

# Dynamic Programming on Tensors for Solving the Problem of Dependency Parsing in NLP

Optimization&NLA course project

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# Outline

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Algorithms

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Conclusion

# Introduction

- Syntactic dependency parsing is the important problem in statistical natural language processing.
- Our goal was to speed up the parsing process.
- Using tensor formulation of the inside-outside algorithm we compared different tensor decompositions.
- Tucker and Tensor Train decompositions were applied.
- We were able to achieve significant performance increase with good accuracy results.

# Setup

## Probabilistic Context-free Grammars

$$G = (\mathcal{N}, \mathcal{L}, \mathcal{R}, \mathcal{P}, \pi),$$

- $\mathcal{N}$  — nonterminal symbols.
- $\mathcal{L}$  — words (lexical tokens).
- $\mathcal{R}$  — set of rules:  $a \rightarrow bc$  or  $a \rightarrow x$ ,  $a, b, c \in \mathcal{N}$ ,  $x \in \mathcal{L}$ .
- $\mathcal{P}$  — transition probabilities  $p(a \rightarrow bc|a)$  and  $p(a \rightarrow x|a)$ .
- $\pi_a$  — probability of  $a$  being the root symbol.
- All probabilities satisfy normalization conditions.

# Setup

Our purpose: given a sentence find the most probable tree of this sentence.

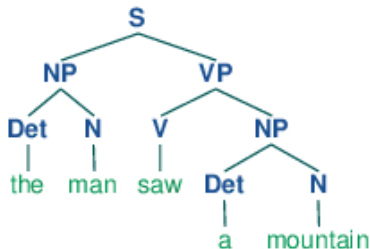
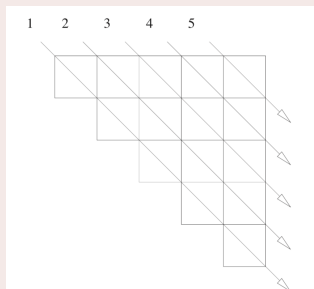


Figure: Parse tree representation of *the man saw a mountain*

# Algorithms

## Inside-Outside



She	N	S	S		S
eats		V	V		V
pizza			N		N, N-P
		without		PP	P
			anchovies		N

Figure: Inside Recursion

# Algorithms

## Inside-Outside

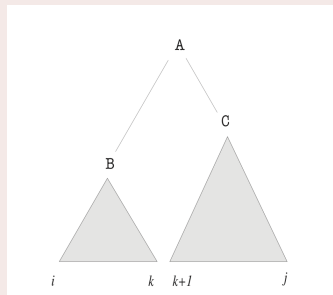
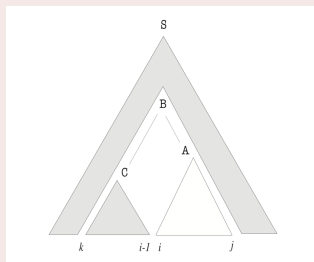


Figure: Inside Recursion

# Algorithms

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## Algorithm 1 Inside-Outside Algorithm in the Tensor Form

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1: **(Inside base case):**  $\forall (a \rightarrow x_i) \in \mathcal{R} [\alpha^{i,i}]_a = p(a \rightarrow x|a)$

2: **(Inside recursion):**

$$[\alpha^{i,j}]_a = \sum_{k=i}^{j-1} \sum_{a \rightarrow bc} T^{a \rightarrow bc}([\alpha^{i,k}]_b, [\alpha^{k+1,j}]_c)$$

3: **(Outside base case):**  $\forall a \in \mathcal{N} [\beta^{1,N}]_a = \pi_a$

4: **(Outside recursion):**

$$[\beta^{i,j}]_a = \sum_{k=1}^{i-1} \sum_{b \rightarrow ca} T_{(1,2)}^{b \rightarrow ca}([\beta^{k,j}]_b, [\alpha^{k,i-1}]_c) + \sum_{k=j+1}^N \sum_{b \rightarrow ac} T_{(1,3)}^{b \rightarrow ac}([\beta^{i,k}]_b, [\alpha^{j+1,k}]_c)$$

5: **(Marginals):**  $\mu(a, i, j) = [\alpha^{ij}]_a \cdot [\beta^{ij}]_a$

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# Tensor Decompositions (3D case)

## Canonical

$$\mathcal{A} = \sum_{i=1}^r \lambda_i \mathbf{a}_i^1 \otimes \mathbf{a}_i^2 \otimes \mathbf{a}_i^3,$$

$\mathcal{A} \in \mathbb{R}^{n_1 \times n_2 \times n_3}$ ,  $\lambda_i \in \mathbb{R}$  and  $\mathbf{a}_i^j \in \mathbb{R}^{n_j}$

**Time of multiplying by vector (of size  $n_3$ ):**  $\mathcal{O}(rn_3)$

## Tensor Train

$$\mathcal{A}(i_1, i_2, i_3) = \sum_{k_1=1}^{r_1} \sum_{k_2=1}^{r_2} G_1(i_1, k_1) G_2(k_1, i_2, k_2) G_3(k_2, i_3),$$

$\mathcal{A} \in \mathbb{R}^{n_1 \times n_2 \times n_3}$ ,  $G_1 \in \mathbb{R}^{n_1 \times r_1}$ ,  $G_2 \in \mathbb{R}^{r_1 \times n_2 \times r_2}$ ,  $G_3 \in \mathbb{R}^{r_2 \times n_3}$ .

**Time of multiplying by vector (of size  $n_3$ ):**  $\mathcal{O}(r_1 r_2 n_3)$

## Tensor Decompositions (3D case)

## Tucker

$$\text{vec}(\mathcal{A}) = (W \otimes V \otimes U) \cdot \text{vec}(\mathcal{C}),$$

$\mathcal{A} \in \mathbb{R}^{n_1 \times n_2 \times n_3}$ ,  $U \in \mathbb{R}^{n_1 \times r_1}$ ,  $V \in \mathbb{R}^{n_2 \times r_2}$ ,  $W \in \mathbb{R}^{n_3 \times r_3}$  and  
 $\mathcal{C} \in \mathbb{R}^{r_1 \times r_2 \times r_3}$

**Time of multiplying by vector (of size  $n_3$ ):**

$$\mathcal{O}(n_3 r_3 + r_1 r_2 r_3 + r_1 r_2 n_1 + r_2 n_2) = \mathcal{O}(nr^2 + r^3).$$

# Experiments

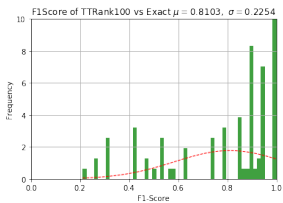
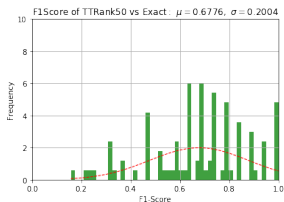
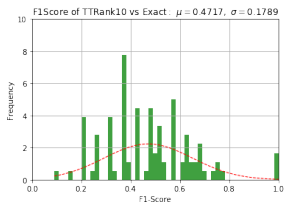


Figure: F1 scores TT

## Experiments

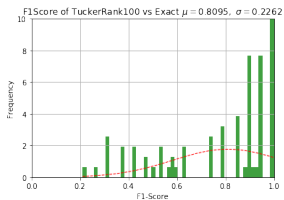
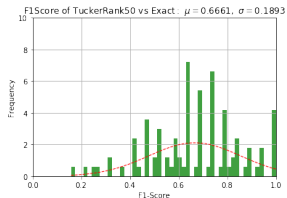
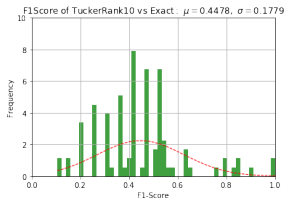


Figure: F1 scores Tucker

## Experiments

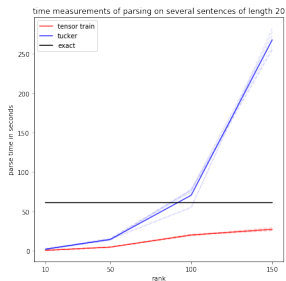


Figure: Time efficiency

# Experiments

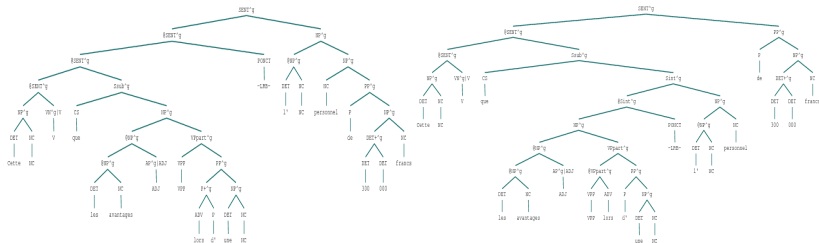


Figure: Tree comparison

## Conclusion

- Using tensor formulation of the inside-outside algorithm we tried different tensor decompositions to speed up parsing process.
- Tucker and Tensor Train decompositions were applied.
- Experiments showed that using tensor decompositions we can achieve significant speed up with the good accuracy of parsing.