

Parsing

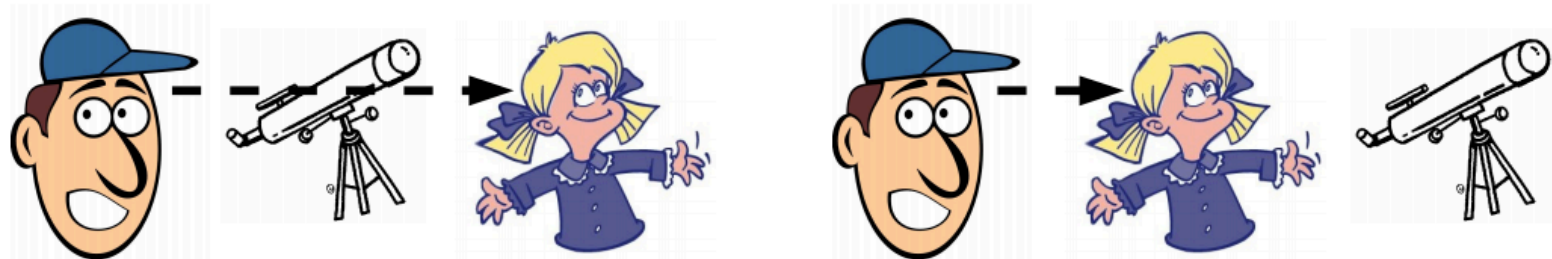
Luciano Barbosa

(baseado nos slides do curso de PLN de Stanford e livro Speech and Language Processing)

Ambiguidades em Interpretar Linguagem

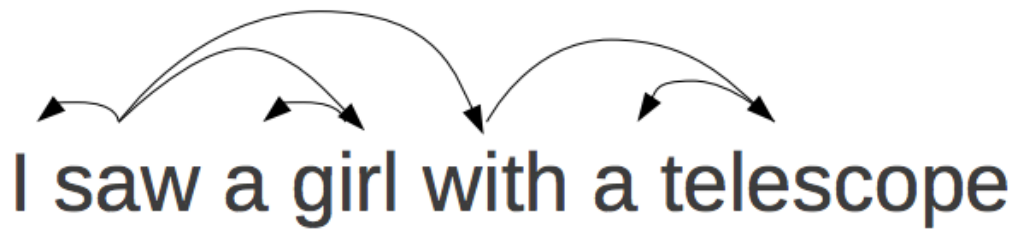
- Parsing resolve ambiguidade estrutural de uma maneira formal

I saw a girl with a telescope

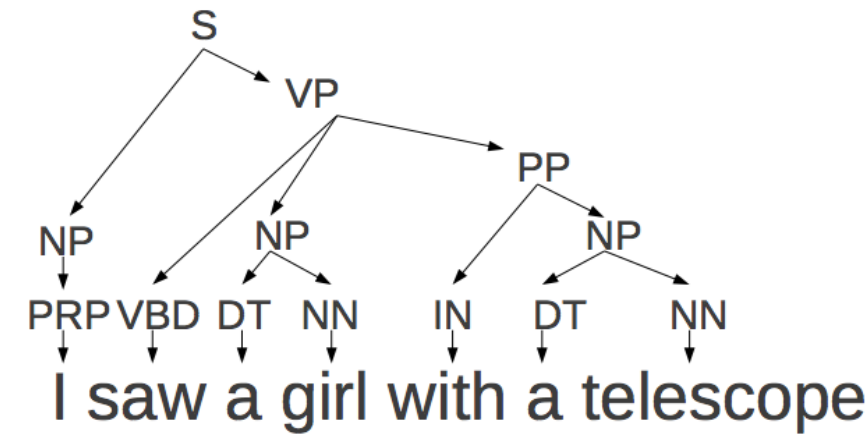


Dois Tipos de Parsing

- Dependency: relações entre palavras



- Constituency: identifica frases e sua estrutura recursiva



Resolvendo Ambiguidade com Dependências

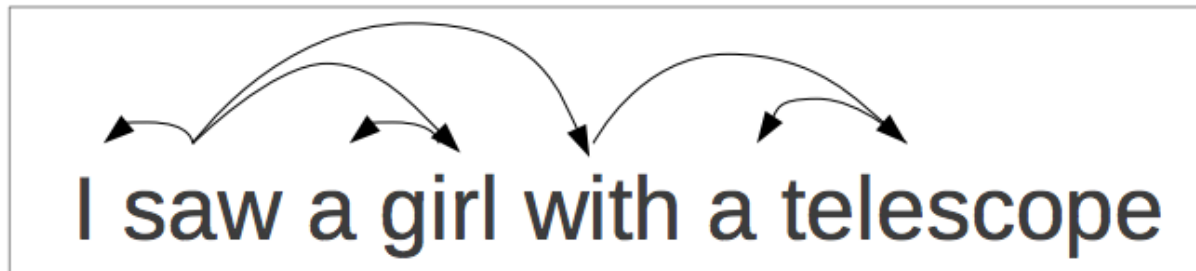


I saw a girl with a telescope

I saw a girl with a telescope

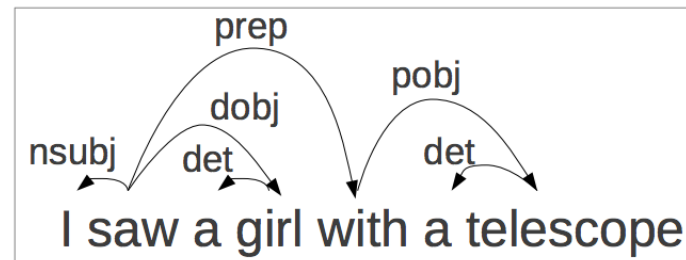
Tipos de Dependências

- Não-tipadas: somente dependências



Tipos de Dependências

- Tipadas: rótulos indicam o relacionamento entre palavras



Clausal Argument Relations	Description
NSUBJ	Nominal subject
DOBJ	Direct object
IOBJ	Indirect object
CCOMP	Clausal complement
XCOMP	Open clausal complement
Nominal Modifier Relations	Description
NMOD	Nominal modifier
AMOD	Adjectival modifier
NUMMOD	Numeric modifier
APPOS	Appositional modifier
DET	Determiner
CASE	Prepositions, postpositions and other case markers
Other Notable Relations	Description
CONJ	Conjunct
CC	Coordinating conjunction

Tipos de Dependências

- Tipadas: rótulos indicam o relacionamento entre palavras

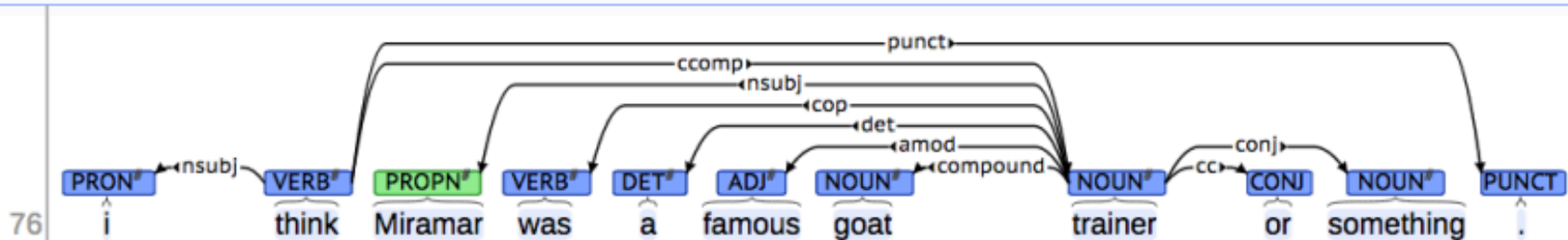
Relation	Examples with <i>head</i> and dependent
NSUBJ	United <i>canceled</i> the flight.
DOBJ	United <i>diverted</i> the flight to Reno. We <i>booked</i> her the first flight to Miami.
IOBJ	We <i>booked</i> her the flight to Miami.
NMOD	We took the morning <i>flight</i> .
AMOD	Book the cheapest <i>flight</i> .
NUMMOD	Before the storm JetBlue canceled 1000 <i>flights</i> .
APPOS	<i>United</i> , a unit of UAL, matched the fares.
DET	The <i>flight</i> was canceled. Which <i>flight</i> was delayed?
CONJ	We <i>flew</i> to Denver and drove to Steamboat.
CC	We flew to Denver and <i>drove</i> to Steamboat.
CASE	Book the flight through <i>Houston</i> .

Preferências na Dependência

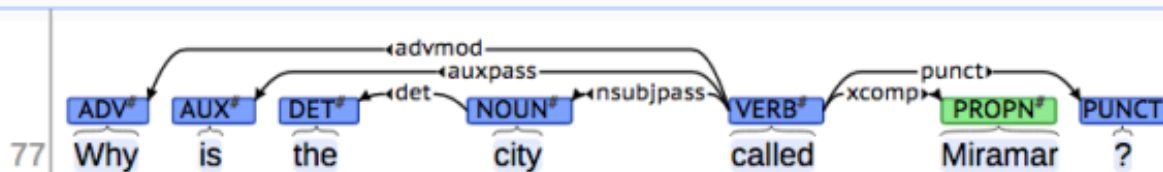
- Distância: palavras próximas
- Raramente vai além dos verbos e pontuações
- Não há ciclos (estrutura de árvore)
- Tem uma raiz

Dados Anotados: Treebank

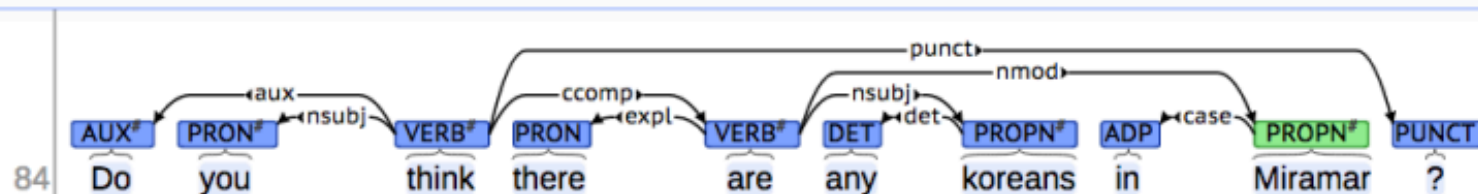
[context] [conllu]



[context] [conllu]



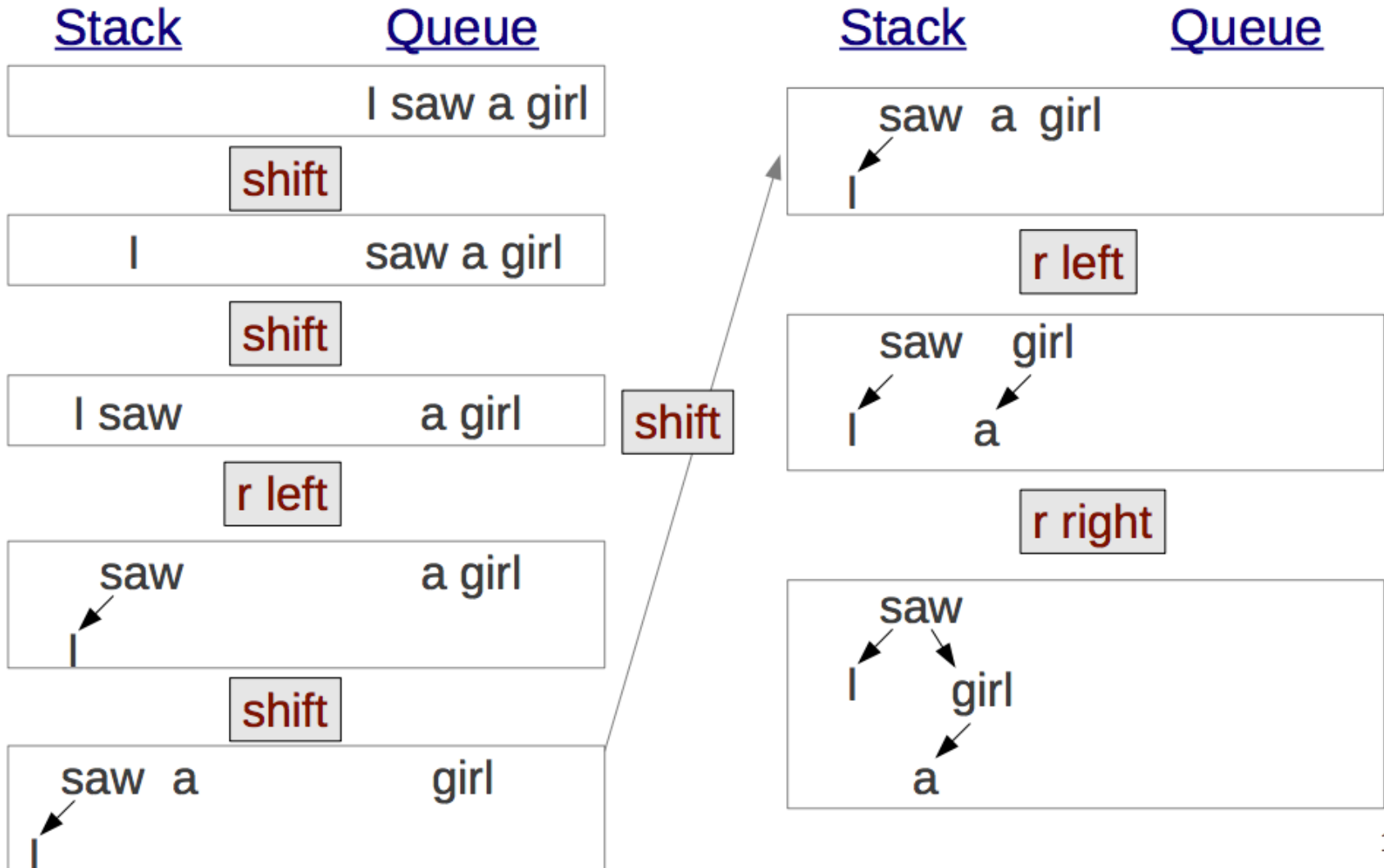
[context] [conllu]



Algoritmo Shift-Reduce

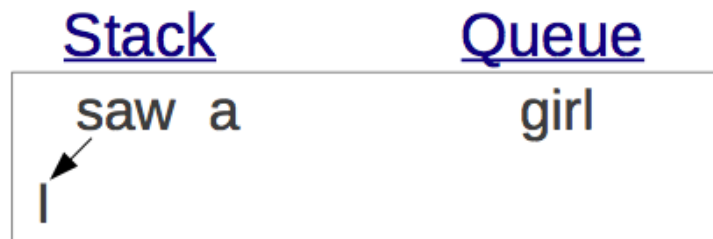
- Processa uma palavra por vez da esquerda para a direita
- Duas estruturas
 - Queue: palavras não processadas
 - Stack: palavras parcialmente processadas
- Em cada ponto escolha
 - shift: move uma palavra do queue para o stack
 - reduce left: palavra no topo do stack é a cabeça da segunda palavra
 - reduce right: segunda palavra no stack é a cabeça da palavra no topo
- Aprender cada tarefa utilizando um classificador

Exemplo do Shift Reduce



Classificação em Shift-Reduce

- Dado o estado:

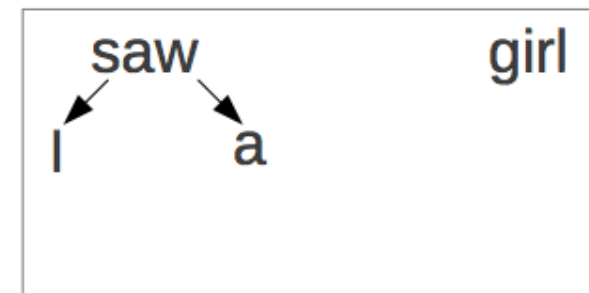
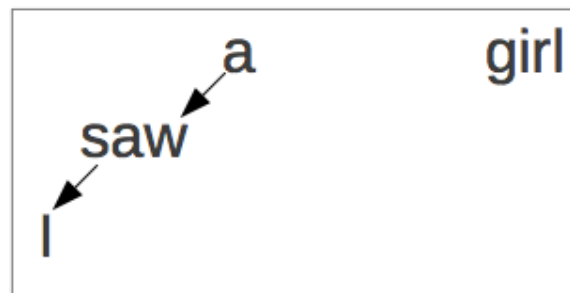
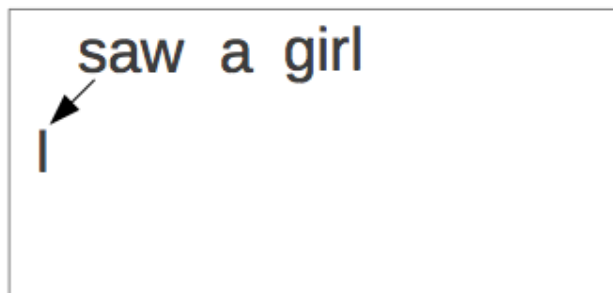


- Qual a próxima ação

shift ?

r left ?

r right ?



Features

- POS
- Lema
- Palavra no topo do stack
- Primeira palavra no queue

	<u>stack[-2]</u>	<u>stack[-1]</u>	<u>queue[0]</u>
Word:	saw	a	girl
POS:	VBD	DET	NN

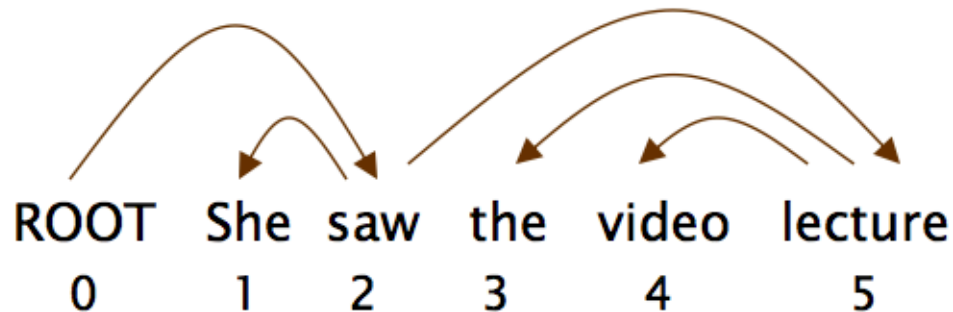
(-2 → second-to-last)

(-1 → last)

(0 → first)

Avaliação

- Acurácia rotulada (LAS) e não rotulada (UAS)



$$\text{Acc} = \frac{\# \text{ correct deps}}{\# \text{ of deps}}$$

$$\text{UAS} = 4 / 5 = 80\%$$

$$\text{LAS} = 2 / 5 = 40\%$$

Gold

1	2	She	nsubj
2	0	saw	root
3	5	the	det
4	5	video	nn
5	2	lecture	obj

Parsed

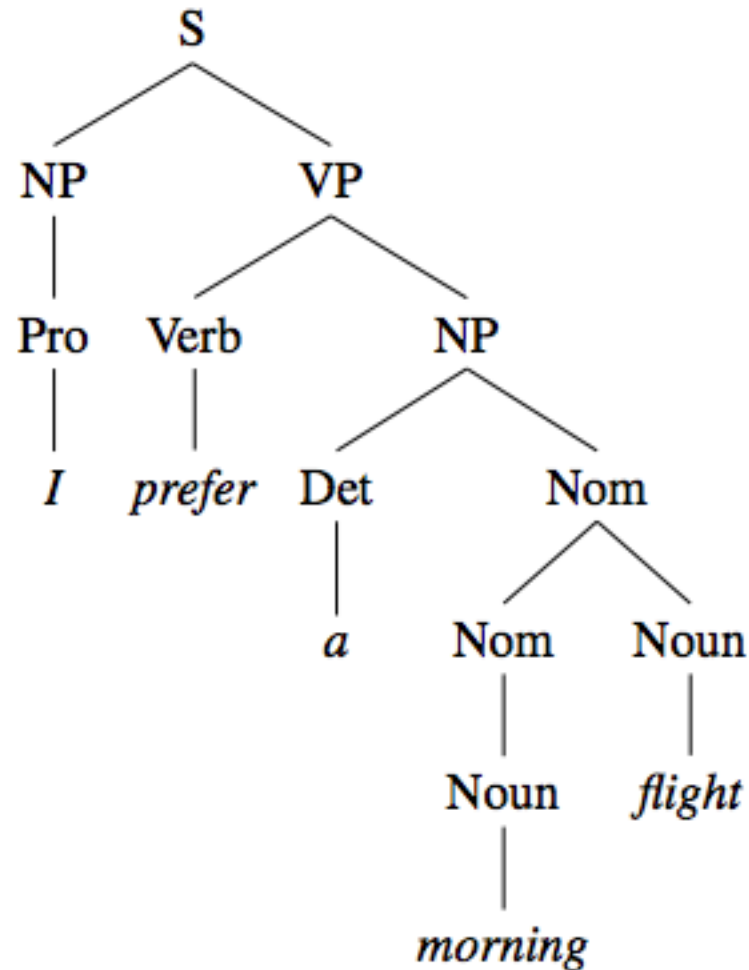
1	2	She	nsubj
2	0	saw	root
3	4	the	det
4	5	video	nsubj
5	2	lecture	ccomp

Context-Free Grammar (CFG)

Noun → *flights* | *breeze* | *trip* | *morning*
Verb → *is* | *prefer* | *like* | *need* | *want* | *fly*
Adjective → *cheapest* | *non-stop* | *first* | *latest*
 | *other* | *direct*
Pronoun → *me* | *I* | *you* | *it*
Proper-Noun → *Alaska* | *Baltimore* | *Los Angeles*
 | *Chicago* | *United* | *American*
Determiner → *the* | *a* | *an* | *this* | *these* | *that*
Preposition → *from* | *to* | *on* | *near*
Conjunction → *and* | *or* | *but*

Grammar Rules	Examples
$S \rightarrow NP VP$	I + want a morning flight
$NP \rightarrow$ <i>Pronoun</i> <i>Proper-Noun</i> <i>Det Nominal</i>	I Los Angeles a + flight
$Nominal \rightarrow$ <i>Nominal Noun</i> <i>Noun</i>	morning + flight flights
$VP \rightarrow$ <i>Verb</i> <i>Verb NP</i> <i>Verb NP PP</i> <i>Verb PP</i>	do want + a flight leave + Boston + in the morning leaving + on Thursday
$PP \rightarrow$ <i>Preposition NP</i>	from + Los Angeles

Context-Free Grammar (CFG)

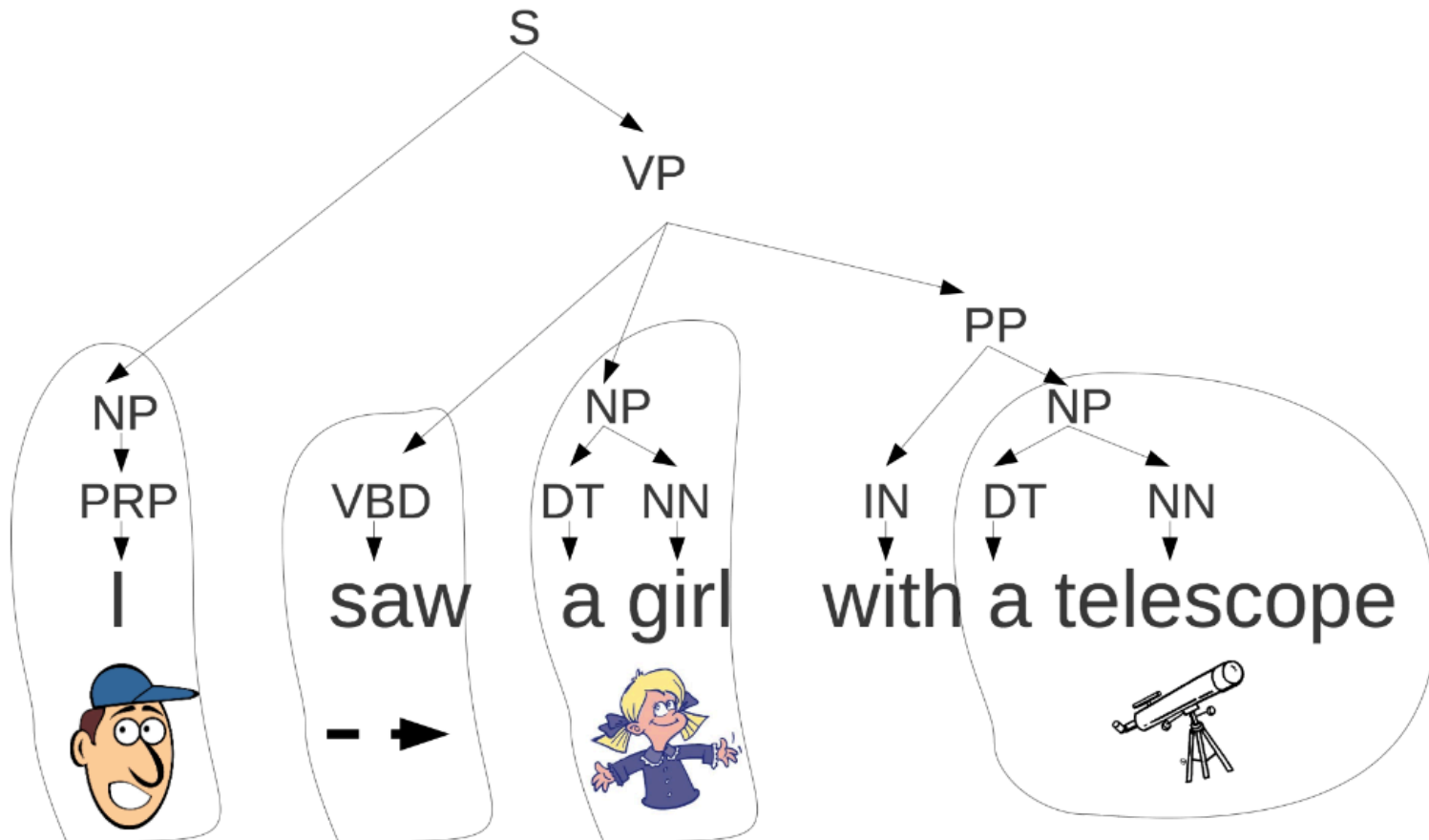


Treebanks

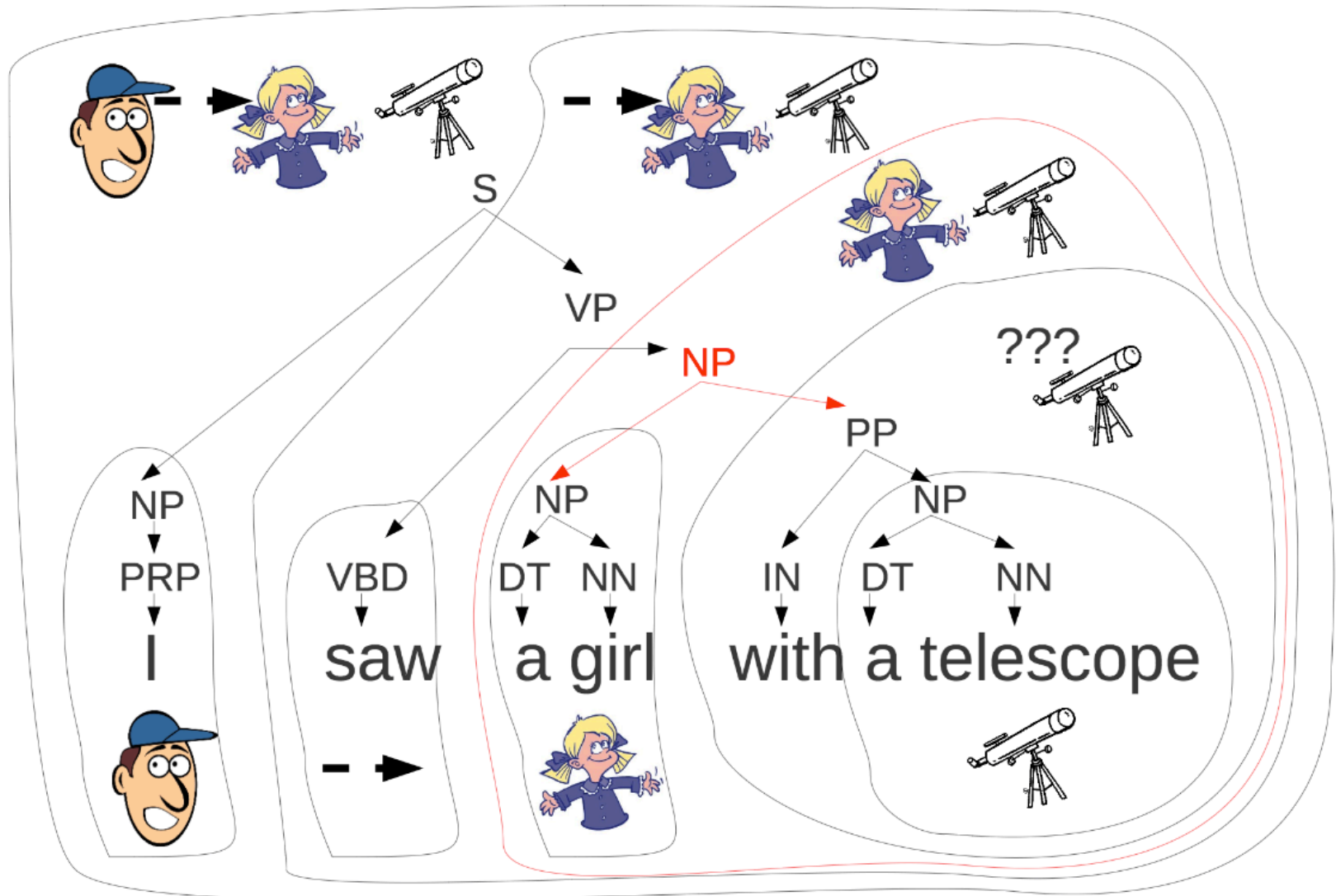
- Cada sentença com parse tree anotada
- Penn Treebank para inglês

```
( (S
  (NP-SBJ
    (NP (NNP Pierre) (NNP Vinken) )
    ( , , )
    (ADJP
      (NP (CD 61) (NNS years) )
      (JJ old) )
    ( , , ) )
  (VP (MD will)
    (VP (VB join)
      (NP (DT the) (NN board) )
      (PP-CLR (IN as)
        (NP (DT a) (JJ nonexecutive) (NN director) ))
      (NP-TMP (NNP Nov.) (CD 29) )))
  ( . . ) ) )
```

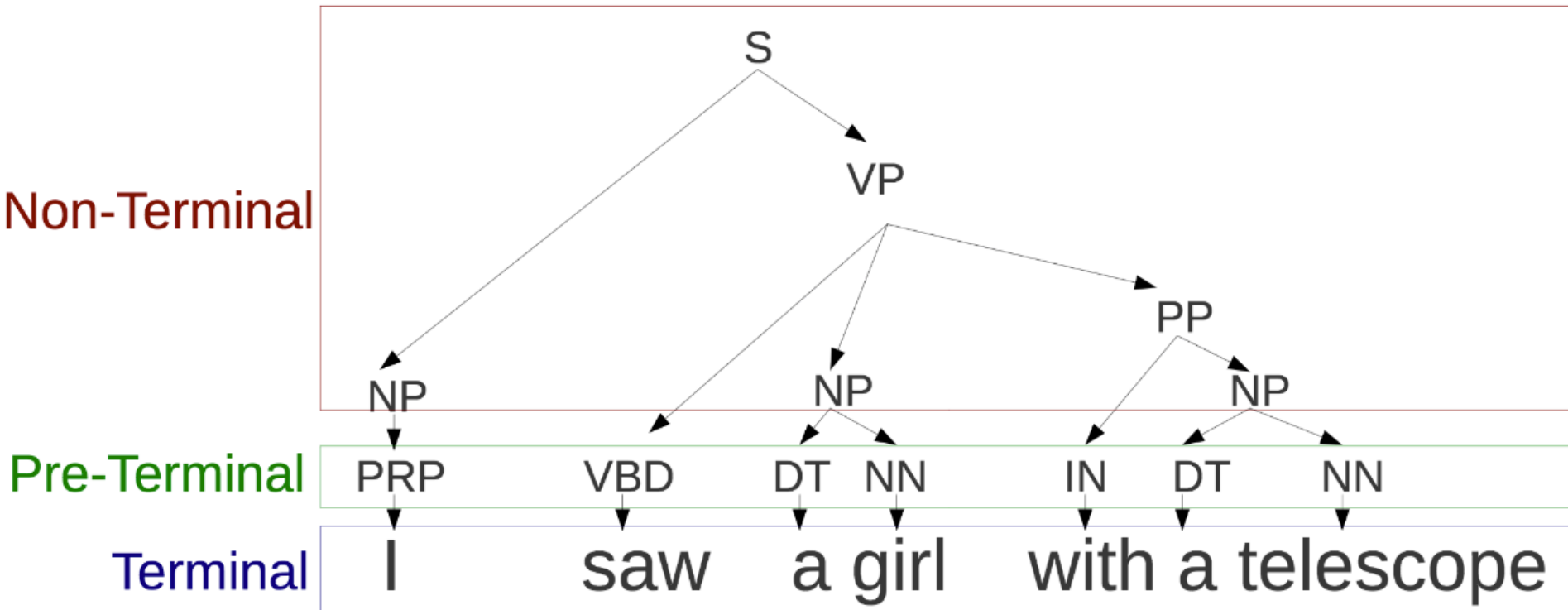
Ambiguidade



Ambiguidade

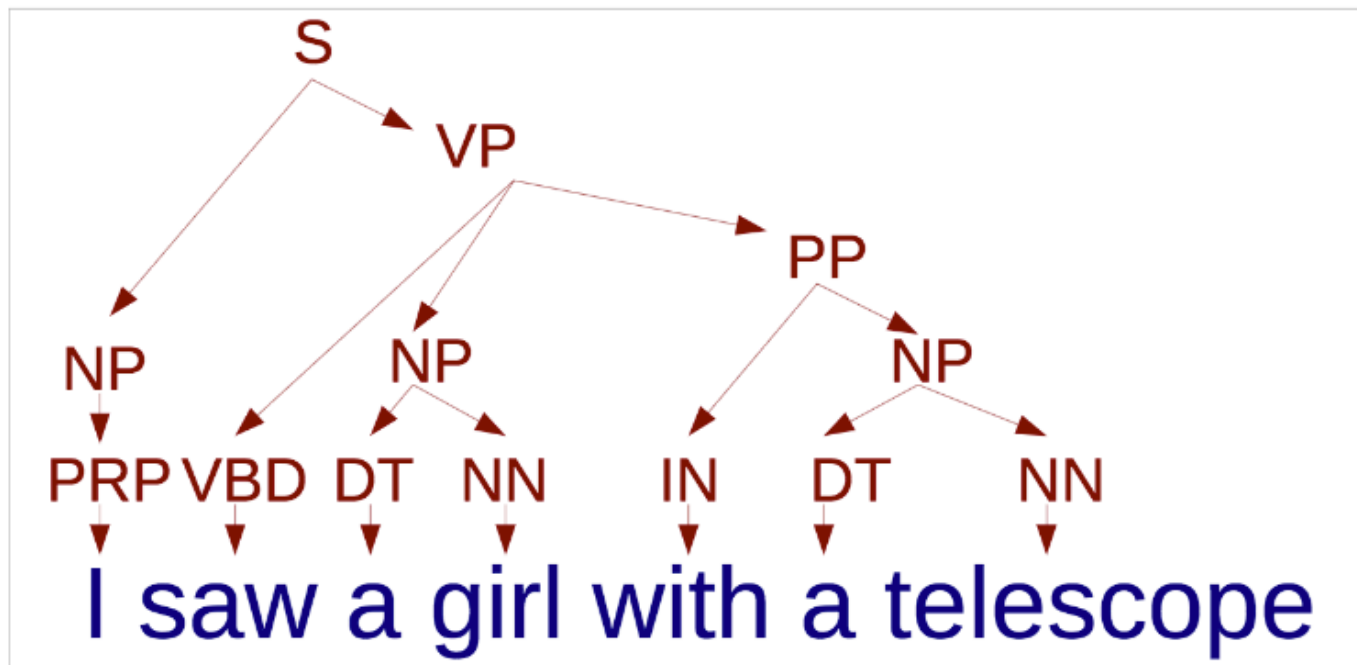


Componentes



Aprendendo uma Parse Tree

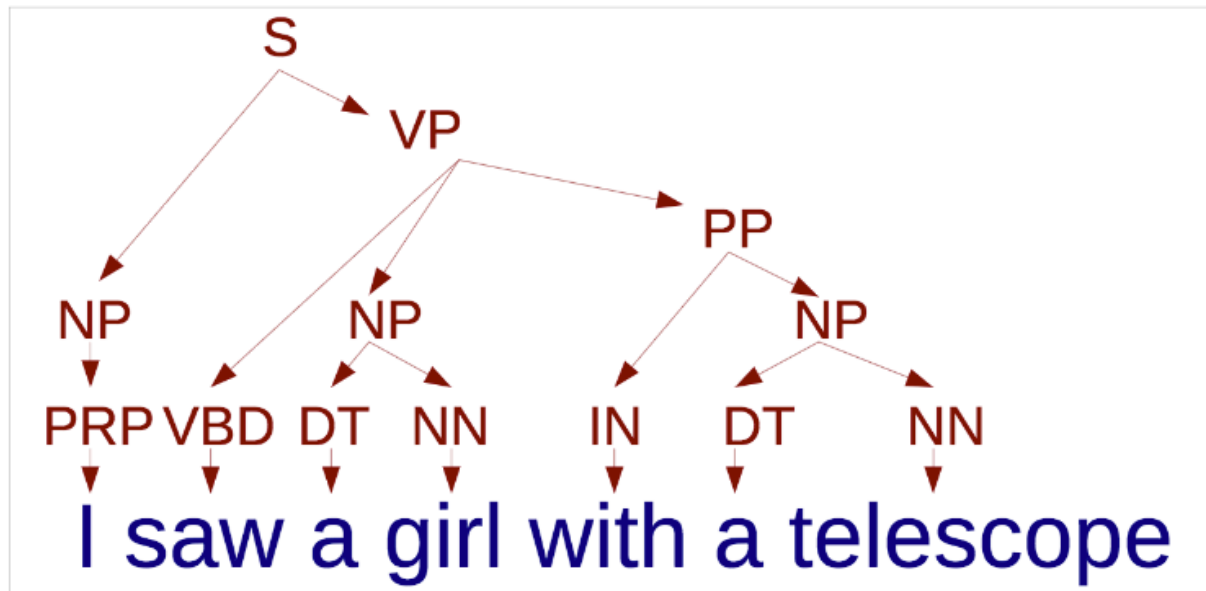
- Dada uma sentença X, predizer a parse tree



- Predição estruturada (similar a POS)

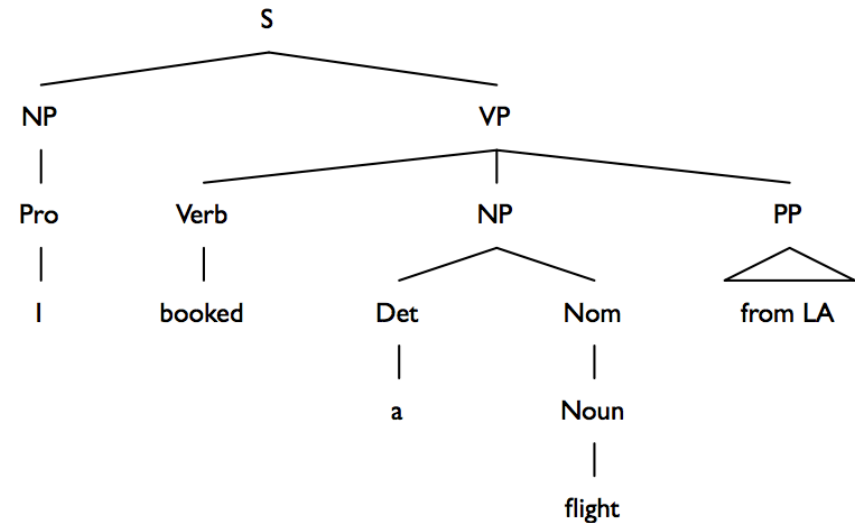
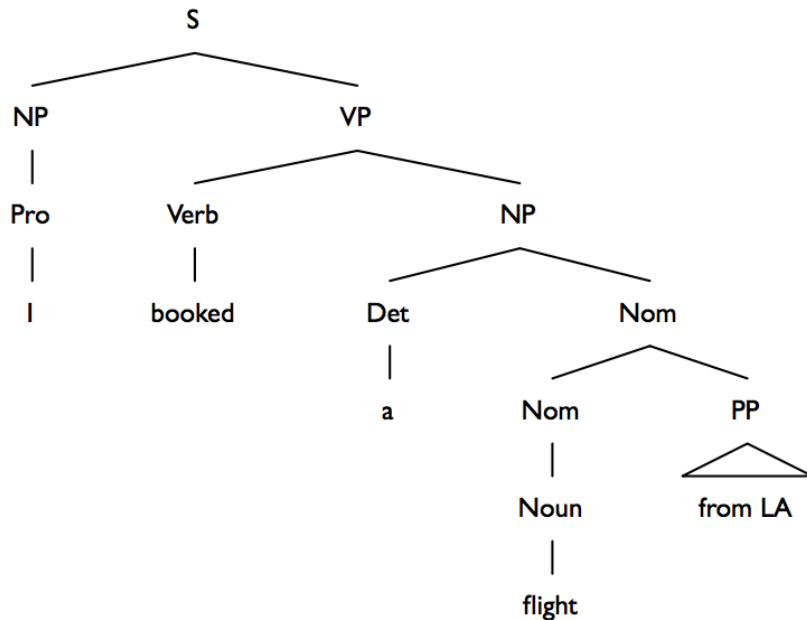
Abordagem Probabilística

- Dada a sentença X, predizer a parse tree Y mais provável



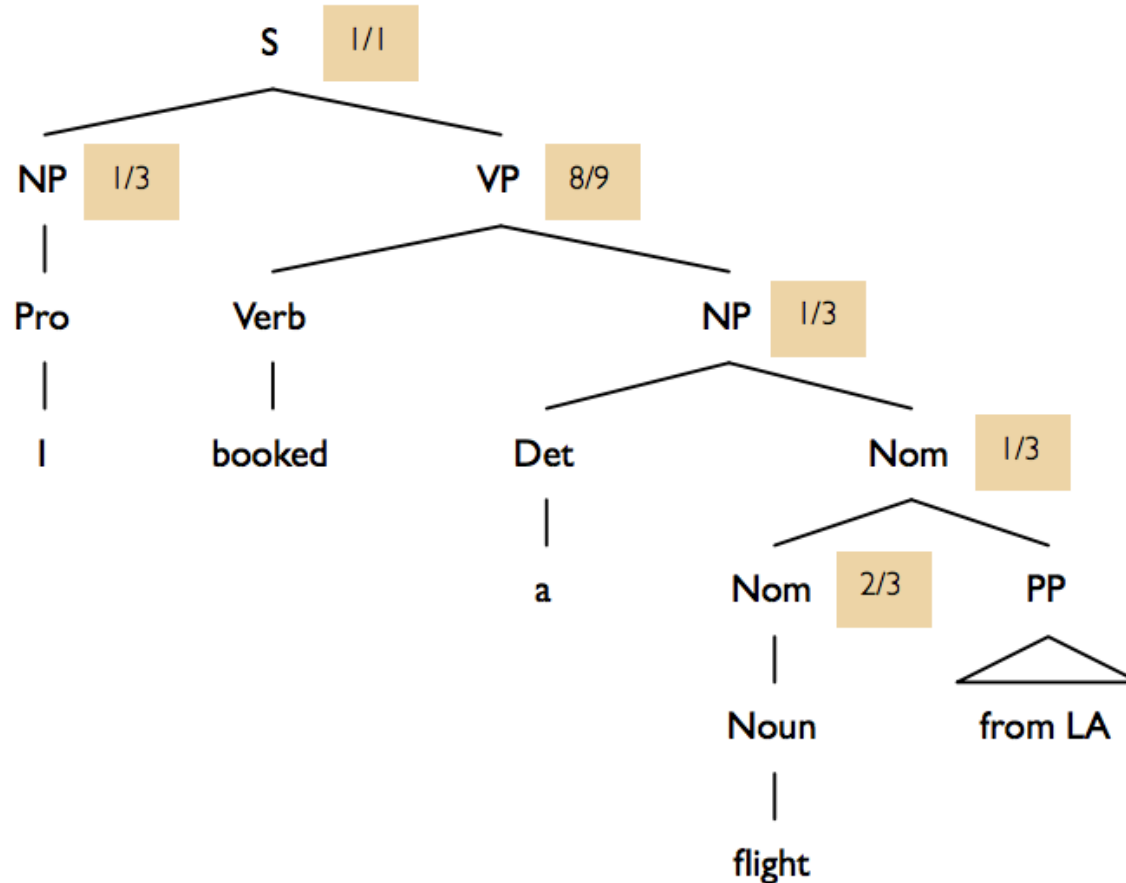
$$\operatorname{argmax}_Y P(Y|X)$$

Escolha da Melhor Parse Tree



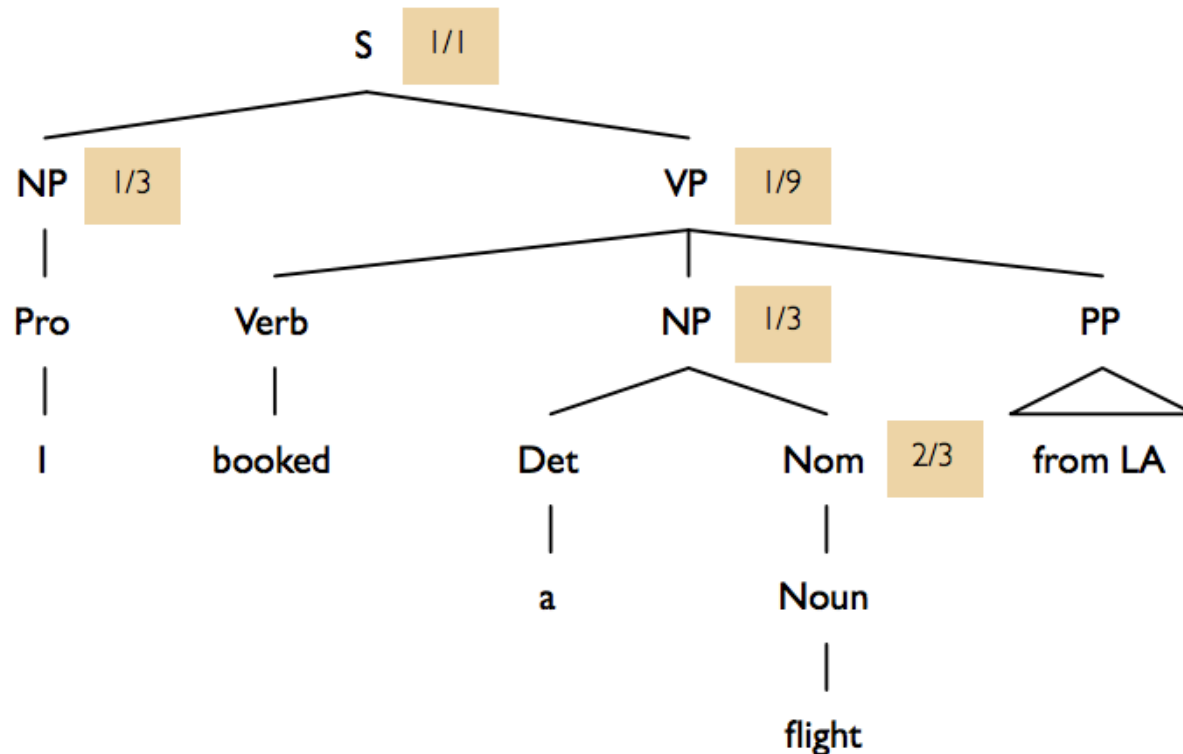
Rule	Probability
$S \rightarrow NPVP$	1
$NP \rightarrow \text{Pronoun}$	1/3
$NP \rightarrow \text{Proper-Noun}$	1/3
$NP \rightarrow \text{Det Nominal}$	1/3
$\text{Nominal} \rightarrow \text{Nominal PP}$	1/3
$\text{Nominal} \rightarrow \text{Noun}$	2/3
$VP \rightarrow \text{Verb NP}$	8/9
$VP \rightarrow \text{Verb NP PP}$	1/9
$PP \rightarrow \text{Preposition NP}$	1

Escolha da Melhor Parse Tree



Probability: 1/6729

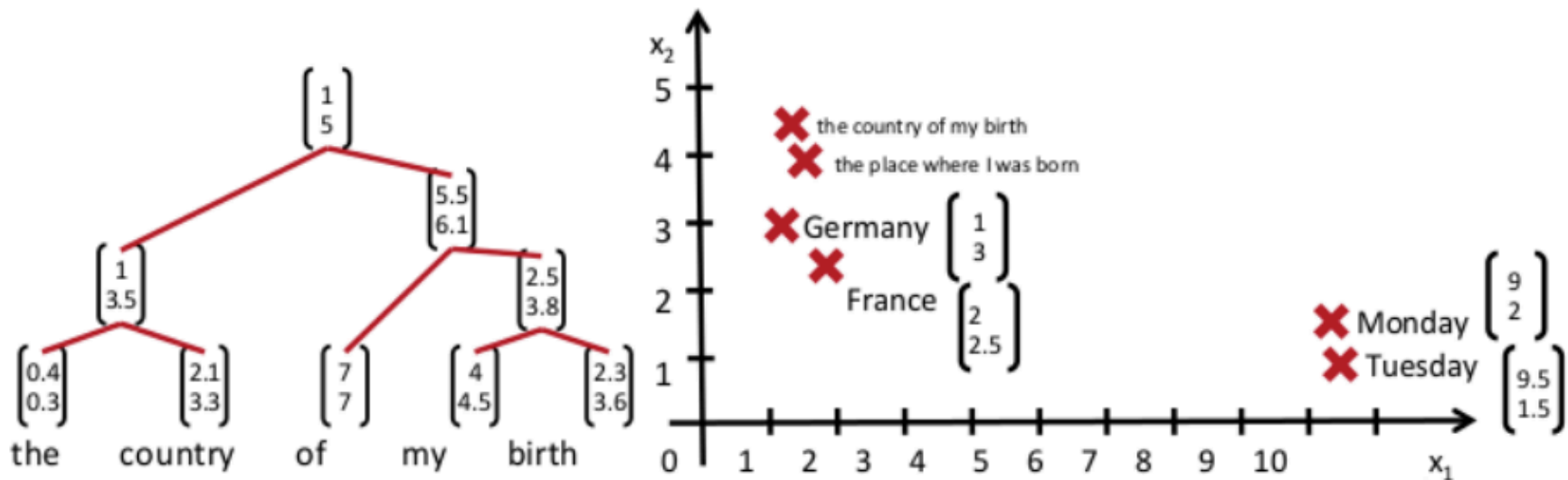
Escolha da Melhor Parse Tree



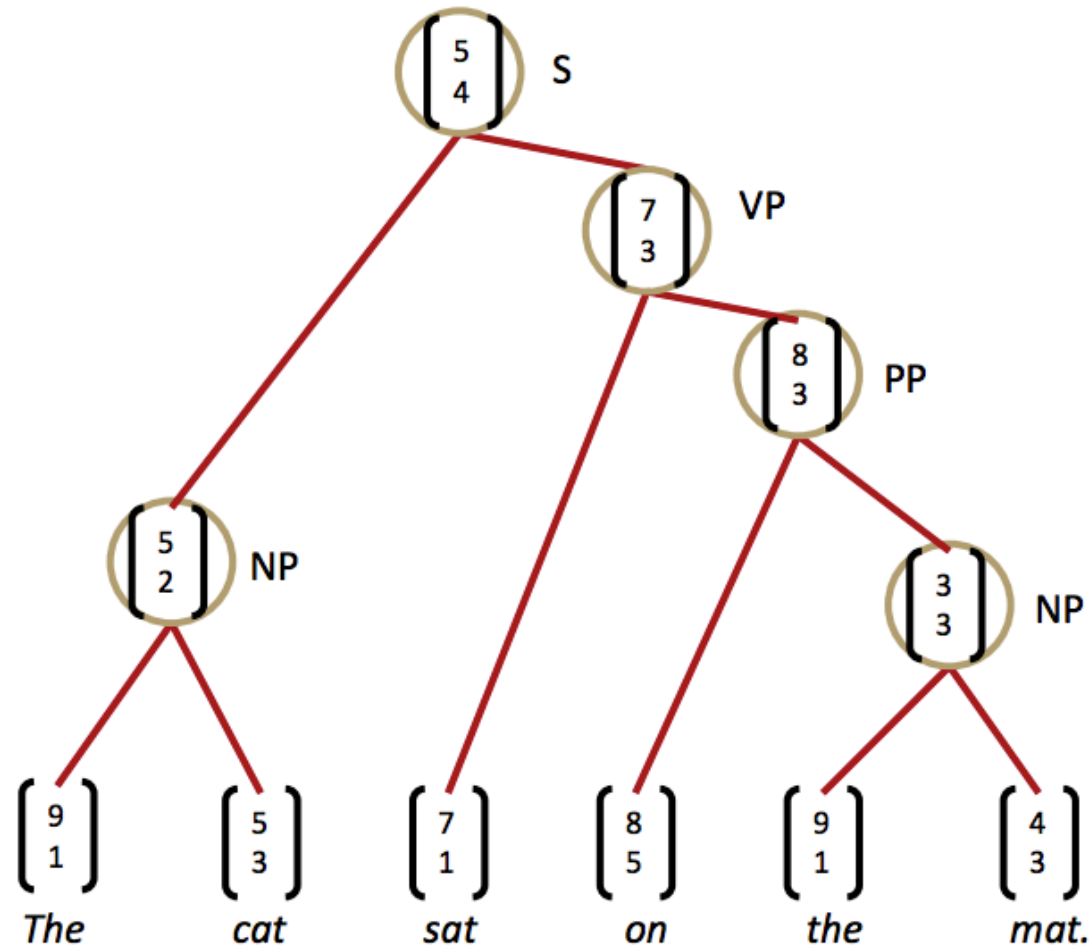
Probability: 6/729

Representando Sentenças

- Significado de uma sentença
 - Significado das palavras
 - Regras que as combinam

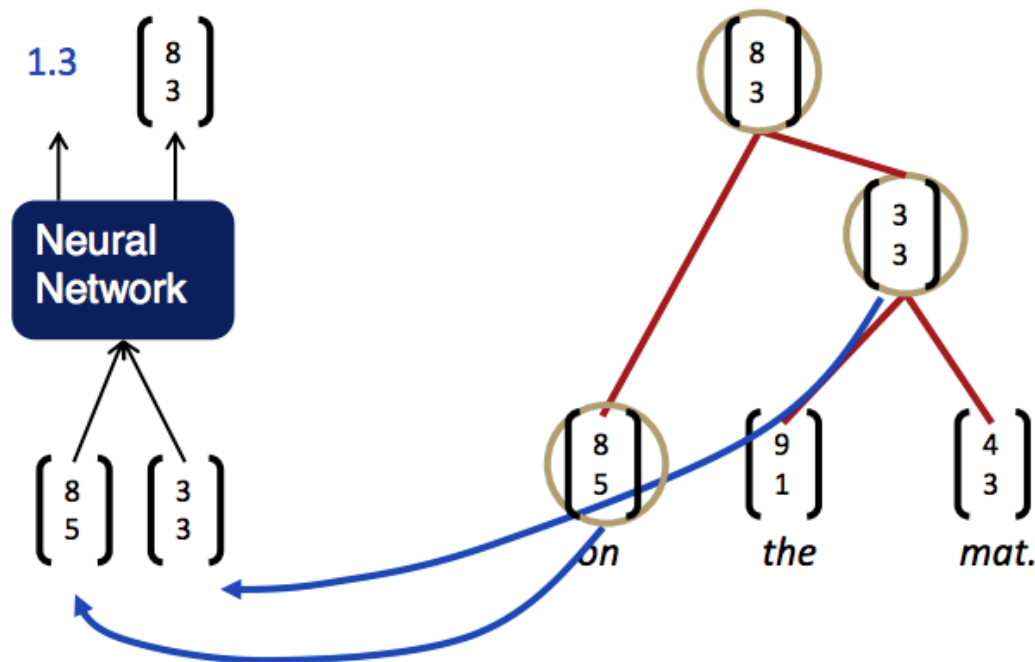


Representando Nós



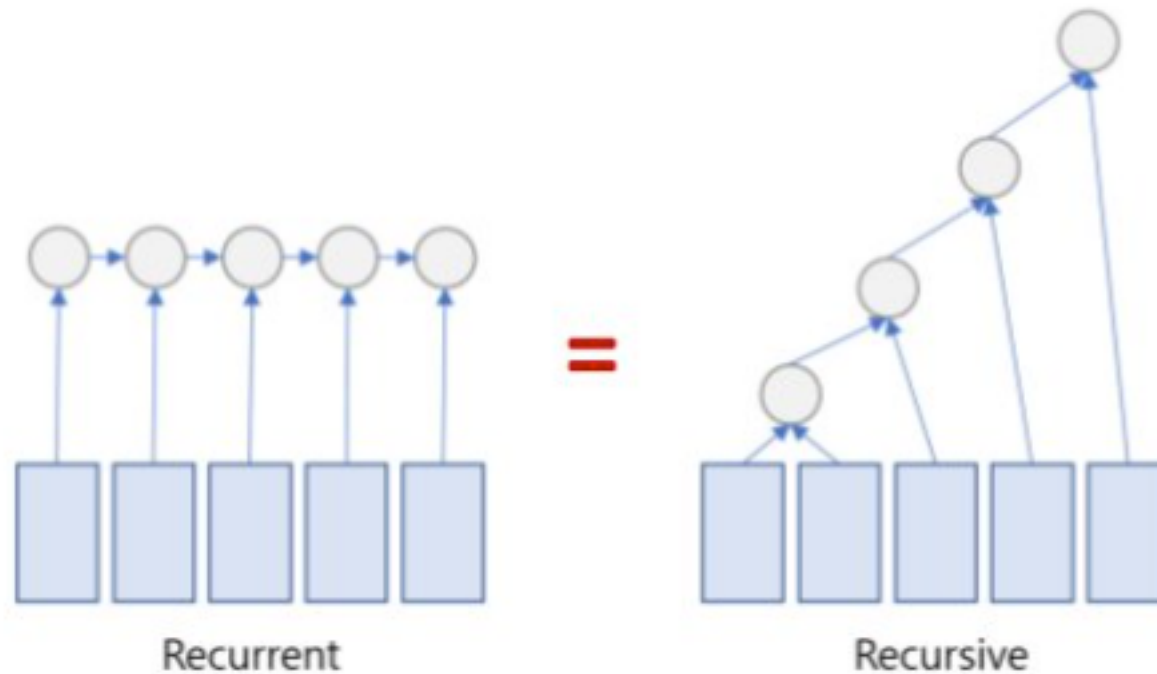
Recursive Neural Networks

- Entrada: representação dos nós filhos
- Saída:
 - A representação semântica dos nós
 - Score de quão plausível é essa junção

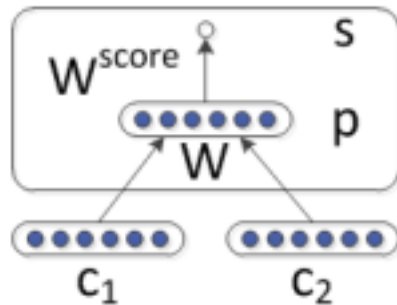


Recursive vs Recurrent

- Tipo de RNN

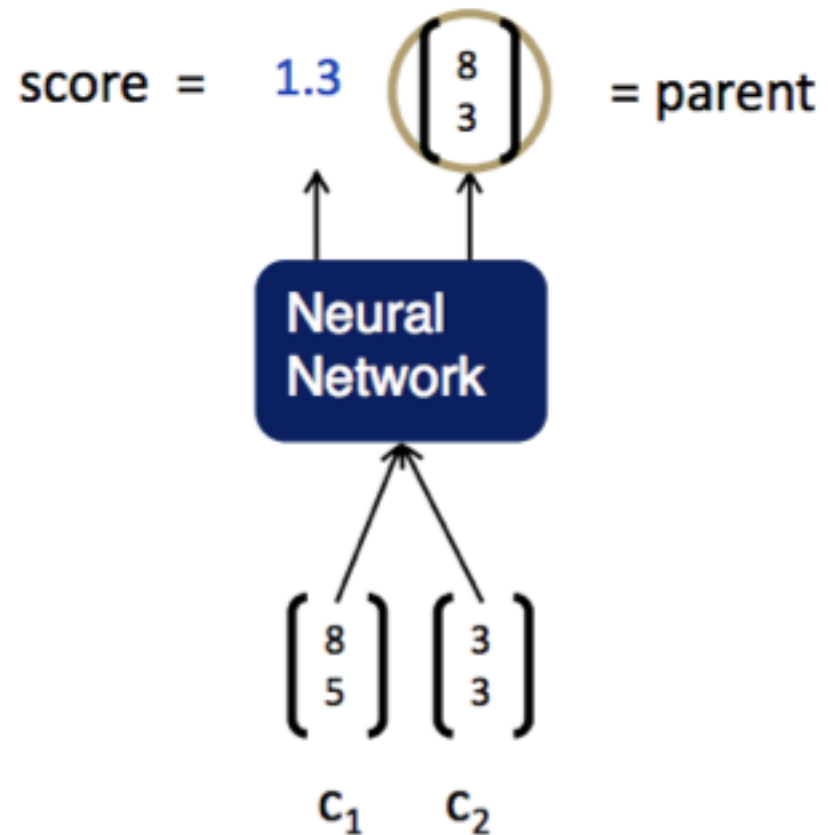


Recursive Neural Networks

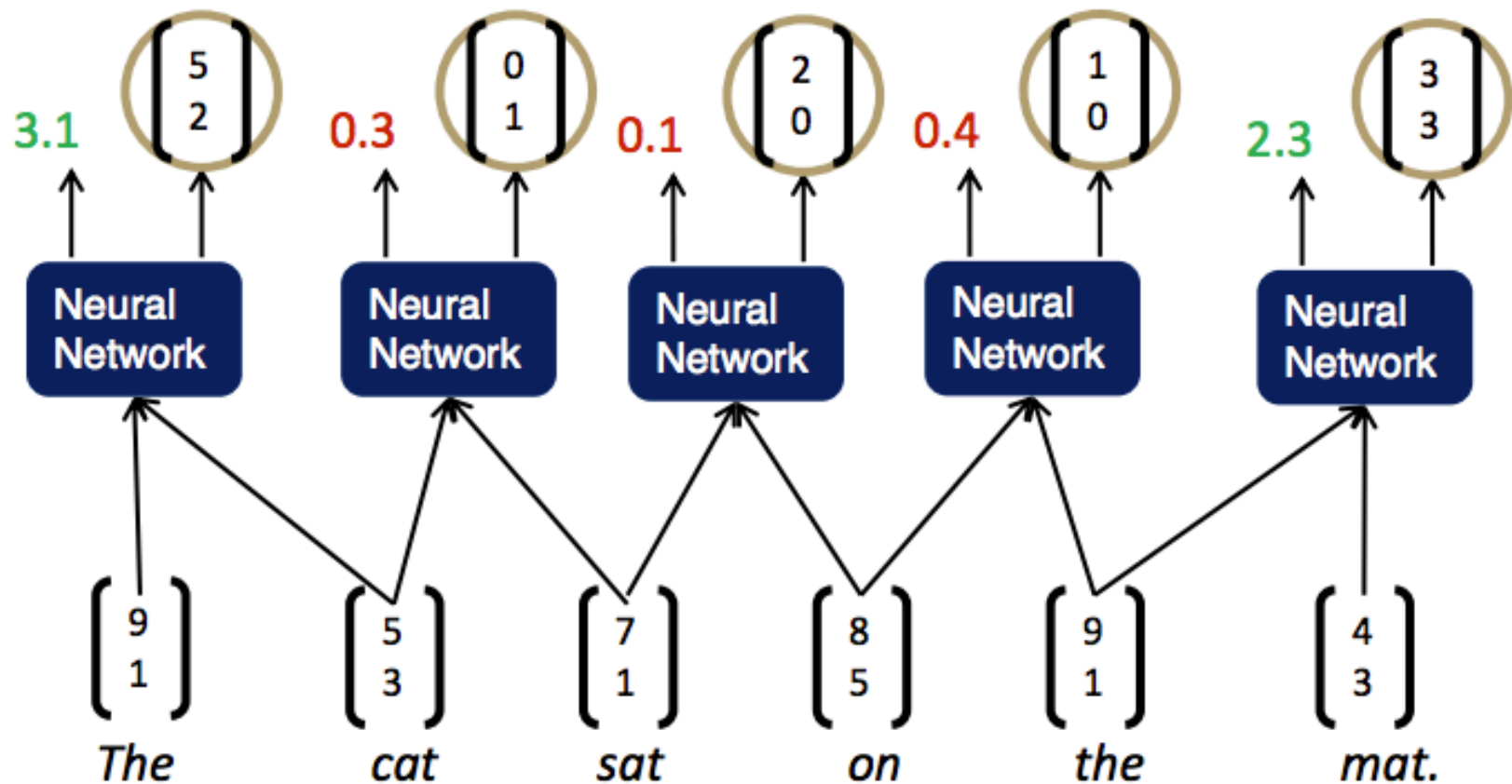


$$s = W^{score} p$$

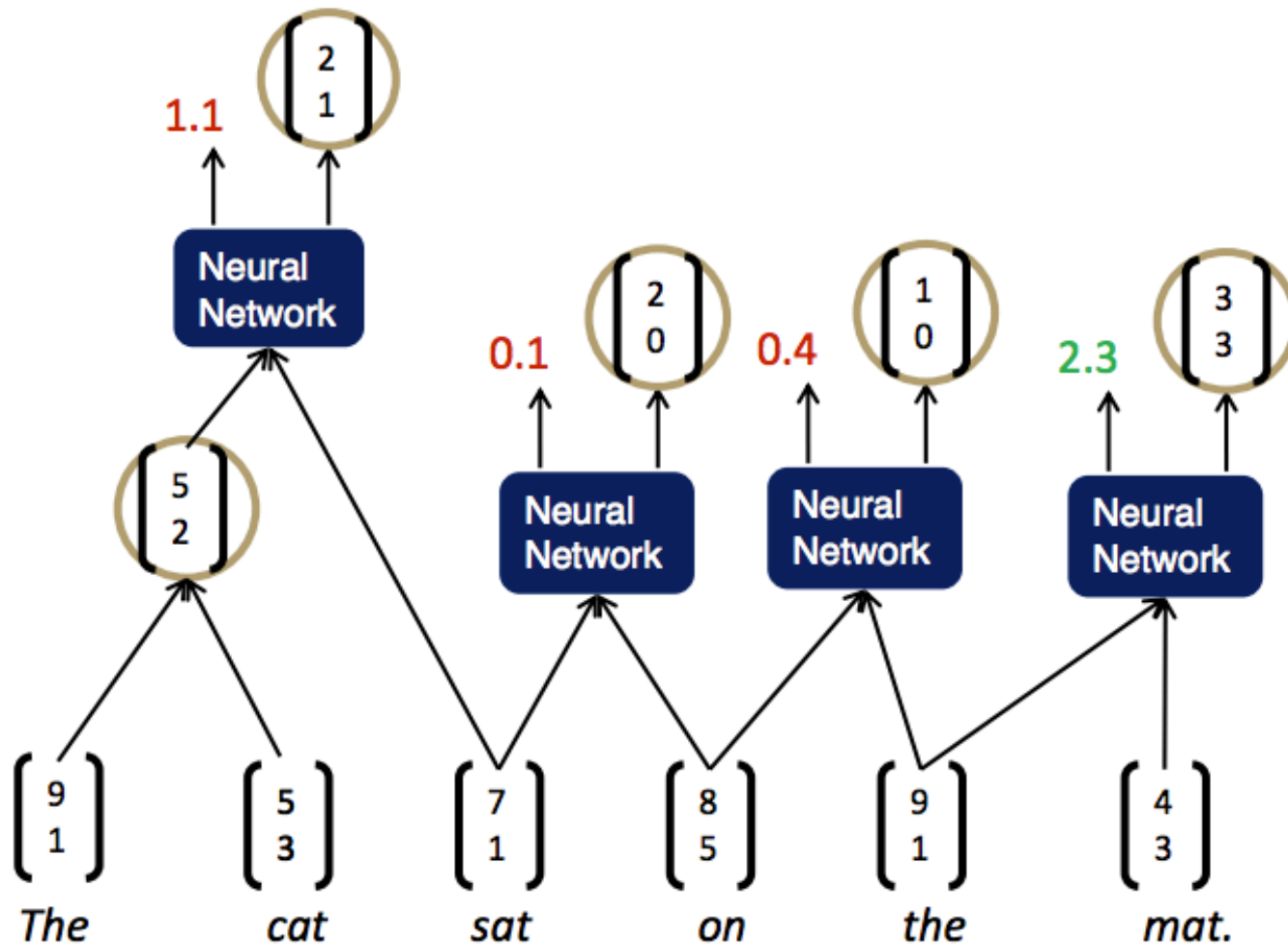
$$p = f(W[c_1; c_2] + b)$$



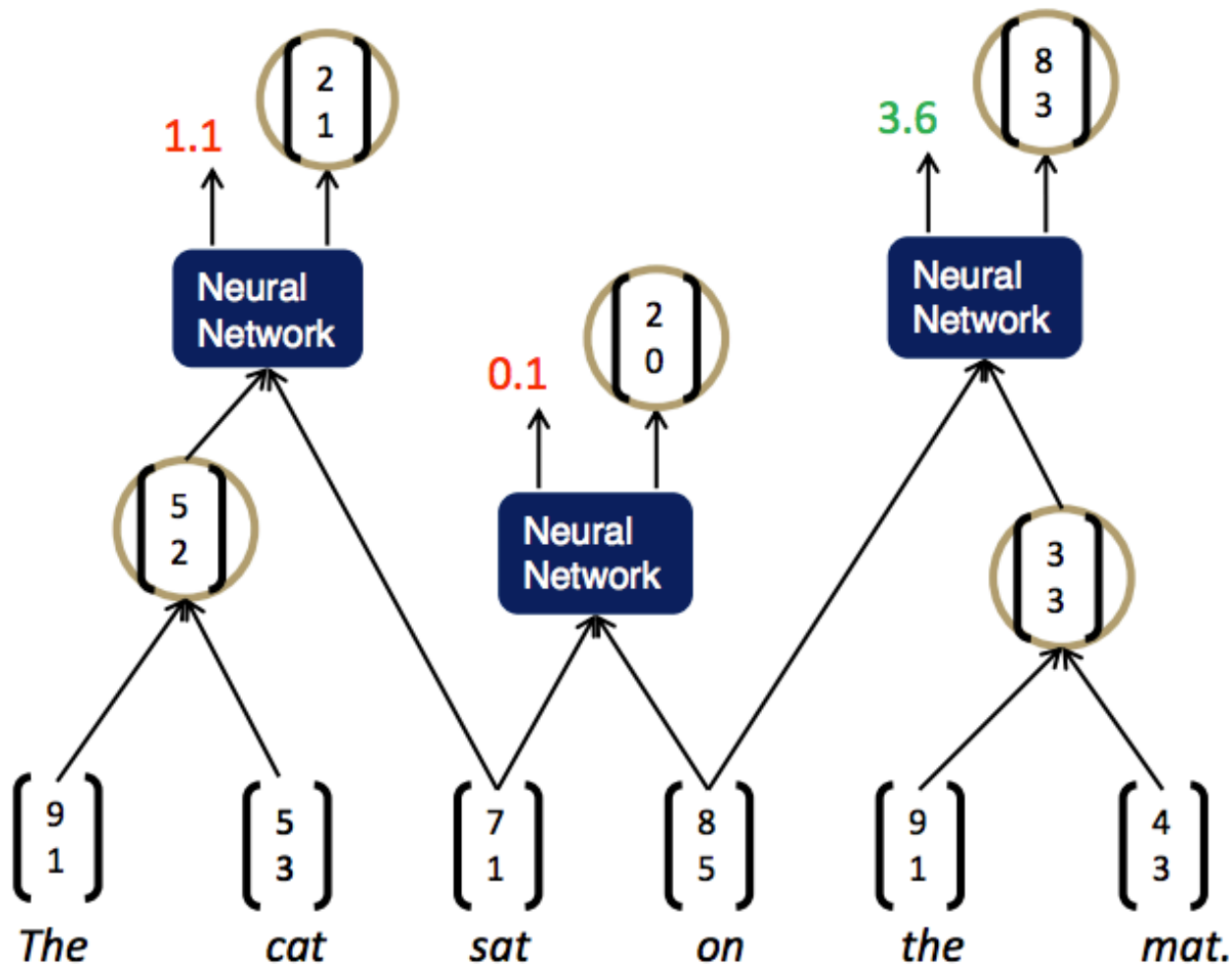
Recursive Neural Networks



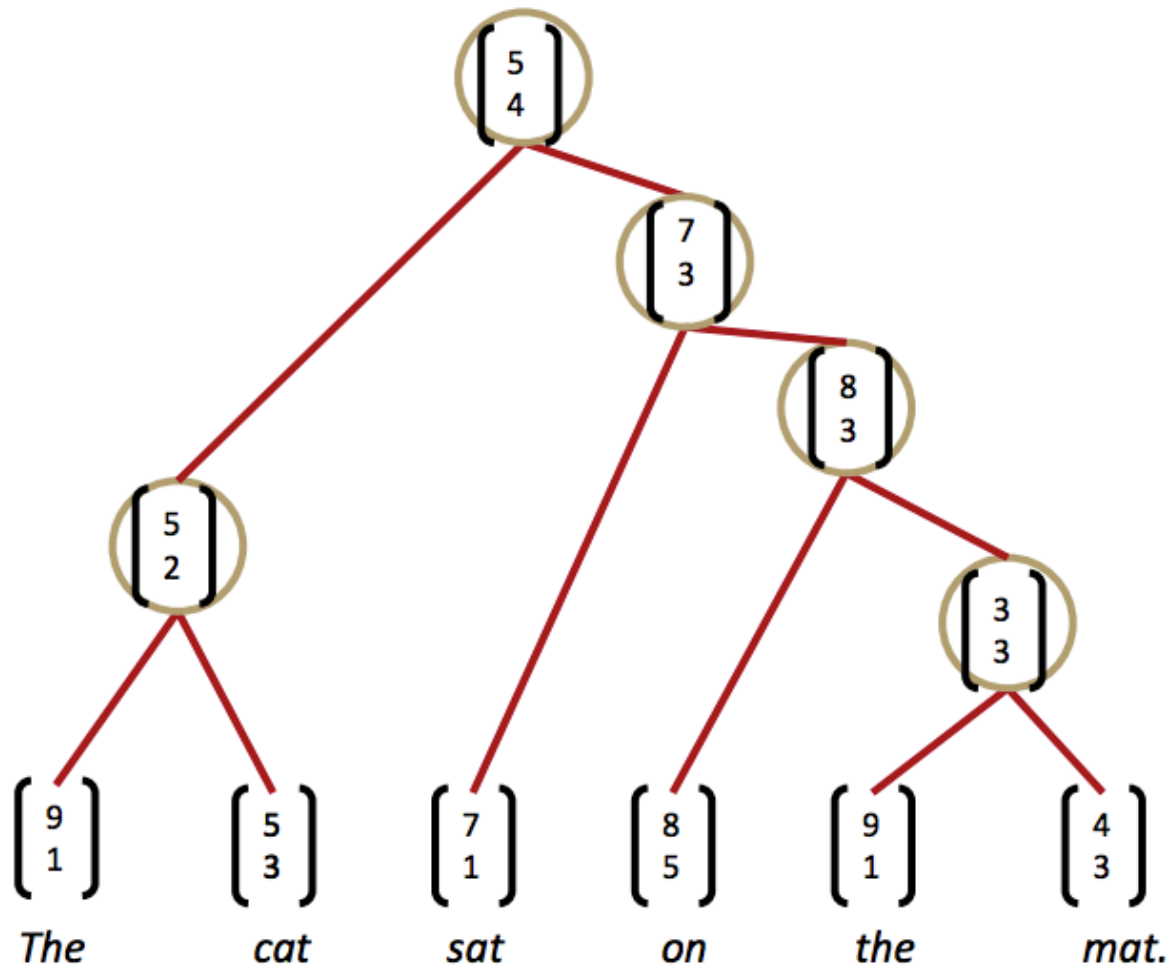
Recursive Neural Networks



Recursive Neural Networks

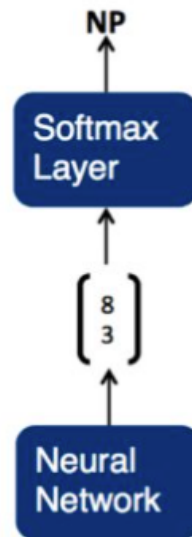


Recursive Neural Networks



Classificando Nós

- Categorias são os tipos gramaticais



Partial Parsing: Chunking

- Criação de parse trees completas pode ser custoso
- Chunking: Identifica as frases de uma sentença

[NP The morning flight] [PP from] [NP Denver] [VP has arrived.]

- Pode focar em um tipo específico de frase

[NP The morning flight] from [NP Denver] has arrived.

- Utiliza sequence labeling para classificar as frases

The morning flight from Denver has arrived
B_NP I_NP I_NP B_PP B_NP B_VP I_VP