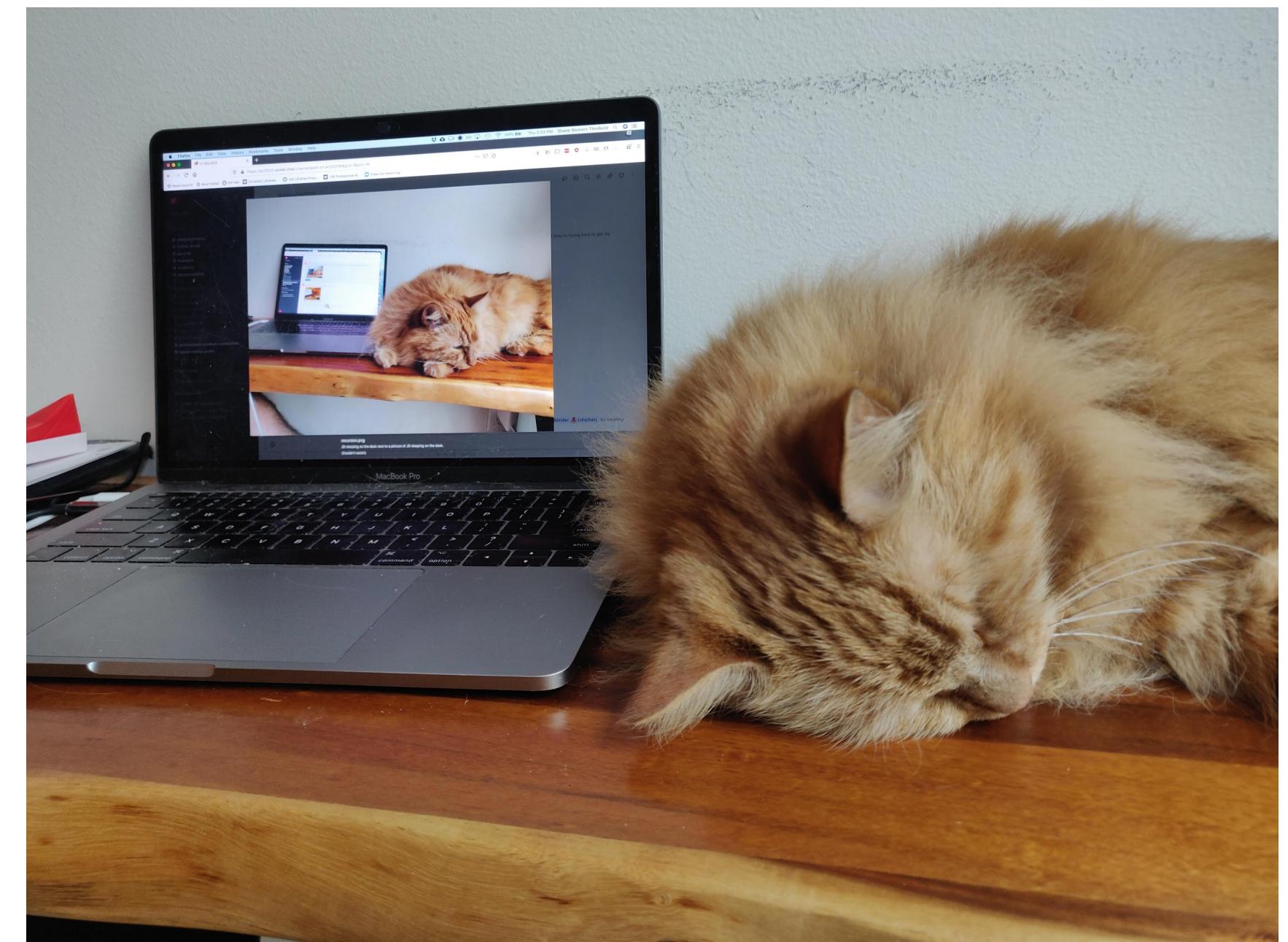
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- Many grammars have recursive rules
- $S \rightarrow S \text{ Conj } S$



# Recursion

- Many grammars have recursive rules
  - $S \rightarrow S \text{ Conj } S$
- In search approaches, recursion is problematic
  - Can yield infinite searches
  - Top-down especially vulnerable



# Roadmap

- Parsing-as-Search
- Parsing Challenges
- **Strategy: Dynamic Programming**
- Grammar Equivalence
- CKY parsing algorithm

# Dynamic Programming

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  - Repeated substructure → Repeated Work

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# Dynamic Programming

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  - Repeated substructure → Repeated Work
- Insight:
  - Global parse composed of sub-parses
  - Can record these sub-parses and re-use
- Dynamic programming avoids repeated work by recording the subproblems
  - Here, stores subtrees

# Parsing with Dynamic Programming

- Avoids repeated work
- Allows implementation of (relatively) efficient parsing algorithms
  - Polynomial time in input length
  - Typically cubic ( $n^3$ ) or less

# Parsing with Dynamic Programming

- Avoids repeated work
- Allows implementation of (relatively) efficient parsing algorithms
  - Polynomial time in input length
  - Typically cubic ( $n^3$ ) or less
- Several different implementations
  - Cocke-Kasami-Younger (CKY) algorithm
  - Earley algorithm
  - Chart parsing

# Roadmap

- Parsing-as-Search
- Parsing Challenges
- Strategy: Dynamic Programming
- **Grammar Equivalence**
- CKY parsing algorithm

# Grammar Equivalence and Form

- *Weak* Equivalence
  - Accepts same language
  - May produce **different** structures
- *Strong* Equivalence
  - Accepts same language
  - Produces **same** structures

# Grammar Equivalence and Form

# Grammar Equivalence and Form

- Reason?
  - We can create a weakly-equivalent grammar that allows for greater efficiency
  - This is required by the CKY algorithm

# Chomsky Normal Form (CNF)

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- Most of our grammars are not of this form:
  - $S \rightarrow Wh\text{-}NP \ Aux \ NP \ VP$
- Need a general conversion procedure

# Chomsky Normal Form (CNF)

# Chomsky Normal Form (CNF)

- Weak equivalence: for every CFG  $G$ , there is a weakly equivalent CNF grammar  $G'$ .
- i.e.: there is a grammar in CNF s.t.  $L(G) = L(G')$ .

# CNF Conversion

Hybrid productions:

$$INF\text{-}VP \rightarrow \text{to } VP$$

Unit productions:

$$A \rightarrow B$$

Long productions:

$$A \rightarrow B \ C \ D \dots$$

# CNF Conversion: Hybrid Productions

- Hybrid production:
  - Replace all terminals with dummy non-terminal
  - $INF\text{-}VP \rightarrow \text{to } VP$
  - $INF\text{-}VP \rightarrow TO\ VP$
  - $TO \rightarrow \text{to}$

# CNF Conversion: Unit Productions

- Unit productions:
  - Rewrite RHS with RHS of all derivable, non-unit productions
  - If  $A \xrightarrow{*} B$  and  $B \rightarrow \gamma$ , **add**  $A \rightarrow \gamma$  [where  $\gamma$  is any non-unit RHS]
  - *[ $A \xrightarrow{*} B$ :  $B$  is reachable from  $A$  by a sequence of unit productions]*
- *Nominal  $\rightarrow$  Noun, Noun  $\rightarrow$  dog*
  - *Nominal  $\rightarrow$  dog*
  - *Noun  $\rightarrow$  dog*
- NB: this example has  $\gamma$  as a single terminal, but the rule applies to all non-unit RHS.

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$$S \rightarrow Aux \ NP \ VP$$

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# CNF Conversion: Long Productions

- Long productions

$$S \rightarrow Aux \ NP \ VP$$
$$S \rightarrow \textcolor{red}{X1} \ VP \quad \textcolor{red}{X1} \rightarrow Aux \ NP$$

- Introduce unique nonterminals, and spread over rules

# CNF Conversion

Convert terminals in hybrid rules to dummy non-terminals

Convert unit productions

Binarize long production rules

## $\mathcal{L}_1$ Grammar

$S \rightarrow NP\ VP$   
 $S \rightarrow Aux\ NP\ VP$

$S \rightarrow VP$

$NP \rightarrow Pronoun$

$NP \rightarrow Proper-Noun$

$NP \rightarrow Det\ Nominal$

$Nominal \rightarrow Noun$

$Nominal \rightarrow Nominal\ Noun$

$Nominal \rightarrow Nominal\ PP$

$VP \rightarrow Verb$

$VP \rightarrow Verb\ NP$

$VP \rightarrow Verb\ NP\ PP$

$VP \rightarrow Verb\ PP$

$VP \rightarrow VP\ PP$

$PP \rightarrow Preposition\ NP$

## $\mathcal{L}_1$ in CNF

$S \rightarrow NP\ VP$   
 $S \rightarrow X1\ VP$   
 $X1 \rightarrow Aux\ NP$   
 $S \rightarrow book\ / include\ / prefer$

$S \rightarrow Verb\ NP$

$S \rightarrow X2\ PP$

$S \rightarrow Verb\ PP$

$S \rightarrow VP\ PP$

$NP \rightarrow I\ / she\ / me$

$NP \rightarrow TWA\ / Houston$

$NP \rightarrow Det\ Nominal$

$Nominal \rightarrow book\ / flight\ / meal\ / money$

$Nominal \rightarrow Nominal\ Noun$

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$VP \rightarrow book\ / include\ / prefer$

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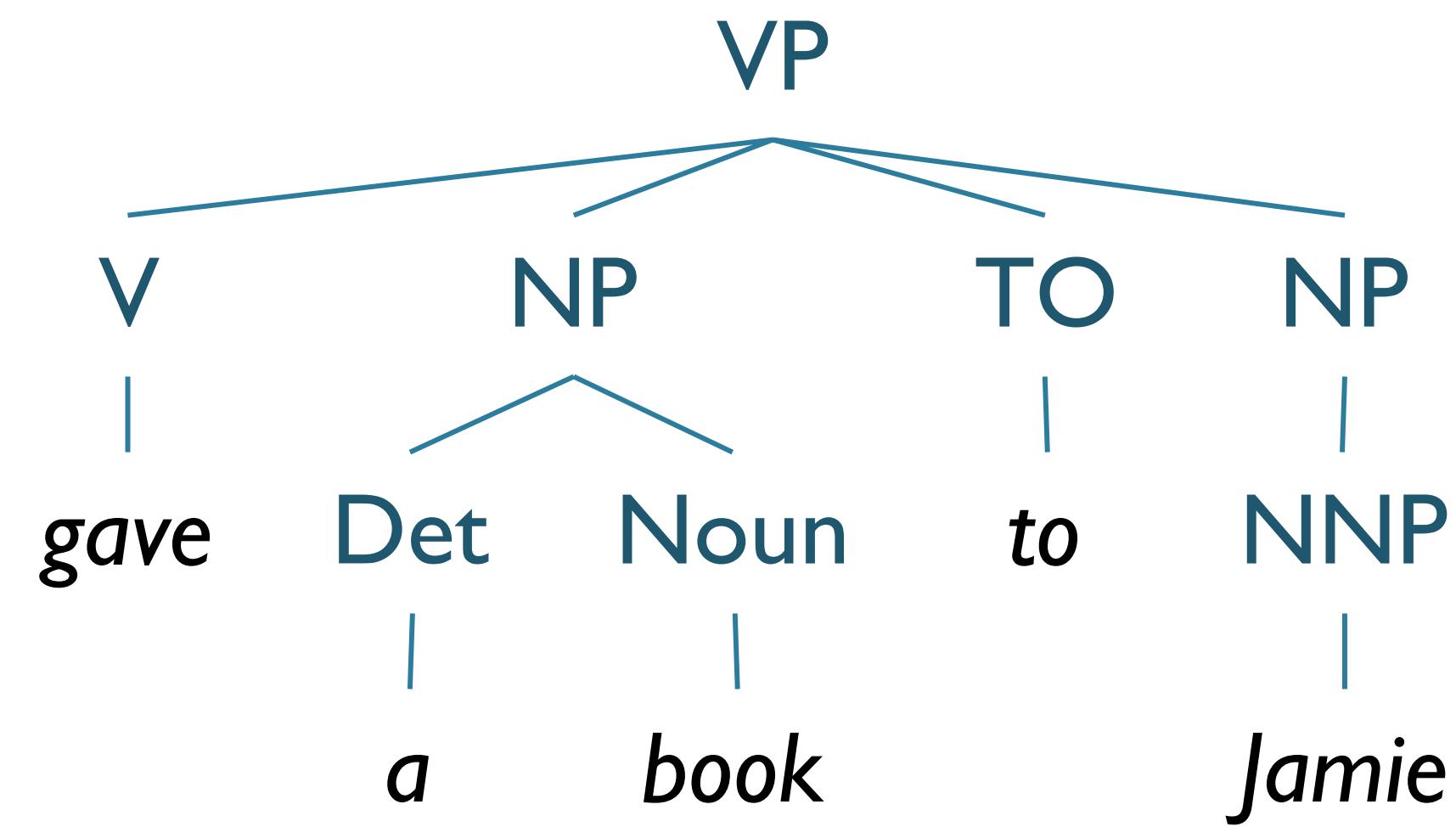
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$\mathcal{L}_1$  Grammar $S \rightarrow NP\ VP$  $S \rightarrow Aux\ NP\ VP$  $S \rightarrow VP$  $NP \rightarrow Pronoun$  $NP \rightarrow Proper-Noun$  $NP \rightarrow Det\ Nominal$  $Nominal \rightarrow Noun$  $Nominal \rightarrow Nominal\ Noun$  $Nominal \rightarrow Nominal\ PP$  $VP \rightarrow Verb$  $VP \rightarrow Verb\ NP$  $VP \rightarrow Verb\ NP\ PP$  $VP \rightarrow Verb\ PP$  $VP \rightarrow VP\ PP$  $PP \rightarrow Preposition\ NP$  $\mathcal{L}_1$  in CNF $S \rightarrow NP\ VP$  $S \rightarrow X1\ VP$  $X1 \rightarrow Aux\ NP$  $S \rightarrow book\ / include\ / prefer$  $S \rightarrow Verb\ NP$  $S \rightarrow X2\ PP$  $S \rightarrow Verb\ PP$  $S \rightarrow VP\ PP$  $NP \rightarrow I\ / she\ / me$  $NP \rightarrow TWA\ / Houston$  $NP \rightarrow Det\ Nominal$  $Nominal \rightarrow book\ / flight\ / meal\ / money$  $Nominal \rightarrow Nominal\ Noun$  $Nominal \rightarrow Nominal\ PP$  $VP \rightarrow book\ / include\ / prefer$  $VP \rightarrow Verb\ NP$  $VP \rightarrow X2\ PP$  $X2 \rightarrow Verb\ NP$  $VP \rightarrow Verb\ PP$  $VP \rightarrow VP\ PP$  $PP \rightarrow Preposition\ NP$

# Variation in CNF: Binarization

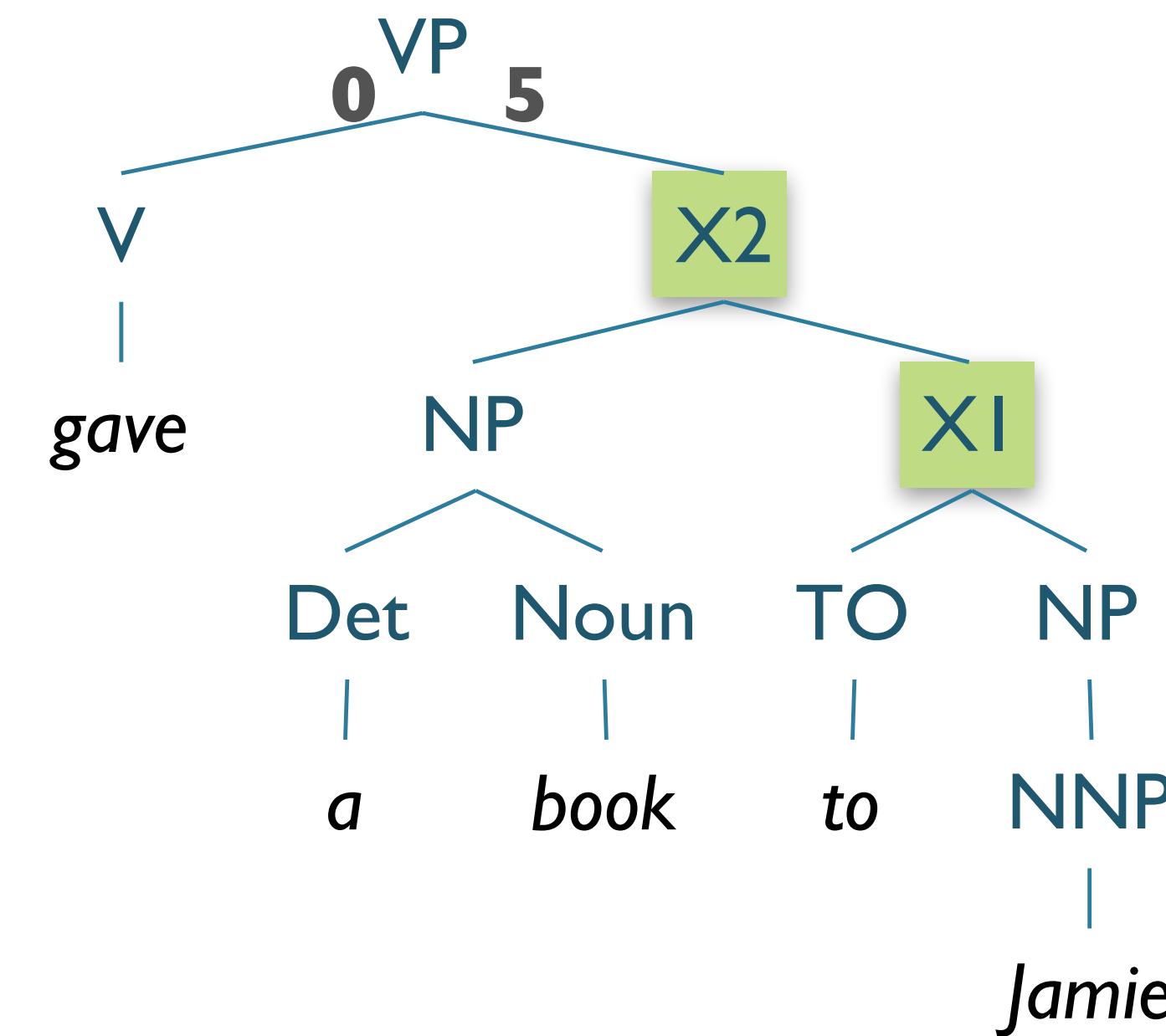
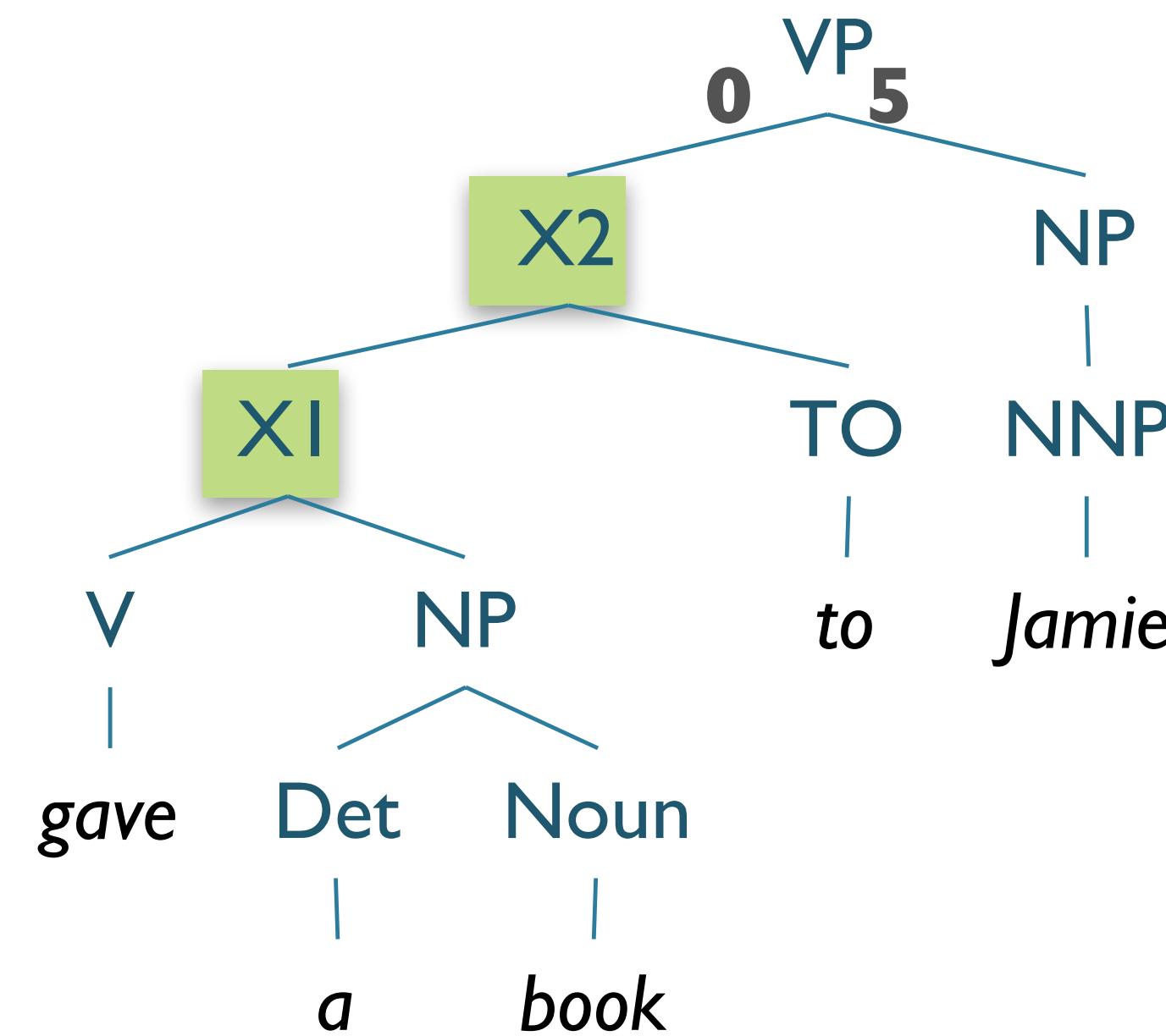
Original Rule

$VP \rightarrow V \ NP \ TO \ NP$



# Variation in CNF: Binarization

Original Rule			
$VP \rightarrow V \ NP \ TO \ NP$			
Left to Right Reduction		Right to Left Reduction	
$VP \rightarrow X_1 \ TO \ NP$	$X_1 \rightarrow V \ NP$	$VP \rightarrow V \ NP \ X_1$	$X_1 \rightarrow TO \ NP$
$VP \rightarrow X_2 \ NP$	$X_2 \rightarrow X_1 \ TO$	$VP \rightarrow V \ X_2$	$X_2 \rightarrow NP \ X_1$



# Roadmap

- Parsing-as-Search
- Parsing Challenges
- Strategy: Dynamic Programming
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- **CKY parsing algorithm**

# CKY Parsing

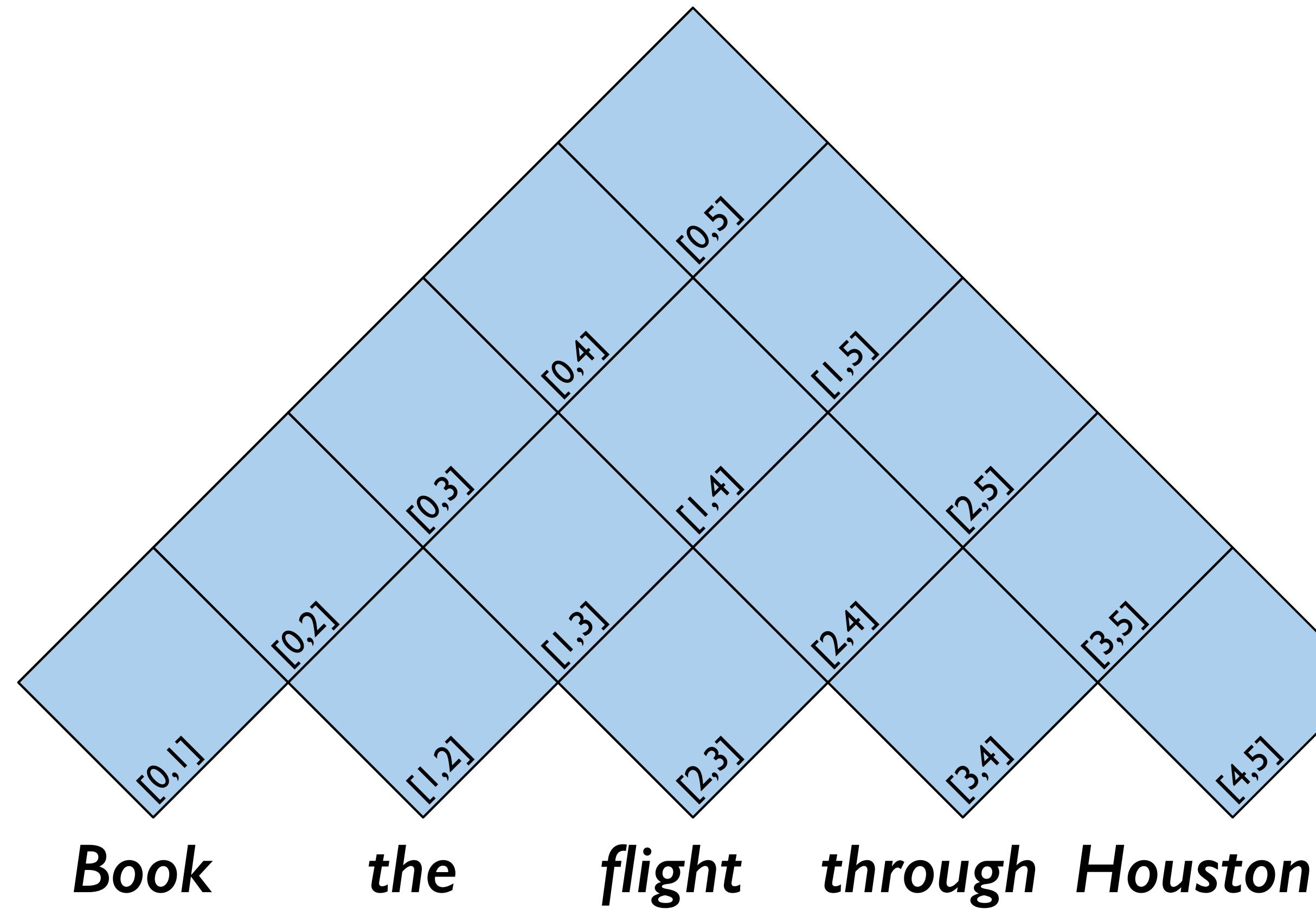
- (Relatively) efficient parsing algorithm
- Based on tabulating substring parses to avoid repeat work
- Approach:
  - Use CNF Grammar
  - Build an  $(n + 1) \times (n + 1)$  matrix to store subtrees
    - Upper triangular portion
  - Incrementally build parse spanning whole input string

# CKY Matrix

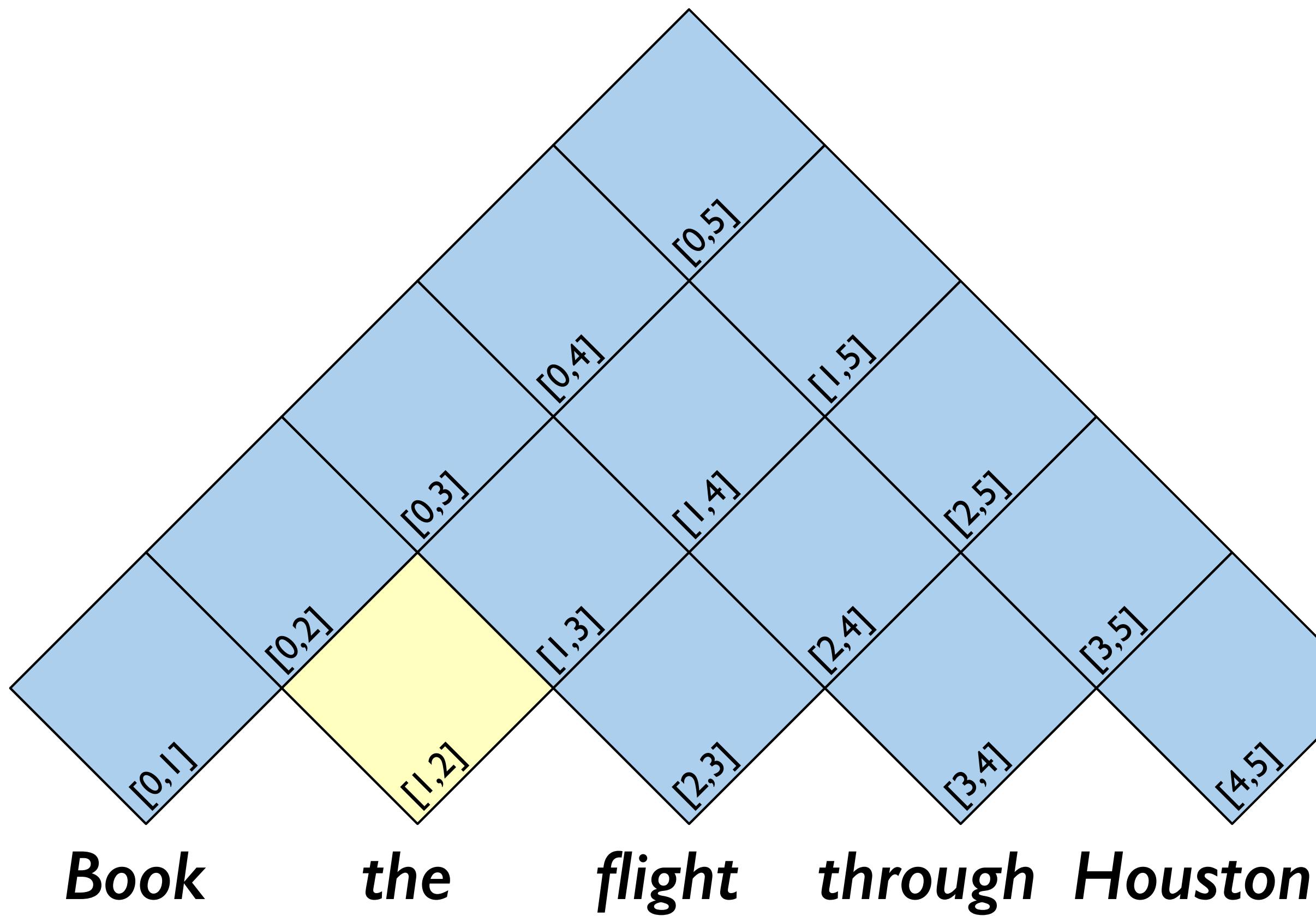
*Book      the      flight    through    Houston*

[0,1]	[0,2]	[0,3]	[0,4]	[0,5]
	[1,2]	[1,3]	[1,4]	[1,5]
		[2,3]	[2,4]	[2,5]
			[3,4]	[3,5]
				[4,5]

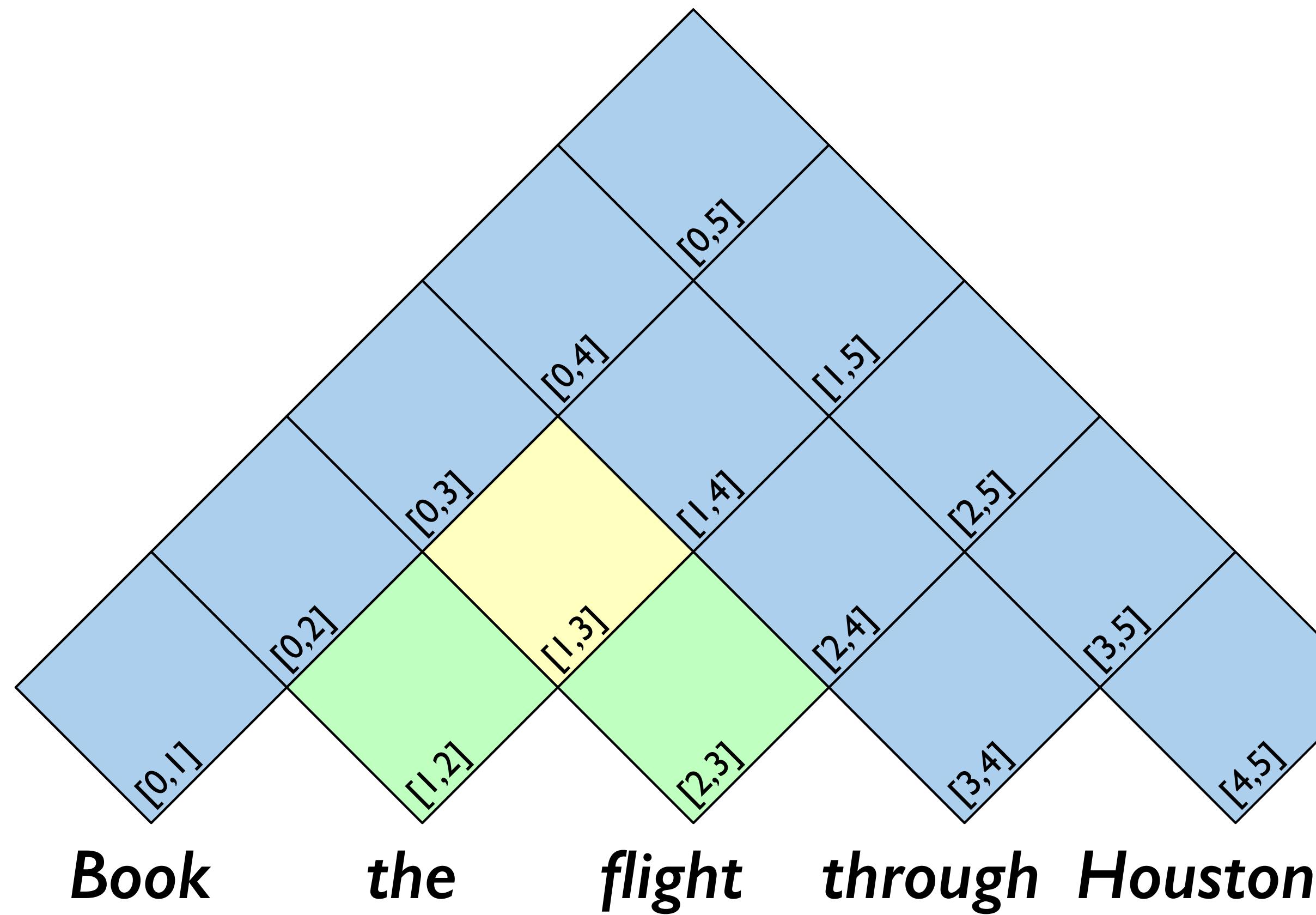
# CKY Matrix



# CKY Matrix



# CKY Matrix



# Dynamic Programming in CKY

- Key idea:
  - for  $i < k < j$
  - ...and a parse spanning substring  $[ i, j ]$
  - There is a  $k$  such that there are parses spanning  $[ i, k ]$  and  $[ k, j ]$
  - We can construct parses for whole sentences by building from these partial parses
- So to have a rule  $A \rightarrow B \ C$  in  $[ i, j ]$ 
  - Must have  $B$  in  $[ i, k ]$  and  $C$  in  $[ k, j ]$  for some  $i < k < j$
  - CNF forces this for all  $j > i + 1$