

“自然语言处理导论”课程讲义

# 句法分析1： 上下文无关文法(Context Free Grammar)

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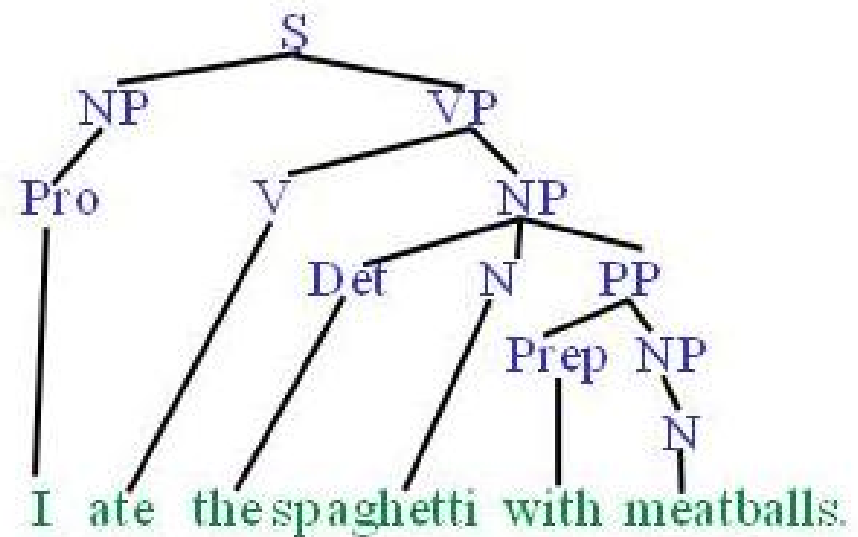
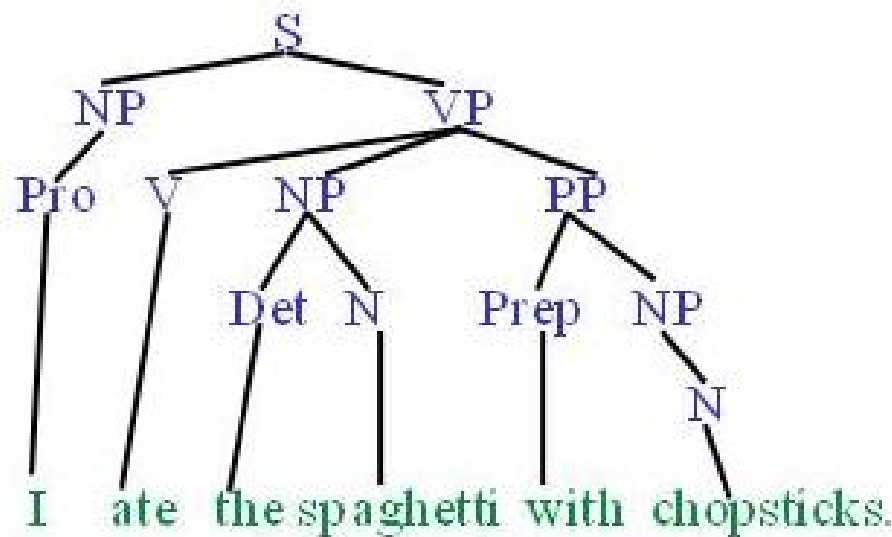
<http://klcl.pku.edu.cn/member/sunxu/index.htm>

# 上下文无关文法(Context Free Grammar)

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# 句法分析(Syntactic Parsing)

- 给定一个句子，输出其正确的句法结构树



# 上下文无关文法 (CFG)

- $N$  是非终结符集合
- $\Sigma$  是终结符集合
- $R$  是一个规则集合
  - 比如一个规则是  $A \rightarrow \beta$ , 其中  $A$  是非终结符, 而  $\beta$  可以是非终结符或终结符
- $S$  是一个特殊非终结符, 代表树结构的根节点, 称为 *start symbol*

## 语法规则

**S** → **NP VP**

**S** → **Aux NP VP**

**S** → **VP**

**NP** → **Pronoun**

**NP** → **Proper-Noun**

**NP** → **Det Nominal**

**Nominal** → **Noun**

**Nominal** → **Nominal Noun**

**Nominal** → **Nominal PP**

**VP** → **Verb**

**VP** → **Verb NP**

**VP** → **VP PP**

**PP** → **Prep NP**

## 词汇规则

**Det** → **the | a | that | this**

**Noun** → **book | flight | meal | money**

**Verb** → **book | include | prefer**

**Pronoun** → **I | he | she | me**

**Proper-Noun** → **Houston | NWA**

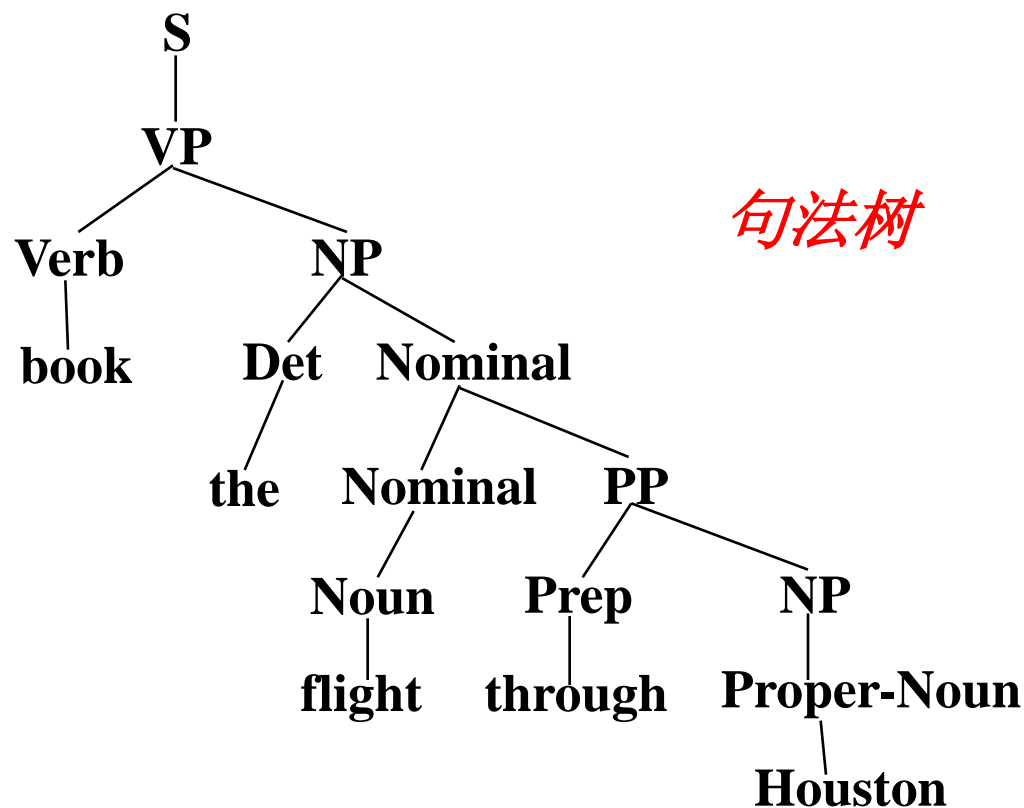
**Aux** → **does**

**Prep** → **from | to | on | near | through**

# 从句法分析看句子是怎么生成的

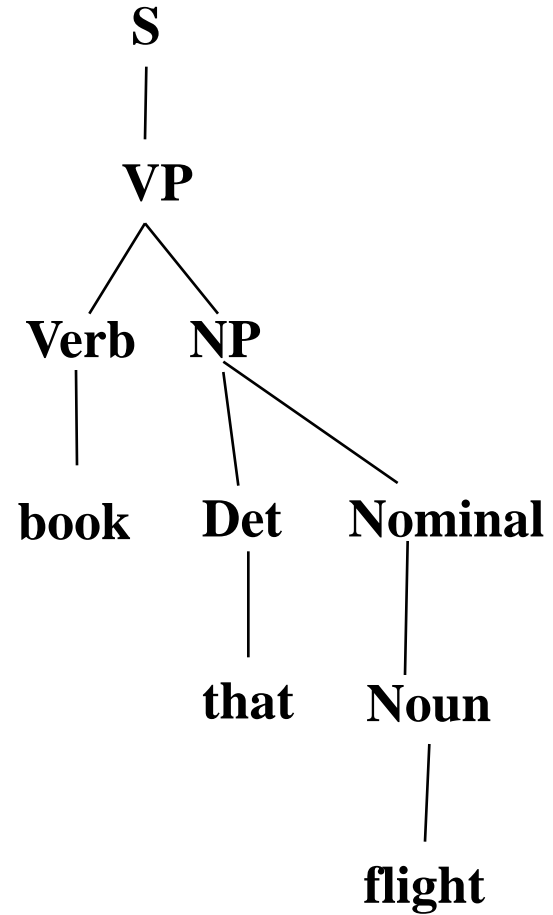
- 从句法分析的角度来看，一个句子是从根节点开始递归生成句法规则的一个过程，直到最后只有终结符存在。

**两大类方法：**  
自上而下句法分析  
自下而上句法分析

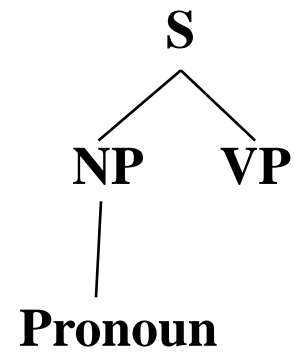


# 句法分析举例

book that flight

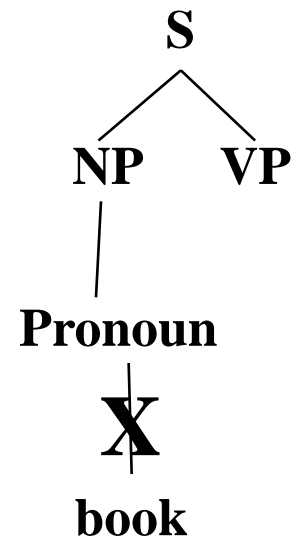


# 自上而下句法分析(Top Down Parsing)

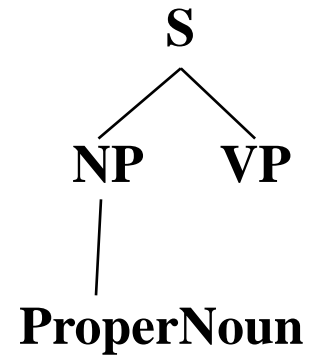




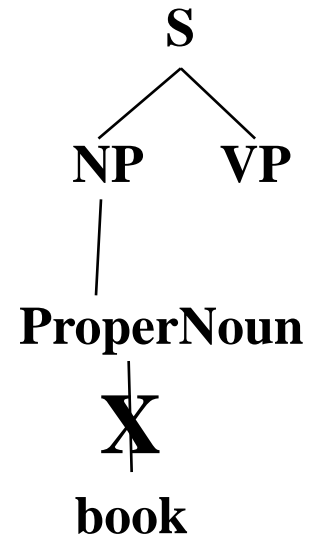
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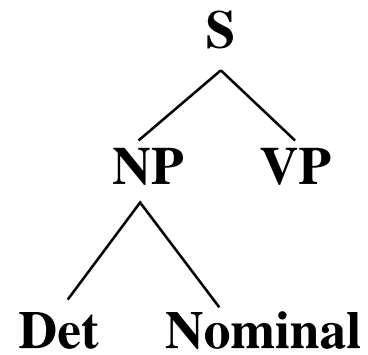
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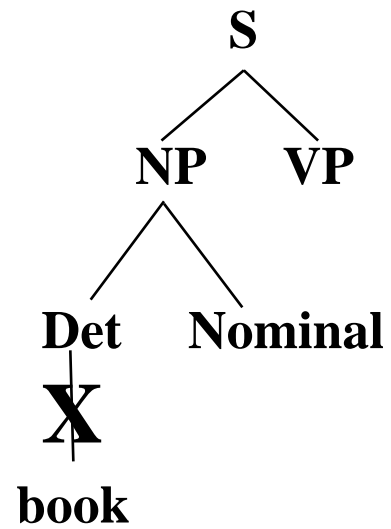
# 自上而下句法分析(Top Down Parsing)



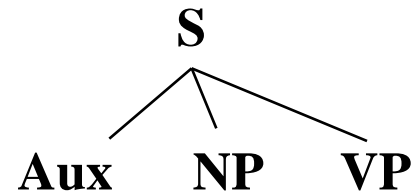
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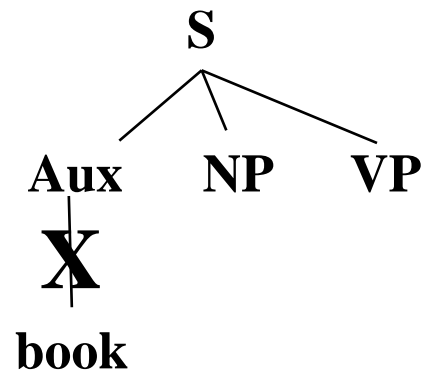
# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)

**S**  
|  
**VP**



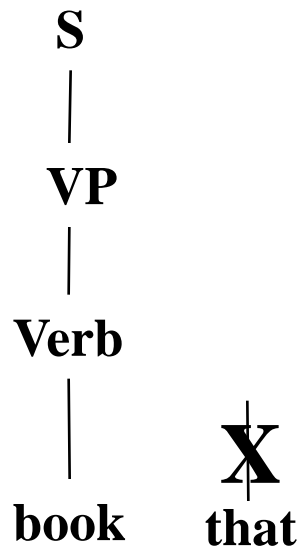
# 自上而下句法分析(Top Down Parsing)

**S**  
|  
**VP**  
|  
**Verb**

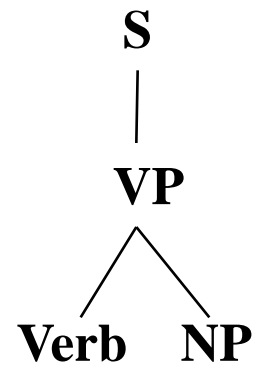
# 自上而下句法分析(Top Down Parsing)



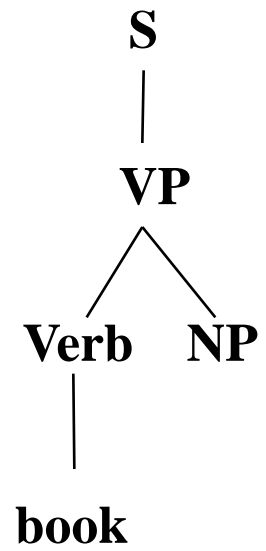
# 自上而下句法分析(Top Down Parsing)



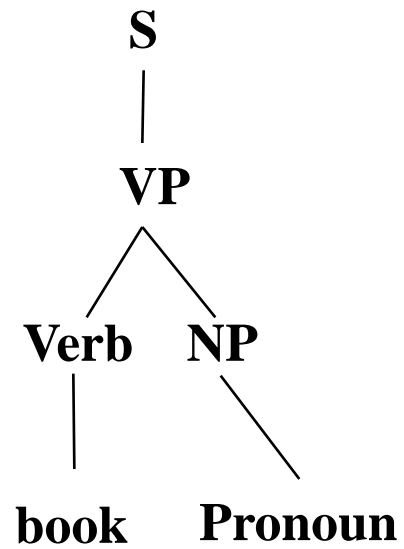
# 自上而下句法分析(Top Down Parsing)



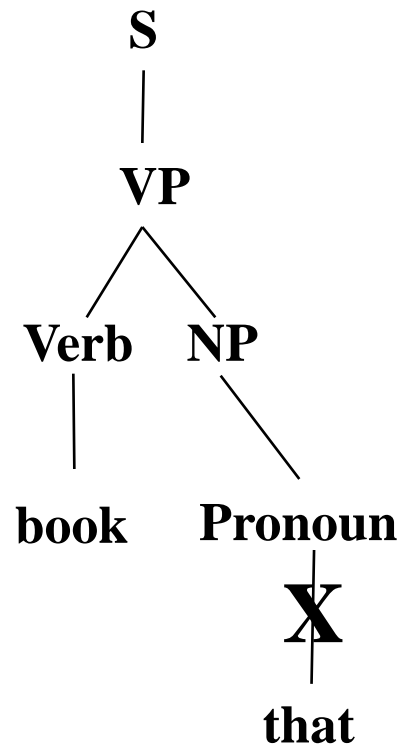
# 自上而下句法分析(Top Down Parsing)



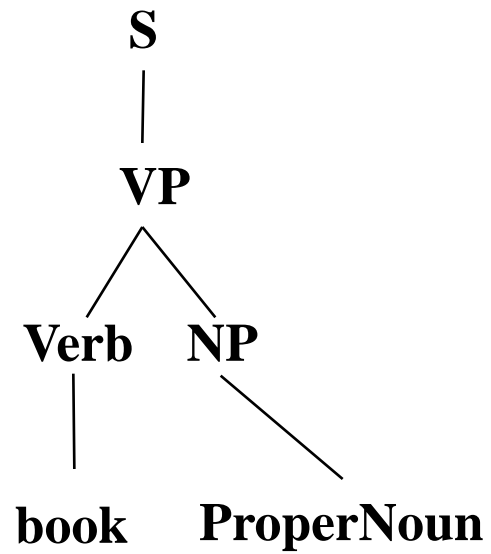
# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)

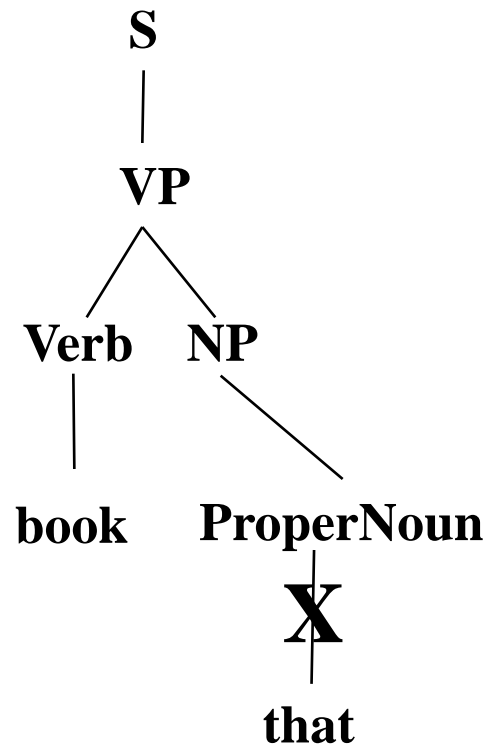


# 自上而下句法分析(Top Down Parsing)

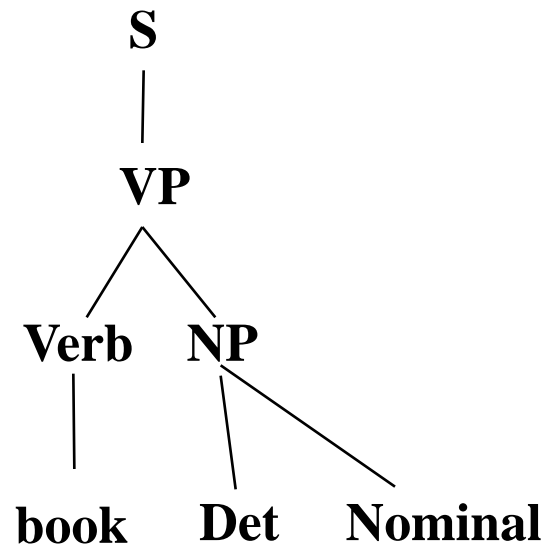




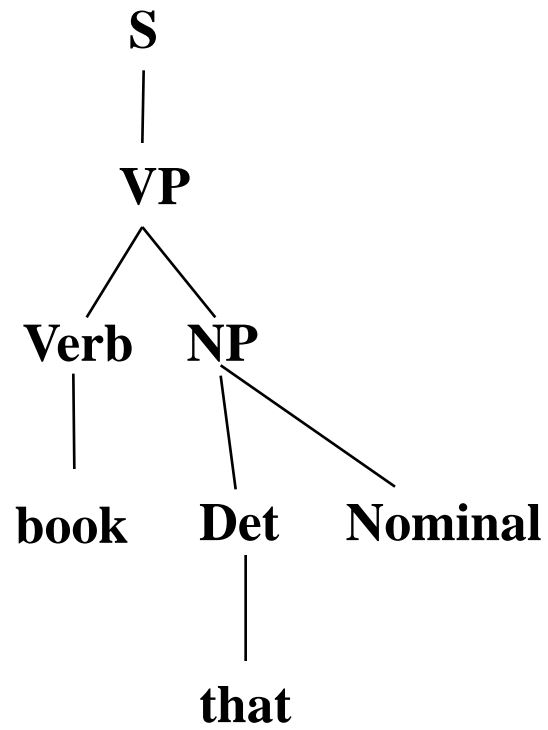
# 自上而下句法分析(Top Down Parsing)



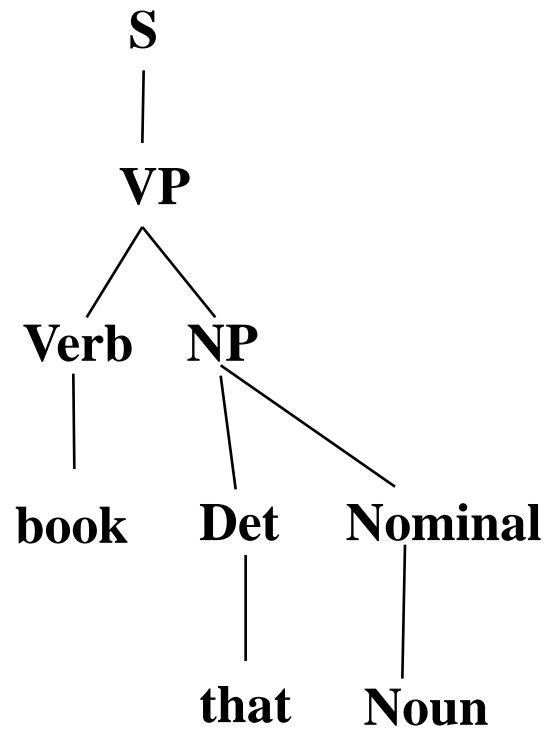
# 自上而下句法分析(Top Down Parsing)



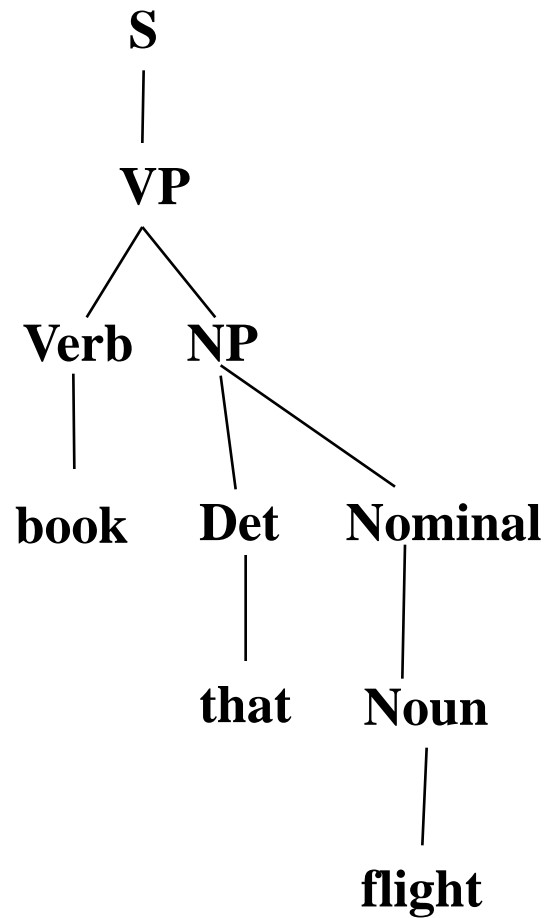
# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)



# 自上而下句法分析(Top Down Parsing)



# 自下而上句法分析(Bottom Up Parsing)

**book**      **that**      **flight**

# 自下而上句法分析(Bottom Up Parsing)

**Noun**

|  
**book**

**that**

**flight**

# 自下而上句法分析(Bottom Up Parsing)

**Nominal**



**Noun**



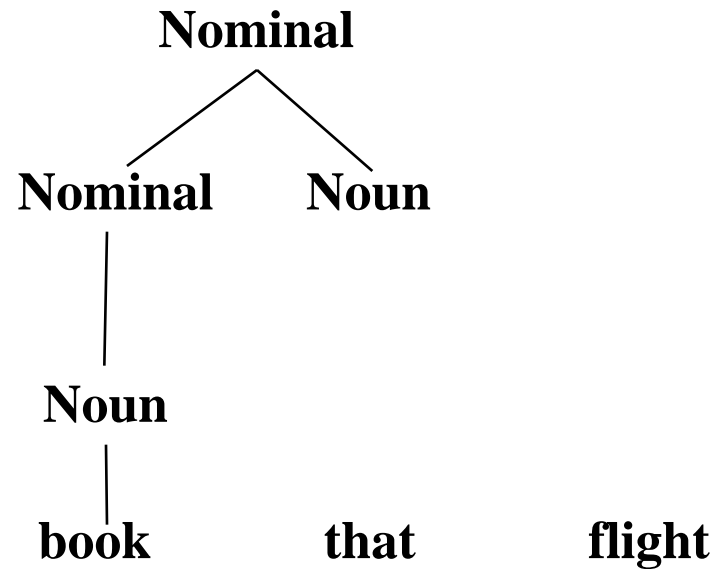
**book**

**that**

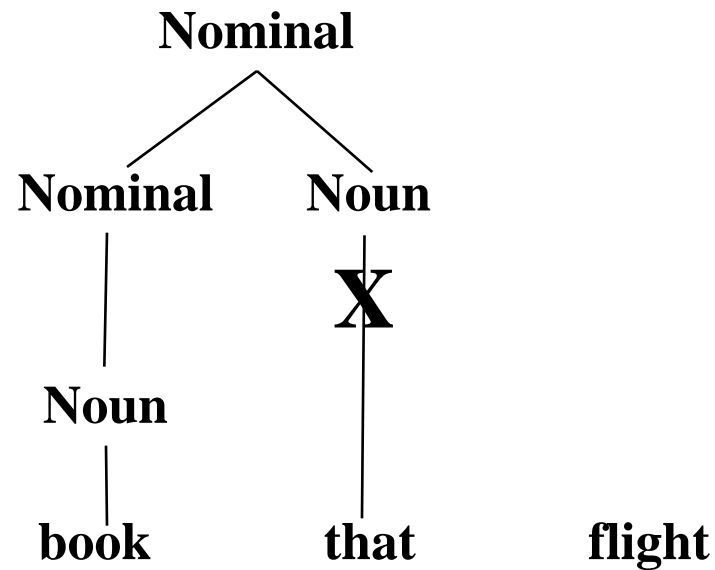
**flight**



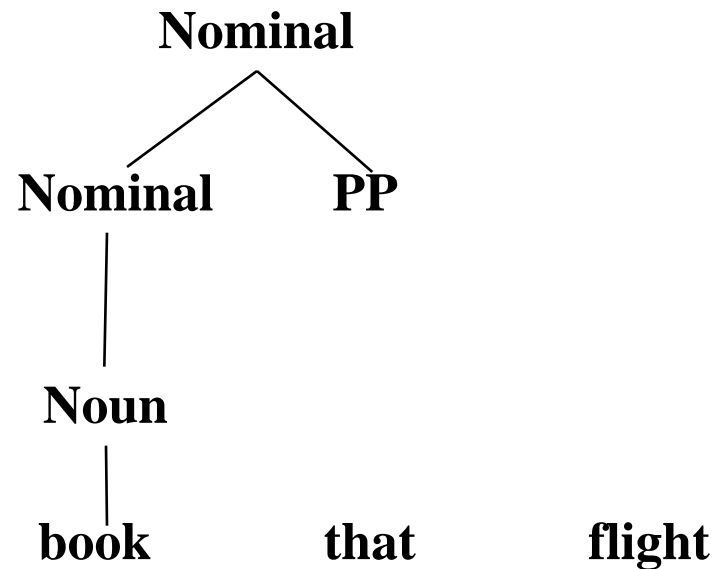
# 自下而上句法分析(Bottom Up Parsing)



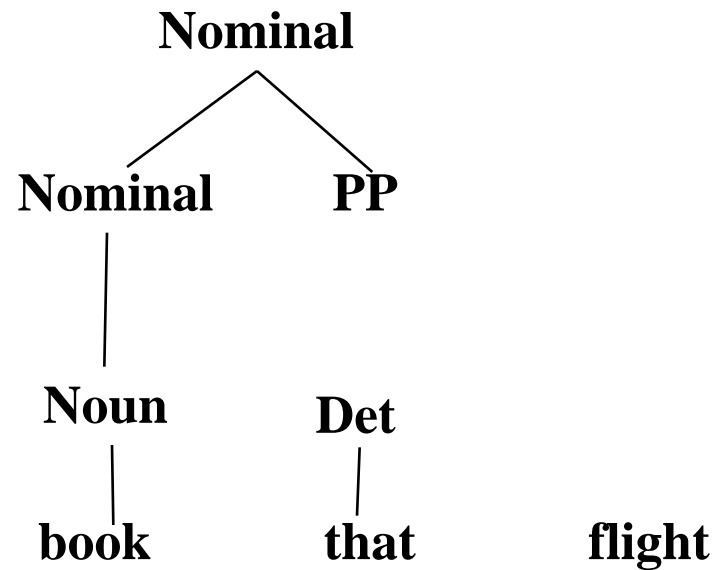
# 自下而上句法分析(Bottom Up Parsing)



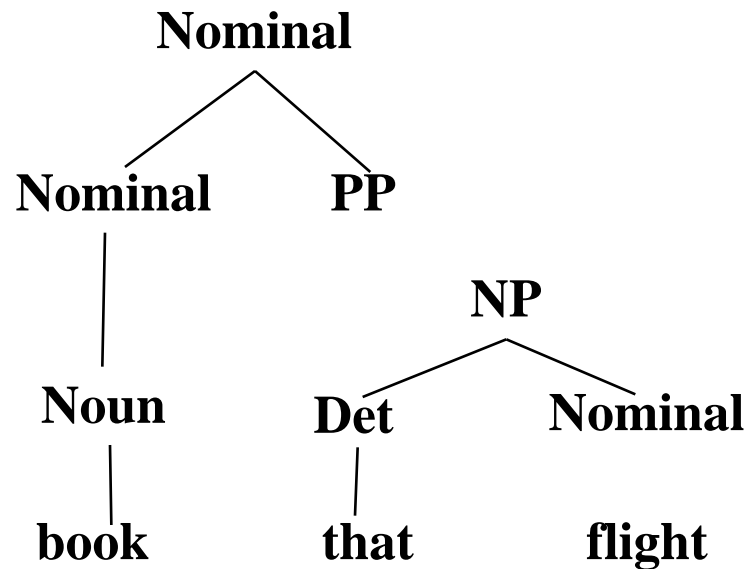
# 自下而上句法分析(Bottom Up Parsing)



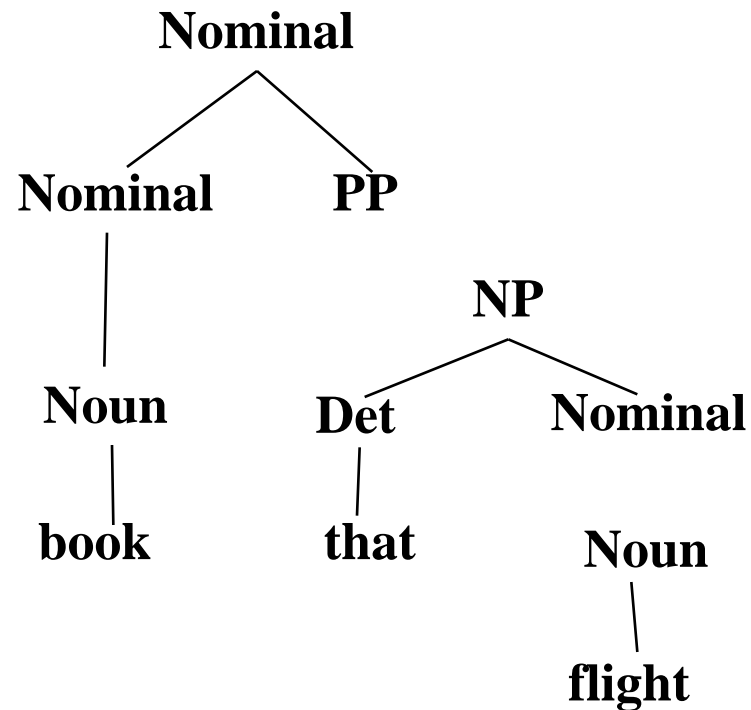
# 自下而上句法分析(Bottom Up Parsing)



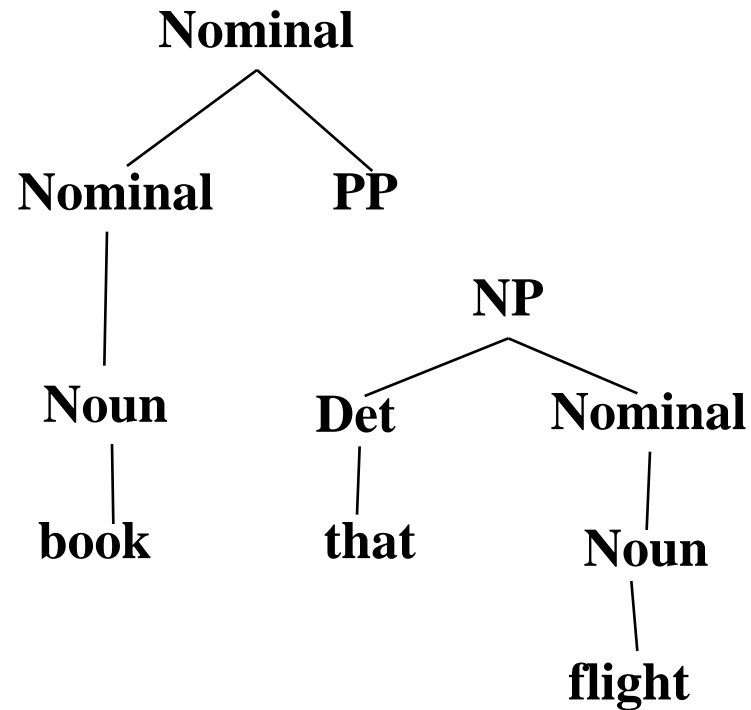
# 自下而上句法分析(Bottom Up Parsing)



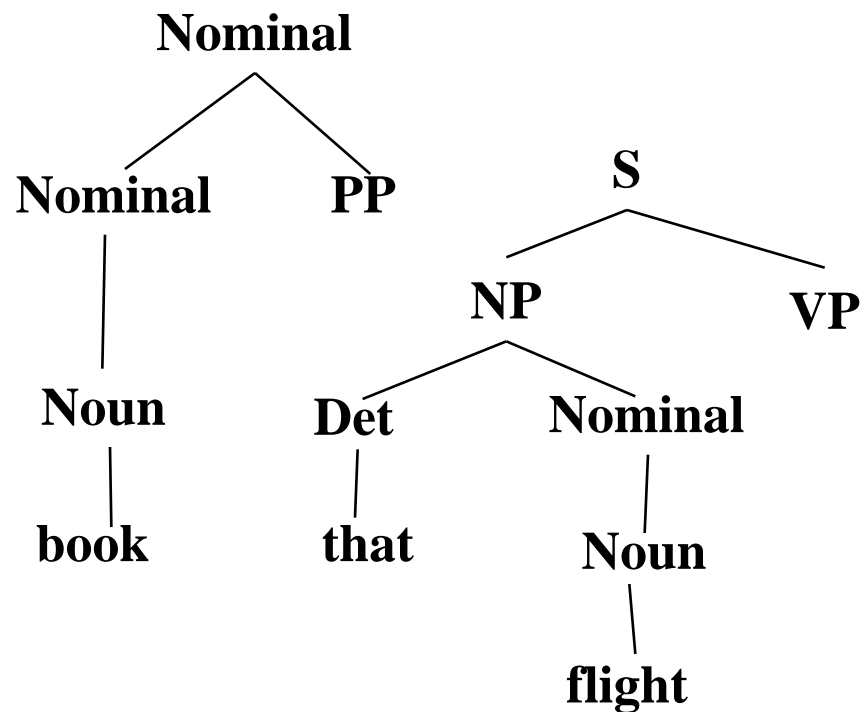
# 自下而上句法分析(Bottom Up Parsing)



# 自下而上句法分析(Bottom Up Parsing)

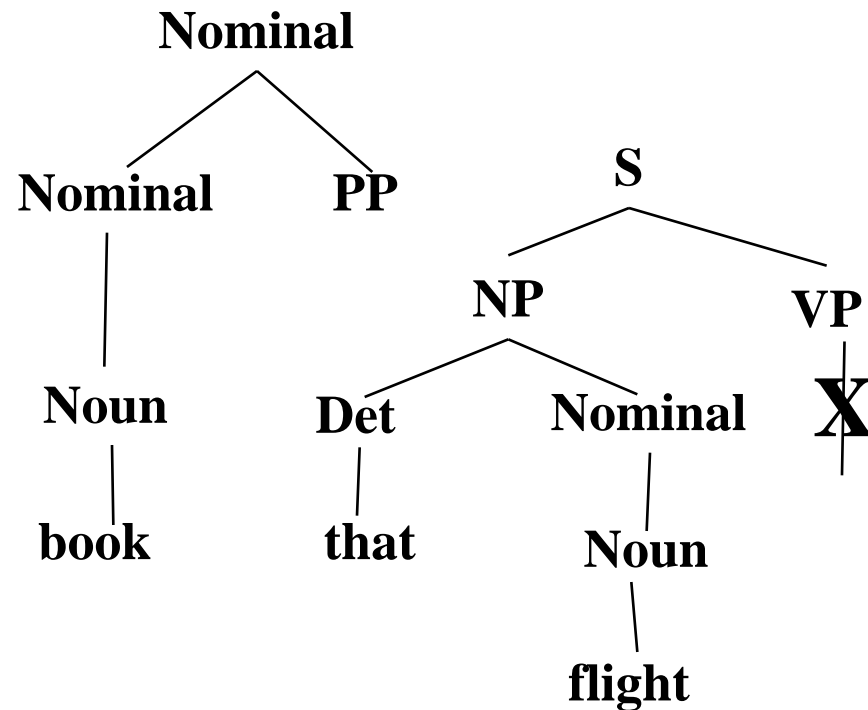


# 自下而上句法分析(Bottom Up Parsing)

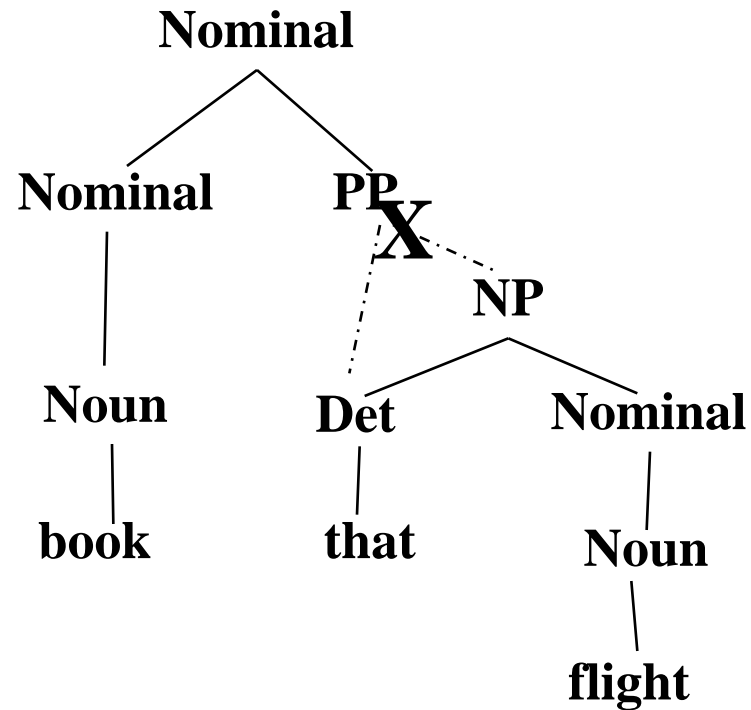




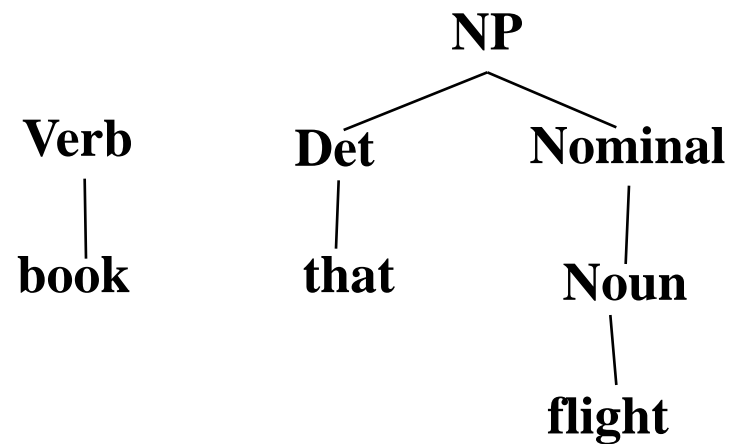
# 自下而上句法分析(Bottom Up Parsing)



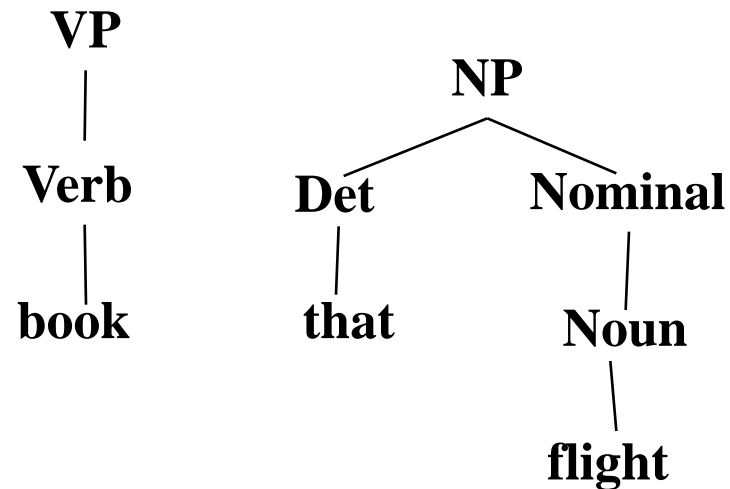
# 自下而上句法分析(Bottom Up Parsing)



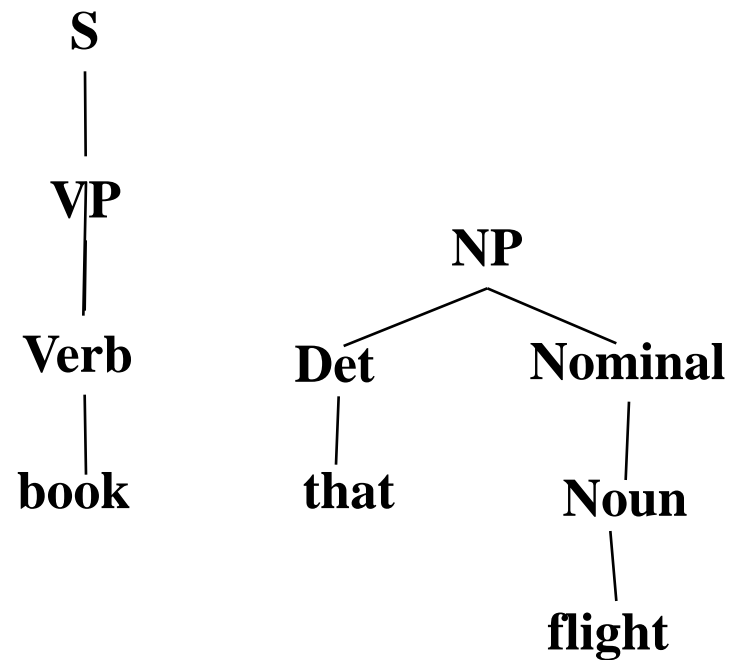
# 自下而上句法分析(Bottom Up Parsing)



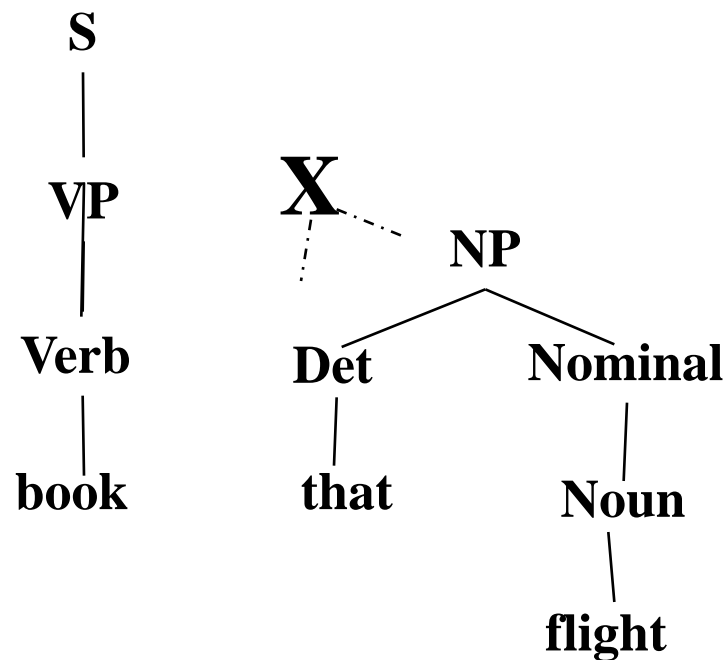
# 自下而上句法分析(Bottom Up Parsing)



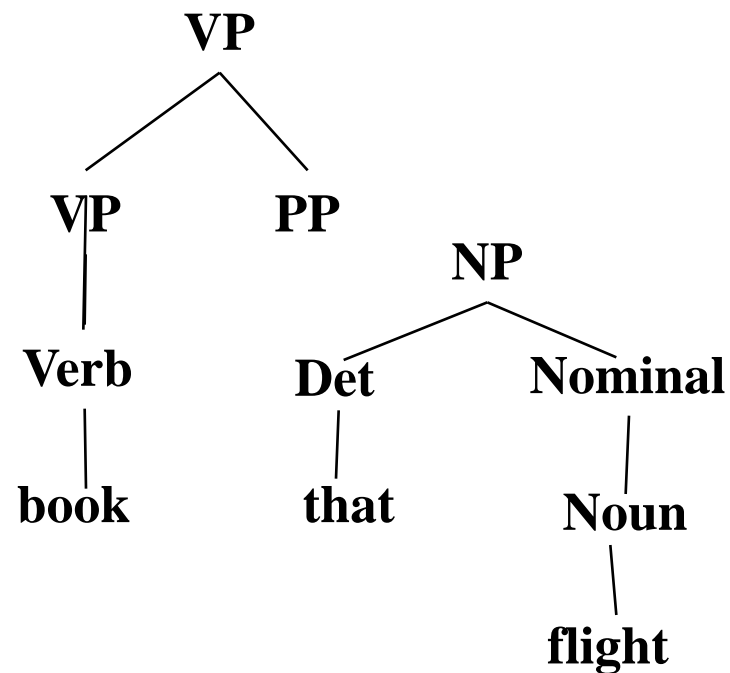
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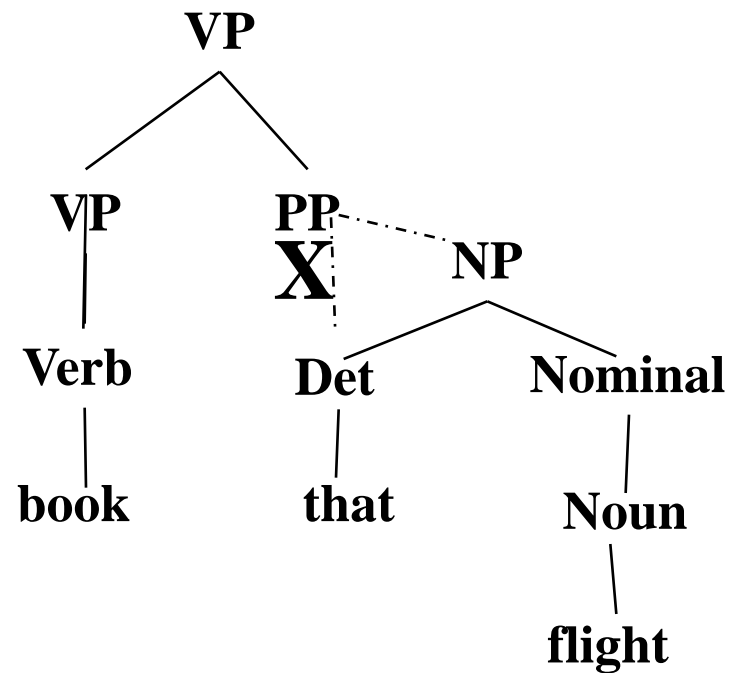
# 自下而上句法分析(Bottom Up Parsing)



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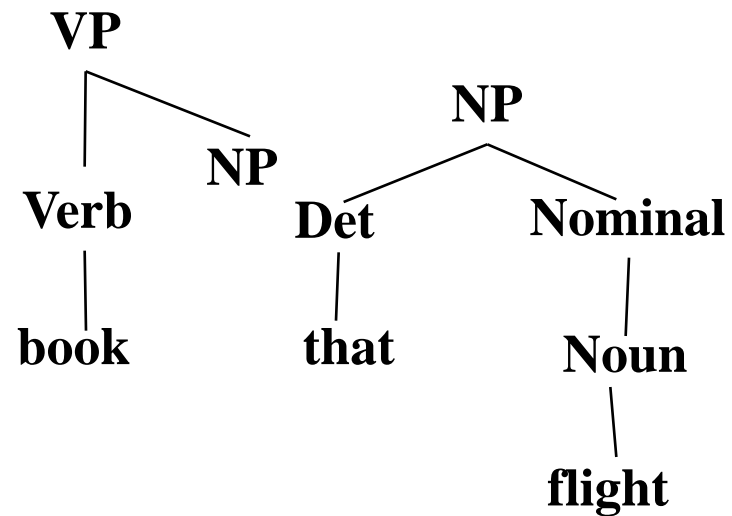


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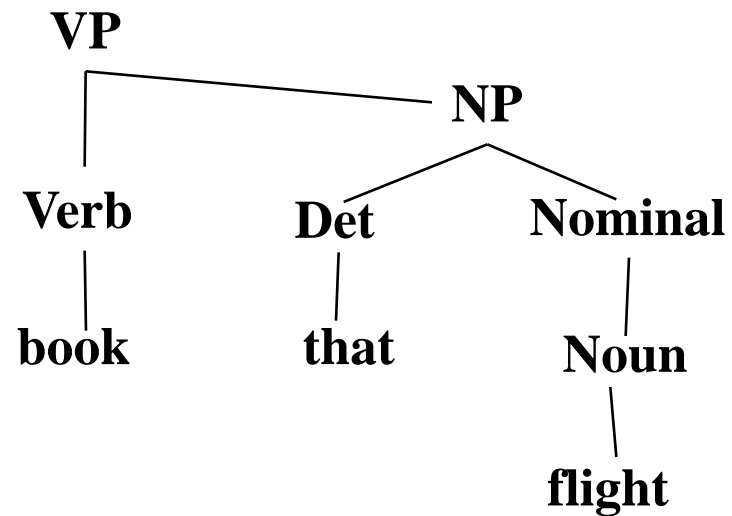




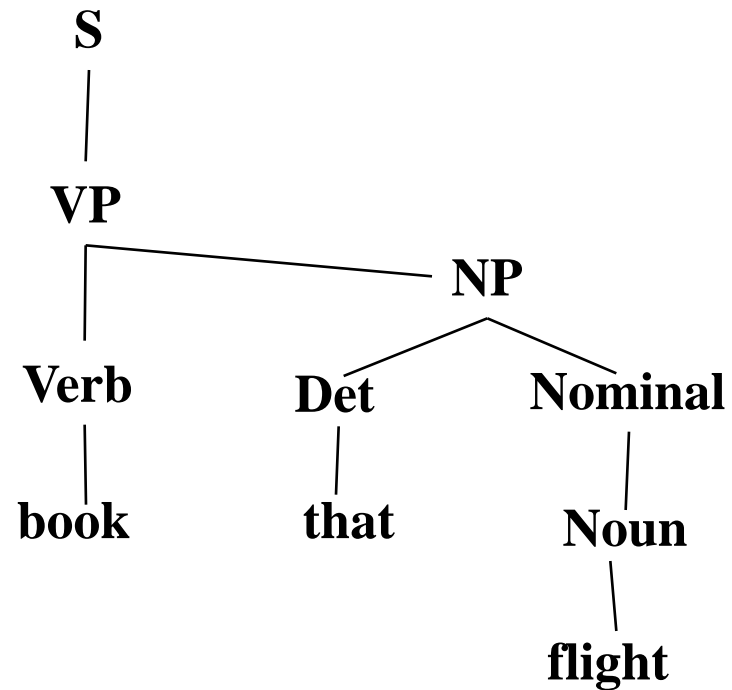
# 自下而上句法分析(Bottom Up Parsing)



# 自下而上句法分析(Bottom Up Parsing)



# 自下而上句法分析(Bottom Up Parsing)



- **自上而下**方法的结果总是一个完整的句法树，但是有可能无法匹配实际的句子
- **自下而上**方法总是能够匹配实际的句子，但是有可能不是一个完整的句法树
- 采用动态规划的方法，这两种方法都可以实现 $O(n^3)$ 的句法分析复杂度， $n$ 是句子长度

- **CKY (Cocke-Kasami-Younger) 算法**是基于自底向上的动态规划句法分析方法
- **Earley 算法**是基于自顶向下的句法分析方法
- **Chart 算法**在图表(**chart**)里面保存了完整的短语信息，可以把自顶向下和自底向上的方法结合起来

## ■ 乔姆斯基范式

- 首先，CKY方法需要把不规则语法转换成乔姆斯基范式 (Chomsky normal form, CNF)
- 在乔姆斯基范式里，一个生成规则必须生成2个非终结符，或者1个终结符

## 原始语法

**S → NP VP**

**S → Aux NP VP**

**S → VP**

**NP → Pronoun**

**NP → Proper-Noun**

**NP → Det Nominal**

**Nominal → Noun**

**Nominal → Nominal Noun**

**Nominal → Nominal PP**

**VP → Verb**

**VP → Verb NP**

**VP → VP PP**

**PP → Prep NP**

## 乔姆斯基范式

**S → NP VP**

**S → X1 VP**

**X1 → Aux NP**

**S → book | include | prefer**

**S → Verb NP**

**S → VP PP**

**NP → I | he | she | me**

**NP → Houston | NWA**

**NP → Det Nominal**

**Nominal → book | flight | meal | money**

**Nominal → Nominal Noun**

**Nominal → Nominal PP**

**VP → book | include | prefer**

**VP → Verb NP**

**VP → VP PP**

**PP → Prep NP**

	Book	the	flight	through	Houston
	j= 1	2	3	4	5
i= 0					
1					
2					
3					
4					

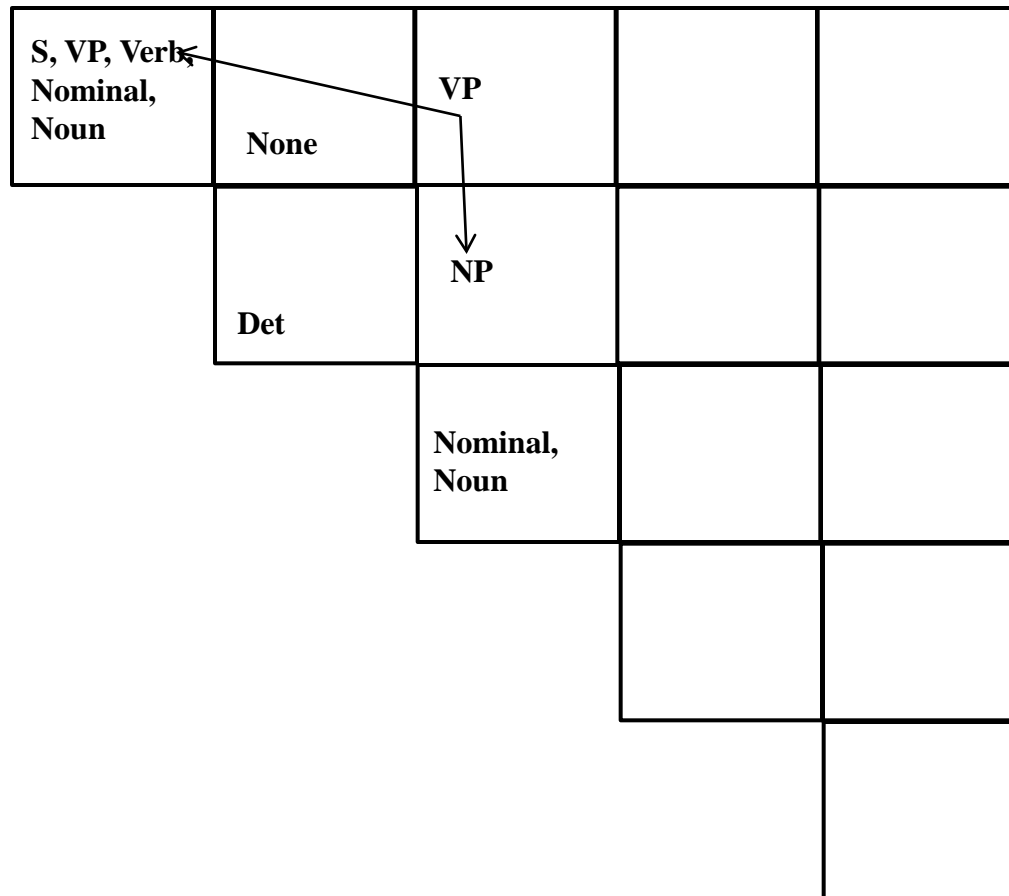
Cell[ $i,j$ ]包含了所有的从  $i+1$ 到  $j$  的非终结符信息



**Book    the    flight    through    Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>				
	<b>Det</b> ←	<b>NP</b>			
		↓			
		<b>Nominal, Noun</b>			

**Book the flight through Houston**



**Book    the    flight    through    Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S</b> <b>VP</b>		
		<b>NP</b>		
	<b>Det</b>	<b>Nominal, Noun</b>		

**Book    the    flight    through    Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S VP</b>		
	<b>Det</b>	<b>NP</b>		
		<b>Nominal, Noun</b>		

**Book the flight through Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S VP</b>	<b>None</b>	
	<b>Det</b>	<b>NP</b>	<b>None</b>	
		<b>Nominal, Noun</b>	<b>None</b>	
			<b>Prep</b>	

**Book    the    flight    through    Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S VP</b>	<b>None</b>	
	<b>Det</b>	<b>NP</b>	<b>None</b>	
		<b>Nominal, Noun</b>	<b>None</b>	
			<b>Prep</b> ←	<b>PP</b>
				↓ <b>NP ProperNoun</b>

**Book the flight through Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S VP</b>	<b>None</b>	
	<b>Det</b>	<b>NP</b>	<b>None</b>	
		<b>Nominal, Noun</b>	<b>None</b>	<b>Nominal</b>
			<b>Prep</b>	<b>PP</b>
				<b>NP ProperNoun</b>

**Book    the    flight    through    Houston**

<b>S, VP, Verb, Nominal, Noun</b>	<b>None</b>	<b>S VP</b>	<b>None</b>	
	<b>Det</b> ←	<b>NP</b>	<b>None</b>	<b>NP</b>
		<b>Nominal, Noun</b>	<b>None</b>	<b>Nominal</b>
			<b>Prep</b>	<b>PP</b>
				<b>NP ProperNoun</b>



**Book the flight through Houston**

S, VP, Verb, Nominal, Noun	None	S VP	None	VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

**Book the flight through Houston**

S, VP, Verb, Nominal, Noun	None	S VP	None	S VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

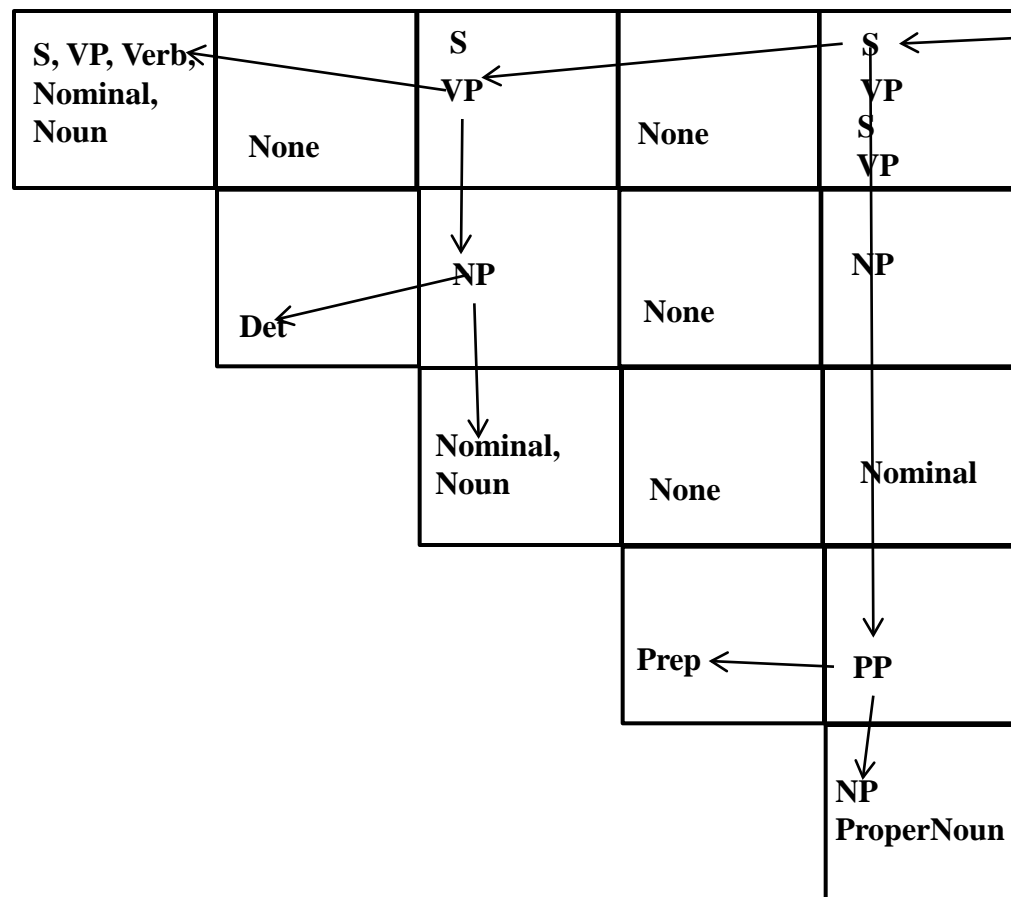
**Book the flight through Houston**

S, VP, Verb, Nominal, Noun	None	S VP ←	None	VP S VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

**Book the flight through Houston**

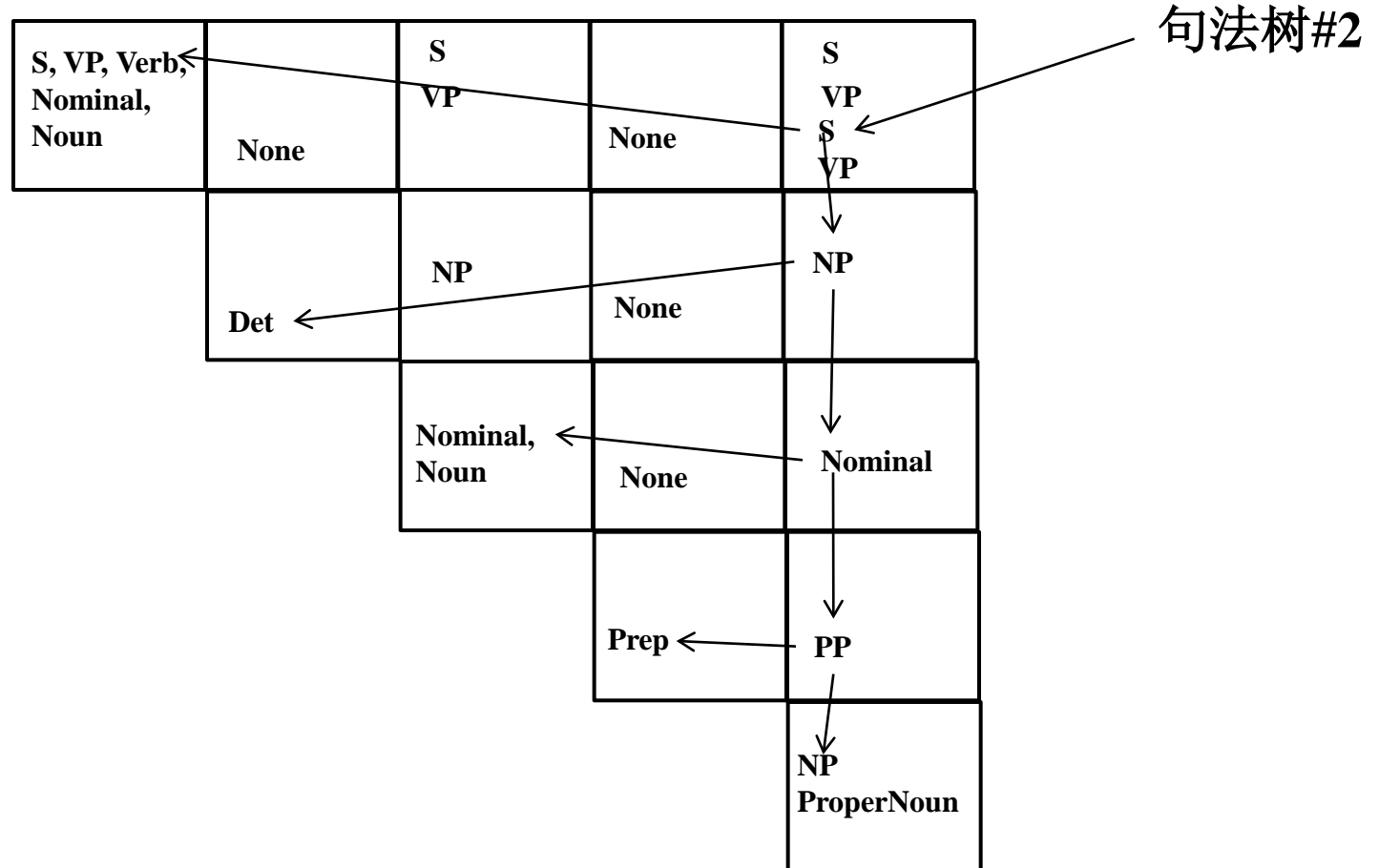
S, VP, Verb, Nominal, Noun	None	S VP ←	None	S VP S VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

**Book the flight through Houston**



句法树#1

**Book the flight through Houston**



# CKY (recognition)的复杂度

- 总共有  $(n(n+1)/2) = O(n^2)$  个单元格
- 总体的时间复杂度是  $O(n^3)$
- 分析得到的句法树 (**parse tree**) 是基于乔姆斯基范式的
- 有必要的話，可以用一个后处理过程把乔姆斯基范式转换回原来的语法规则结构

- 自动句法分析有助于计算机理解句子的意思
  - John ate the spaghetti with meatballs with chopsticks.
    - How did John eat the spaghetti?
    - What did John eat?
- 动态规划算法可以在**3**次方时间内计算一棵句法树，或指数时间内计算所有的句法树
- 问题：只是单纯的输出符合规则的句法树，无法确定哪个句法树的概率最大
  - 解决方法：基于概率的句法分析



# 基于概率的上下文无关文法

(Probabilistic Context Free Grammar)

- **使用概率模型对每个句法树赋予一个概率信息**
  - 通过概率信息消解句法分析中的歧义现象
  - 在标注好的树库的基础上，实现有监督学习
  - 也可以实现无监督学习，但是目前的无监督学习效果比较有限
- **基于概率的上下文无关文法 (Probabilistic Context Free Grammar, PCFG)**
  - 基于概率的上下文无关文法(PCFG)是上下文无关文法(CFG)的概率版本
  - 每个生成规则都带有概率信息

## 语法规则

S → NP VP	0.8	}	+	1.0
S → Aux NP VP	0.1			
S → VP	0.1			
NP → Pronoun	0.2	}	+	1.0
NP → Proper-Noun	0.2			
NP → Det Nominal	0.6			
Nominal → Noun	0.3	}	+	1.0
Nominal → Nominal Noun	0.2			
Nominal → Nominal PP	0.5			
VP → Verb	0.2	}	+	1.0
VP → Verb NP	0.5			
VP → VP PP	0.3			
PP → Prep NP	1.0			

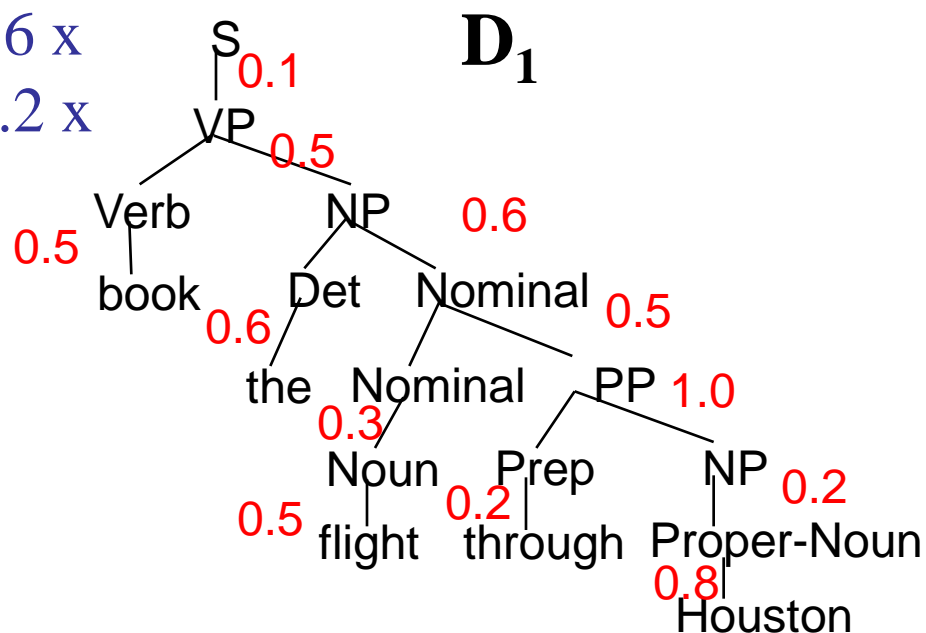
## 概率

## 词汇规则

Det → the   a   that   this	0.6	0.2	0.1	0.1	
Noun → book   flight   meal   money	0.1	0.5	0.2	0.2	
Verb → book   include   prefer	0.5	0.2	0.3		
Pronoun → I   he   she   me	0.5	0.1	0.1	0.3	
Proper-Noun → Houston   NWA	0.8	0.2			
Aux → does	1.0				
Prep → from   to   on   near   through	0.25	0.25	0.1	0.2	0.2

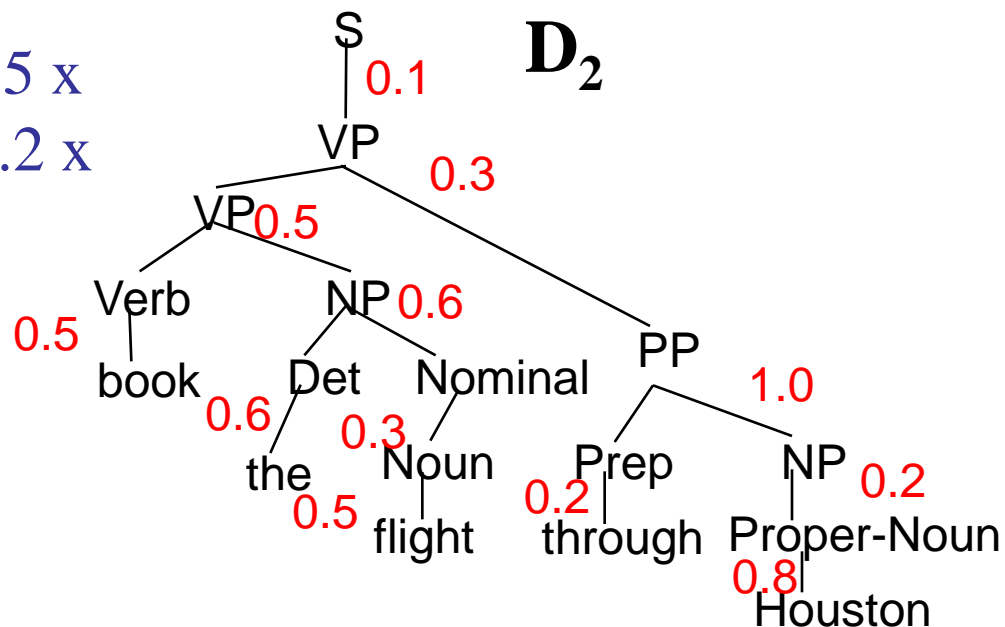
- 独立性假设
- 概率相乘

$$\begin{aligned} P(D_1) &= 0.1 \times 0.5 \times 0.5 \times 0.6 \times 0.6 \times \\ &\quad 0.5 \times 0.3 \times 1.0 \times 0.2 \times 0.2 \times \\ &\quad 0.5 \times 0.8 \\ &= 0.0000216 \end{aligned}$$



- 挑选概率最大的句法树作为句法分析的结果

$$\begin{aligned} P(D_2) &= 0.1 \times 0.3 \times 0.5 \times 0.6 \times 0.5 \times \\ &\quad 0.6 \times 0.3 \times 1.0 \times 0.5 \times 0.2 \times \\ &\quad 0.2 \times 0.8 \\ &= 0.00001296 \end{aligned}$$



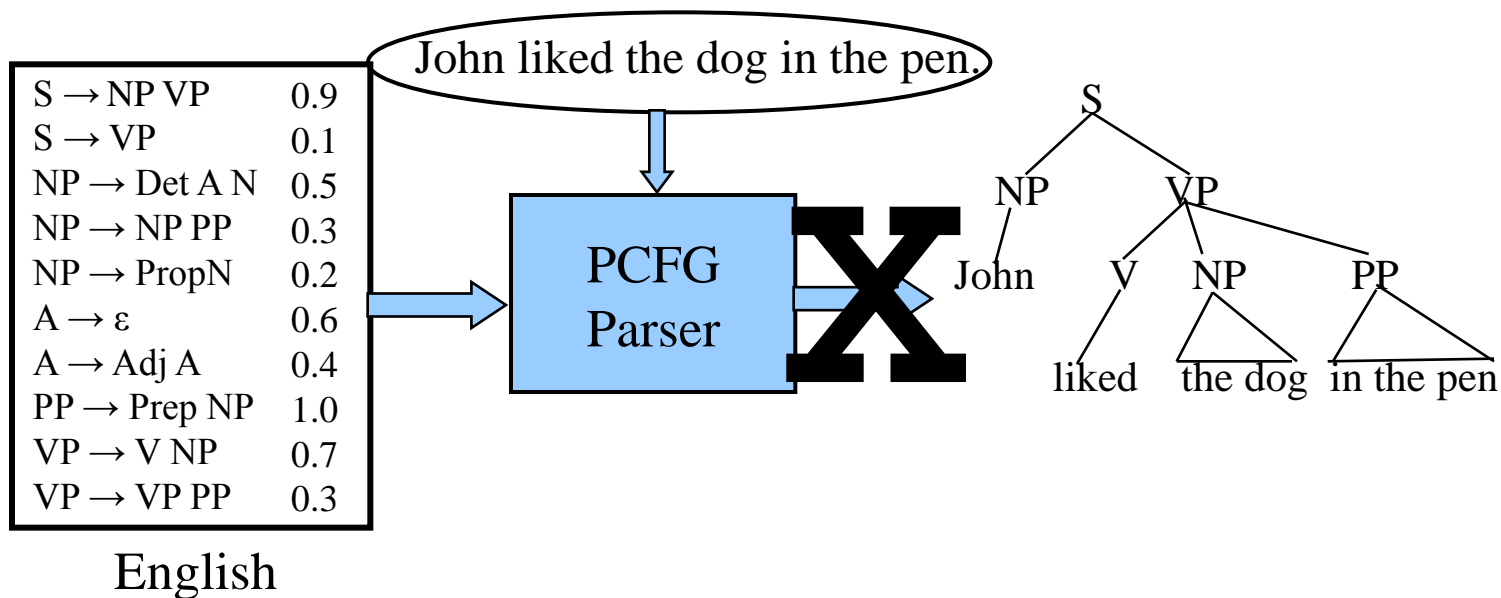
- 也可以计算句子本身的概率，句子本身的概率是其所有句法树概率之和

$$\begin{aligned} P(\text{"book the flight through Houston"}) &= \\ P(D_1) + P(D_2) &= 0.0000216 + 0.00001296 \\ &= 0.00003456 \end{aligned}$$

- **观测概率 (Observation likelihood)**
  - 用于对句子排序等
- **最大概率句法树 (Most likely derivation)**
  - 找出最大概率句法树
- **最大似然训练 (Maximum likelihood training)**
  - 基于训练数据训练一个句法分析器

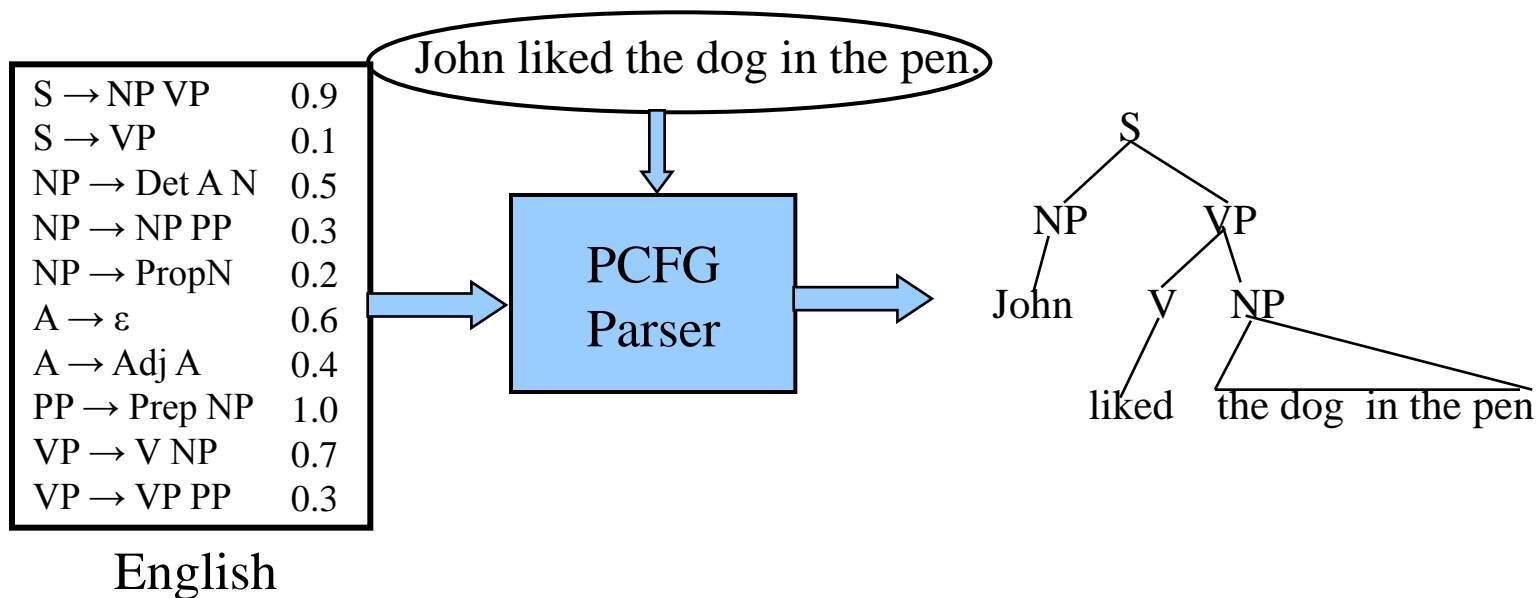
# PCFG: 最大概率的句法树

- 可以使用维特比（**Viterbi**）算法确定最大可能的句法树





- 可以使用维特比（**Viterbi**）算法确定最大可能的句法树



# 基于概率的CKY算法 ( Probabilistic CKY )

- 在原来的**CKY**算法的基础上加入概率信息
- 当转换为乔姆斯基范式的时候，需要重新设置概率信息从而保证原有的概率分布

# 基于概率的乔姆斯基范式转换

## 原始语法规则

<b>S</b> → <b>NP VP</b>	0.8
<b>S</b> → <b>Aux NP VP</b>	0.1
<b>S</b> → <b>VP</b>	0.1
<b>NP</b> → <b>Pronoun</b>	0.2
<b>NP</b> → <b>Proper-Noun</b>	0.2
<b>NP</b> → <b>Det Nominal</b>	0.6
<b>Nominal</b> → <b>Noun</b>	0.3
<b>Nominal</b> → <b>Nominal Noun</b>	0.2
<b>Nominal</b> → <b>Nominal PP</b>	0.5
<b>VP</b> → <b>Verb</b>	0.2
<b>VP</b> → <b>Verb NP</b>	0.5
<b>VP</b> → <b>VP PP</b>	0.3
<b>PP</b> → <b>Prep NP</b>	1.0

## 乔姆斯基范式

<b>S</b> → <b>NP VP</b>	0.8
<b>S</b> → <b>X1 VP</b>	0.1
<b>X1</b> → <b>Aux NP</b>	1.0
<b>S</b> → <b>book   include   prefer</b>	
<b>0.01 0.004 0.006</b>	
<b>S</b> → <b>Verb NP</b>	0.05
<b>S</b> → <b>VP PP</b>	0.03
<b>NP</b> → <b>I   he   she   me</b>	
<b>0.1 0.02 0.02 0.06</b>	
<b>NP</b> → <b>Houston   NWA</b>	
<b>0.16 .04</b>	
<b>NP</b> → <b>Det Nominal</b>	0.6
<b>Nominal</b> → <b>book   flight   meal   money</b>	
<b>0.03 0.15 0.06 0.06</b>	
<b>Nominal</b> → <b>Nominal Noun</b>	0.2
<b>Nominal</b> → <b>Nominal PP</b>	0.5
<b>VP</b> → <b>book   include   prefer</b>	
<b>0.1 0.04 0.06</b>	
<b>VP</b> → <b>Verb NP</b>	0.5
<b>VP</b> → <b>VP PP</b>	0.3
<b>PP</b> → <b>Prep NP</b>	1.0

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None			
	Det:.6	NP:.6*.6*.15 =.054		
		Nominal:.15 Noun:.5		

**Book      the      flight      through      Houston**

S :.01, VP:.1, Verb:.5 ← Nominal:.03 Noun:.1	None	VP:.5*.5*.054 =.0135		
	Det:.6	NP:.6*.6*.15 =.054		
		Nominal:.15 Noun:.5		

**Book      the      flight      through      Houston**

<b>S :.01, VP:.1, Verb:.5 ← Nominal:.03 Noun:.1</b>	<b>None</b>	<b>S:.05*.5*.054 =.00135</b> <b>VP:.5*.5*.054 =.0135</b>		
	<b>Det:.6</b>	<b>NP:.6*.6*.15 =.054</b>		
		<b>Nominal:.15 Noun:.5</b>		

**Book      the      flight      through      Houston**

<b>S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1</b>	<b>None</b>	<b>S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135</b>	<b>None</b>	
	<b>Det:.6</b>	<b>NP:.6*.6*.15 =.054</b>	<b>None</b>	
		<b>Nominal:.15 Noun:.5</b>	<b>None</b>	
			<b>Prep:.2</b>	

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	
	Det:.6	NP:.6*.6*.15 =.054	None	
		Nominal:.15 Noun:.5	None	
			Prep:.2 ←	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8



**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	
	Det:.6	NP:.6*.6*.15 =.054	None	
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8

# 基于概率的CKY算法

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	
	Det:.6 ←	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8

# 基于概率的CKY算法

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1		S:.05*.5*.054 =.00135		S:.05*.5* .000864 =.0000216
	None	VP:.5*.5*.054 =.0135	None	
	Det:.6	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun: 8

# 基于概率的CKY算法

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	S:.03*.0135* .032 =.00001296 S:.0000216
	Det:.6	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8

# 基于概率的CKY算法

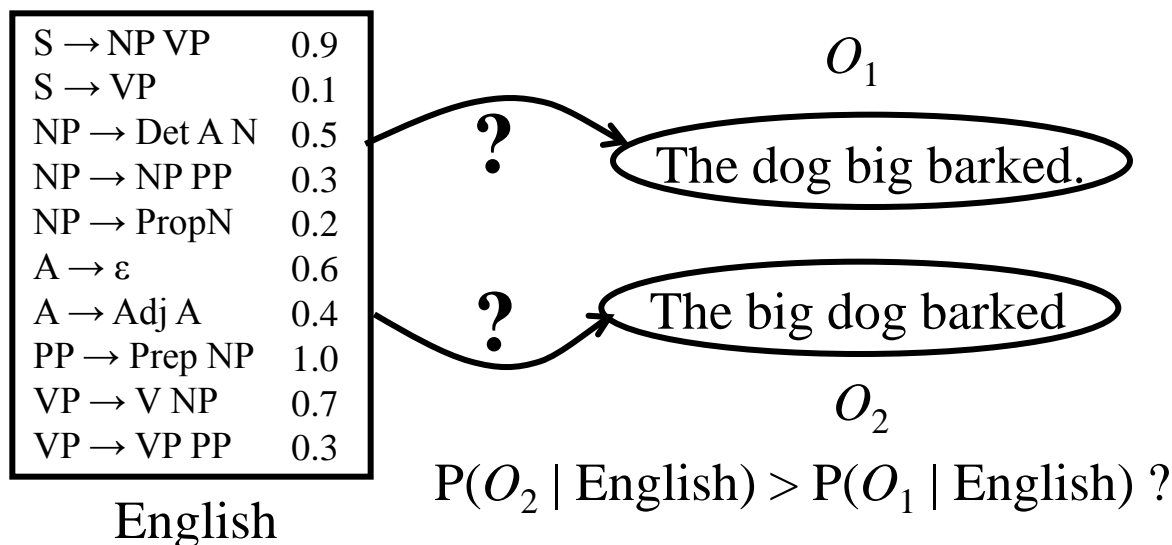
**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1		S:.05*.5*.054 =.00135		S:.0000216
	None	VP:.5*.5*.054 =.0135	None	
	Det:.6	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun: 8

计算最大概率的句法树

# PCFG: 观测概率的计算(Observation Likelihood)

- 类似于序列标注问题中的观测概率的计算方法—前向算法(**Forward algorithm**), 可以使用 **Inside algorithm** 这个算法进行动态规划计算观测概率
  - 把Viterbi算法的max计算改为sum计算
- 可以使用**PCFG**作为一个语言模型, 从而对句子的概率进行计算, 用于语音识别、机器翻译等



# PCFG: 观测概率的计算(Observation Likelihood)

## Probabilistic CKY Parser for Inside Computation

**Book the flight through Houston**

S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	S:..00001296 S:.0000216
	Det:.6	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8

# PCFG: 观测概率的计算(Observation Likelihood)

Probabilistic CKY Parser for Inside Computation

**Book the flight through Houston**

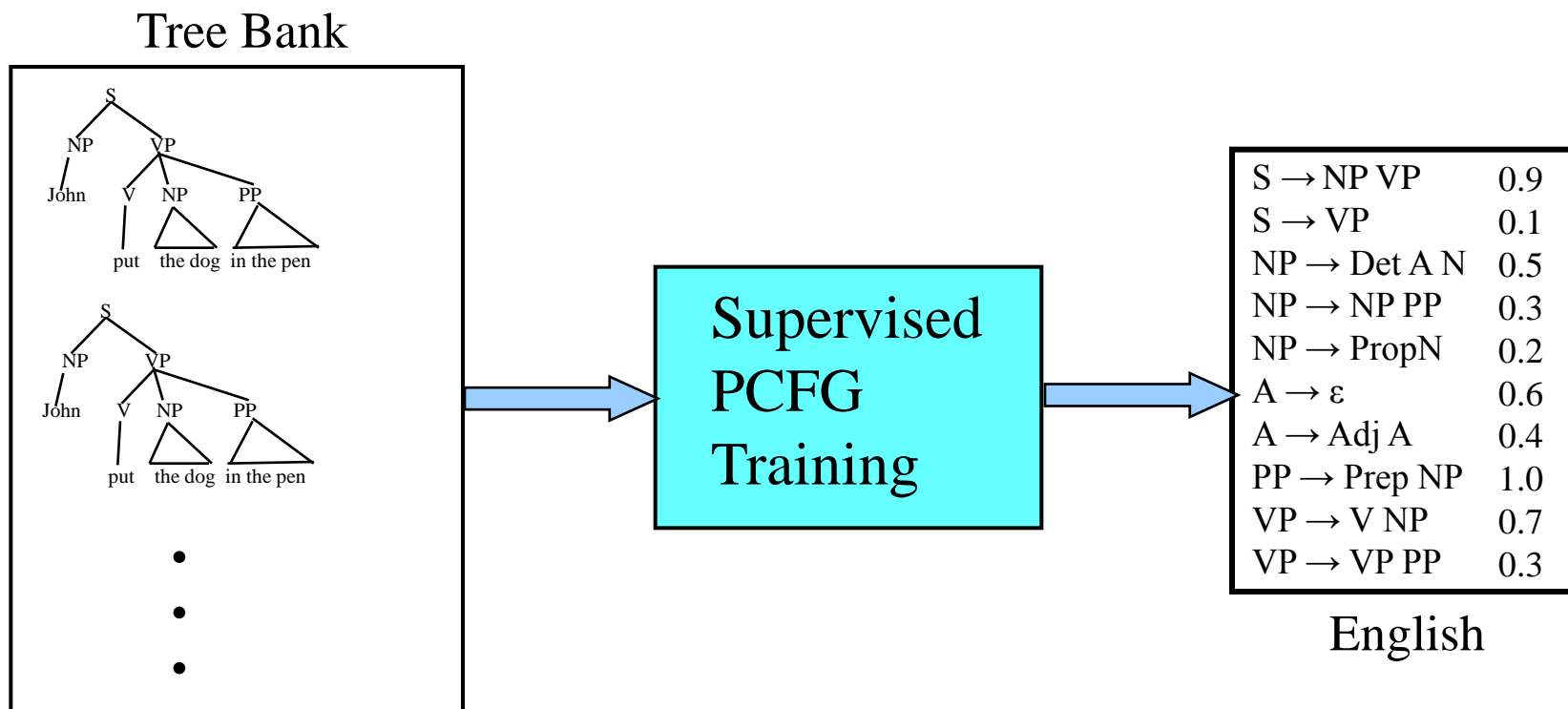
S :.01, VP:.1, Verb:.5 Nominal:.03 Noun:.1	None	S:.05*.5*.054 =.00135 VP:.5*.5*.054 =.0135	None	S: .00001296 +.0000216 =.00003456
	Det:.6	NP:.6*.6*.15 =.054	None	NP:.6*.6* .0024 =.000864
		Nominal:.15 Noun:.5	None	Nominal: .5*.15*.032 =.0024
			Prep:.2	PP:1.0*.2*.16 =.032
				NP:.16 PropNoun:. 8

计算观测概率



# PCFG: 有监督学习(Supervised Training)

- 给定了训练数据的话（一般是标注好的树库**tree bank**），**PCFG**的有监督学习相对比较简单
  - 可以通过相对频率来计算



- 可以从树库收集语法规则
- 语法规则对应的概率可以通过相对频率来计算

$$P(\alpha \rightarrow \beta | \alpha) = \frac{\text{count}(\alpha \rightarrow \beta)}{\sum_{\gamma} \text{count}(\alpha \rightarrow \gamma)} = \frac{\text{count}(\alpha \rightarrow \beta)}{\text{count}(\alpha)}$$

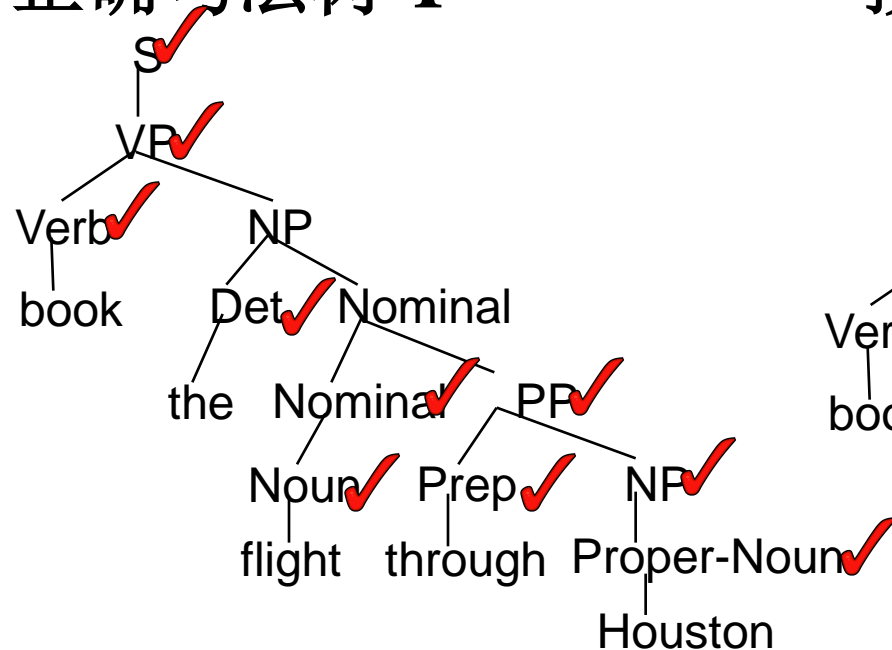
# 句法分析的效果打分

- 句法分析器给出了预测的句法树之后，可以计算句法树和已经标注好的句法树的相似度
- 假设 **P** 是系统输出的句法树，假设 **T** 是标注好的句法树：
  - **召回率(Recall)** = (#P中正确的元素) / (#T中正确的元素)
  - **准确率(Precision)** = (#P中正确的元素) / (#P中总的元素)
- **F值(F-score, F1)**是召回率和准确率之间的调和平均数 (harmonic mean)

目前好的句法分析系统在标准数据集上可以达到**90%**以上的Precision, Recall, F-score

# 句法分析的效果打分

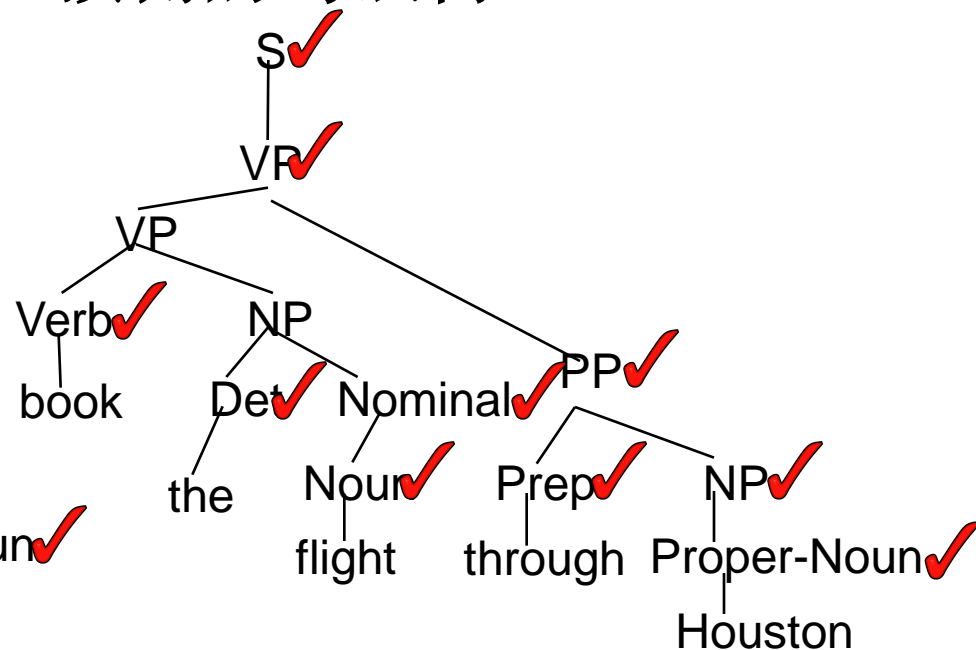
## 正确句法树 T



# 元素: 12

# 正确的元素: 10

## 预测的句法树P

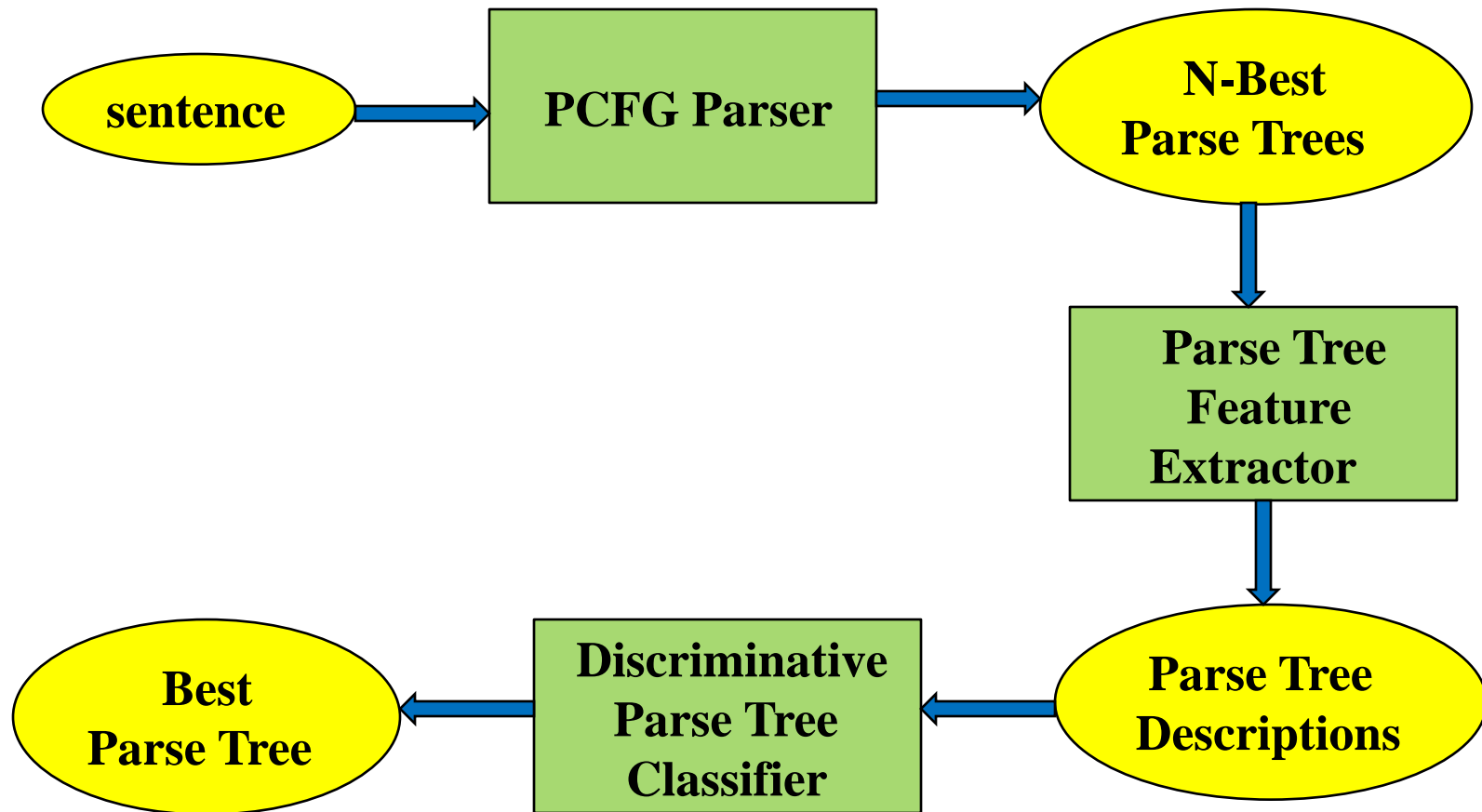


# 元素: 12

Recall =  $10/12 = 83.3\%$  Precision =  $10/12 = 83.3\%$

$F_1 = 83.3\%$

# 基于重排序的句法分析(re-ranking parser)



- 通过概率信息，消解句法分析中的歧义，得到最大概率的句法分析树等
- 通过标注好的树库，可以学习到概率句法分析器
- 现有的概率句法分析技术已经有很高的准确度

## □ 参考书

- 《统计自然语言处理》第8章：句法分析
  - 8.1 句法结构分析概述
  - Page 179 – 184
  - 8.2 基于PCFG的基本分析方法
  - Page 184 - 192