

# Conjunctive Queries on Probabilistic Graphs: Combined Complexity

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# Tuple-independent databases (TID)

- **Probabilistic databases:** model **uncertainty** about data
- Simplest model: **tuple-independent databases (TID)**
  - A **relational database**  $I$
  - A **probability valuation**  $\pi$  mapping each fact of  $I$  to  $[0, 1]$
- **Semantics** of a TID  $(I, \pi)$ : a **probability distribution** on  $I' \subseteq I$ :
  - Each fact  $F \in I$  is either **present** or **absent** with probability  $\pi(F)$
  - Assume **independence** across facts

## Example: TID

<b>S</b>		
<i>a</i>	<i>b</i>	<i>.5</i>
<i>a</i>	<i>c</i>	<i>.2</i>

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This TID  $(I, \pi)$  represents the following **probability distribution**:

$.5 \times .2$		$.5 \times (1 - .2)$	
<b>S</b>		<b>S</b>	
<i>a</i>	<i>b</i>	<i>a</i>	<i>b</i>
<i>a</i>	<i>c</i>		

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<b>S</b>	<b>S</b>	<b>S</b>
<i>a</i> <i>b</i>	<i>a</i> <i>b</i>	
<i>a</i> <i>c</i>		<i>a</i> <i>c</i>

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<b>S</b>	<b>S</b>	<b>S</b>	<b>S</b>
<i>a</i> <i>b</i>	<i>a</i> <i>b</i>		
<i>a</i> <i>c</i>		<i>a</i> <i>c</i>	



# Probabilistic query evaluation (PQE)

Let us fix:

- **Relational signature**  $\sigma$
- Class  $\mathcal{I}$  of **relational instances** on  $\sigma$  (e.g., acyclic, treelike)
- Class  $\mathcal{Q}$  of **Boolean queries** (e.g., paths, trees)

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$$\rightarrow \Pr((I, \pi) \models q) = \sum_{J \subseteq I, J \models q} \Pr(J)$$

# Complexity of probabilistic query evaluation (PQE)

**Question:** what is the (data, combined) **complexity** of PQE depending on the class  $\mathcal{Q}$  of **queries** and class  $\mathcal{I}$  of **instances**?

## Data complexity results

- Existing **data dichotomy result** on queries [Dalvi & Suciu, 2012]
  - $\mathcal{Q} = \text{UCQs}$
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What about **combined** complexity?

## Restrict to CQs on graph signatures

$\exists x y z t R(x, y) \wedge S(y, z) \wedge S(t, z)$

<b>R</b>		
<i>a</i>	<i>b</i>	.1
<i>b</i>	<i>c</i>	.1
<i>c</i>	<i>d</i>	.05
<i>d</i>	<i>a</i>	1.
<i>d</i>	<i>b</i>	.8

<b>S</b>		
<i>b</i>	<i>d</i>	.7

## Restrict to CQs on graph signatures

$$\exists x y z t R(x, y) \wedge S(y, z) \wedge S(t, z) \quad \rightarrow \quad x \xrightarrow{R} y \xrightarrow{S} z \xleftarrow{S} t$$

---

**R**

---

<i>a</i>	<i>b</i>	.1
<i>b</i>	<i>c</i>	.1
<i>c</i>	<i>d</i>	.05
<i>d</i>	<i>a</i>	1.
<i>d</i>	<i>b</i>	.8

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**S**

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<i>b</i>	<i>d</i>	.7
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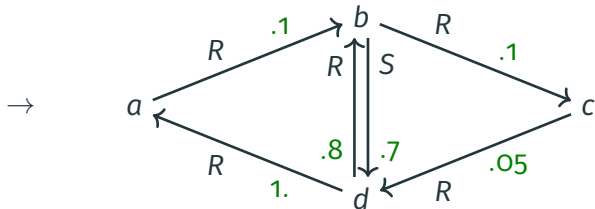
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$$\exists xyz t R(x,y) \wedge S(y,z) \wedge S(t,z) \quad \rightarrow \quad x \xrightarrow{R} y \xrightarrow{S} z \xleftarrow{S} t$$

R		
a	b	.1
b	c	.1
c	d	.05
d	a	1.
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S		
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## Restrict instances to trees

$\mathcal{Q}$  = one-way paths (1WP),  $\mathcal{I}$  = polytrees (PT)

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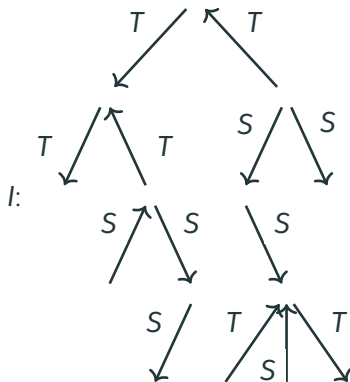
$\mathcal{Q}$  = one-way paths (1WP),  $\mathcal{I}$  = polytrees (PT)

Q:  $\xrightarrow{T} \xrightarrow{S} \xrightarrow{S} \xrightarrow{S} \xrightarrow{T}$



# Restrict instances to trees

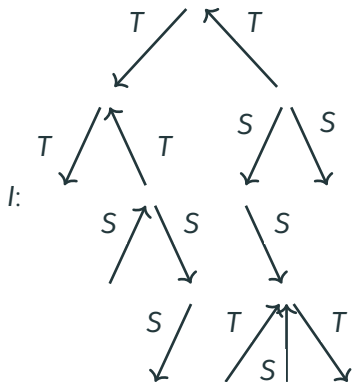
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+ prob. for each edge

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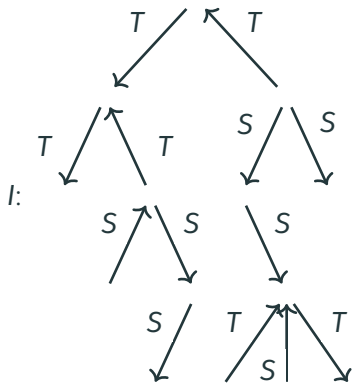
## Proposition

PQE of 1WP on PT is **#P-hard**

+ prob. for each edge

# $\mathcal{Q}$ = one-way paths, $\mathcal{I}$ = polytrees, without labels

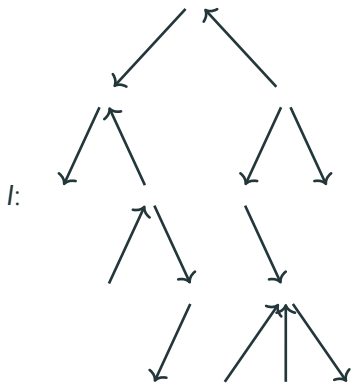
- What if we **do not have labels**?



+ prob. for each edge

## $\mathcal{Q} = \text{one-way paths}, \mathcal{I} = \text{polytrees, without labels}$

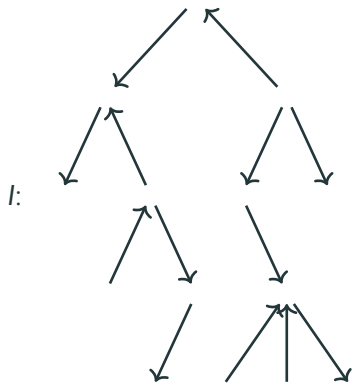
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## $Q = \text{one-way paths}, \mathcal{I} = \text{polytrees, without labels}$

- What if we **do not have labels**?
- Probability that the instance graph has a path of length  $|Q|$

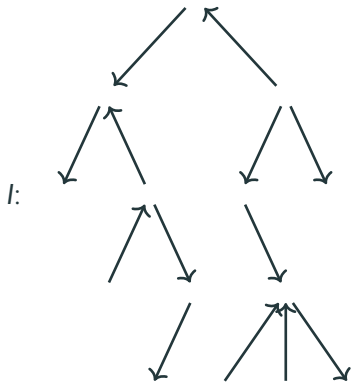


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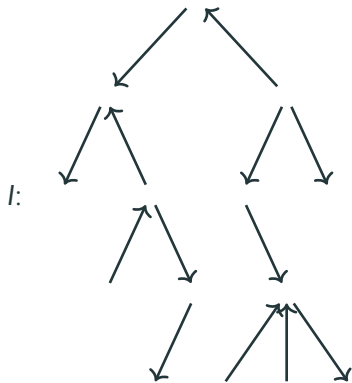
- What if we **do not have labels**?
- Probability that the instance graph has a path of length  $|Q|$
- **PTIME**: Bottom-up, e.g., tree automaton
- **Labels** have an impact!



+ prob. for each edge

# $\mathcal{Q}$ = two-way paths, $\mathcal{I}$ = polytrees, without labels

- $\mathcal{Q}$  = one-way paths (1WP),  $\mathcal{I}$  = polytrees (PT)

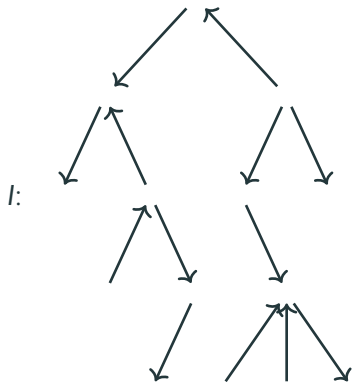


+ prob. for each edge



# $\mathcal{Q}$ = two-way paths, $\mathcal{I}$ = polytrees, without labels

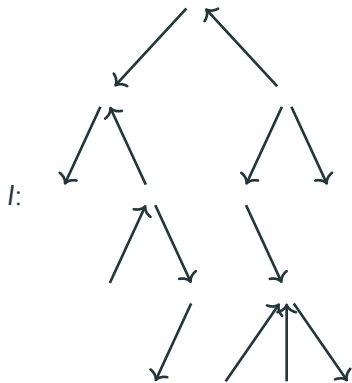
- $\mathcal{Q}$  = **two-way paths** (2WP),  $\mathcal{I}$  = **polytrees** (PT)



+ prob. for each edge

# $\mathcal{Q}$ = two-way paths, $\mathcal{I}$ = polytrees, without labels

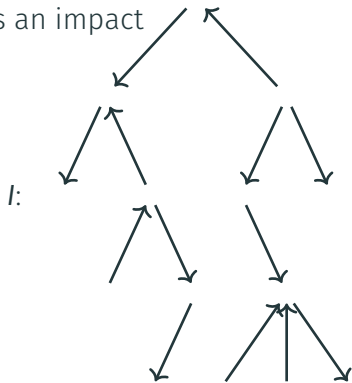
- $\mathcal{Q}$  = **two**-way paths (2WP),  $\mathcal{I}$  = **polytrees** (PT)
- **#P-hard**



+ prob. for each edge

# $\mathcal{Q}$ = two-way paths, $\mathcal{I}$ = polytrees, without labels

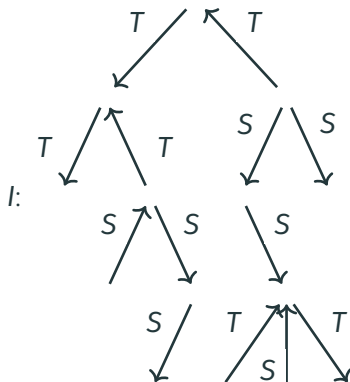
- $\mathcal{Q}$  = **two**-way paths (2WP),  $\mathcal{I}$  = **polytrees** (PT)
- **#P-hard**
- **Global orientation** of the **query** has an impact



+ prob. for each edge

$\mathcal{Q} = \text{one-way paths}, \mathcal{I} = \text{downwards trees}$

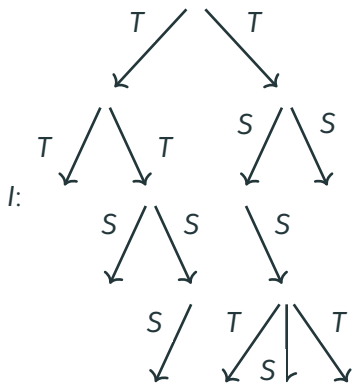
- $\mathcal{Q} = \text{one-way paths (1WP)}, \mathcal{I} = \text{polytrees (PT)}$



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$\mathcal{Q} = \text{one-way paths}, \mathcal{I} = \text{downwards trees}$

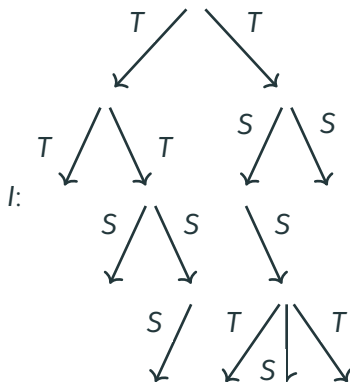
- $\mathcal{Q} = \text{one-way paths (1WP)}, \mathcal{I} = \text{downwards trees (DWT)}$



+ prob. for each edge 10/17

$\mathcal{Q} = \text{one-way paths}, \mathcal{I} = \text{downwards trees}$

- $\mathcal{Q} = \text{one-way paths}$  (1WP),  $\mathcal{I} = \text{downwards trees}$  (DWT)
- **PTIME** also:  $\beta$ -acyclicity of the lineage

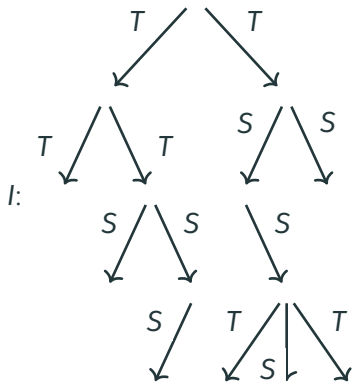


+ prob. for each edge 10/17



$\mathcal{Q}$  = downwards trees,  $\mathcal{I}$  = downwards trees, with labels

- $\mathcal{Q}$  = one-way paths (1WP),  $\mathcal{I}$  = downwards trees

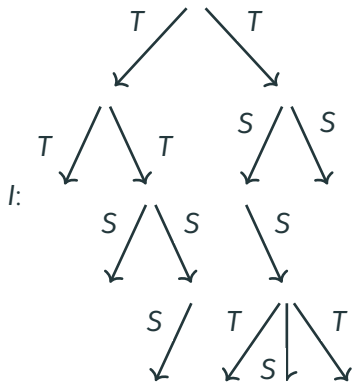
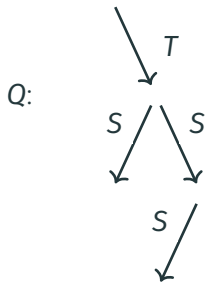


+ prob. for each edge



$\mathcal{Q}$  = downwards trees,  $\mathcal{I}$  = downwards trees, with labels

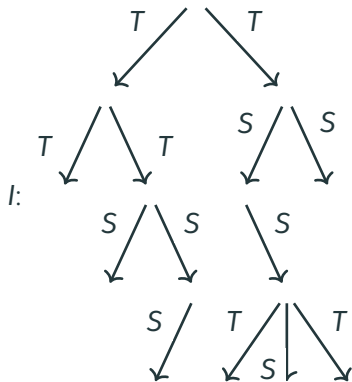
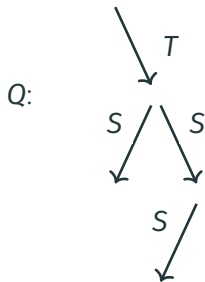
- $\mathcal{Q}$  = **downwards trees** (DWT),  $\mathcal{I}$  = downwards trees



+ prob. for each edge

# $\mathcal{Q}$ = downwards trees, $\mathcal{I}$ = downwards trees, with labels

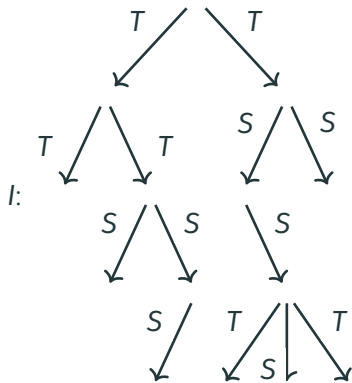
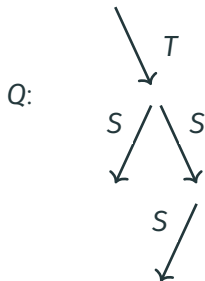
- $\mathcal{Q}$  = **downwards trees** (DWT),  $\mathcal{I}$  = **downwards trees**
- **#P-hard**



+ prob. for each edge

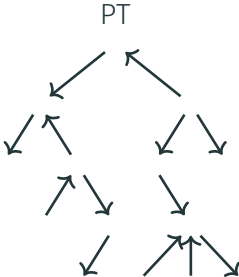
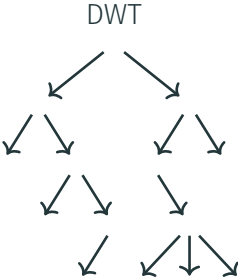
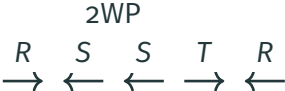
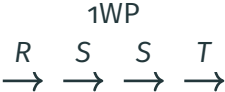
# $\mathcal{Q}$ = downwards trees, $\mathcal{I}$ = downwards trees, with labels

- $\mathcal{Q}$  = **downwards trees** (DWT),  $\mathcal{I}$  = **downwards trees**
- **#P-hard**
- **Branching** has an impact!



+ prob. for each edge

# Our graph classes



$1WP \subsetneq 2WP$   
 $1WP \subsetneq DWT$   
 $PT \subseteq \text{Connected} \subseteq \text{All}$

# Results

$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected
1WP						
2WP						
DWT			PTIME			
PT						
Connected						

$\geq 2$  labels

#P-hard

# Results

$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected	
1WP							$\geq 2$ labels
2WP							
DWT			PTIME				
PT						#P-hard	
Connected							
$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected	
1WP							No labels
2WP							
DWT			PTIME				
PT						#P-hard	
Connected							

# Results

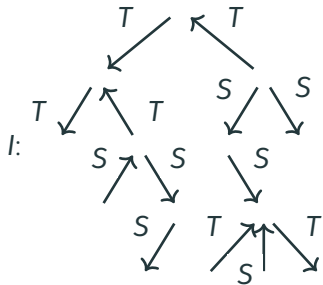
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1WP				•	•		$\geq 2$ labels
2WP				•			
DWT			PTIME	•			
PT						#P-hard	
Connected			•				
$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected	
1WP						•	No labels
2WP					•		
DWT			PTIME		•		
PT						#P-hard	
Connected			•	•			

# Results

$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected	
1WP					•		$\geq 2$ labels
2WP							
DWT			PTIME				
PT						#P-hard	
Connected							
$\downarrow Q$	$I \rightarrow$	1WP	2WP	DWT	PT	Connected	
1WP							No labels
2WP							
DWT			PTIME				
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Connected							



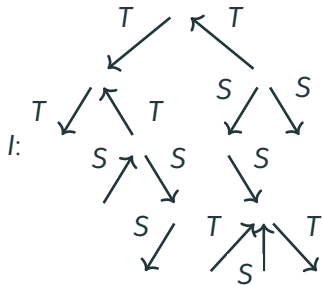
## Reduction for $Q = \text{one-way paths}$ , $\mathcal{I} = \text{polytrees}$



+ prob. for each edge



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Reduction from **#P-hard** problem **#PP2DNF**:

- INPUT:** Boolean formula  $\varphi = \bigvee_{j=1 \dots m} (X_{x_j} \wedge Y_{y_j})$  on variables  $\{X_1, \dots, X_{n_1}\} \sqcup \{Y_1, \dots, Y_{n_2}\}$
- OUTPUT:** number of satisfying assignments of  $\varphi$

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$$\varphi = X_1 Y_2 \vee X_1 Y_1 \vee X_2 Y_2$$

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$I$ :

$Q$ :

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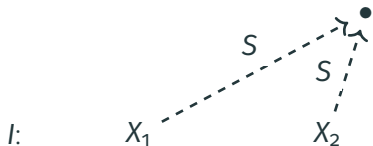
•

I:

Q:

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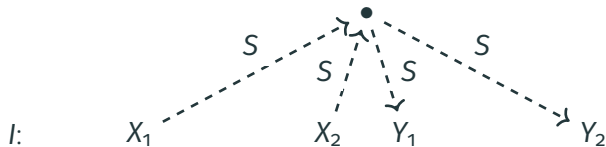
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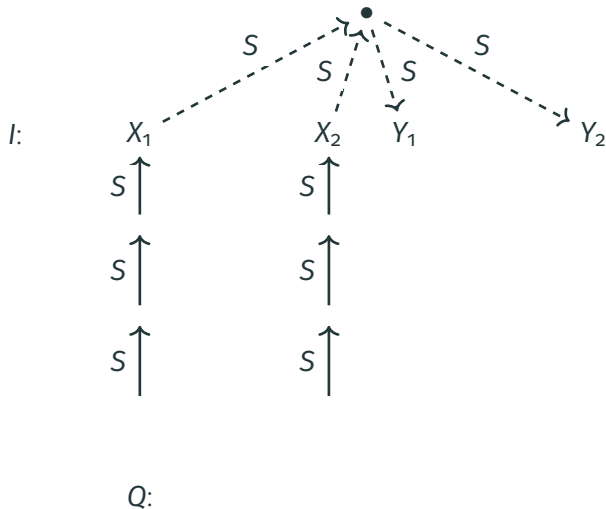


Q:



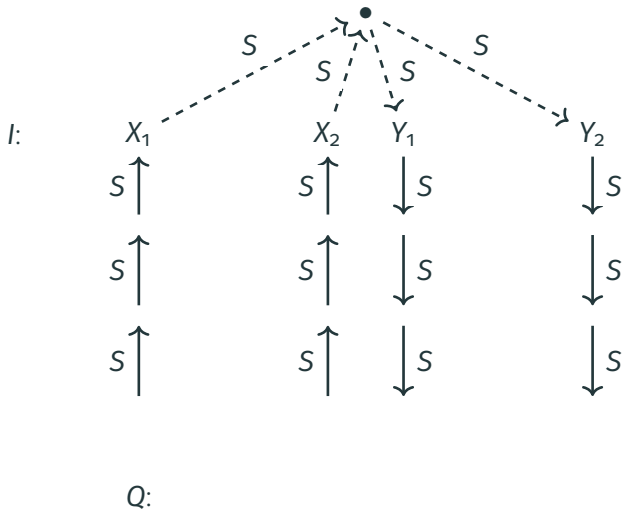
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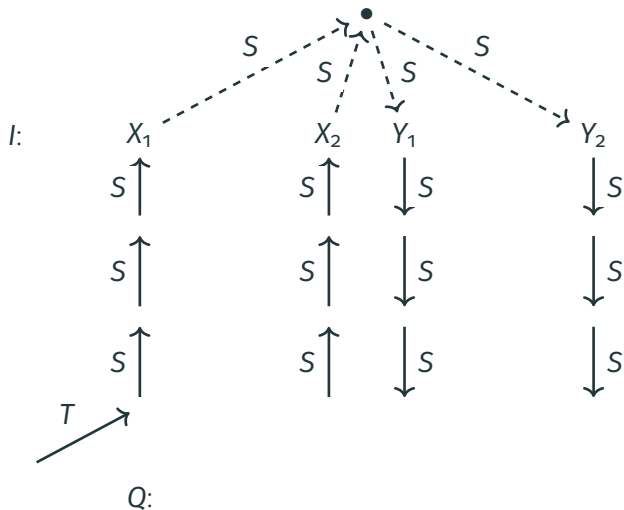
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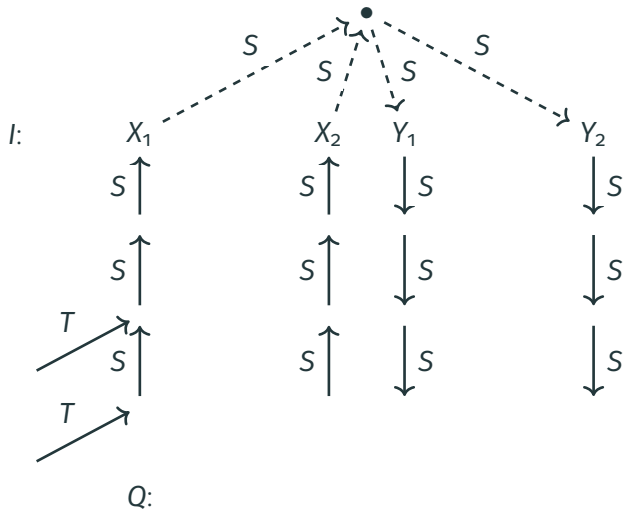
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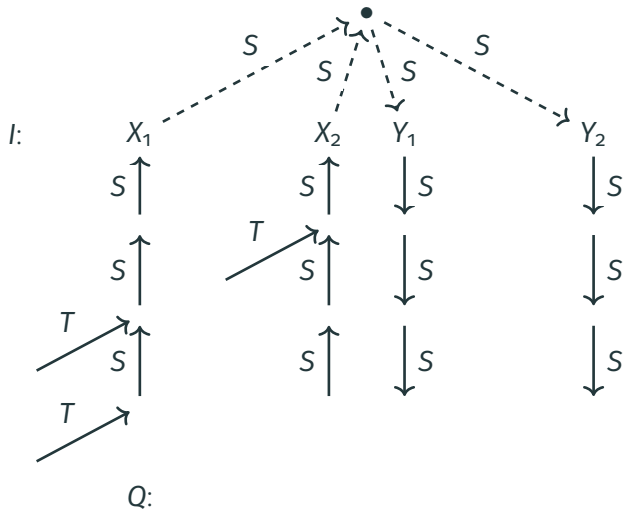
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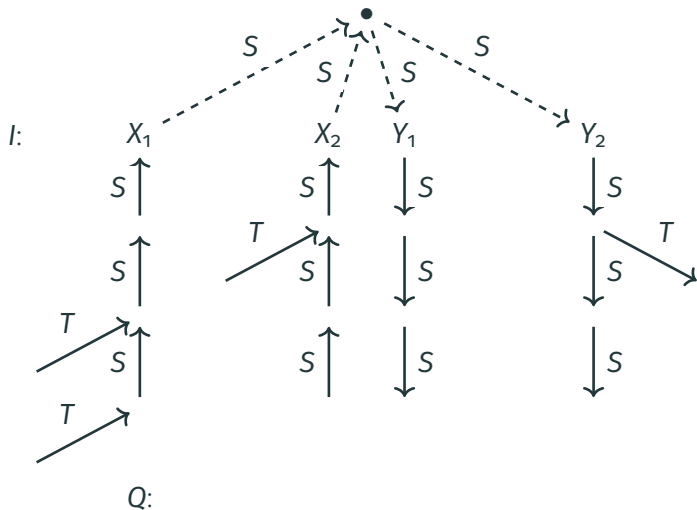
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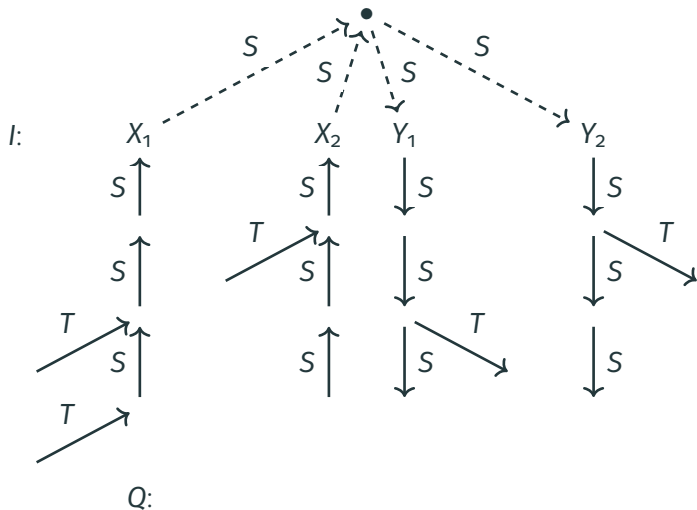
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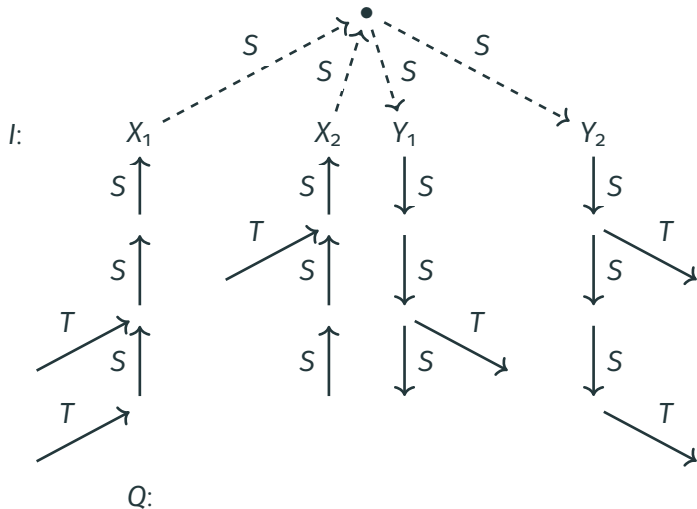
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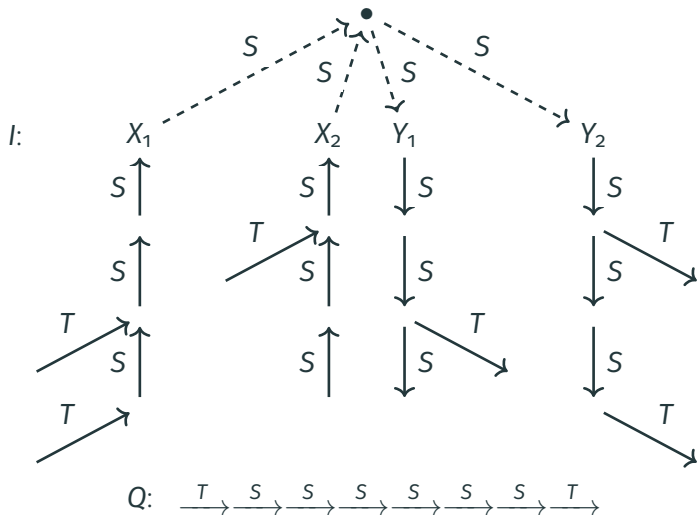
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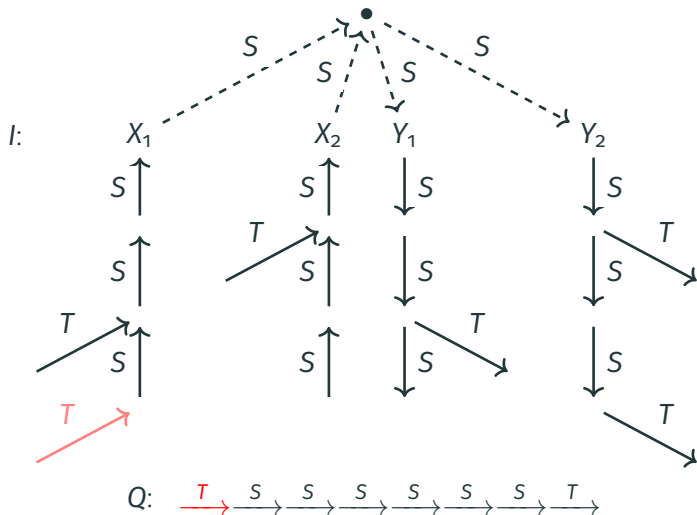
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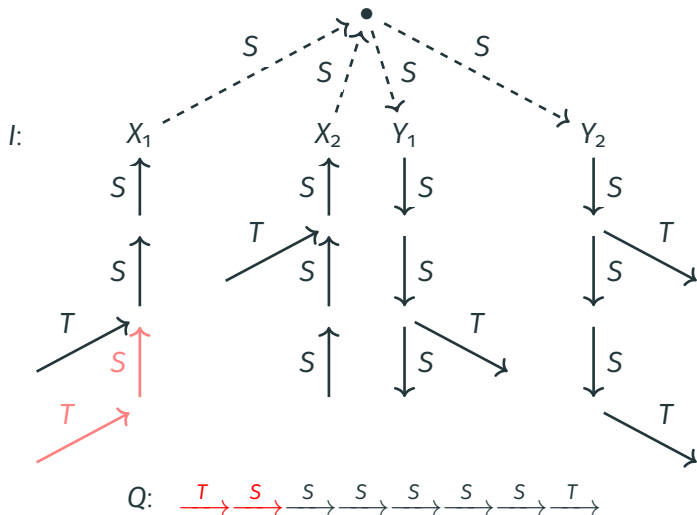
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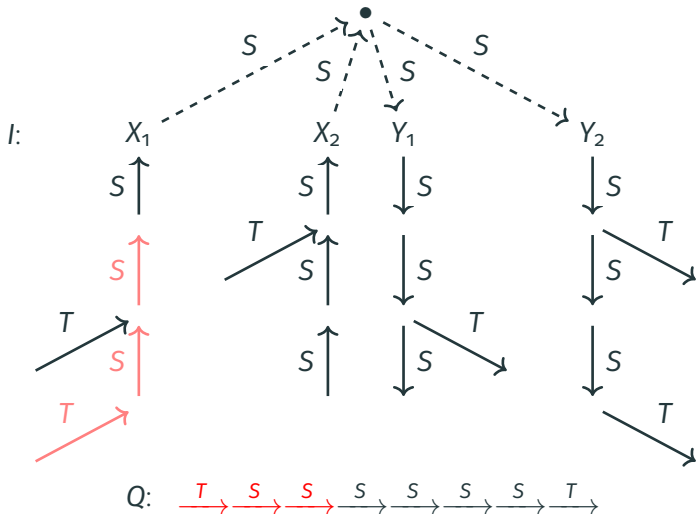
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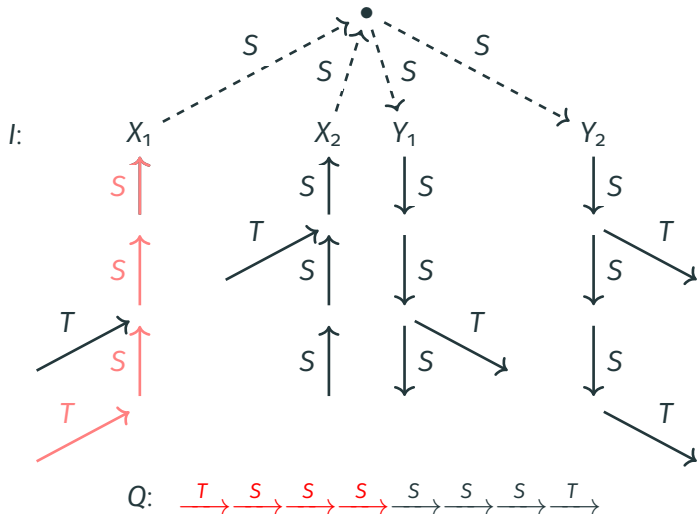
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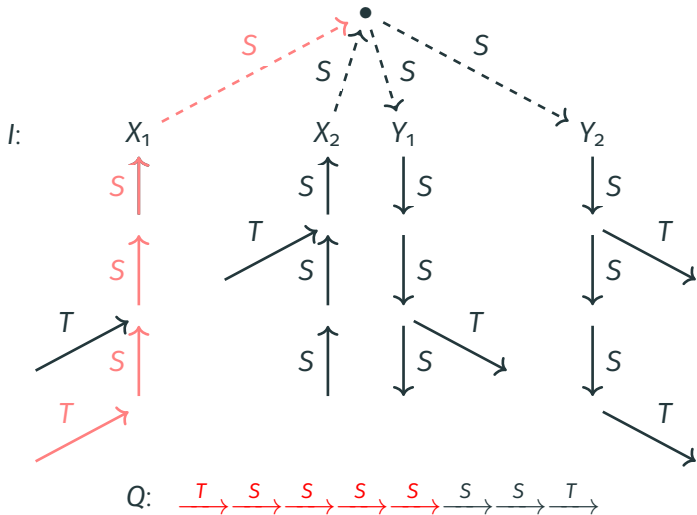
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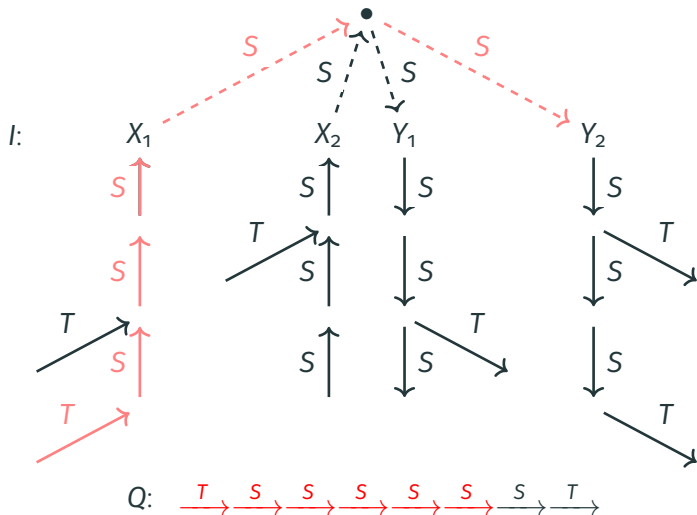
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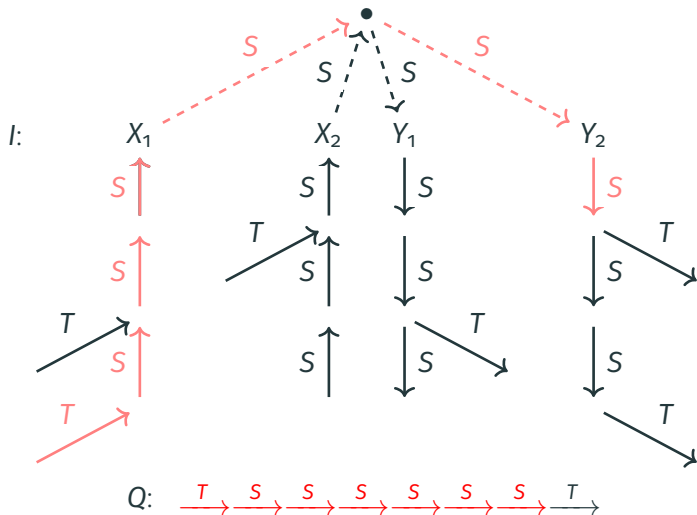
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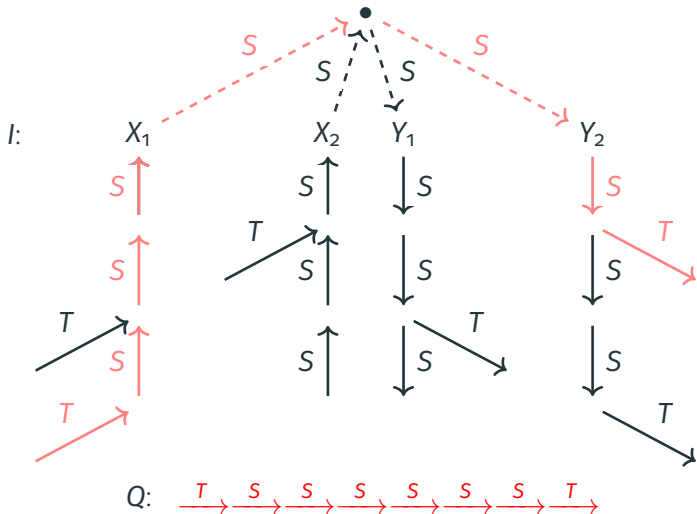
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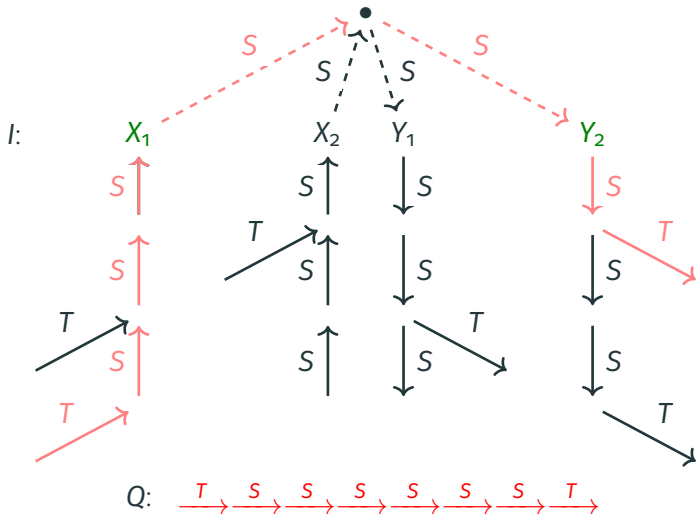
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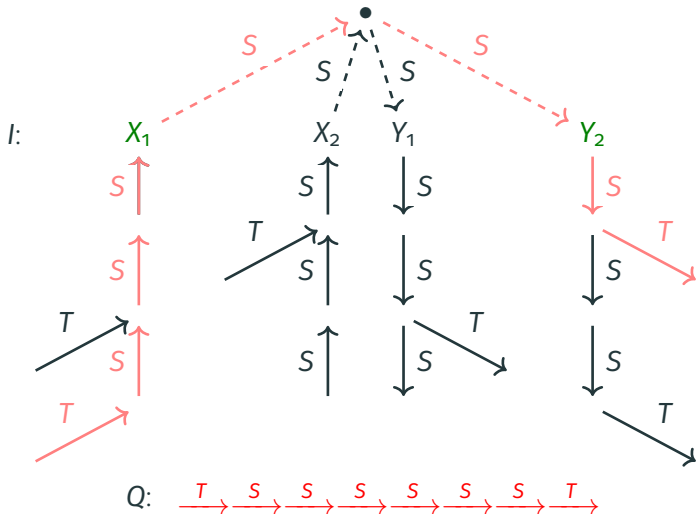
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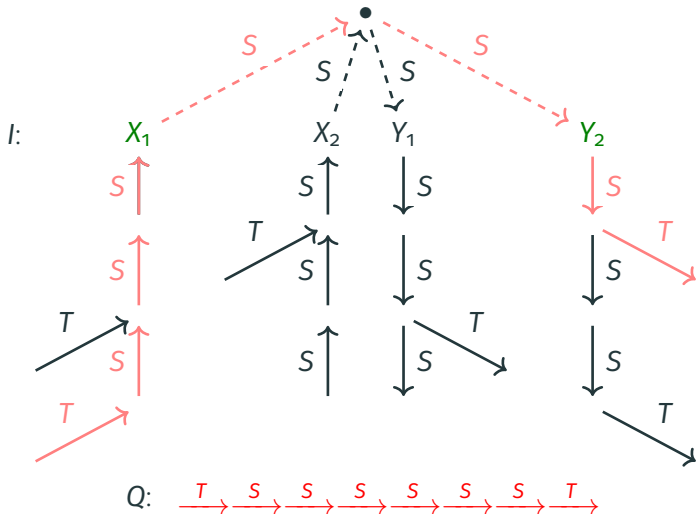
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$$\#\varphi = \Pr((I, \pi) \models Q) \times 2^{|\text{vars}(\varphi)|}$$



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With labels, PQE of  $\sqcup 1WP$  on  $1WP$  is already **#P-hard!**

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Thanks for your attention!