

Optimizations and Extensions for Weighted CFG Parsers

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Outline

Pruning

- Pruning for CKY parsing

- Pruning for deductive parsing

k-best parsing

A*-parsing

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Pruning

- ▶ During CKY or deductive parsing many items are explored which are not part of the best derivation
- ▶ Idea: avoid items that are not part of the best derivation to speed up parsing
- ▶ Problem: How can we know these items in advance?
- ▶ Practical solution: Use simple methods but take the risk of finding suboptimal derivation.

Pruning for CKY parsing

Require: weighted binary cfg (N, Σ, P, S, μ) , word $t_1 \dots t_n$ where $t_1, \dots, t_n \in \Sigma$

Ensure: family $(c_{i,j,A} \in \mathbb{R} \mid 0 \leq i < j \leq n, A \in N)$ such that, for all i, j, A ,

$$c_{i,j,A} = \max\{\mu(d) \mid d \in D_G^A(t_{i+1} \dots t_j)\} \cup \{0\}$$

```
1: function CKY( $P, \mu, t_1 \dots t_n$ )
2:   ( $c_{i,j,A} := 0 \mid 0 \leq i < j \leq n, A \in N$ )
3:   for  $1 \leq i \leq n$  do
4:     for  $A \rightarrow t_i \in P$  do
5:        $c_{i-1,i,A} := \max\{c_{i-1,i,A}, \mu(A \rightarrow t_i)\}$ 
6:   for  $2 \leq r \leq n$  do
7:     for  $0 \leq i \leq n - r$  do
8:        $j := i + r$ 
9:       for  $m \in \{i + 1, i + 2, \dots, j - 1\}$  do
10:        for  $B, C \in N$  do
11:          for  $A \in N$  such that  $A \rightarrow BC \in R$  do
12:             $c_{i,j,A} := \max\{c_{i,j,A}, \mu(A \rightarrow BC) \cdot c_{i,m,B} \cdot c_{m,j,C}\}$ 
13:   return  $c$ 
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5:        $c_{i-1,i,A} := \max\{c_{i-1,i,A}, \mu(A \rightarrow t_i)\}$ 
6:       ( $c_{i-1,i,A} \mid A \in N$ ) := prune( $(c_{i-1,i,A} \mid A \in N)$ )
7:   for  $2 \leq r \leq n$  do
8:     for  $0 \leq i \leq n - r$  do
9:        $j := i + r$ 
10:      for  $m \in \{i + 1, i + 2, \dots, j - 1\}$  do
11:        for  $B, C \in N$  do
12:          if  $c_{i,m,B} = 0$  or  $c_{m,j,C} = 0$  then continue
13:          for  $A \in N$  such that  $A \rightarrow BC \in R$  do
14:             $c_{i,j,A} := \max\{c_{i,j,A}, \mu(A \rightarrow BC) \cdot c_{i,m,B} \cdot c_{m,j,C}\}$ 
15:            ( $c_{i,j,A} \mid A \in N$ ) := prune( $(c_{i,j,A} \mid A \in N)$ )
16:   return  $c$ 
```


Pruning for CKY parsing

► Threshold beam

Require: family $c = (c_{i,j,A} \in \mathbb{R} \mid A \in N)$, threshold $\theta \in [0, 1]$

Ensure: family $(c_{i,j,A} \in \mathbb{R} \mid A \in N)$

```
1: function PRUNE( $c$ )
2:    $m = \max_{A \in N} \{c_{i,j,A} \mid A \in N\}$ 
3:   for  $A \in N$  do
4:     if  $c_{i,j,A} < m \cdot \theta$  then
5:        $c_{i,j,A} := 0$ 
6:   return  $c$ 
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► Fixed-sized beam

Require: family $c = (c_{i,j,A} \in \mathbb{R} \mid A \in N)$, size $1 \leq n \leq |N|$

Ensure: family $(c_{i,j,A} \in \mathbb{R} \mid A \in N)$

```
1: function PRUNE( $c$ )
2:    $[s_1, \dots, s_n] = n\text{-best}\{c_{i,j,A} \mid A \in N\}$ 
3:   for  $A \in N$  do
4:     if  $c_{i,j,A} < s_n$  then
5:        $c_{i,j,A} := 0$ 
6:   return  $c$ 
```

Pruning for CKY parsing– implementation considerations

- ▶ No changes to data structures required.
- ▶ More speed-ups might be obtained by not adding items to chart which for sure would later be pruned:
 - ▶ Threshold beam: store weight m of currently best item. If new item has weight m below $\theta \cdot m$, it is safe to prune immediately.
 - ▶ Fixed-size beam: store weights of the n best items. If the weight of new is below of worst item, prune immediately.

Pruning for deductive parsing

Require: weighted binary cfg (N, Σ, P, S, μ) , word $t_1 \dots t_n$ where $t_1, \dots, t_n \in \Sigma$

Ensure: family $(c_{i,j,A} : \mathbb{R} \mid 0 \leq i < j \leq n, A \in N)$ such that

$$c_{i,j,A} = \max\{\mu(d) \mid d \in D_G^A(t_{i+1} \dots t_j)\} \cup \{0\}$$

```
1: function DEDUCE( $P, \mu, t_1 \dots t_n$ )
2:    $queue := \{(i-1, A, i, \mu(A \rightarrow t_i)) \mid 1 \leq i \leq n, A \rightarrow t_i \in P\}$ 
3:    $(c_{i,j,A} := 0 \mid 0 \leq i < j \leq n, A \in N)$ 
4:   while  $queue \neq \emptyset$  do
5:      $(i, A, j, w) := \operatorname{argmax}_{(i,A,j,w) \in queue} w$ 
6:      $queue \setminus = \{(i, A, j, w)\}$ 
7:     if  $c_{i,j,A} = 0$  then
8:        $c_{i,j,A} := w$ 
9:        $queue \cup = \{(i, A', j', \mu(A' \rightarrow AC) \cdot w \cdot c_{j,j',C}) \mid A' \rightarrow AC \in P\}$ 
10:       $queue \cup = \{(i', A', j, \mu(A' \rightarrow BA) \cdot c_{i',i,B} \cdot w) \mid A' \rightarrow BA \in P\}$ 
11:       $queue \cup = \{(i, A', j, \mu(A' \rightarrow A) \cdot w) \mid A' \rightarrow A \in P\}$ 
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- 10: $queue \cup= \{(i', A', j, \mu(A' \rightarrow BA) \cdot c_{i',i,B} \cdot w) \mid A' \rightarrow BA \in P\}$
- 11: $queue \cup= \{(i, A', j, \mu(A' \rightarrow A) \cdot w) \mid A' \rightarrow A \in P\}$
- 12: **prune**($queue$)
- 13: **return** c

Pruning for deductive parsing

► Threshold beam

Require: set $queue \subseteq \mathbb{N} \times \mathcal{N} \times \mathbb{N} \times \mathbb{R}$, threshold $\theta \in [0, 1]$

Ensure: set $queue' \subseteq \mathbb{N} \times \mathcal{N} \times \mathbb{N} \times \mathbb{R}$

- 1: **function** PRUNE($queue$)
- 2: $m = \max_{(i,A,j,w) \in queue} w$
- 3: **return** $\{(i, A, j, w) \in queue \mid w > \theta \cdot m\}$

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▶ Fixed-sized beam

Require: set $queue \subseteq \mathbb{N} \times \mathcal{N} \times \mathbb{N} \times \mathbb{R}$, size $n \in \mathbb{N}$

Ensure: set $queue' \subseteq \mathbb{N} \times \mathcal{N} \times \mathbb{N} \times \mathbb{R}$

- 1: **function** PRUNE($queue$)
- 2: $[i_1, \dots, i_n] = n\text{-best}(queue)$ w.r.t
- 3: **return** $\{i_1, \dots, i_n\}$

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k -best parsing

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 - ▶ “with the telescope” modifies “saw”
 - ▶ “with the telescope” modifies “the astronomer”
- ▶ Goal: given a sentence w , a PCFG G , and a positive integer k , find the k most probable derivations of G for w

k -best parsing – naïve

Require: $k \in \mathbb{N}$, weighted binary CFG (N, Σ, S, R, μ) , word $t_1 \cdots t_n$

Ensure: k most probable parse trees of PCFG for $t_1 \cdots t_n$

```
1: function KBEST( $k, R, \mu, t_1, \dots, t_n$ )
2:    $b[i, j, A] := []$  for each cell  $(i, j, A)$ 
3:   for  $i \in \{0, \dots, n-1\}$  do
4:      $c := \{(A(t_{i+1}), w) \mid A \rightarrow t_{i+1} \text{ in } R, w = \mu(A \rightarrow t_{i+1})\}$ 
5:      $b[i, j, A] = \text{take}(k, \text{sort}(c))$ 
6:   for  $z \in \{2, \dots, n\}$  do
7:     for  $i \in \{0, \dots, n-z\}$  do
8:        $j := i + z$ 
9:       for  $A \in N$  do
10:         $c := \{(A(d_1, d_2), w) \mid A \rightarrow BC \text{ in } R, m \in \{i+1, \dots, j-1\},$ 
            $(d_1, w_1) \in b[i, m, B], (d_2, w_2) \in b[m, j, C],$ 
            $w = \mu(A \rightarrow BC) \cdot w_1 \cdot w_2\}$ 
11:         $b[i, j, A] = \text{take}(k, \text{sort}(c))$ 
12:   return  $b[0, n, S]$ 
```

k -best parsing – implementation of merging

- 10: $c := \{(A(d_1, d_2), w) \mid A \rightarrow BC, m \in \{i+1, \dots, j-1\},$
 $(d_1, w_1) \in b[i, m, B], (d_2, w_2) \in b[m, j, C],$
 $w = \mu(A \rightarrow BC) \cdot w_1 \cdot w_2\}$
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can be implemented as

- 10: $b[i, j, A] := []$
- 11: **for** $m \in \{i+1, \dots, j-1\}$ **do**
- 12: **for** $A \rightarrow BC$ **in** R **do**
- 13: $c := \{(A(d_1, d_2), w) \mid (d_1, w_1) \in b[i, m, B], (d_2, w_2) \in b[m, j, C],$
 $w = \mu(A \rightarrow BC) \cdot w_1 \cdot w_2\}$
- 14: $b[i, j, A] := \text{mergeAndTakeK}(k, b[i, j, A], c)$

k-best parsing – merging more efficient

2			
2	↑		
0	1	⇒	
	1	2	4

(a)

2			
2	3	↑	
0	1	2	⇒
	1	2	4

(b)

2			
2	3	4	
0	1	2	4
	1	2	4

(c)

- ```

10: $b[i, j, A] := []$
11: for $m \in \{i + 1, \dots, j - 1\}$ do
12: for $A \rightarrow BC$ in R do
13: denote $w_{u,v} := \mu(A \rightarrow BC) \cdot w_u^1 \cdot w_v^2$ where
 $(d_u^1, w_u^1) := b[i, m, B][u]$ and
 $(d_v^2, w_v^2) := b[m, j, C][v]$
14: $F := \{(1, 1)\}$
15: while $\max_{(u,v) \in F} w_{u,v} > \min_{(d,w) \in b[i,j,A]} w$ or $|b[i, j, A]| < k$ do
16: $(u, v) = \operatorname{argmax}_{(u,v) \in F} w_{u,v}$
17: insertAndTakeK($k, (A(d_u^1, d_v^2), w_{u,v}), b[i, j, A]$)
18: $F := (F \setminus \{(u, v)\}) \cup \{(u + 1, v), (u, v + 1)\}$

```

[HC05]

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- ▶ Klein and Manning [KM03] propose several admissible heuristics.

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  - ▶ Why?: Usually items with small spans are more probable than items with large spans.
  - ▶ This is counteracted by future costs which are higher for items with small spans.
- ▶ Klein and Manning [KM03] propose several admissible heuristics.
  
- ▶ A heuristic may also be useful when pruning items during CKY parsing.

## A\*-parsing– Viterbi outside score

We use the admissible heuristic out:

$$\text{out}(A) = \max_{d \in D_G, u, w \in \Sigma^* : S \xrightarrow{d}_G uAw} \text{weight}(d)$$

It can be computed by a variant of the inside/outside algorithm:

```
1: function INSIDE
2: for $A \in N$ do
3: $\text{in}(A) := \max(\{\mu(A \rightarrow \alpha) \mid A \rightarrow \alpha \in R\} \cup \{0\})$
4: while not converged do
5: for $A \in N$ do
6: $\text{in}(A) = \max(\{\text{in}(A)\} \cup \{\mu(A \rightarrow BC) \cdot \text{in}(B) \cdot \text{in}(C) \mid A \rightarrow BC \text{ in } R\})$
7: function OUTSIDE
8: set $\text{out}(B) := \begin{cases} 1 & B = S \\ 0 & \text{otherwise} \end{cases}$ for each $B \in N$
9: while not converged do
10: for $B \in N$ do
11: $\text{out}(B) := \max(\{\text{out}(B)\} \cup \{\text{out}(A) \cdot \mu(A \rightarrow BC) \cdot \text{in}(C) \mid A \rightarrow BC \text{ in } R\}$

 $\cup \{\text{out}(A) \cdot \mu(A \rightarrow CB) \cdot \text{in}(C) \mid A \rightarrow CB \text{ in } R\})$
```

# A\*-parsing – parsing algorithm with heuristic

**Require:** weighted binary cfg  $(N, \Sigma, P, S, \mu)$ , word  $t_1 \dots t_n$  where  $t_1, \dots, t_n \in \Sigma$

**Ensure:** family  $(c_{i,j,A} : \mathbb{R} \mid 0 \leq i < j \leq n, A \in N)$  such that

$$c_{i,j,A} = \max\{\mu(d) \mid d \in D_G^A(t_{i+1} \dots t_j)\} \cup \{0\}$$

```
1: function DEDUCE($P, \mu, t_1 \dots t_n$)
2: $queue := \{(i-1, A, i, \mu(A \rightarrow t_i)) \mid 1 \leq i \leq n, A \rightarrow t_i \in P\}$
3: $(c_{i,j,A} := 0 \mid 0 \leq i < j \leq n, A \in N)$
4: while $queue \neq \emptyset$ do
5: $(i, A, j, w) := \operatorname{argmax}_{(i,A,j,w) \in queue} w \cdot \text{out}(A)$
6: $queue \setminus= \{(i, A, j, w)\}$
7: if $c_{i,j,A} = 0$ then
8: $c_{i,j,A} := w$
9: $queue \cup= \{(i, A', j', \mu(A' \rightarrow AC) \cdot w \cdot c_{j,j',C}) \mid A' \rightarrow AC \in P\}$
10: $queue \cup= \{(i', A', j, \mu(A' \rightarrow BA) \cdot c_{i',i,B} \cdot w) \mid A' \rightarrow BA \in P\}$
11: $queue \cup= \{(i, A', j, \mu(A' \rightarrow A) \cdot w) \mid A' \rightarrow A \in P\}$
12: return c
```

- [Atk+86] M. D. Atkinson et al. “Min-max Heaps and Generalized Priority Queues”. In: *Commun. ACM* 29.10 (Oct. 1986), pp. 996–1000. ISSN: 0001-0782. DOI: 10.1145/6617.6621. URL: <http://doi.acm.org/10.1145/6617.6621>.
- [HC05] Liang Huang and David Chiang. “Better k-best parsing”. In: *Proceedings of the Ninth International Workshop on Parsing Technology*. Association for Computational Linguistics. 2005, pp. 53–64.
- [KM03] Dan Klein and Christopher D. Manning. “A\* Parsing: Fast Exact Viterbi Parse Selection”. In: *Proceedings of the 2003 Human Language Technology Conference of the North American Chapter of the Association for Computational Linguistics*. 2003, pp. 119–126. URL: <https://www.aclweb.org/anthology/N03-1016>.