



Removing Implicit Places Using Regions for Process Discovery



Seminar - Selected Topics in Process Mining

Faizan Zafar

Chair of Process and Data Science

RWTH Aachen University

`faizan.zafar@rwth-aachen.de`

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Outline

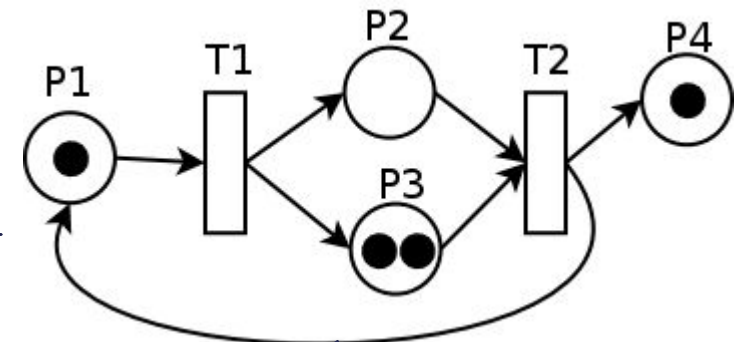
- **Introduction**
- **Motivation**
- **Preliminaries**
- **Main Idea**
- **Running Example**
- **Limitations**
- **Application**
- **Evaluation**
- **Conclusion**
- **Future Work**
- **References**

Input: Event Log



Process Discovery

Output: Process model

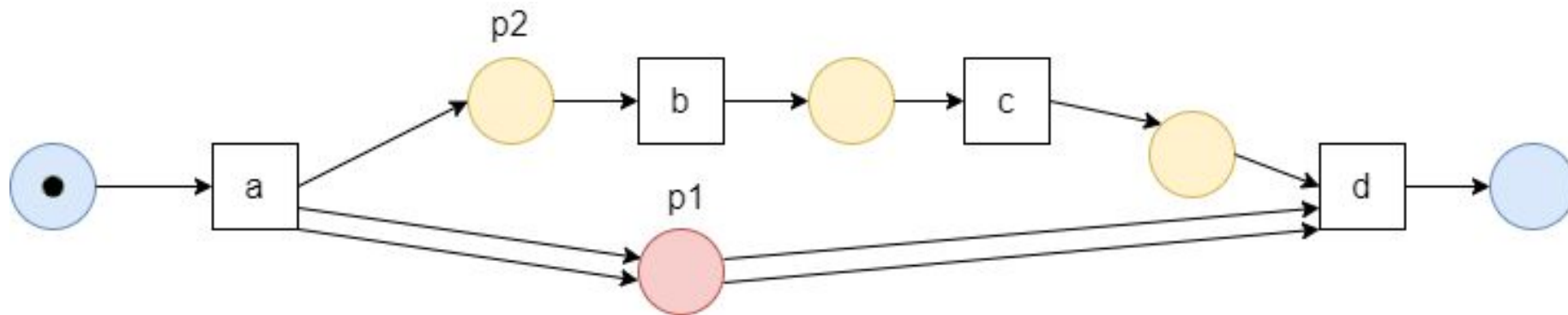


- Process discovery aims to discover simple Petri nets
- Petri nets may contain several *implicit* places
- Implicit places increase model complexity
- A novel technique to identify and remove implicit places
- Integrating the approach with the eST-Miner

- An activity is simply *work* performed in a business process
- A trace is a sequence of activities
- An event log is a multiset of traces

$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$

- A marked Petri net (P, T, F, m_0) is a process model composed of places P , transitions T and a multiset of arcs $F \rightarrow \mathbb{N}_0$
- An implicit place p does not contribute to a Petri net's behaviour



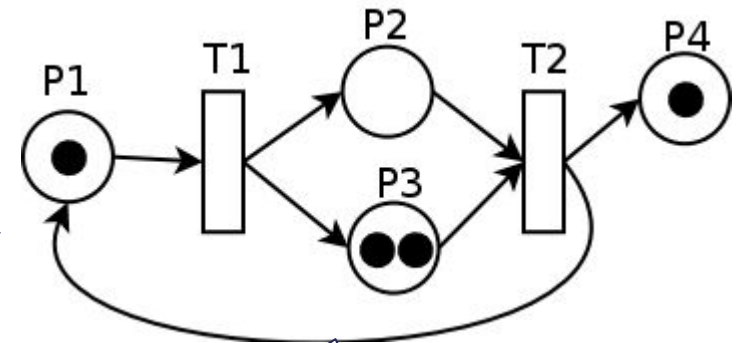
$\langle a \rangle$
$\langle a, b \rangle$
$\langle a, b, c \rangle$
$\langle a, b, c, d \rangle$

The language L of a Petri net is every possible firing sequence in the Petri net

	ϵ	a	b	c	d
x_{p_2}	0	1	0	0	0

The token count function x_{p_i} defines the number of tokens in place p_i

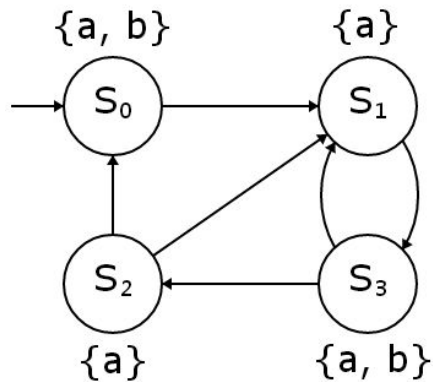
Preliminaries: Region Theory



Preliminaries: Region Theory

- Identifying *regions* in the event log
- Each region corresponds to a feasible place in the resulting Petri net

State-based region theory makes
use of a *transition system*



Language-based region theory makes
use of a *prefix-closed language*

$\langle a \rangle$
$\langle a, b \rangle$
$\langle a, b, c \rangle$
$\langle a, b, c, d \rangle$

- **Given an event log and the corresponding Petri net N :**
 - $p_1, p_2 \in P$ are two individual places in N ,
 - **Comparing p_1 and p_2 , we verify the conditions:**

$$\forall \sigma \in L(N) : x_{p_1}(\sigma) \geq x_{p_2}(\sigma) \quad (1)$$

$$\exists \sigma \in L(N) : x_{p_1}(\sigma) > x_{p_2}(\sigma) \quad (2)$$

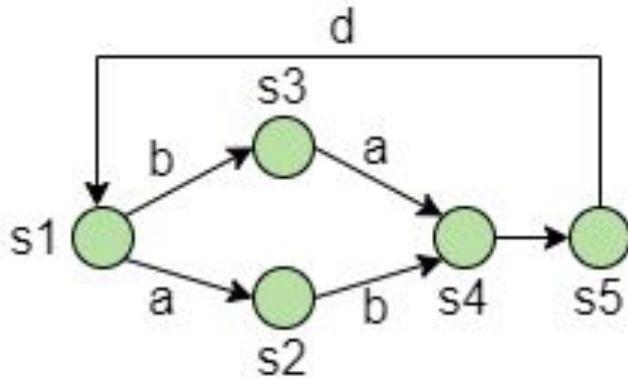
- If $x_{p_1} > x_{p_2}$ for the prefix-closed language of N , then

$$x_{p_3} = x_{p_1} - x_{p_2} \quad (3)$$

is a region and p_1 can be regarded as *implicit* if p_3 exists

- Incoming arc added for increase in the token count of place p_3
- Outgoing arc added for decrease in the token count of place p_3

- Minimal regions method avoids adding implicit places to a model



$$r_1 = \{s1, s2\}$$

$$r_2 = \{s1, s3\}$$

$$r_3 = \{s2, s4\}$$

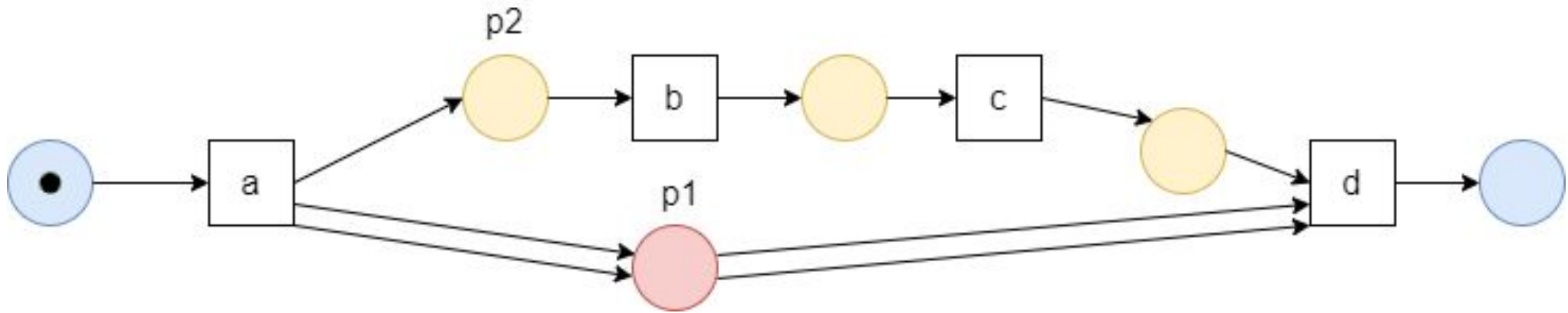
$$r_4 = \{s3, s4\}$$

$$r_5 = \{s5\}$$

- Transitioning directly from the event log to the resulting model
- The goal is to remove implicit places from a discovered set of places

Running Example: Simple Case

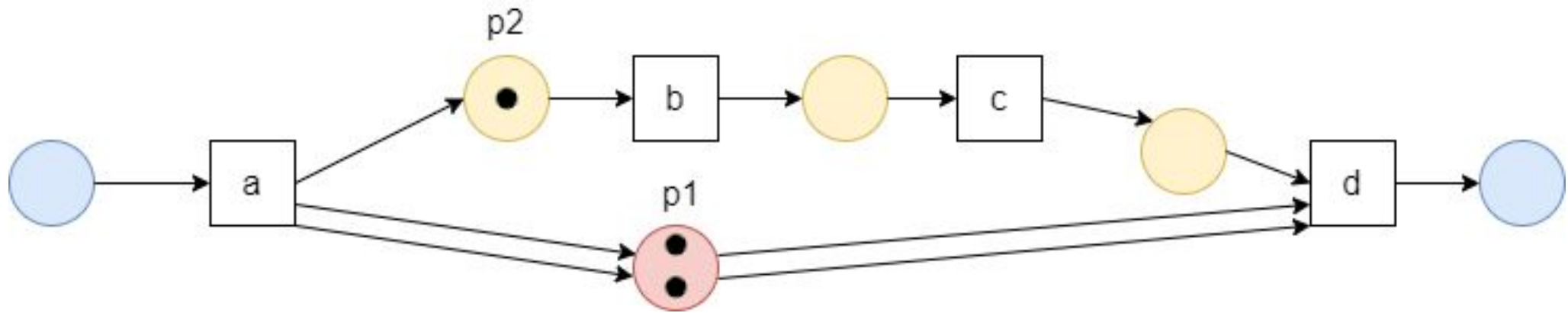
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0				
x_{p_2}	0				
x_{p_3}					

Running Example: Simple Case

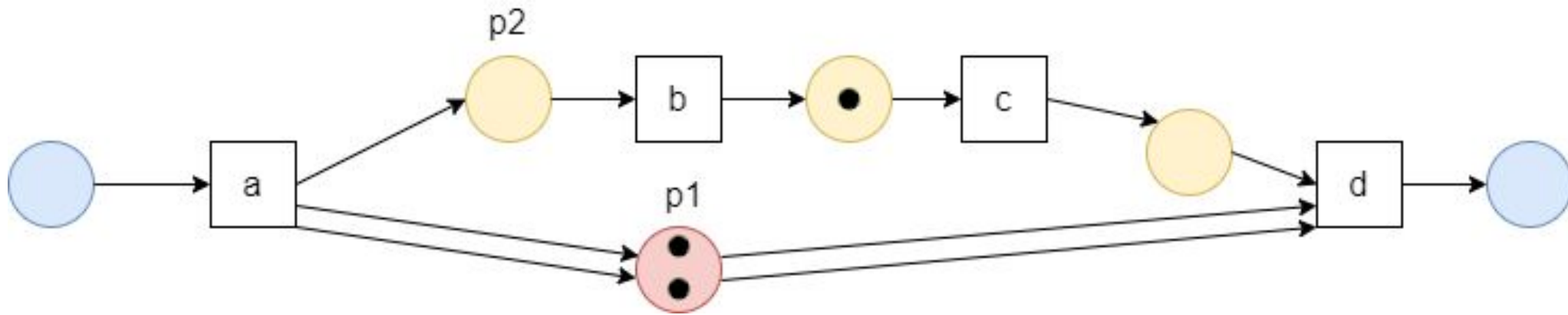
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0	2			
x_{p_2}	0	1			
x_{p_3}					

Running Example: Simple Case

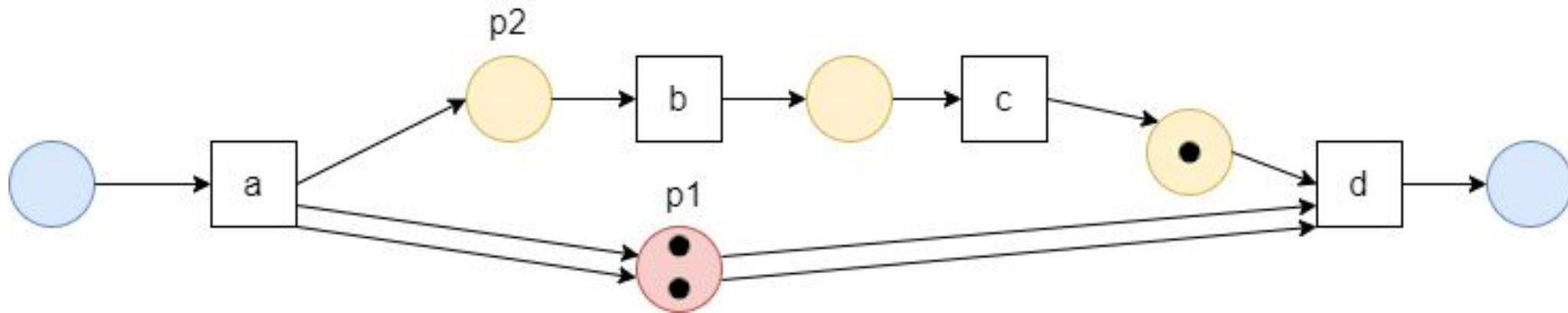
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0	2	2		
x_{p_2}	0	1	0		
x_{p_3}					

Running Example: Simple Case

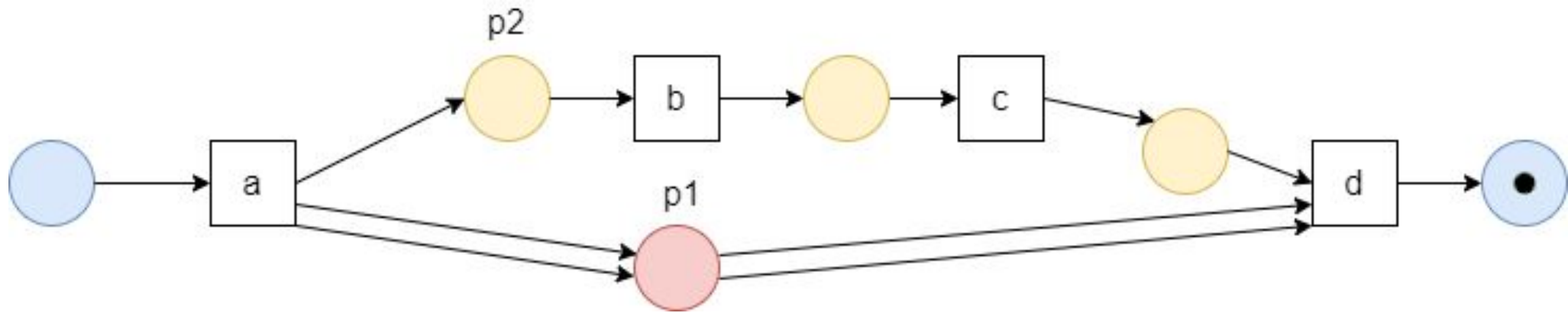
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0	2	2	2	
x_{p_2}	0	1	0	0	
x_{p_3}					

Running Example: Simple Case

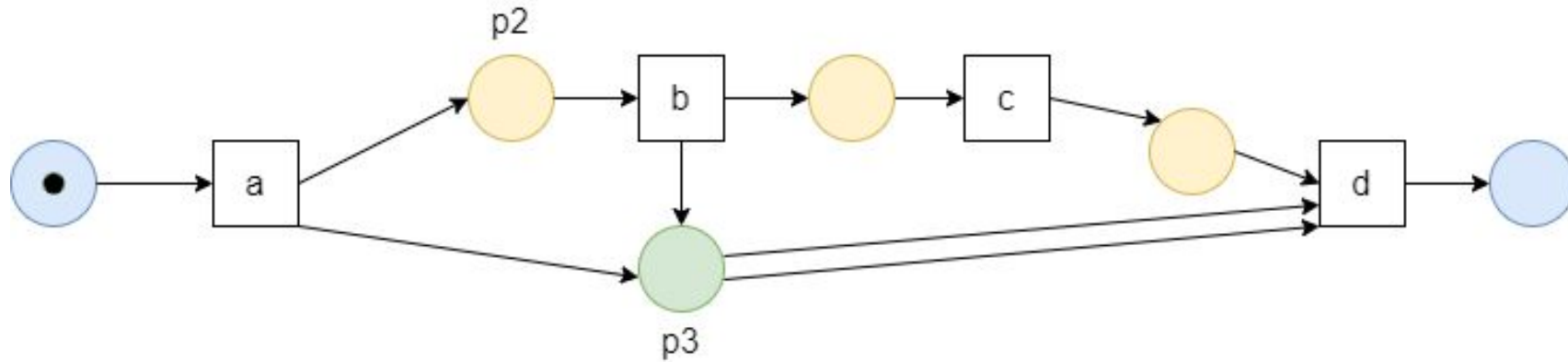
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0	2	2	2	0
x_{p_2}	0	1	0	0	0
x_{p_3}					

Running Example: Simple Case

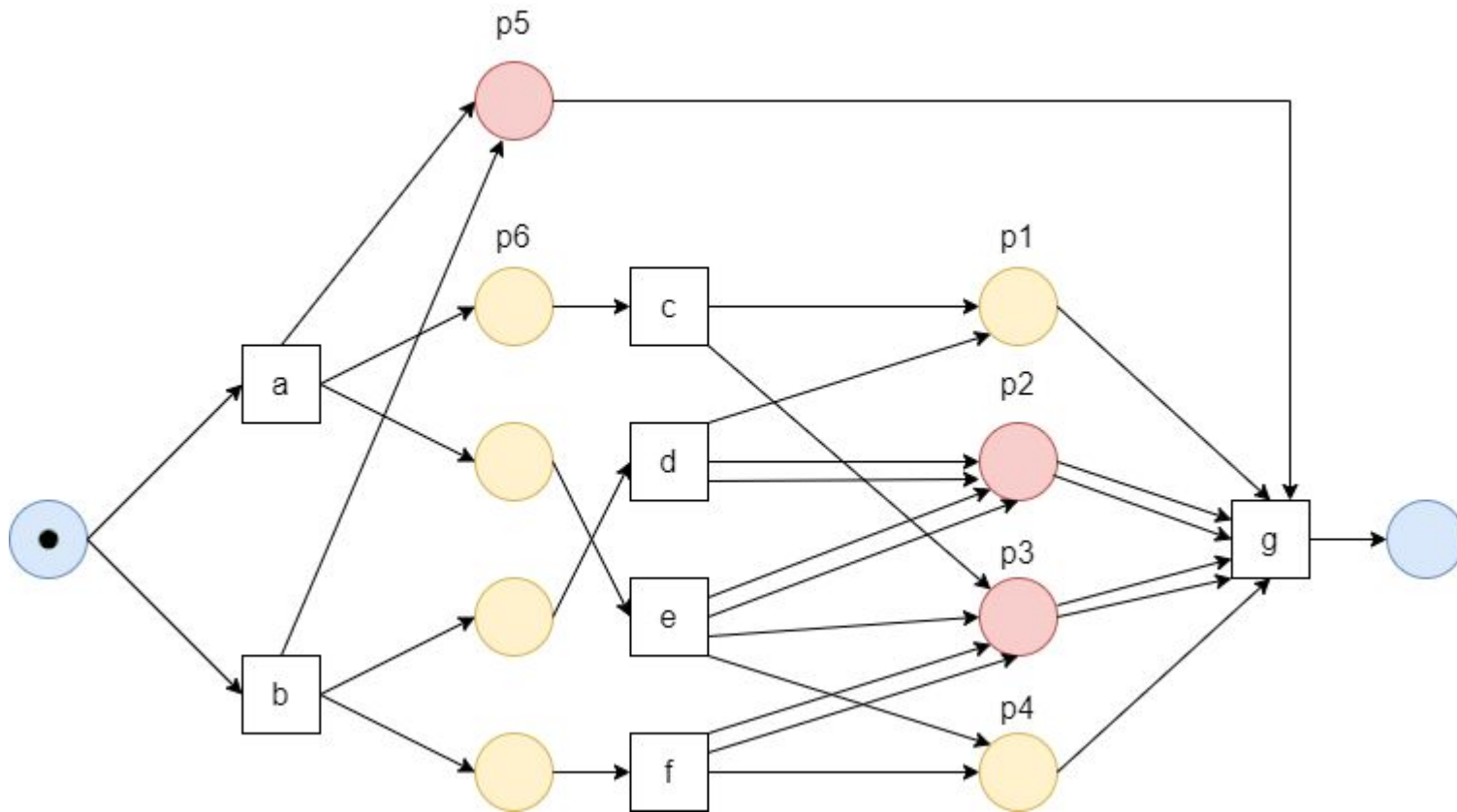
$$L = [\epsilon, \langle a, b, c, d \rangle^4]$$



	ϵ	a	b	c	d
x_{p_1}	0	2	2	2	0
x_{p_2}	0	1	0	0	0
x_{p_3}	0	1	2	2	0

Running Example: Intermediate Case

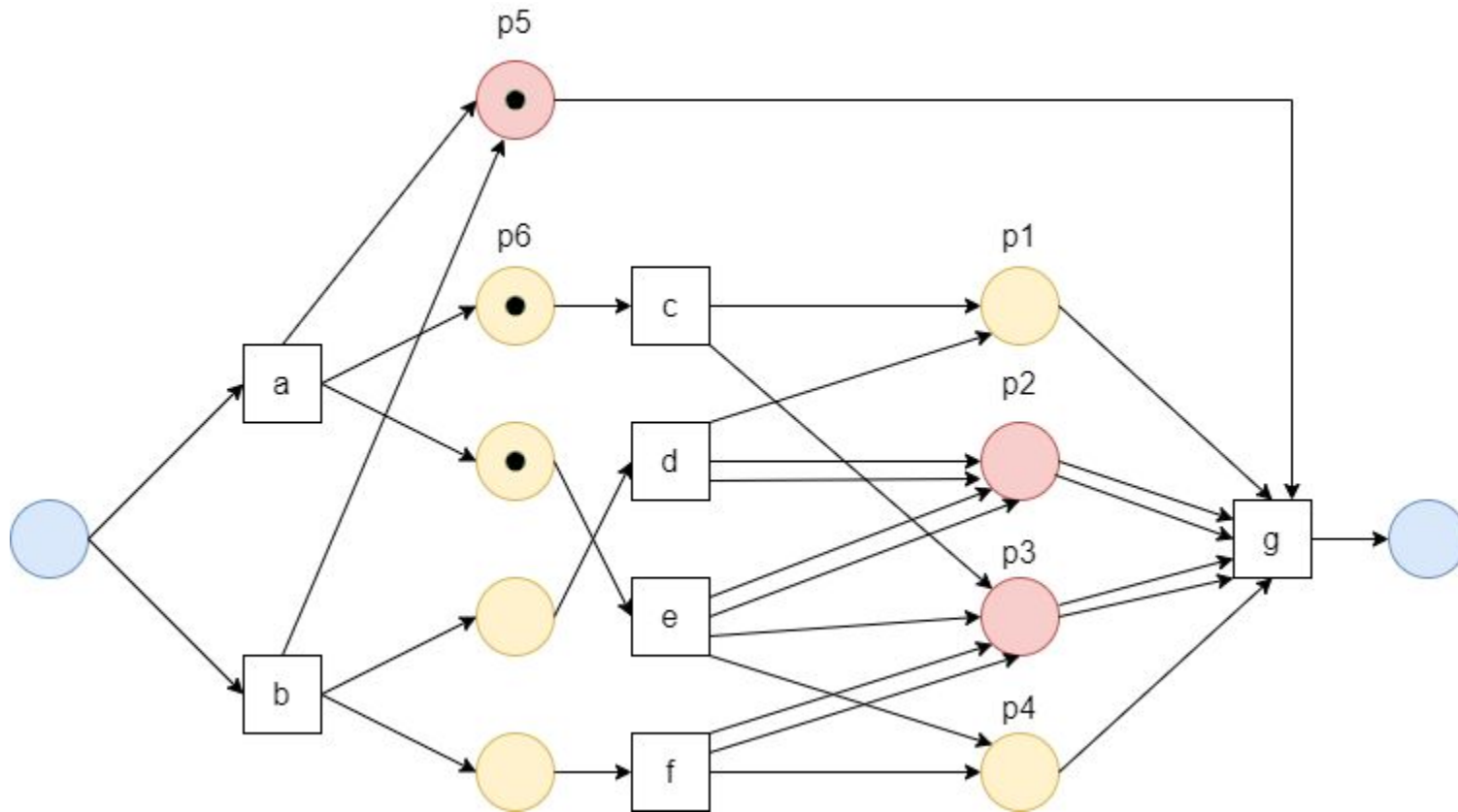
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0				
x_{p6}	0				
x_{p7}					

Running Example: Intermediate Case

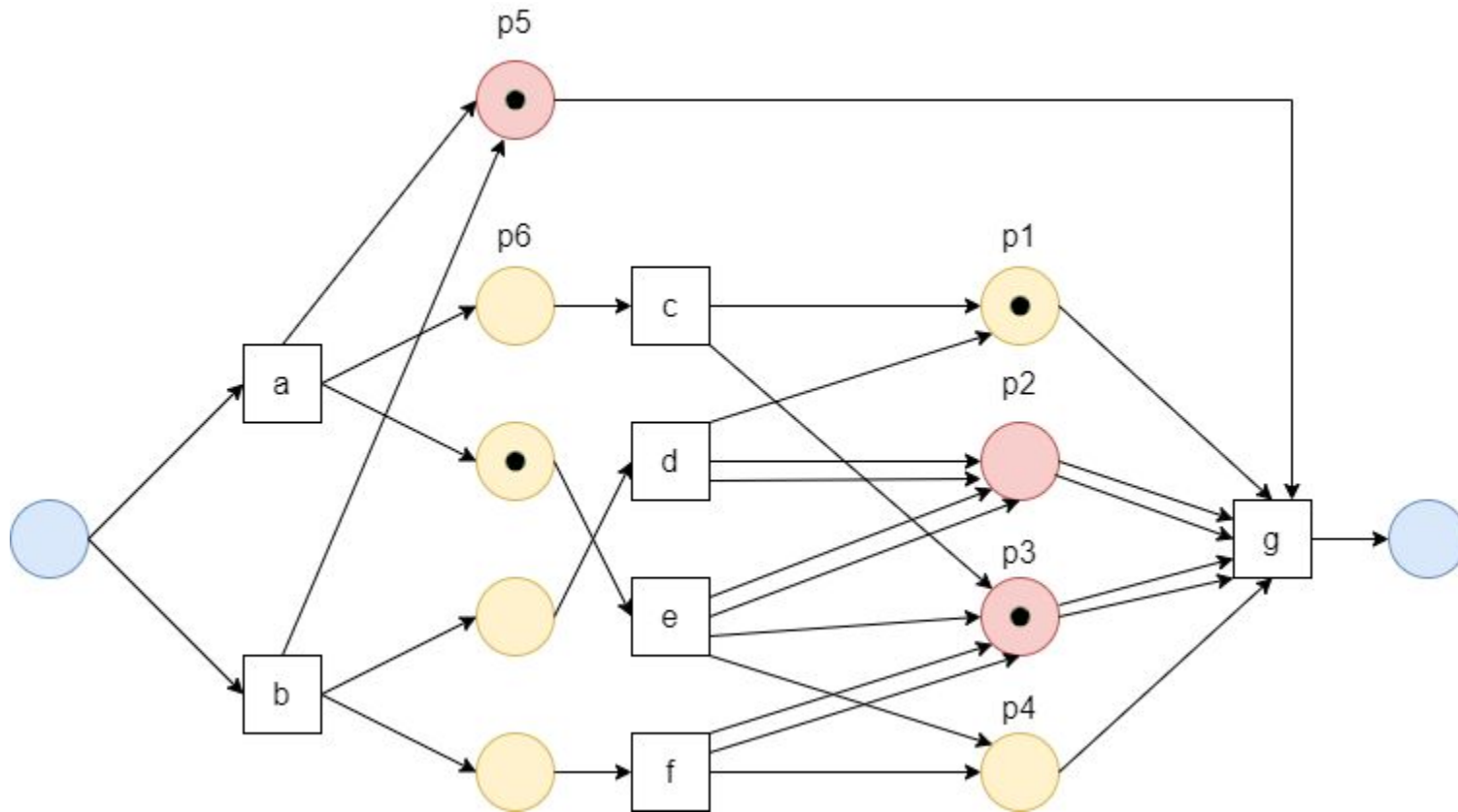
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0	1			
x_{p6}	0	1			
x_{p7}					

Running Example: Intermediate Case

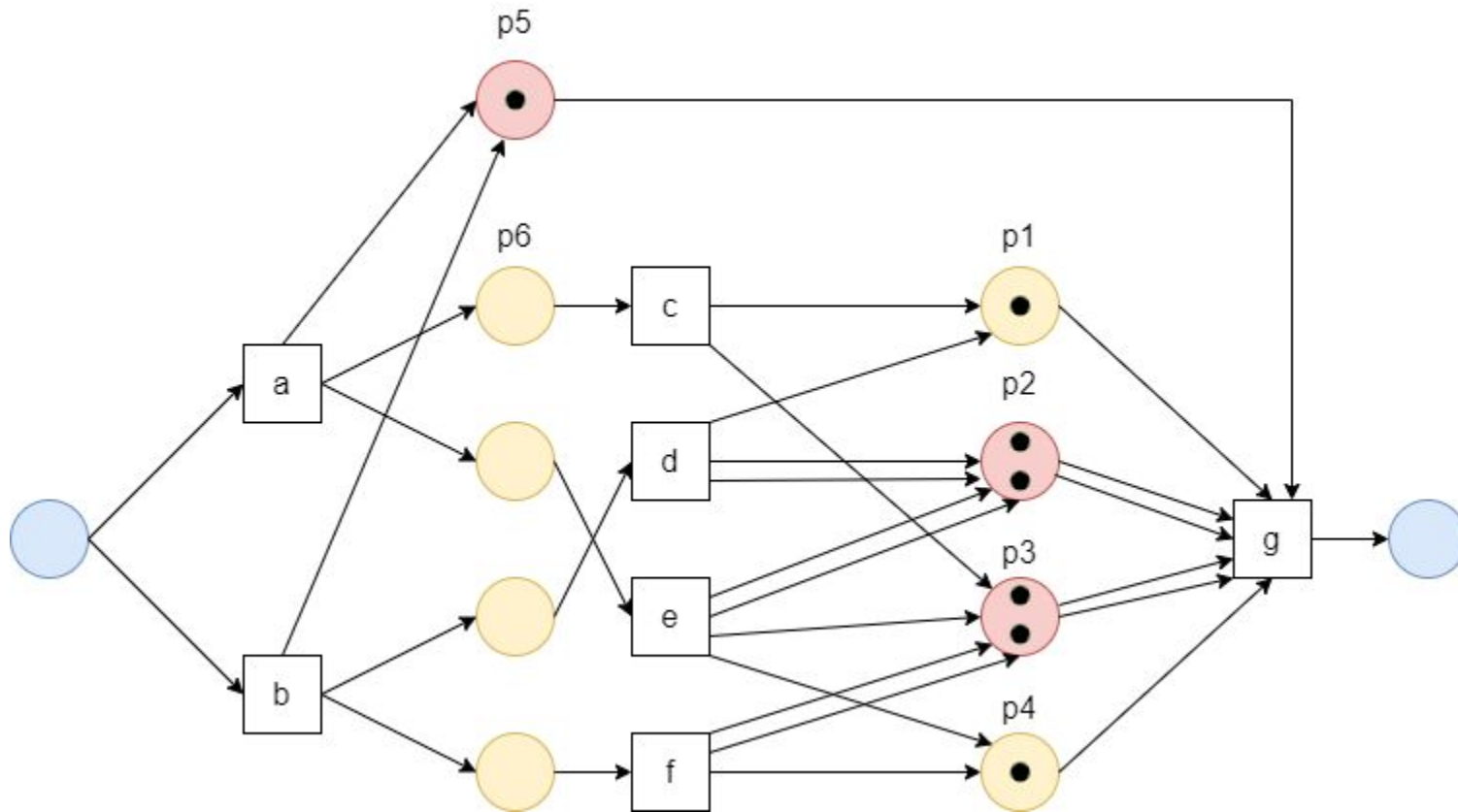
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0	1	1		
x_{p6}	0	1	0		
x_{p7}					

Running Example: Intermediate Case

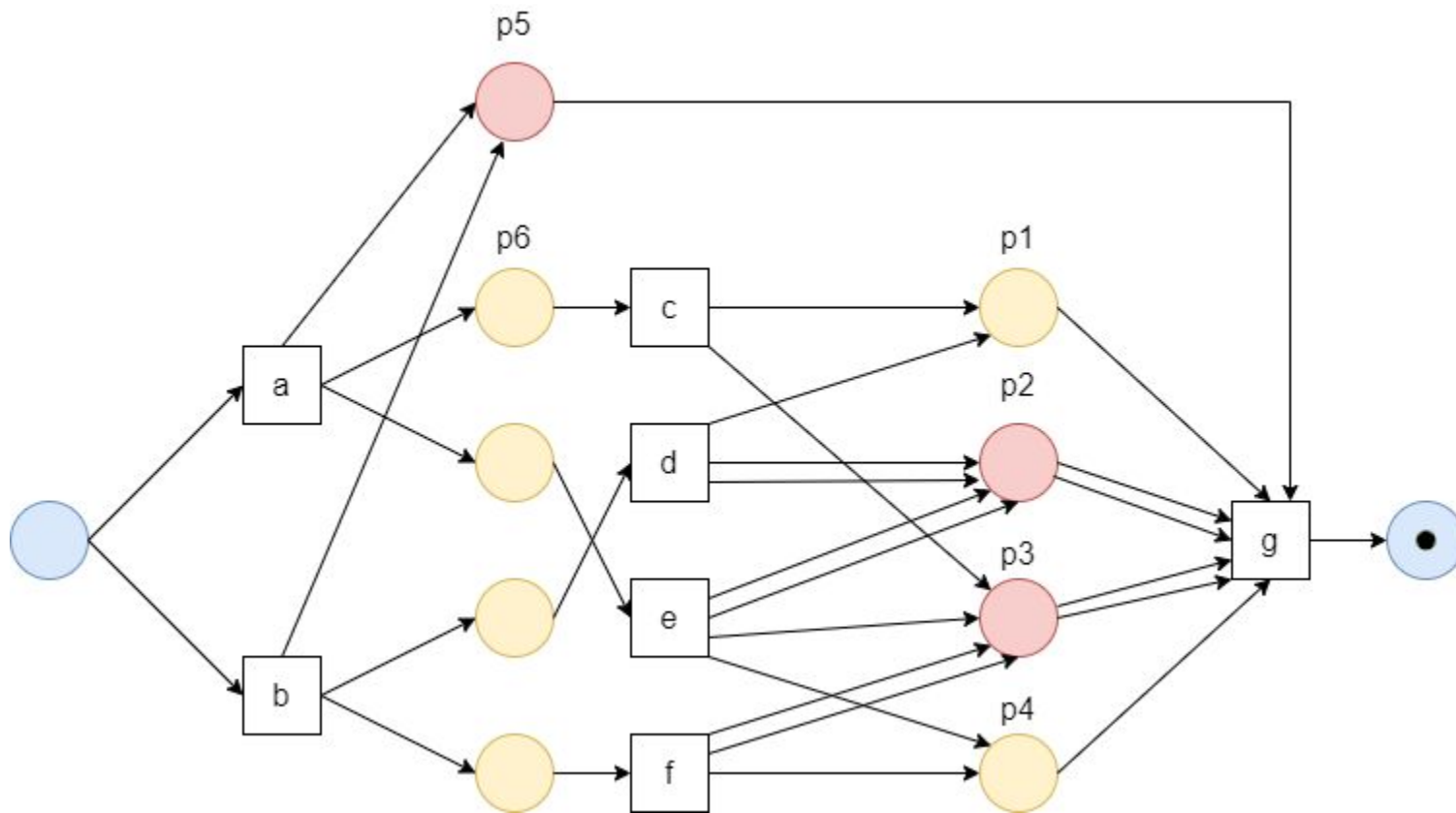
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0	1	1	1	
x_{p6}	0	1	0	0	
x_{p7}					

Running Example: Intermediate Case

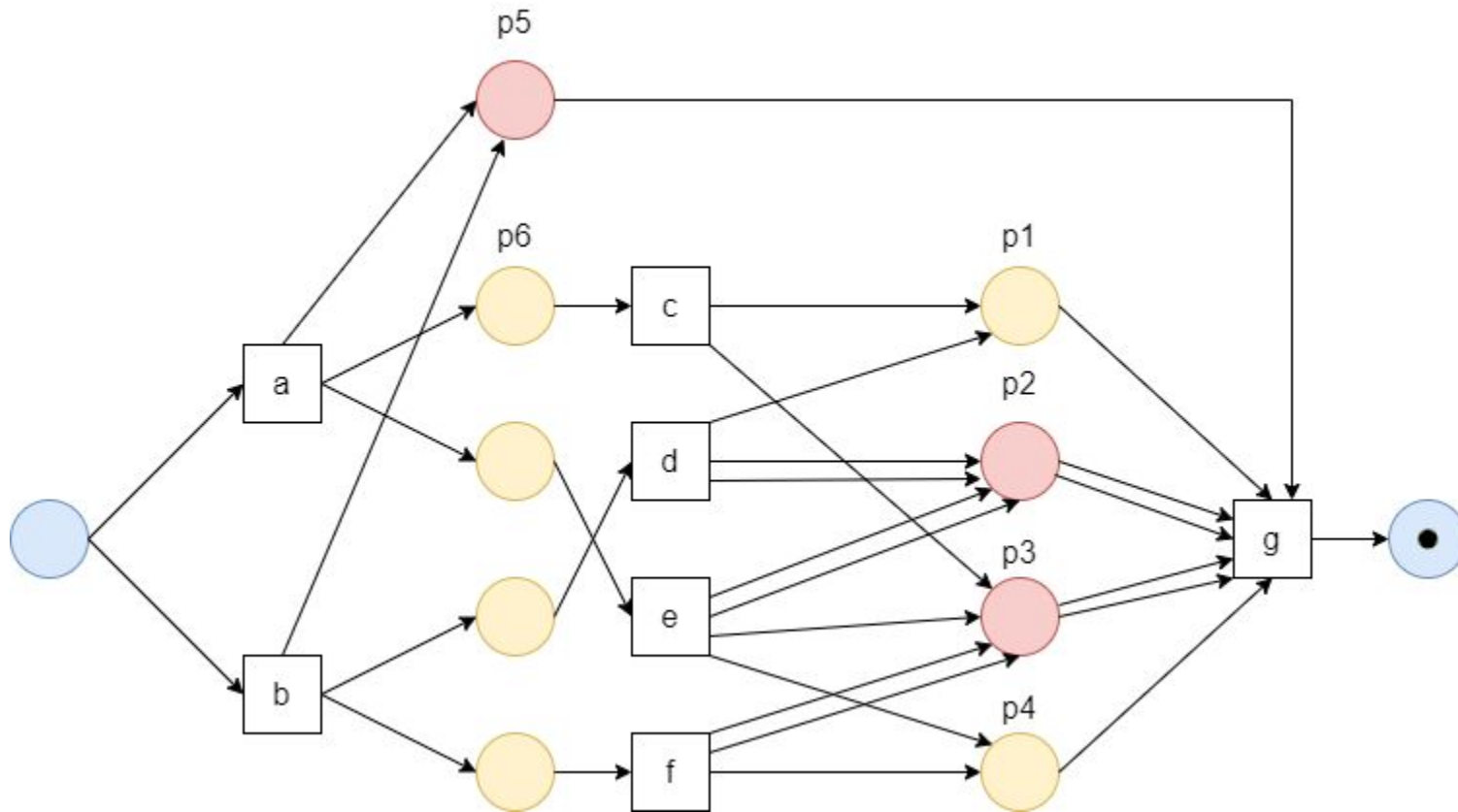
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0	1	1	1	0
x_{p6}	0	1	0	0	0
x_{p7}					

Running Example: Intermediate Case

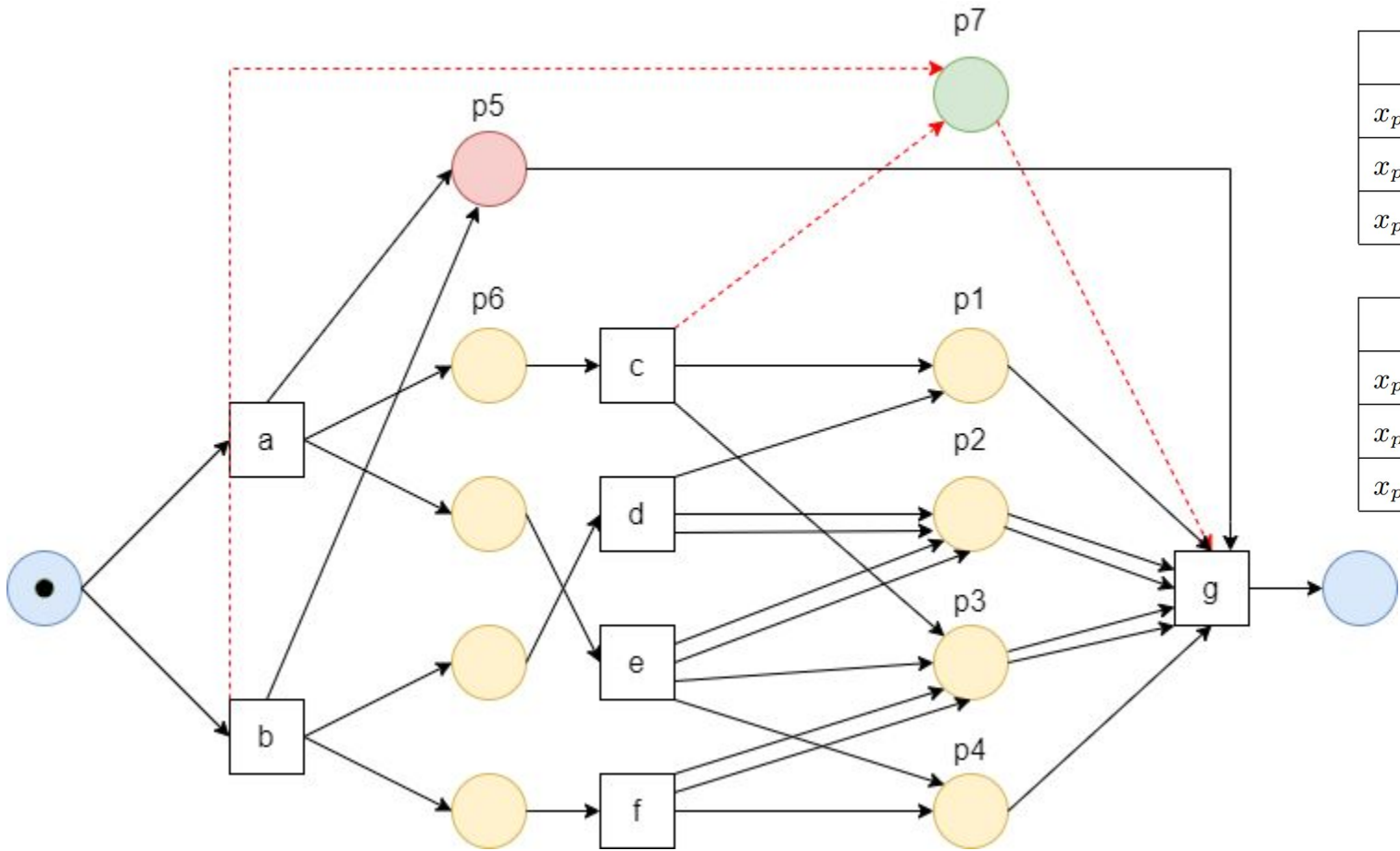
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p5}	0	1	1	1	0
x_{p6}	0	1	0	0	0
x_{p7}	0	0	1	1	0

Running Example: Intermediate Case

$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_5}	0	1	1	1	0
x_{p_6}	0	1	0	0	0
x_{p_7}	0	0	1	1	0

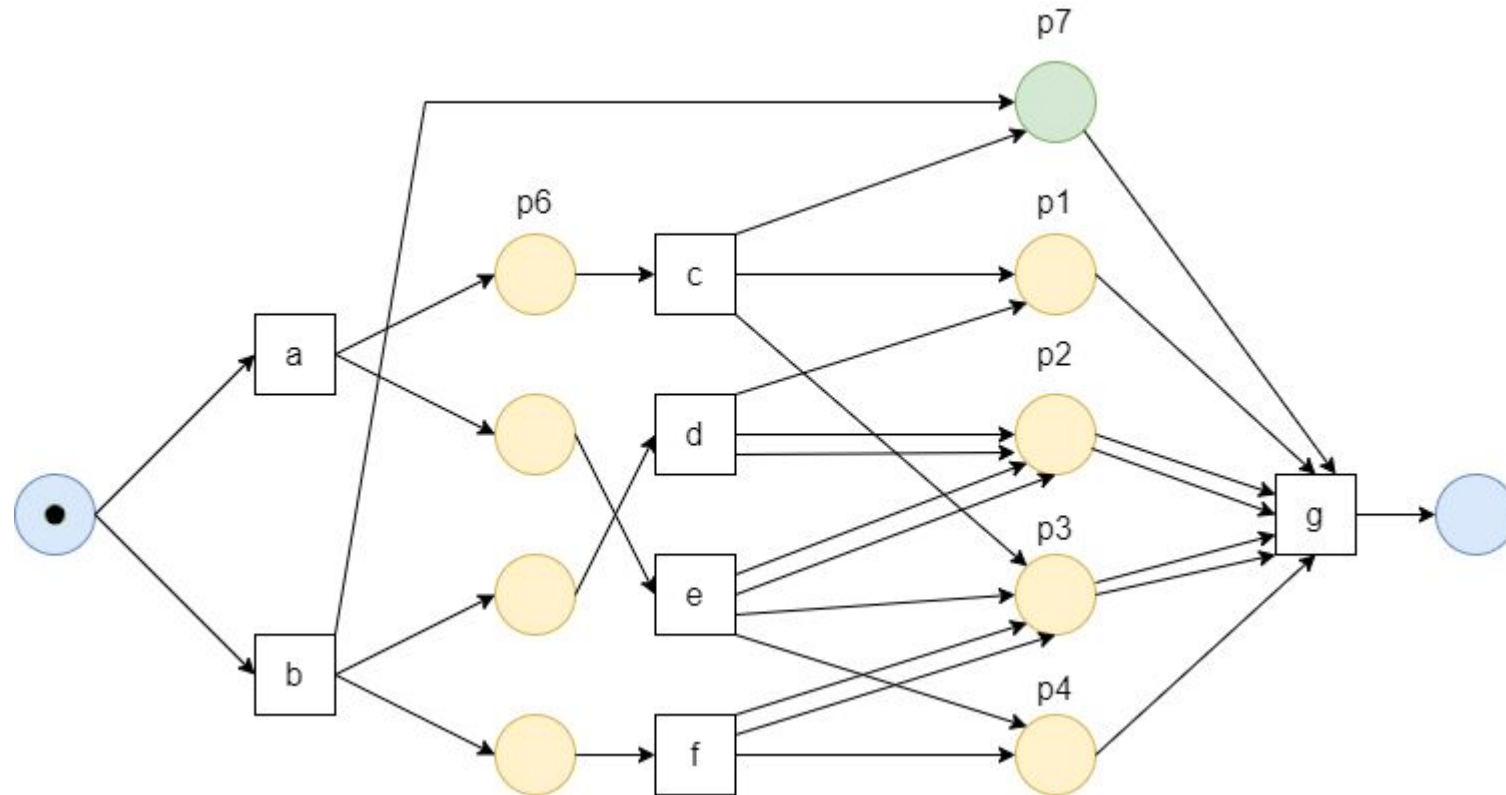
	ϵ	a	e	c	g
x_{p_5}	0	1	1	1	0
x_{p_6}	0	1	1	0	0
x_{p_7}	0	0	0	1	0

	ϵ	b	d	f	g
x_{p_5}	0	1	1	1	0
x_{p_6}	0	0	0	0	0
x_{p_7}	0	1	1	1	0

	ϵ	b	f	d	g
x_{p_5}	0	1	1	1	0
x_{p_6}	0	0	0	0	0
x_{p_7}	0	1	1	1	0

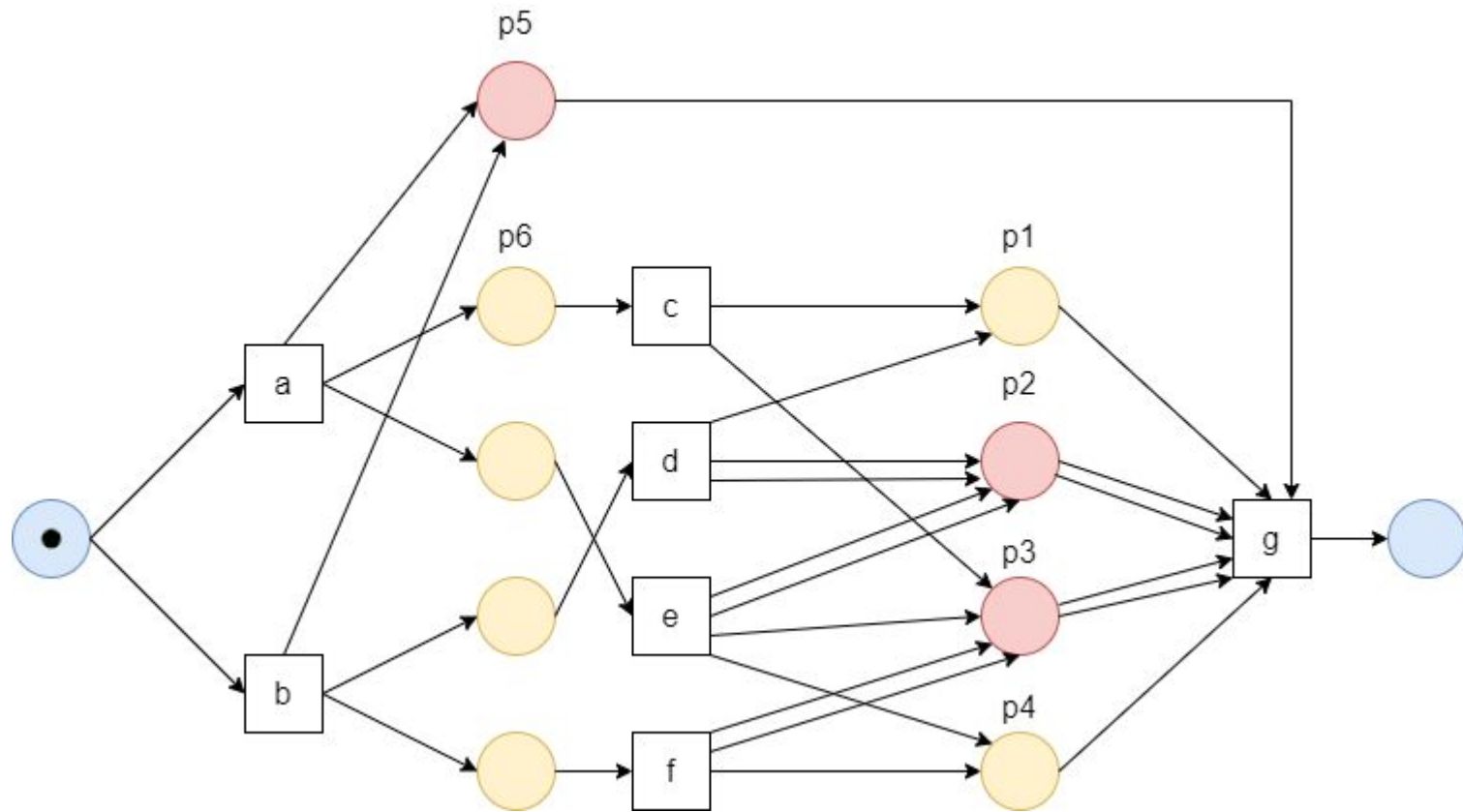
Running Example: Intermediate Case

$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



Running Example: Extended Case

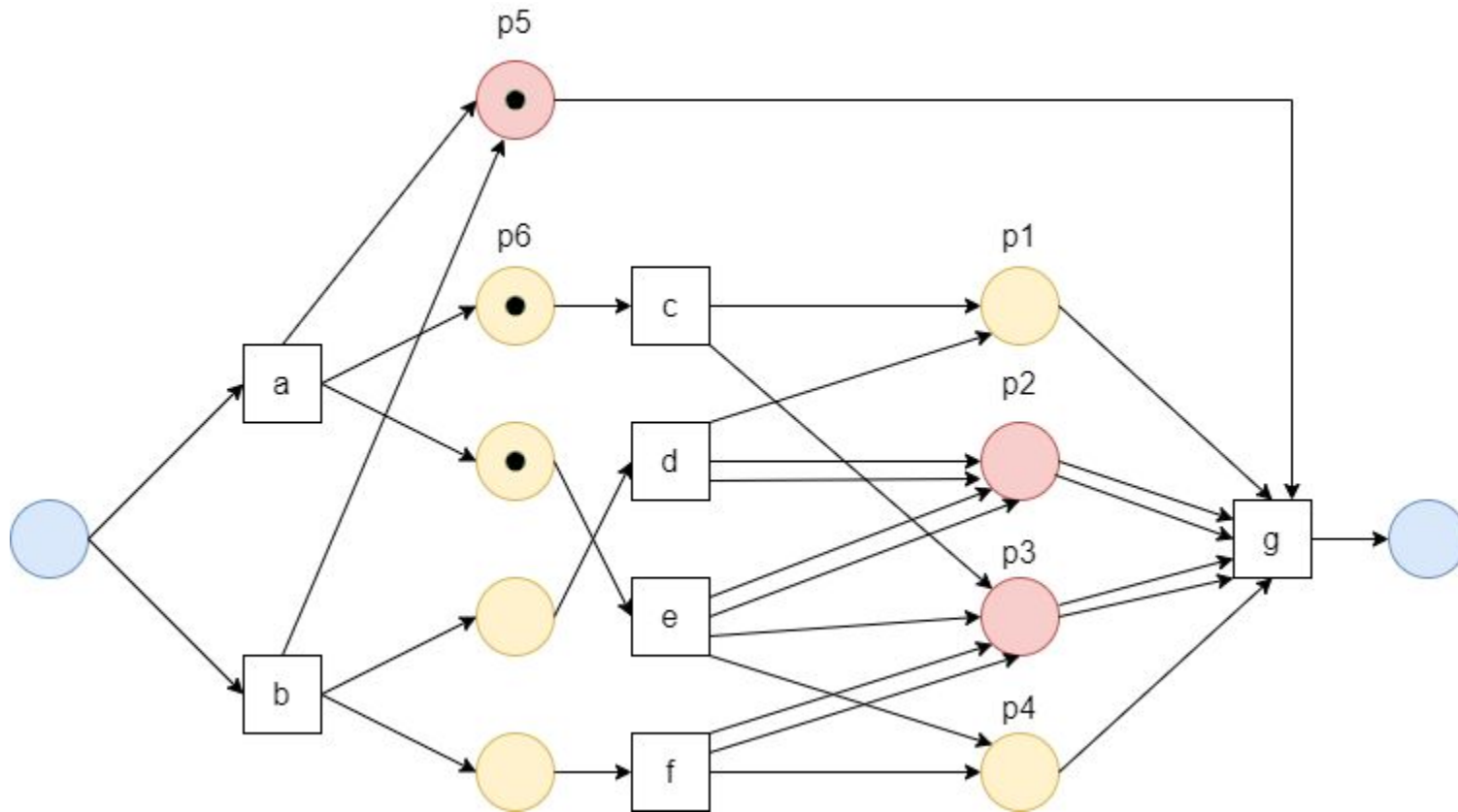
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p3}	0				
x_{p4}	0				
x_{p8}					

Running Example: Extended Case

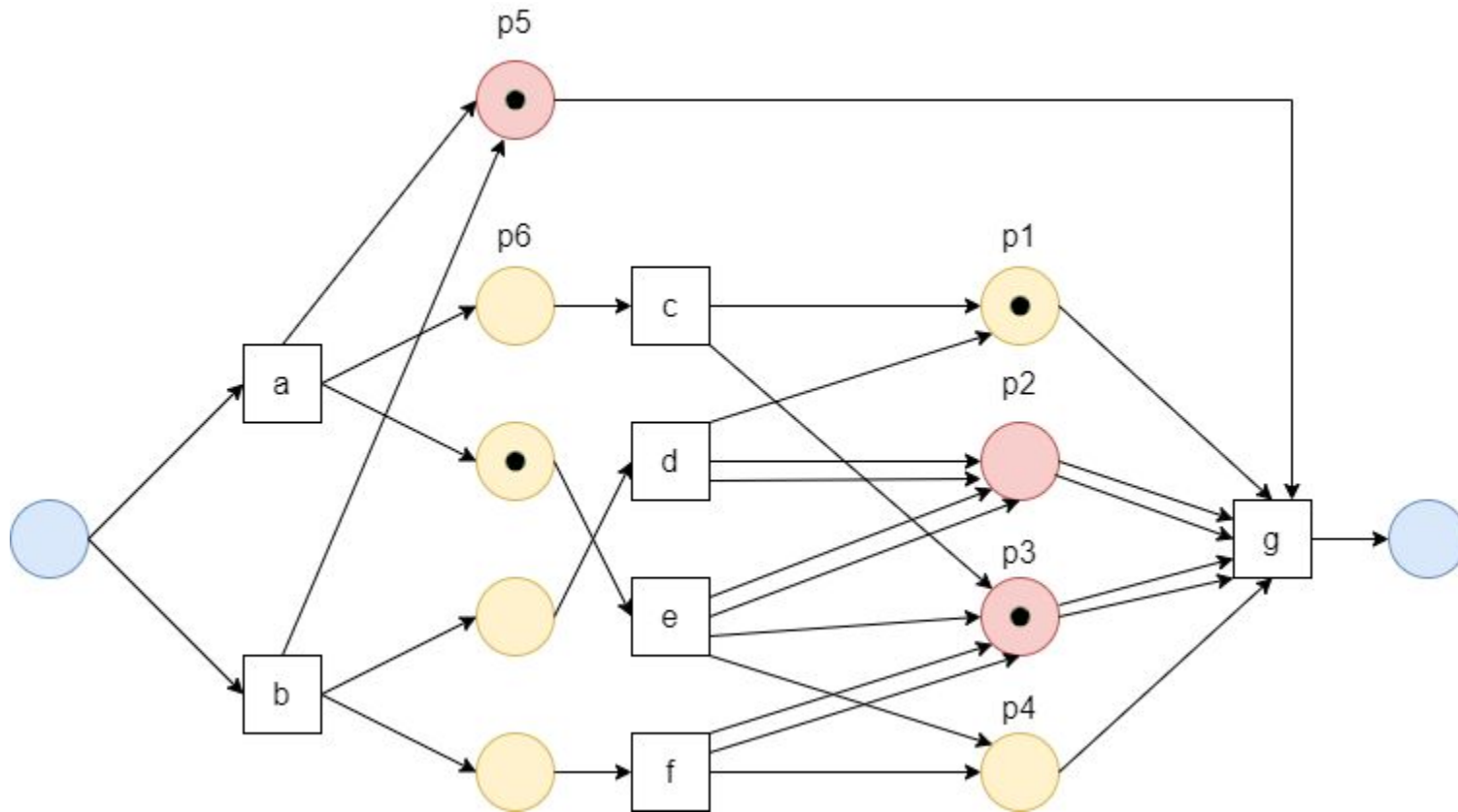
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	gg
x_{p3}	0	0			
x_{p4}	0	0			
x_{p8}					

Running Example: Extended Case

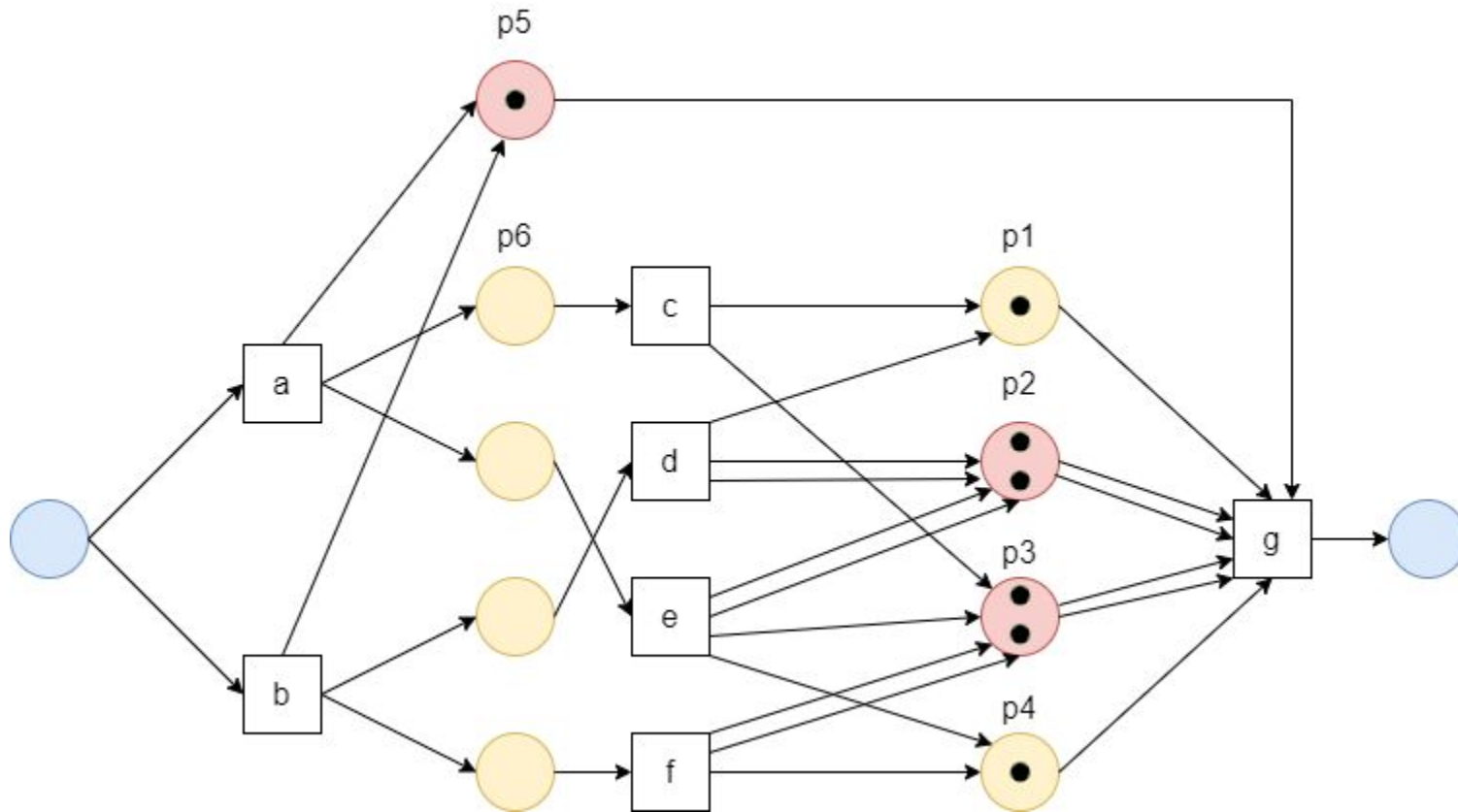
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	gg
x_{p3}	0	0	1		
x_{p4}	0	0	0		
x_{p8}					

Running Example: Extended Case

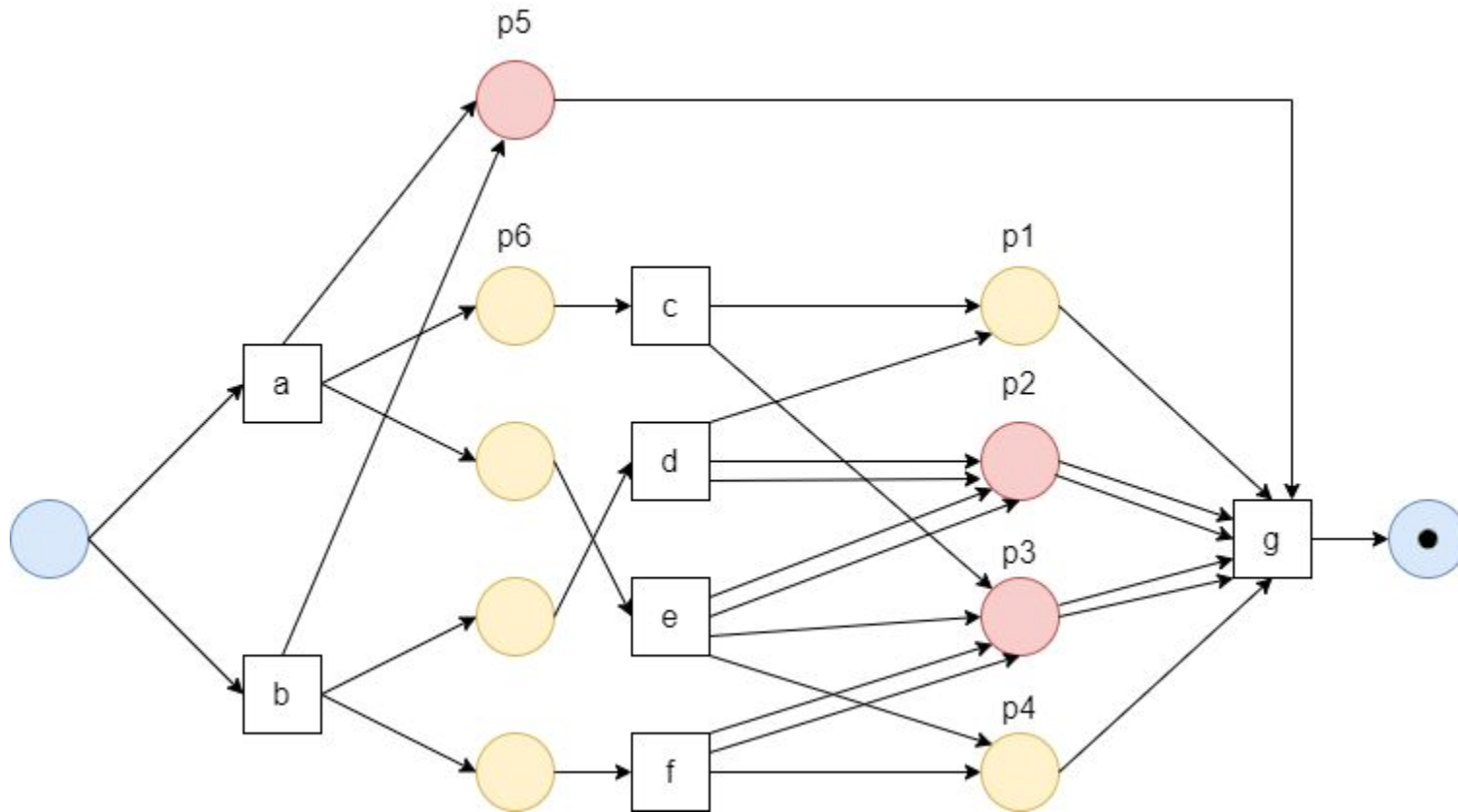
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	gg
x_{p3}	0	0	1	2	
x_{p4}	0	0	0	1	
x_{p8}					

Running Example: Extended Case

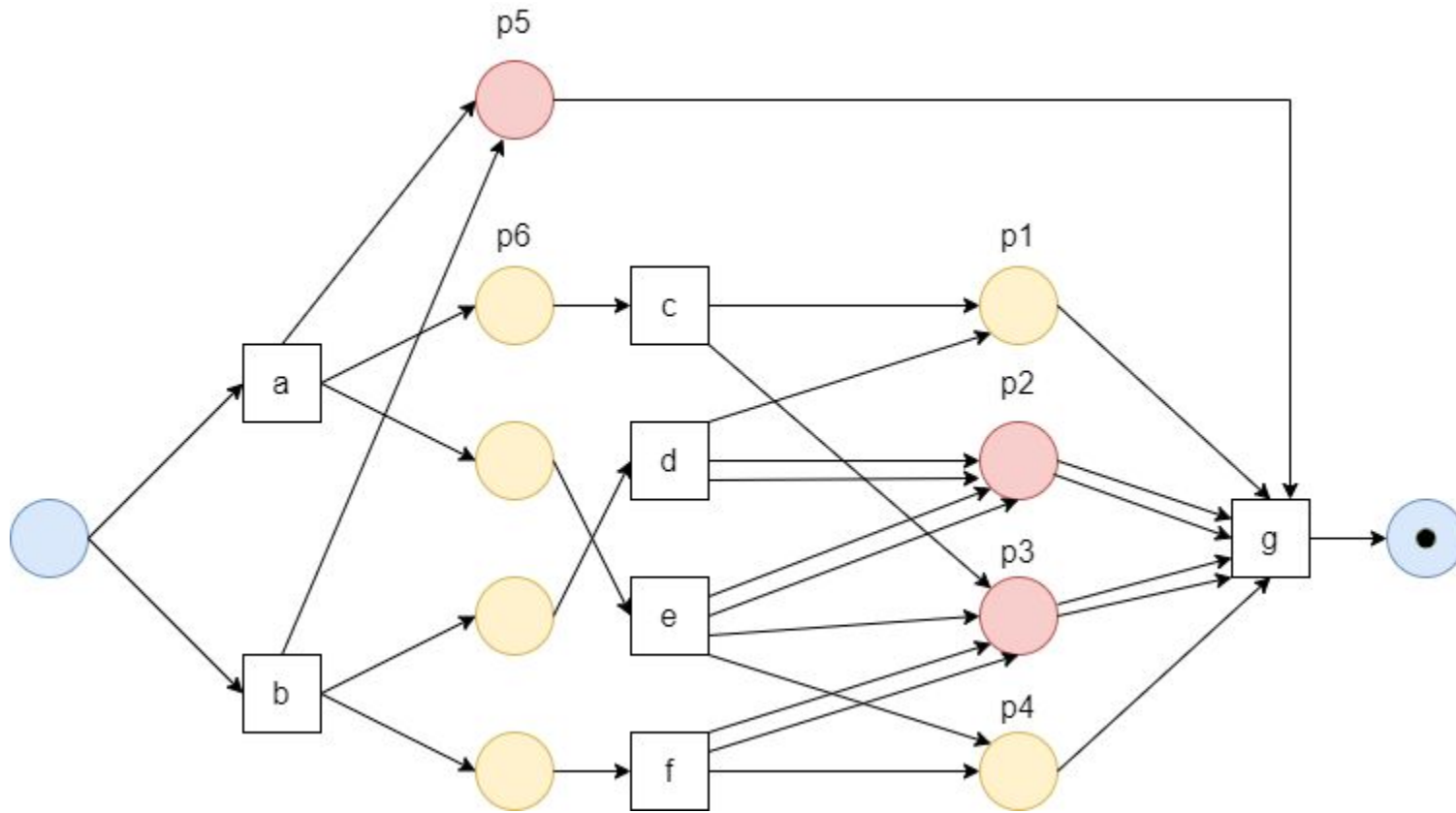
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_3}	0	0	1	2	0
x_{p_4}	0	0	0	1	0
x_{p_8}					

Running Example: Extended Case

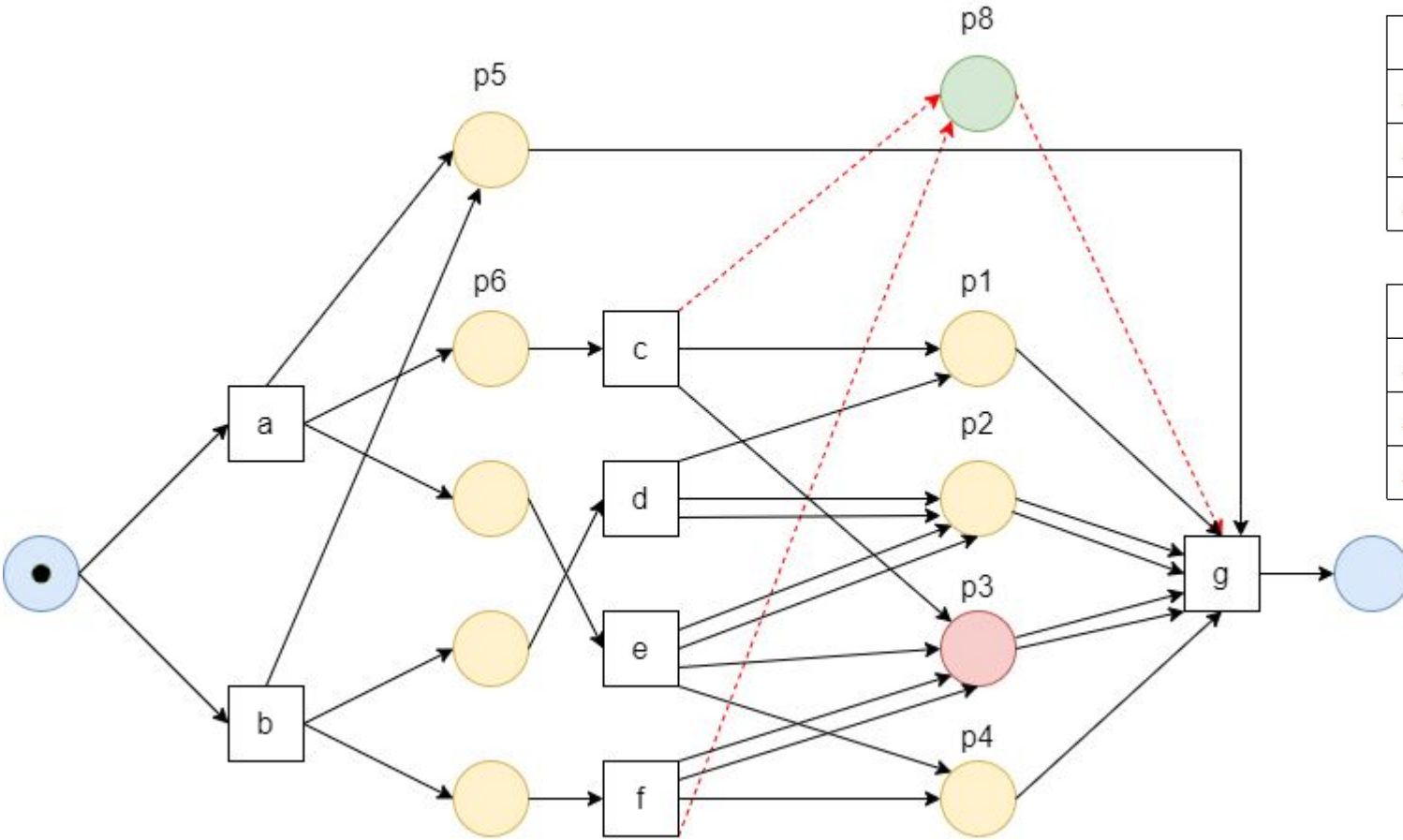
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_3}	0	0	1	2	0
x_{p_4}	0	0	0	1	0
x_{p_8}	0	0	1	1	0

Running Example: Extended Case

$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_3}	0	0	1	2	0
x_{p_4}	0	0	0	1	0
x_{p_8}	0	0	1	1	0

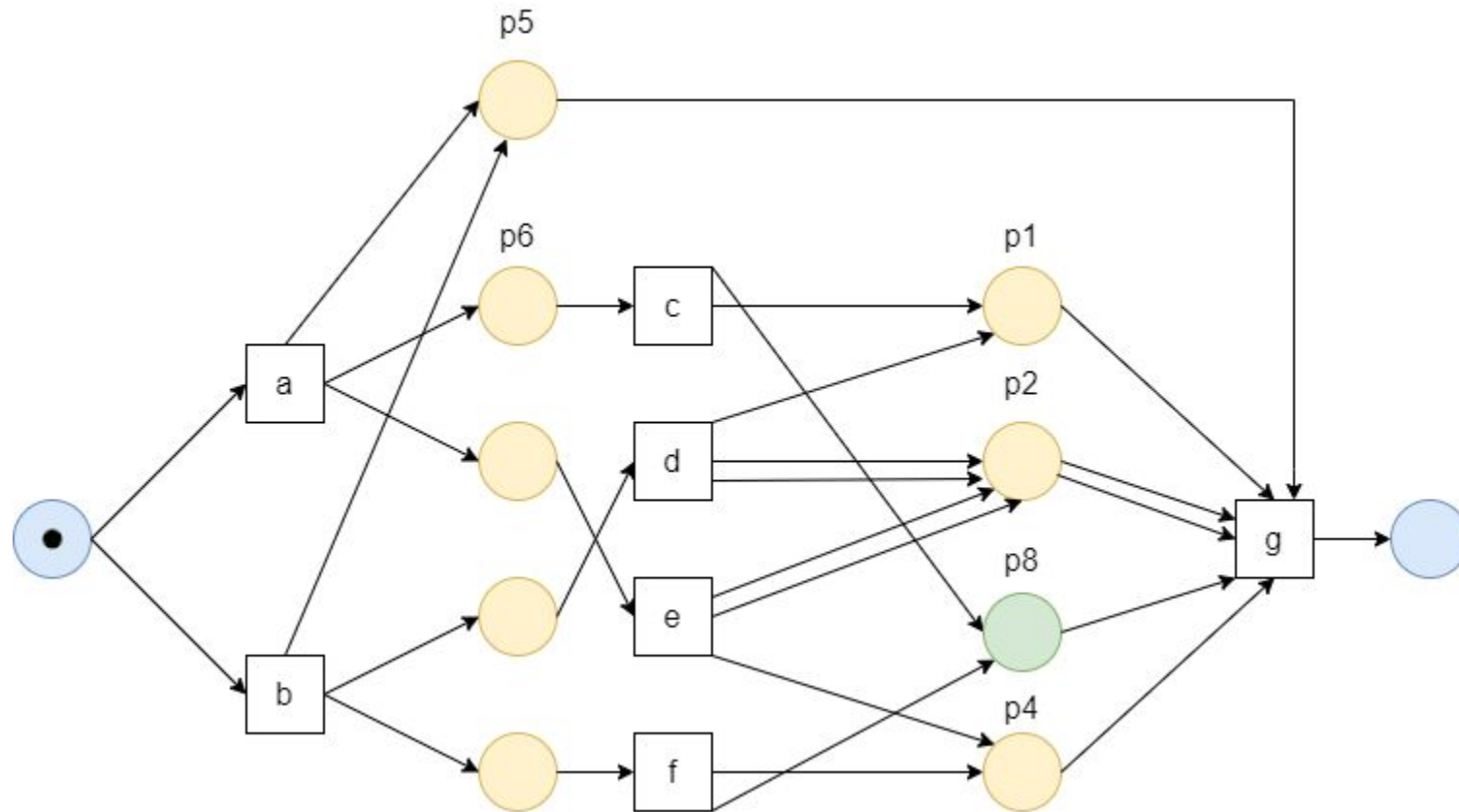
	ϵ	a	e	c	g
x_{p_3}	0	0	1	2	0
x_{p_4}	0	0	1	1	0
x_{p_8}	0	0	0	1	0

	ϵ	b	d	f	g
x_{p_3}	0	0	0	2	0
x_{p_4}	0	0	0	1	0
x_{p_8}	0	0	0	1	0

	ϵ	b	f	d	g
x_{p_3}	0	0	2	2	0
x_{p_4}	0	0	1	1	0
x_{p_8}	0	0	1	1	0

Running Example: Extended Case

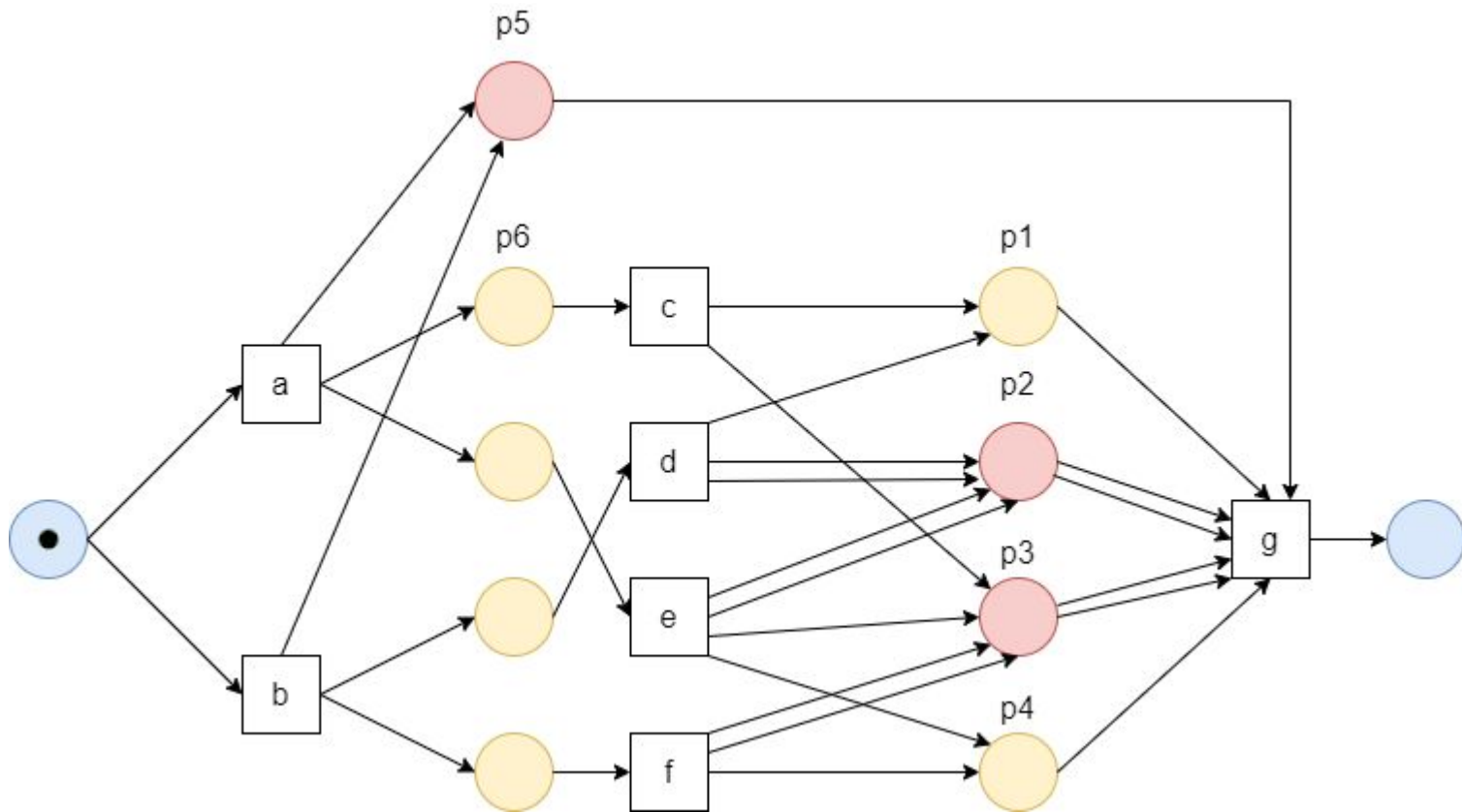
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



- **Realized the algorithm via its simple, intermediate and extended cases**
- **However, can all implicit places be discovered using this method?**
- **Uncovering the limitations of our implicit place removal technique**

Limitations

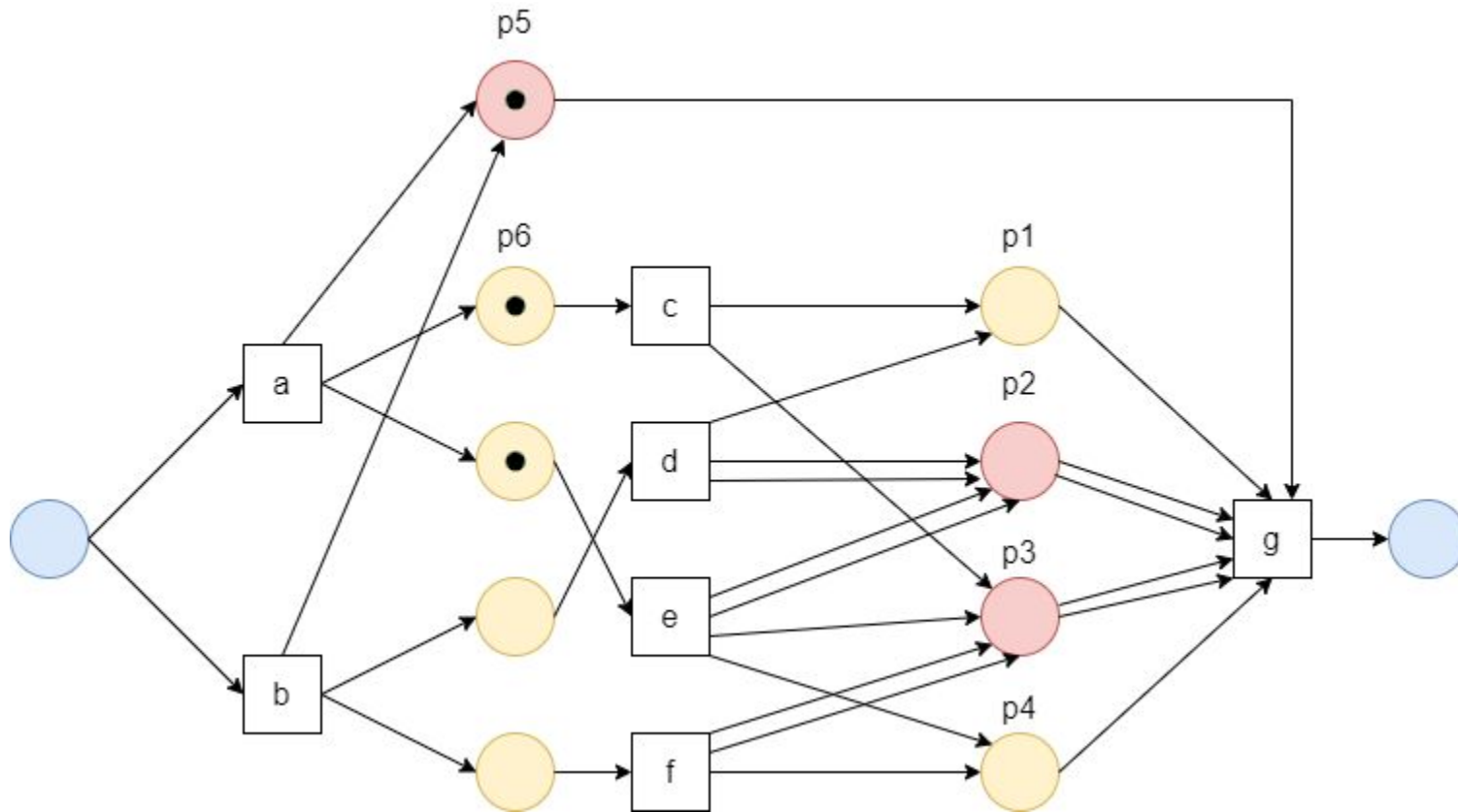
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_2}	0				
x_{p_1}	0				
x_{p_9}					

Limitations

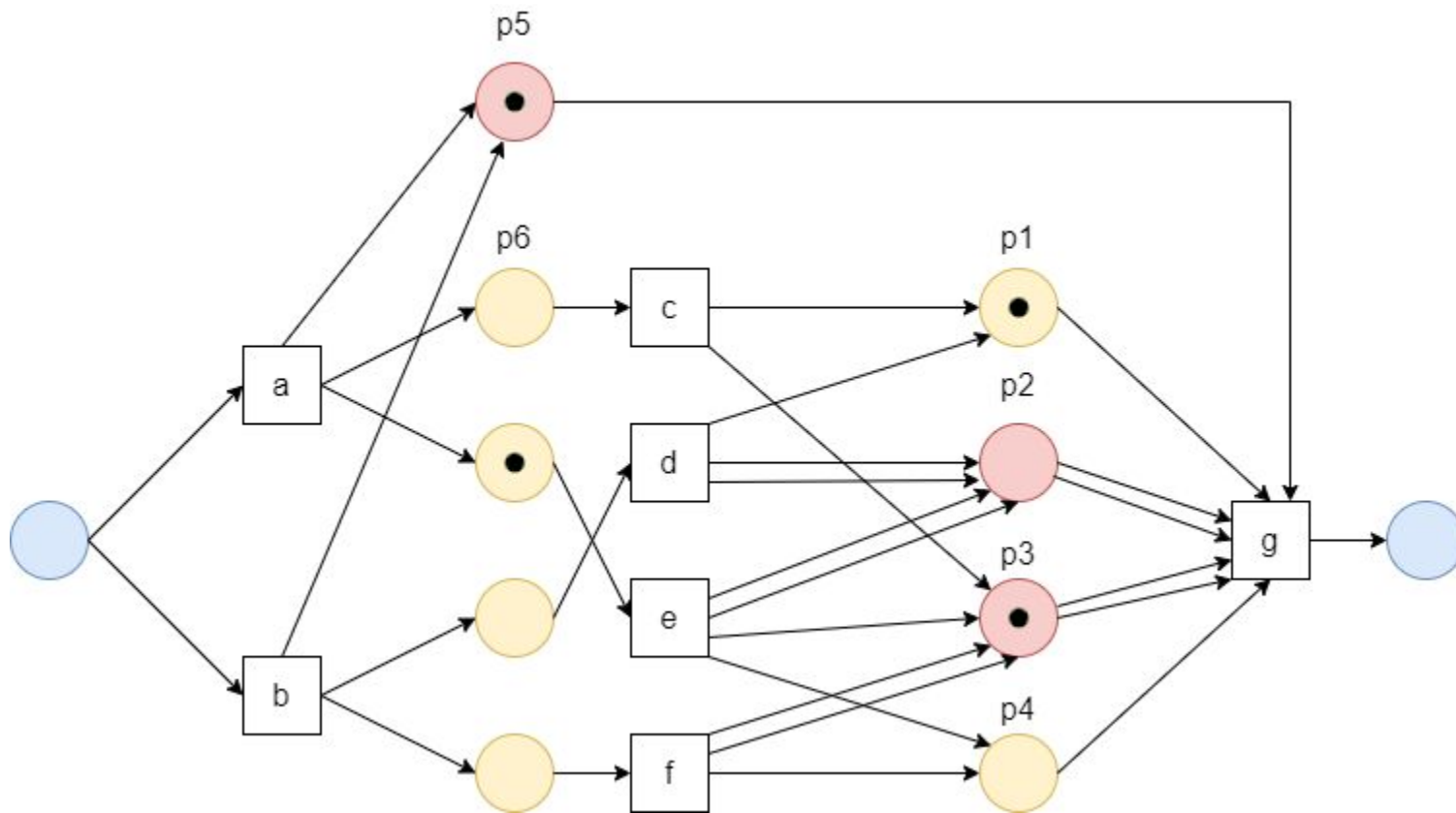
$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_2}	0	0			
x_{p_1}	0	0			
x_{p_9}					

Limitations

$$L = [\epsilon, \langle a, c, e, g \rangle^2, \langle a, e, c, g \rangle^3, \langle b, d, f, g \rangle^2, \langle b, f, d, g \rangle^4]$$



	ϵ	a	c	e	g
x_{p_2}	0	0	0		
x_{p_1}	0	0	1		
x_{p_9}	0	0	-1		

$x_{p_2} > x_{p_i}$ does not hold for any i

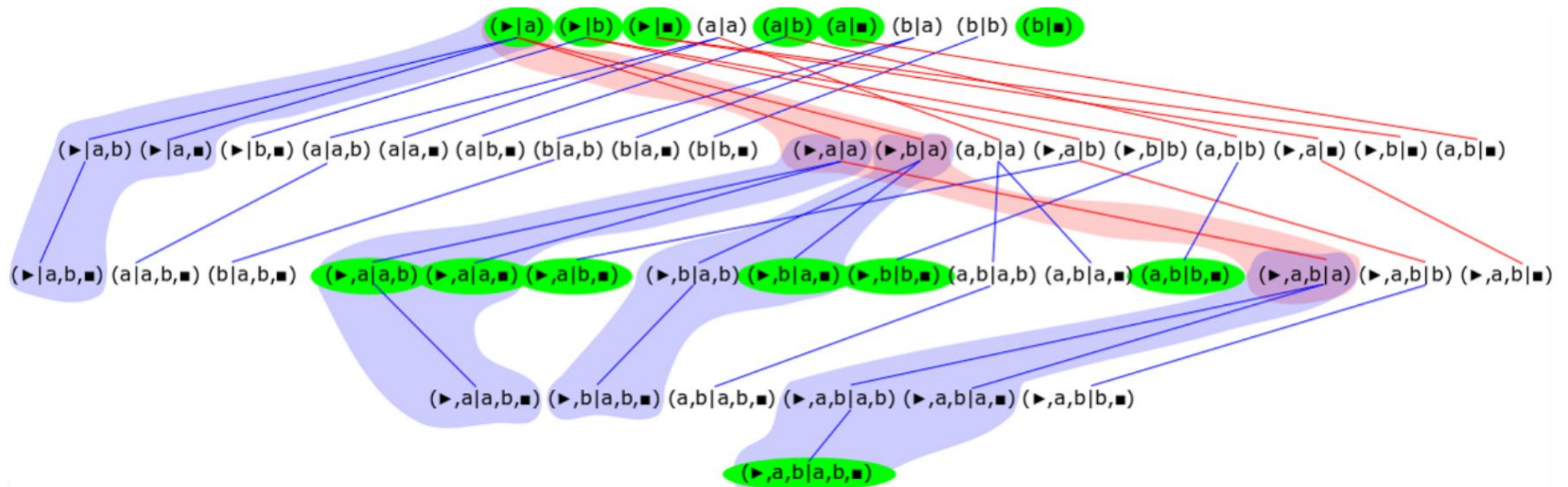
- **Understood the motivation behind removing implicit places**
- **Examined the main idea of our method**
- **Witnessed running examples of the algorithm**
- **Recognized limitations of our technique**

Next Steps

- **When and where is this algorithm actually beneficial?**
- **Diving into the application to the eST-Miner**
- **Are there any associated implementation challenges?**

- **Using the eST-Miner as a process discovery algorithm**
- **Only the event log is used**
- **Candidate places are traversed and evaluated using token replay**
- **Replaying the event log is crucial to removing implicit places**

➤ A clever way to enumerate places using a *tree* like structure



- The discovery of a set of all *fitting* places is guaranteed
- A significant number of implicit places are also discovered
- Removed by solving an Integer Linear Programming Problem
- Immense time and space complexity
- Reason for applying our implicit place removal technique

Final Place Removal (FPR)

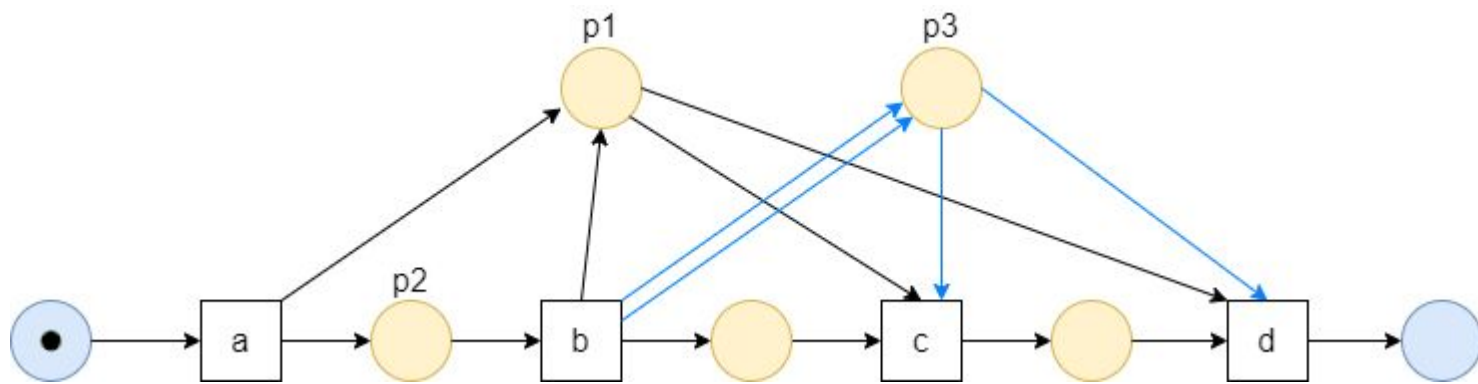
- **The final set of places is computed first**
- **Then, all places are compared and implicit places are removed**

Concurrent Place Removal (CPR)

- **Every discovered place is directly compared to existing places**
- **Removed immediately if found to be implicit**

Application: Challenges

- The eST-Miner returns a Petri net without arc weights
- However, arc weights are required



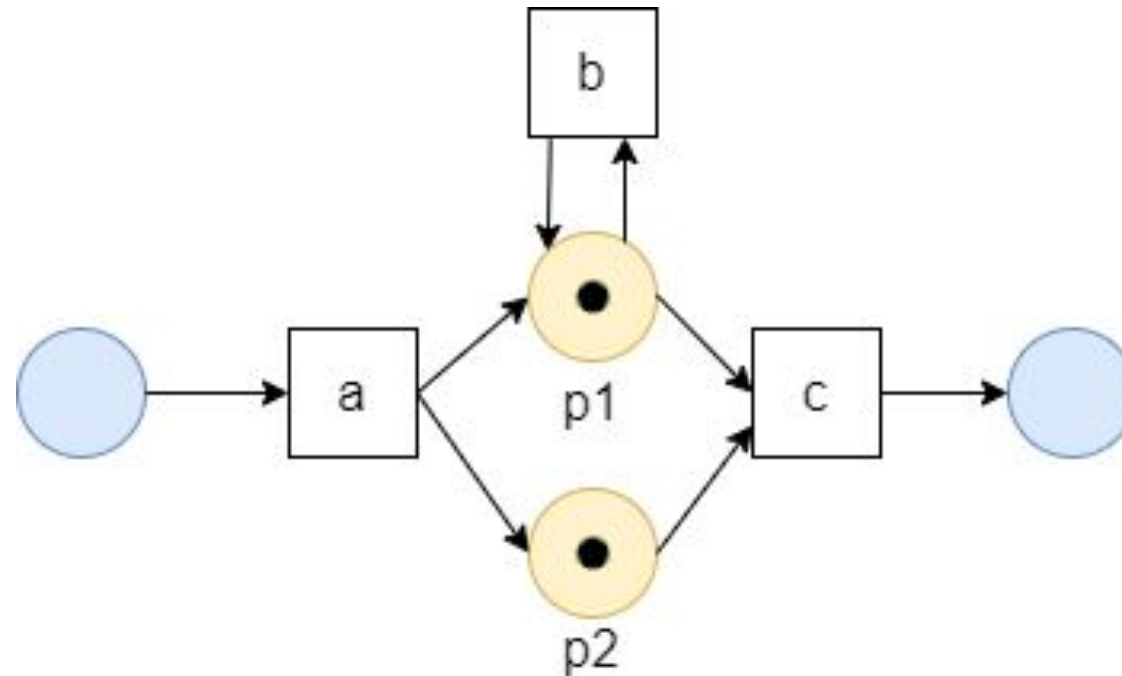
	ϵ	a	b	c	d
x_{p_1}	0	1	2	1	0
x_{p_2}	0	1	0	0	0
x_{p_3}	0	0	2	1	0

- **The eST-Miner discovers only fitting places**
- **Places empty at the beginning and end of token replay**
- **Place emptiness is guaranteed by construction**

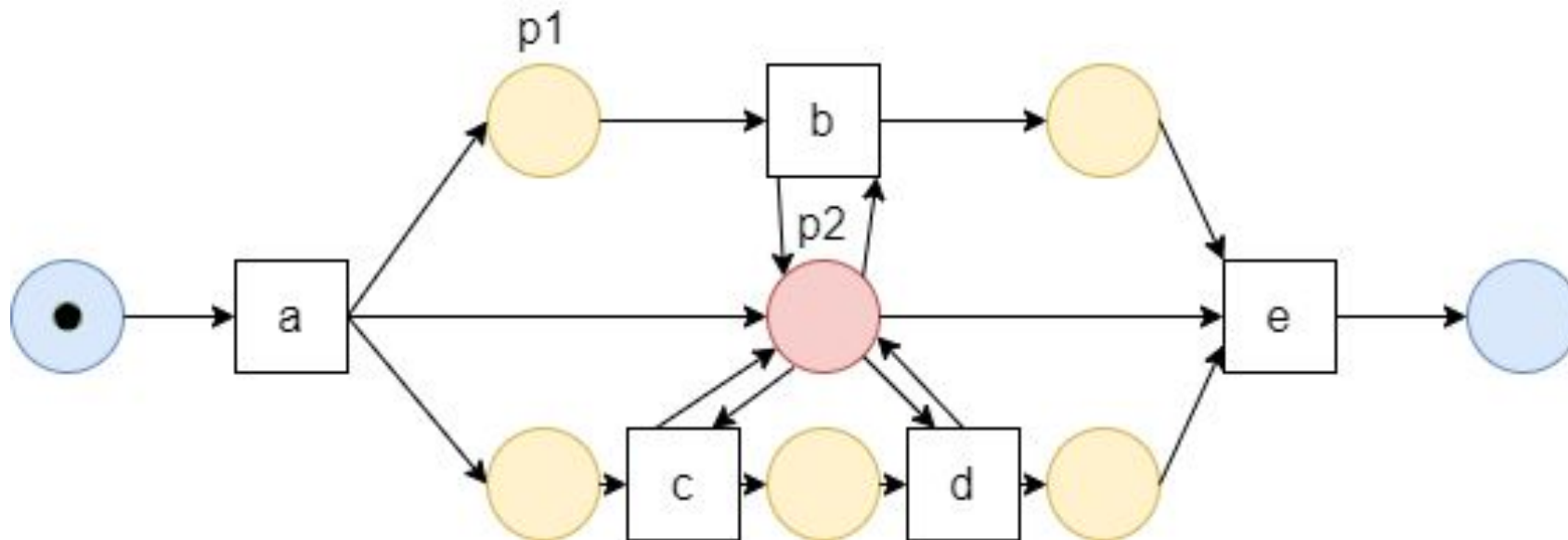
$$x_{p_1} = x_{p_2} = 0 \quad (4)$$

$$x_{p_3} = 0 \quad (5)$$

eST-Miner allows for Petri nets with self-loops

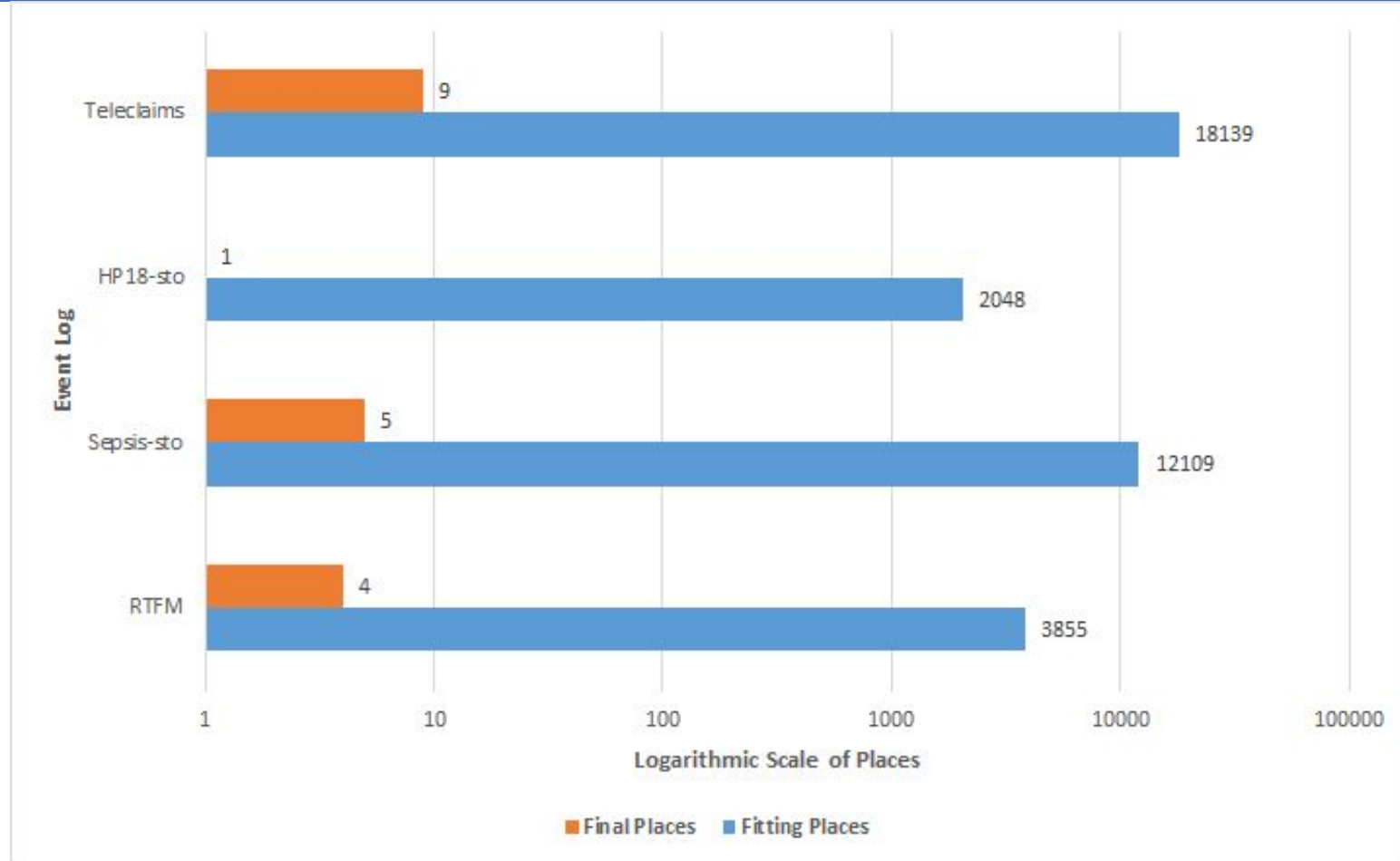


Special case of places with self-loops in parallel constructs

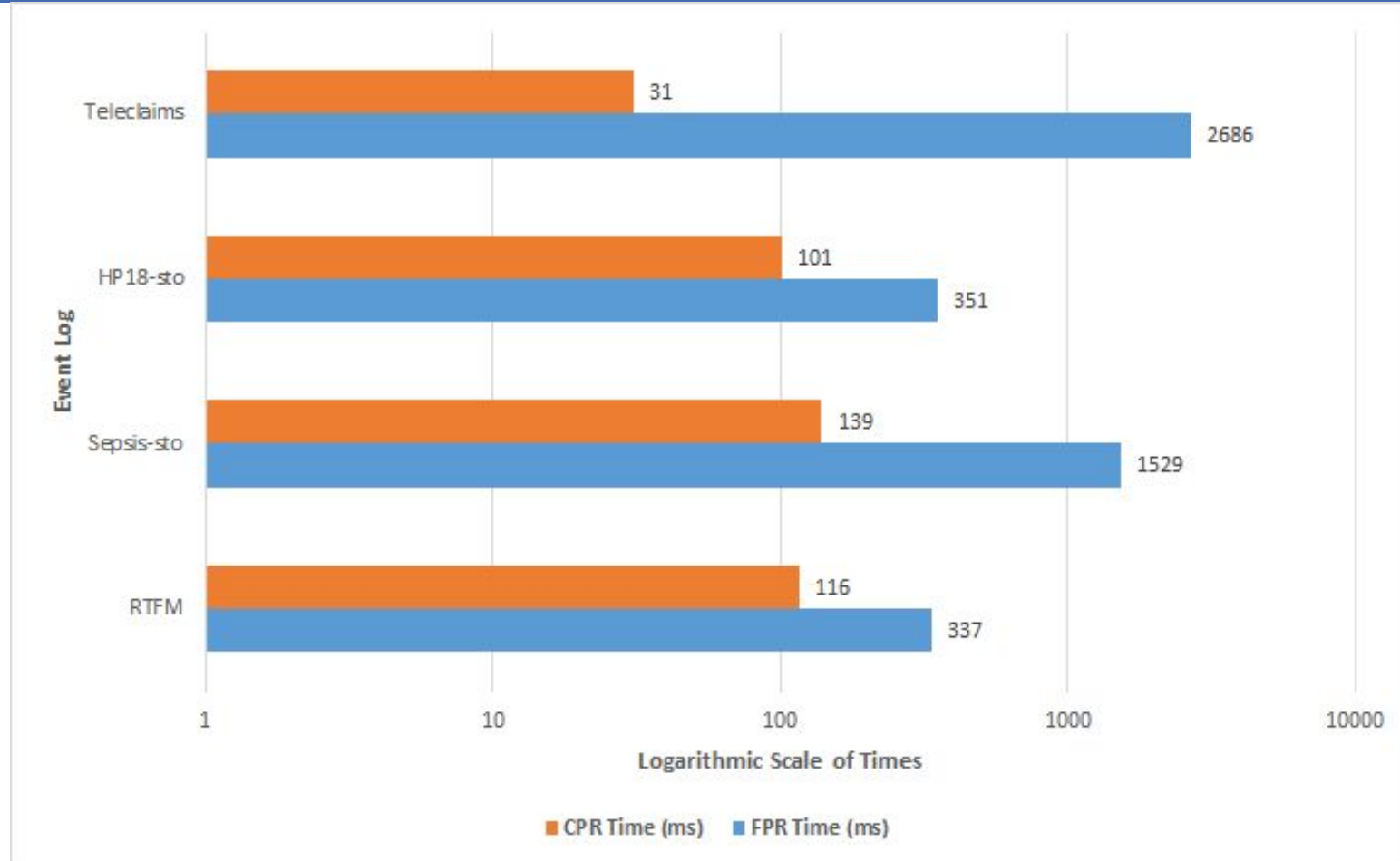


Evaluation: Space Efficiency

Remaining final places are significantly less than total fitting places



CPR is always significantly faster than FPR



- **An approach to identify and remove implicit places from Petri nets**
- **Combination with the eST-Miner process discovery algorithm**
- **Sequential and concurrent application schemes**
- **Robust time and space efficiency of the CPR variant**

- **Further investigations with the eST-Miner**
 - **Increasing the efficiency of the candidate traversal step**
 - **Returning results after a certain running time**
- **Choosing the order of place comparisons**
- **Solving the problem of self-loop places in parallel constructs**
- **Application to the Inductive-Miner process discovery algorithm**

Thank you!

Questions?

References

- [1] L. Mannel. Removing Implicit Places Using Regions for Process Discovery. 06 2020.
- [2] W. Aalst and A. Ter. Verification of workflow task structures: A petri-net-based approach. *Information Systems*, 25:43–69, 03 2000.
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