

Efficient Algorithms for Frequently Asked Questions

9. Solving Functional Aggregate Queries

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DaST 
Data • (Systems+Theory)

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Agenda for this Lecture

Solving Functional Aggregate Queries (FAQs)

- We start with FAQs over a single semiring, dubbed FAQ-SS
 - Uses LFTJ to compute the bags of a hypertree decomposition
 - Uses Yannakakis to aggregate away bound variables eagerly
- We then continue with FAQs over multiple semirings
 - Can we swap marginalisation order for variables under different semirings?
 - How can we deal with product aggregates?

$$\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}} \cdots \bigoplus_{x_{n-1}} \bigoplus_{x_n} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

with hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ and semiring $(\mathbf{D}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$

- Input: Hypertree decomposition \mathcal{T} of Φ , factors $(\psi_S)_{S \in \mathcal{E}}$
- Runtime: $\mathcal{O}(N^{\text{fhtw}(\mathcal{T})} + \text{OUT})$ where OUT is the output size of Φ
- By choosing an optimal hypertree decomposition for the hypergraph \mathcal{H} of Φ , the runtime becomes $\mathcal{O}(N^{\text{fhtw}(\mathcal{H})} + \text{OUT})$

Main Steps of the FAQ-SS Solver

Input: FAQ-SS Φ with hypergraph \mathcal{H} and a hypertree decomposition \mathcal{T} of \mathcal{H}

Step 1: Query Rewriting

1.1 Turn Φ into an equivalent α -acyclic FAQ Φ' by materialising the bags of \mathcal{T}

1.2 Choose an order σ for variable marginalisation *compatible* with \mathcal{T}

1.3 Rewrite Φ' into an FAQ Φ'' following σ by moving sums past products

Step 2: Query Evaluation

- Evaluate Φ'' by marginalising the bound variables *inside out*

We next explain the above steps using an example

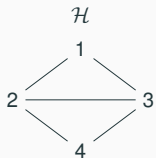
FAQ-SS Solver Example

Example of Solving FAQ-SS

Sum-product FAQ-SS:

$$\Phi(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{23}(x_2, x_3) \cdot \psi_{24}(x_2, x_4) \cdot \psi_{34}(x_3, x_4)$$

with hypergraph \mathcal{H} and hypertree decomposition \mathcal{T} :

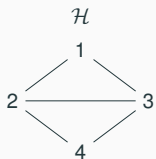


Example of Solving FAQ-SS

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with hypergraph \mathcal{H} and hypertree decomposition \mathcal{T} :



Φ can be computed in time $\mathcal{O}(N^2)$:

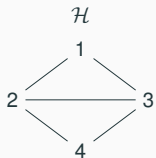
- Compute the join of the factors in time $\mathcal{O}(N^2)$ (since $\rho_{\mathcal{H}}^*(\{1, 2, 3, 4\}) = 2$)
- Marginalise out the variables X_2 and X_4

Example of Solving FAQ-SS

Sum-product FAQ-SS:

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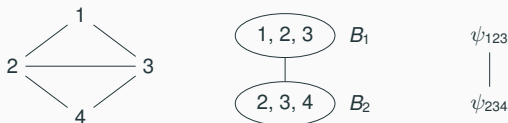
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- Compute the join of the factors in time $\mathcal{O}(N^2)$ (since $\rho_{\mathcal{H}}^*(\{1, 2, 3, 4\}) = 2$)
- Marginalise out the variables X_2 and X_4

Next: a strategy to compute Φ in time $\mathcal{O}(N^{\frac{3}{2}})$

Step 1.1: Turning Φ into an α -acyclic FAQ

$$\Phi(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{23}(x_2, x_3) \cdot \psi_{24}(x_2, x_4) \cdot \psi_{34}(x_3, x_4)$$



Turn Φ into an α -acyclic FAQ by **materialising** the bags of \mathcal{T} :

- Construct a factor $\psi_{123}(x_1, x_2, x_3)$ for B_1
- Construct a factor $\psi_{234}(x_2, x_3, x_4)$ for B_2
- Construct the α -acyclic query that has the join tree above:

$$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4)$$

Step 1.1: Turning Φ into an α -acyclic FAQ

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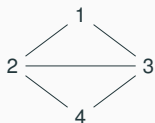
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$$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4)$$

How to construct ψ_{123} and ψ_{234} ?

Constructing the Factors for the Bags (1/2)



$$\begin{array}{c} \psi_{123} \\ | \\ \psi_{234} \end{array}$$

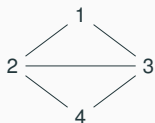
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First strategy:

compute time

- $\psi_{123}(x_1, x_2, x_3) := \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3)$ $\mathcal{O}(N^2)$
- $\psi_{234}(x_2, x_3, x_4) := \psi_{23}(x_2, x_3) \cdot \psi_{24}(x_2, x_4) \cdot \psi_{34}(x_3, x_4)$ $\mathcal{O}(N^{\frac{3}{2}})$

Constructing the Factors for the Bags (1/2)



$$\begin{array}{c} \psi_{123} \\ | \\ \psi_{234} \end{array}$$

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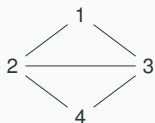
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Q: How can we reduce the compute time for ψ_{123} ?

Constructing the Factors for the Bags (1/2)



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First strategy:

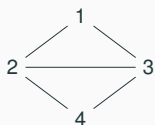
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Q: How can we reduce the compute time for ψ_{123} ?

A: Include ψ_{23} into the computation of ψ_{123}

Constructing the Factors for the Bags (2/2)



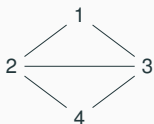
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Second strategy:

compute time

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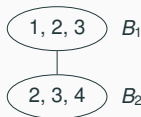
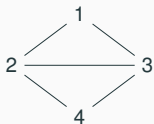
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$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4)$ is **NOT** equivalent to Φ

- The payloads of tuples in ψ_{23} are used twice in Φ' !

Constructing the Factors for the Bags (2/2)



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Second strategy:

compute time

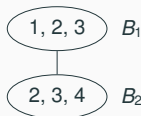
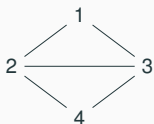
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Q: How to ensure Φ' is equivalent to Φ while keeping the compute time $\mathcal{O}(N^{\frac{3}{2}})$?

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- The payloads of tuples in ψ_{23} are used twice in Φ' !

Q: How to ensure Φ' is equivalent to Φ while keeping the compute time $\mathcal{O}(N^{\frac{3}{2}})$?

A: Use the **indicator projection** of ψ_{23} in ψ_{234}

Indicator Projections

$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4)$, where

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In ψ_{234} we only need to know which tuples **exist** in ψ_{23}

Indicator Projections

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In ψ_{234} we only need to know which tuples **exist** in ψ_{23}

Indicator projection: New factor $\psi_{23/234}$ maps the tuples of ψ_{23} to 1:

$$\psi_{23/234}(x_2, x_3) = \begin{cases} \mathbf{1}, & \text{if } \psi_{23}(x_2, x_3) \neq \mathbf{0} \\ \mathbf{0}, & \text{otherwise} \end{cases}$$

Notation 23/234: $\psi_{23/234}$ only retains values for the variables in ψ_{234}

Indicator Projections

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Notation 23/234: $\psi_{23/234}$ only retains values for the variables in ψ_{234}

The factor for the second bag over variables 2, 3, 4 is then:

$$\psi_{234}(x_2, x_3, x_4) := \psi_{23/234}(x_2, x_3) \cdot \psi_{24}(x_2, x_4) \cdot \psi_{34}(x_3, x_4)$$

Using Indicator Projections

Next strategy uses all indicator projections:

- Same asymptotic complexity as previous strategy: $\mathcal{O}(N^{\frac{3}{2}})$
- It may be more beneficial in practice

$$\psi_{123}(x_1, x_2, x_3) = \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{23}(x_2, x_3) \cdot \underbrace{\psi_{24/123}(x_2) \cdot \psi_{34/123}(x_3)}_{\text{indicator projections}}$$

$$\psi_{234}(x_2, x_3, x_4) = \psi_{24}(x_2, x_4) \cdot \psi_{34}(x_3, x_4) \cdot \underbrace{\psi_{12/234}(x_2) \cdot \psi_{13/234}(x_3) \cdot \psi_{23/234}(x_2, x_3)}_{\text{indicator projections}}$$

$$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4) \text{ is equivalent to } \Phi$$

Step 1.2: Find Good Marginalisation Order and Rewrite Φ'

$$\Phi'(X_1, X_3) = \sum_{X_2} \sum_{X_4} \psi_{123}(X_1, X_2, X_3) \cdot \psi_{234}(X_2, X_3, X_4)$$

Step 1.2: Find Good Marginalisation Order and Rewrite Φ'

$$\Phi'(x_1, x_3) = \sum_{x_2} \sum_{x_4} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{234}(x_2, x_3, x_4)$$

An approach to compute Φ' in time $\mathcal{O}(N^{\frac{3}{2}})$:

- The summation over x_4 is only relevant to the second bag factor
- First aggregate away x_4 over ψ_{234} , then join and finally aggregate away x_2

$$\Phi''(x_1, x_3) = \sum_{x_2} \left(\psi_{123}(x_1, x_2, x_3) \cdot \left(\sum_{x_4} \psi_{234}(x_2, x_3, x_4) \right) \right)$$

Steps 1.3 + 2 Compute Φ'' by Marginalising the Bound Variables

$$\Phi''(x_1, x_3) = \sum_{x_2} \left(\psi_{123}(x_1, x_2, x_3) \cdot \left(\sum_{x_4} \psi_{234}(x_2, x_3, x_4) \right) \right)$$

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Steps 1.3 + 2 Compute Φ'' by Marginalising the Bound Variables

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Steps 1.3 + 2 Compute Φ'' by Marginalising the Bound Variables

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General Case

Query Rewriting: Indicator Projections

Example

Factor ψ_S			
1	2	3	ψ_S
1	3	3	5
2	3	3	2
3	1	2	4

$$T = \{2, 3, 4\}$$

Indicator projection of

ψ_S onto T

2	3	$\psi_{S/T}$
3	3	1
1	2	1

Query Rewriting: Indicator Projections

Example

Factor ψ_S			
1	2	3	ψ_S
1	3	3	5
2	3	3	2
3	1	2	4

$$T = \{2, 3, 4\}$$

Indicator projection of

ψ_S onto T		
2	3	$\psi_{S/T}$
3	3	1
1	2	1

Let ψ_S be a factor and T a set with $S \cap T \neq \emptyset$

$$\psi_{S/T}(\mathbf{x}_{S \cap T}) = \begin{cases} \mathbf{1}, & \text{if } \exists \mathbf{x}_{S-T} \text{ s.t. } \psi_S(\mathbf{x}_{S-T}, \mathbf{x}_{S \cap T}) \neq \mathbf{0} \\ \mathbf{0}, & \text{otherwise} \end{cases}$$

is the **indicator projection** of ψ_S onto T

Which Copies of a Factor Become Indicator Projections?

Factor ψ_S can be used for computing several bags in the decomposition \mathcal{T}

- Consider each bag B_i in \mathcal{T} with variables T such that $S \cap T \neq \emptyset$
- If B_i is the highest bag of \mathcal{T} such that $S \subseteq T$, then include the actual factor ψ_S into the computation of the bag factor ψ_{B_i}
- Otherwise, include the indicator projection ψ_{S/B_i} into the computation of the bag factor ψ_{B_i}

The above strategy is just a convention, any other convention is OK as long as the factor ψ_S is only used once for computing the bags in \mathcal{T}

Query Rewriting: Marginalisation Orders Compatible with Decompositions

Consider a hypertree decomposition \mathcal{T}

- A bag B of \mathcal{T} **owns** a variable i if B is the **highest** bag in \mathcal{T} containing i

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- For $i \in [m]$, let σ_i be the list of bound variables owned by B_i in any order
- $\sigma = \sigma_1 \dots \sigma_m$ is a **marginalisation order** compatible with \mathcal{T}

Query Rewriting: Eagerly Moving Sums Past Products

Input:

- FAQ-SS $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}} \cdots \bigoplus_{x_{n-1}} \bigoplus_{x_n} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

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We rewrite Φ by repeatedly moving the sums past products

- Move \bigoplus_{x_n} : Let $\partial(n) = \{S \in \mathcal{E} \mid n \in S\}$ and $U = \bigcup_{S \in \partial(n)} (S \setminus \{n\})$

$$\begin{aligned}\Phi(\mathbf{x}_{[f]}) &= \bigoplus_{x_{f+1}} \cdots \bigoplus_{x_{n-1}} \bigotimes_{S \in \mathcal{E} \setminus \partial(n)} \psi_S(\mathbf{x}_S) \otimes \underbrace{\bigoplus_{x_n} \bigotimes_{S \in \partial(n)} \psi_S(\mathbf{x}_S)}_{\psi_U} \\ &= \bigoplus_{x_{f+1}} \cdots \bigoplus_{x_{n-1}} \bigotimes_{S \in \mathcal{E} \setminus \partial(n)} \psi_S(\mathbf{x}_S) \otimes \psi_U(\mathbf{x}_U)\end{aligned}$$

- We now have a new factor $\psi_U(\mathbf{x}_U)$ and a new hypergraph for Φ
- We next move $\bigoplus_{x_{n-1}}$ and so on until \bigoplus_{x_f}

Query Evaluation: Computing the FAQ-SS

Consider an FAQ-SS Φ where sums are moved past products

- Use Leapfrog Triejoin to compute each of the new factors ψ_U

We are left with an α -acyclic FAQ $\Phi(\mathbf{x}_{[f]}) = \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$ where all bound variables are marginalised

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- Use Yannakakis' algorithm to compute the join of all factors ψ_S in Φ

Further FAQ-SS Solver Example

Example 1: The Grid (1/2)

$$\Phi() = \sum_{x_1, x_2, x_3, x_4, x_5, x_6} \psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi_{45}(x_4, x_5) \cdot \psi_{56}(x_5, x_6) \cdot \psi_{16}(x_1, x_6) \cdot \psi_{25}(x_2, x_5)$$

Example 1: The Grid (1/2)

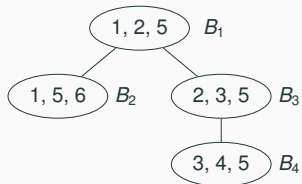
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$$\psi_{16}(x_1, x_6) \cdot \psi_{25}(x_2, x_5)$$

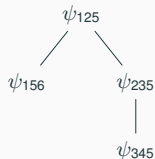
Hypergraph \mathcal{H}



Hypertree decomposition \mathcal{T} for \mathcal{H}



factors materialising
the bags of \mathcal{T}



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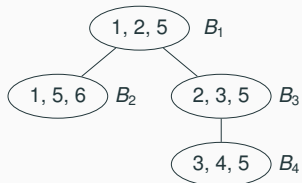
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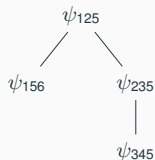
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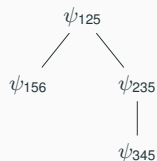


- Factor for B_1 : $\psi_{125}(x_1, x_2, x_5) =$
 $\psi_{12}(x_1, x_2) \cdot \psi_{23/125}(x_2) \cdot \psi_{45/125}(x_5) \cdot \psi_{56/125}(x_5) \cdot \psi_{16/125}(x_1) \cdot \psi_{25}(x_2, x_5)$
- Factor for B_2 : $\psi_{156}(x_1, x_5, x_6) =$
 $\psi_{12/156}(x_1) \cdot \psi_{45/156}(x_5) \cdot \psi_{56}(x_5, x_6) \cdot \psi_{16}(x_1, x_6) \cdot \psi_{25/156}(x_5)$
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Example 1: The Grid (2/2)

$$\Phi'() = \sum_{x_1, x_2, x_5, x_6, x_3, x_4} \psi_{125}(x_1, x_2, x_5) \cdot \psi_{156}(x_1, x_5, x_6) \cdot \psi_{235}(x_2, x_3, x_5) \cdot \psi_{345}(x_3, x_4, x_5)$$

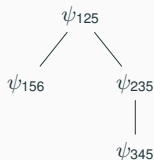
Marginalisation order: (1,2,5,6,3,4)



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FAQs over Multiple Semirings

Differences to FAQ-SS

Consider an FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$

- (1) The operators $\bigoplus^{(f+1)}, \dots, \bigoplus^{(n)}$ can be from **different semirings**
 - Application: e.g. Count SAT for quantified formulas (Exercise Sheet 1)
- (2) We can have $\bigoplus^{(k)} = \bigotimes$ for some $k \in \{f+1, \dots, n\}$
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Product aggregates

- Marginalising X_k with $\bigoplus^{(k)} = \bigotimes$ can be easier than in case $\bigoplus^{(k)} \neq \bigotimes$

Marginalisation Orders

Swapping Aggregate Operators

Consider an FAQ

$$\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

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- A marginalisation order (i_1, \dots, i_n) is **valid** for Φ if

$$\Phi'(\mathbf{x}_{[f]}) = \bigoplus_{x_{i_1}}^{(i_1)} \cdots \bigoplus_{x_{i_n}}^{(i_n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

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Not all marginalisation orders are valid for an FAQ

Example: Swapping Aggregate Operators not Always Correct

Consider the factor ψ_{12} and the two FAQs Φ_1 and Φ_2

X_1	X_2	ψ_{12}
1	2	2
1	3	4
2	2	3

$$\Phi_1() = \sum_{x_1} \max_{x_2} \psi_{12}(x_1, x_2)$$

$$\Phi_2() = \max_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2)$$

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result of Φ_1

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X_2	$\sum_{x_1} \psi_{12}$
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3	4

result of Φ_2

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$()$	5

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3	4

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Φ_1 and Φ_2 are not equivalent

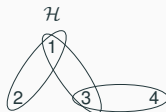
Example: Marginalisation Orders Affect Evaluation Time

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$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

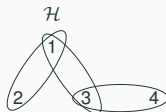


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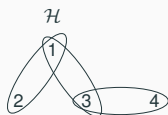
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- The marginalisation orders $\sigma_1 = (3, 2, 1)$ and $\sigma_2 = (3, 1, 2)$ are valid for Φ
- We will see that σ_1 and σ_2 lead to different evaluation times

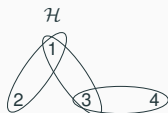
Example: A Bad Marginalisation Order



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Marginalisation order: $\sigma_1 = (3, 2, 1)$

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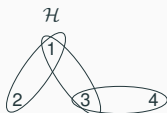


$\partial(1) = \{\{1, 2\} \{1, 3\}\}$ The hyperedges containing 1

$U = \{1, 2, 3\}$ Set of nodes occurring in the hyperedges containing 1

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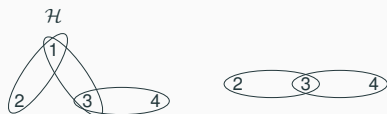
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$$= \max_{x_3} \sum_{x_2} \psi_{34}(x_3, x_4) \cdot \underbrace{\left(\sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \right)}_{\psi'_{23}} \quad \rho_{\mathcal{H}}^*(\{1, 2, 3\}) = 2$$

$\rho_{\mathcal{H}}^*(\{1, 2, 3\})$: ρ^* for the subgraph of \mathcal{H} induced by the nodes $\{1, 2, 3\}$

Example: A Bad Marginalisation Order



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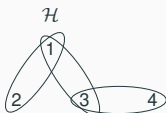
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compute time

$$= \max_{x_3} \sum_{x_2} \psi_{34}(x_3, x_4) \cdot \psi'_{23}(x_2, x_3)$$

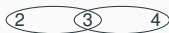
$\mathcal{O}(N^2)$

Example: A Bad Marginalisation Order



$$\partial(1) = \{\{1, 2\}, \{1, 3\}\}$$

$$U = \{1, 2, 3\}$$



$$\partial(2) = \{\{2, 3\}\}$$

$$U = \{2, 3\}$$

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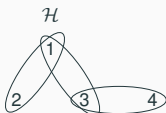
$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \underbrace{\left(\sum_{x_2} \psi'_{23}(x_2, x_3) \right)}_{\psi'_3}$$

compute time

$$\mathcal{O}(N^2)$$

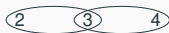
$$\rho_{\mathcal{H}}^*(\{2, 3\}) = 2$$

Example: A Bad Marginalisation Order



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$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi_{34}(x_3, x_4) \cdot \psi'_{23}(x_2, x_3)$$

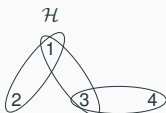
$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

compute time

$$\mathcal{O}(N^2)$$

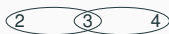
$$\mathcal{O}(N^2)$$

Example: A Bad Marginalisation Order



$$\partial(1) = \{\{1, 2\}, \{1, 3\}\}$$

$$U = \{1, 2, 3\}$$



$$\partial(2) = \{\{2, 3\}\}$$

$$U = \{2, 3\}$$



$$\partial(3) = \{\{3\}, \{3, 4\}\}$$

$$U = \{3, 4\}$$

$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi_{34}(x_3, x_4) \cdot \psi'_{23}(x_2, x_3)$$

$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

$$= \underbrace{\left(\max_{x_3} \psi_{34}(x_4, x_3) \cdot \psi'_3(x_3) \right)}_{\psi'_4}$$

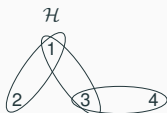
compute time

$$\mathcal{O}(N^2)$$

$$\mathcal{O}(N^2)$$

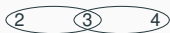
$$\rho_{\mathcal{H}}^*(\{3, 4\}) = 1$$

Example: A Bad Marginalisation Order



$$\partial_4(1) = \{\{1, 2\}, \{1, 3\}\}$$

$$U_4 = \{1, 2, 3\}$$



$$\partial(2) = \{\{2, 3\}\}$$

$$U = \{2, 3\}$$



$$\partial(3) = \{\{3\}, \{3, 4\}\}$$

$$U = \{3, 4\}$$



$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi_{34}(x_3, x_4) \cdot \psi'_{23}(x_2, x_3)$$

$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

$$= \psi'_4(x_4)$$

compute time

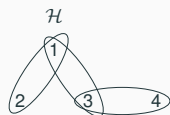
$$\mathcal{O}(N^2)$$

$$\mathcal{O}(N^2)$$

$$\mathcal{O}(N)$$

overall $\mathcal{O}(N^2)$

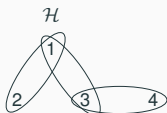
Example: A Better Marginalisation Order



$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

Marginalisation order: $\sigma_1 = (3, 1, 2)$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

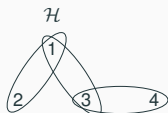
$$U = \{1, 2\}$$

$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \underbrace{\left(\sum_{x_2} \psi_{12}(x_1, x_2) \right)}_{\psi'_1}$$

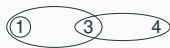
$$\rho_{\mathcal{H}}^*(\{1, 2\}) = 1$$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

$$U = \{1, 2\}$$



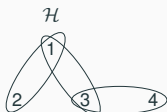
$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

compute time

$$= \max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi'_1(x_1)$$

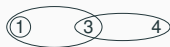
$\mathcal{O}(N)$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

$$U = \{1, 2\}$$



$$\partial(1) = \{\{1\}, \{1, 3\}\}$$

$$U = \{1, 3\}$$

$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi'_1(x_1)$$

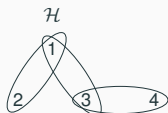
$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \underbrace{\left(\sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi'_1(x_1) \right)}_{\psi'_3}$$

compute time

$\mathcal{O}(N)$

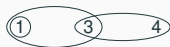
$$\rho_{\mathcal{H}}^*(\{1, 3\}) = 1$$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

$$U = \{1, 2\}$$



$$\partial(1) = \{\{1\}, \{1, 3\}\}$$

$$U = \{1, 3\}$$



$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

compute time

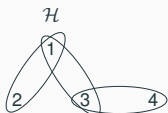
$$= \max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi'_1(x_1)$$

$\mathcal{O}(N)$

$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

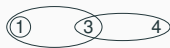
$\mathcal{O}(N)$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

$$U = \{1, 2\}$$



$$\partial(1) = \{\{1\}, \{1, 3\}\}$$

$$U = \{1, 3\}$$



$$\partial(3) = \{\{3\}, \{3, 4\}\}$$

$$U = \{3, 4\}$$

$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

compute time

$$= \max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi'_1(x_1)$$

$\mathcal{O}(N)$

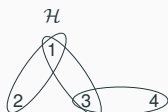
$$= \max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

$\mathcal{O}(N)$

$$= \underbrace{\left(\max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3) \right)}_{\psi'_4}$$

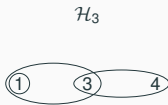
$$\rho_{\mathcal{H}}^*(\{3, 4\}) = 1$$

Example: A Better Marginalisation Order



$$\partial(2) = \{\{1, 2\}\}$$

$$U = \{1, 2\}$$



$$\partial(1) = \{\{1\}, \{1, 3\}\}$$

$$U = \{1, 3\}$$



$$\partial(3) = \{\{3\}, \{3, 4\}\}$$

$$U = \{3, 4\}$$

④

$$\Phi(x_4) = \max_{x_3} \sum_{x_1} \sum_{x_2} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

compute time

$$\max_{x_3} \sum_{x_1} \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi'_1(x_1)$$

$\mathcal{O}(N)$

$$\max_{x_3} \psi_{34}(x_3, x_4) \cdot \psi'_3(x_3)$$

$\mathcal{O}(N)$

$$= \psi'_4(x_4)$$

$\mathcal{O}(N)$

overall $\mathcal{O}(N)$

Finding the Best Marginalisation Order

- Different marginalisation orders can lead to different evaluation times
- To find the **optimal marginalisation order** we need to go over the entire space of valid marginalisation orders
- In the general case, the existing algorithms are not better than enumerating over all possible orders
- Focus of this lecture: What is the evaluation time for a given FAQ and marginalisation order?

Product Aggregates

Example: Moving Product Marginalisation Past Factor Products 1/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
1	2	2	2	2	4	2	3
1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Example: Moving Product Marginalisation Past Factor Products 1/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Assume $\text{Dom}(X_2) = \{2, 3\}$

$$\Phi(1, 2) = \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right)$$

Example: Moving Product Marginalisation Past Factor Products 1/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
1	2	2	2	2	4	2	3
1	3	3	3	2	5	5	4

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Assume $\text{Dom}(X_2) = \{2, 3\}$

$$\begin{aligned} \Phi(1, 2) &= \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right) \\ &= (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 1/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

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$$\begin{aligned} \Phi(1, 2) &= \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right) \\ &= (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \\ &= (2 \cdot 3) \cdot (4 \cdot 5) \cdot (3 \cdot 3) \end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 1/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Assume $\text{Dom}(X_2) = \{2, 3\}$

$$\begin{aligned} \Phi(1, 2) &= \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right) \\ &= (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \\ &= (2 \cdot 3) \cdot (4 \cdot 5) \cdot (3 \cdot 3) \\ &= \prod_{x_2} \psi_{12}(1, x_2) \cdot \prod_{x_2} \psi_{23}(x_2, 2) \cdot \psi_3(2)^{|\text{Dom}(X_2)|} \end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 2/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
1	2	2	2	2	4	2	3
1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Assume now $\text{Dom}(X_2) = \{1, 2, 3\}$

$$\Phi(1, 2) = \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right)$$

Example: Moving Product Marginalisation Past Factor Products 2/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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$$\Phi(x_1, x_3) = \prod_{x_2} (\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3))$$

Assume now $\text{Dom}(X_2) = \{1, 2, 3\}$

$$\begin{aligned}\Phi(1, 2) &= \prod_{x_2} (\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2)) \\ &= 0 \cdot (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3)\end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 2/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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1	3	3	3	2	5	5	4

$$\Phi(x_1, x_3) = \prod_{x_2} (\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3))$$

Assume now $\text{Dom}(X_2) = \{1, 2, 3\}$

$$\begin{aligned}\Phi(1, 2) &= \prod_{x_2} (\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2)) \\ &= 0 \cdot (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \\ &= 0\end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 2/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Assume now $\text{Dom}(X_2) = \{1, 2, 3\}$

$$\begin{aligned} \Phi(1, 2) &= \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right) \\ &= 0 \cdot (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \\ &= 0 \\ &= \prod_{x_2} \psi_{12}(1, x_2) \cdot \prod_{x_2} \psi_{23}(x_2, 2) \cdot \psi_3(2)^{|\text{Dom}(X_2)|} \end{aligned}$$

Example: Moving Product Marginalisation Past Factor Products 2/2

Consider input factors ψ_{12} , ψ_{23} , and ψ_3 and FAQ Φ

X_1	X_2	ψ_{12}	X_2	X_3	ψ_{23}	X_3	ψ_3
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$$\Phi(x_1, x_3) = \prod_{x_2} \left(\psi_{12}(x_1, x_2) \cdot \psi_{23}(x_2, x_3) \cdot \psi_3(x_3) \right)$$

Assume now $\text{Dom}(X_2) = \{1, 2, 3\}$

$$\begin{aligned} \Phi(1, 2) &= \prod_{x_2} \left(\psi_{12}(1, x_2) \cdot \psi_{23}(x_2, 2) \cdot \psi_3(2) \right) \\ &= 0 \cdot (2 \cdot 4 \cdot 3) \cdot (3 \cdot 5 \cdot 3) \\ &= 0 \\ &= \prod_{x_2} \psi_{12}(1, x_2) \cdot \prod_{x_2} \psi_{23}(x_2, 2) \cdot \psi_3(2)^{|\text{Dom}(X_2)|} \end{aligned}$$

- We will use this rewriting strategy to deal with product aggregates

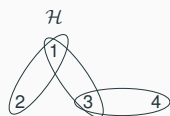
Product Aggregates

$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \sum_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$



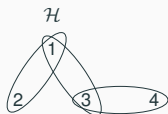
- Consider again the marginalisation order (3, 2, 1) for Φ_4
- We have seen: The marginalisation of X_1 requires $\mathcal{O}(N^2)$ compute time
- Assume now: \sum_{x_1} is replaced by \prod_{x_1}
- We will see: In this case, X_1 can be marginalised in $\mathcal{O}(N)$ time (with an additional logarithmic factor)

Example: Product Aggregates



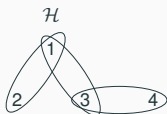
$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

Example: Product Aggregates



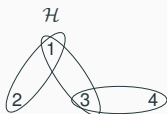
$$\begin{aligned}\Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \left(\prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \right)\end{aligned}$$

Example: Product Aggregates



$$\begin{aligned}\Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \left(\prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \right) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \left(\underbrace{\prod_{x_1} \psi_{12}(x_1, x_2)}_{\psi'_2} \cdot \underbrace{\prod_{x_1} \psi_{13}(x_1, x_3)}_{\psi'_3} \right)\end{aligned}$$

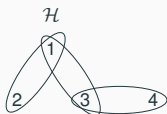
Example: Product Aggregates



$$\begin{aligned}\Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \left(\prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \right) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \left(\underbrace{\prod_{x_1} \psi_{12}(x_1, x_2)}_{\psi'_2} \cdot \underbrace{\prod_{x_1} \psi_{13}(x_1, x_3)}_{\psi'_3} \right)\end{aligned}$$

- Compute time for ψ'_2 (same for ψ'_3): $\mathcal{O}(N)$ since we need to iterate over a single factor of size N

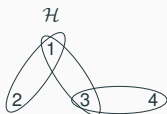
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- $\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}$ is $\psi_{34}(x_3, x_4)$ to the power of $|\text{Dom}(X_1)|$

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- $\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}$ is $\psi_{34}(x_3, x_4)$ to the power of $|\text{Dom}(X_1)|$
- Next we look closer at the computation of $\left(\psi_{34}(x_3, x_4) \right)^{|\text{Dom}(X_1)|}$

Powering Factors

Consider a factor ψ_S

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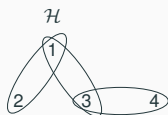
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- ψ'_S can be computed in $\mathcal{O}(|\psi_S| \cdot \log_2 n)$ time by repeated squaring
- If the range of ψ_S is idempotent, i.e. contains only 0 and 1, then $\psi'_S = \psi_S$
 \implies Compute time is $\mathcal{O}(1)$

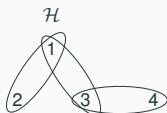
Example: Product Aggregates



$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

Marginalisation order: $\sigma = (3, 2, 1)$

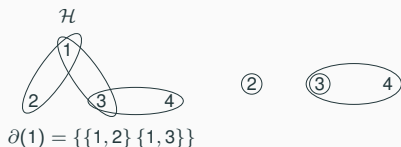
Example: Product Aggregates



$$\partial(1) = \{\{1, 2\} \{1, 3\}\}$$

$$\begin{aligned}\Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \\ &= \max_{x_3} \sum_{x_2} \underbrace{\left(\psi_{34}(x_3, x_4)\right)^{|\text{Dom}(X_1)|}}_{\psi'_{34}} \cdot \underbrace{\left(\prod_{x_1} \psi_{12}(x_1, x_2)\right)}_{\psi'_2} \cdot \underbrace{\left(\prod_{x_1} \psi_{13}(x_1, x_3)\right)}_{\psi'_3}\end{aligned}$$

Example: Product Aggregates

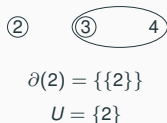
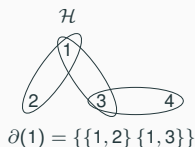


$$\begin{aligned}\Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) \\ &= \max_{x_3} \sum_{x_2} \psi'_{34}(x_3, x_4) \cdot \psi'_2(x_2) \cdot \psi'_3(x_3)\end{aligned}$$

compute time

$\mathcal{O}(N)$

Example: Product Aggregates



$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi'_{34}(x_3, x_4) \cdot \psi'_2(x_2) \cdot \psi'_3(x_3)$$

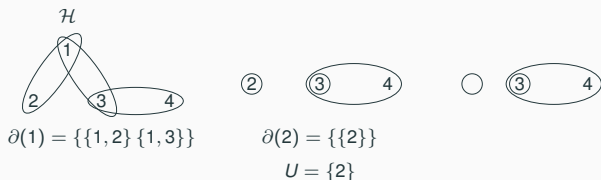
$$= \max_{x_3} \psi'_{34}(x_3, x_4) \cdot \psi'_3(x_3) \cdot \underbrace{\left(\sum_{x_2} \psi'_2(x_2) \right)}_{\psi'}$$

compute time

$\mathcal{O}(N)$

$$\rho_{\mathcal{H}}^*(\{2\}) = 1$$

Example: Product Aggregates



$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi'_{34}(x_3, x_4) \cdot \psi'_2(x_2) \cdot \psi'_3(x_3)$$

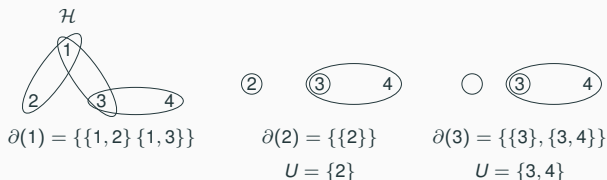
$$= \max_{x_3} \psi'_{34}(x_3, x_4) \cdot \psi'_3(x_3) \cdot \psi'()$$

compute time

$\mathcal{O}(N)$

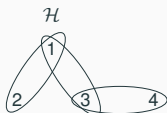
$\mathcal{O}(N)$

Example: Product Aggregates



$$\begin{aligned}
 \Phi(x_4) &= \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4) && \text{compute time} \\
 &= \max_{x_3} \sum_{x_2} \psi'_{34}(x_3, x_4) \cdot \psi'_2(x_2) \cdot \psi'_3(x_3) && \mathcal{O}(N) \\
 &= \max_{x_3} \psi'_{34}(x_3, x_4) \cdot \psi'_3(x_3) \cdot \psi'() && \mathcal{O}(N) \\
 &= \psi'() \cdot \underbrace{\left(\max_{x_3} \psi'_{34}(x_3, x_4) \cdot \psi'_3(x_3) \right)}_{\psi'_4} && \rho_{\mathcal{H}}^*({3,4}) = 1
 \end{aligned}$$

Example: Product Aggregates



$$\partial_4(1) = \{\{1, 2\}, \{1, 3\}\}$$

②



$$\partial(2) = \{\{2\}\}$$

$$U = \{2\}$$

○



$$\partial(3) = \{\{3\}, \{3, 4\}\}$$

$$U = \{3, 4\}$$

○

④

$$\Phi(x_4) = \max_{x_3} \sum_{x_2} \prod_{x_1} \psi_{12}(x_1, x_2) \cdot \psi_{13}(x_1, x_3) \cdot \psi_{34}(x_3, x_4)$$

$$= \max_{x_3} \sum_{x_2} \psi'_{34}(x_3, x_4) \cdot \psi'_2(x_2) \cdot \psi'_3(x_3)$$

$$= \max_{x_3} \psi'_{34}(x_3, x_4) \cdot \psi'_3(x_3) \cdot \psi'()$$

$$= \psi'() \cdot \psi'_4(x_4)$$

compute time

$\mathcal{O}(N)$

$\mathcal{O}(N)$

$\mathcal{O}(N)$

overall $\mathcal{O}(N)$

InsideOut Algorithm

Main Steps of the InsideOut Algorithm

Input: FAQ Φ and marginalisation order σ for bound variables

Step 1: Marginalisation

- Marginalise the bound variables following σ
 \implies results in an FAQ with hypergraph \mathcal{H}'

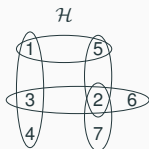
Step 2: Evaluation

- Construct a hypertree decomposition \mathcal{T} for \mathcal{H}'
- Turn Φ into an α -acyclic FAQ Φ' by materialising the bags of \mathcal{T}
- Run Yannakakis' algorithm on Φ'

We next explain the above steps using an example

InsideOut Algorithm Example

Example: InsideOut Algorithm 1/3

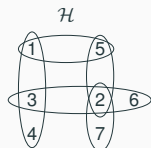


$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

(for brevity, input variables of factors are skipped)

Marginalisation order: $\sigma = (3, 4, 5, 6)$

Example: InsideOut Algorithm 1/3



$$\partial(6) = \{\{2, 3, 6\}\}$$

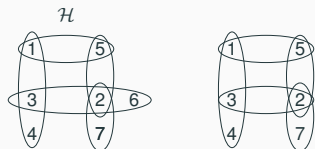
$$U = \{2, 3, 6\}$$

$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \underbrace{\left(\max_{x_6} \underbrace{\psi_{25/236} \cdot \psi_{134/236} \cdot \psi_{27/236}}_{\text{indicator projections}} \cdot \psi_{236} \right)}_{\psi'_{23}}$$

$$\rho_{\mathcal{H}}^*({2, 3, 6}) = 1$$

Example: InsideOut Algorithm 1/3



$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$

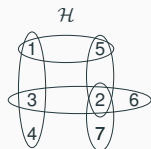
$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

compute time

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$

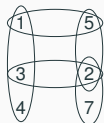
$\mathcal{O}(N)$

Example: InsideOut Algorithm 1/3



$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

$$U = \{1, 2, 5\}$$

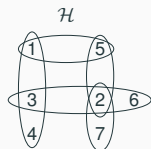
$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27} \quad \text{compute time}$$

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \quad \mathcal{O}(N)$$

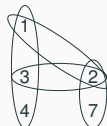
$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \underbrace{\left(\max_{x_5} \psi_{134/125} \cdot \psi'_{23/125} \cdot \psi'_{27/125} \cdot \psi_{15} \cdot \psi_{25} \right)}_{\psi'_{12}}$$

$$\rho_{\mathcal{H}}^*(\{1, 2, 5\}) = 2$$

Example: InsideOut Algorithm 1/3



$$\partial(6) = \{\{2, 3, 6\}\} \quad \partial(5) = \{\{1, 5\}, \{2, 5\}\}$$
$$U = \{2, 3, 6\} \quad U = \{1, 2, 5\}$$



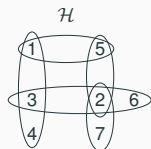
$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$
$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$
$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12}$$

compute time

$\mathcal{O}(N)$

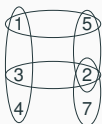
$\mathcal{O}(N^2)$

Example: InsideOut Algorithm 1/3



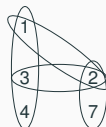
$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

$$U = \{1, 2, 5\}$$



$$\partial(4) = \{\{1, 3, 4\}\}$$

$$U = \{1, 3, 4\}$$

$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

compute time

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$

$\mathcal{O}(N)$

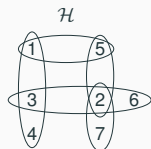
$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12}$$

$\mathcal{O}(N^2)$

$$= \prod_{x_3} \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \cdot \underbrace{\left(\sum_{x_4} \psi'_{23/134} \cdot \psi'_{12/134} \cdot \psi_{134} \right)}_{\psi'_{13}}$$

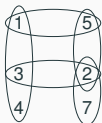
$$\rho_{\mathcal{H}}^*({1, 3, 4}) = 1$$

Example: InsideOut Algorithm 1/3



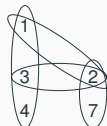
$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



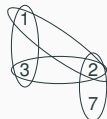
$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

$$U = \{1, 2, 5\}$$



$$\partial(4) = \{\{1, 3, 4\}\}$$

$$U = \{1, 3, 4\}$$



$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

compute time

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$

$\mathcal{O}(N)$

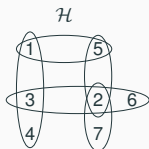
$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12}$$

$\mathcal{O}(N^2)$

$$= \prod_{x_3} \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \cdot \psi'_{13}$$

$\mathcal{O}(N)$

Example: InsideOut Algorithm 1/3



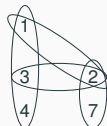
$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



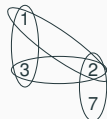
$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

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$$\partial(4) = \{\{1, 3, 4\}\}$$

$$U = \{1, 3, 4\}$$



$$\partial(3) = \{\{1, 3\}, \{2, 3\}\}$$

$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27} \quad \text{compute time}$$

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \quad \mathcal{O}(N)$$

$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \quad \mathcal{O}(N^2)$$

$$= \prod_{x_3} \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \cdot \psi'_{13} \quad \mathcal{O}(N)$$

$$= \underbrace{\left(\psi'_{27}\right)^{|\text{Dom}(X_3)|}}_{\psi'_{27}} \cdot \underbrace{\left(\psi'_{12}\right)^{|\text{Dom}(X_3)|}}_{\psi'_{12}} \cdot \underbrace{\left(\prod_{x_3} \psi'_{23}\right)}_{\psi'_2} \cdot \underbrace{\left(\prod_{x_3} \psi'_{13}\right)}_{\psi'_1}$$

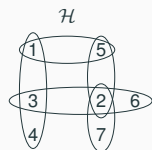
$$\rho_{\mathcal{H}}^*(\{2, 7\}) = \rho_{\mathcal{H}}^*(\{2, 3\}) = \rho_{\mathcal{H}}^*(\{1, 3\}) = 1$$

sizes of ψ_{27} , ψ'_{23} , and ψ'_{13} : $\mathcal{O}(N)$
 compute times for ψ'_{27} , ψ'_2 , and ψ'_1 : $\mathcal{O}(N)$

$$\rho_{\mathcal{H}}^*(\{1, 2\}) = 2$$

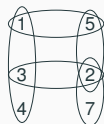
size of ψ'_{12} : $\mathcal{O}(N^2)$
 compute time for ψ'_{12} : $\mathcal{O}(N^2)$

Example: InsideOut Algorithm 1/3



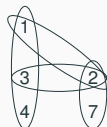
$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



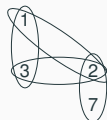
$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

$$U = \{1, 2, 5\}$$



$$\partial(4) = \{\{1, 3, 4\}\}$$

$$U = \{1, 3, 4\}$$



$$\partial(3) = \{\{1, 3\}, \{2, 3\}\}$$



$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

compute time

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$

$\mathcal{O}(N)$

$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12}$$

$\mathcal{O}(N^2)$

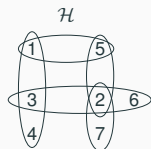
$$= \prod_{x_3} \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \cdot \psi'_{13}$$

$\mathcal{O}(N)$

$$= \psi'_{27} \cdot \psi''_{12} \cdot \psi'_2 \cdot \psi'_1$$

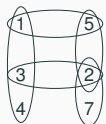
$\mathcal{O}(N^2)$

Example: InsideOut Algorithm 1/3



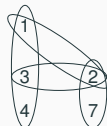
$$\partial(6) = \{\{2, 3, 6\}\}$$

$$U = \{2, 3, 6\}$$



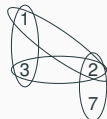
$$\partial(5) = \{\{1, 5\}, \{2, 5\}\}$$

$$U = \{1, 2, 5\}$$



$$\partial(4) = \{\{1, 3, 4\}\}$$

$$U = \{1, 3, 4\}$$



$$\partial(3) = \{\{1, 3\}, \{2, 3\}\}$$



$$\Phi(x_1, x_2, x_7) = \prod_{x_3} \sum_{x_4} \max_{x_5} \max_{x_6} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{236} \cdot \psi_{27}$$

compute time

$$= \prod_{x_3} \sum_{x_4} \max_{x_5} \psi_{15} \cdot \psi_{25} \cdot \psi_{134} \cdot \psi_{27} \cdot \psi'_{23}$$

$\mathcal{O}(N)$

$$= \prod_{x_3} \sum_{x_4} \psi_{134} \cdot \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12}$$

$\mathcal{O}(N^2)$

$$= \prod_{x_3} \psi_{27} \cdot \psi'_{23} \cdot \psi'_{12} \cdot \psi'_{13}$$

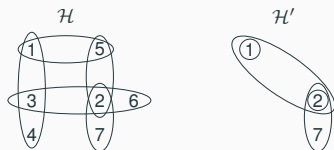
$\mathcal{O}(N)$

$$= \psi'_{27} \cdot \psi''_{12} \cdot \psi'_2 \cdot \psi'_1$$

$\mathcal{O}(N^2)$

- We are left with an FAQ with hypergraph \mathcal{H}' and all variables free

Example: InsideOut Algorithm 2/3

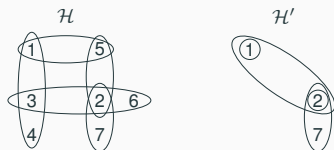


$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

After having marginalised all bound variables:

- Construct a **hypertree decomposition** \mathcal{T} for \mathcal{H}'
- **Materialise** the bags of \mathcal{T} using **Leapfrog Triejoin** (as in case of FAQ-SS)
- Compute the result of Φ using **Yannakakis' algorithm**

Example: InsideOut Algorithm 2/3



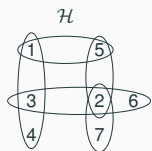
$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

After having marginalised all bound variables:

- Construct a **hypertree decomposition** \mathcal{T} for \mathcal{H}'
- **Materialise** the bags of \mathcal{T} using **Leapfrog Triejoin** (as in case of FAQ-SS)
- Compute the result of Φ using **Yannakakis' algorithm**

Next, we go through the above three steps.

Example: InsideOut Algorithm 3/3

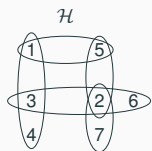


Hypertree decomposition \mathcal{T} for \mathcal{H}'



$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

Example: InsideOut Algorithm 3/3



Hypertree decomposition \mathcal{T} for \mathcal{H}'

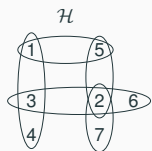


$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

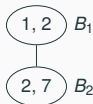
Materialisation of the bags of \mathcal{T}

- $\psi_{B_1}(x_1, x_2) = \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1) \cdot \psi'_{27/12}(x_2)$
 $\rho_{\mathcal{H}}^*(\{1, 2\}) = 2 \Rightarrow \mathcal{O}(N^2)$ compute time

Example: InsideOut Algorithm 3/3



Hypertree decomposition \mathcal{T} for \mathcal{H}'

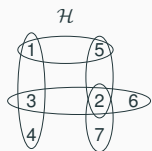


$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

Materialisation of the bags of \mathcal{T}

- $\psi_{B_1}(x_1, x_2) = \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1) \cdot \psi'_{27/12}(x_2)$
 $\rho_{\mathcal{H}}^*(\{1, 2\}) = 2 \Rightarrow \mathcal{O}(N^2)$ compute time
- $\psi_{B_2}(x_2, x_7) = \psi'_{27} \cdot \psi''_{12/27}(x_2) \cdot \psi'_{2/27}(x_2)$
 $\rho_{\mathcal{H}}^*(\{2, 7\}) = 1 \Rightarrow \mathcal{O}(N)$ compute time

Example: InsideOut Algorithm 3/3



Hypertree decomposition \mathcal{T} for \mathcal{H}'



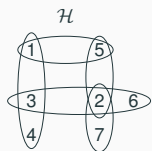
$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

Materialisation of the bags of \mathcal{T}

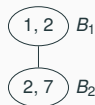
- $\psi_{B_1}(x_1, x_2) = \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1) \cdot \psi'_{27/12}(x_2)$
 $\rho_{\mathcal{H}}^*(\{1, 2\}) = 2 \Rightarrow \mathcal{O}(N^2)$ compute time
- $\psi_{B_2}(x_2, x_7) = \psi'_{27} \cdot \psi''_{12/27}(x_2) \cdot \psi'_{2/27}(x_2)$
 $\rho_{\mathcal{H}}^*(\{2, 7\}) = 1 \Rightarrow \mathcal{O}(N)$ compute time

$$\Phi'(x_1, x_2, x_7) = \psi_{B_1}(x_1, x_2) \cdot \psi_{B_2}(x_2, x_7)$$

Example: InsideOut Algorithm 3/3



Hypertree decomposition \mathcal{T} for \mathcal{H}'



$$\Phi(x_1, x_2, x_7) = \psi'_{27}(x_2, x_7) \cdot \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1)$$

Materialisation of the bags of \mathcal{T}

- $\psi_{B_1}(x_1, x_2) = \psi''_{12}(x_1, x_2) \cdot \psi'_2(x_2) \cdot \psi'_1(x_1) \cdot \psi'_{27/12}(x_2)$
 $\rho_{\mathcal{H}}^*(\{1, 2\}) = 2 \Rightarrow \mathcal{O}(N^2)$ compute time
- $\psi_{B_2}(x_2, x_7) = \psi'_{27} \cdot \psi''_{12/27}(x_2) \cdot \psi'_{2/27}(x_2)$
 $\rho_{\mathcal{H}}^*(\{2, 7\}) = 1 \Rightarrow \mathcal{O}(N)$ compute time

$$\Phi'(x_1, x_2, x_7) = \psi_{B_1}(x_1, x_2) \cdot \psi_{B_2}(x_2, x_7)$$

- Φ' is equivalent to Φ
- Compute the result of Φ' using **Yannakakis' algorithm**

InsideOut Algorithm in General

InsideOut Algorithm

Input:

- FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

InsideOut Algorithm

Input:

- FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

$$\Phi_n = \Phi$$

$$\mathcal{H}_n = \mathcal{H}$$

InsideOut Algorithm

Input:

- FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

$$\Phi_n = \Phi$$

$$\mathcal{H}_n = \mathcal{H}$$

for $k = n$ down to $f + 1$ do

if $\bigoplus^{(k)} \neq \bigotimes$

$$(\Phi_{k-1}, \mathcal{H}_{k-1}) = \text{SemiringMarginalisation}(\Phi_k, \mathcal{H}_k)$$

InsideOut Algorithm

Input:

- FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{s \in \mathcal{E}} \psi_s(\mathbf{x}_s)$
- Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

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for $k = n$ down to $f + 1$ do

if $\bigoplus^{(k)} \neq \bigotimes$

$$(\Phi_{k-1}, \mathcal{H}_{k-1}) = \text{SemiringMarginalisation}(\Phi_k, \mathcal{H}_k)$$

else

$$(\Phi_{k-1}, \mathcal{H}_{k-1}) = \text{ProductMarginalisation}(\Phi_k, \mathcal{H}_k)$$

InsideOut Algorithm

Input:

- FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$
- Marginalisation order $\sigma = (f + 1, \dots, n)$

$$\Phi_n = \Phi$$

$$\mathcal{H}_n = \mathcal{H}$$

for $k = n$ down to $f + 1$ do

if $\bigoplus^{(k)} \neq \bigotimes$

$$(\Phi_{k-1}, \mathcal{H}_{k-1}) = \text{SemiringMarginalisation}(\Phi_k, \mathcal{H}_k)$$

else

$$(\Phi_{k-1}, \mathcal{H}_{k-1}) = \text{ProductMarginalisation}(\Phi_k, \mathcal{H}_k)$$

Construct for \mathcal{H}_f a hypertree decomposition \mathcal{T} with bags $(B_i)_{i \in [m]}$

Compute the bags $(\psi_{B_i})_{i \in [m]}$ of \mathcal{T} using **Leapfrog Triejoin**

Run **Yannakakis' algorithm** on $\Phi'(\mathbf{x}_{[f]}) = \bigotimes_{i \in [m]} \psi_{B_i}(\mathbf{x}_{B_i})$

Procedure for Semiring Marginalisation

Procedure: **SemiringMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

Procedure for Semiring Marginalisation

Procedure: **Semiring Marginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

$$U_k = \bigcup_{S \in \partial_k(k)} S$$

Procedure for Semiring Marginalisation

Procedure: **Semiring Marginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

$$U_k = \bigcup_{S \in \partial_k(k)} S$$

$$\psi_{U_k \setminus \{k\}}(\mathbf{x}_{U_k \setminus \{k\}}) = \bigoplus_{x_k} \left(\bigotimes_{S \in \partial_k(k)} \psi_S(\mathbf{x}_S) \otimes \bigotimes_{\substack{S \in \mathcal{E}_k \setminus \partial_k(k) \\ S \cap U_k \neq \emptyset}} \underbrace{\psi_{S/U_k}(\mathbf{x}_{S \cap U_k})}_{\text{indicator projection}} \right)$$

marginalisation

Procedure for Semiring Marginalisation

Procedure: **Semiring Marginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

$$U_k = \bigcup_{S \in \partial_k(k)} S$$

$$\psi_{U_k \setminus \{k\}}(\mathbf{x}_{U_k \setminus \{k\}}) = \bigoplus_{x_k} \left(\bigotimes_{S \in \partial_k(k)} \psi_S(\mathbf{x}_S) \otimes \bigotimes_{\substack{S \in \mathcal{E}_k \setminus \partial_k(k) \\ S \cap U_k \neq \emptyset}} \underbrace{\psi_{S/U_k}(\mathbf{x}_{S \cap U_k})}_{\text{indicator projection}} \right)$$

marginalisation

$$\mathcal{V}_{k-1} = [k-1]$$

$\mathcal{E}_{k-1} = (\mathcal{E}_k \setminus \partial_k(k)) \cup (U_k \setminus \{k\})$ (skip hyperedges $\partial_k(k)$, add hyperedge $U_k \setminus \{k\}$)

$$\Phi_{k-1}(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_{k-1}}^{(k-1)} \bigotimes_{S \in \mathcal{E}_{k-1}} \psi_S(\mathbf{x}_S) \text{ (uses the new factor } \psi_{U_k \setminus \{k\}})$$

Procedure for Semiring Marginalisation

Procedure: **Semiring Marginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

$$U_k = \bigcup_{S \in \partial_k(k)} S$$

$$\psi_{U_k \setminus \{k\}}(\mathbf{x}_{U_k \setminus \{k\}}) = \bigoplus_{x_k} \left(\bigotimes_{S \in \partial_k(k)} \psi_S(\mathbf{x}_S) \otimes \bigotimes_{\substack{S \in \mathcal{E}_k \setminus \partial_k(k) \\ S \cap U_k \neq \emptyset}} \underbrace{\psi_{S/U_k}(\mathbf{x}_{S \cap U_k})}_{\text{indicator projection}} \right)$$

marginalisation

$$\mathcal{V}_{k-1} = [k-1]$$

$$\mathcal{E}_{k-1} = (\mathcal{E}_k \setminus \partial_k(k)) \cup (U_k \setminus \{k\}) \text{ (skip hyperedges } \partial_k(k), \text{ add hyperedge } U_k \setminus \{k\})$$

$$\Phi_{k-1}(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_{k-1}}^{(k-1)} \bigotimes_{S \in \mathcal{E}_{k-1}} \psi_S(\mathbf{x}_S) \text{ (uses the new factor } \psi_{U_k \setminus \{k\}})$$

$$\text{return } (\Phi_{k-1}, \mathcal{H}_{k-1} = (\mathcal{V}_{k-1}, \mathcal{E}_{k-1}))$$

Procedure for Product Marginalisation

Procedure: **ProductMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

Procedure for Product Marginalisation

Procedure: **ProductMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

Procedure for Product Marginalisation

Procedure: **ProductMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

for each $S \in \partial_k(k)$

marginalisation

$$\psi_{S \setminus \{k\}}(\mathbf{x}_{S \setminus \{k\}}) = \bigotimes_{x_k} \psi_S(\mathbf{x}_S)$$

for each $S \in \mathcal{E}_k \setminus \partial_k(k)$

$$\psi_S(\mathbf{x}_S) = \left(\psi_S(\mathbf{x}_S) \right)^{|\text{Dom}(X_k)|}$$

Procedure for Product Marginalisation

Procedure: **ProductMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

for each $S \in \partial_k(k)$

marginalisation

$$\psi_{S \setminus \{k\}}(\mathbf{x}_{S \setminus \{k\}}) = \bigotimes_{x_k} \psi_S(\mathbf{x}_S)$$

for each $S \in \mathcal{E}_k \setminus \partial_k(k)$

$$\psi_S(\mathbf{x}_S) = \left(\psi_S(\mathbf{x}_S) \right)^{|\text{Dom}(X_k)|}$$

$$\mathcal{V}_{k-1} = [k-1]$$

$$\mathcal{E}_{k-1} = \{S \setminus \{k\} \mid S \in \mathcal{E}_k\}$$

$$\Phi_{k-1}(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_{k-1}}^{(k-1)} \bigotimes_{S \in \mathcal{E}_{k-1}} \psi_S(\mathbf{x}_S) \text{ (uses the new factors } \psi_{S \setminus \{k\}}, \psi_S)$$

Procedure for Product Marginalisation

Procedure: **ProductMarginalisation**

Input:

- FAQ $\Phi_k(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_k}^{(k)} \bigotimes_{S \in \mathcal{E}_k} \psi_S(\mathbf{x}_S)$
- Hypergraph $\mathcal{H}_k = (\mathcal{V}_k, \mathcal{E}_k)$

$$\partial_k(k) = \{S \mid S \in \mathcal{E}_k \text{ with } k \in S\}$$

for each $S \in \partial_k(k)$

marginalisation

$$\psi_{S \setminus \{k\}}(\mathbf{x}_{S \setminus \{k\}}) = \bigotimes_{x_k} \psi_S(\mathbf{x}_S)$$

for each $S \in \mathcal{E}_k \setminus \partial_k(k)$

$$\psi_S(\mathbf{x}_S) = \left(\psi_S(\mathbf{x}_S) \right)^{|\text{Dom}(X_k)|}$$

$$\mathcal{V}_{k-1} = [k-1]$$

$$\mathcal{E}_{k-1} = \{S \setminus \{k\} \mid S \in \mathcal{E}_k\}$$

$$\Phi_{k-1}(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_{k-1}}^{(k-1)} \bigotimes_{S \in \mathcal{E}_{k-1}} \psi_S(\mathbf{x}_S) \text{ (uses the new factors } \psi_{S \setminus \{k\}}, \psi_S)$$

return $(\Phi_{k-1}, \mathcal{H}_{k-1} = (\mathcal{V}_{k-1}, \mathcal{E}_{k-1}))$

Runtime of InsideOut

Consider an FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$ with hypergraph \mathcal{H}

Runtime of InsideOut

Consider an FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$ with hypergraph \mathcal{H}

Let $\mathcal{K} := \{k \mid k > f \text{ and } \bigoplus^{(k)} \neq \bigotimes\}$

Runtime of InsideOut

Consider an FAQ $\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \dots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$ with hypergraph \mathcal{H}

Let $\mathcal{K} := \{k \mid k > f \text{ and } \bigoplus^{(k)} \neq \bigotimes\}$

Given the marginalisation order $\sigma = (f + 1, \dots, n)$, InsideOut constructs

- a sequence $\mathcal{H} = \mathcal{H}_n, \dots, \mathcal{H}_f$ of hypergraphs
- a set U_k for each $k \in \mathcal{K}$
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The InsideOut Algorithm runs in time

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$\text{faqw}(\sigma, \mathcal{T}) = \max_{k \in \mathcal{K}, i \in [m]} \{\rho_{\mathcal{H}}^*(U_k), \rho_{\mathcal{H}}^*(B_i)\}$ is the FAQ-width of (σ, \mathcal{T})

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For optimal σ and \mathcal{T} , InsideOut computes Φ in time

$$\mathcal{O}\left(N^{\text{faqw}(\Phi)} + \text{OUT}\right)$$