

Counting Popular Matchings in House Allocation Problems

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CSR 2014

The Problem

Agents

a_1 ●

a_2 ●

a_3 ●

a_4 ●

a_5 ●

a_6 ●

Houses

 h_1

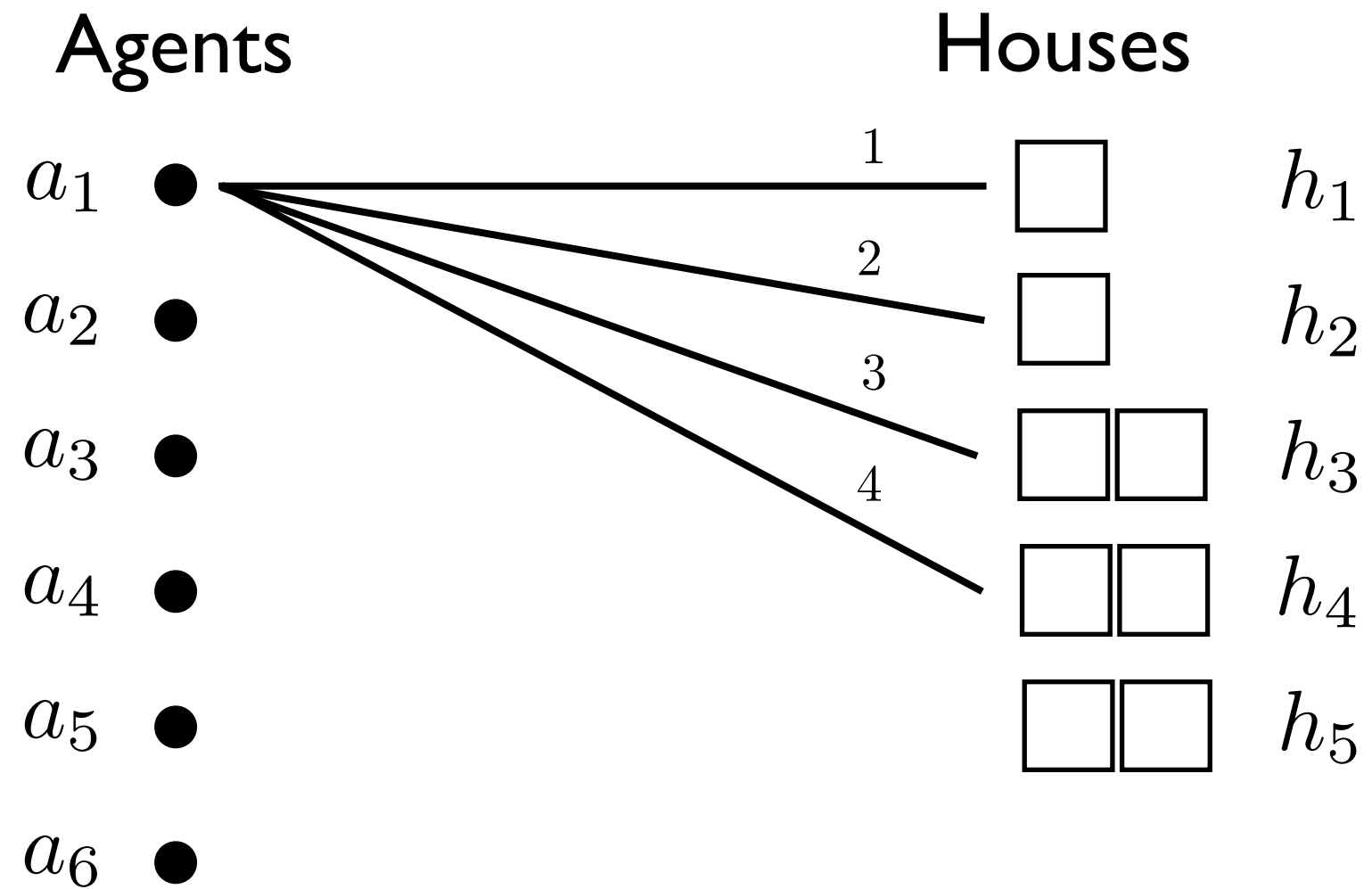
 h_2

 h_3

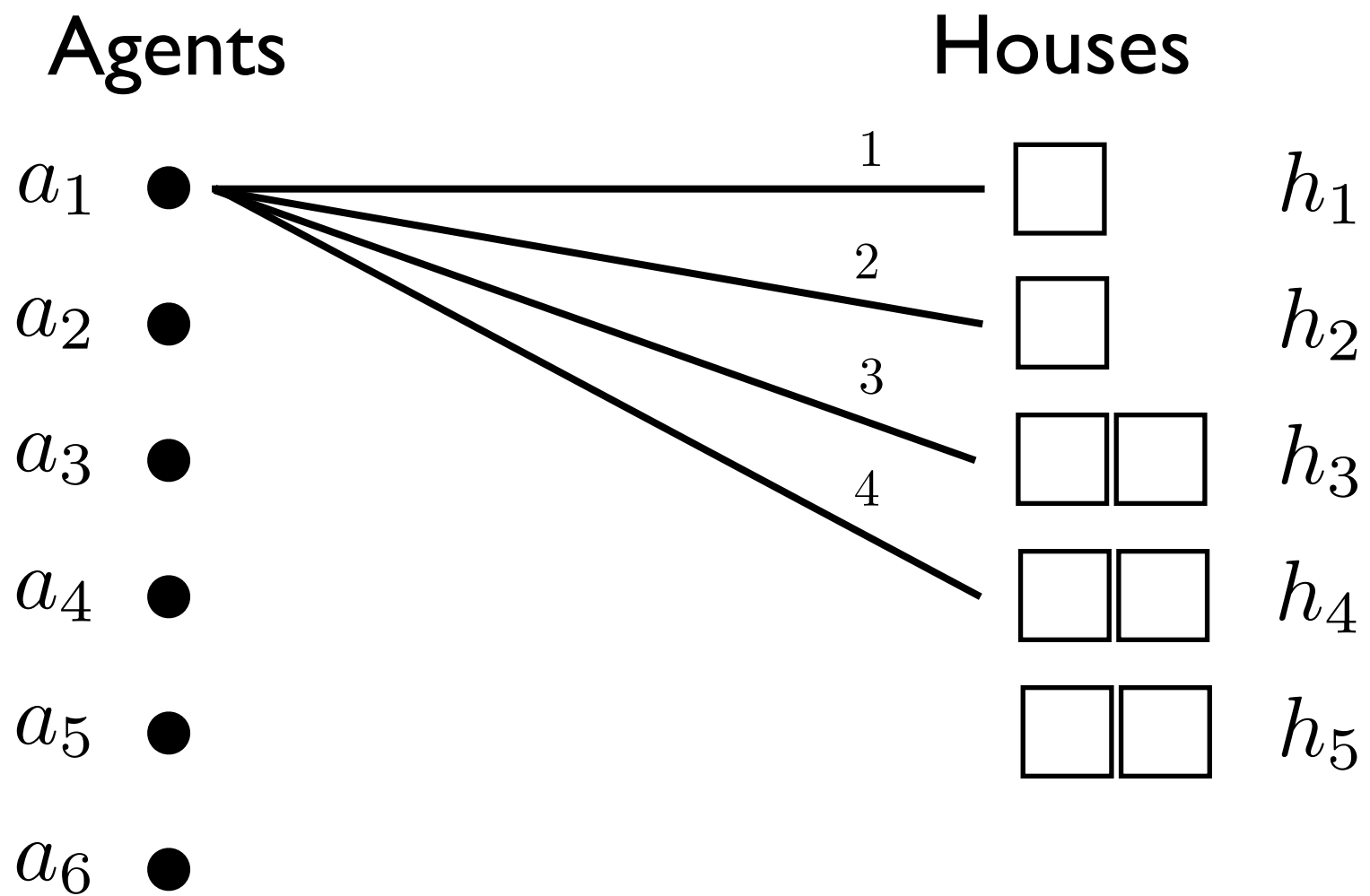
 h_4

 h_5

The Problem



The Problem



a_1 : h_1 h_2 h_3 h_4

The Problem Instance

Preferences

a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	
a_5	h_5	h_1		
a_6	h_5	h_2		

The Problem Instance

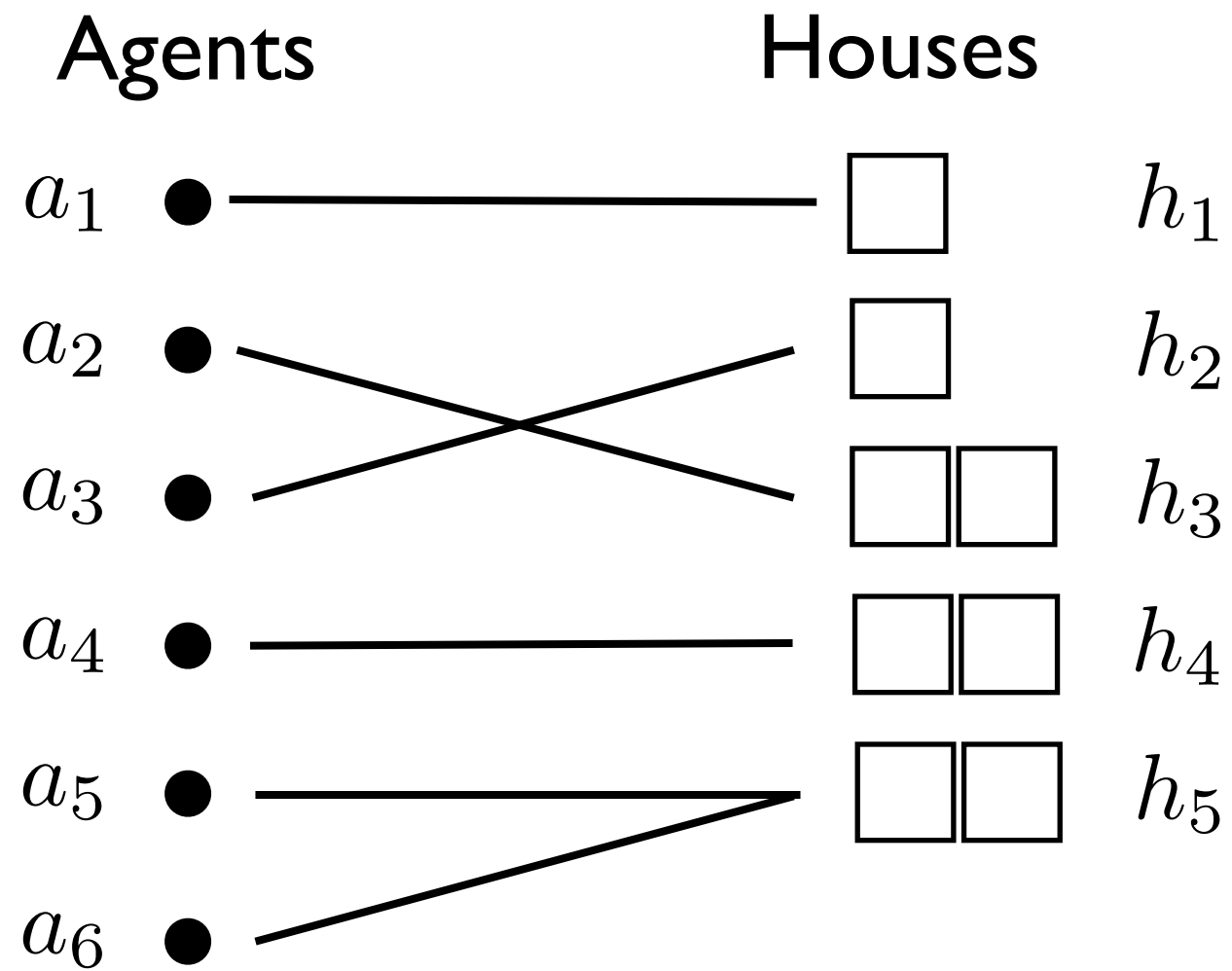
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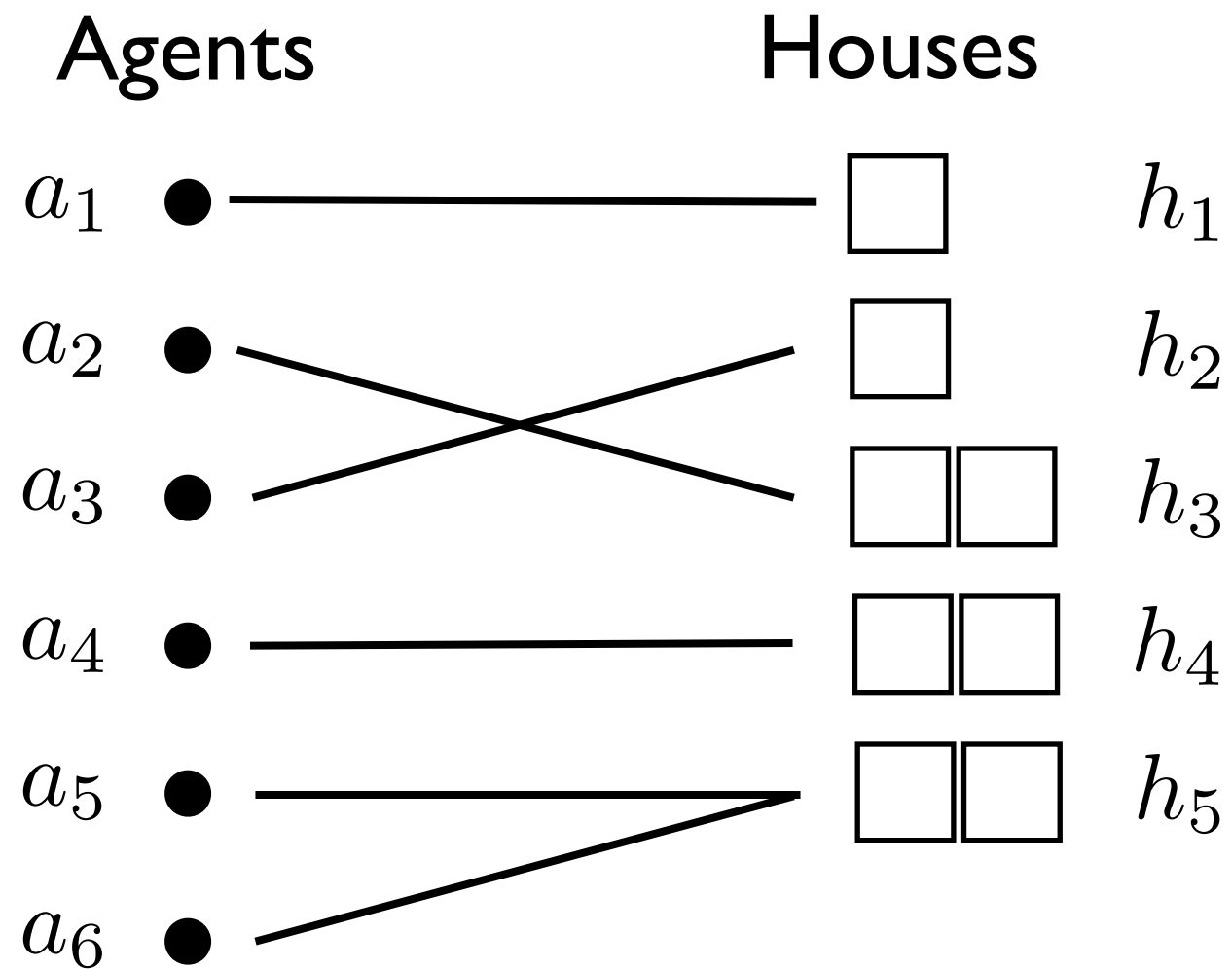
Capacities

h_1	1
h_2	1
h_3	2
h_4	2
h_5	2

A Matching



A Matching



Is it a good matching?

Comparing two matchings

M_1	M_2
$a_1 \quad h_1$	$a_1 \quad h_4$
$a_2 \quad h_3$	$a_2 \quad h_1$
$a_3 \quad h_2$	$a_3 \quad h_3$
$a_4 \quad h_4$	$a_4 \quad h_4$
$a_5 \quad h_5$	$a_5 \quad h_5$
$a_6 \quad h_5$	$a_6 \quad h_5$

Voting

M_1

a_1 h_1

a_2 h_3

a_3 h_2

a_4 h_4

a_5 h_5

a_6 h_5

M_2

a_1 h_4

a_2 h_1

a_3 h_3

a_4 h_4

a_5 h_5

a_6 h_5

a_1 | h_1 | h_2 | h_3 | h_4

Voting

M_1		M_2						
a_1	h_1	a_1	h_4	a_1	h_1	h_2	h_3	h_4
a_2	h_3	a_2	h_1					
a_3	h_2	a_3	h_3					
a_4	h_4	a_4	h_4					
a_5	h_5	a_5	h_5					
a_6	h_5	a_6	h_5					

a_1
 M_1

Voting

M_1		M_2	
a_1	h_1	a_1	h_4
a_2	h_3	a_2	h_1
a_3	h_2	a_3	h_3
a_4	h_4	a_4	h_4
a_5	h_5	a_5	h_5
a_6	h_5	a_6	h_5

a_1	a_2	a_3	a_4	a_5	a_6
M_1	M_2	M_1	*	*	*

“more popular than”

a_1	a_2	a_3	a_4	a_5	a_6
M_1	M_2	M_1	*	*	*

M_1 more popular than M_2

Popular Matching - Definition

M is popular if $\nexists M'$ more popular than M

Popular Matching - Definition

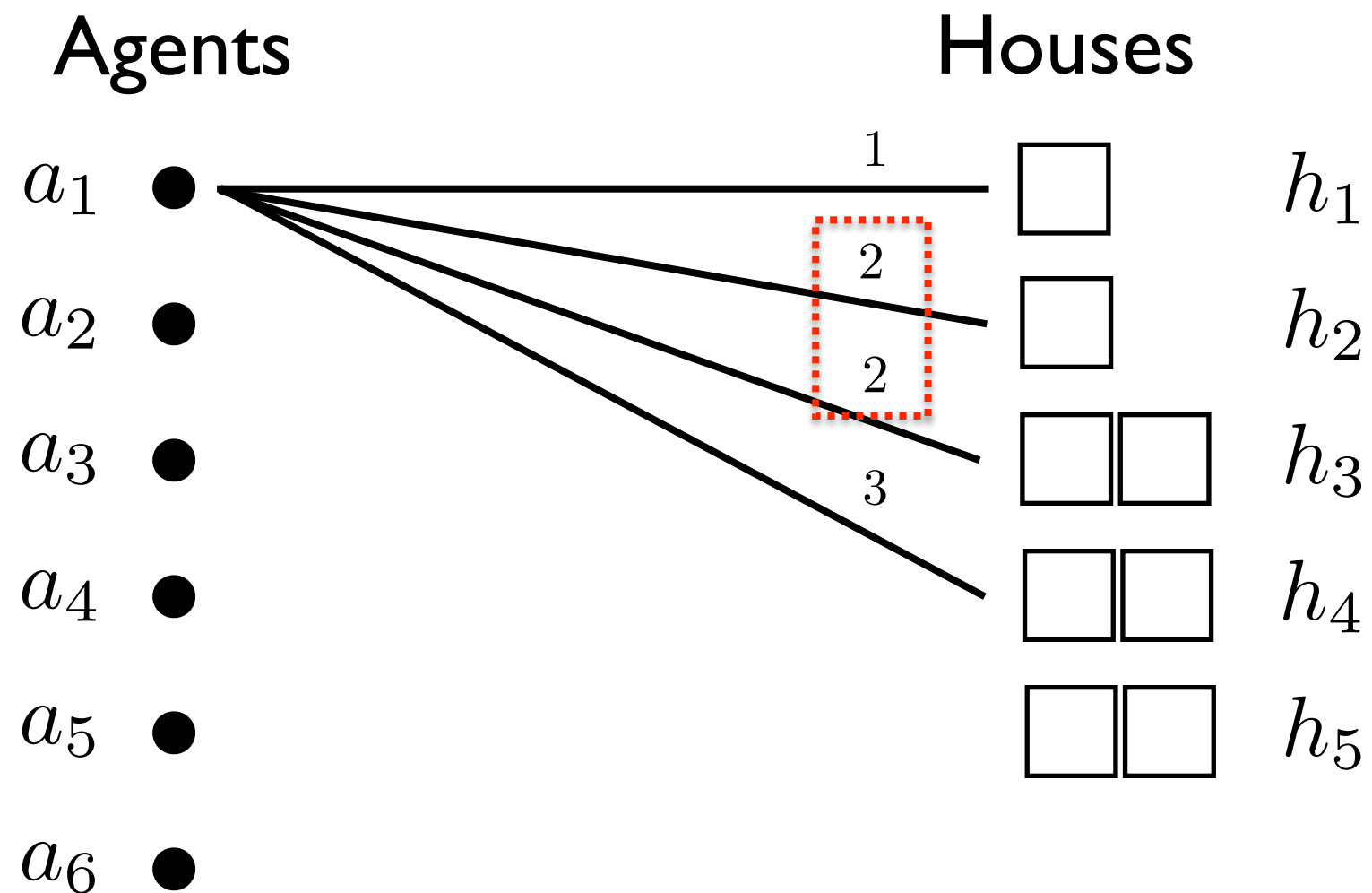
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Questions

Does there always exist one? No.

Can there exist more than one? Yes.

Instances with Ties



$a_1 : h_1 \underline{(h_2 \ h_3)} h_4$

History

[Abraham et al. 2005] gave a poly-time algorithm to output a popular matching

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Studied Extensively in Different Settings

- Random Popular Matching [Mahdian 2006]
- Different Optimality Criteria [KNN2010, KMN2011]
- Preferences on both sides [Kavitha 2012]
- Games on Popular Matching [Nasre 2013]

Counting Popular Matchings

	Hardness	Approximation

Counting Popular Matchings

	Hardness	Approximation
no ties, capacities = 1	poly-time	Exact Count in $O(n)$ [McDermid and Irving 2011]

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Counting Popular Matchings

	Hardness	Approximation
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ties, capacities = 1	#P-hard [Nasre 2013]	FPRAS [Acharyya, Chakraborty, J]
no ties, integer capacities	#P-hard [Acharyya, Chakraborty, J]	OPEN

Fully Polynomial Randomised Polynomial Scheme

For problem $f : \Sigma^* \rightarrow \mathbb{N}$, input $x \in \Sigma^*$, tolerance $\epsilon > 0$, error probability $\delta > 0$ outputs N such that for all x :

$$P[(1 - \epsilon)f(x) \leq N \leq (1 + \epsilon)f(x)] \geq \delta$$

in time $\text{poly}(|x|, 1/\epsilon, \log 1/\delta)$.

Our Focus

Strict Ordering, Integer Capacities

Preferences

a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	

Capacities

h_1	1
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Preferences

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Capacities

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f-houses = $\{h_1, h_2\}$

Preferences

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a_4	h_2	h_4	h_3	

Capacities

h_1	1
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Preferences

a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	

Capacities

h_1	1
h_2	1
h_3	2
h_4	2

f-houses = $\{h_1, h_2\}$

s-houses = $\{h_3, h_4\}$

Popular Matching Characterisation

In every popular matching:

- every agent gets either its f-house or s-house

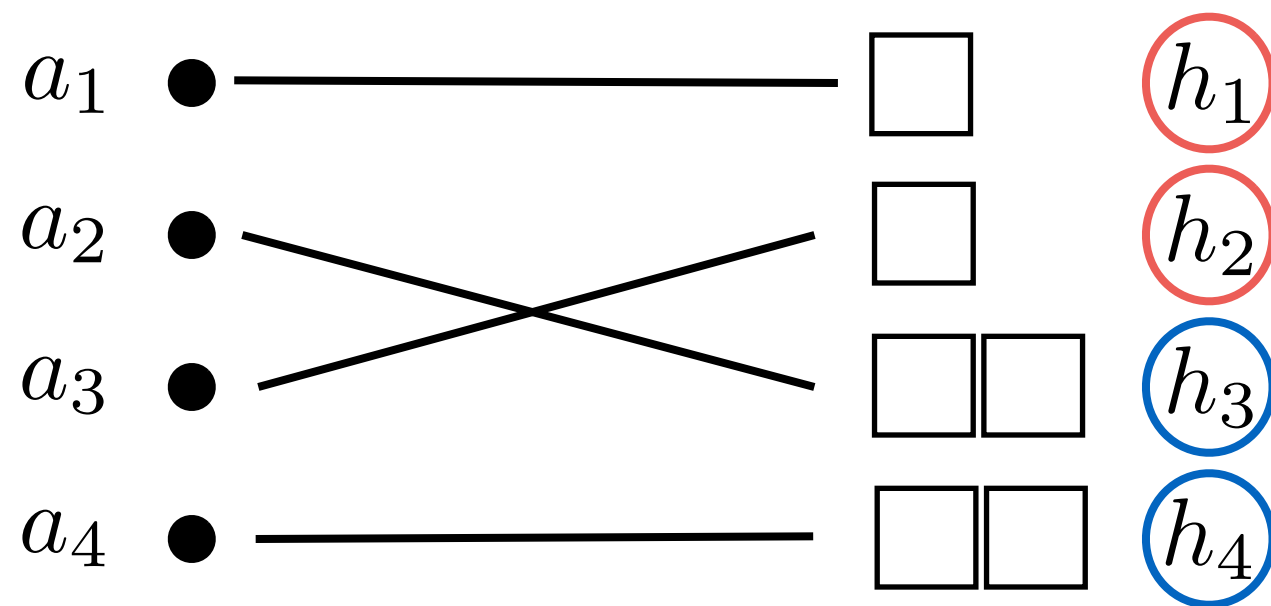
Popular Matching Characterisation

In every popular matching:

- every agent gets either its f-house or s-house
- every f-house is used to maximum capacity

a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	

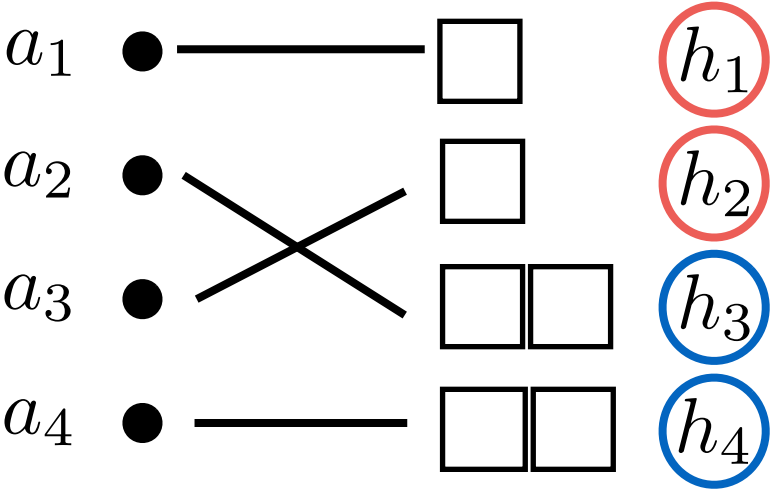
h_1	1
h_2	1
h_3	2
h_4	2



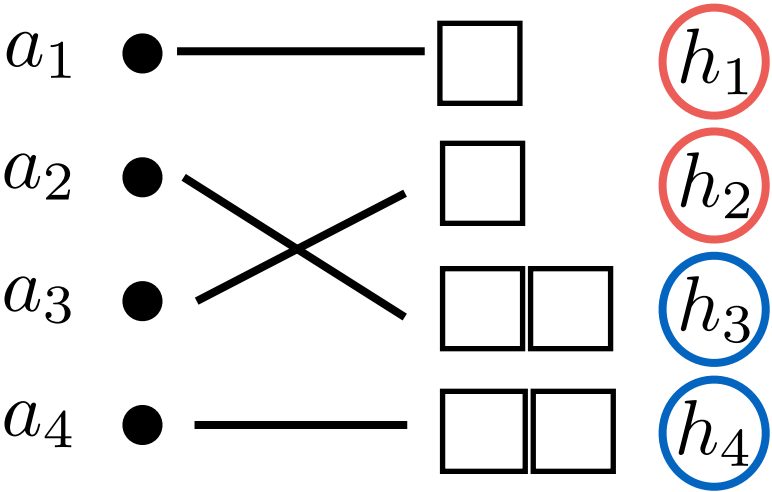
Switching Graph Characterisation

Another way to look at popular matchings!

a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	



a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
a_4	h_2	h_4	h_3	



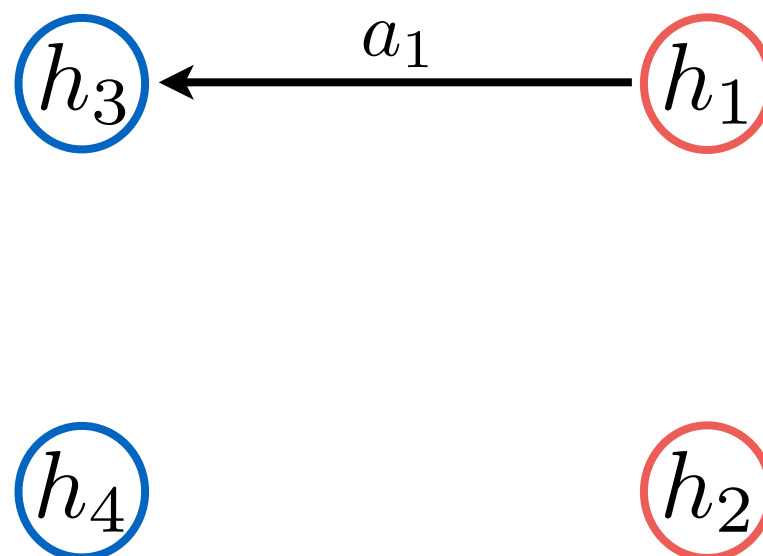
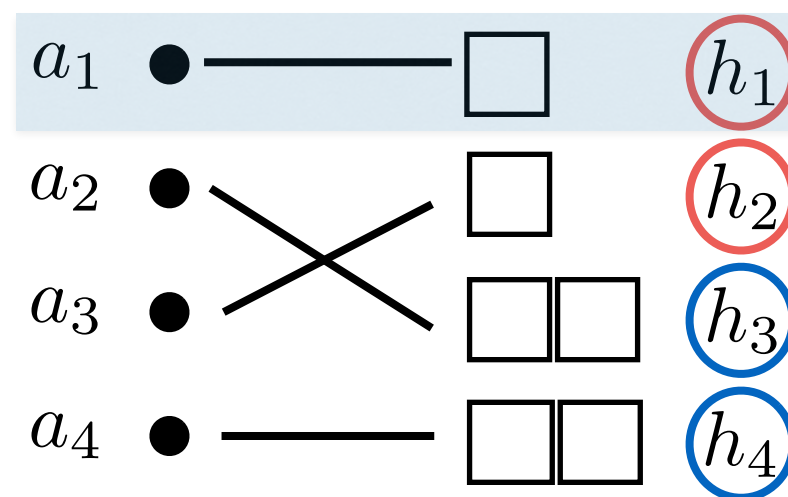
h_3

h_1

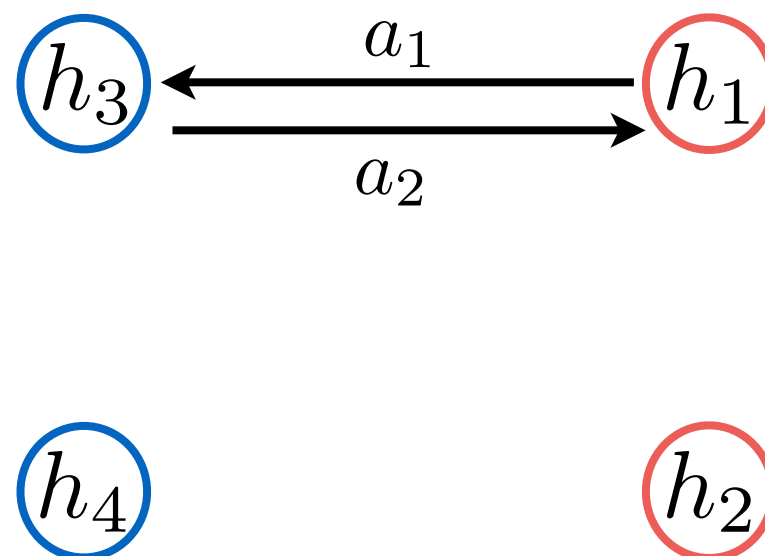
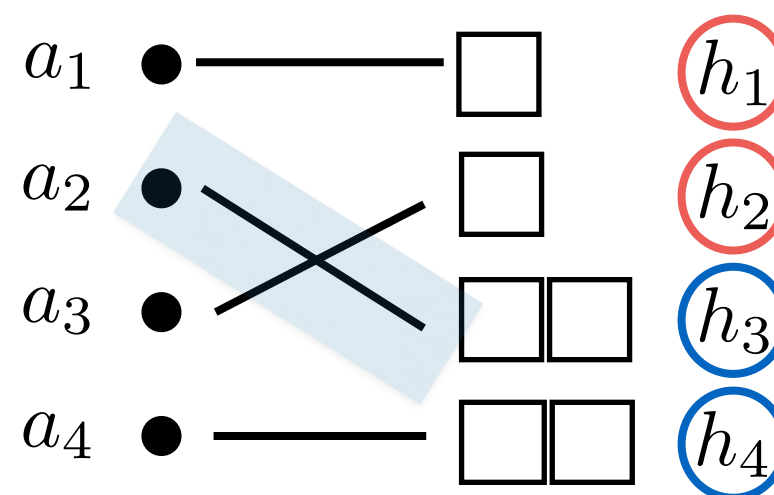
h_4

h_2

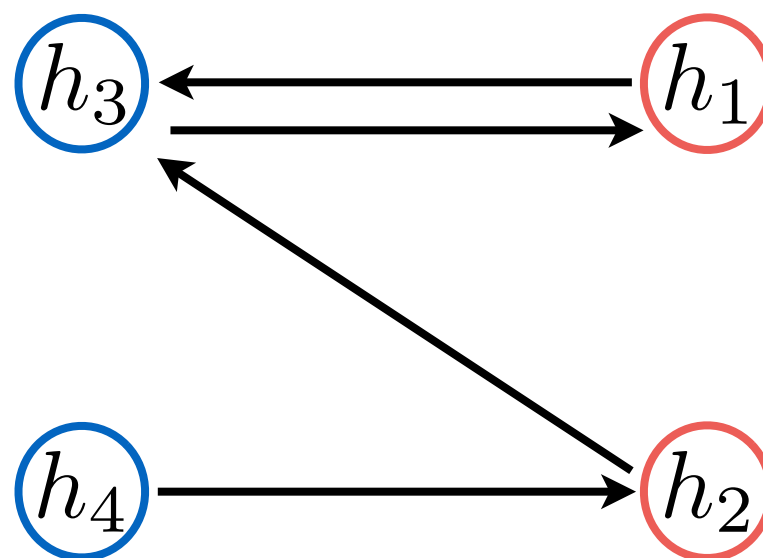
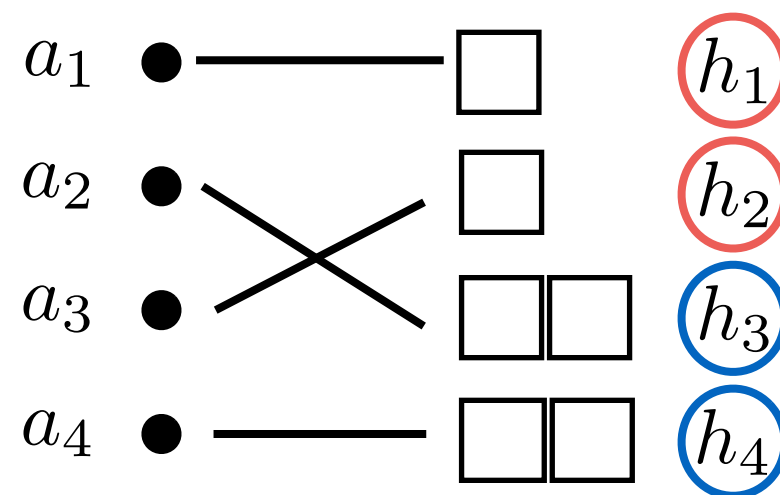
a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
a_3	h_2	h_1	h_3	
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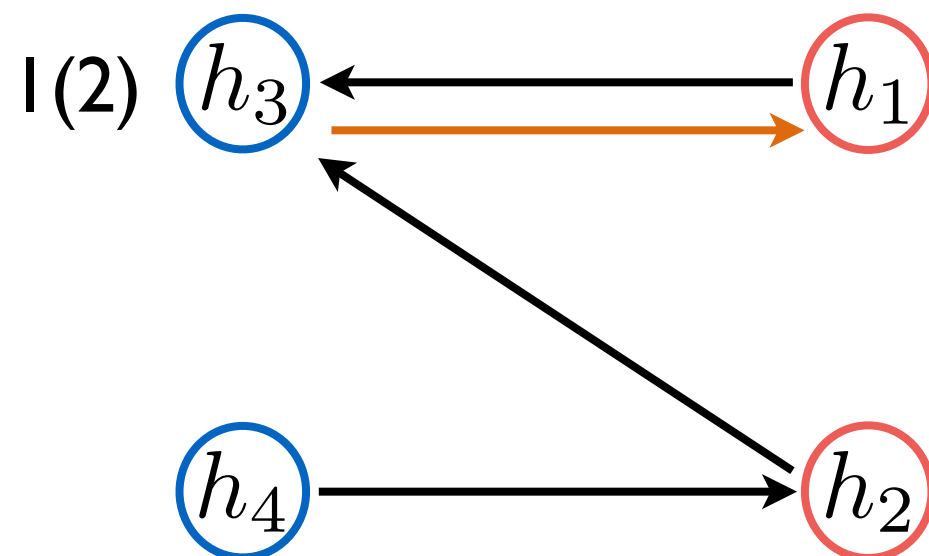
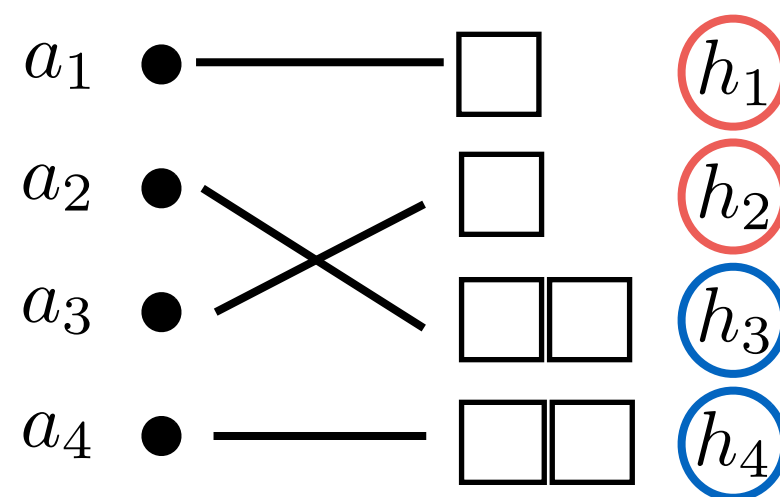
a_1	h_1	h_2	h_3	h_4
a_2	h_1	h_3	h_2	
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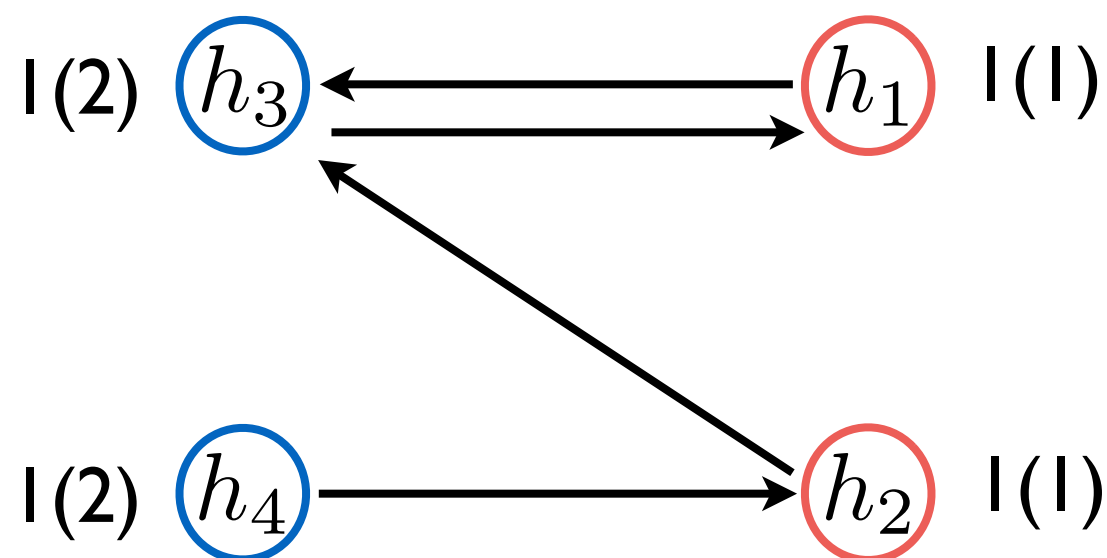
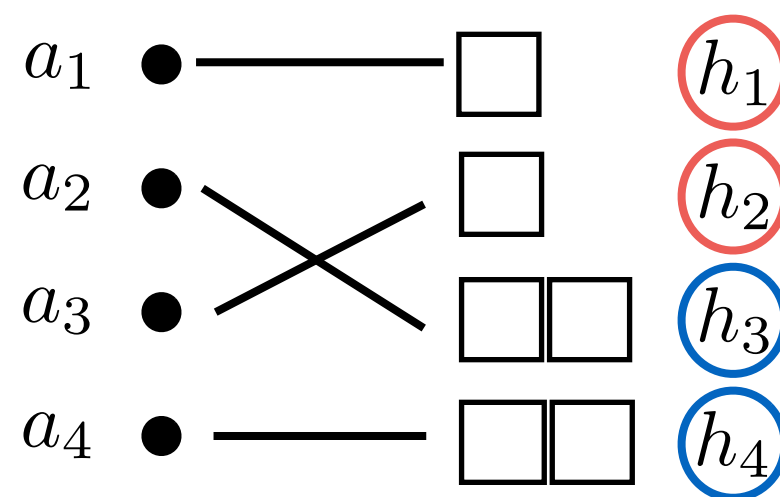
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a_4	h_2	h_4	h_3	



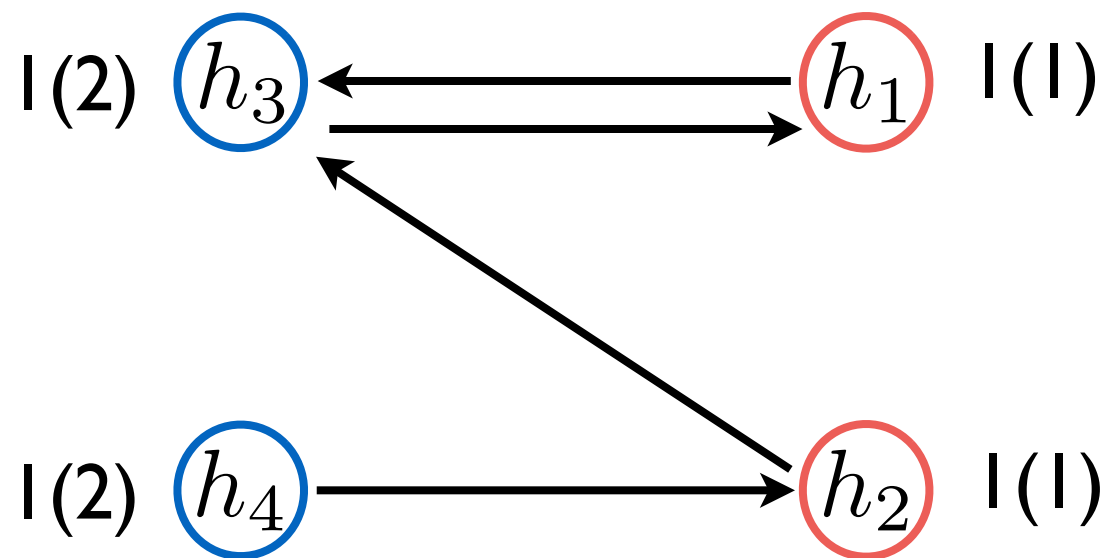
h_1	1
h_2	1
h_3	2
h_4	2



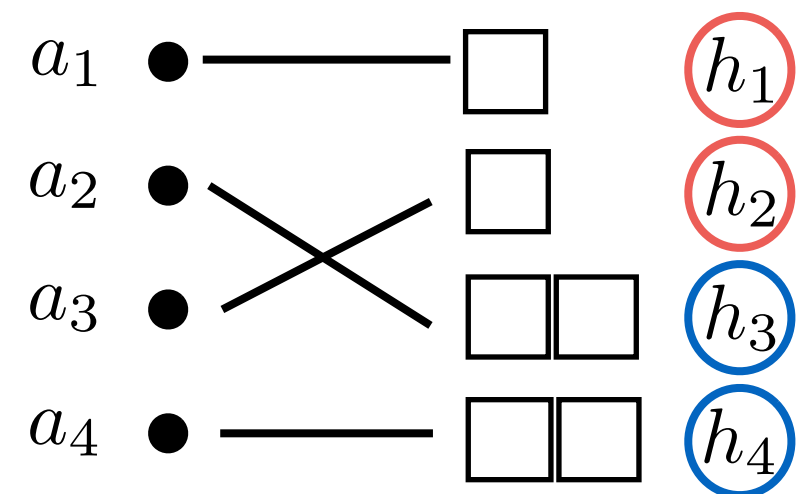
h_1	1
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Moving from one Popular Matching to Another



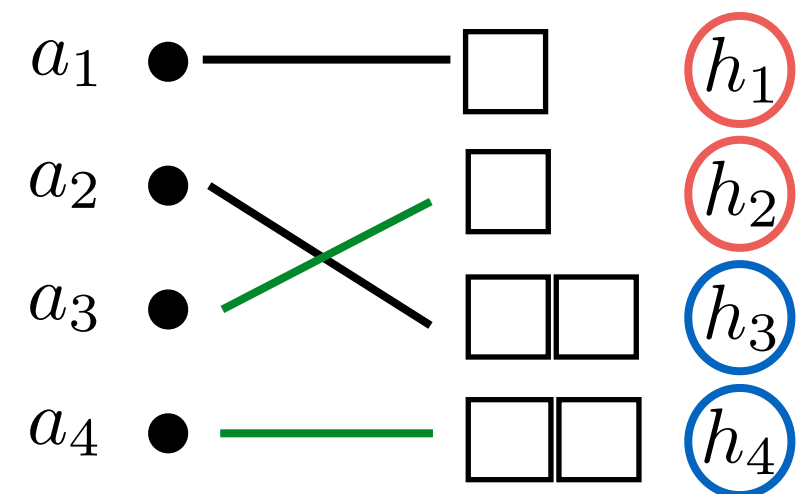
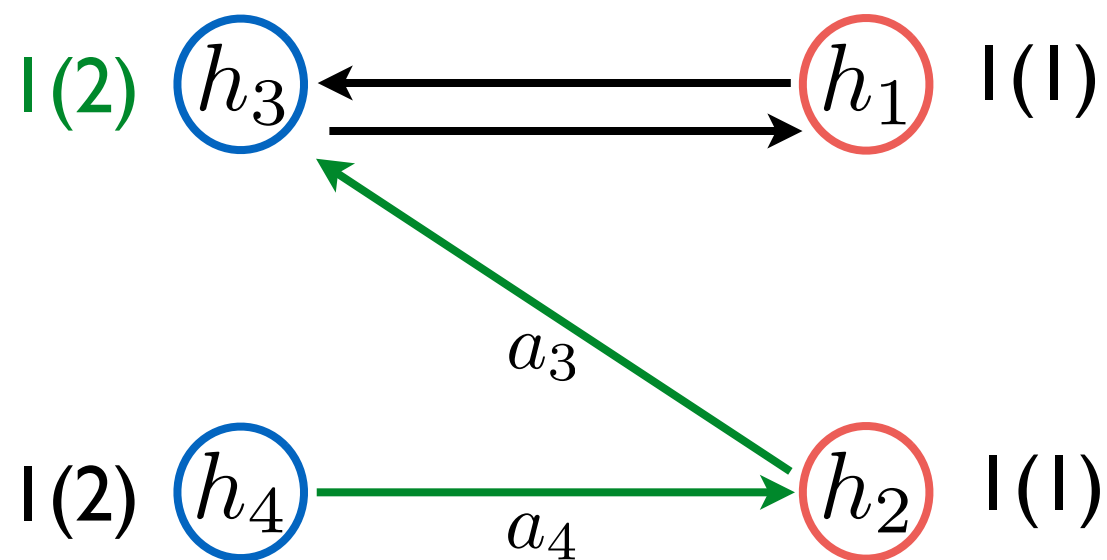
switching graph



popular matching

Moving from one Popular Matching to Another

Trick : Path Reversal



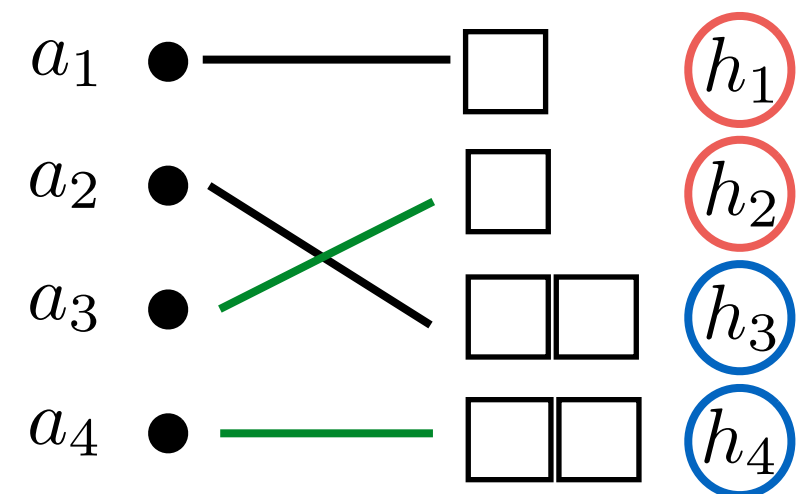
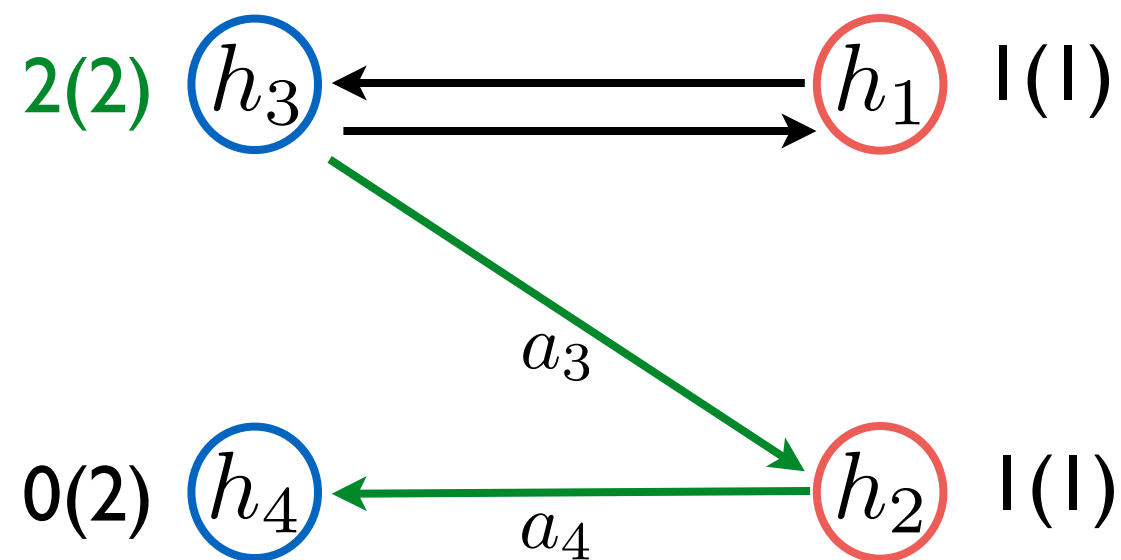
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popular matching

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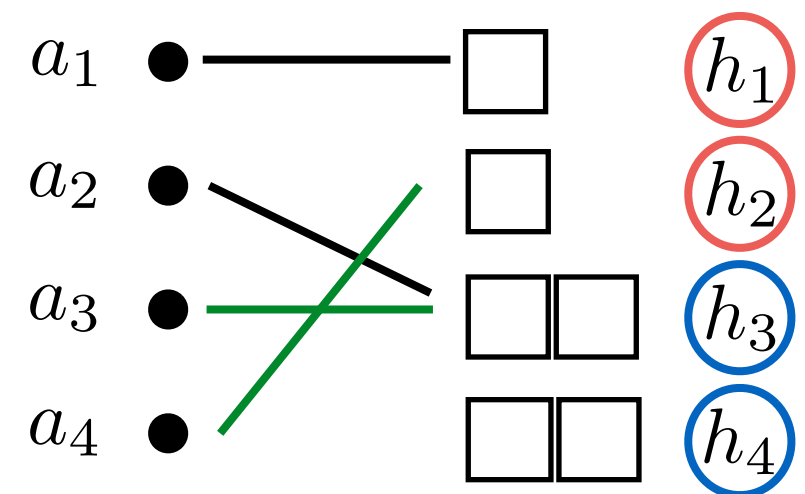
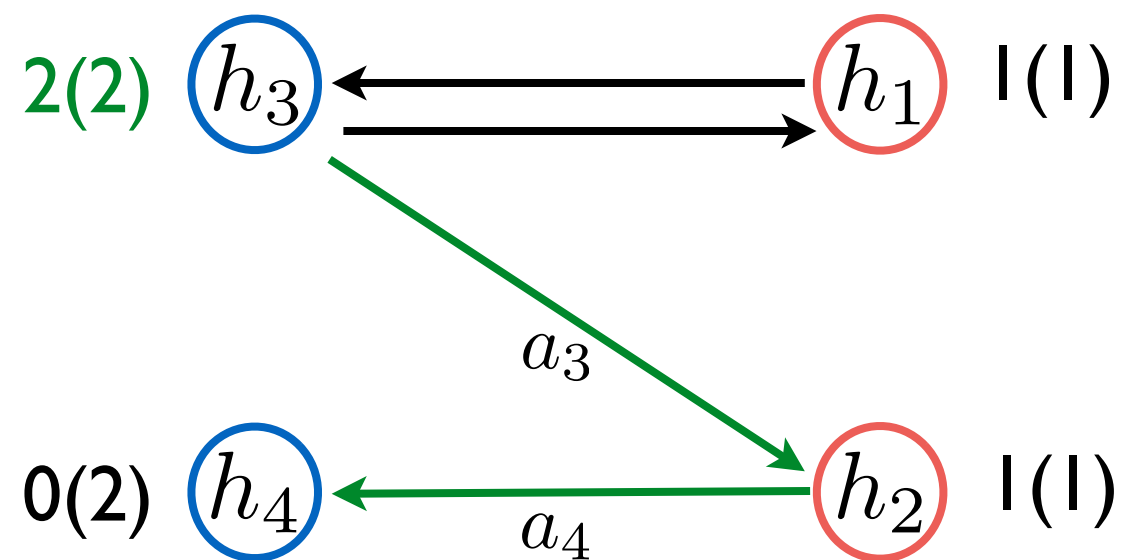
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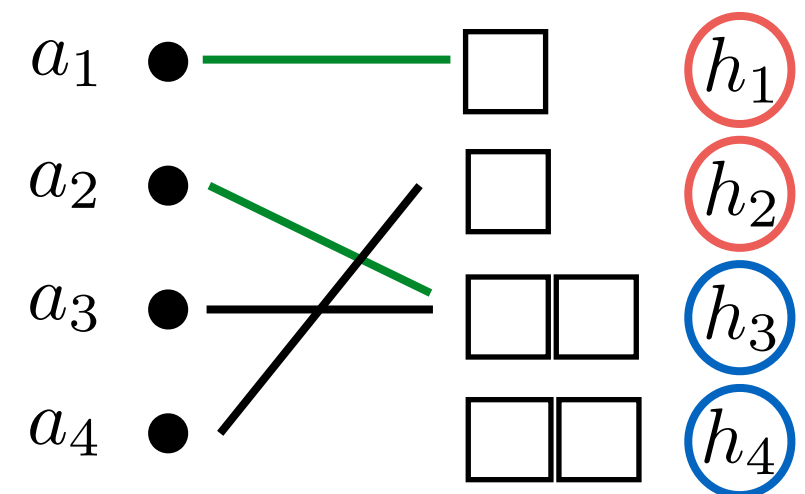
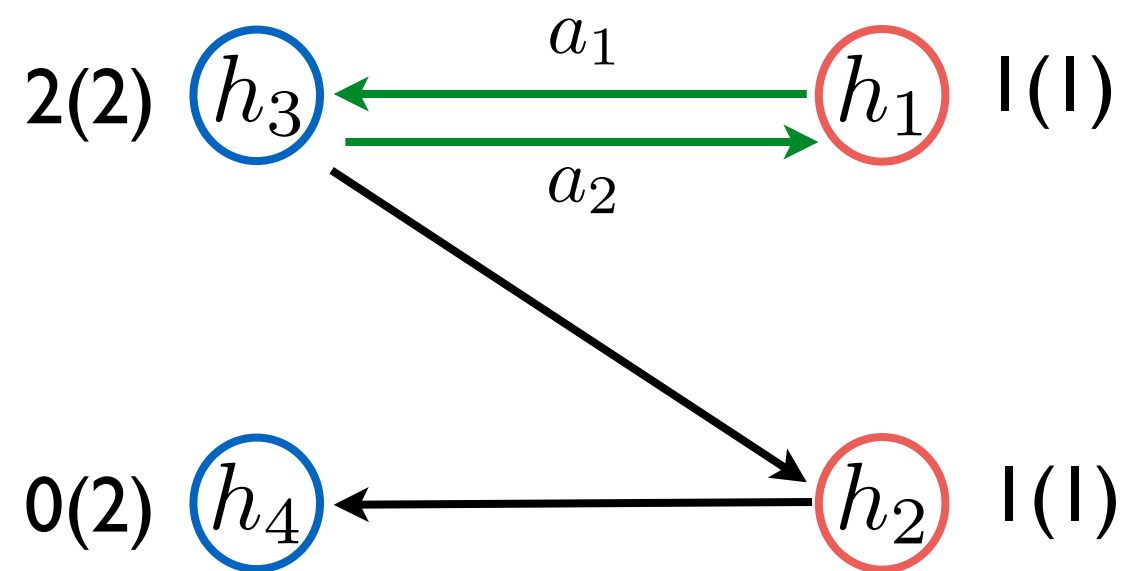
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popular matching

Moving from one Popular Matching to Another

Trick : Cycle Reversal



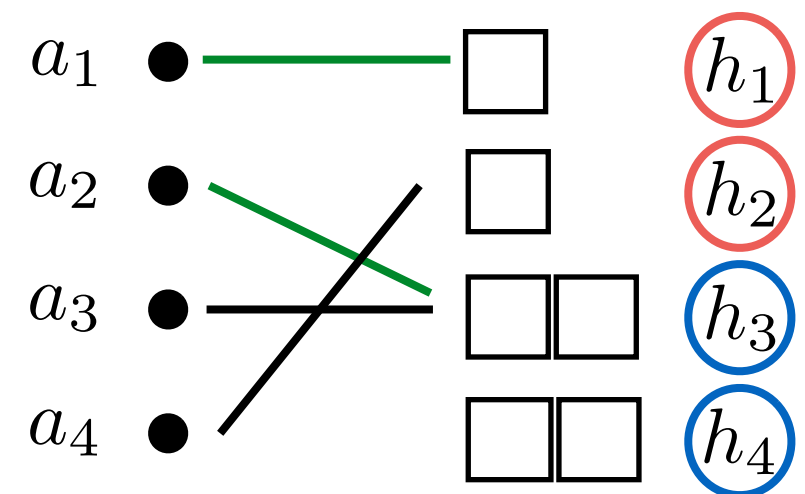
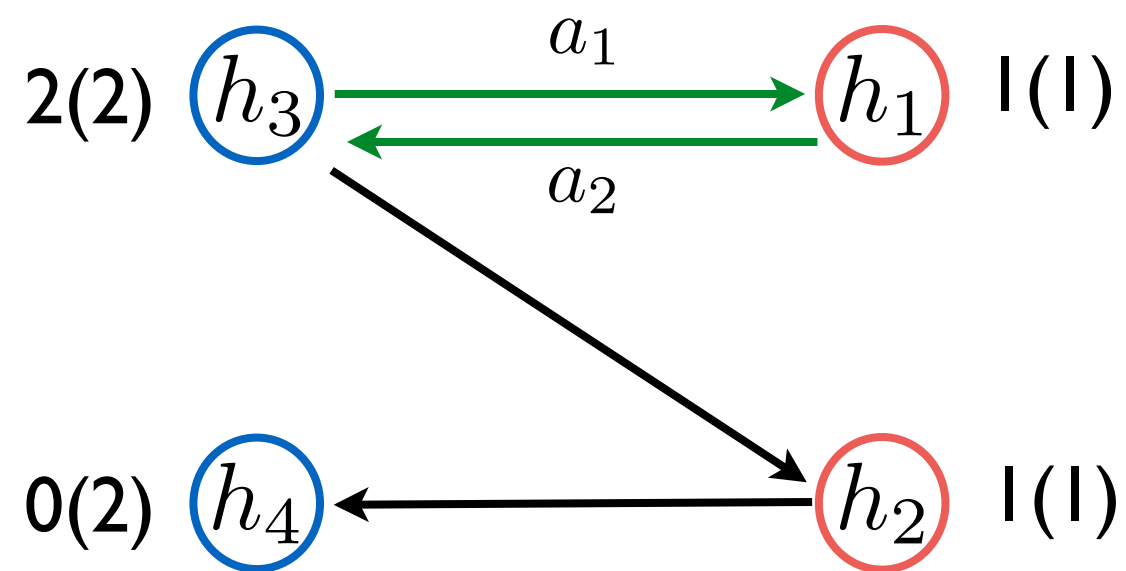
switching graph



popular matching

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Trick : Cycle Reversal



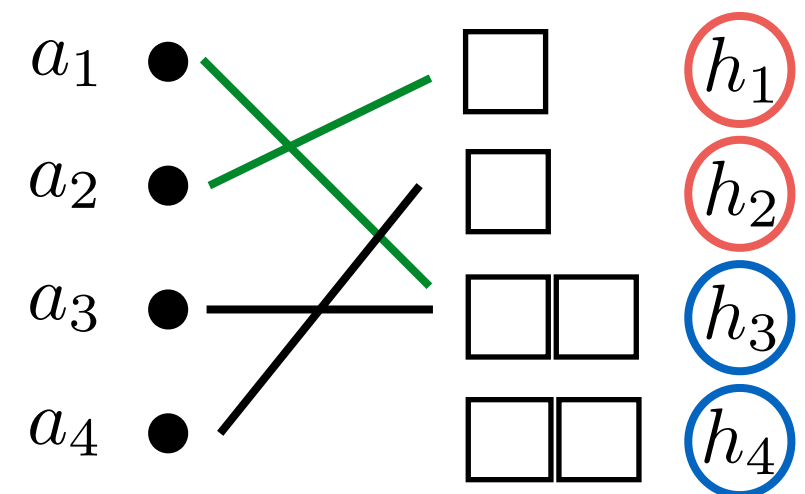
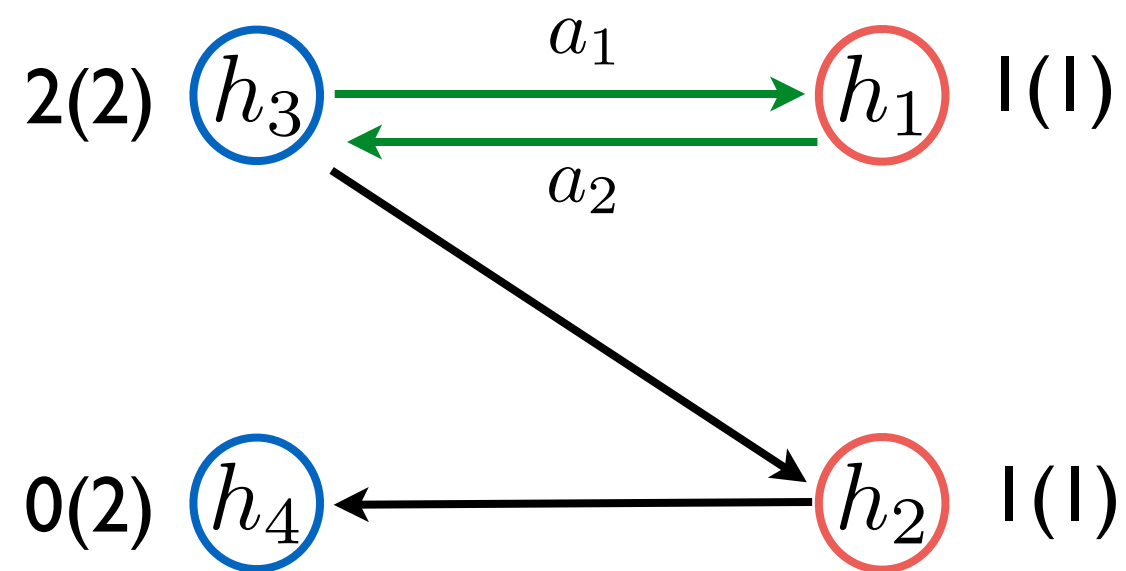
switching graph



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switching graph



popular matching

Moving from one Popular Matching to Another

edge-disjoint union of **switching paths** and
switching cycles could be reversed

Moving from one Popular Matching to Another

edge-disjoint union of **switching paths** and **switching cycles** could be reversed

Careful: **number of switching paths** ending at a house should be less than its **remaining capacity**

Theorem *If G_M is the switching graph of the CHA instance G with respect to a popular matching M , then*

- (i) every switching move on G_M generates another popular matching, and*
- (ii) every popular matching of G can be generated by a switching move on M .*

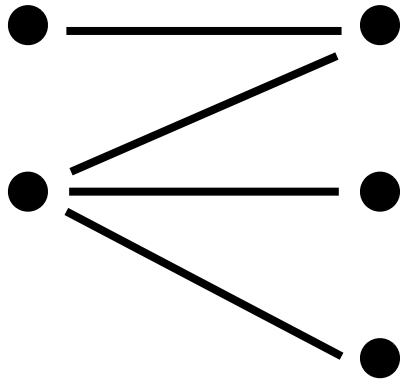
#P-Hardness of Counting Popular Matchings

matchings in
bipartite graph

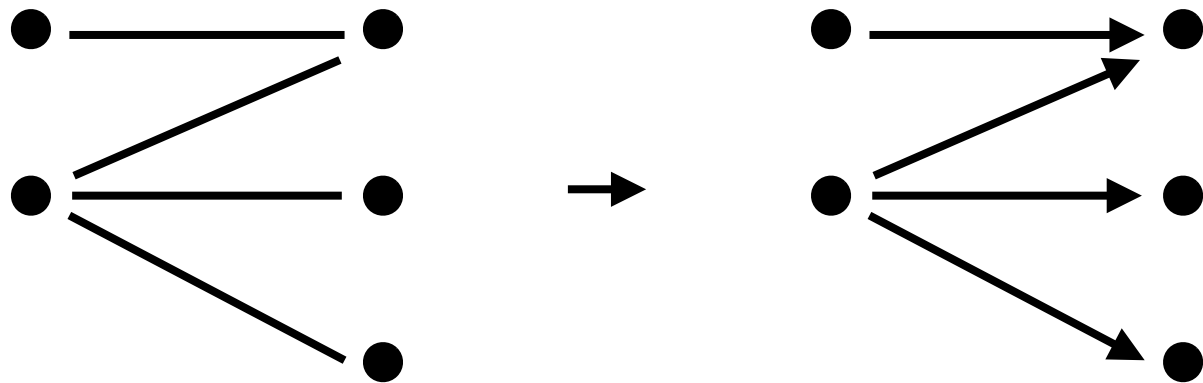


popular matchings
in no-ties, integer
capacity instance

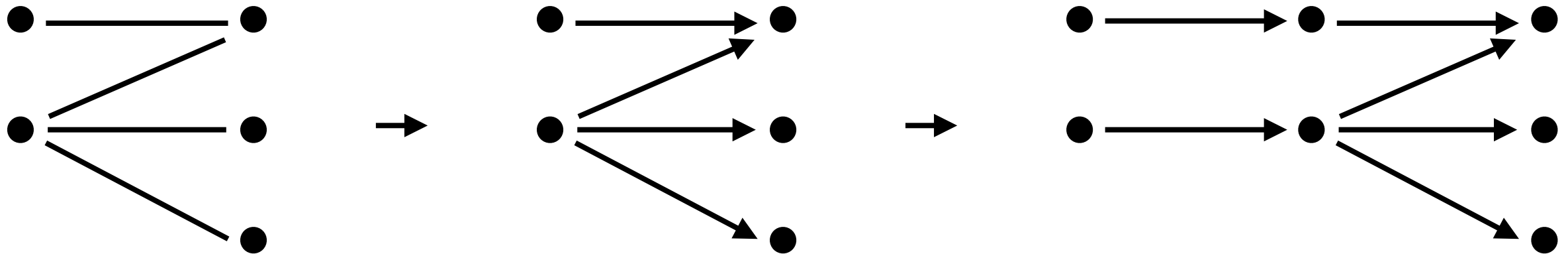
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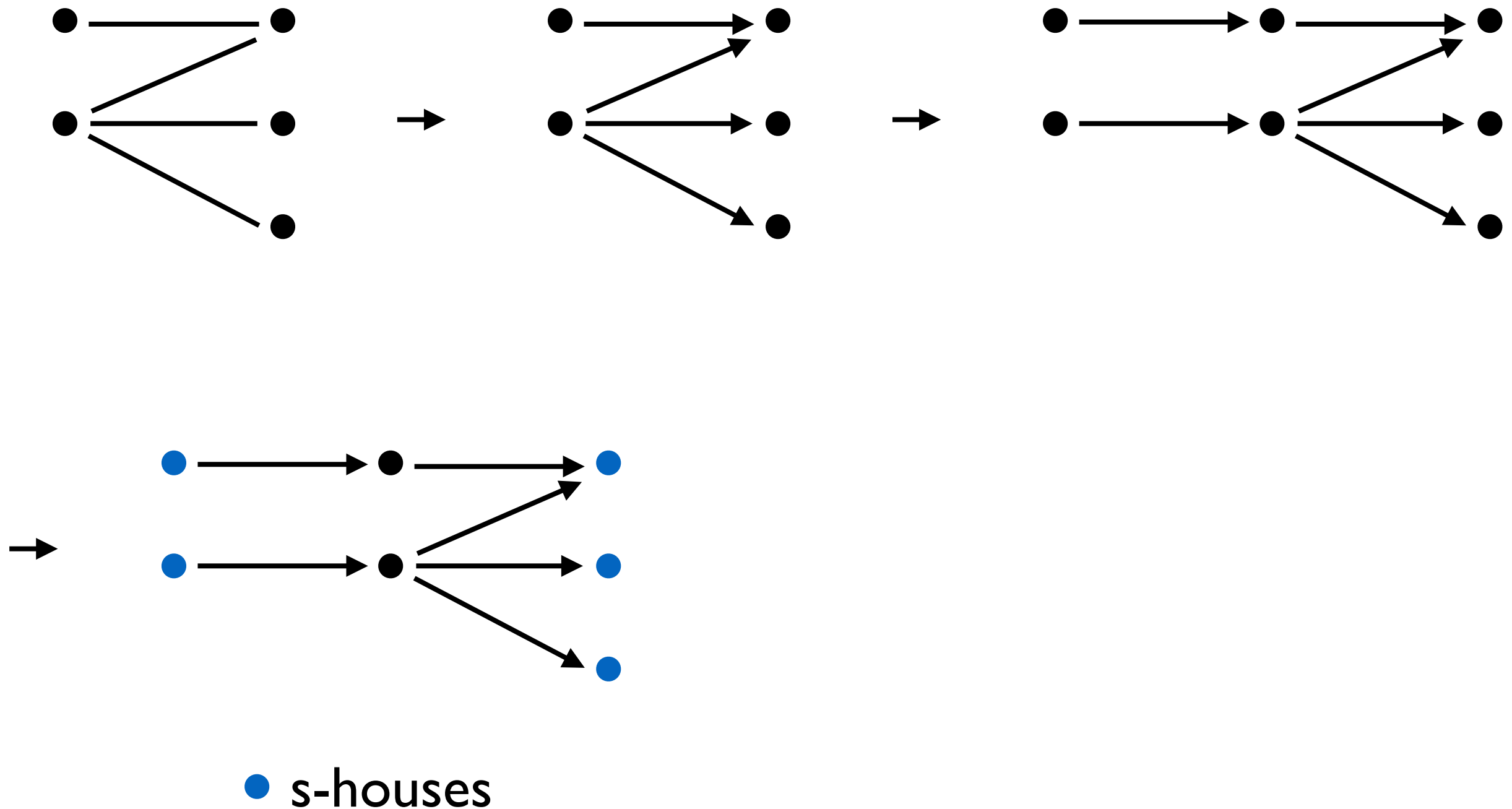
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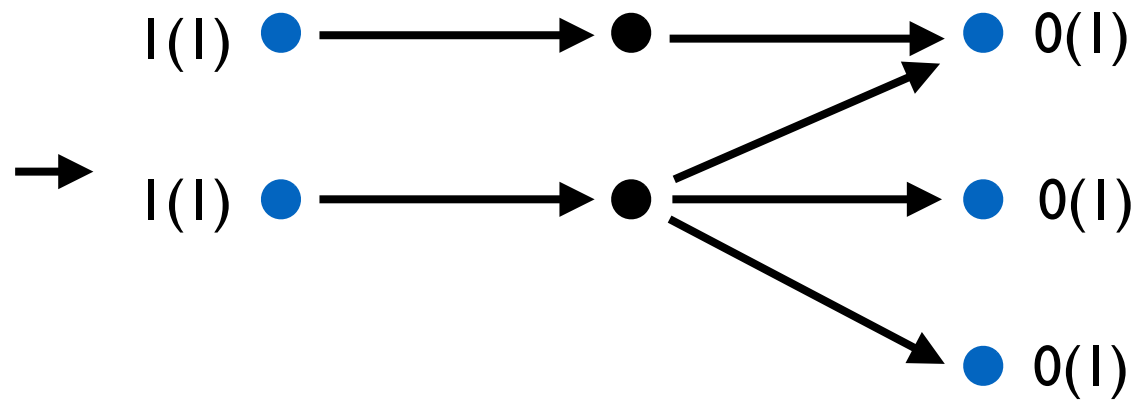
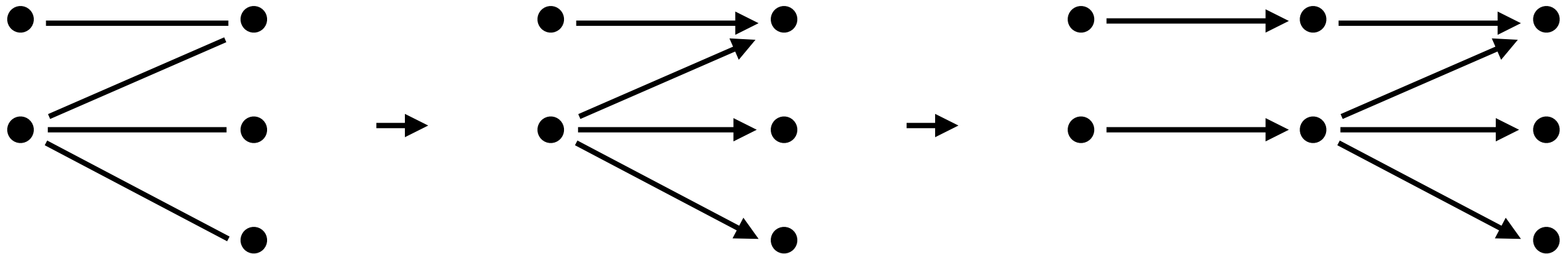
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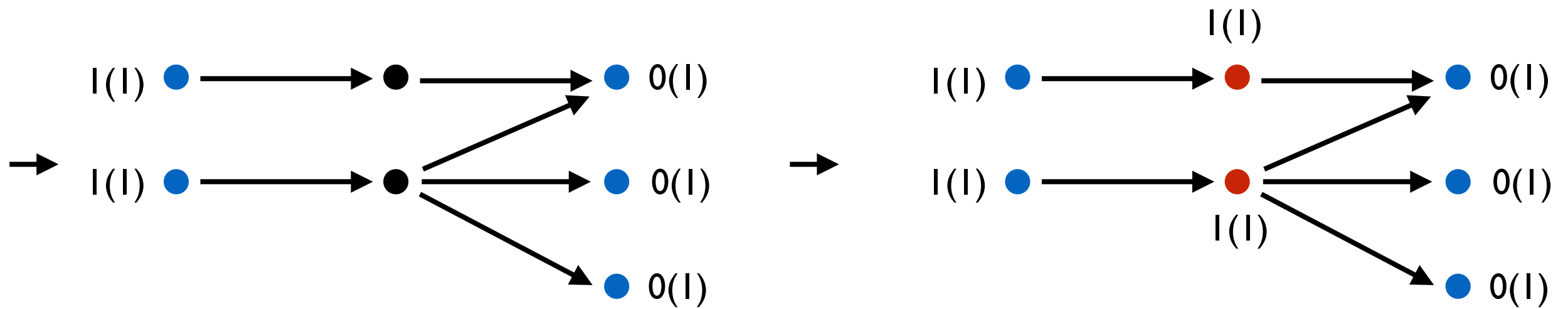
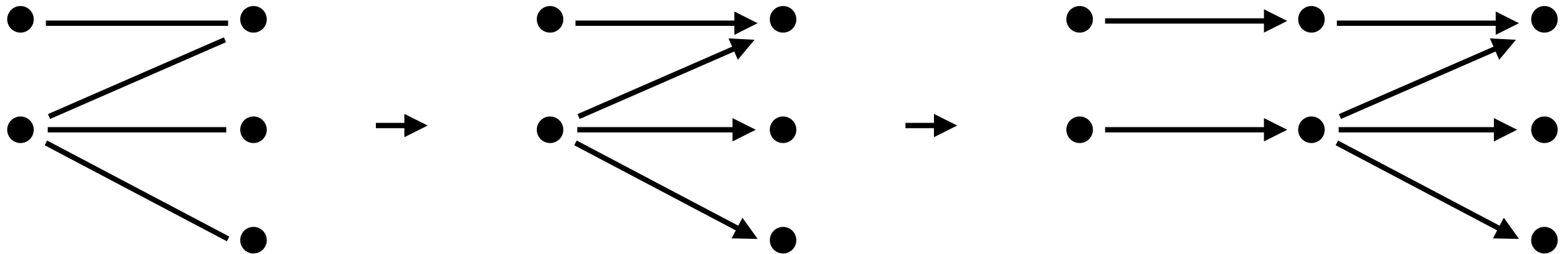


#P-Hardness of Counting Popular Matchings



● s-houses

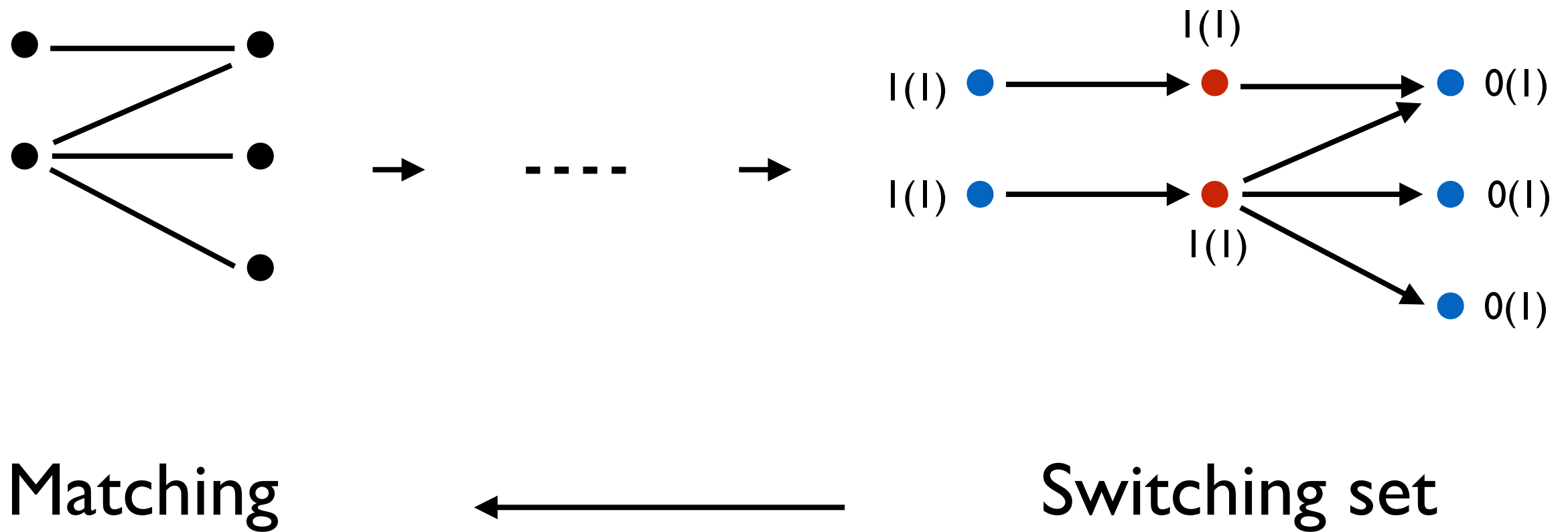
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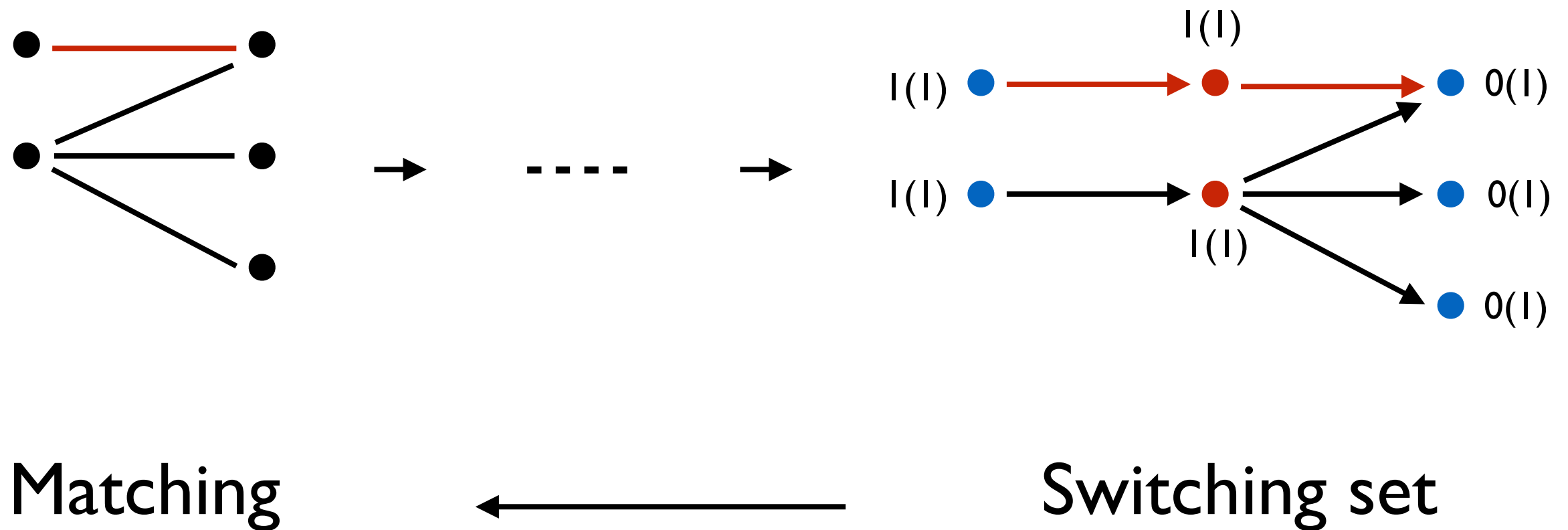
● s-houses

● f-houses

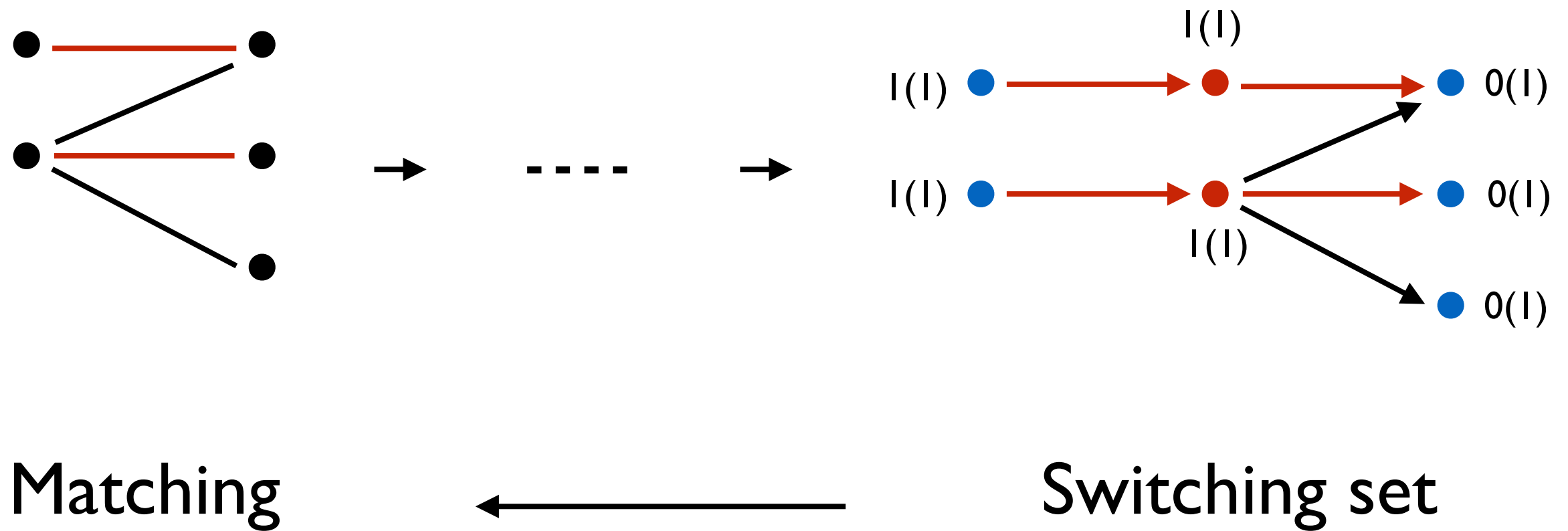
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#P-Hardness of Counting Popular Matchings



Summary

We Give

- A ‘Switching Graph’ Characterisation for no-ties, capacitated case and prove #P-hardness of counting
- We give an FPRAS for ties case

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Open

Algorithm for Counting in no-ties, capacitated case