

CKY Parsing & CNF Conversion

LING 571 — Deep Processing Techniques for NLP

October 3, 2018

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Announcements

- **HW #2** will be extended to Monday, 11/8 at **11:00pm**.
 - Then we will be caught up, so HW #3 will still be due that Friday.
- If you want to use `python3.6` on Patas:
 - `/opt/python-3.6/bin/python3`
 - `nltk` is installed.

Type Hinting in Python

- Supported in ≥ 3.6 [[tutorial](#)]

```
from typing import List
from nltk.grammar import Production
```

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

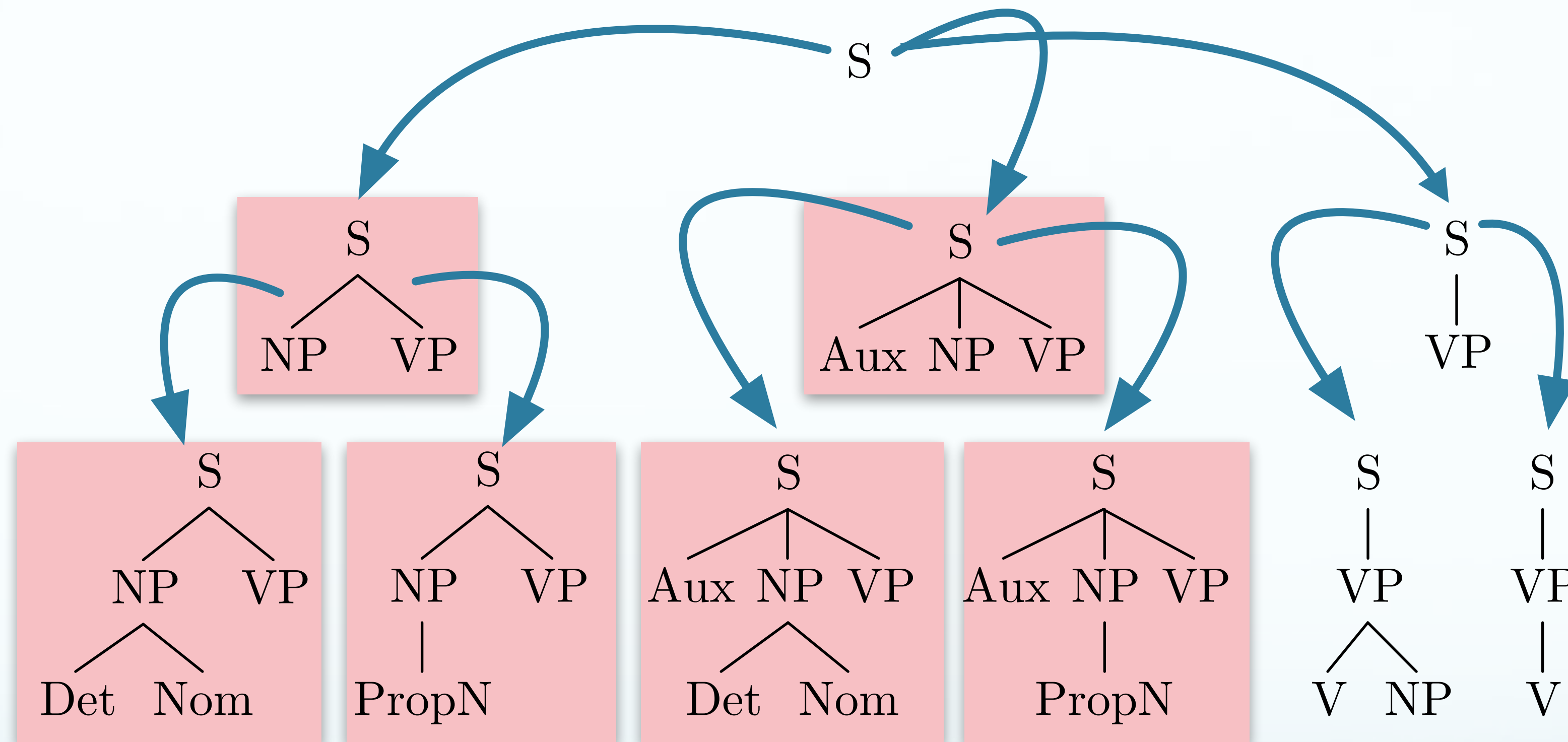
- Also available in PyCharm through [docstrings and/or comments](#):

```
def fix_hybrid_productions(hybrid_prod):
    """
    This function takes a hybrid production and
    returns a list of new CNF productions
    :type hybrid_prod: Production
    :rtype: list[Production]
    """
```

Roadmap

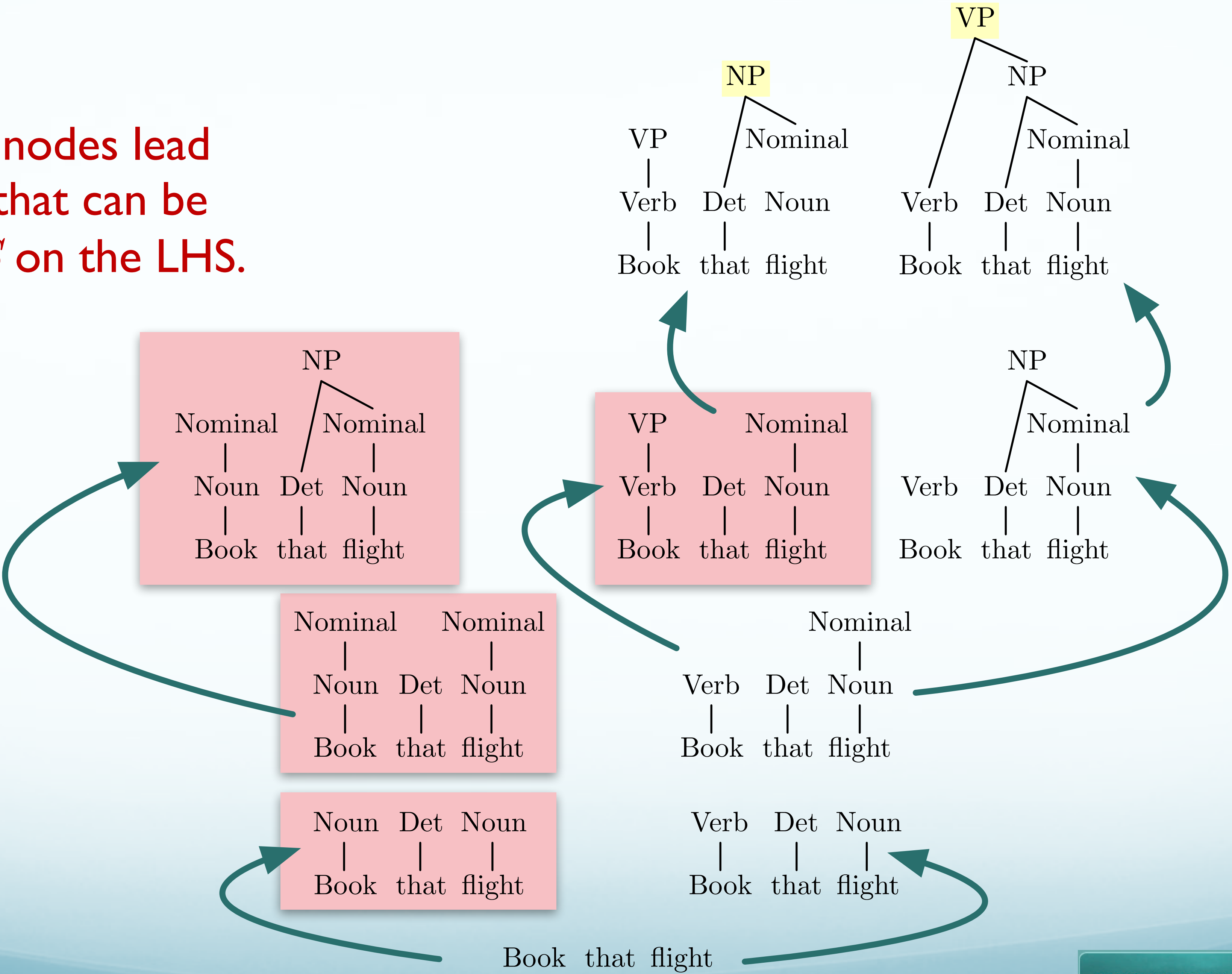
- **Recap: Parsing-as-Search**
- Parsing Challenges
- Strategy: Dynamic Programming
- Grammar Equivalence
- CKY parsing algorithm

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

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

Parsing Ambiguity

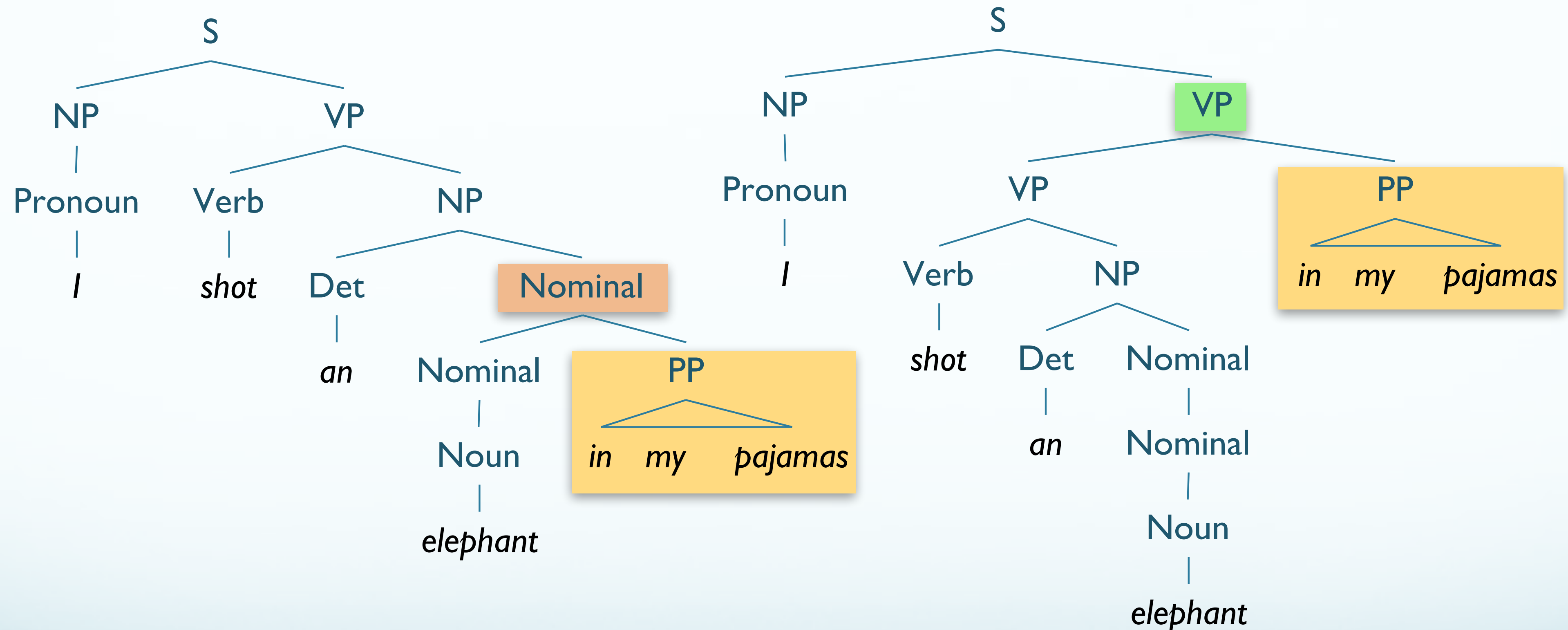
- **Lexical Ambiguity:**
 - **Book/NN** → *I left a **book** on the table.*
 - **Book/VB** → ***Book** that flight.*
- Structural Ambiguity

Attachment Ambiguity

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



Attachment Ambiguity



“We saw the Eiffel Tower flying to Paris”

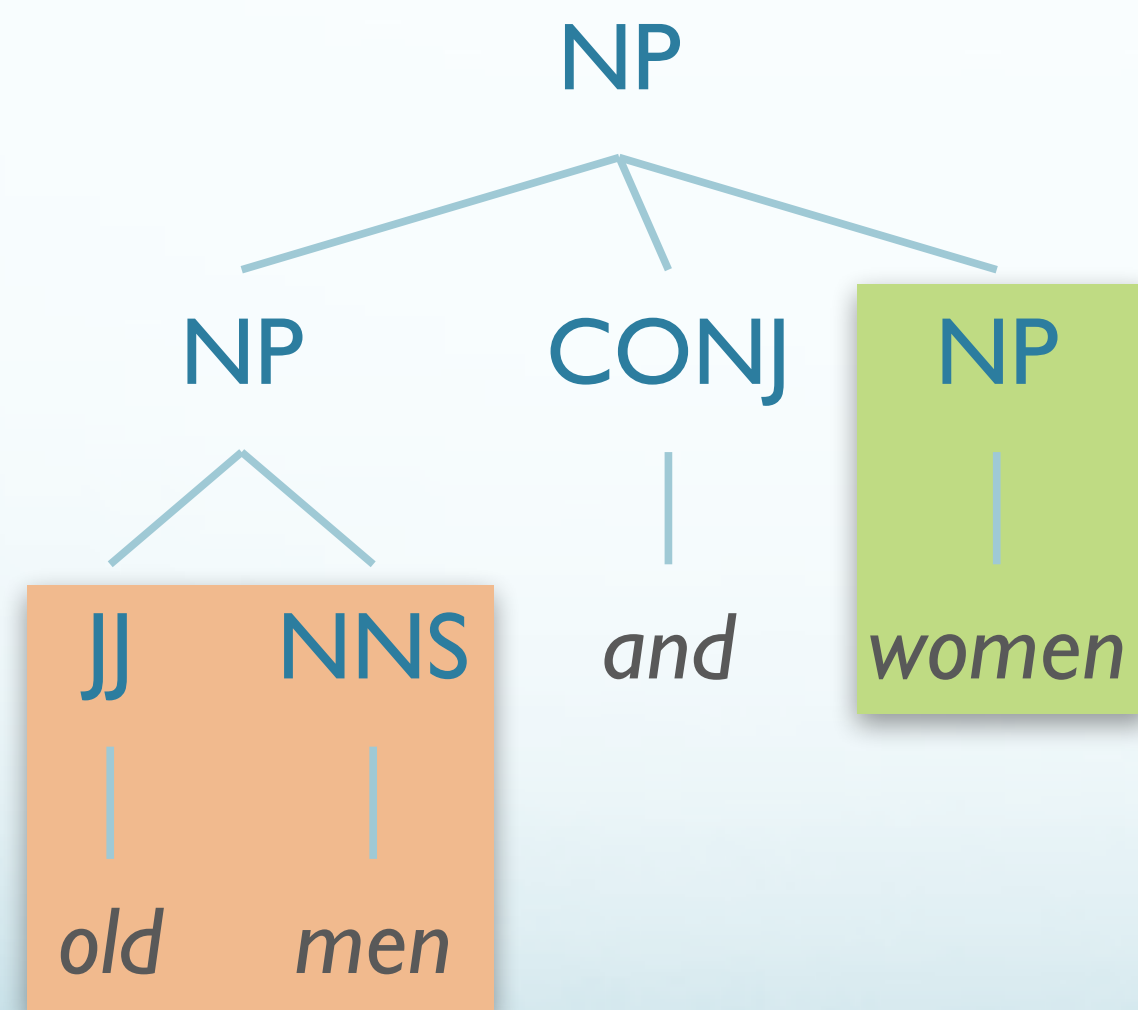


Coordination Ambiguity:

“old men and women”

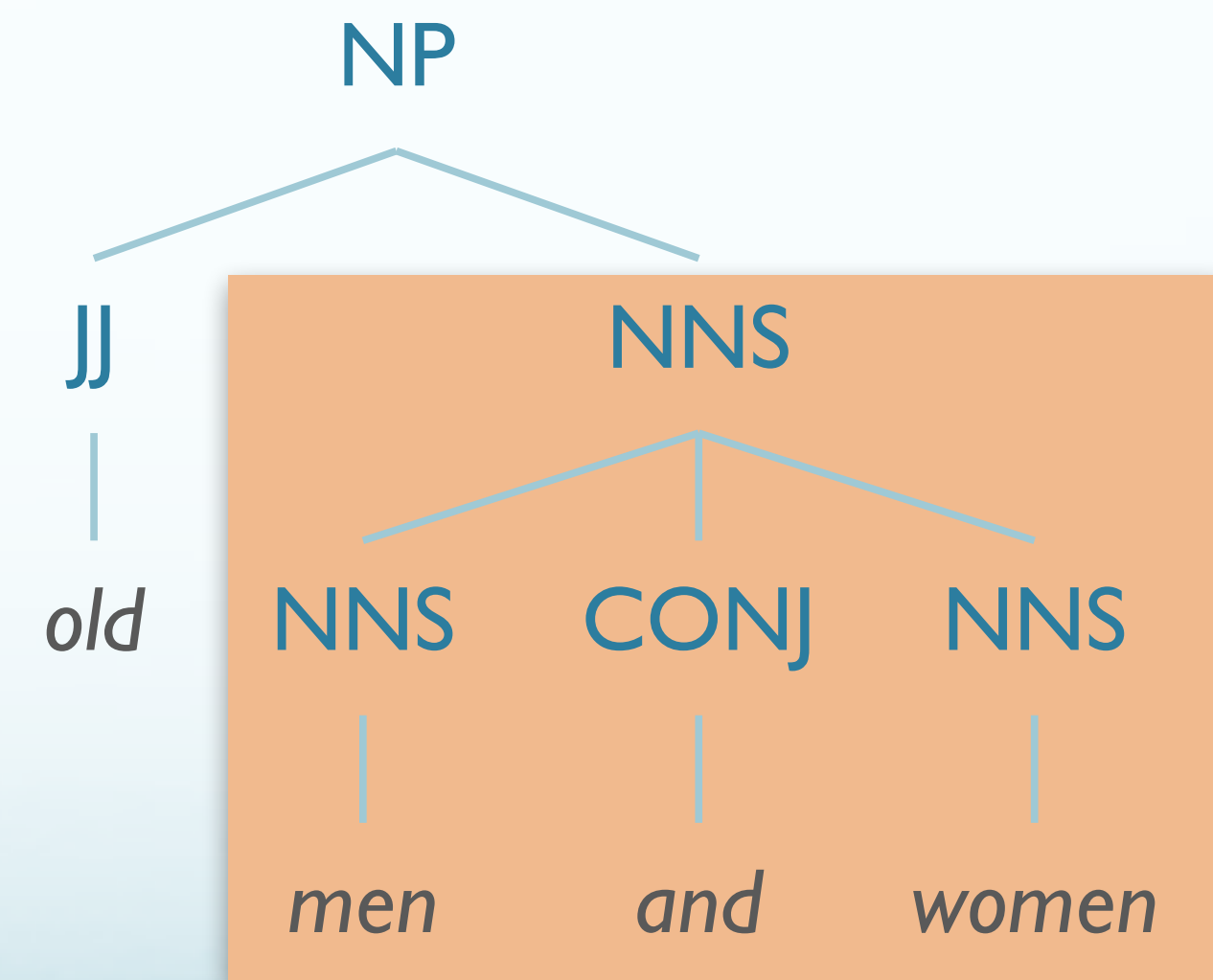
[old men] and **[women]**

(Only the men are old)



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

(Both the men and women are old)



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?

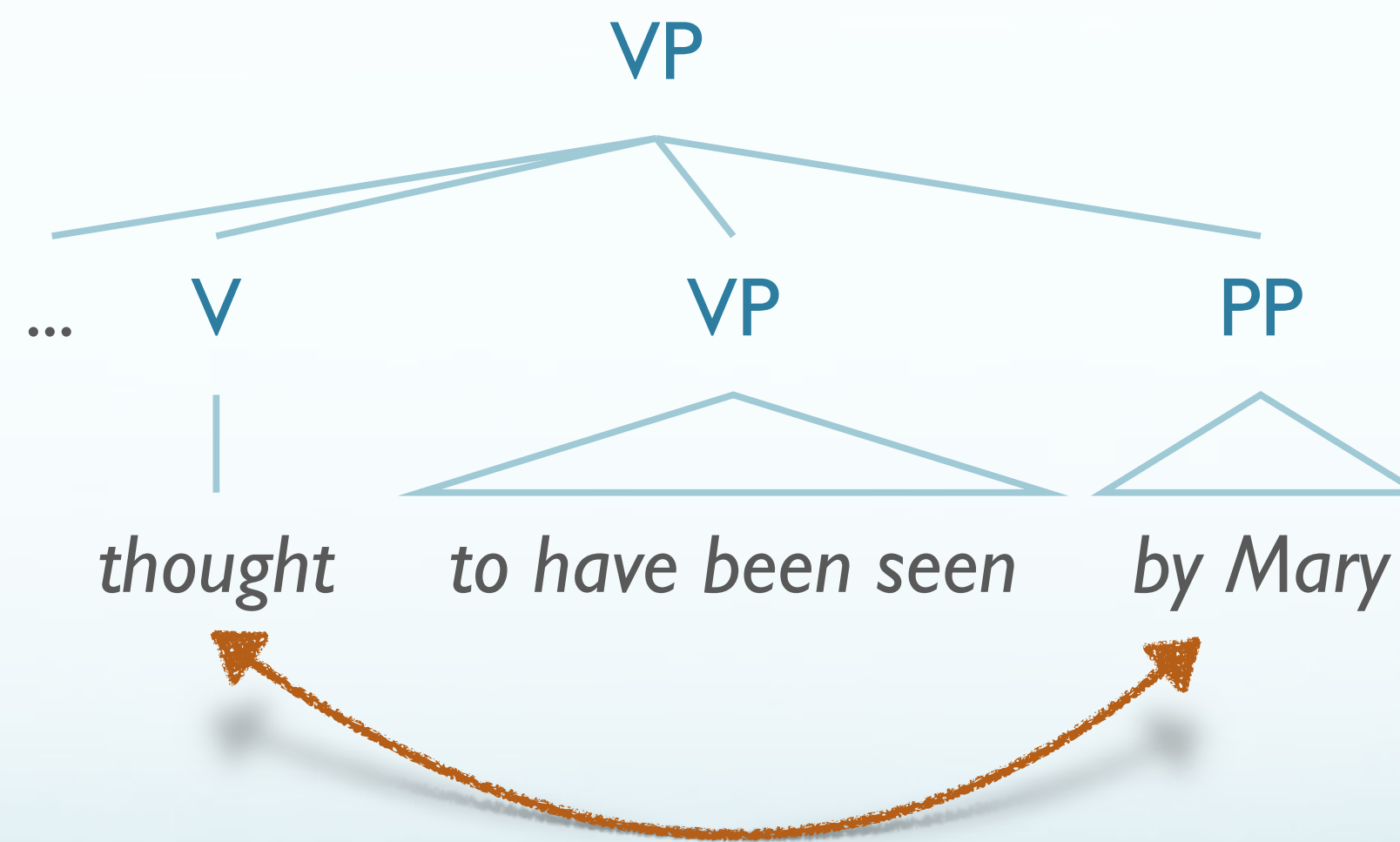
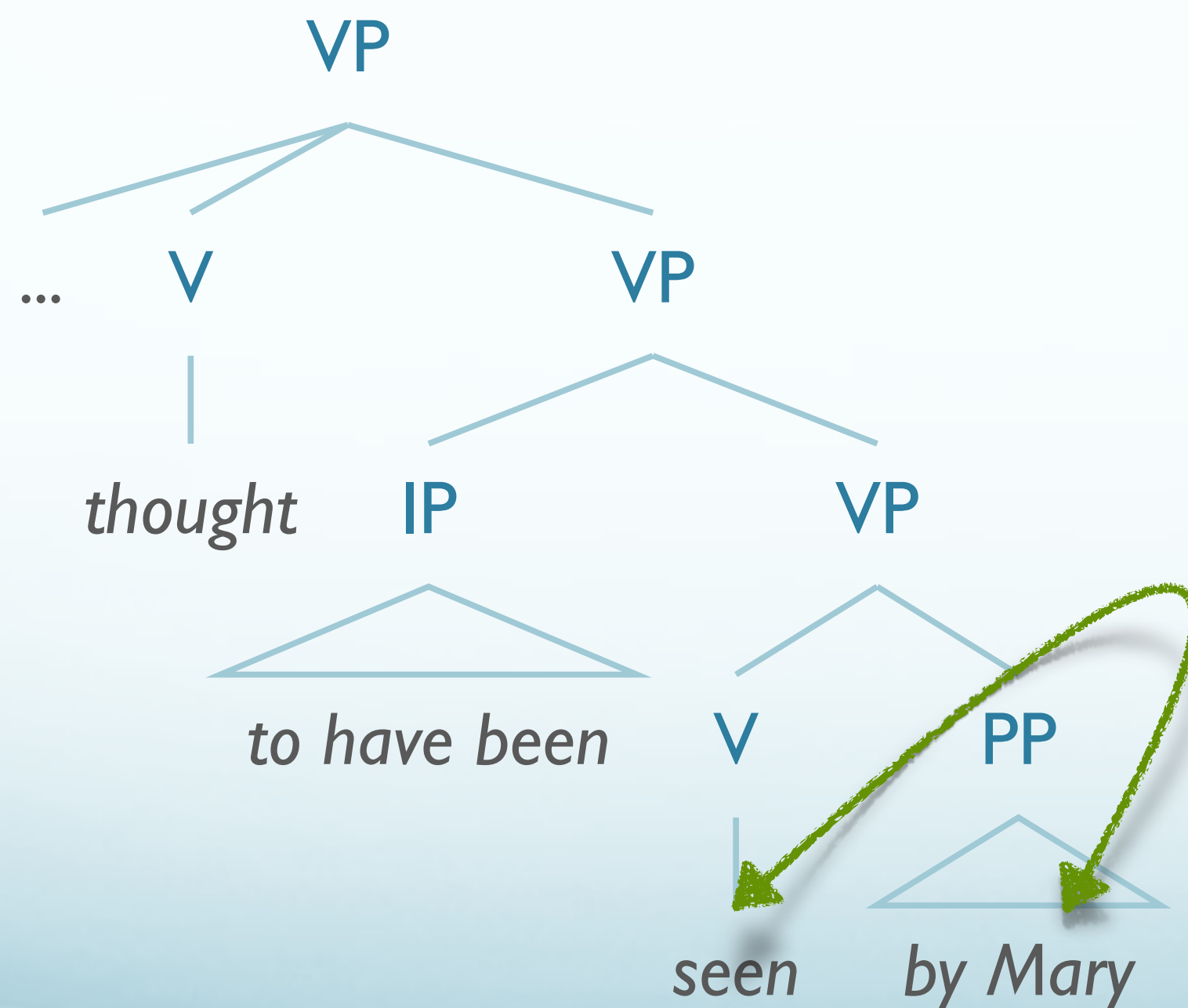
- *Local* ambiguity:
 - increased processing time
- *Global* ambiguity:
 - Would like to yield only “reasonable” parses
 - Ideally, the one that was intended*

Solution to Ambiguity?

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

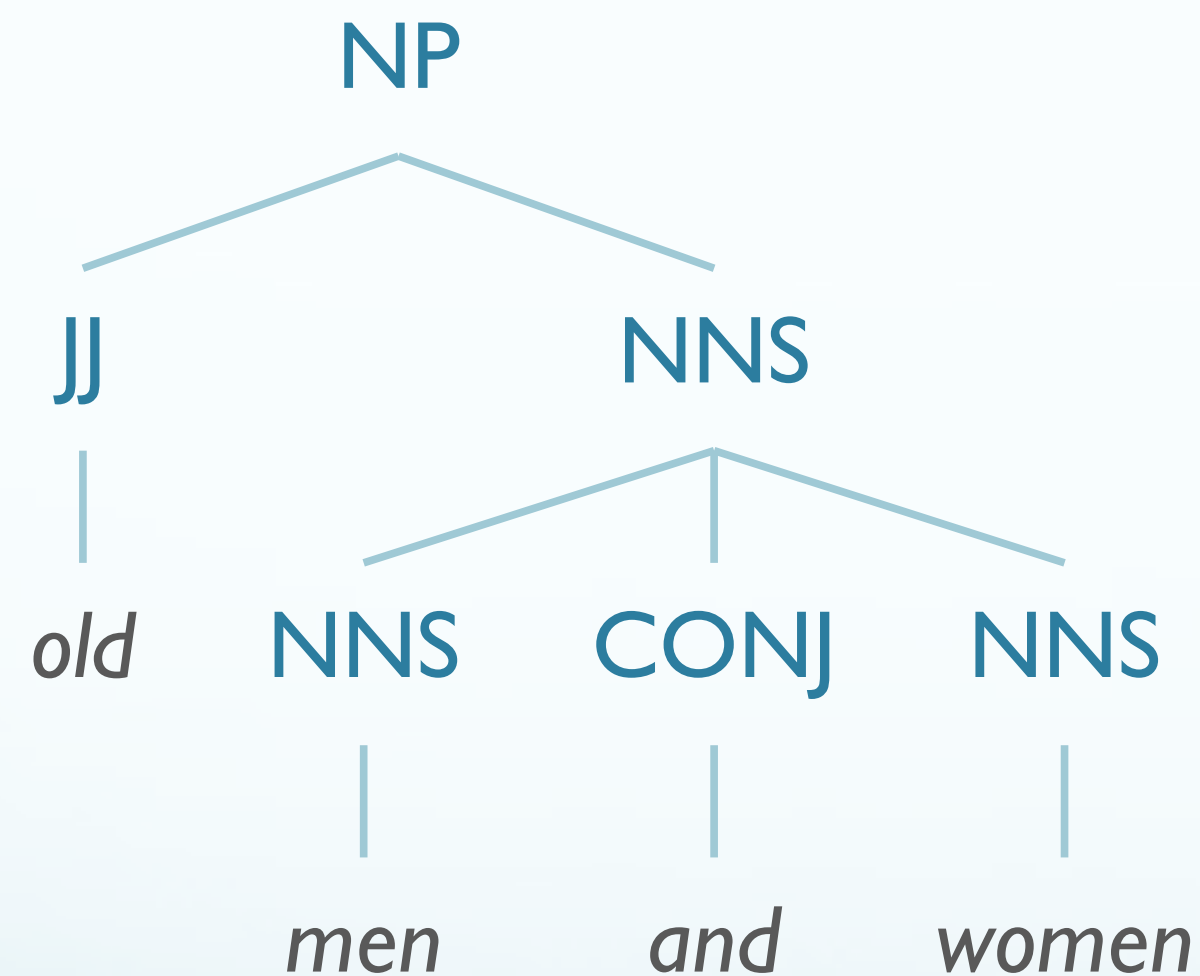
Disambiguation Strategy: Statistical

- Some prepositional structs more likely to attach high/low
- *John was thought to have been seen by Mary*
 - Mary could be doing the **seeing** or **thinking** — seeing more likely

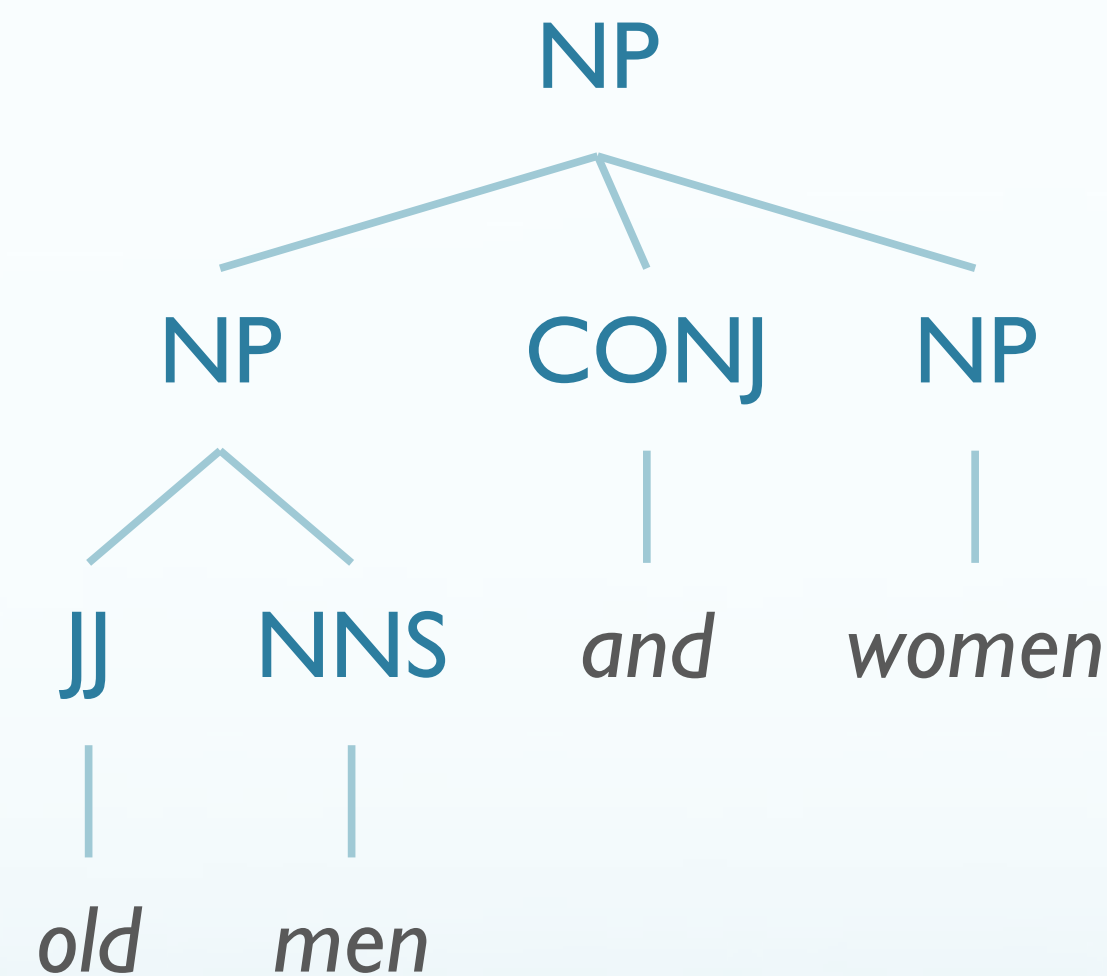


Disambiguation Strategy: Statistical

- 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
 - *Eiffel tower* as subject of *fly*

Disambiguation Strategy: **Pragmatic**

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

Disambiguation Strategy:



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

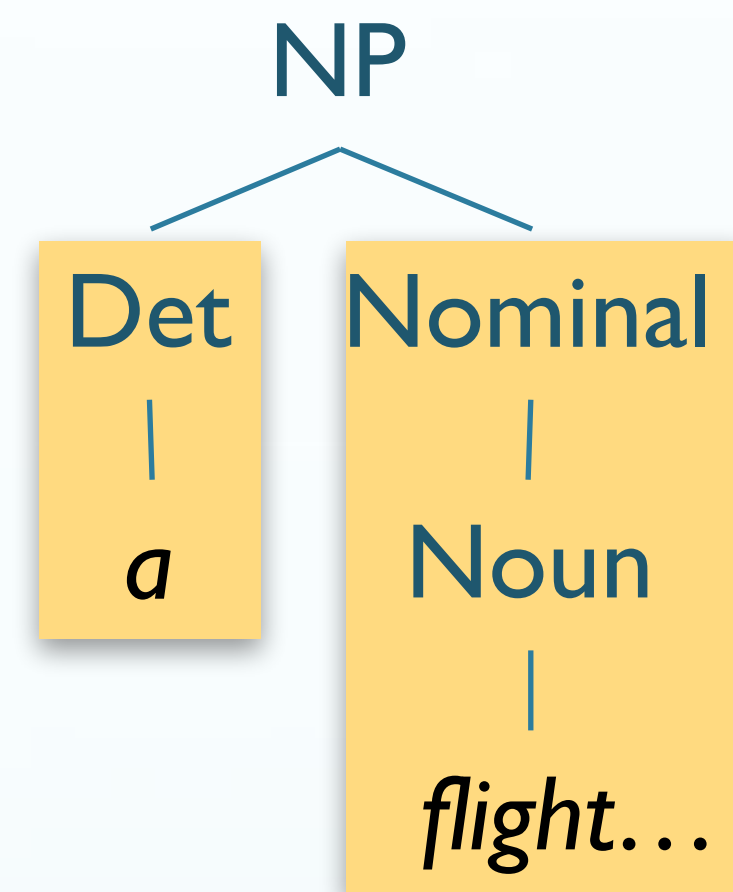
Parsing Challenges

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

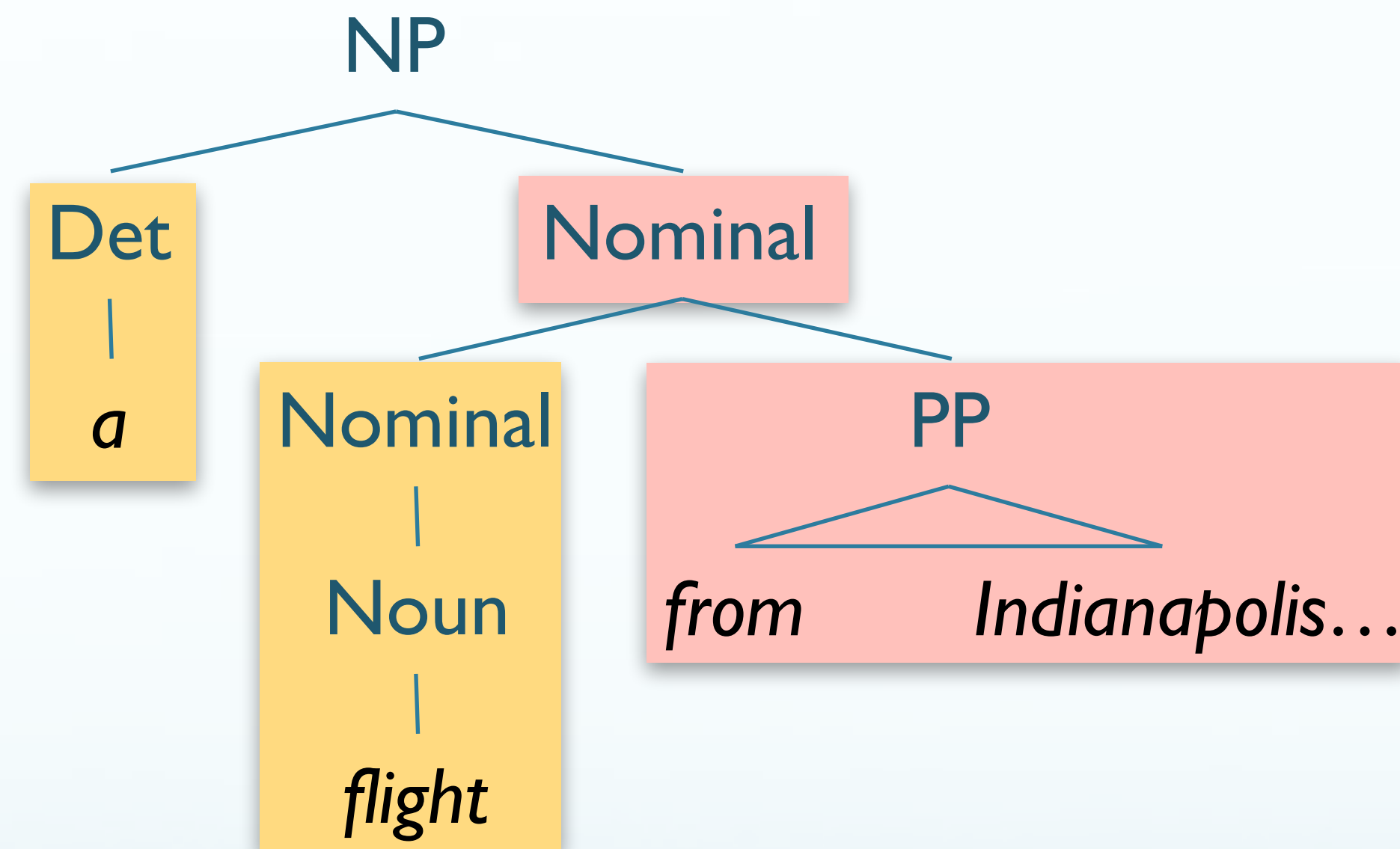
Repeated Work

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

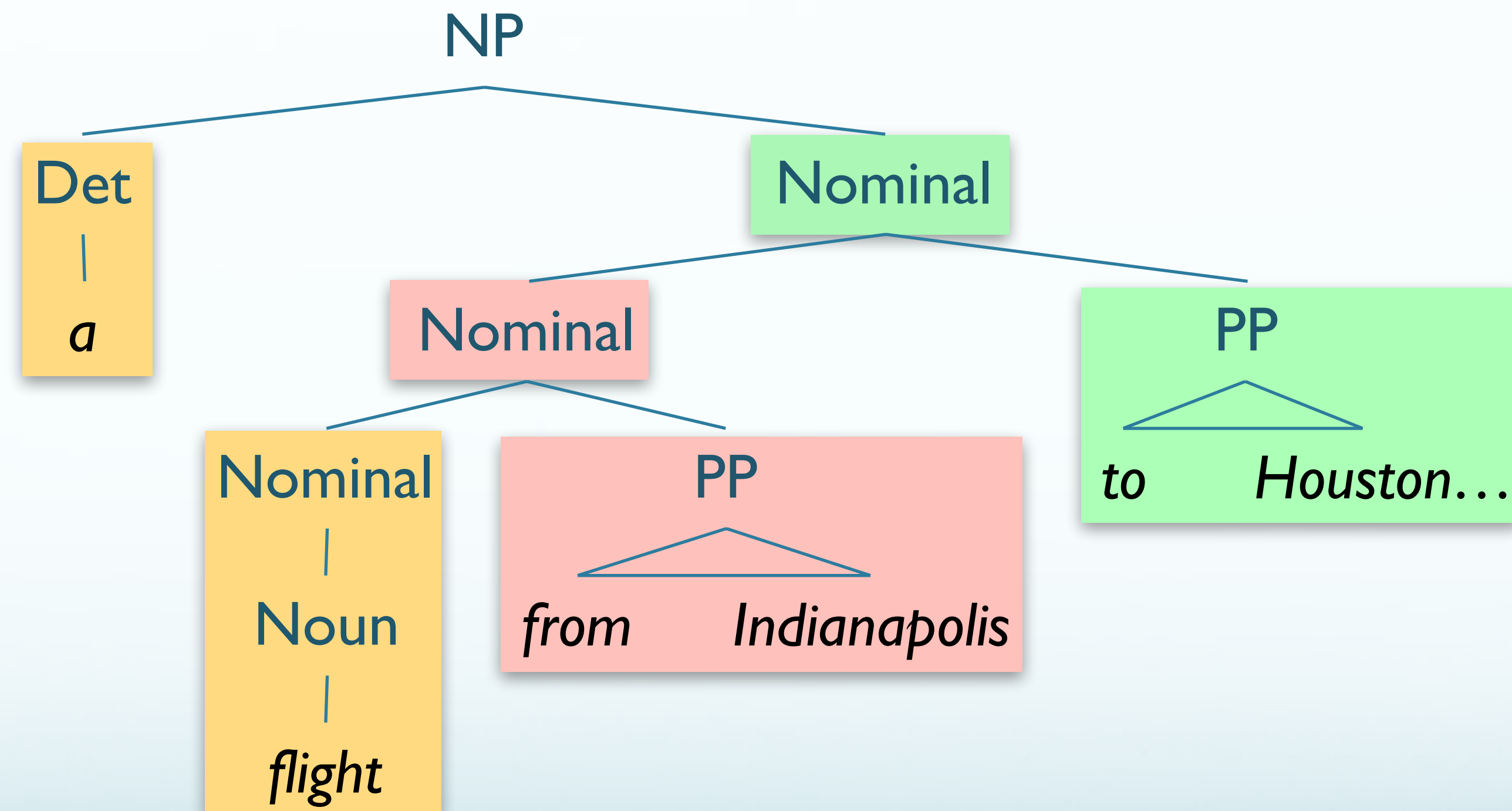
Shared Sub-Problems



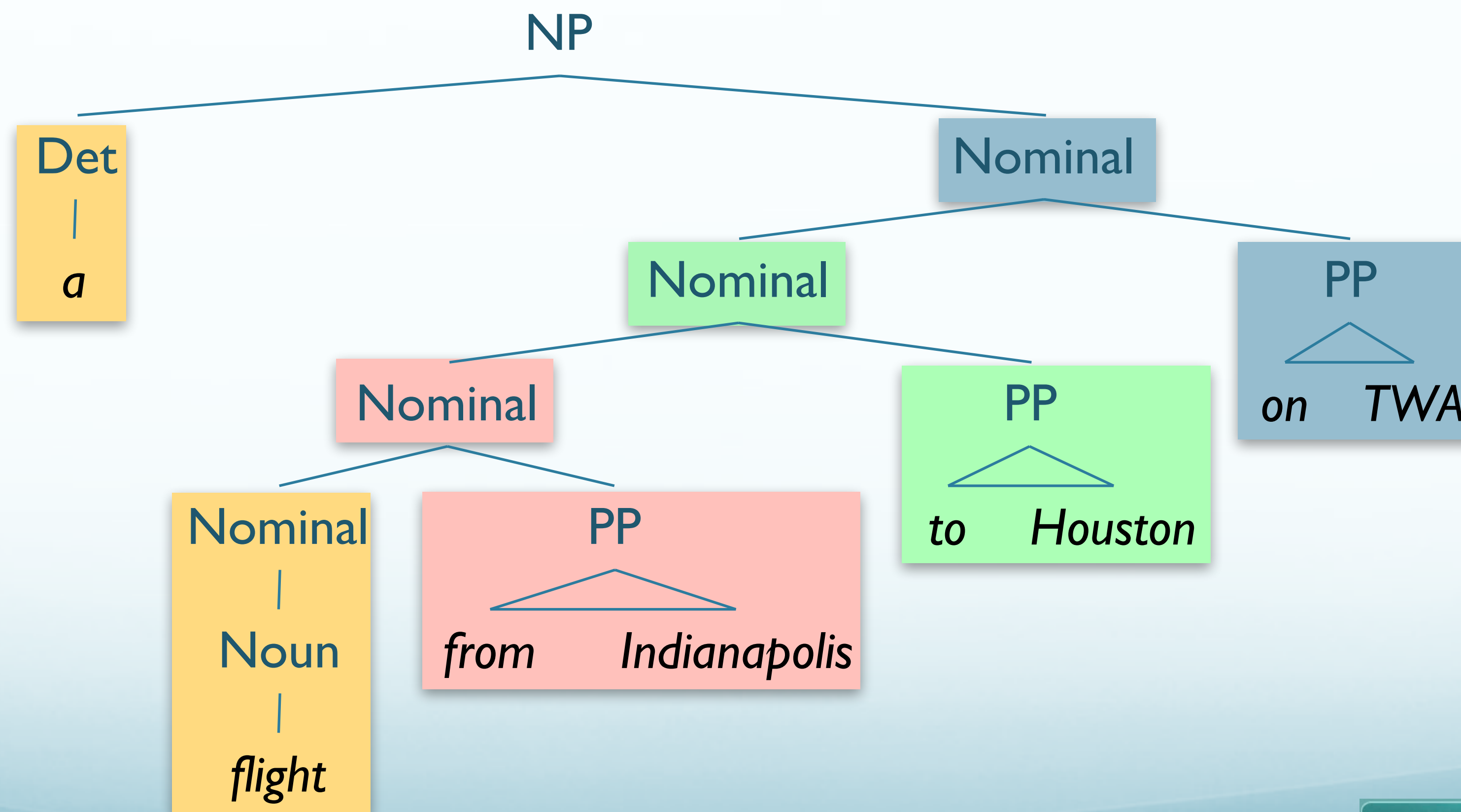
Shared Sub-Problems



Shared Sub-Problems



Shared Sub-Problems



Parsing Challenges

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 - Ambiguity
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 - **Recursion**
- Strategy: Dynamic Programming
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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

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

Dynamic Programming

- Challenge:
 - 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 w/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

- Recap: 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

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

Chomsky Normal Form (CNF)

- Required by CKY Algorithm
- All productions are of the form:
 - $A \rightarrow B C$
 - $A \rightarrow a$
- Most of our grammars are not of this form:
 - $S \rightarrow Wh-NP Aux NP VP$
- Need a general conversion procedure

CNF Conversion

1) Hybrid productions:

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

2) Unit productions:

$$A \rightarrow B$$

3) Long productions:

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

CNF Conversion: Hybrid Productions

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

CNF Conversion: Unit Productions

- Unit productions:
 - Rewrite RHS with RHS of all derivable, non-unit productions
 - If $A \xRightarrow{*} B$ and $B \rightarrow w$, add $A \rightarrow w$
- $Nominal \rightarrow Noun, Noun \rightarrow \mathbf{dog}$
 - $Nominal \rightarrow \mathbf{dog}$
 - $Noun \rightarrow \mathbf{dog}$

CNF Conversion: Long Productions

- Long productions
 - Introduce unique nonterminals, and spread over rules

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

CNF Conversion

- 1) Convert terminals in hybrid rules to dummy non-terminals
- 2) Convert unit productions
- 3) 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

$S \rightarrow NP VP$

$S \rightarrow X1 VP$

$X1 \rightarrow Aux NP$

$S \rightarrow book / include / prefer$

$S \rightarrow Verb NP$

$S \rightarrow X2 PP$

$S \rightarrow Verb PP$

$S \rightarrow VP PP$

$NP \rightarrow I / she / me$

$NP \rightarrow TWA / Houston$

$NP \rightarrow Det Nominal$

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

$Nominal \rightarrow Nominal Noun$

$Nominal \rightarrow Nominal PP$

$VP \rightarrow book / include / prefer$

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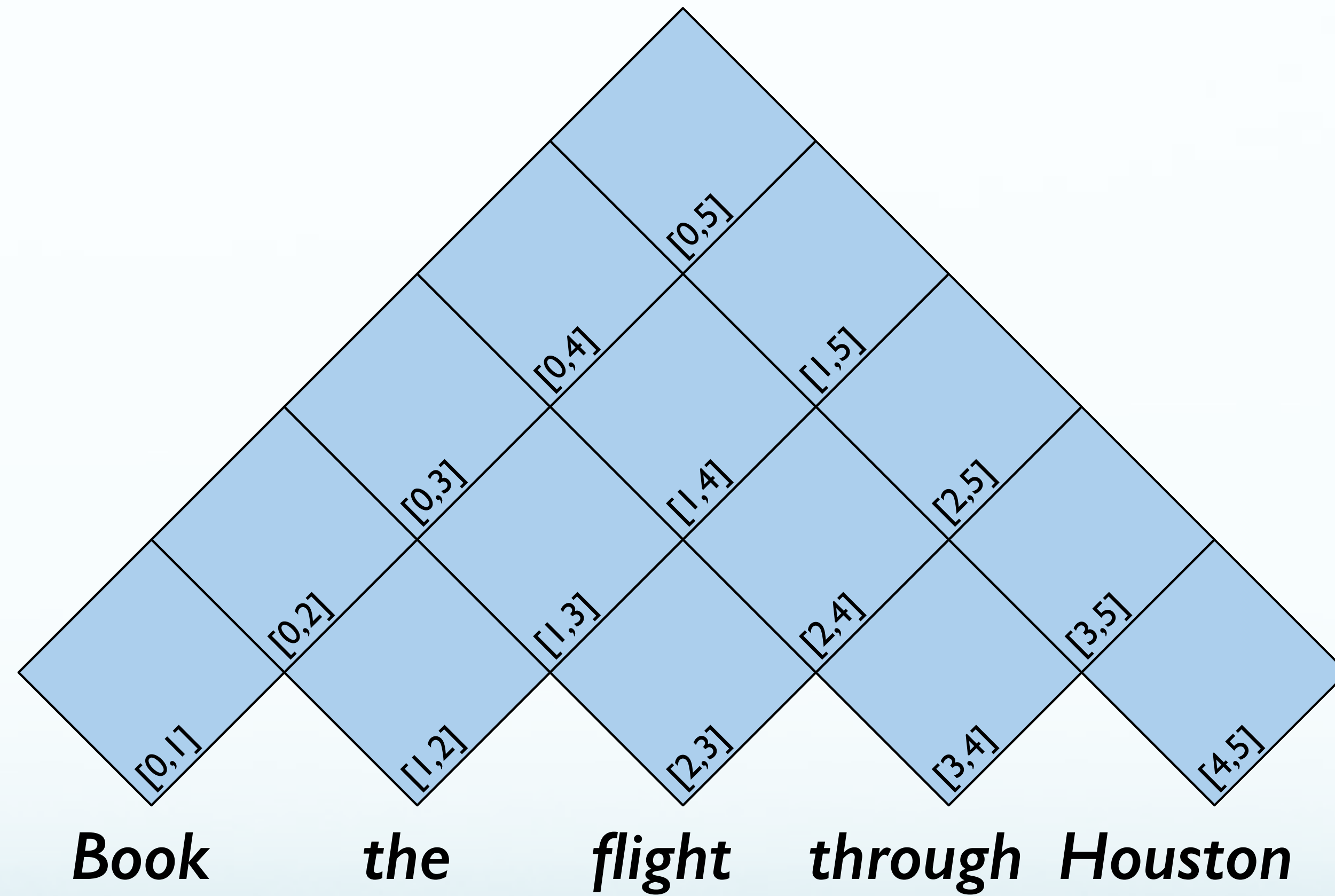
CKY Parsing

- (Relatively) efficient bottom-up 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

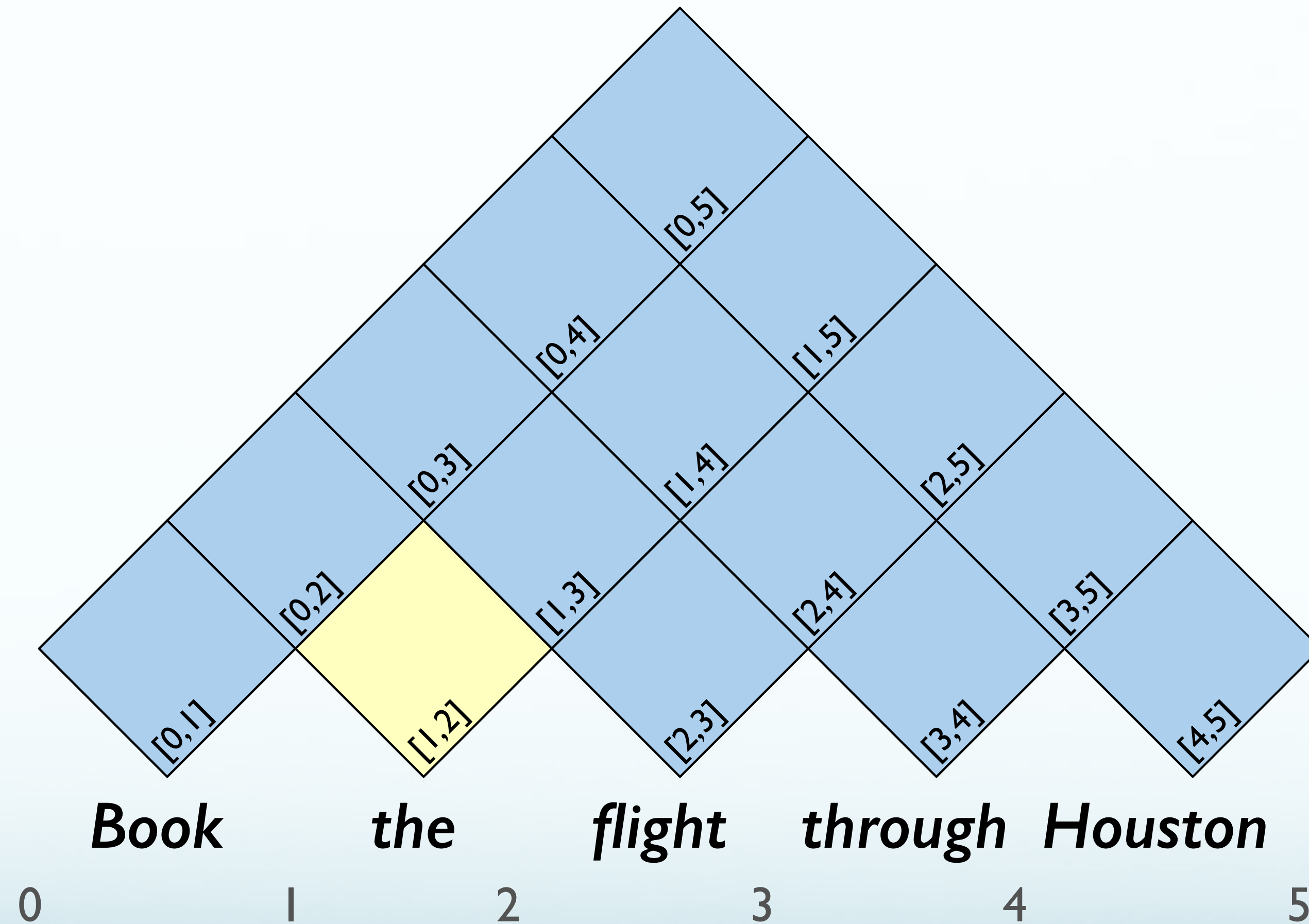
CKY Matrix

<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
[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]

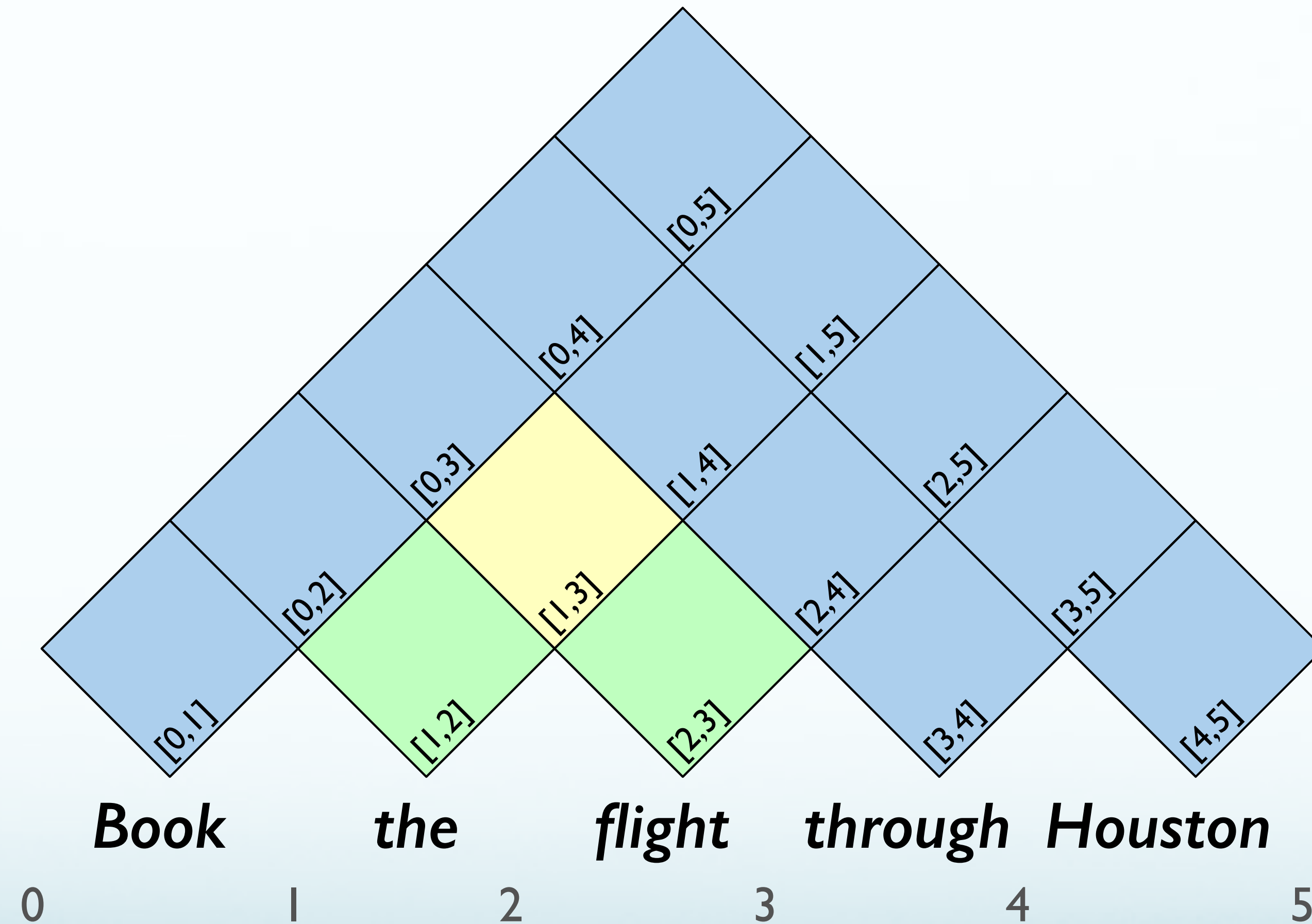
CKY Matrix



CKY Matrix



CKY Matrix



Dynamic Programming in CKY

- Key idea:
 - for $i < k < j$
 - ...and a parse spanning substring $[i, j]$
 - $\exists k$ 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 $[i, j]$ and C in $[k, j]$ for some $i < k < j$
 - CNF forces this for all $j > i + 1$

HW #2

LING 571

Deep Processing Techniques for NLP

January 10, 2018

Goals

- Begin development of CKY parser
- First stage: Conversion to CNF
 - Develop Representation for CFG
 - Manipulate/Transform Grammars
 - Investigate weakly equivalent grammars

Task

- Conversion:
 - Read in grammar rules from arbitrary CFG
 - Convert to CNF
 - Write out new grammar
- Validation:
 - Parse test sentences with original CFG
 - Parse test sentences with CFG in CNF

Approach

- May use any programming language
 - In keeping with [course policies](#)
- May use existing models/packages to represent rules
 - Need RULE, RHS, LHS, etc
 - NLTK, Stanford
- ***Conversion code must be your own***

Data

- ATIS (Air Travel Information System) data
 - Grammar provided in nltk-data
 - Terminals in double-quotes
 - *the* → “the”
 - All required files on patas dropbox
- **NOTE:**
 - Grammar is fairly large (~193K Productions)
 - Grammar is fairly ambiguous (Test sentences may have 100 parses)
 - You will likely want to develop against a smaller grammar

NLTK Grammars

```
>>> gr1 = nltk.data.load('grammars/large_grammars/atis.cfg')
```

```
>>> gr1.productions()[0]
```

```
ABBCL_NP -> QUANP_DTI QUANP_DTI QUANP_CD AJP_JJ NOUN_NP  
PRPRTCL_VBG
```

```
>>> gr1.productions()[0].lhs()
```

```
ABBCL_NP
```

```
>>> gr1.productions(lhs=gr1.productions()[1].lhs())
```

```
[ADJ_ABL -> only, ADJ_ABL->such]
```