

Constraint Satisfaction Problems (CSPs)

Agenda

- CSP Problem Modeling
- N-ary Constraints
- Elimination Example

- **CSP Problem Modeling**
- N-ary Constraints
- Elimination Example

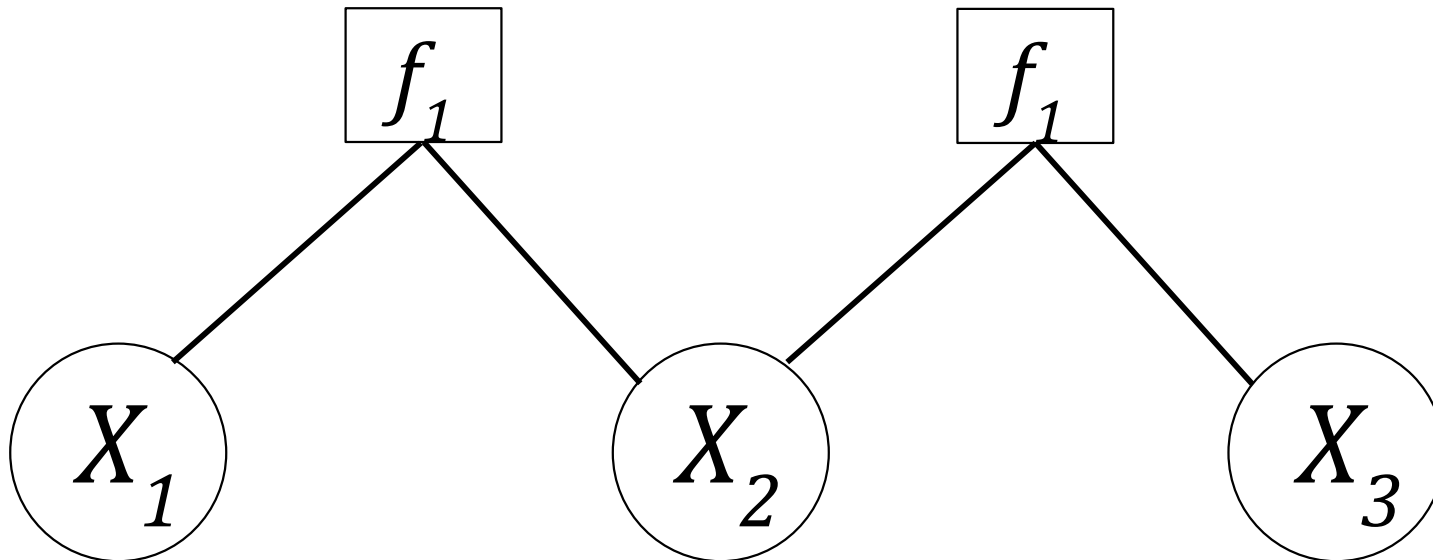
Definition: Factor Graph

Variables:

$$X = (X_1, \dots, X_n), \text{ where } X_i \in \text{Domain}_i$$

Factors:

$$f_1, \dots, f_m, \text{ with each } f_j(X) \geq 0$$



Definition: Constraint Satisfaction Problem (CSP)

A CSP is a factor graph where all factors are **constraints**:

for all $j = 1, \dots, m$.

The constraint is satisfied iff $f_j(x) = 1$.

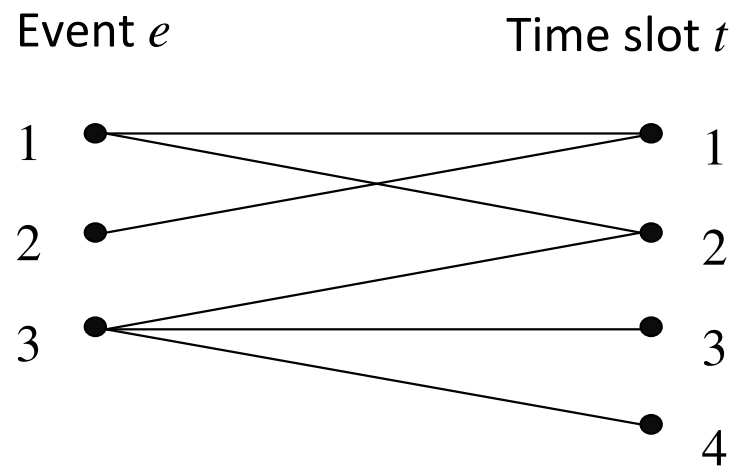
Definition: Consistent Assignments

An assignment x if $Weight(x) = 1$ (i.e., all constraints are satisfied.)

Event Scheduling

Setup:

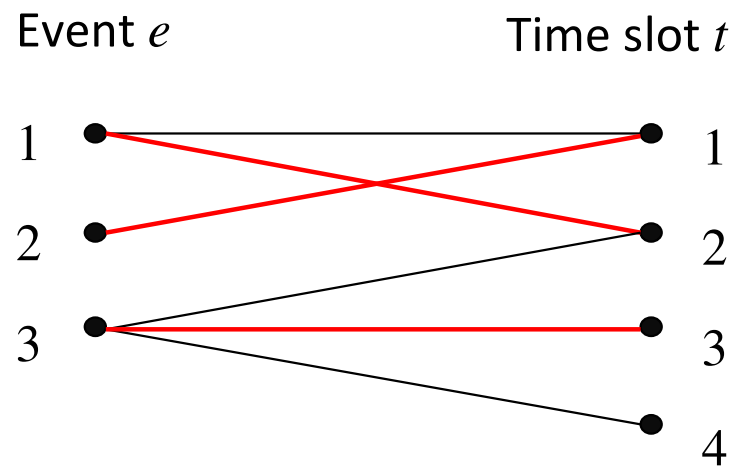
- Have E events and T time slots
- Each event e must be put in **exactly one** time slot
- Each time slot t can have **at most one** event
- Event e only allowed at time slot t if (e, t) in A



Event Scheduling

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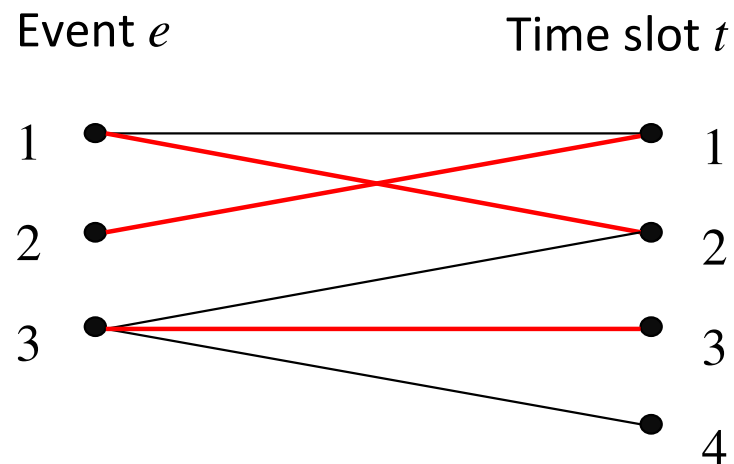
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Event Scheduling

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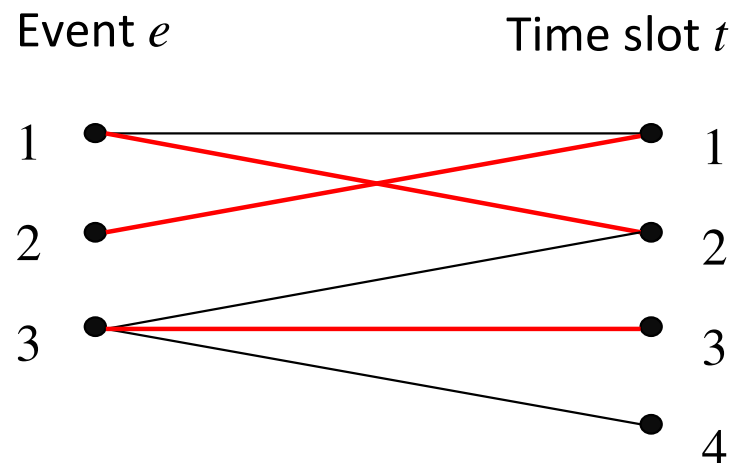
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Event Scheduling

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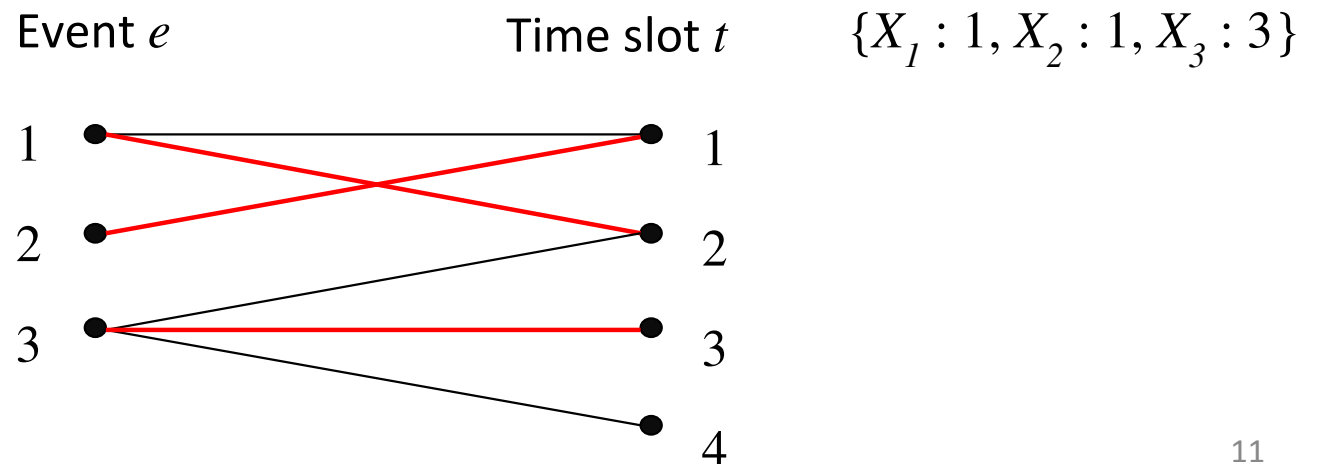
- Variables for each event e , $X_e \in \{1, \dots, T\}$
- Constraints (only one event per time slot): for each pair of events $e \neq e'$, enforce $[X_e \neq X_{e'}]$
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Event Scheduling

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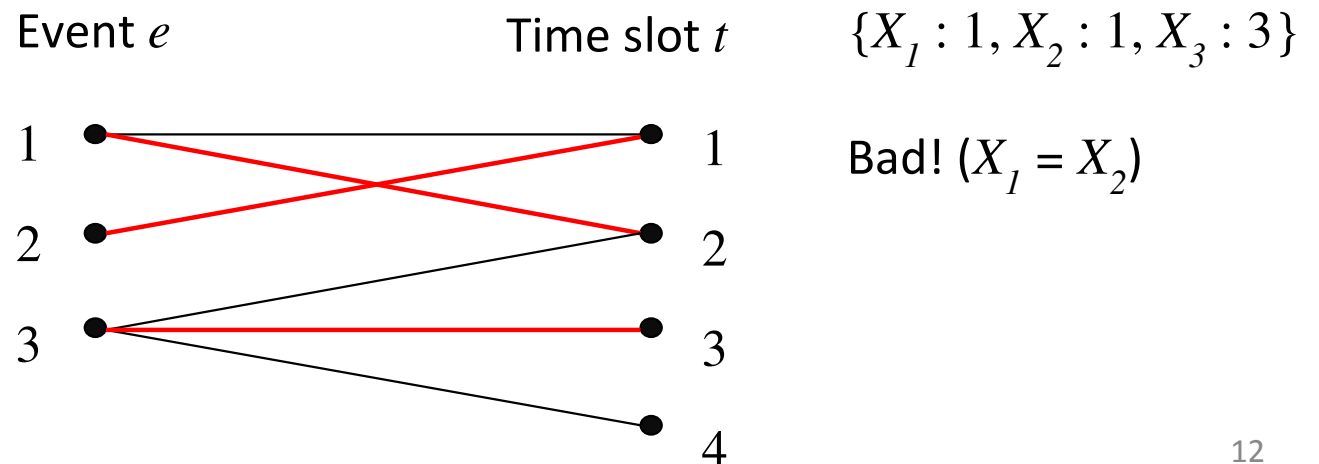
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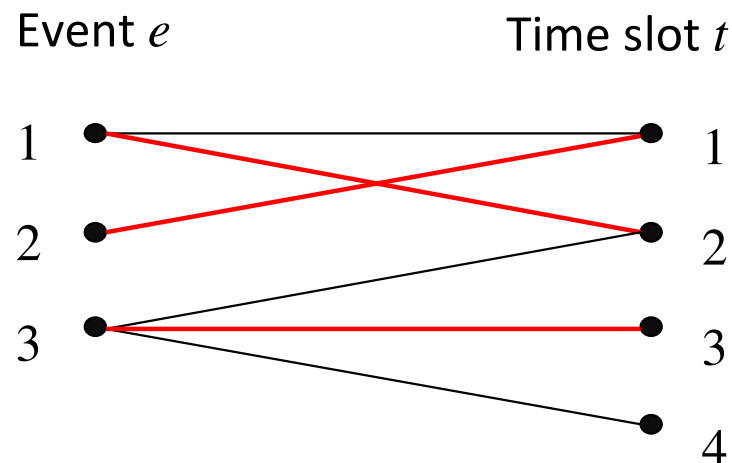
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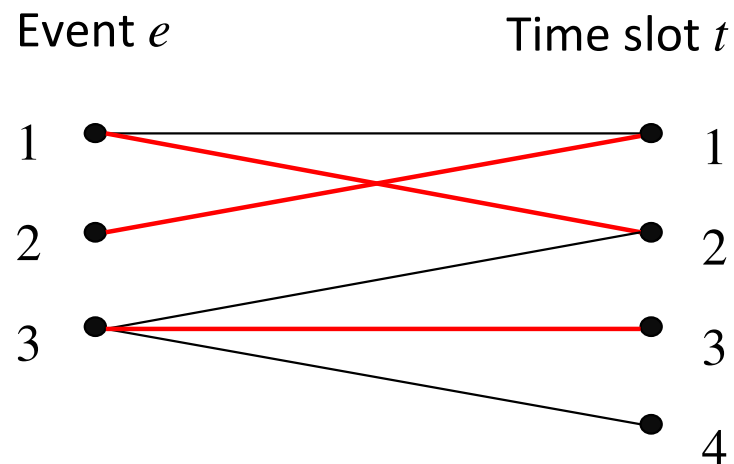
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Event Scheduling

Formulation 1b:

- Variables for each event e , X_1, \dots, X_E

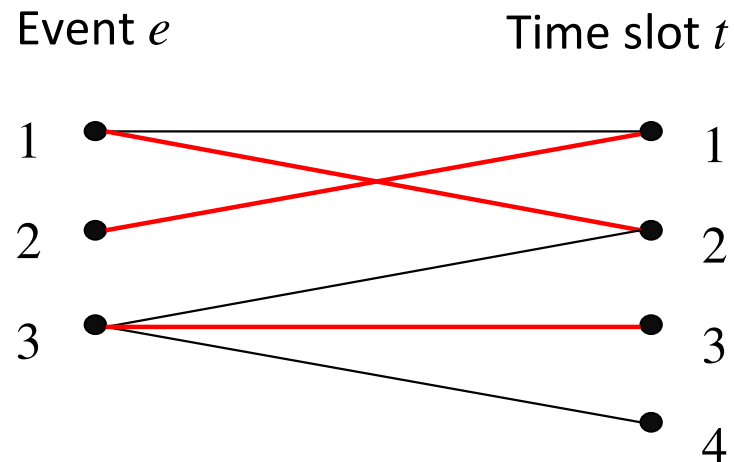


Event Scheduling

Formulation 1b:

- Variables for each event e , X_1, \dots, X_E

$$\text{Domain}_i = \{t : (i, t) \in A\}$$



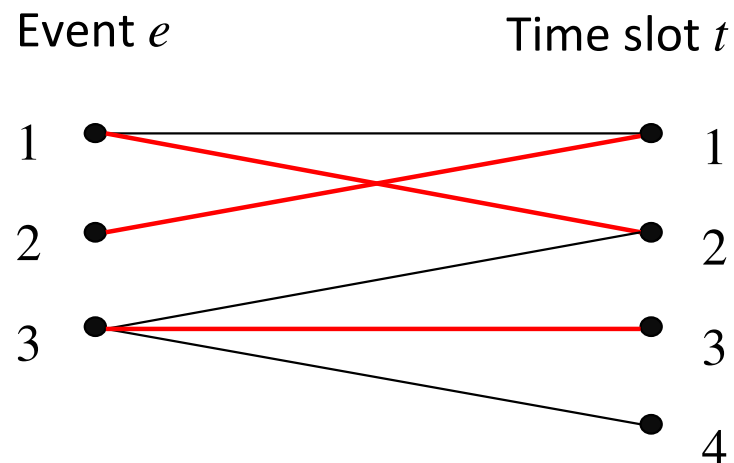
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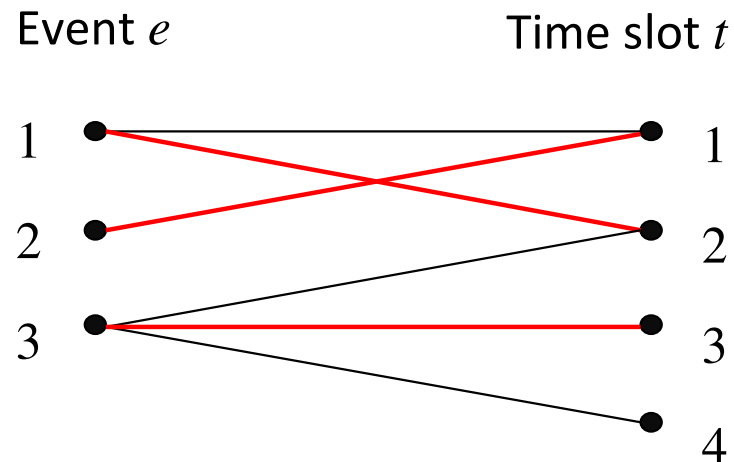
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Event Scheduling

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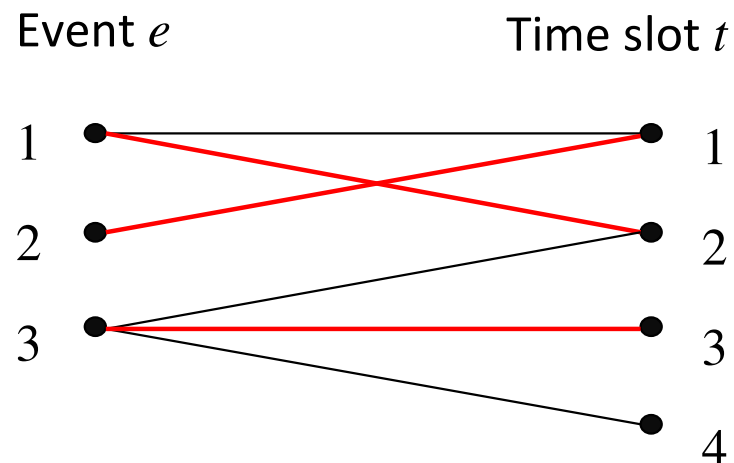
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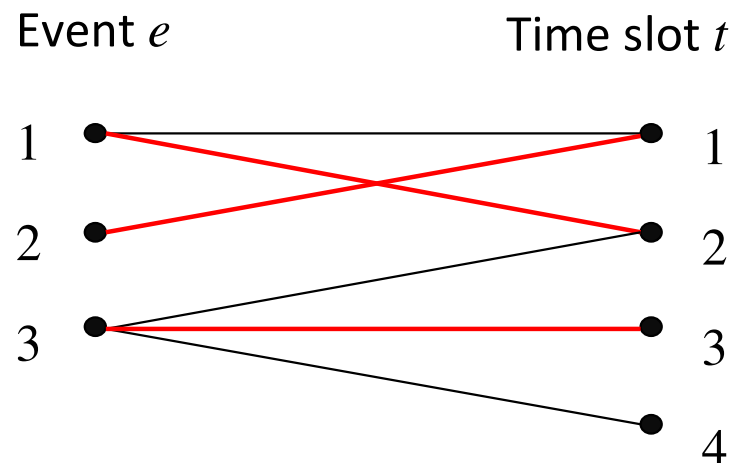
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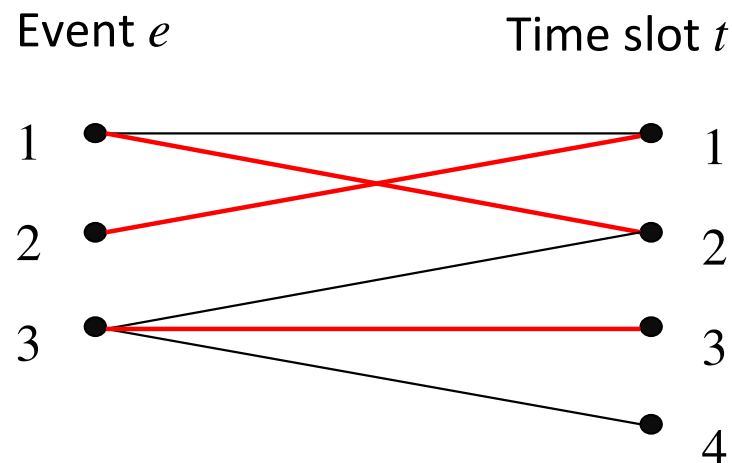
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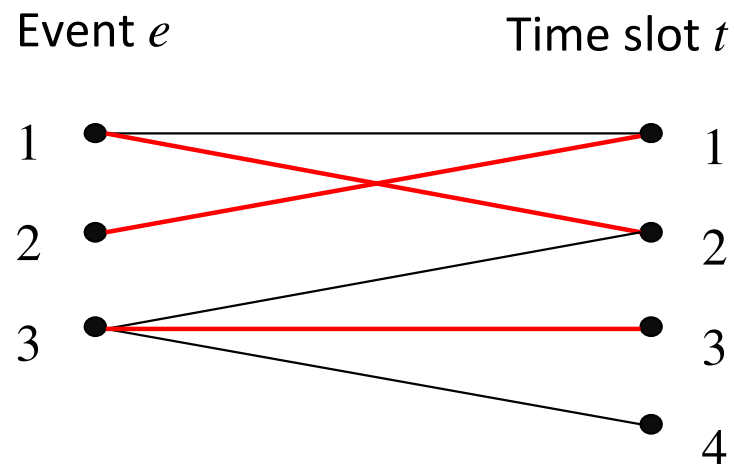
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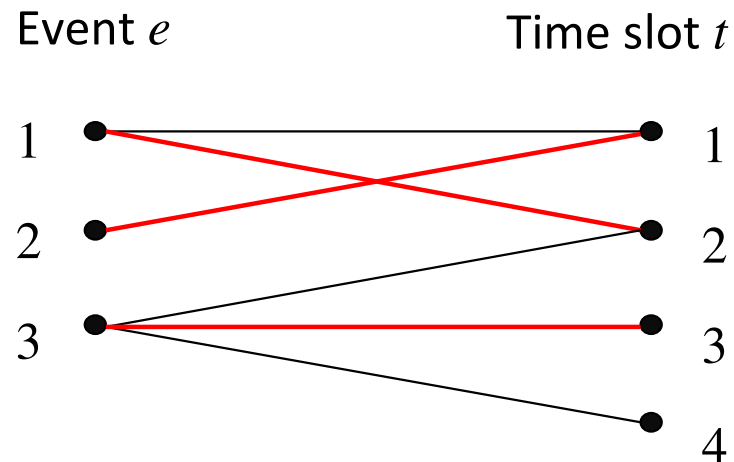


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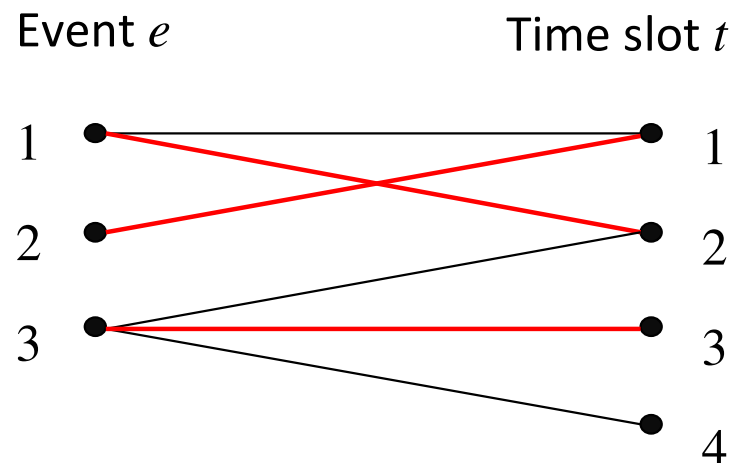
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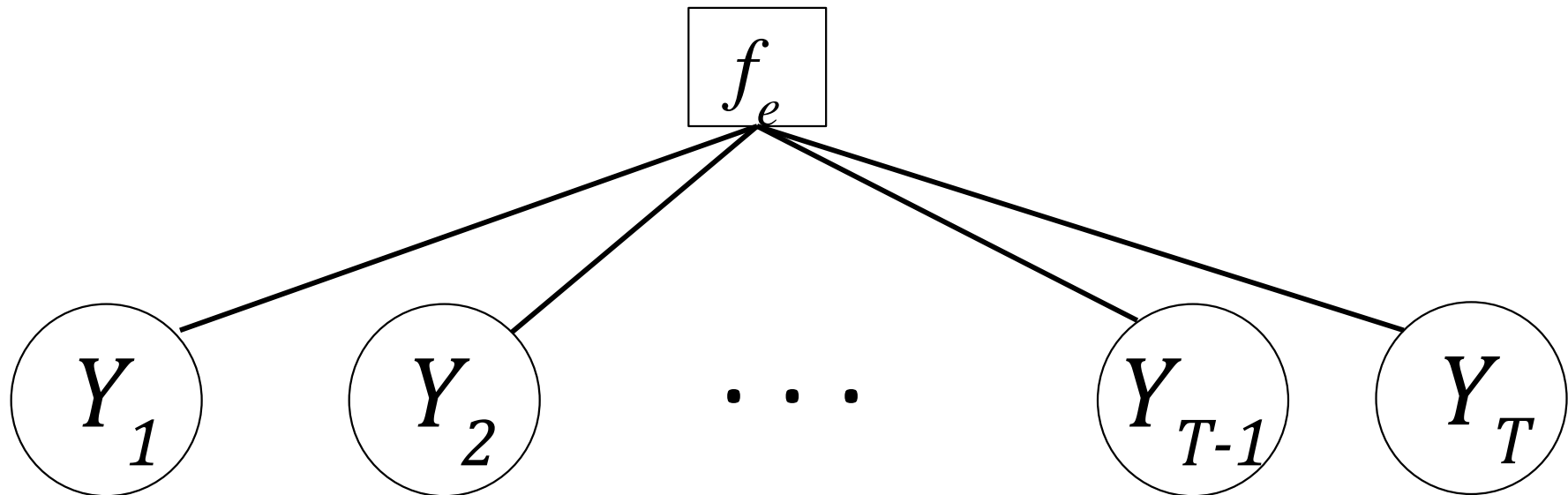
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- Problem Modeling
- **N-ary Constraints**
- Elimination Example

N-ary Constraints

- From event scheduling:
 - Constraints (each event is scheduled exactly once): for each event e , enforce
 $[Y_t = e \text{ for exactly one } t]$



N-ary Constraints

Key Idea: Auxiliary Variables

Auxiliary Variables hold intermediate computation.

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Factors:

Initialization: $[A_0 = 0]$

i	0	1	2	3	4
Y_i		3	1	2	1
A_i	0				

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Initialization: $[A_0 = 0]$

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Final Output: $1[A_T = 1]$

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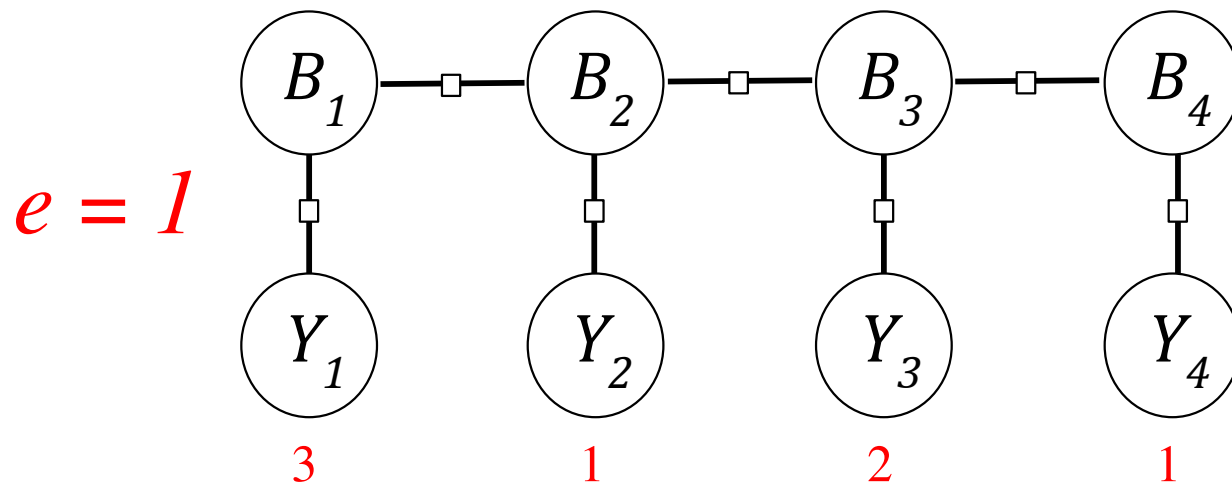
Still have factors with three variables...

N-ary Constraints

Key idea: Combine A_{i-1} and A_i into one variable B_i

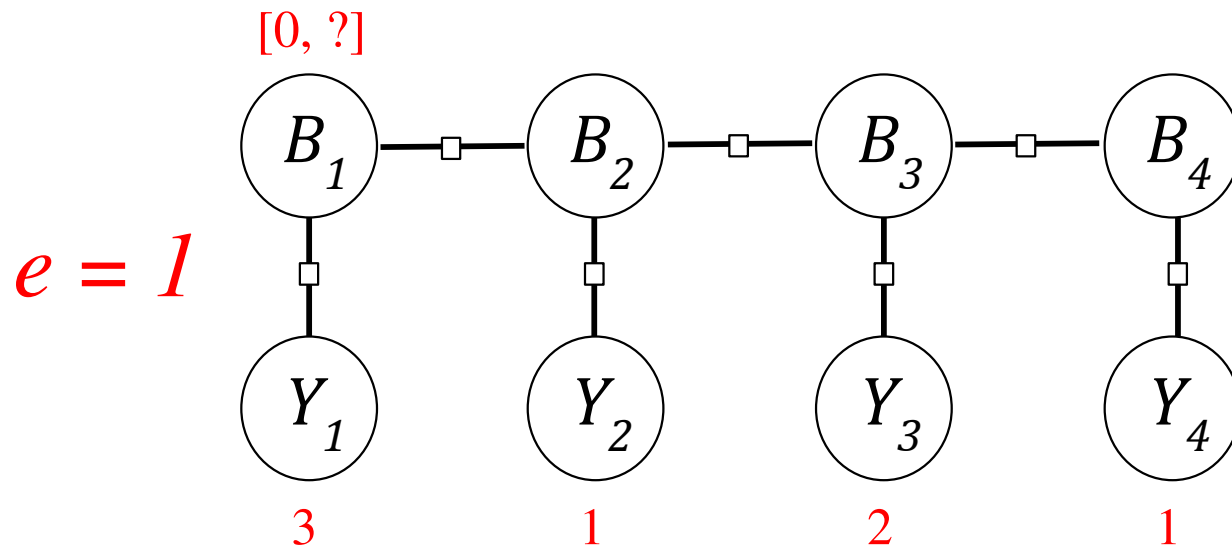
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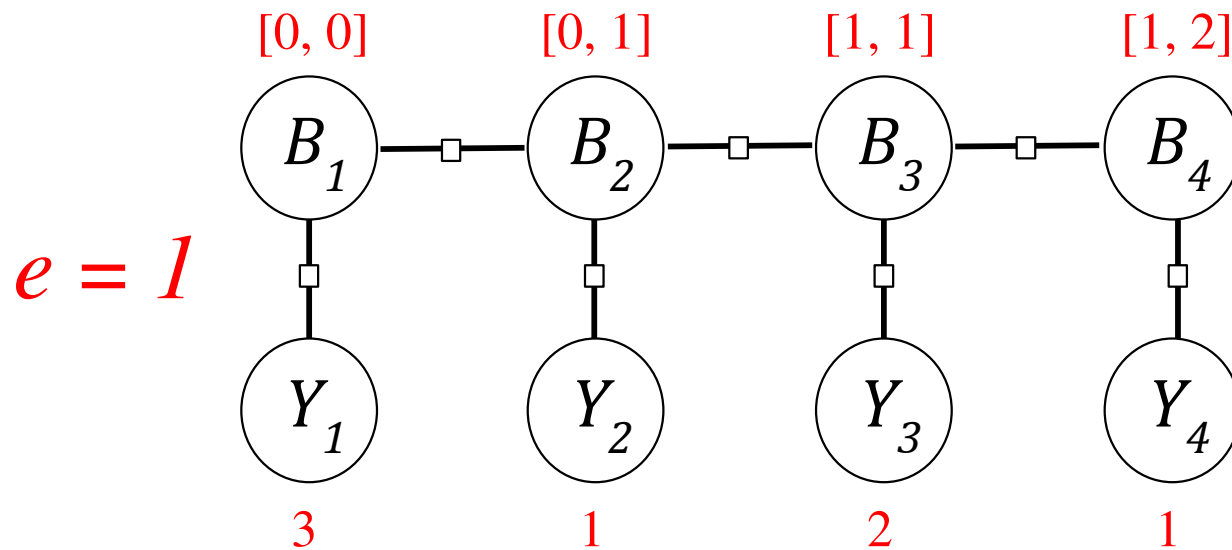


Factors:

Initialization: $[B_1[0] = 0]$

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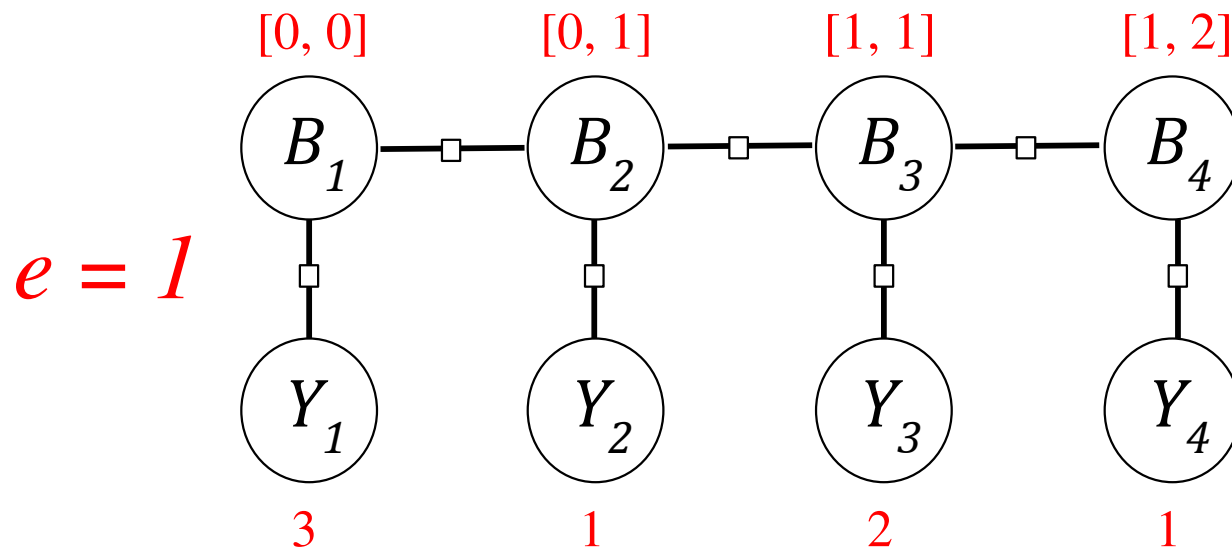
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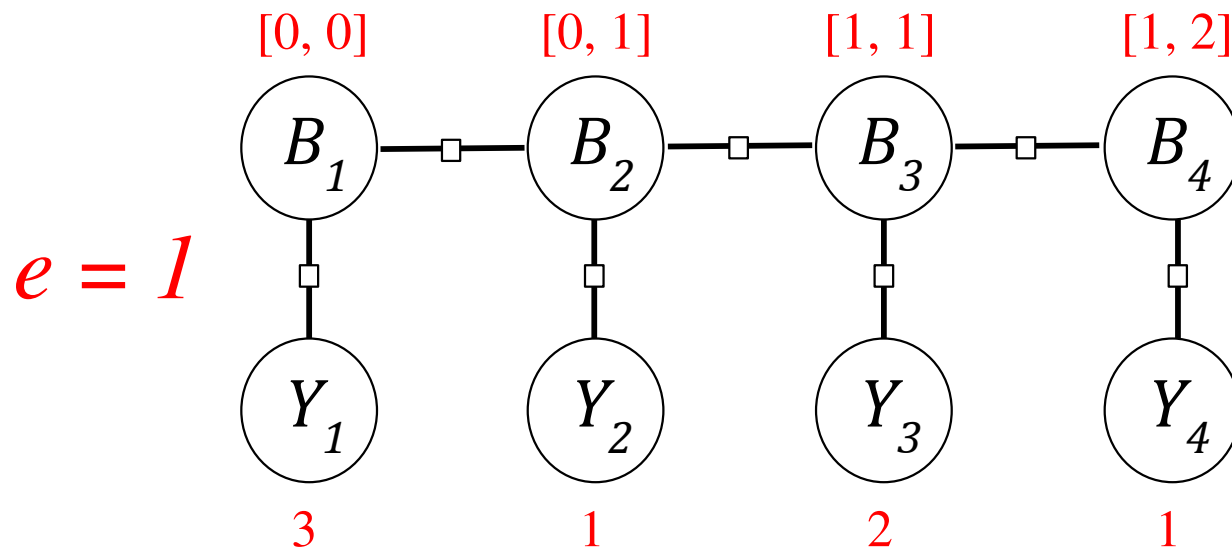
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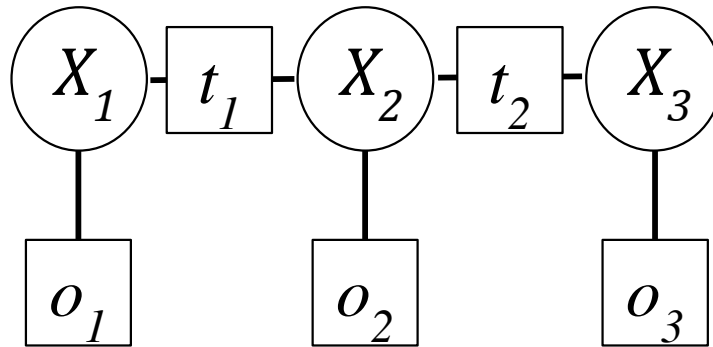
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Final Output: $1[B_T[1] = 1]$

Consistency: $[B_{i-1}[1] = B_i[0]]$

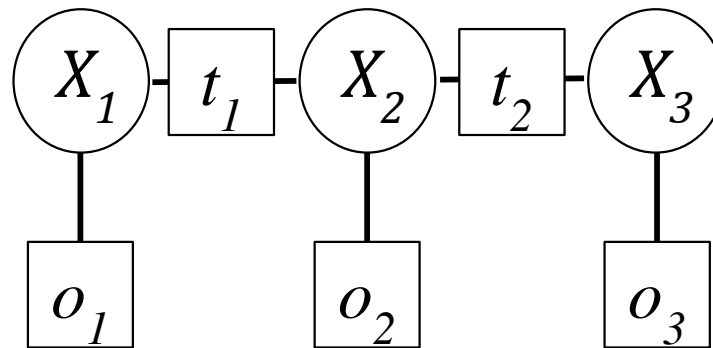
- Problem Modeling
- N-ary Constraints
- **Elimination Example**

Person Tracking Example



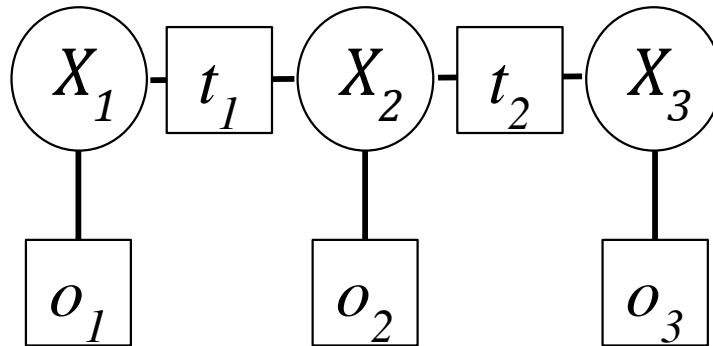
- Variables X_i : Location of object at position i

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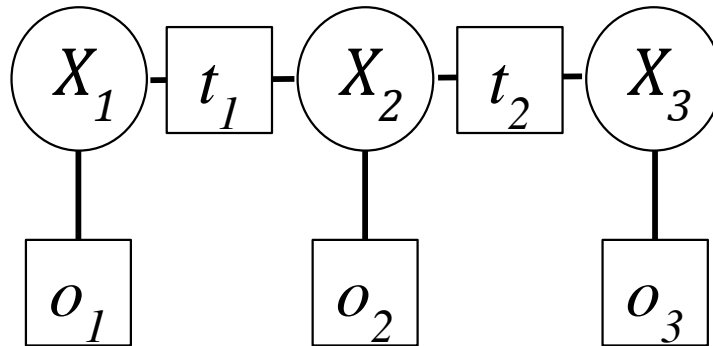
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- Observation Factors $o_i(x_i)$: noisy information compatible with position

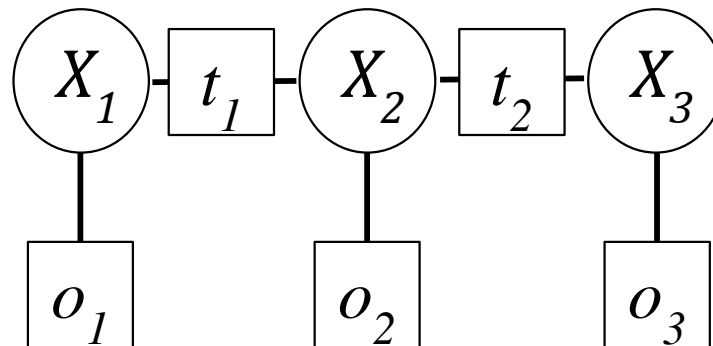
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```
def t(x, y):  
  if x == y: return 2  
  if abs(x - y) == 1: return 1  
  return 0
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Person Tracking Example



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def t(x, y):  
  if x == y: return 2  
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```
def o1(x): return t(x, 0)  
def o2(x): return t(x, 2)  
def o3(x): return t(x, 2)
```

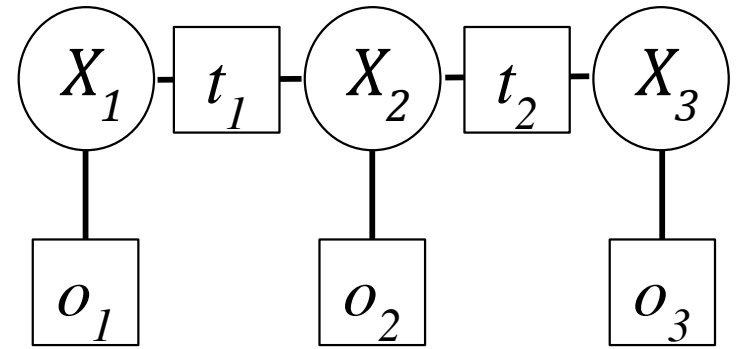
Variable Elimination

Definition: Elimination

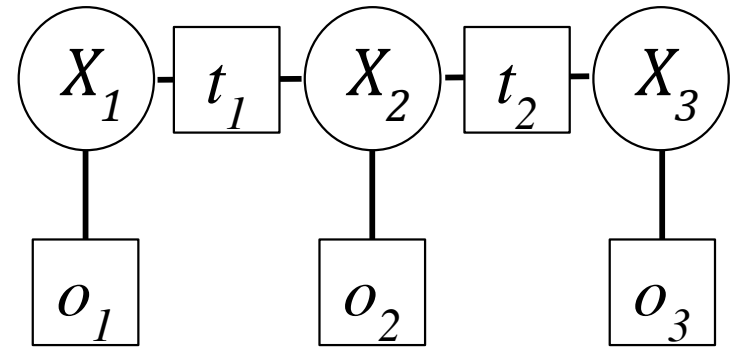
- To **eliminate** a variable X_i , consider all factors f_1, \dots, f_k that depend on X_i
- Remove X_i and f_1, \dots, f_k

- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- Eliminate X_1



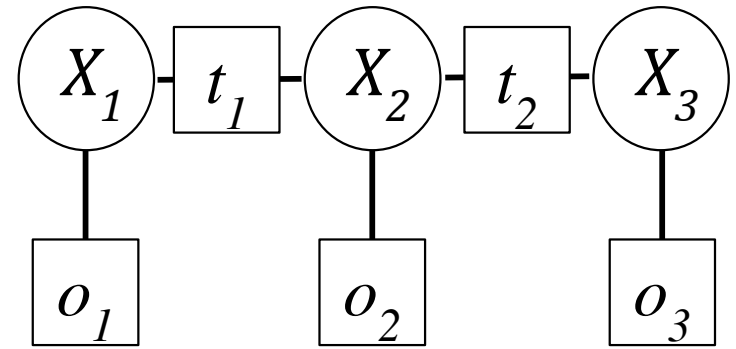
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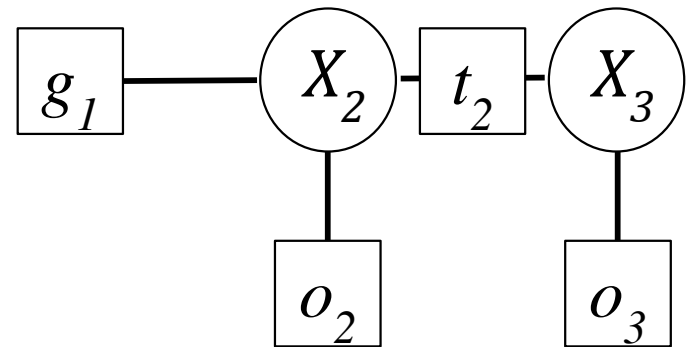
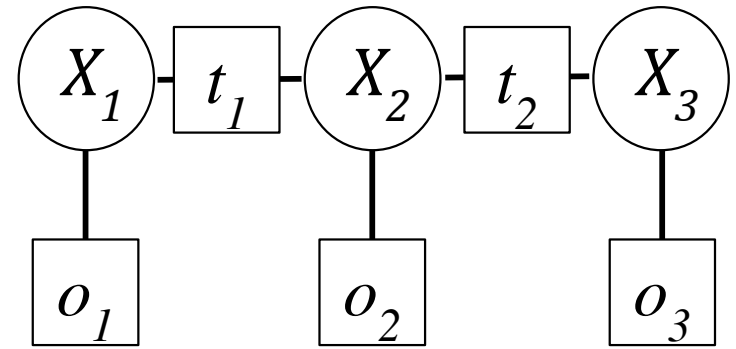


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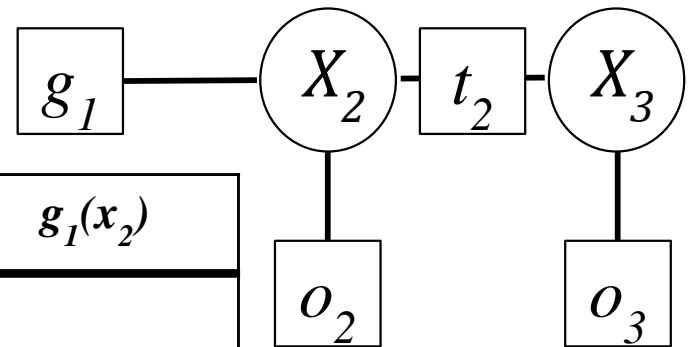
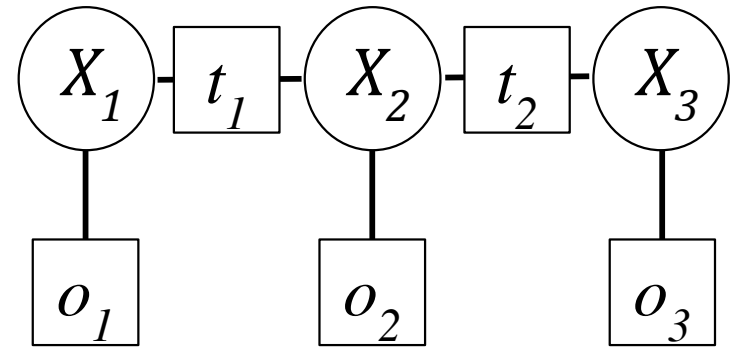
- $g_1(x_2) = \max_{x_1 \in \{0,1,2\}} o_1(x_1) \cdot t_1(x_1, x_2)$



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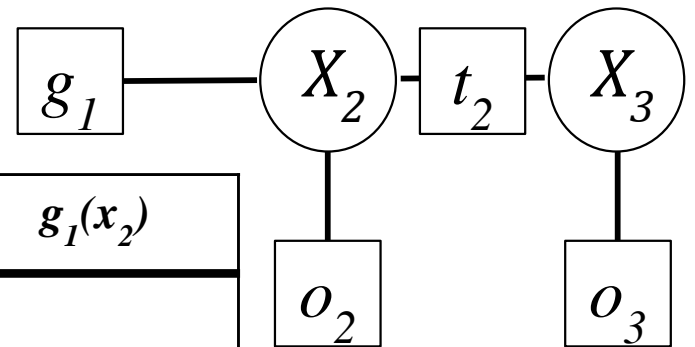
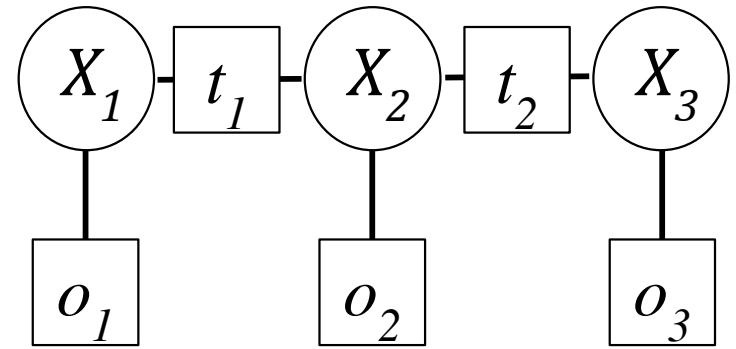


x_2	x_1	$o_1(x_1)$	$t_1(x_1, x_2)$	$o_1(x_1) t_1(x_1, x_2)$	$g_1(x_2)$
0	0				
0	1				
0	2				
1	0				
1	1				
1	2				
2	0				
2	1				
2	2				

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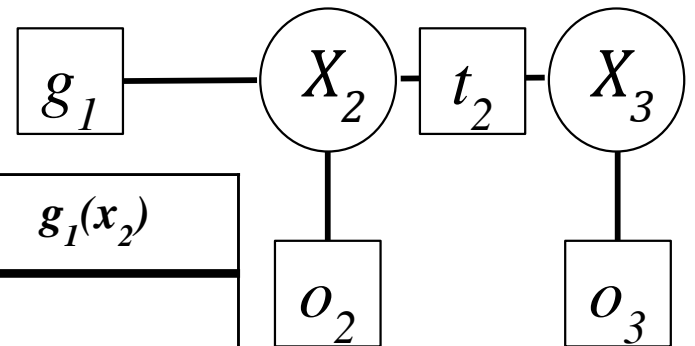
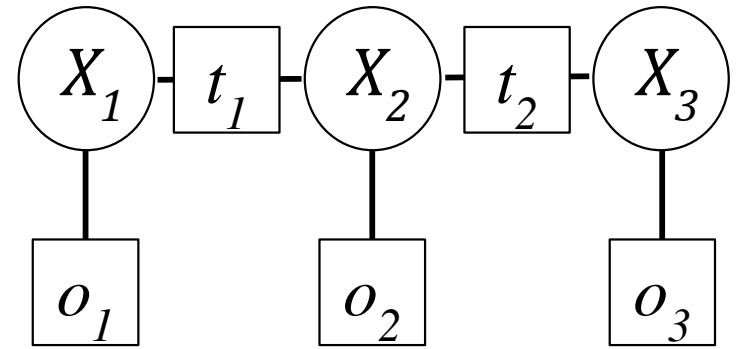


x_2	x_1	$o_1(x_1)$	$t_1(x_1, x_2)$	$o_1(x_1) t_1(x_1, x_2)$	$g_1(x_2)$
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0	1	1			
0	2	0			
1	0	2			
1	1	1			
1	2	0			
2	0	2			
2	1	1			
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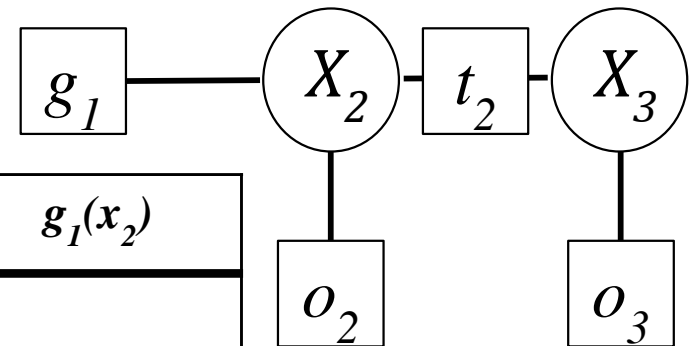
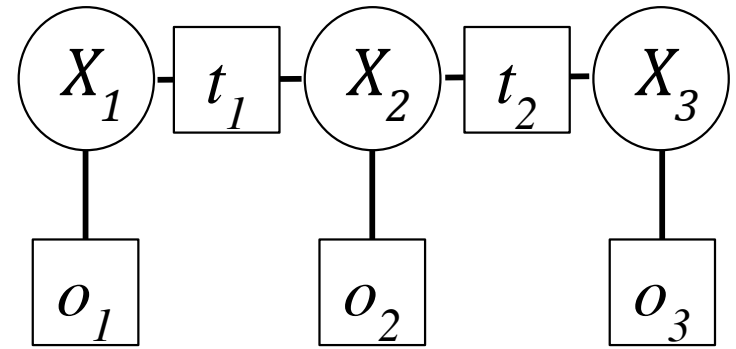


x_2	x_1	$o_1(x_1)$	$t_1(x_1, x_2)$	$o_1(x_1) t_1(x_1, x_2)$	$g_1(x_2)$
0	0	2	2		
0	1	1	1		
0	2	0	0		
1	0	2	1		
1	1	1	2		
1	2	0	1		
2	0	2	0		
2	1	1	1		
2	2	0	2		

- Eliminate X_1
- Factors that depend on X_1 :
 - o_1, t_1

- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_1(x_2) = \max_{x_1 \in \{0,1,2\}} o_1(x_1) \cdot t_1(x_1, x_2)$

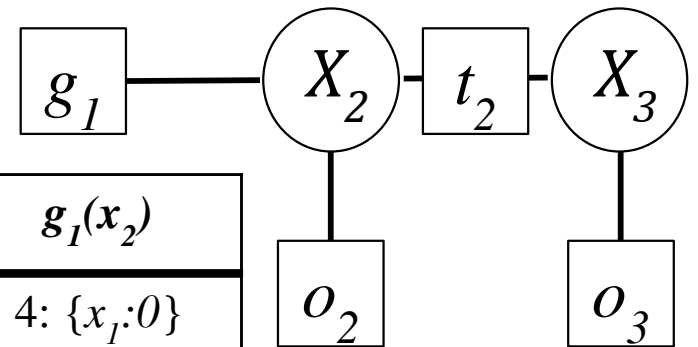
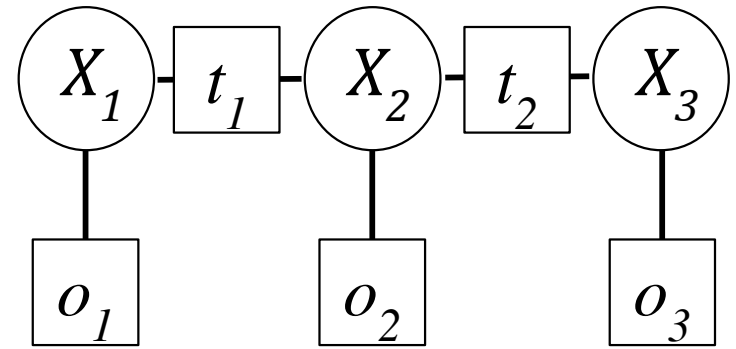


x_2	x_1	$o_1(x_1)$	$t_1(x_1, x_2)$	$o_1(x_1) t_1(x_1, x_2)$	$g_1(x_2)$
0	0	2	2	4	
0	1	1	1	1	
0	2	0	0	0	
1	0	2	1	2	
1	1	1	2	2	
1	2	0	1	0	
2	0	2	0	0	
2	1	1	1	1	
2	2	0	2	0	

- Eliminate X_1
- Factors that depend on X_1 :
 - o_1, t_1

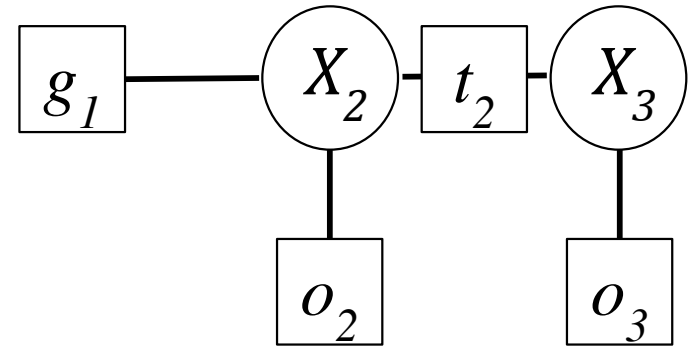
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_1(x_2) = \max_{x_1 \in \{0,1,2\}} o_1(x_1) \cdot t_1(x_1, x_2)$

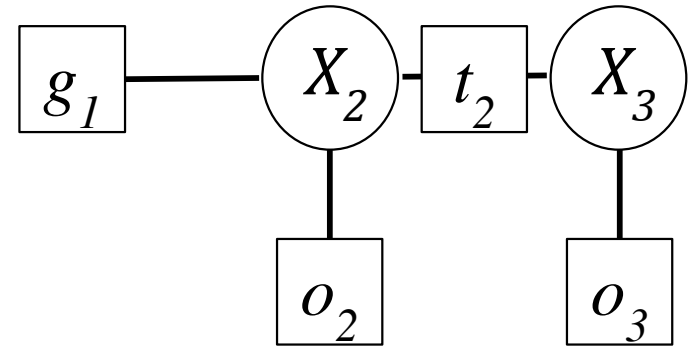


x_2	x_1	$o_1(x_1)$	$t_1(x_1, x_2)$	$o_1(x_1) t_1(x_1, x_2)$	$g_1(x_2)$
0	0	2	2	4	4: $\{x_1: 0\}$
0	1	1	1	1	
0	2	0	0	0	
1	0	2	1	2	2: $\{x_1: 1\}$
1	1	1	2	2	
1	2	0	1	0	
2	0	2	0	0	1: $\{x_1: 1\}$
2	1	1	1	1	
2	2	0	2	0	

- Eliminate X_2



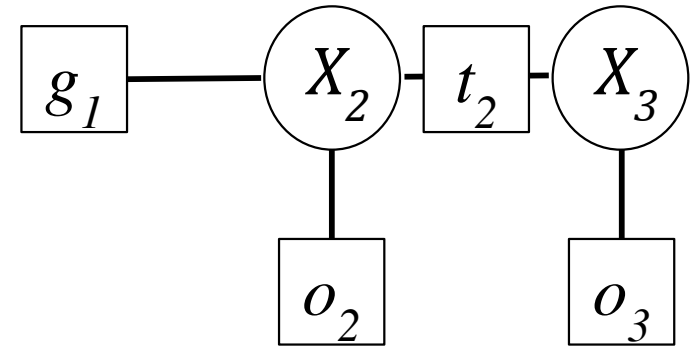
- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



- Eliminate X_2
- Factors that depend on X_2 :

- o_2, t_2, g_1

- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

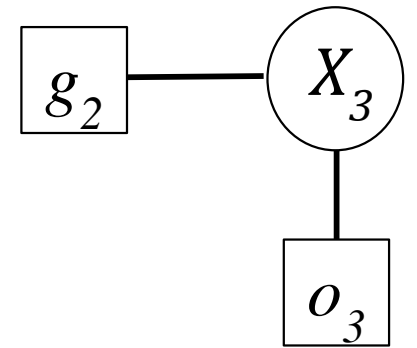
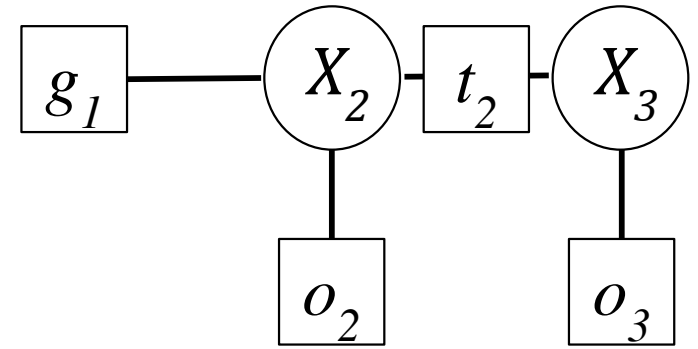


- Eliminate X_2
- Factors that depend on X_2 :

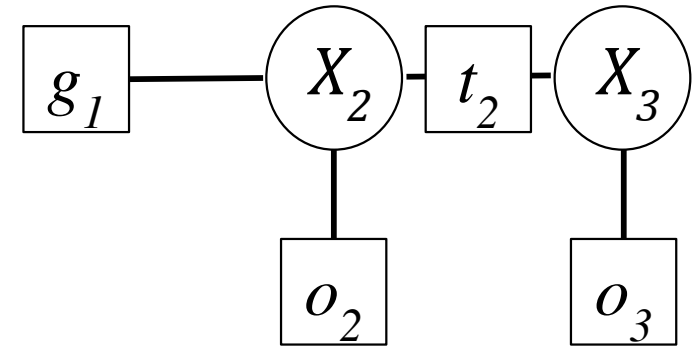
- o_2, t_2, g_1

- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$

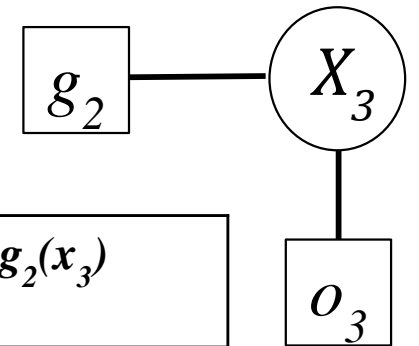


- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



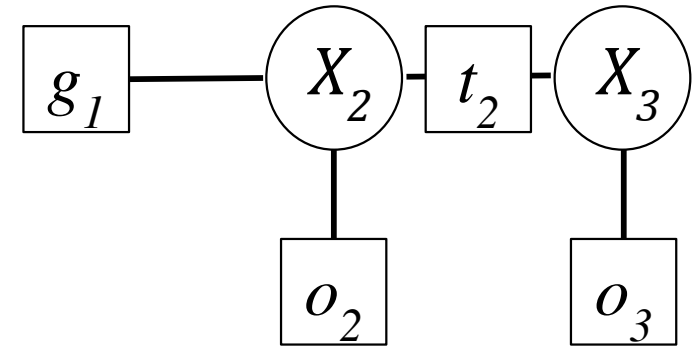
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



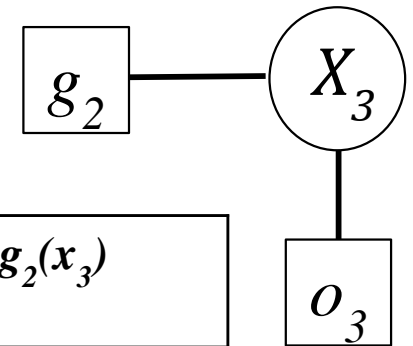
x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0					
0	1					
0	2					
1	0					
1	1					
1	2					
2	0					
2	1					
2	2					

- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



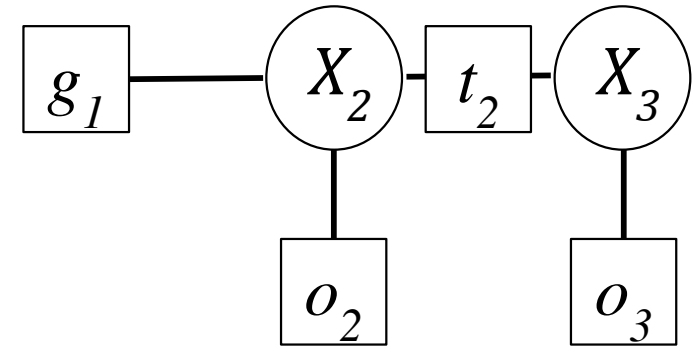
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



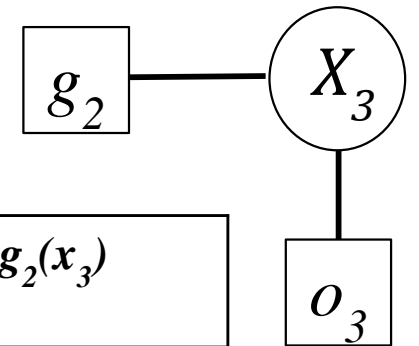
x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0	4: $\{x_1: 0\}$				
0	1	2: $\{x_1: 1\}$				
0	2	1: $\{x_1: 1\}$				
1	0	4: $\{x_1: 0\}$				
1	1	2: $\{x_1: 1\}$				
1	2	1: $\{x_1: 1\}$				
2	0	4: $\{x_1: 0\}$				
2	1	2: $\{x_1: 1\}$				
2	2	1: $\{x_1: 1\}$				

- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



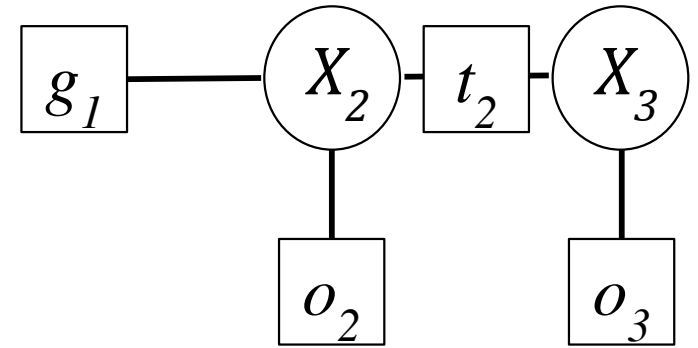
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



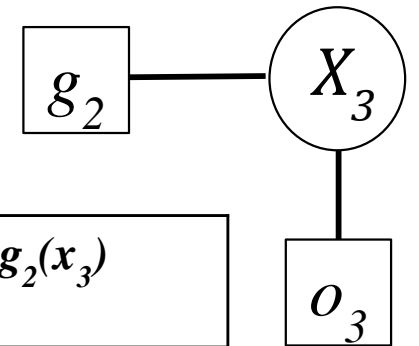
x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0	4: $\{x_1: 0\}$	0			
0	1	2: $\{x_1: 1\}$	1			
0	2	1: $\{x_1: 1\}$	2			
1	0	4: $\{x_1: 0\}$	0			
1	1	2: $\{x_1: 1\}$	1			
1	2	1: $\{x_1: 1\}$	2			
2	0	4: $\{x_1: 0\}$	0			
2	1	2: $\{x_1: 1\}$	1			
2	2	1: $\{x_1: 1\}$	2			

- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



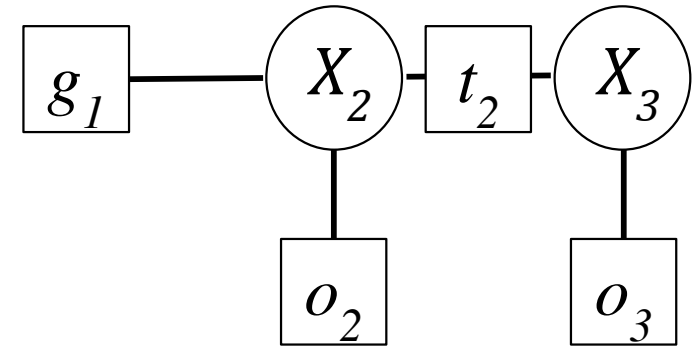
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



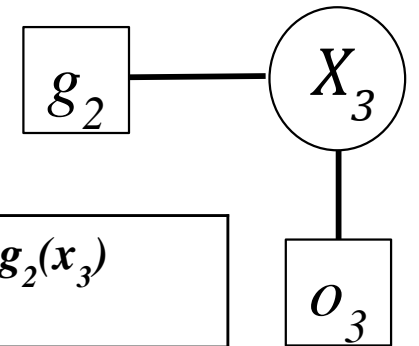
x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0	4: $\{x_1: 0\}$	0	2		
0	1	2: $\{x_1: 1\}$	1	1		
0	2	1: $\{x_1: 1\}$	2	0		
1	0	4: $\{x_1: 0\}$	0	1		
1	1	2: $\{x_1: 1\}$	1	2		
1	2	1: $\{x_1: 1\}$	2	1		
2	0	4: $\{x_1: 0\}$	0	0		
2	1	2: $\{x_1: 1\}$	1	1		
2	2	1: $\{x_1: 1\}$	2	2		

- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



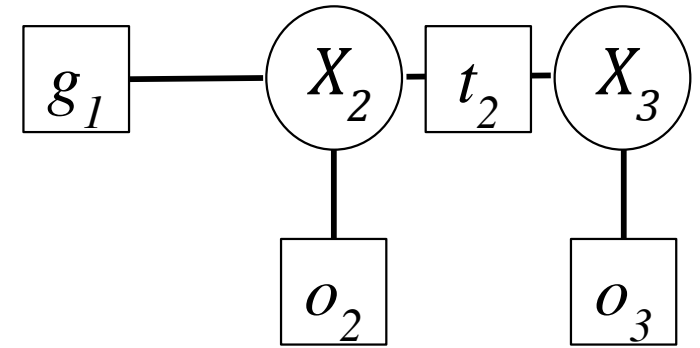
- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



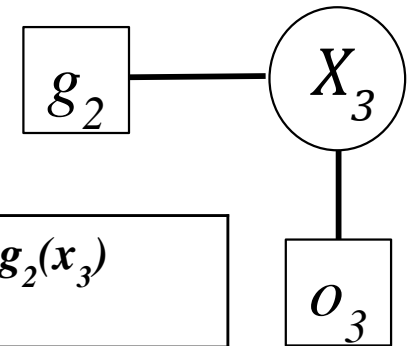
x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0	4: $\{x_1: 0\}$	0	2	0	
0	1	2: $\{x_1: 1\}$	1	1	2	
0	2	1: $\{x_1: 1\}$	2	0	2	
1	0	4: $\{x_1: 0\}$	0	1	4	
1	1	2: $\{x_1: 1\}$	1	2	4	
1	2	1: $\{x_1: 1\}$	2	1	2	
2	0	4: $\{x_1: 0\}$	0	0	0	
2	1	2: $\{x_1: 1\}$	1	1	2	
2	2	1: $\{x_1: 1\}$	2	2	4	

- Eliminate X_2
- Factors that depend on X_2 :
 - o_2, t_2, g_1



- Add $f_{new}(x) = \max_{x_i} \prod_{j=1}^k f_j(x)$

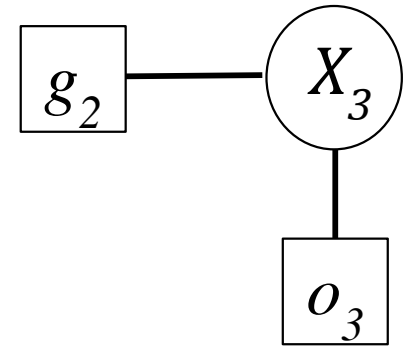
- $g_2(x_3) = \max_{x_2 \in \{0,1,2\}} g_1(x_2) \cdot o_2(x_2) \cdot t_2(x_2, x_3)$



x_3	x_2	$g_1(x_2)$	$o_2(x_2)$	$t_2(x_2, x_3)$	$g_1(x_2) o_2(x_2) t_2(x_2, x_3)$	$g_2(x_3)$
0	0	4: $\{x_1: 0\}$	0	2	0	2: $\{x_1: 1, x_2: 2\}$
0	1	2: $\{x_1: 1\}$	1	1	2	
0	2	1: $\{x_1: 1\}$	2	0	2	
1	0	4: $\{x_1: 0\}$	0	1	4	4: $\{x_1: 1, x_2: 1\}$
1	1	2: $\{x_1: 1\}$	1	2	4	
1	2	1: $\{x_1: 1\}$	2	1	2	
2	0	4: $\{x_1: 0\}$	0	0	0	4: $\{x_1: 1, x_2: 2\}$
2	1	2: $\{x_1: 1\}$	1	1	2	
2	2	1: $\{x_1: 1\}$	2	2	4	

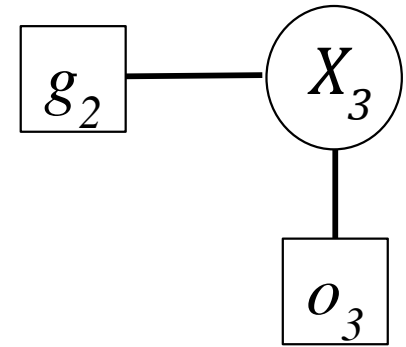
- We are left with:

$$\max_{x_3 \in \{0,1,2\}} g_2(x_3) \cdot o_3(x_3)$$



- We are left with:

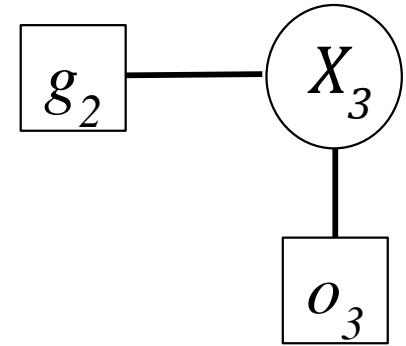
$$\max_{x_3 \in \{0,1,2\}} g_2(x_3) \cdot o_3(x_3)$$



x_3	$g_2(x_3)$	$o_3(x_3)$	$g_2(x_3) o_3(x_3)$	<i>Optimal Weight</i>
0				
1				
2				

- We are left with:

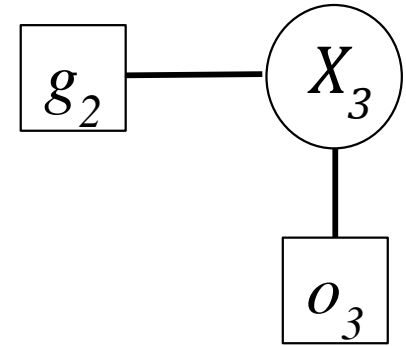
$$\max_{x_3 \in \{0,1,2\}} g_2(x_3) \cdot o_3(x_3)$$



x_3	$g_2(x_3)$	$o_3(x_3)$	$g_2(x_3) o_3(x_3)$	<i>Optimal Weight</i>
0	2: $\{x_1: 1, x_2: 2\}$	0		
1	4: $\{x_1: 1, x_2: 1\}$	1		
2	4: $\{x_1: 1, x_2: 2\}$	2		

- We are left with:

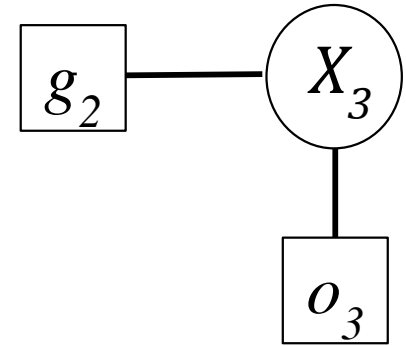
$$\max_{x_3 \in \{0,1,2\}} g_2(x_3) \cdot o_3(x_3)$$



x_3	$g_2(x_3)$	$o_3(x_3)$	$g_2(x_3) o_3(x_3)$	<i>Optimal Weight</i>
0	2: $\{x_1: 1, x_2: 2\}$	0	2	
1	4: $\{x_1: 1, x_2: 1\}$	1	4	
2	4: $\{x_1: 1, x_2: 2\}$	2	8	

- We are left with:

$$\max_{x_3 \in \{0,1,2\}} g_2(x_3) \cdot o_3(x_3)$$



x_3	$g_2(x_3)$	$o_3(x_3)$	$g_2(x_3) o_3(x_3)$	<i>Optimal Weight</i>
0	2: $\{x_1: 1, x_2: 2\}$	0	2	8: $\{x_1: 1, x_2: 2, x_3: 2\}$
1	4: $\{x_1: 1, x_2: 1\}$	1	4	
2	4: $\{x_1: 1, x_2: 2\}$	2	8	

