Deep Learning for Natural Language Processing

The Eisner algorithm

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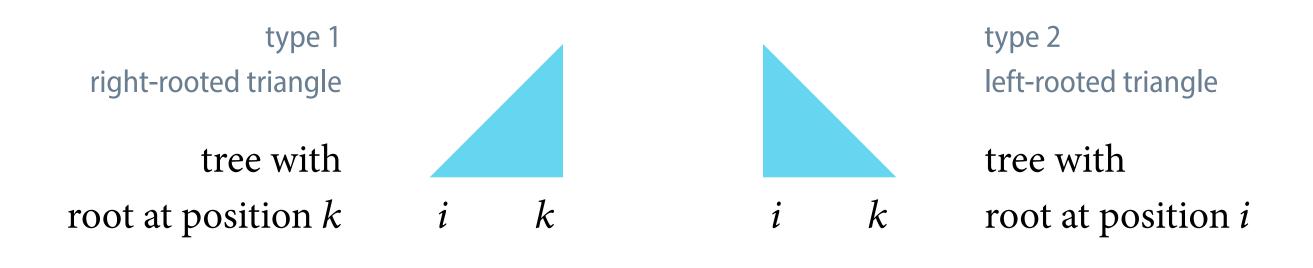
The Eisner algorithm

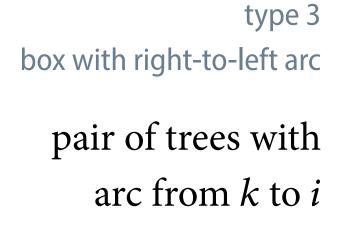
- The Eisner algorithm is an algorithm for computing the highestscoring projective dependency tree under an arc-factored model.
- It solves this problem using bottom—up dynamic programming, storing solutions to sub-problems in a table.

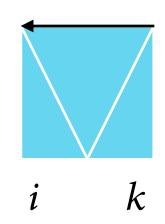
Overview of the Eisner algorithm

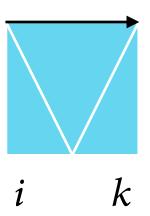
- We are given an input sentence x and a table A that holds the score of each possible arc: $A[h][d] = \text{score}(x, h \rightarrow d)$
- We will fill four tables T_t such that each entry $T_t[i][k]$ will hold the maximal possible score of a certain type of graph, where i and k identify the leftmost and the rightmost words in the graph.
- Once we are done, we will use the tables to compute the maximal possible score of a projective tree for the full sentence.

Sub-problems in the Eisner algorithm









box with left-to-right arc pair of trees with arc from i to k

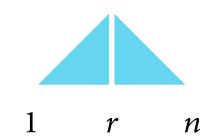
Filling the score tables

```
# Assume that all table entries are initialised with -inf
foreach i from 1 to n:
     T_1[i][i] = 0
     T_2[i][i] = 0
foreach k from 2 to n:
     foreach i from k-1 downto 1:
           T_4[i][k] = \max \text{ over } j \text{ from } i \text{ to } k-1 \text{ of } T_2[i][j] + T_1[j+1][k] + A[i][k]
           T_3[i][k] = \max \text{ over } j \text{ from } i \text{ to } k-1 \text{ of } T_2[i][j] + T_1[j+1][k] + A[k][i]
           T_2[i][k] = \max \text{ over } j \text{ from } i+1 \text{ to } k \text{ of } T_4[i][j] + T_2[j][k]
           T_1[i][k] = \max \text{ over } j \text{ from } i \text{ to } k-1 \text{ of } T_1[i][j] + T_3[j][k]
```

Computing the maximal score for the full sentence

• To obtain the maximal possible score of a projective dependency tree for a full sentence of length *n*, we compute

$$\max_{r} (T_1[1][r] + T_2[r][n])$$



• Alternatively, we can introduce a special 'pseudo-root' with position 0 and directly extract $T_2[0][n]$, the maximal possible score of a triangle rooted at the pseudo-root.

allows solutions with multiple 'real roots'

Filling the backpointer tables

```
# Assume that all table entries are initialised with None
foreach i from 1 to n:
    B_1[i][i] = None
    B_2[i][i] = None
foreach k from 2 to n:
    foreach i from k-1 downto 1:
         B_4[i][k] = argmax over j from i to k-1 of <math>T_2[i][j] + T_1[j+1][k] + A[i][k]
         B_3[i][k] = argmax over j from i to k-1 of <math>T_2[i][j] + T_1[j+1][k] + A[k][i]
         B_2[i][k] = argmax over j from i+1 to k of <math>T_4[i][j] + T_2[j][k]
         B_1[i][k] = argmax over j from i to k-1 of <math>T_1[i][j] + T_3[j][k]
```

Constructing the tree from the backpointers

```
def build(t, i, k):
    j = B_t[i][k] # retrieve the entry from the backpointer table
    if j == None:
                  return Ø
    if t == 4: return build(2, i, j) \cup build(1, j+1, k) \cup {(i, k)}
    if t == 3: return build(2, i, j) \cup build(1, j+1, k) \cup {(k, i)}
    if t == 2: return build(4, i, j) U build(2, j, k)
    if t == 1: return build(1, i, j) \cup build(3, j, k)
# To construct the full tree (right-rooted triangle + left-rooted triangle):
arcs = build(1, 1, r) \cup build(2, r, n)
```

Complexity analysis of the Eisner algorithm

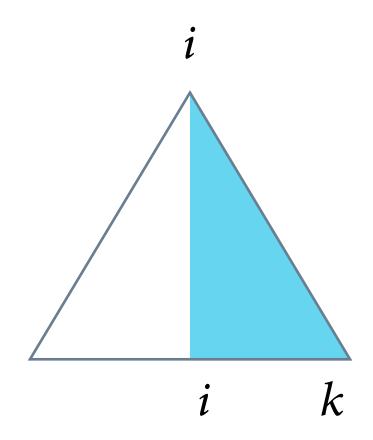
Let *n* be the length of the input sentence.

- The space complexity of the Eisner algorithm is in $O(n^2)$; this corresponds to the number of cells in a table T_t .
- The runtime complexity of the Eisner algorithm is in $O(n^3)$; this corresponds to the number of nested *for* loops that we need to enumerate sub-problems and compute maximal values.

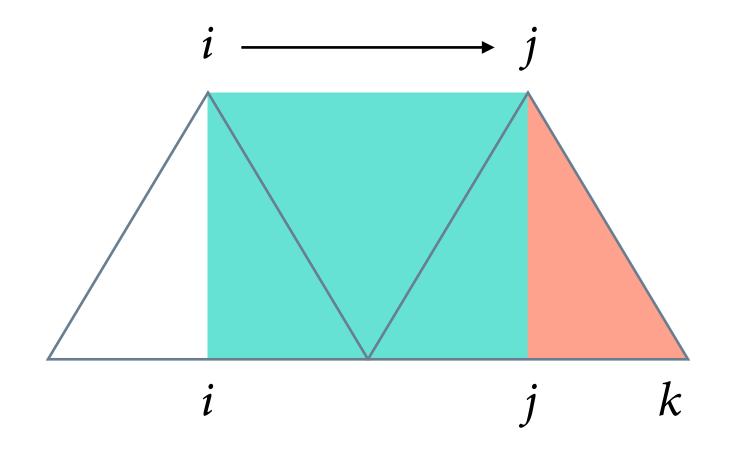
Lemma: For every type t and all $i \le k$, the value $T_t[i][k]$ is the maximal possible score of a graph of type t with endpoints i and k.

Proof: by induction on the size of the graph, k - i

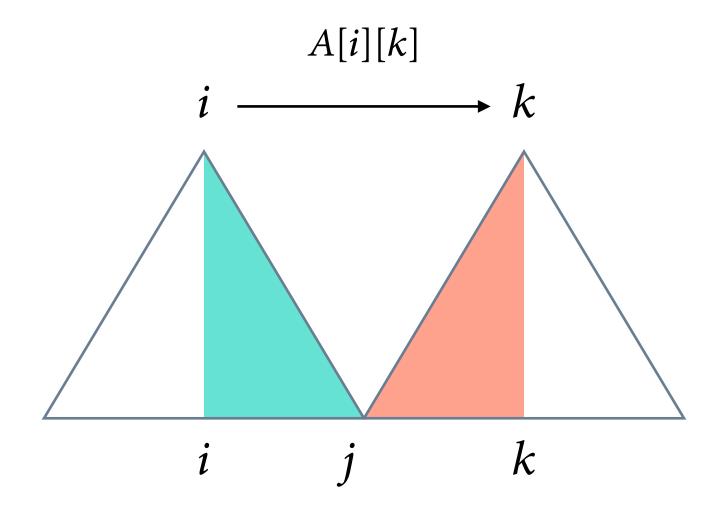
- In the simplest case, we have a graph with a single vertex and no arcs. The score of such a graph is zero.
- In the general case, we have a graph with at least one arc. We can then decompose the graph into two smaller graphs.



 $T_2[i][k]$?



$$T_2[i][k] = \max_{j} (T_4[i][j] + T_2[j][k])$$



$$T_4[i][k] = \max_j \left(\frac{T_2[i][j]}{T_1[j+1][k]} + A[i][k] \right)$$