# Calculation of Inside and Outside Weights of Weighted Hypergraphs by Newton's Method 

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## Motivation

- Calculation of the corpus likelihood.
- Inside weight is needed in the E-step of the EM-algorithm
- Normalization, transformation of pCFGs [2]


## Motivation

Task

Preliminaries
Hypergraphs
Inside and outside weights

Approximation
Fixed-point iteration
Newton's method
Decomposed Newton's method

Performance

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- Calculation of Inside and Outside weights in weighted hypergraphs


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- by Newton's method


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- by Newton's method
- embedded in Vanda


## Hypergraphs



- weighted $\Sigma$-hypergraph $H=(V, E, \mu)$
- vertices $V$
- edges $E \subseteq V^{*} \times \Sigma \times V$
- weights $\mu: E \rightarrow \mathbb{R}_{\geq 0}$


## Hypergraphs


hyperpath tree of edges, connected according to the hypergraph, leaves have a rank of 0

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context tree of edges and one vertex, connected according to the hypergraph, leaves have a rank of 0 , exactly one leave is a vertex

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## Inside and outside weights

inside weight for a vertex $v$ : sum of the weights of all hyperpaths in $H$ leading to $v$

$$
\operatorname{inside}(v)=\sum_{t \in \operatorname{HP}(H, v)} w t(t)
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outside weight for a vertex $v^{\prime}$ and a target vertex $v$ : sum of the weights of all $v$-rooted $v^{\prime}$-contexts in $H$

$$
\text { outside }^{v}\left(v^{\prime}\right)=\sum_{c \in C^{v}\left(H, v^{\prime}\right)} w t(c)
$$

## Fixed-point iteration

- Recursive system

$$
\operatorname{inner}(v)=\sum_{\substack{e \in E \\ e=(w, \sigma, v)}} \mu(e) \cdot \prod_{i \in[|w|]} \operatorname{inner}\left(w_{i}\right)
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e=(w, \sigma, v)}} \mu(e) \cdot \prod_{i \in[|w|]} \operatorname{inner}\left(w_{i}\right) \\
\text { outer }^{v^{\prime}}(v) & =s(v)+\sum_{\substack{e \in E \\
e=(w, \sigma, \hat{v}) \\
i \in \mathbb{N}: w_{i}=v}} \text { outer }^{v^{\prime}}(\hat{v}) \cdot \mu(e) \cdot \prod_{\substack{j \in[|w|] \\
j \neq i}} \operatorname{inner}\left(w_{j}\right) \\
s(v) & = \begin{cases}1 & \text { if } v=v^{\prime} \\
0 & \text { otherwise }\end{cases}
\end{aligned}
$$

## Newton's method

- inside and outside weights as root

$$
0=-i_{v}+\sum_{\substack{e \in E \\ e=(w, \sigma, v)}} \mu(e) \cdot \prod_{i \in[|w|]} i_{w_{i}}
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Input a function $f: \mathbb{R} \rightarrow \mathbb{R}$, a starting value $x_{0} \in \mathbb{R}$

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3. If $x_{n+1}=x_{n}$, output $x_{n}$.

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Input a function $f: \mathbb{R}^{m} \rightarrow \mathbb{R}^{m}$, a starting value $x_{0} \in \mathbb{R}^{m}$

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x_{n+1}=x_{n}-\left(J f\left(x_{n}\right)\right)^{-1} \cdot f\left(x_{n}\right)
$$

## Inside weights as root of a function



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$$
\begin{array}{r}
{i n^{\prime}(i)_{v}=}=-i_{v}+\sum_{\substack{e \in E \\
e=(w, \sigma, v)}} \mu(e) \cdot \prod_{i \in[|w|]} i_{w_{i}} \\
i n^{\prime}(i)=\left(\begin{array}{c}
-i_{A}+\mu\left(e_{0}\right)+\mu\left(e_{4}\right) \cdot i_{C} \cdot i_{B} \\
-i_{B}+\mu\left(e_{1}\right) \\
-i_{C}+\mu\left(e_{2}\right)+\mu\left(e_{5}\right) \cdot i_{D} \cdot i_{A} \\
-i_{D}+\mu\left(e_{3}\right)
\end{array}\right)
\end{array}
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4. (Compute polynomials.)
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For the hypergraph $H_{\{B\}}$ we get

$$
\begin{aligned}
i n_{\{B\}}^{\prime} & =\left(-i_{B}+\mu\left(e_{1}\right)\right) \\
& =\left(1-i_{B}\right) \\
\left(\operatorname{Jin}_{\{B\}}^{\prime}\right)^{-1} & =(-1) .
\end{aligned}
$$

Then by Newton's method in:

$$
i_{B}=1
$$

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4. (Compute polynomials.)
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For the hypergraph $H_{\{D\}}$ we get

$$
\begin{aligned}
i n_{\{D\}}^{\prime} & =\left(-i_{D}+\mu\left(e_{3}\right)\right) \\
& =\left(1-i_{D}\right) \\
\left(\operatorname{Jin}_{\{D\}}^{\prime}\right)^{-1} & =(-1) .
\end{aligned}
$$

Then by Newton's method:

$$
i_{D}=1
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## Decomposed Newton's method

4. (Compute polynomials.)
5. Substitute known inside weights in the polynomials with values.
6. (Apply Newton's method.)

For the hypergraph $H_{\{A, C\}}$ we get

$$
\begin{aligned}
\operatorname{in}_{\{A, C\}}^{\prime} & =\binom{-i_{A}+\mu\left(e_{0}\right)+\mu\left(e_{4}\right) \cdot i_{C} \cdot i_{B}}{-i_{C}+\mu\left(e_{2}\right)+\mu\left(e_{5}\right) \cdot i_{D} \cdot i_{A}} \\
& =\binom{-i_{A}+0.25+0.25 \cdot i_{C}}{-i_{C}+0.25+0.25 \cdot i_{A}} \\
\left(\operatorname{Jin}_{\{B\}}^{\prime}\right)^{-1} & =-\frac{4}{15}\left(\begin{array}{ll}
4 & 1 \\
1 & 4
\end{array}\right) .
\end{aligned}
$$

Then by Newton's method:

$$
i_{A}=\frac{1}{3}, i_{C}=\frac{1}{3} .
$$

## Performance



## Performance



## Decomposed fixed-point method



## The End

Thank you for your attention!

## References

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