

Natural Language Processing

CSCI 4152/6509 — Lecture 23

DCG and PCFG

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Time and date: 16:05 – 17:25, 28-Nov-2023

Location: Rowe 1011

Previous Lecture

- Natural language syntax:
 - ▶ phrase structure, clauses, sentences
 - ▶ Parsing, parse tree examples
- Context-Free Grammars review:
 - ▶ formal definition
 - ▶ inducing a grammar from parse trees
 - ▶ derivations, and other notions
- Bracket representation of a parse tree
- Parsing NL in Prolog using Difference Lists
- Reading: [JM] Ch 12

Basic Definite Clause Grammar (DCG)

- DCG — Prolog built-in mechanism for parsing

Example

s --> np, vp.

np --> d, n.

d --> [the].

n --> [dog].

n --> [dogs].

vp --> [run].

vp --> [runs].

Building a Parse Tree

A parse tree can be built in the following way:

```
s(s(Tn, Tv)) --> np(Tn), vp(Tv).
```

```
np(np(Td, Tn)) --> d(Td), n(Tn).
```

```
d(d(the)) --> [the].
```

```
n(n(dog)) --> [dog].
```

```
n(n(dogs)) --> [dogs].
```

```
vp(vp(run)) --> [run].
```

```
vp(vp(runs)) --> [runs].
```

At Prolog prompt we type and obtain:

```
?- s(X, [the, dog, runs], []).
```

```
   X = s(np(d(the), n(dog)), vp(runs));
```

Handling Agreement

$s(s(T_n, T_v)) \quad \rightarrow \quad np(T_n, A), \quad vp(T_v, A).$

$np(np(T_d, T_n), A) \quad \rightarrow \quad d(T_d), \quad n(T_n, A).$

$d(d(\text{the})) \quad \rightarrow \quad [\text{the}].$

$n(n(\text{dog}), \text{sg}) \quad \rightarrow \quad [\text{dog}].$

$n(n(\text{dogs}), \text{pl}) \quad \rightarrow \quad [\text{dogs}].$

$vp(vp(\text{run}), \text{pl}) \quad \rightarrow \quad [\text{run}].$

$vp(vp(\text{runs}), \text{sg}) \quad \rightarrow \quad [\text{runs}].$

This grammar will accept sentences “the dog runs” and “the dogs run” but not “the dog run” and “the dogs runs”. Other phenomena can be modeled in a similar fashion.

Embedded Code

We can embed additional Prolog code using braces, e.g.:

```
s(T) --> np(Tn), vp(Tv), {T = s(Tn,Tv)}.
```

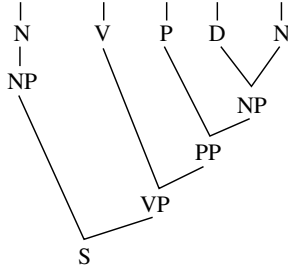
and so on, is another way of building the parse tree.

Probabilistic Context-Free Grammar (PCFG)

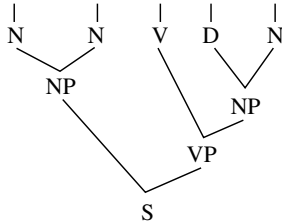
- Reading: Chapters 13 and 14
- also known as Stochastic Context-Free Grammar (SCFG)
- Handles ambiguous trees using a probabilistic model

Ambiguity Example

Time flies like an arrow.



Time flies like an arrow.



S	→	NP VP	VP	→	V NP	N	→	time	V	→	like
NP	→	N	VP	→	V PP	N	→	arrow	V	→	flies
NP	→	N N	PP	→	P NP	N	→	flies	P	→	like
NP	→	D N				D	→	an			

PCFG as a Probabilistic Model

- A generative model based on probabilistic derivation, for example:

$$S \Rightarrow NP VP \Rightarrow D N VP \Rightarrow \dots$$

- Each step is probabilistic use of one production

Probabilistic Context-Free Grammar Example

S	→	NP VP	/1	VP	→	V NP	/.5	N	→	time	/.5
NP	→	N	/.4	VP	→	V PP	/.5	N	→	arrow	/.3
NP	→	N N	/.2	PP	→	P NP	/1	N	→	flies	/.2
NP	→	D N	/.4					D	→	an	/1
V	→	like	/.3								
V	→	flies	/.7								
P	→	like	/1								

- The following condition must be satisfied for each nonterminal N :

$$\sum_{i=1}^n P(N \rightarrow \alpha_i) = 1$$

Computational Tasks for PCFG Model

- Evaluation

$$P(\text{tree}) = ?$$

- Generation

- Learning

- Inference

- ▶ Marginalization

$$P(\text{sentence}) = ?$$

- ▶ Conditioning

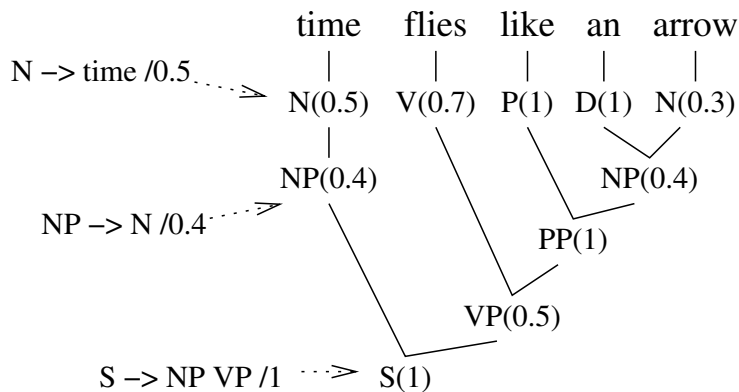
$$P(\text{tree}|\text{sentence}) = ?$$

- ▶ Completion

$$\arg \max_{\text{tree}} P(\text{tree}|\text{sentence})$$

Evaluation example: time flies like an arrow (1st meaning)

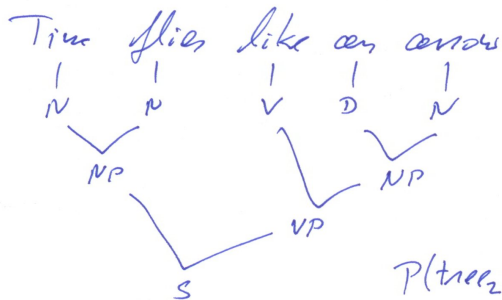
Evaluation



$$P(\text{tree}) = 0.5 \times 0.7 \times 1 \times 1 \times 0.3 \times 0.4 \times 0.4 \times 1 \times 0.5 \times 1 = 0.0084$$

Evaluation example: time flies like an arrow (2nd meaning)

Similarly



$$P(\text{tree}_2) = 0.00036$$

Generation (sampling)

S \Rightarrow NP VP \Rightarrow N VP \Rightarrow flies VP \Rightarrow ...

S \rightarrow NPVP / 1

NP \rightarrow N / 0.5

N \rightarrow time / 0.5

NP \rightarrow NN / 0.2

N \rightarrow sand / 0.3

NP \rightarrow DN / 0.4

N \rightarrow flies / 0.2

- choose rule randomly according to the given distribution

Question: Is the process going to stop?

A: Stops with probability 1 if the grammar is proper.

Good News: A grammar learned from a corpus is always proper.

Learning and Inference

Expressing PCFGs in DCGs

Let us consider the previous example of a PCFG:

S	→	NP VP	/1	VP	→	V NP	/.5	N	→	time	/.5
NP	→	N	/.4	VP	→	V PP	/.5	N	→	arrow	/.3
NP	→	N N	/.2	PP	→	P NP	/1	N	→	flies	/.2
NP	→	D N	/.4					D	→	an	/1
V	→	like	/.3								
V	→	flies	/.7								
P	→	like	/1								

The probabilities can be calculated as an addition argument:

$s(T,P) \rightarrow np(T1,P1), vp(T2,P2),$
 $\{T = s(T1,T2), P \text{ is } P1 * P2 * 1\}.$
 $np(T,P) \rightarrow n(T1,P1), \{T = n(T1), P \text{ is } P1 * 0.4\}.$
and so on.

Full PCFG Expressed in DCG

```
s(s(Tn,Tv),P) --> np(Tn,P1), vp(Tv,P2), {P is P1 * P2}.
np(np(T),P) --> n(T,P1), {P is P1 * 0.4}.
np(np(T1,T2),P) --> n(T1,P1), n(T2,P2),
                    {P is P1 * P2 * 0.2}.
np(np(Td,Tn),P) --> d(Td,P1), n(Tn,P2),
                    {P is P1 * P2 * 0.4}.
v(v(like), 0.3) --> [like].
v(v(flies), 0.7) --> [flies].
p(p(like), 1.0) --> [like].
vp(vp(Tv,Tn), P) --> v(Tv, P1), np(Tn, P2),
                    {P is P1 * P2 * 0.5}.
vp(vp(Tv,Tp), P) --> v(Tv, P1), pp(Tp, P2),
                    {P is P1 * P2 * 0.5}.
pp(pp(Tp,Tn), P) --> p(Tp, P1), np(Tn, P2),
                    {P is P1 * P2}.
n(n(time), 0.5) --> [time].
n(n(arrow), 0.3) --> [arrow].
```

Example Run in Prolog Interpreter

```
?- s(T,P,[time,flies,like,an,arrow],[ ]).
```

the interpreter would reply with: $T = s(np(n(time)),$
 $vp(v(flies), pp(p(like), np(d(an), n(arrow))))))$

$P = 0.0084$

and after typing ; (semi-colon), we get: $T = s(np(n(time), n(flies)),$
 $vp(v(like), np(d(an), n(arrow))))$

$P = 0.00036$

After typing second ';', the interpreter reports 'No' since there are no more parse trees.