

Faster Simulated Annealing Algorithms for Combinatorial Counting

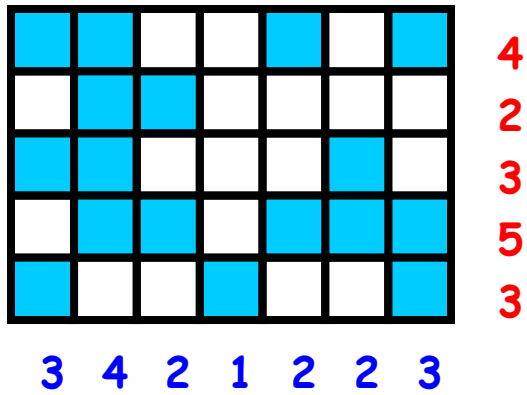
Ivona Bezáková

Rochester Institute of Technology

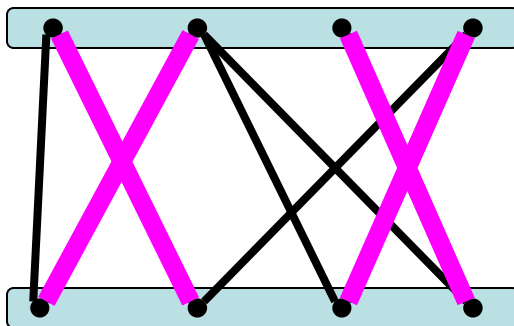
September 22, 2006

Problems

Binary contingency tables

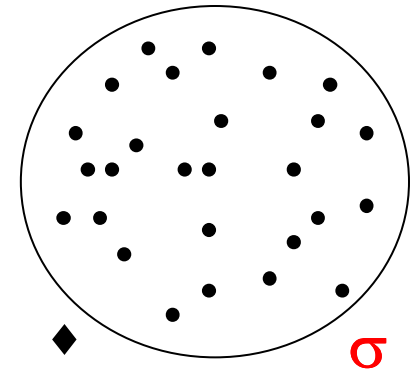


Permanent



Heuristics

Importance sampling

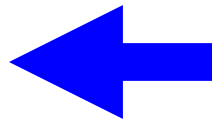
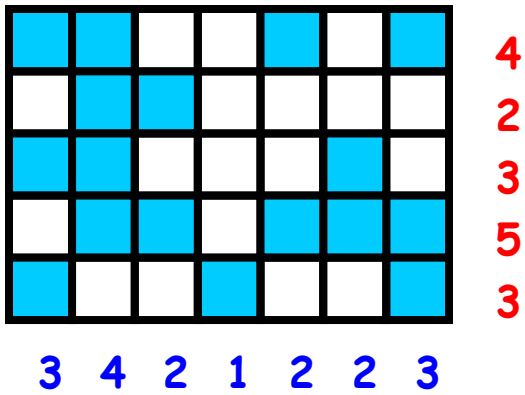


Simulated annealing



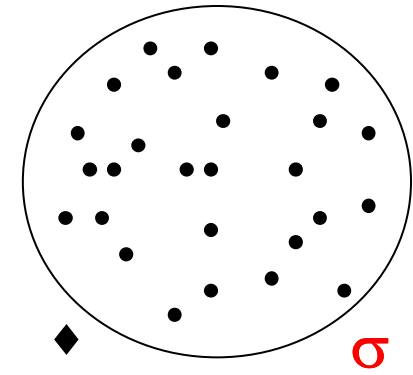
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Binary contingency tables



Heuristics

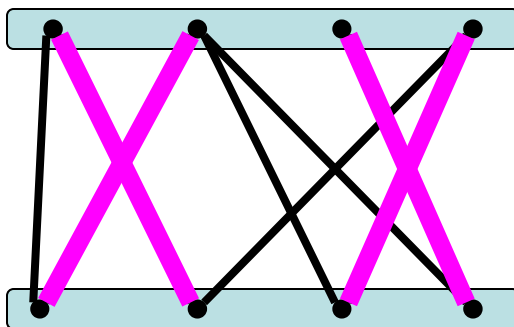
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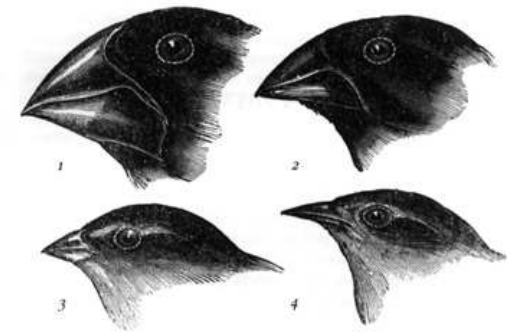
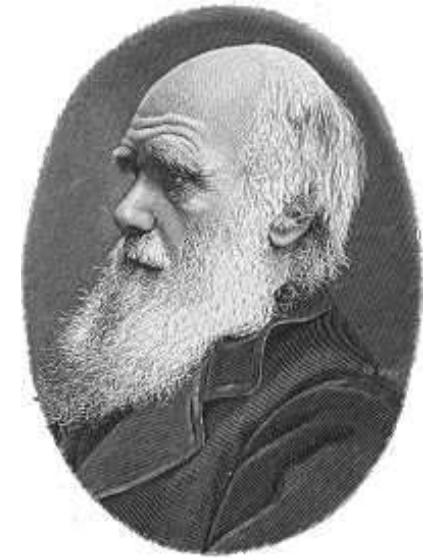


Darwin's Finches



The Voyage of the Beagle

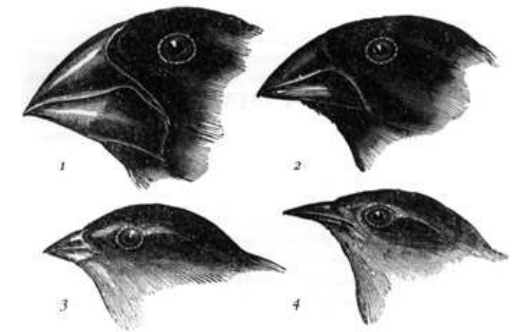
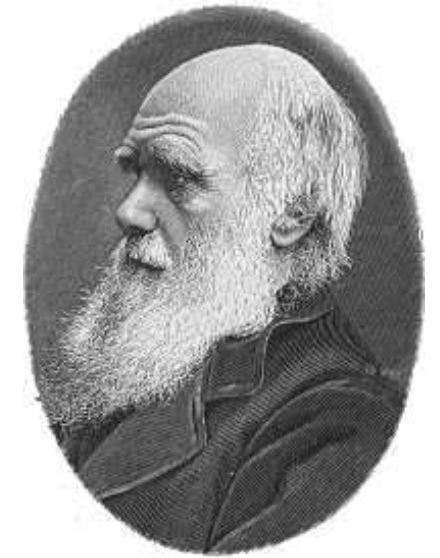
Galápagos archipelago (1835)



Darwin's Finches

Distribution of Darwin's Finches on Visitor Islands

	Santa Cruz	Plaza	Santa Fe	San Cristobal	Espanola	Floreana	Isabela	Fernandina	Santiago	Rabida
Small Ground Finch	●	●	●	●	●	●	●	●	●	●
Medium Ground Finch	●	●	●	●		●	●	●	●	●
Large Ground Finch	●						●	●	●	●
Cactus Ground Finch	●	●	●	●		●	●		●	●
Large Cactus Ground Finch					●					
Sharp-beaked Ground Finch								●	●	
Vegetarian Finch	●			●		●	●	●	●	●
Small Tree Finch	●		●	●		●	●	●	●	●
Medium Tree Finch						●				
Large Tree Finch	●					●	●	●	●	●
Woodpecker Finch	●			●			●		●	
Mangrove Finch							●	●		
Warbler Finch	●		●	●	●	●	●	●	●	●



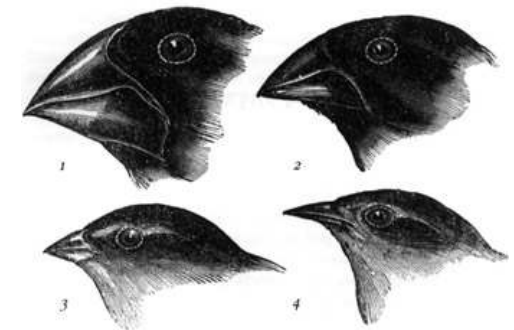
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Small Ground Finch	•	•	•	•	•	•	•	•	•	•
Medium Ground Finch	•	•	•	•	•	•	•	•	•	•
Large Ground Finch	•						•	•	•	•
Cactus Ground Finch	•	•	•	•	•	•	•	•	•	•
Large Cactus Ground Finch					•					
Sharp-beaked Ground Finch								•	•	
Vegetarian Finch	•			•		•	•	•	•	•
Small Tree Finch	•		•	•		•	•	•	•	•
Medium Tree Finch						•				
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Warbler Finch	•		•	•	•	•	•	•	•	•



8



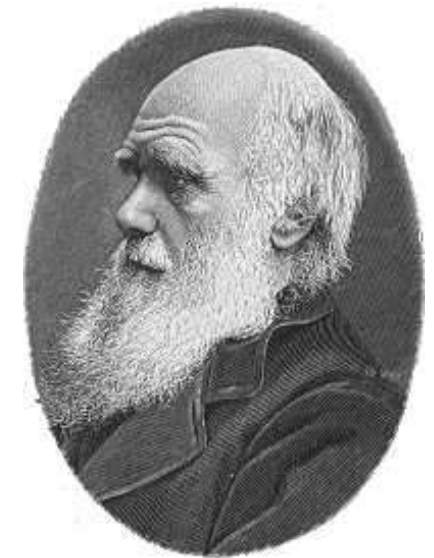
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Darwin's Finches

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Woodpecker Finch	●			●			●		●	
Mangrove Finch							●	●		
Warbler Finch	●		●	●	●	●	●	●	●	●

9 3 5 7 3 8 10 9 10 8



10
9
6
8
2
3
7
8
1
6
4
2
10

chance

OR

competitive pressures

?

Binary Contingency Tables

Given: marginals (**row sums**, **column sums**)

Goal:

- sample tables uniformly at random
- count tables

							4
							2
							3
							5
							3
3	4	2	1	2	2	3	

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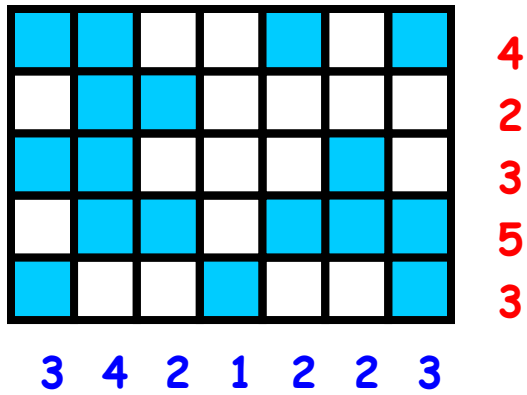
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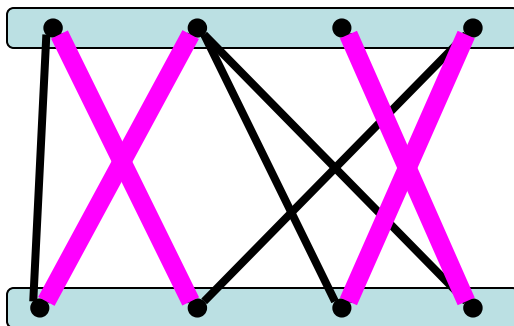
■	■	□	□	■	□	■	4
□	■	■	□	□	□	□	2
■	■	□	□	□	■	□	3
□	■	■	□	■	■	■	5
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Binary contingency tables

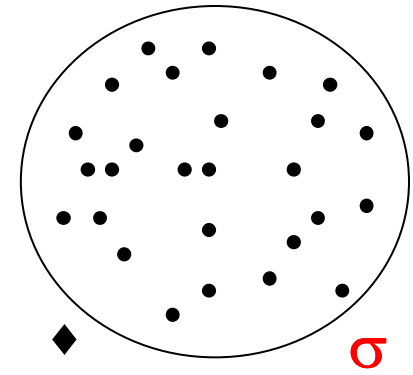


Permanent

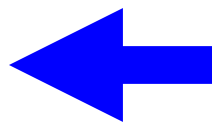


Heuristics

Importance sampling



Simulated annealing



Permanent of an $n \times n$ matrix A

$$\text{Per}(A) = \sum_{\pi \in S_n} \prod_{i=1}^n a_{i, \pi(i)}$$

in 1812:



Augustin-Louis
Cauchy



Jacques
Binet

$A =$

5	1	0	3	4
2	0	1	4	4
5	7	9	2	2
1	1	6	1	0
3	3	0	8	7

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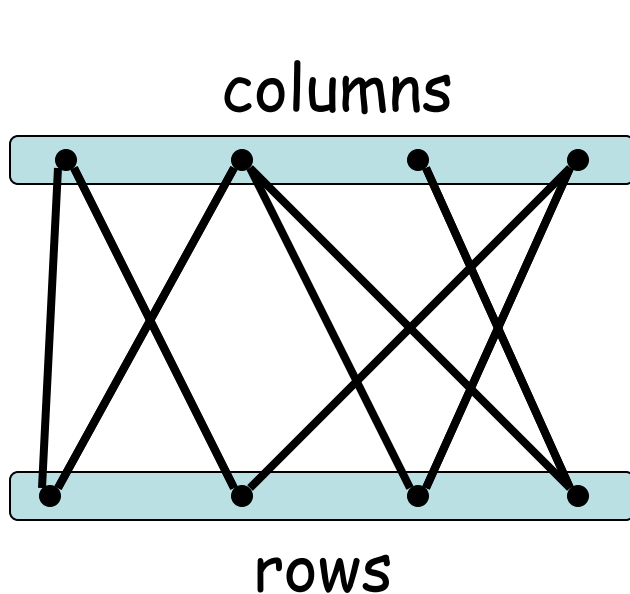
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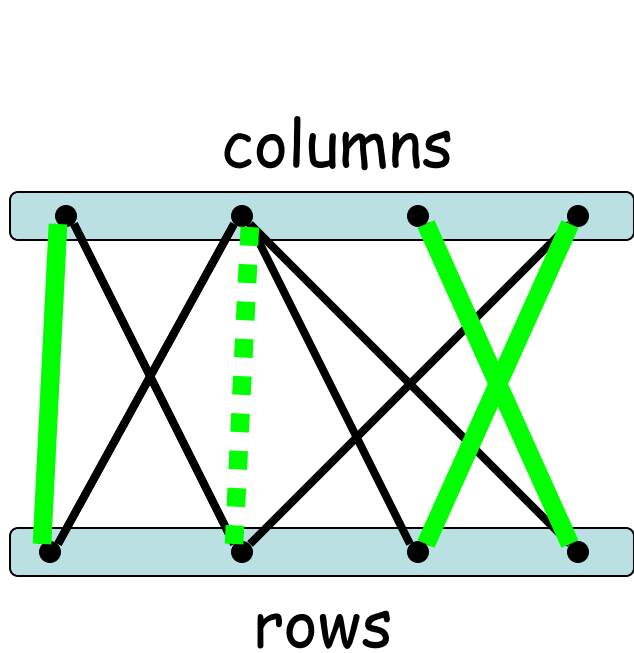
columns

1	1	0	0
1	0	0	1
0	1	0	1
0	1	1	0

rows

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columns π

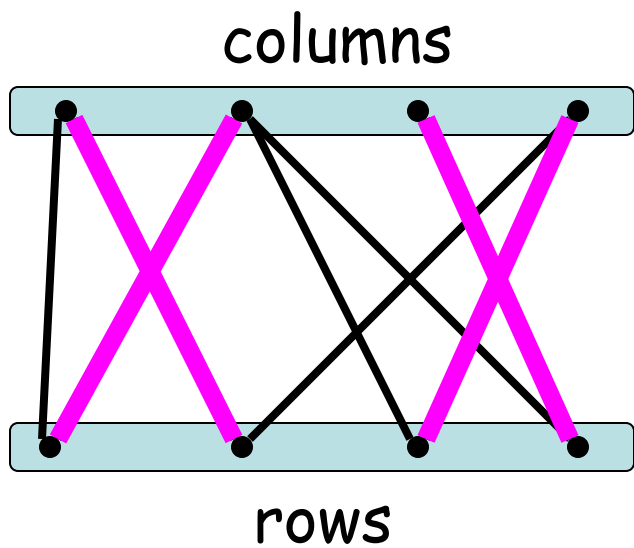
1	1	0	0
1	0	0	1
0	1	0	1
0	1	1	0

rows

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1



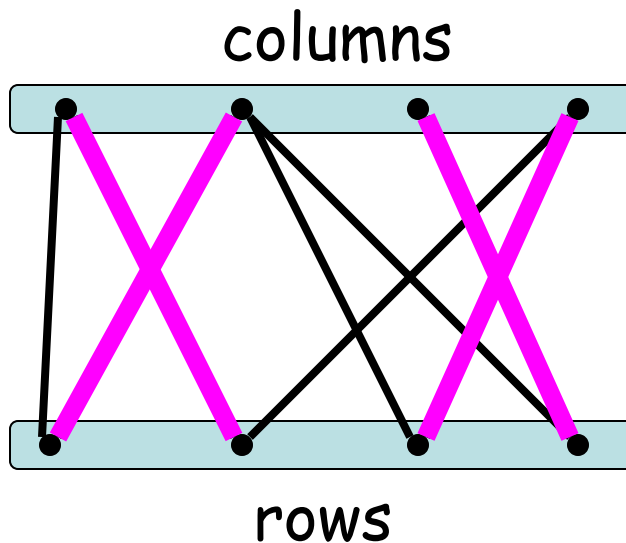
columns π

1	1	0	0
1	0	0	1
0	1	0	1
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ROWS

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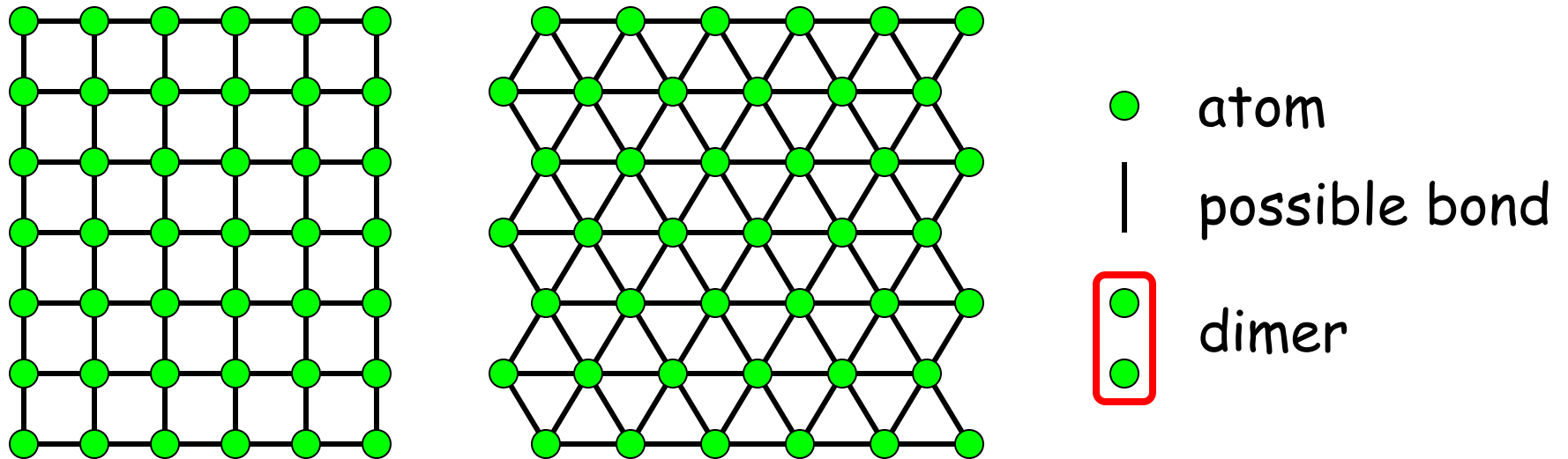


Count
perfect matchings
of a bipartite graph

Permanent: Counting Perfect Matchings

[Kasteleyn '63] planar graphs in polytime

Dimer model



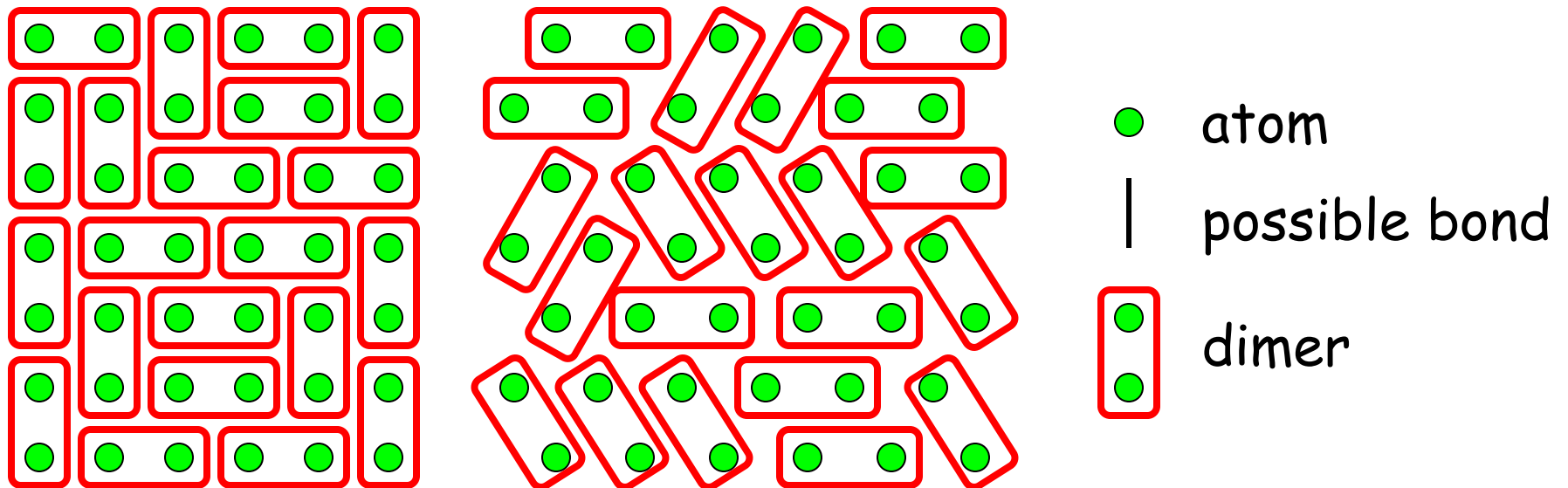
Partition function = permanent (in bipartite case)

- for computation of thermodynamic properties

Permanent: Counting Perfect Matchings

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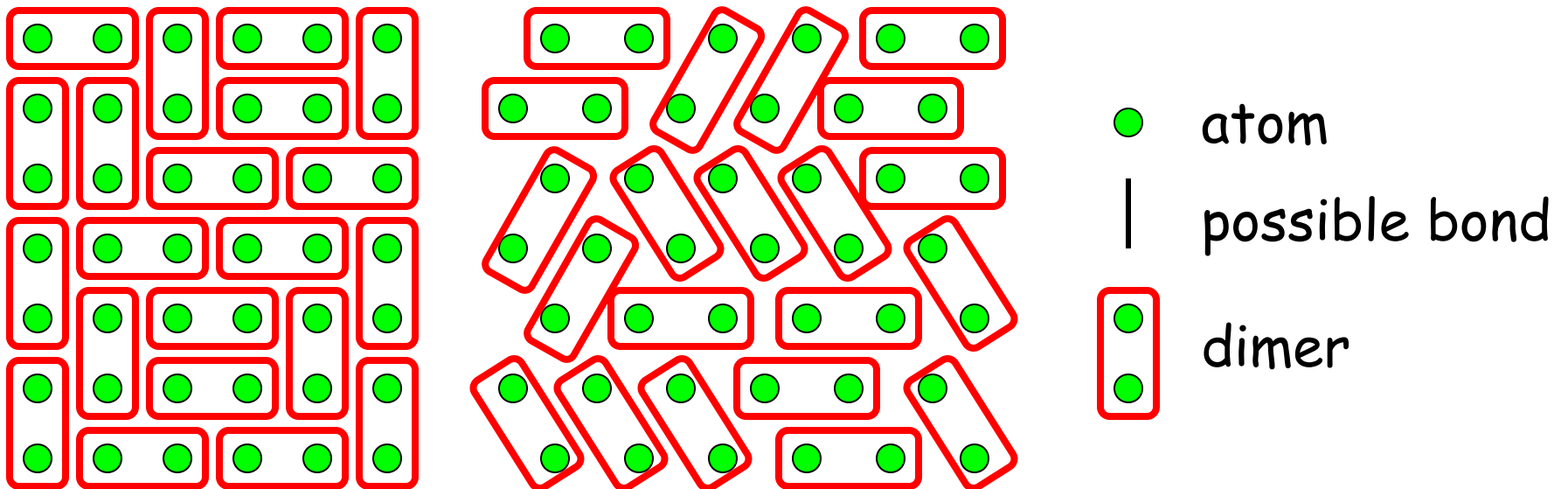
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Permanent: Counting Perfect Matchings

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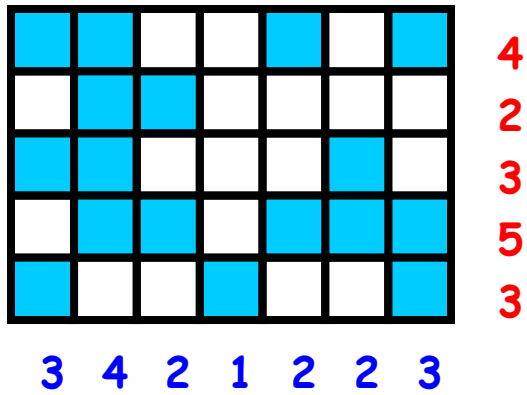
Dimer model



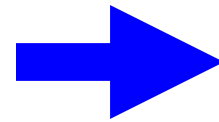
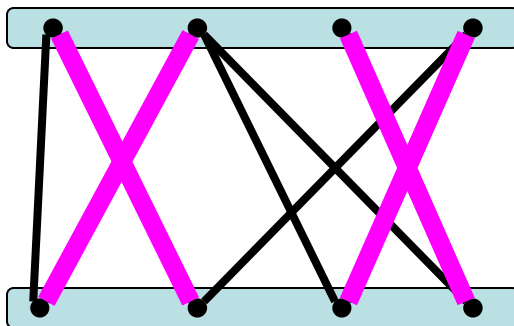
[Valiant '79] **#P-complete** for general (bipartite) graphs

Problems

Binary contingency tables

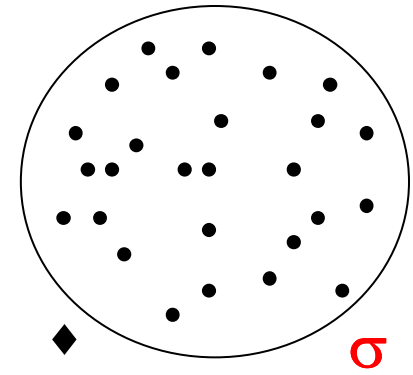


Permanent



Heuristics

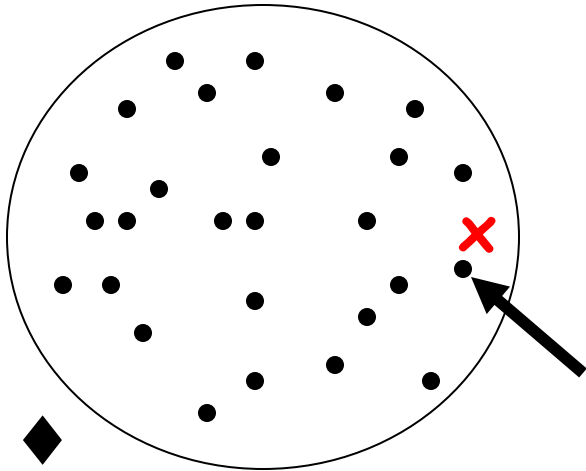
Importance sampling



Simulated annealing



Importance Sampling for counting problems



Probability distribution σ
on the points + \diamond

with **positive** probability $\sigma(x) > 0$

Random variable

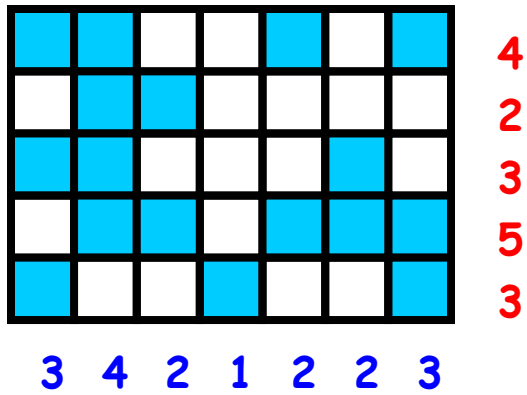
$$\eta(s) = \begin{cases} 1/\sigma(s) & \text{if } s \text{ in the set} \\ 0 & \text{if } s \text{ is } \diamond \end{cases}$$

Unbiased estimator

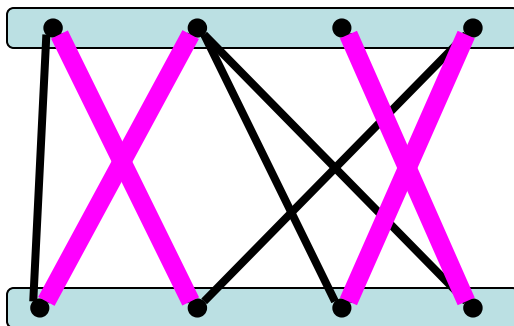
$$E[\eta] = \sum \sigma(x) \cdot 1/\sigma(x) = \text{size of the set}$$

Problems

Binary contingency tables

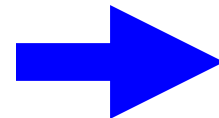
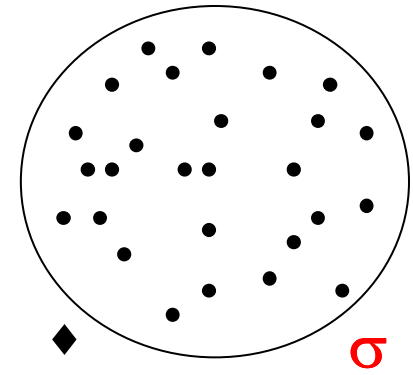


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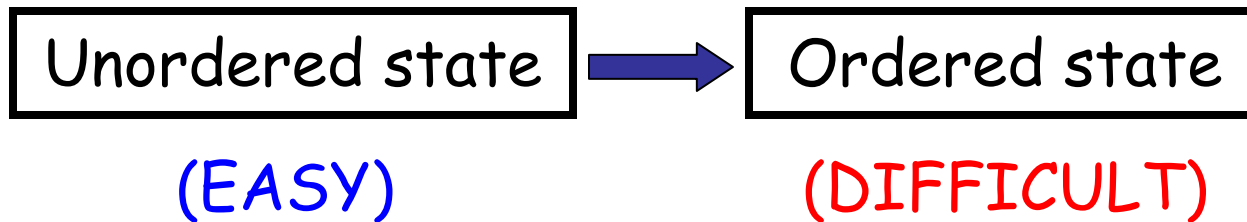
Simulated annealing



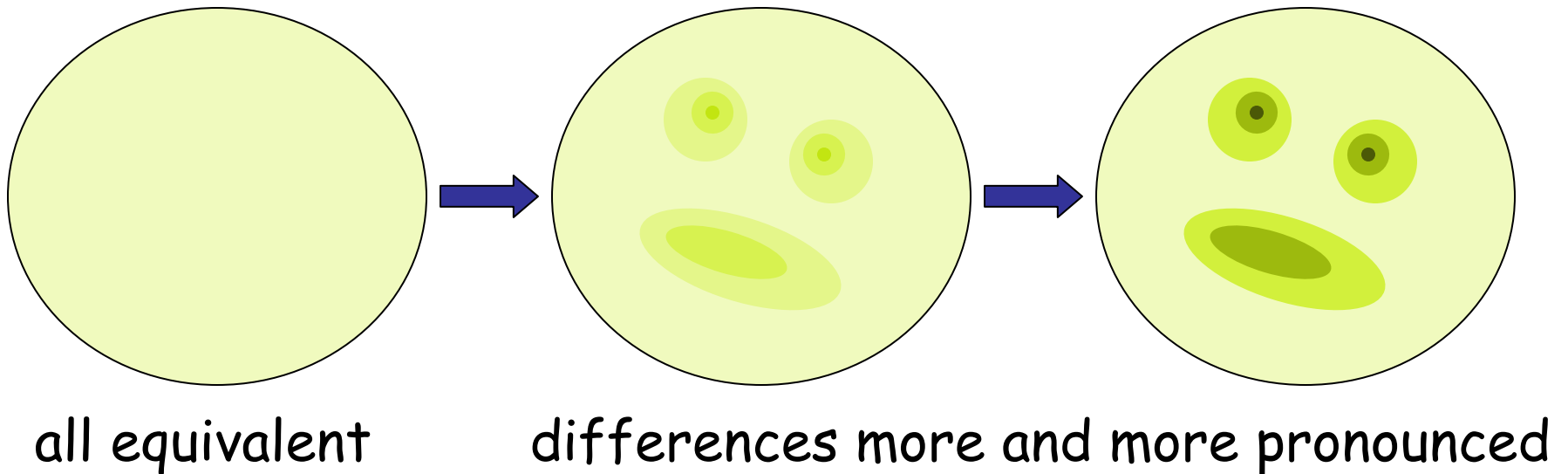
Simulated Annealing



[Kirkpatrick-Gelatt-Vecchi '83]
[Černý '85]

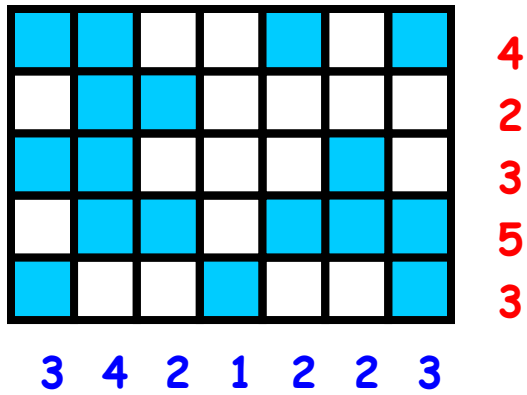


Cooling:

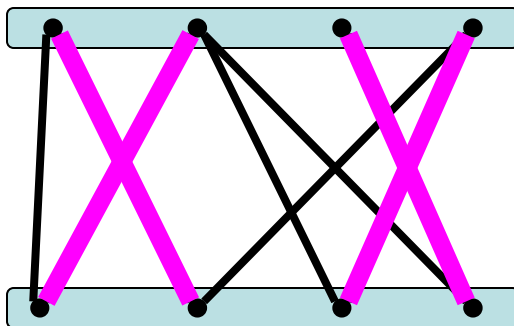


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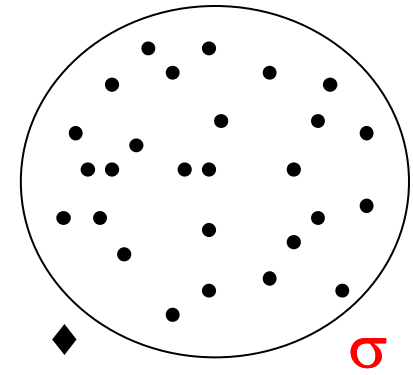


Permanent



Heuristics

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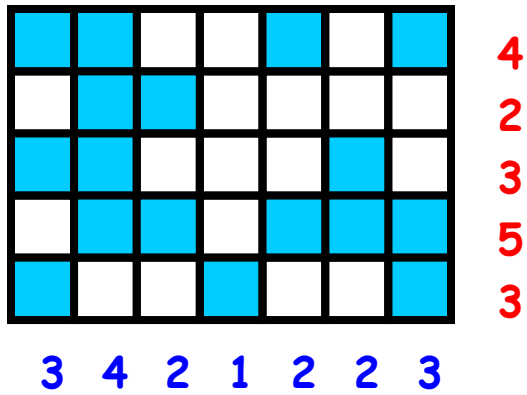


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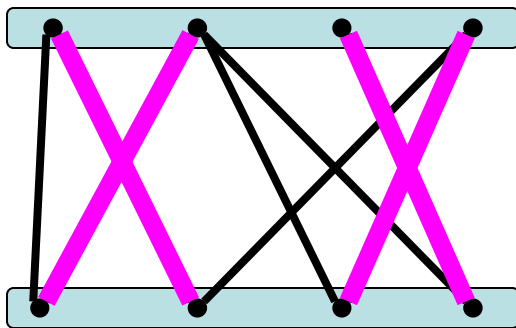


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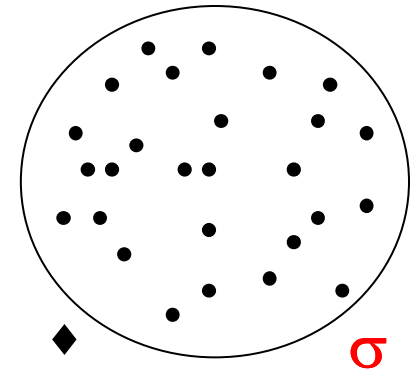


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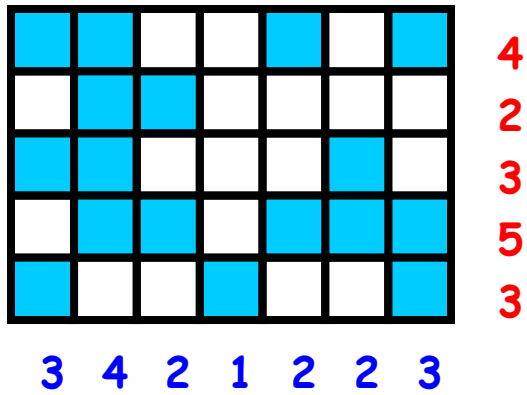


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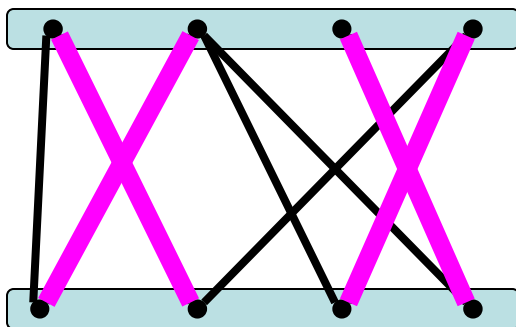


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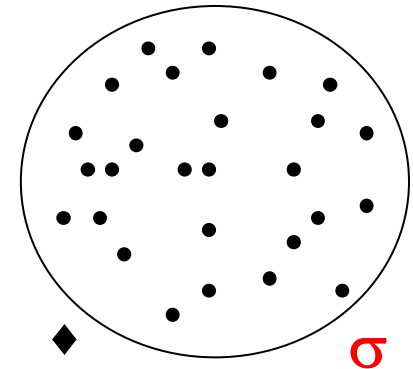


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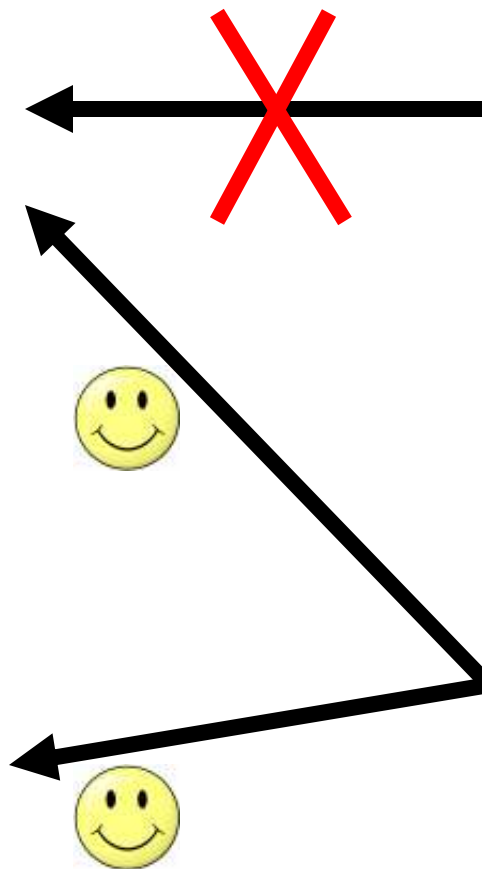


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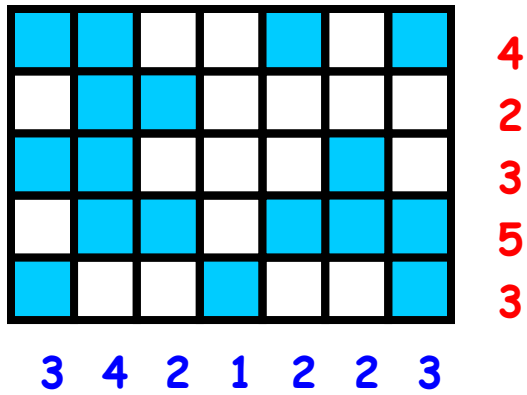


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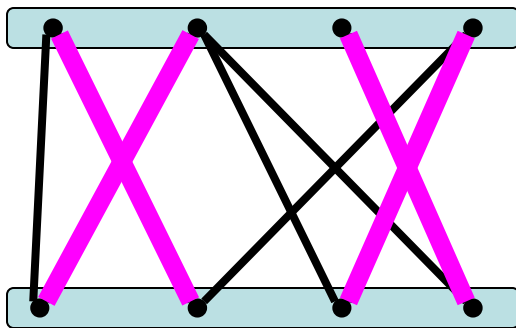


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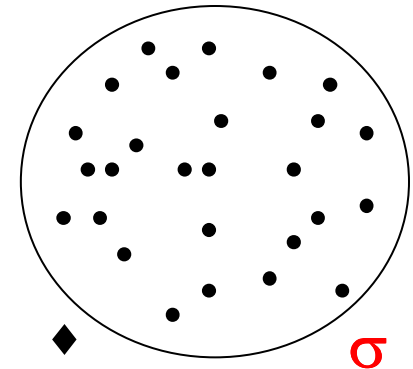


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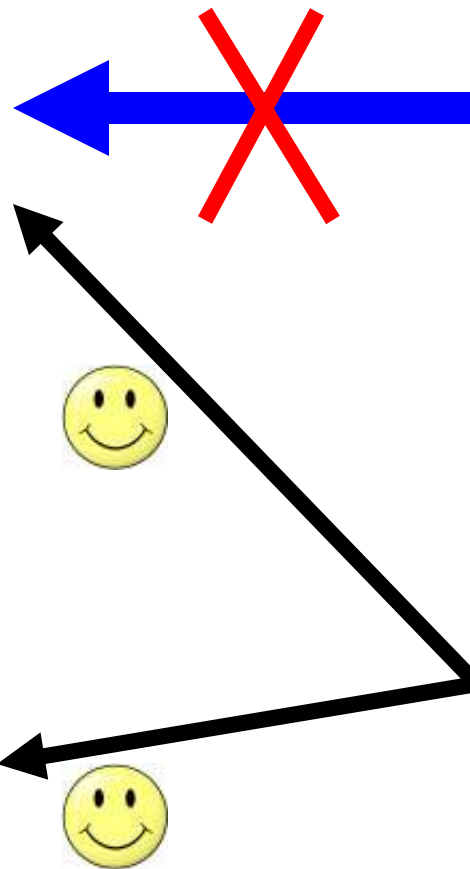


Heuristics

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Sequential Importance Sampling for BCT

[Chen-Diaconis-Holmes-Liu '05]

a specific σ

- fill table column-by-column
- assign each column ignoring other column sums

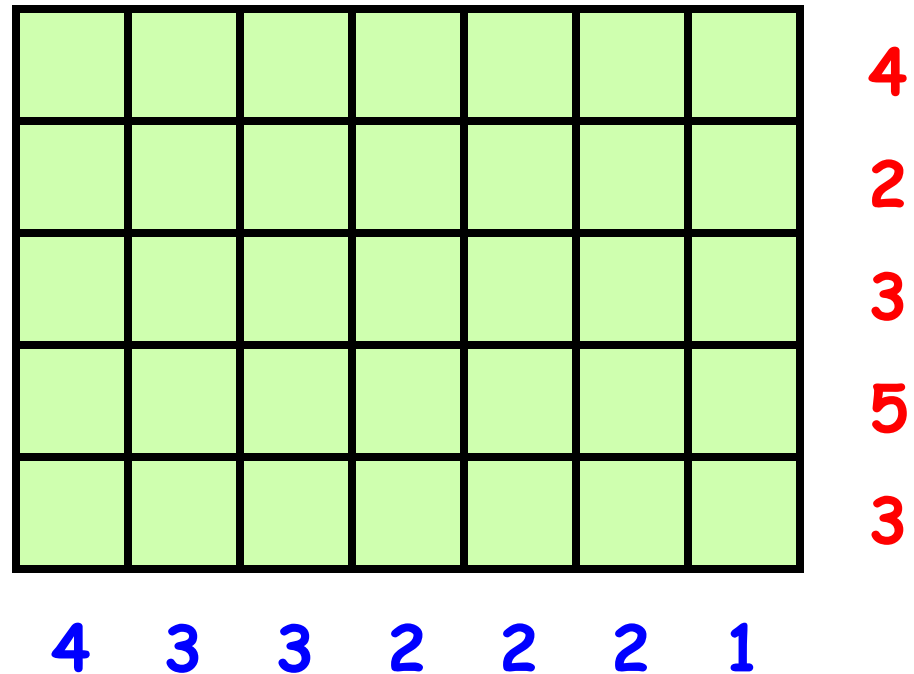
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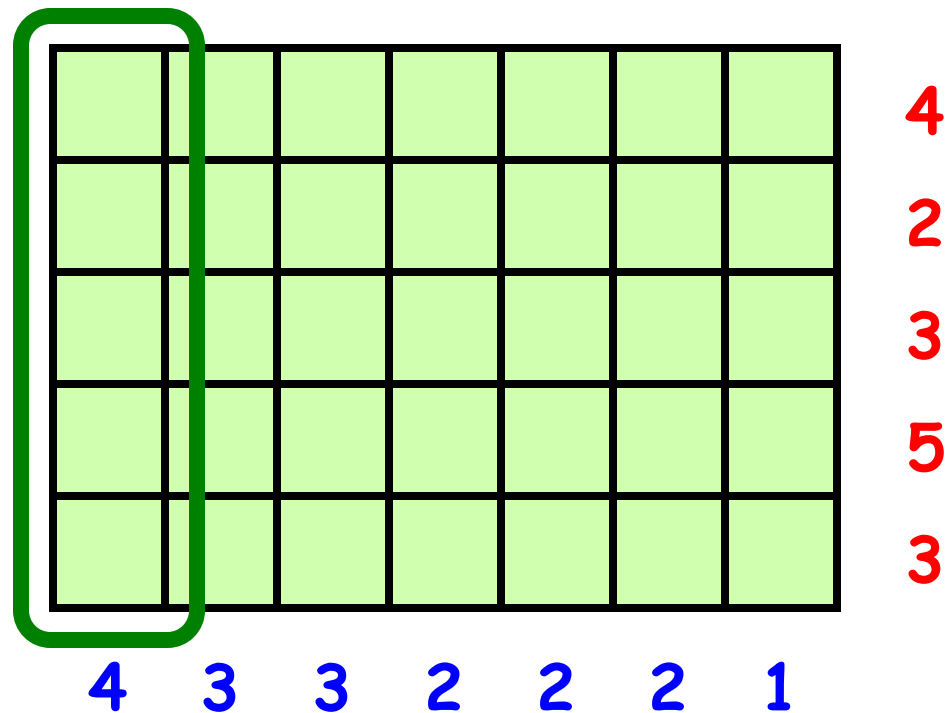


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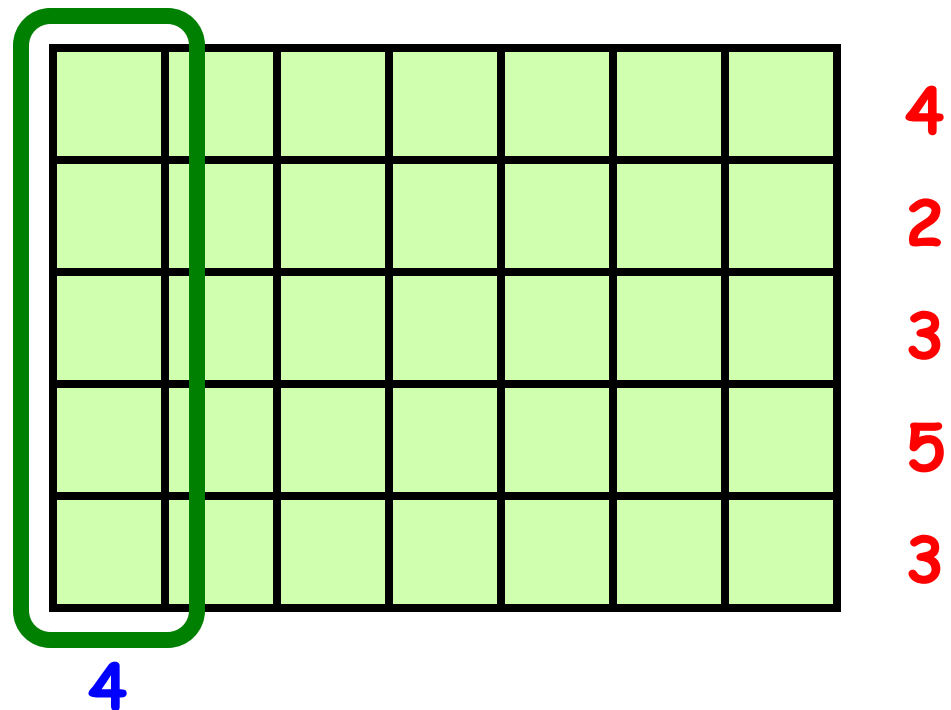


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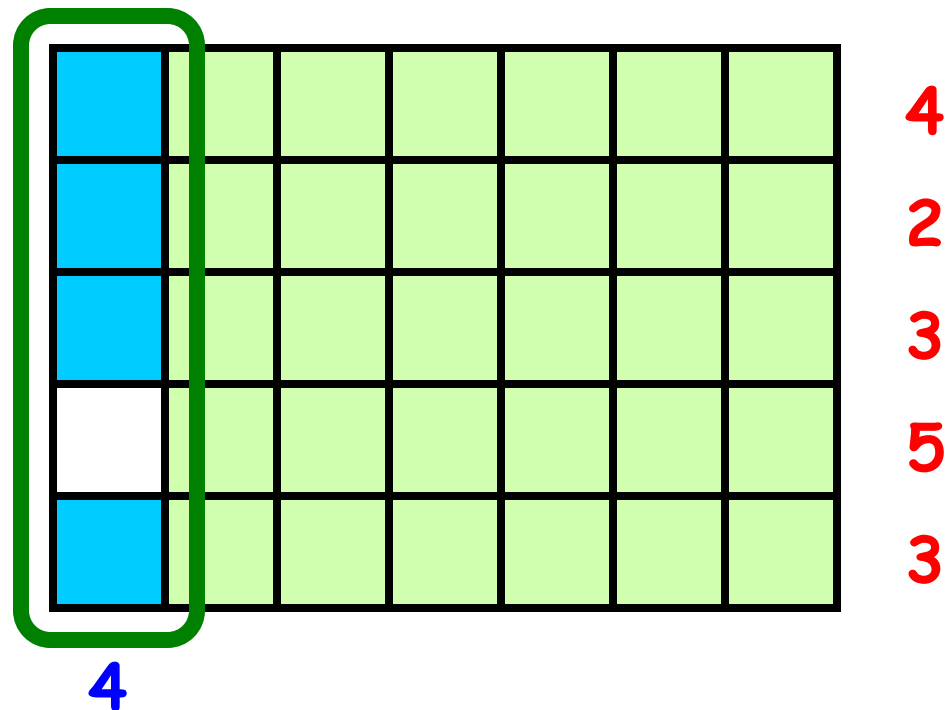
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assign the column
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proportional to

$$\prod r_i / (n - r_i)$$

where product
ranges over i : rows
with assignment 1



Sequential Importance Sampling for BCT

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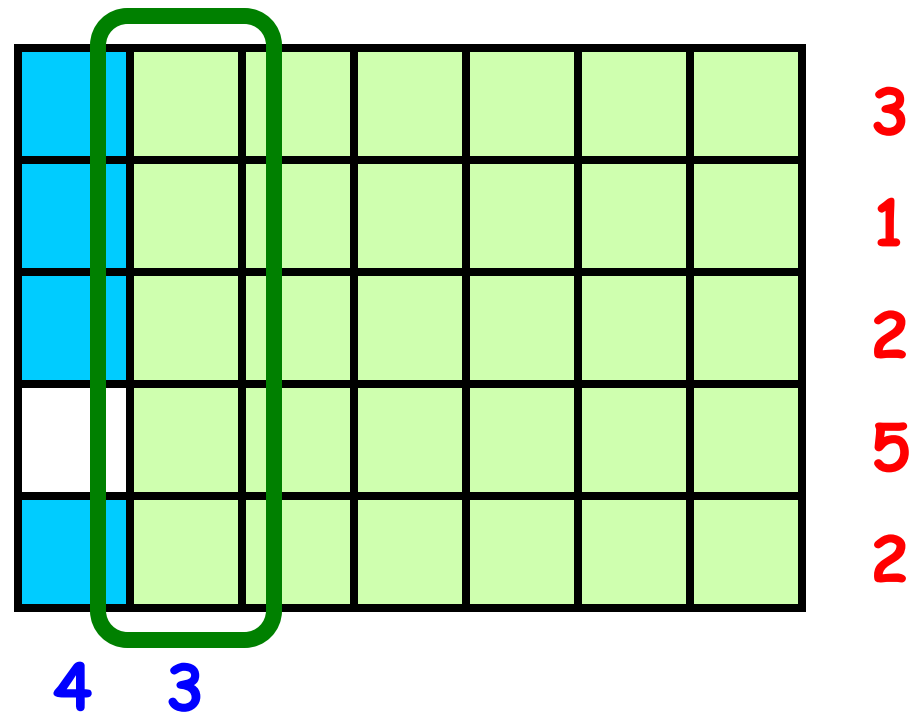
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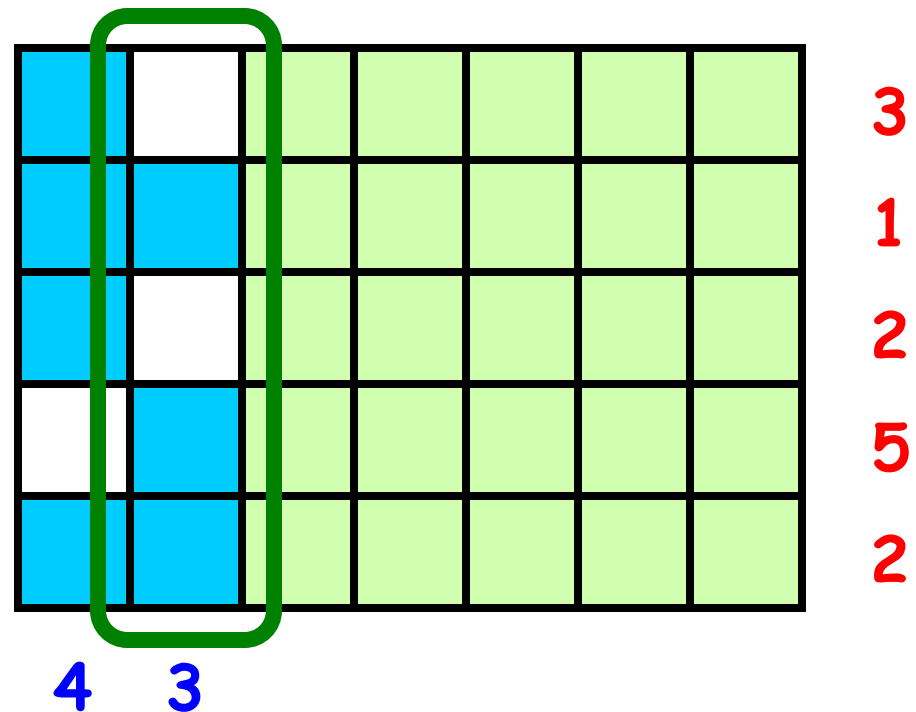
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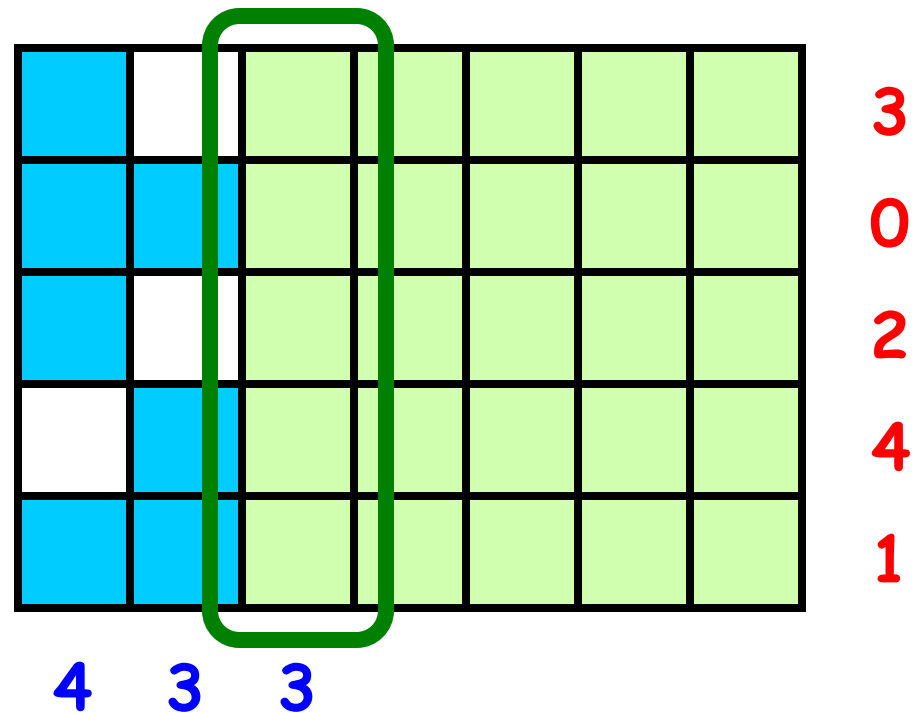
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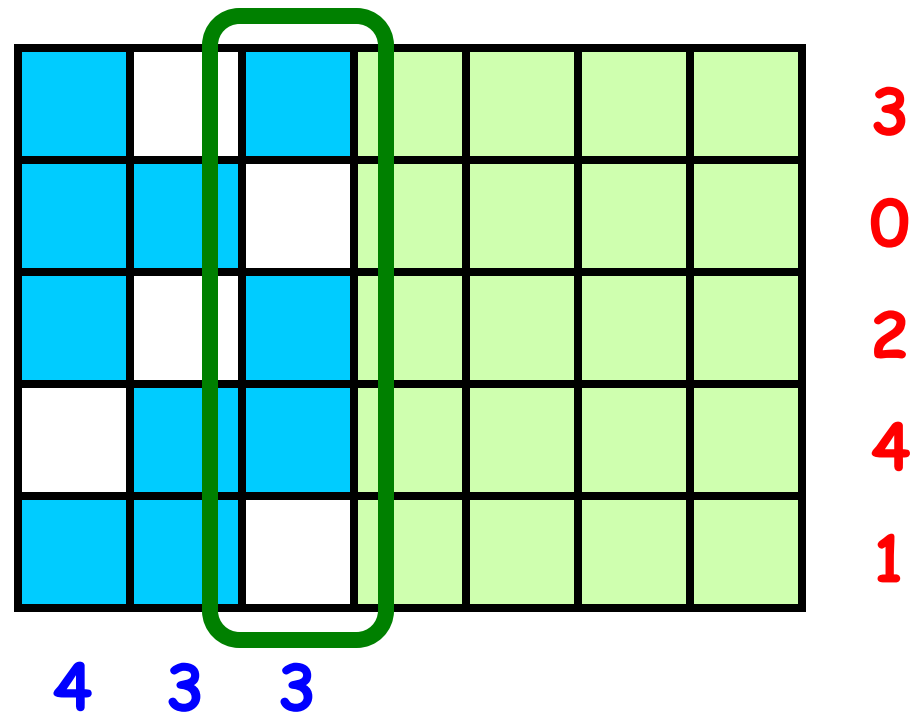
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ranges over i : rows
with assignment 1



Sequential Importance Sampling for BCT

[Chen-Diaconis-Holmes-Liu '05]

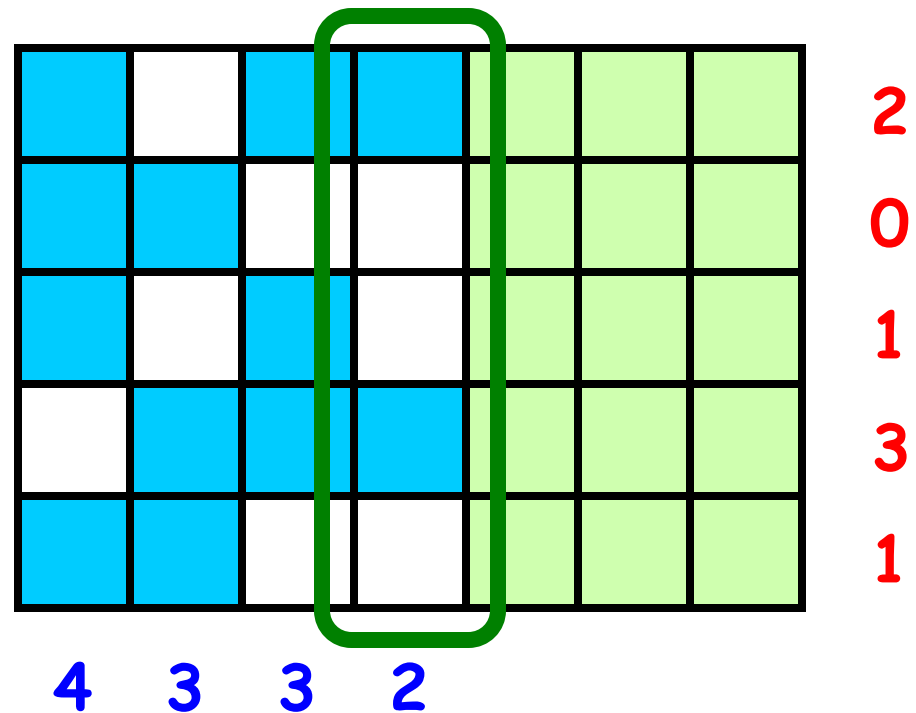
a specific σ

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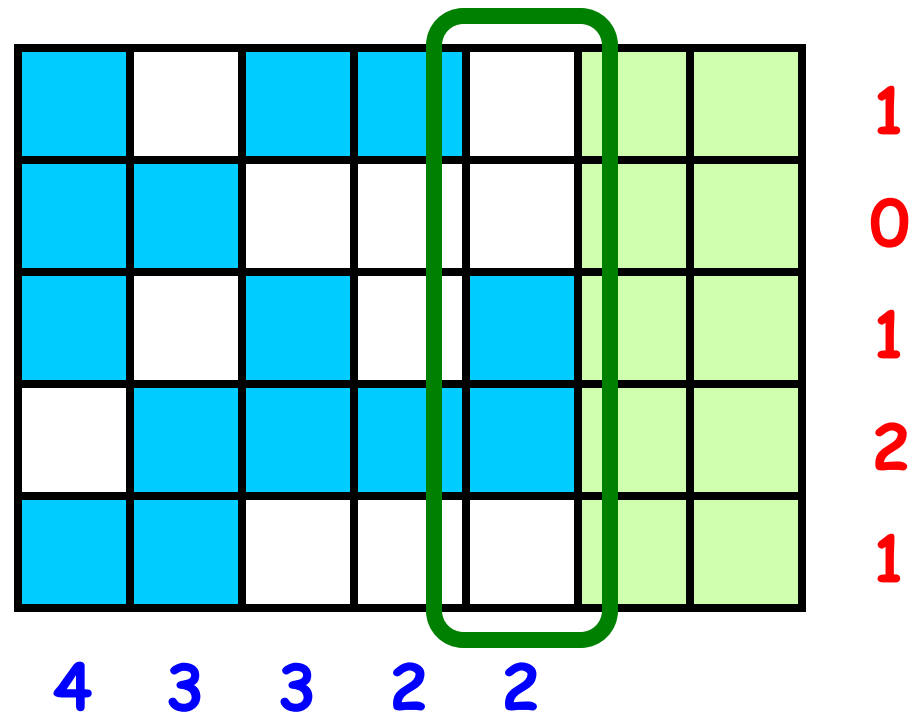
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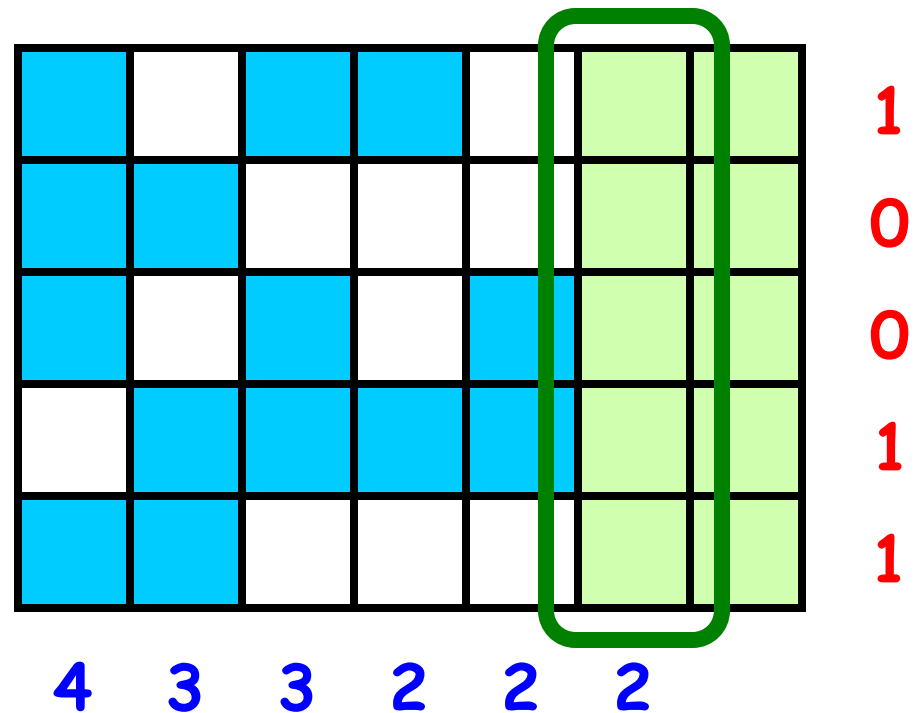
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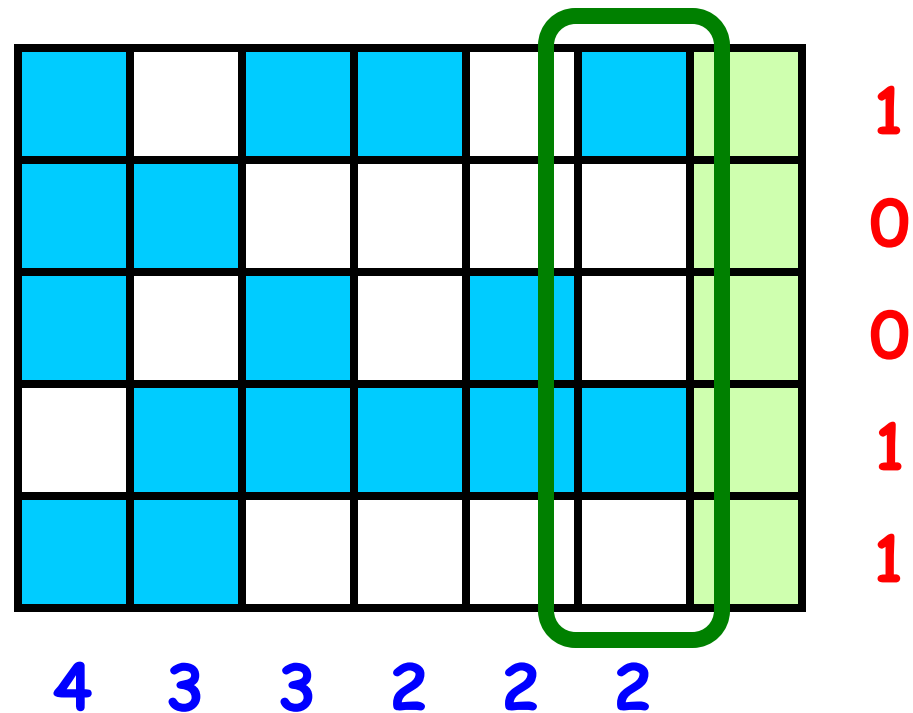
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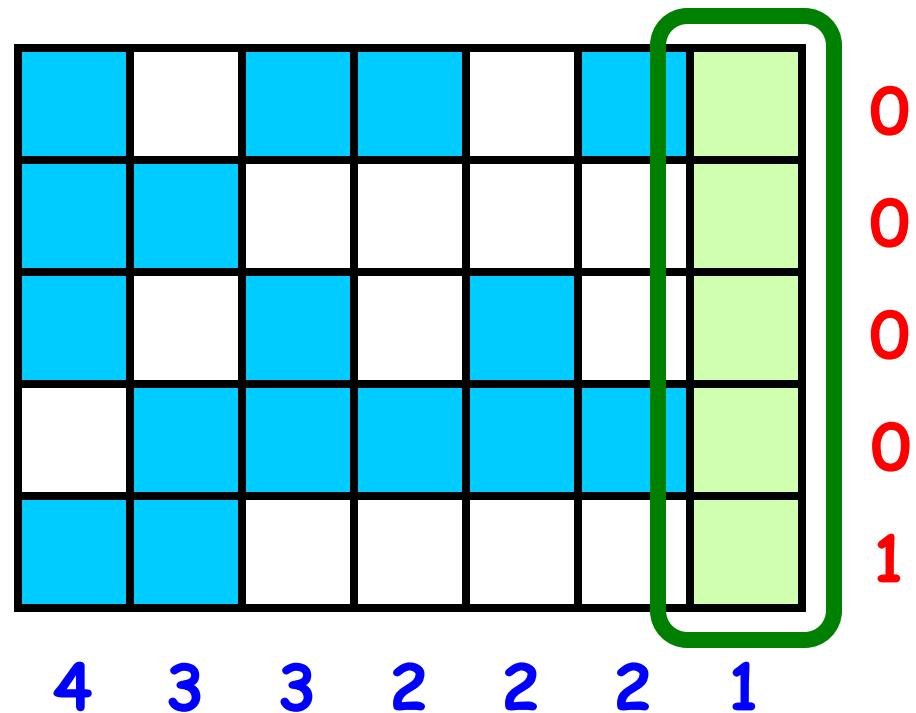
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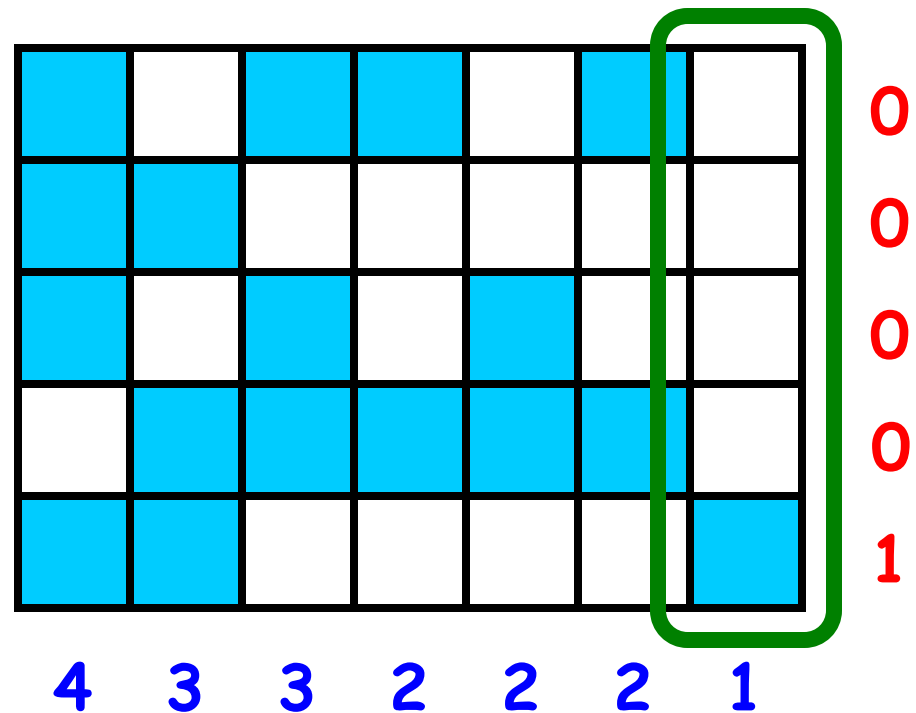
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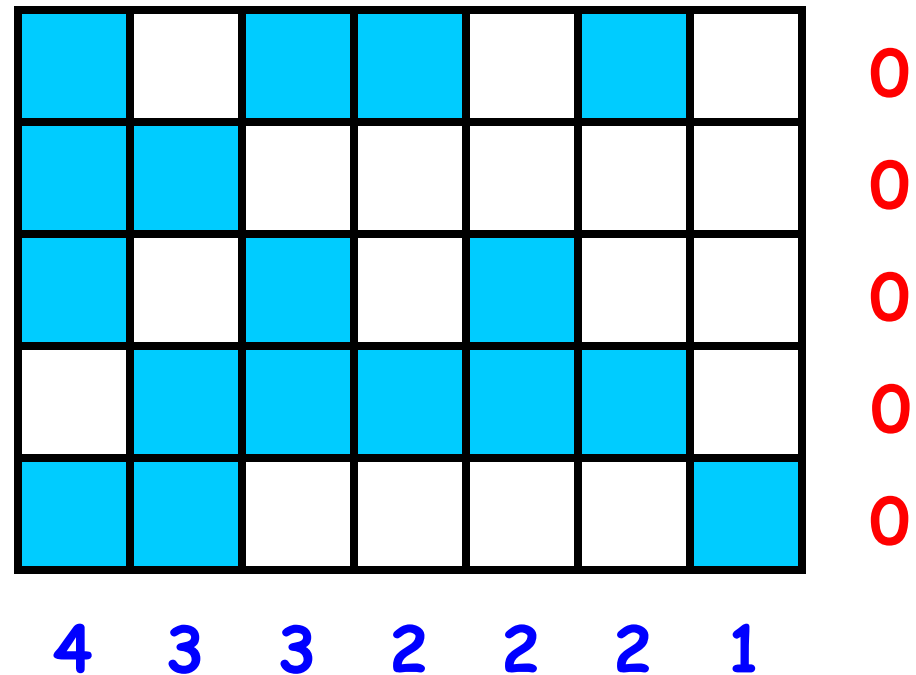
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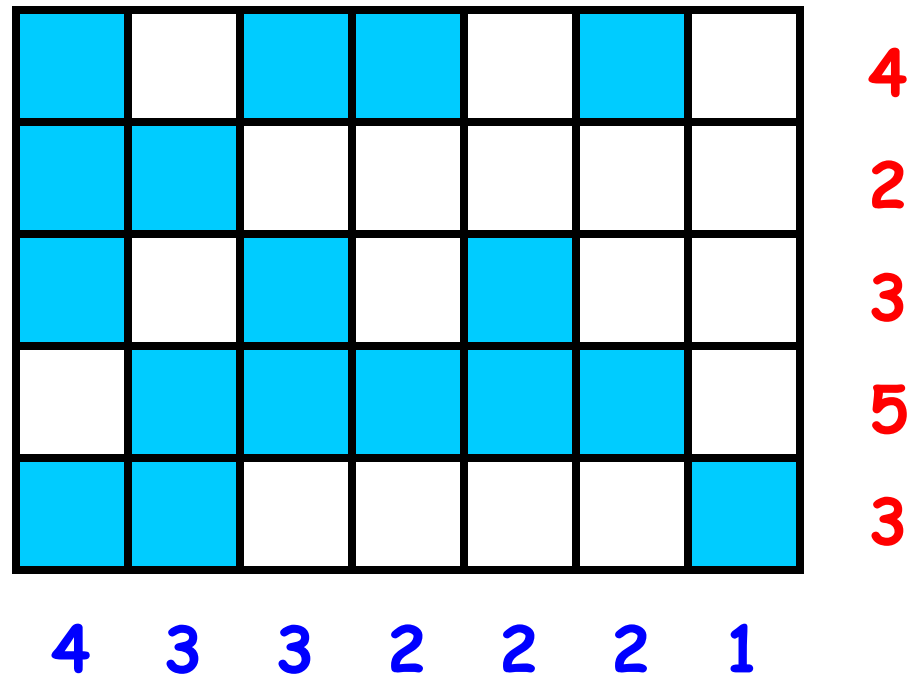
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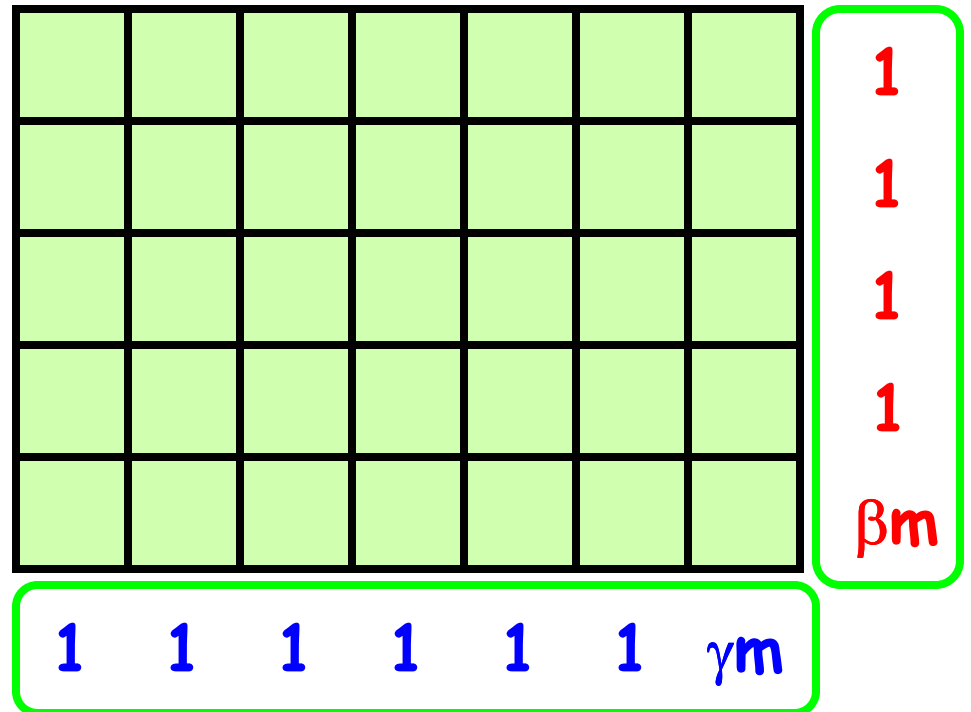
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A Counterexample for SIS

Thm [Bezáková-Sinclair-Štefankovič-Vigoda '06]:

For any $\beta \neq \gamma$, SIS output after any subexponential number of trials is **off by an exponential factor** (with high probability).

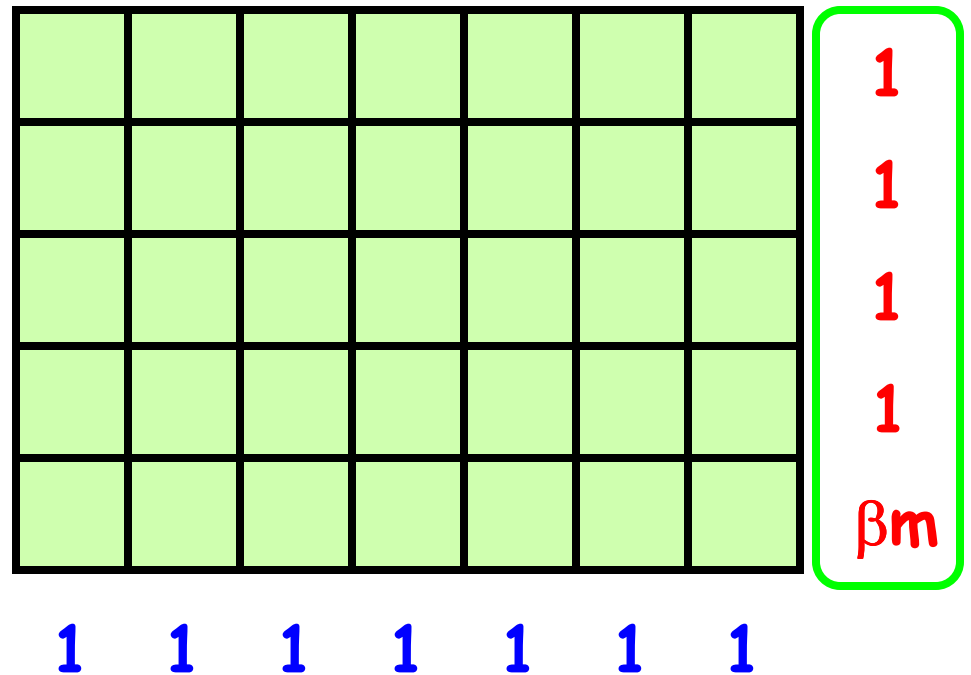


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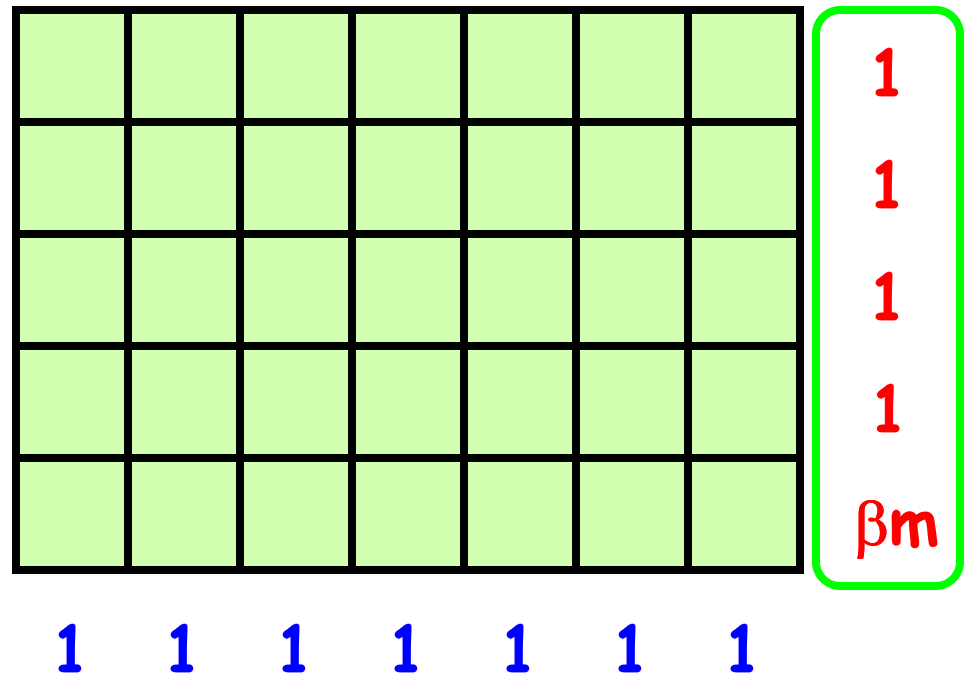
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Random table:

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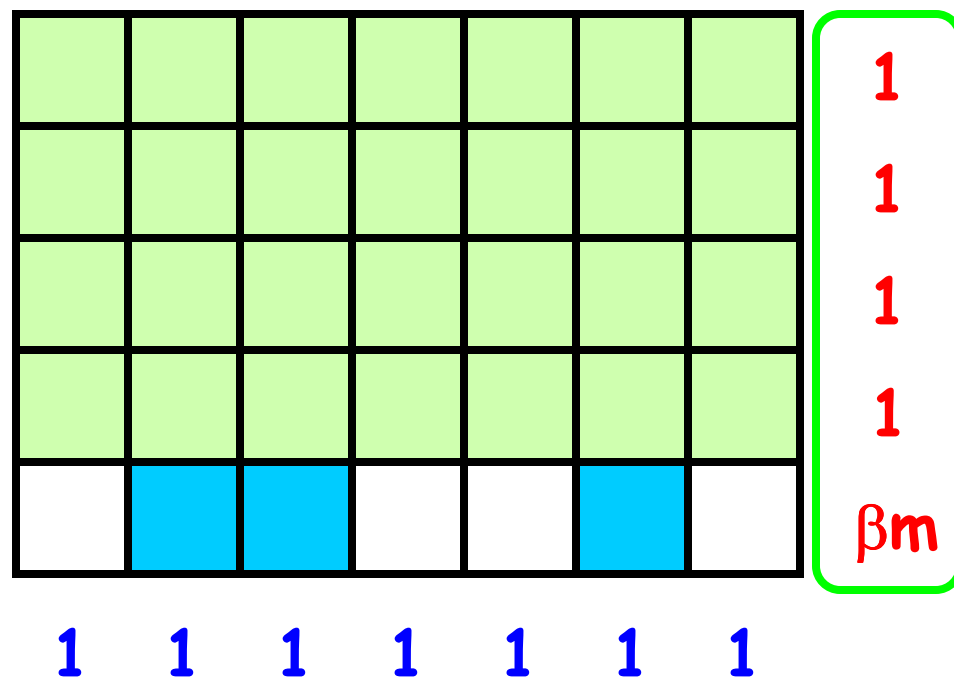
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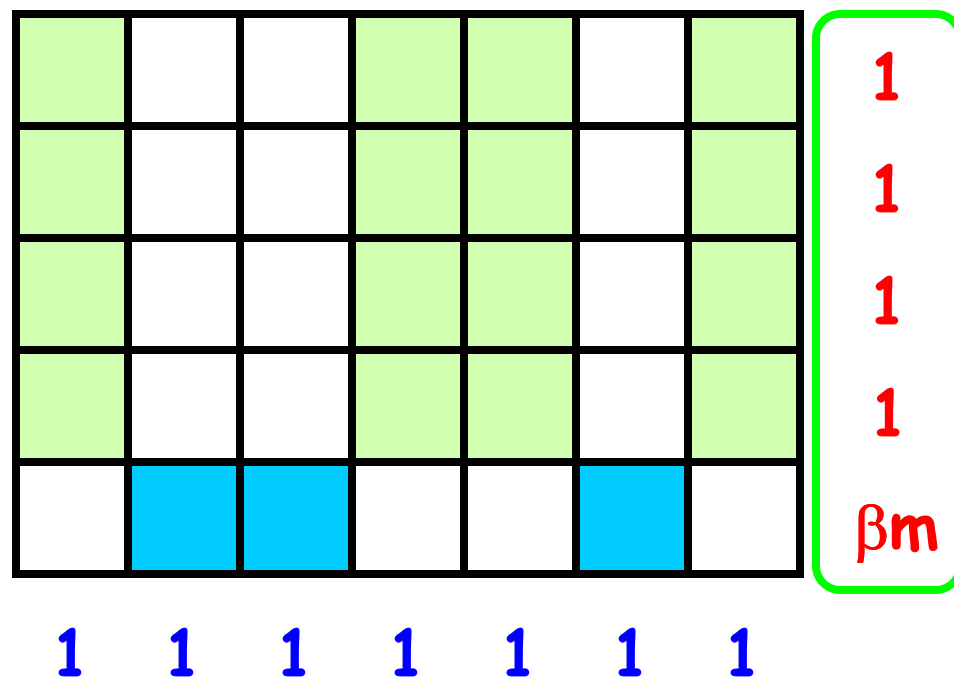
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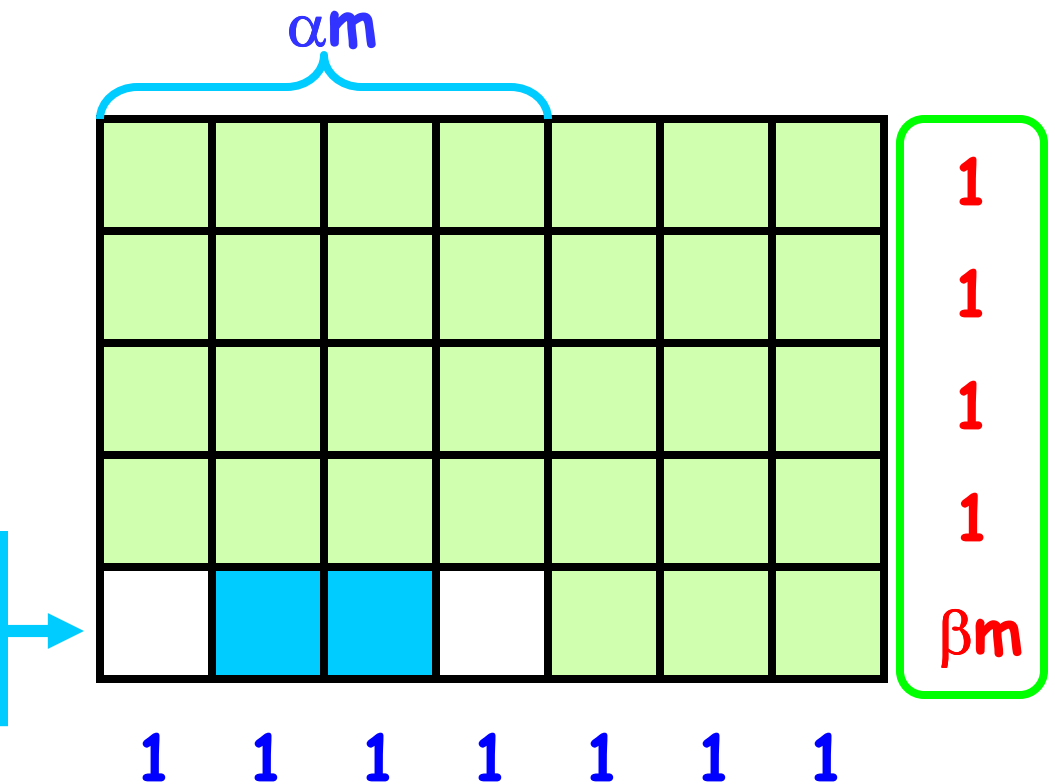
Intuition

Expect: $\alpha\beta m$ ones

SIS: asymptotically fewer

Random table:

- randomly choose βm ones



A Counterexample for SIS

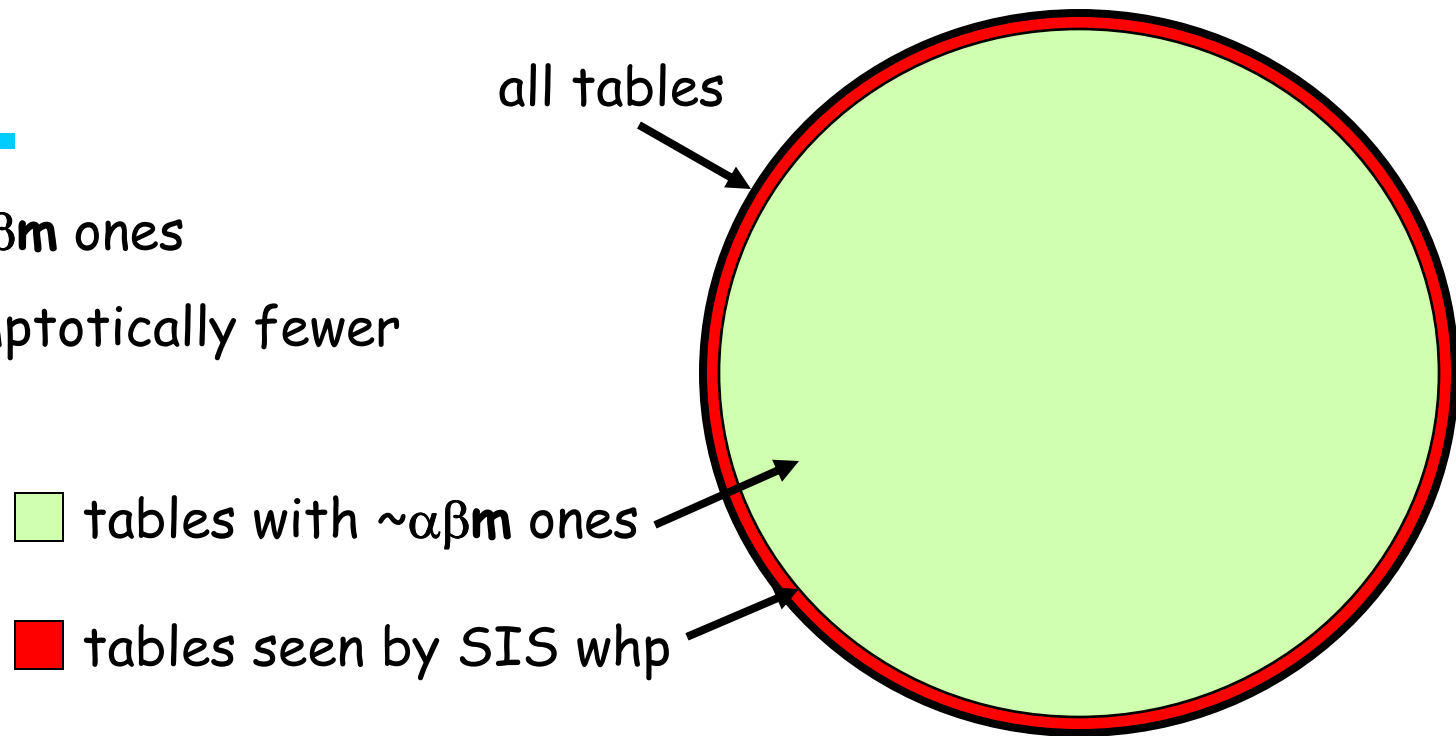
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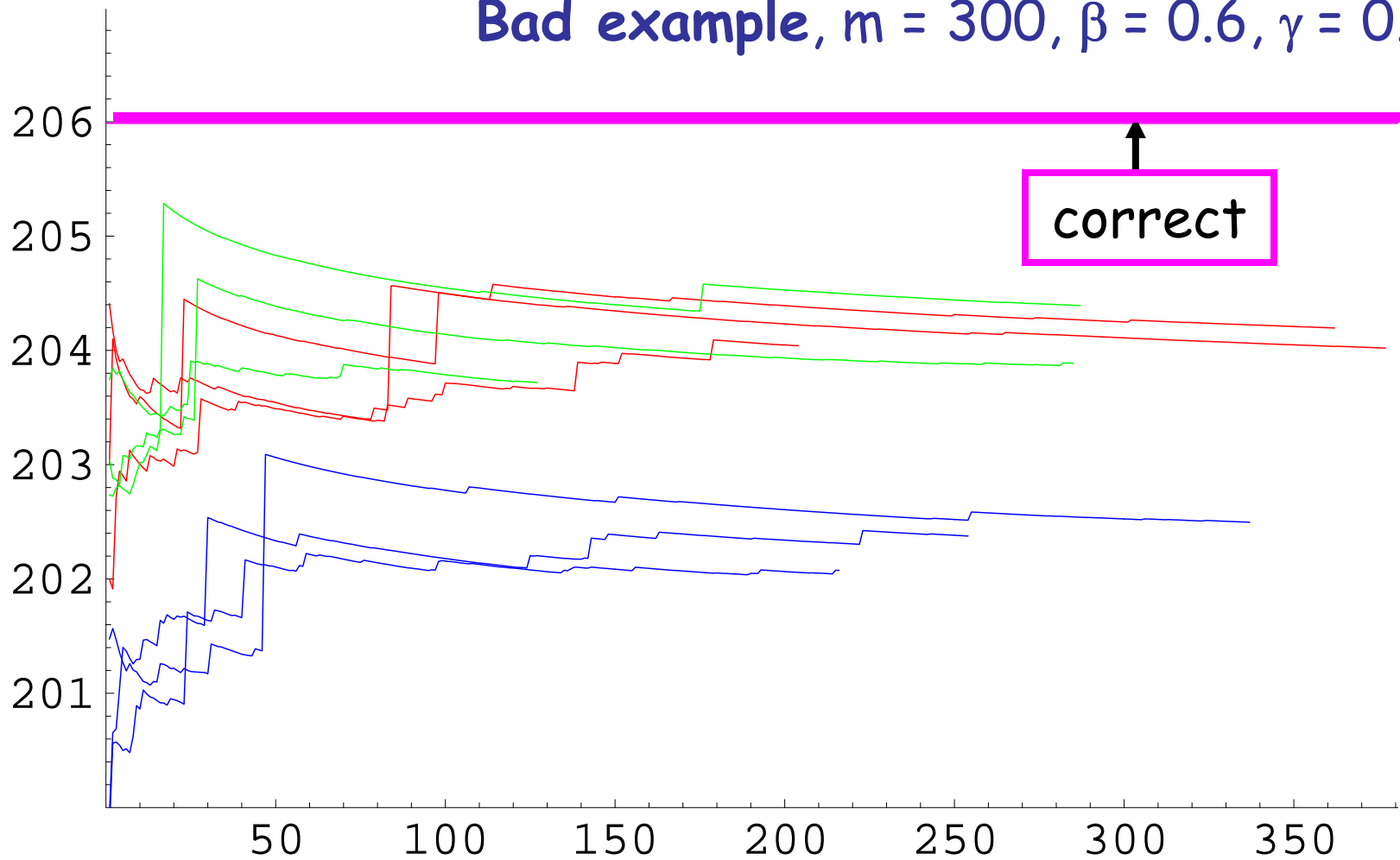
SIS: asymptotically fewer



SIS - Experimental Results

Bad example, $m = 300$, $\beta = 0.6$, $\gamma = 0.7$

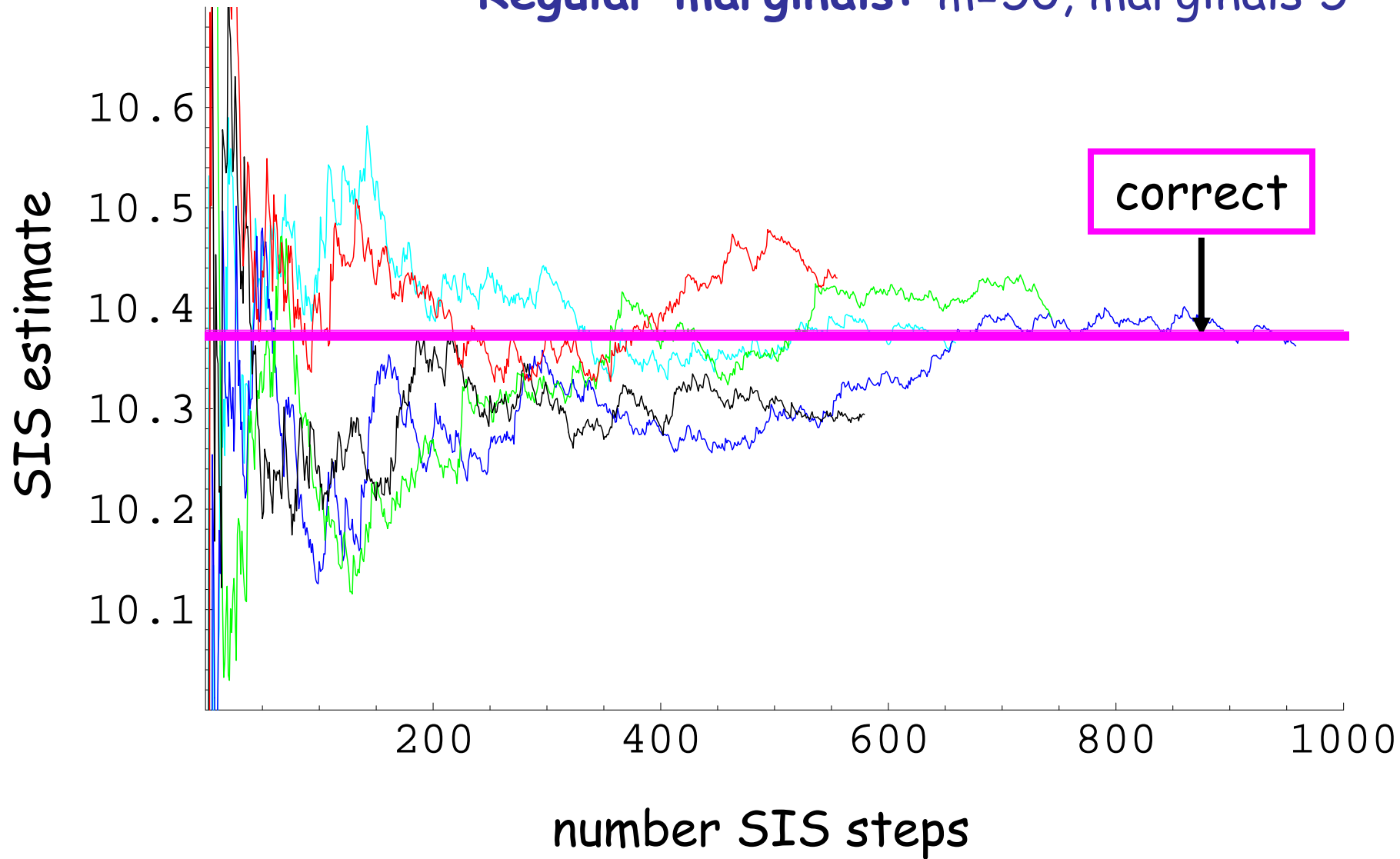
log-scale of SIS estimate



number SIS steps

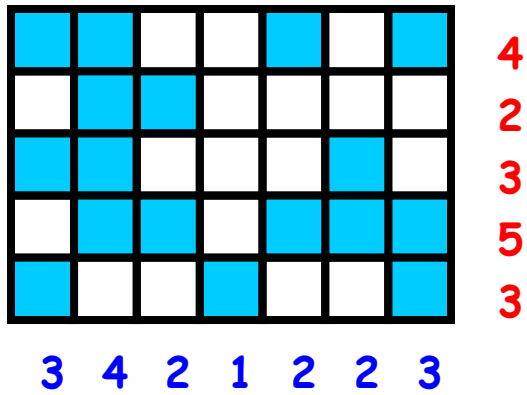
SIS - Experimental Results

Regular marginals: $m=50$, marginals 5

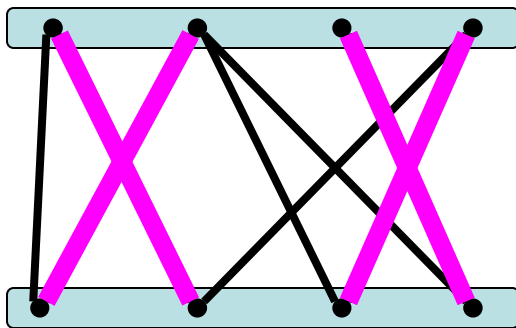


Problems

Binary contingency tables

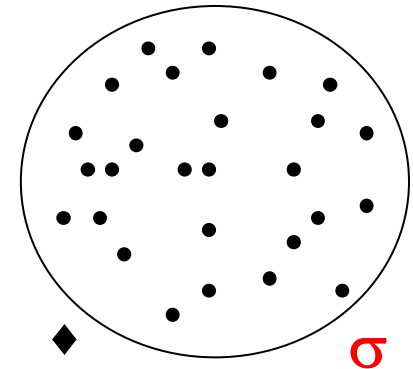


Permanent

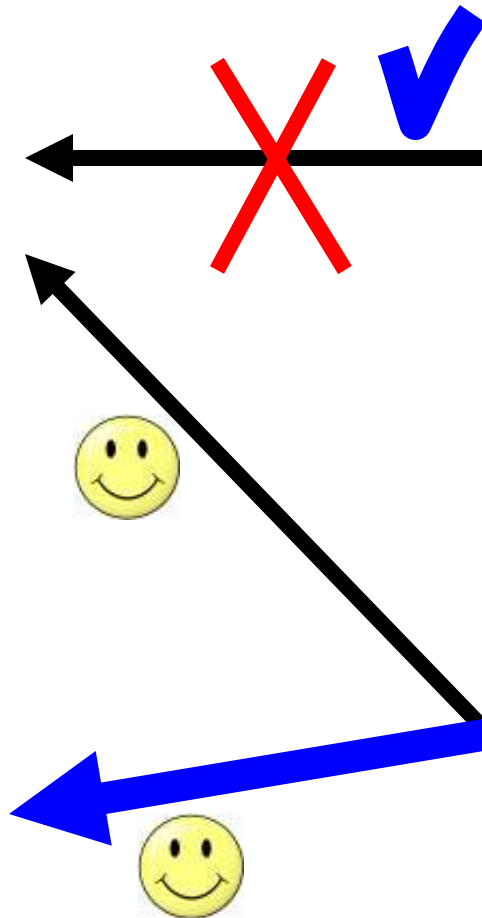


Heuristics

Importance sampling



Simulated annealing

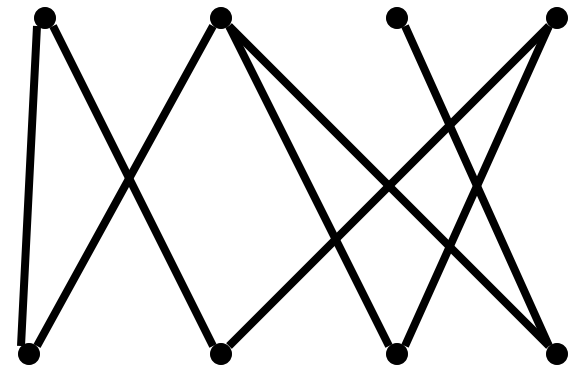


Broder Chain

[Broder '88]

What for: uniform sampling of perfect matchings

How: Markov chain on perfect + **near-perfect** matchings

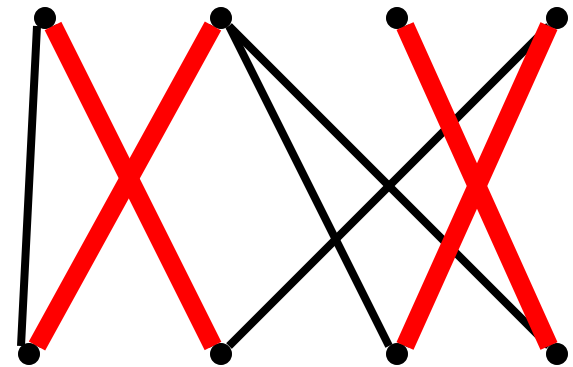


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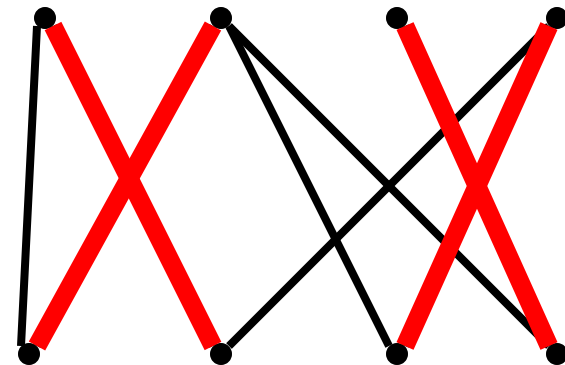
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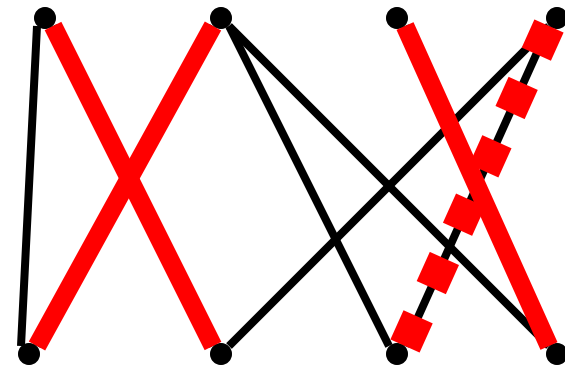
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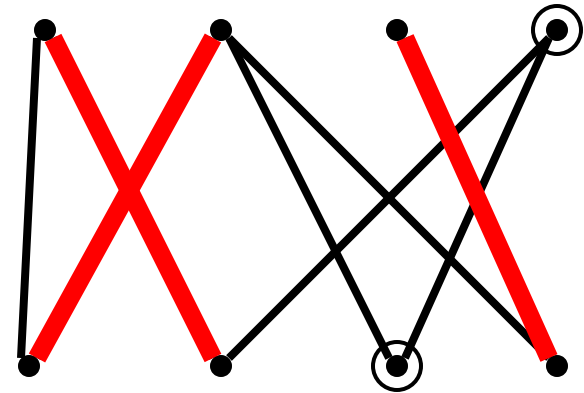
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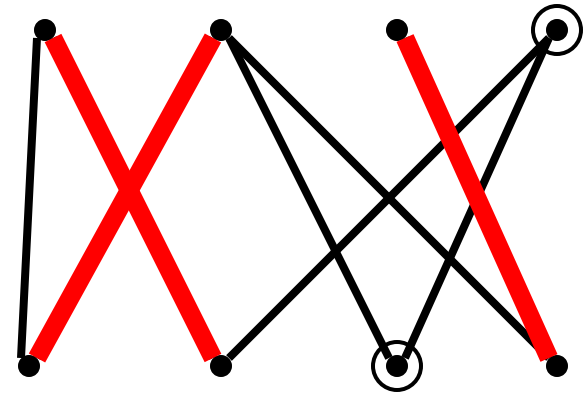
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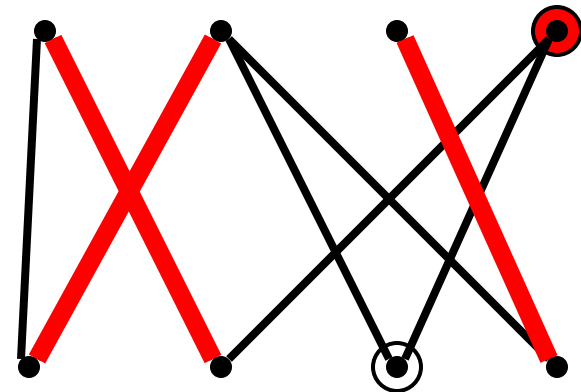
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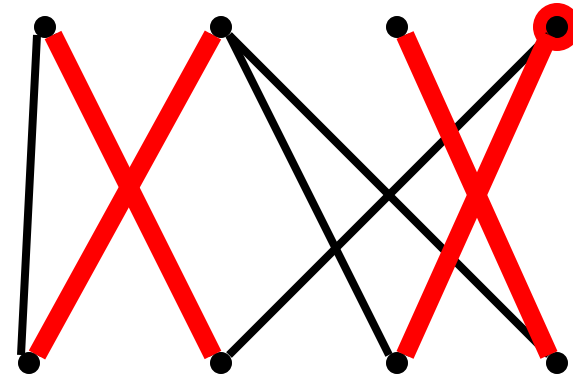
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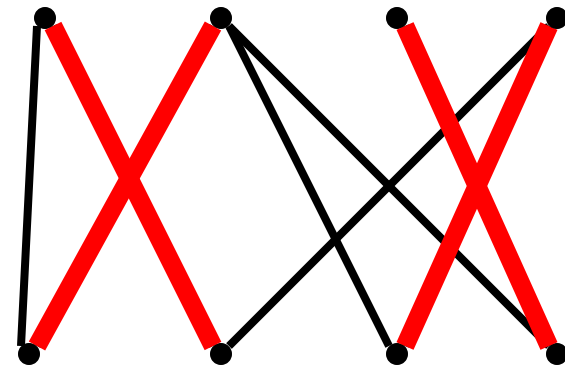
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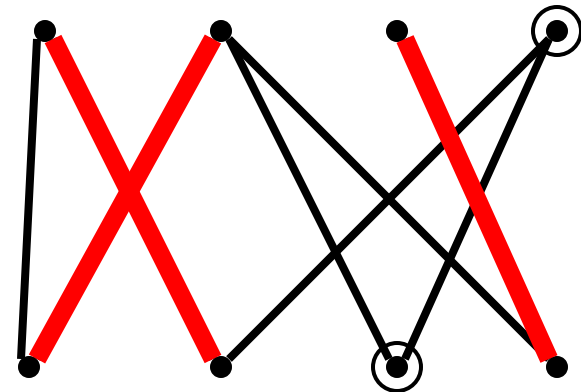
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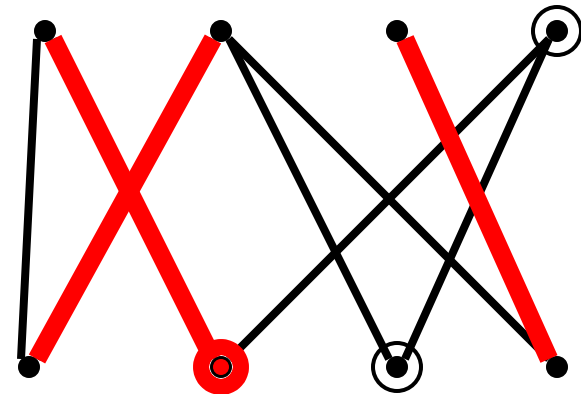
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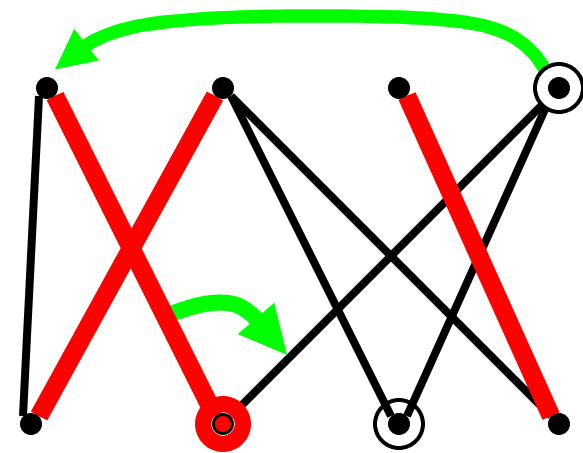
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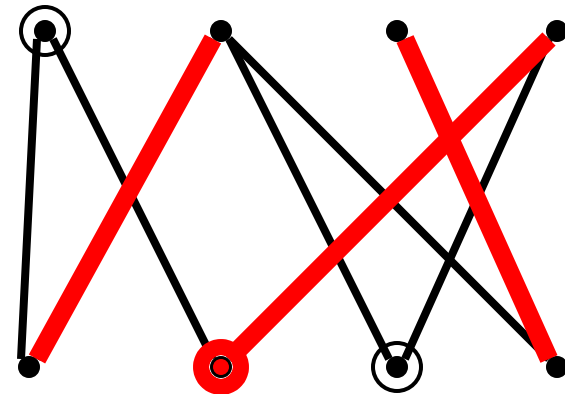
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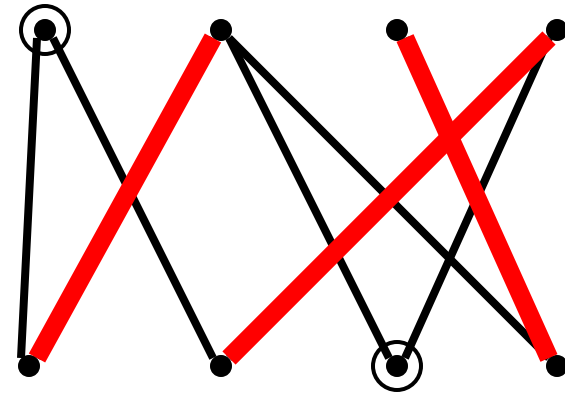
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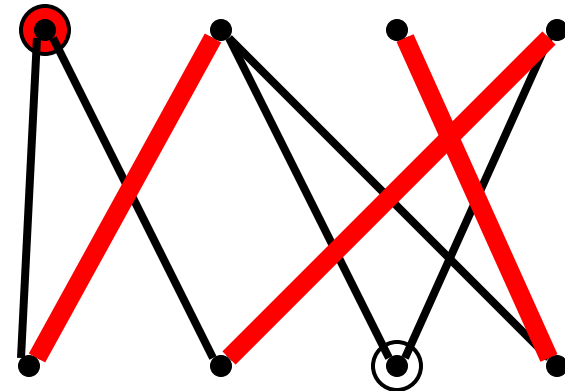
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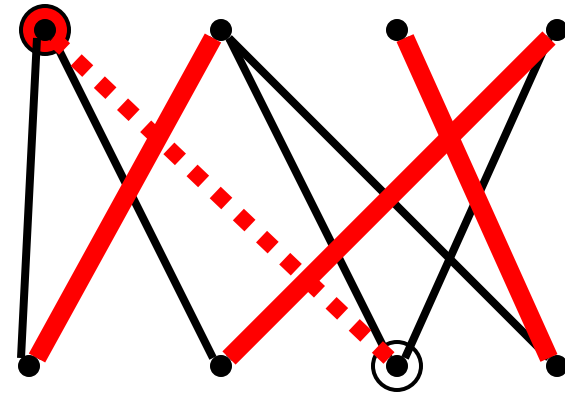
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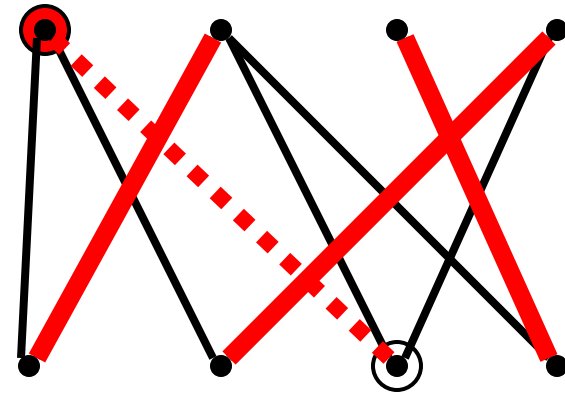
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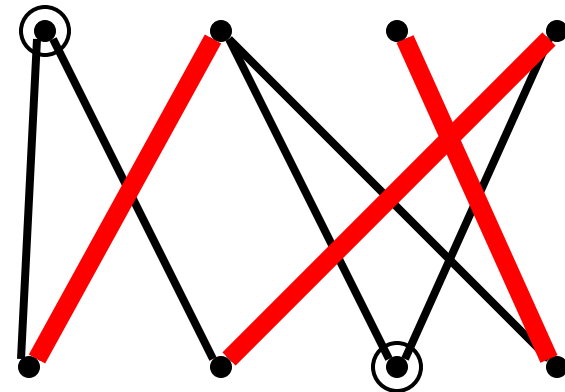
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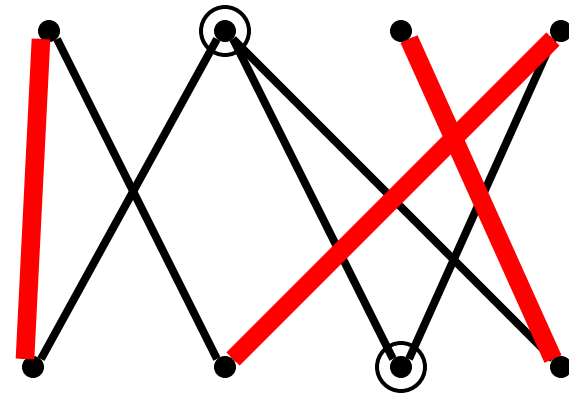
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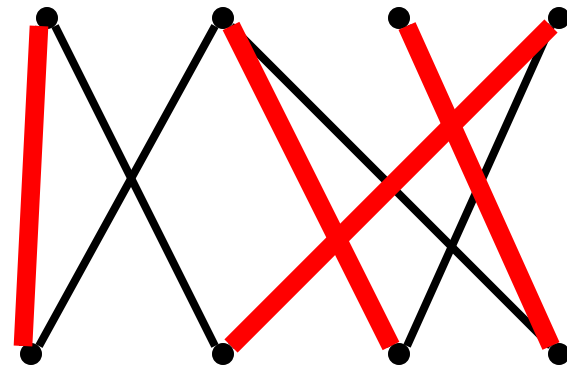
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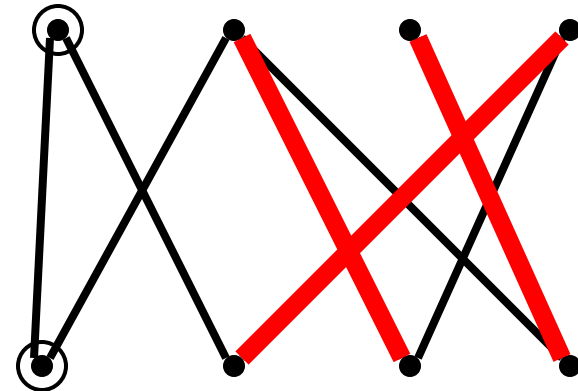
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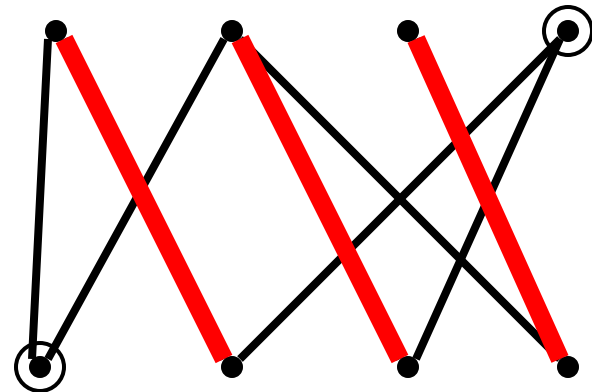
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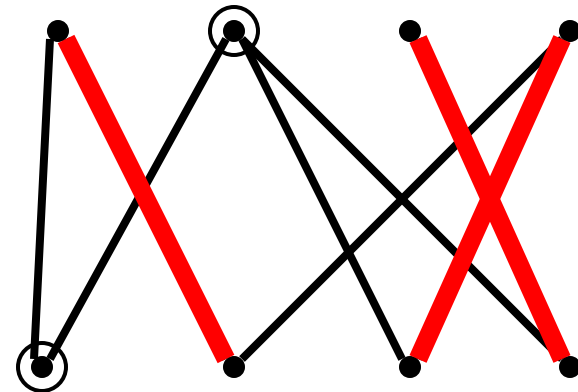
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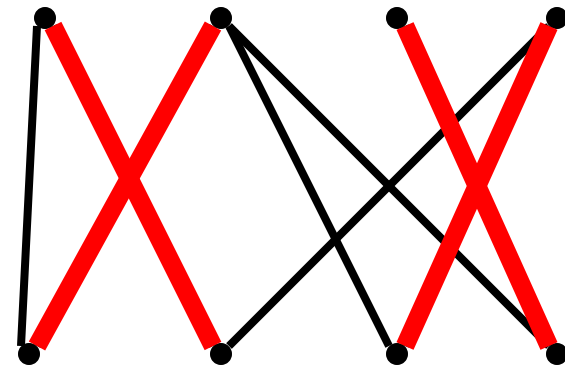
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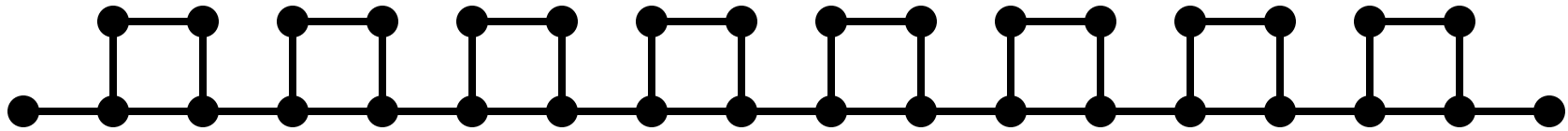
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Broder Chain

Mixes in polynomial time ?

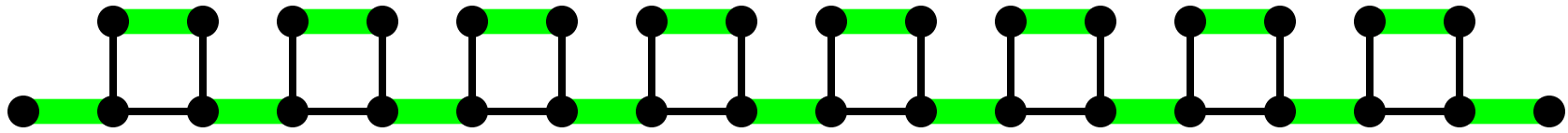
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Broder Chain

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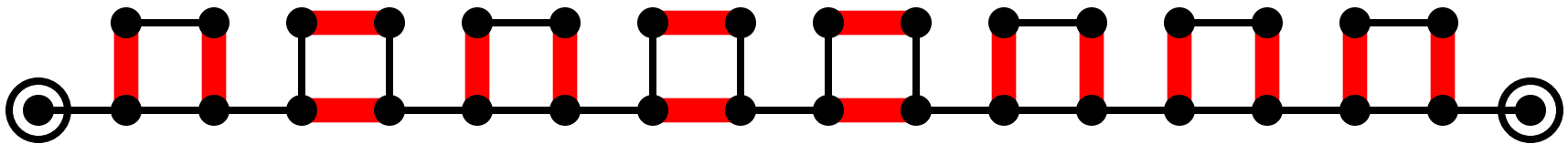


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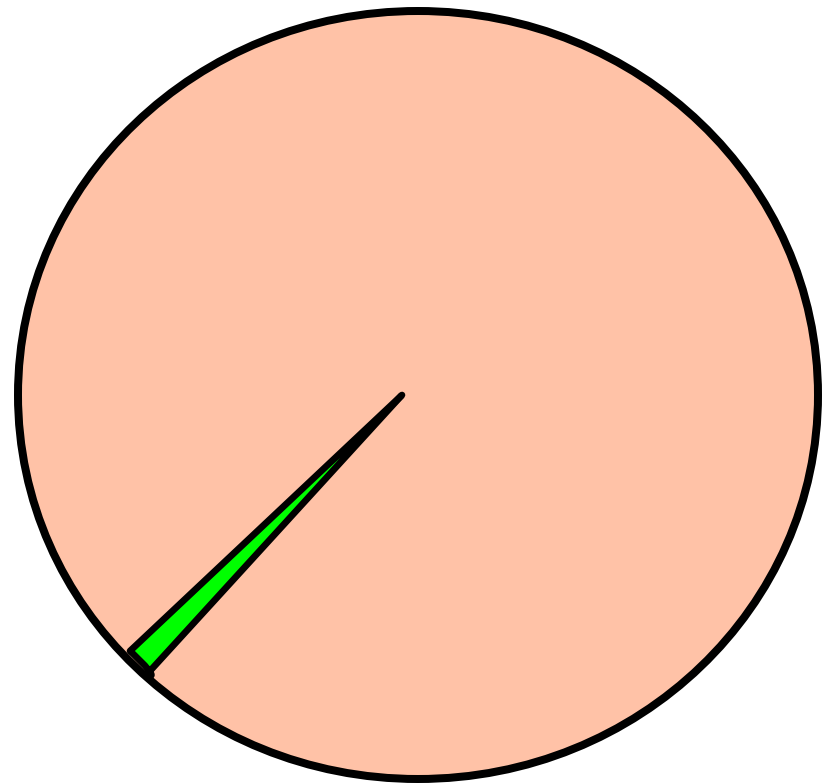
Broder Chain

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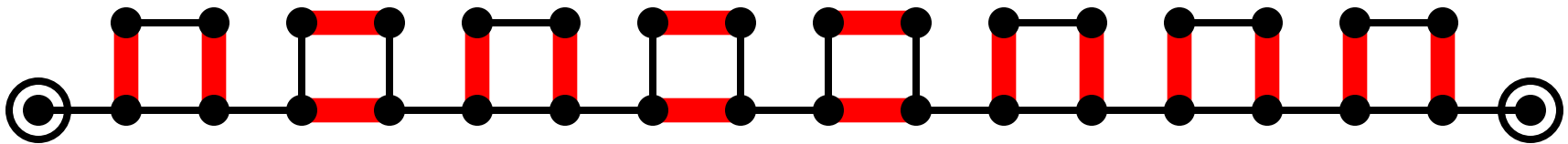
1 perfect matching
 $\geq 2^{(n/4)}$ near matchings



Broder Chain

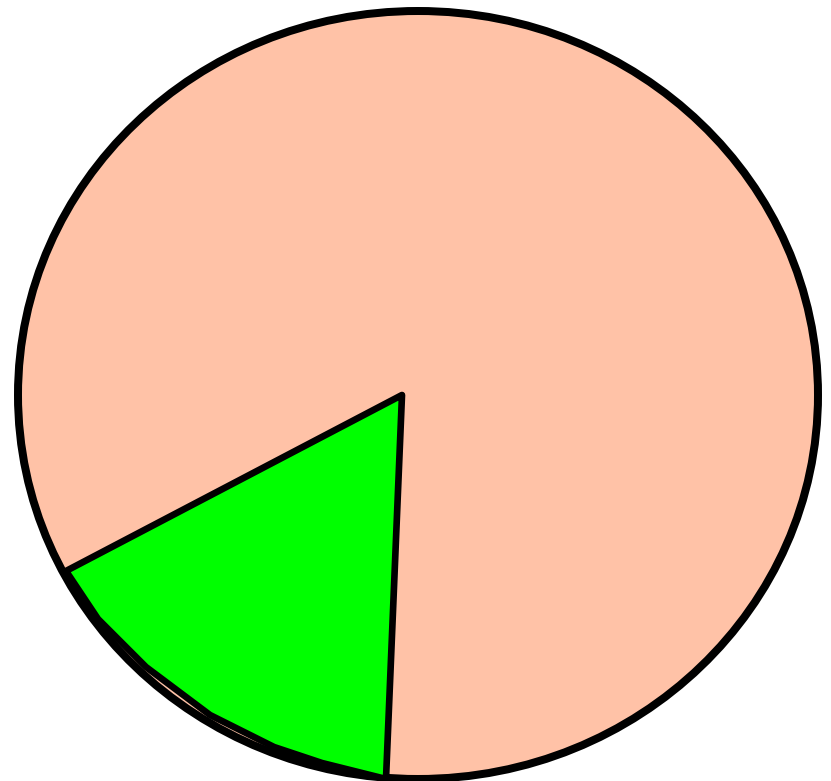
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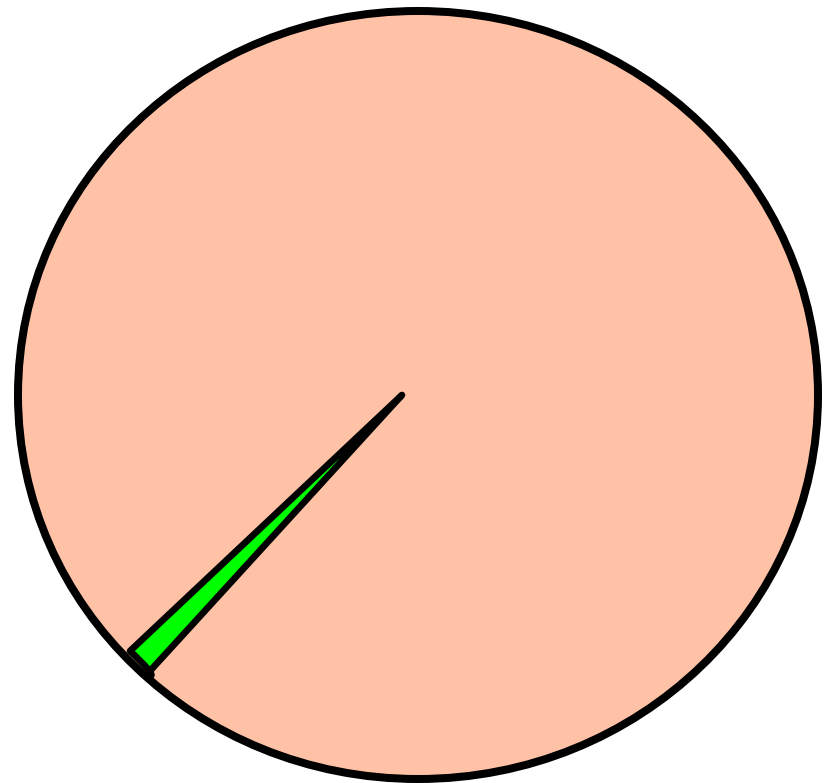
Thm [Jerrum-Sinclair '89]:
Rapid mixing if perfects
polynomially related to nears.



Broder Chain

Idea [Jerrum-Sinclair-Vigoda '01]:

Change the **weight** so that **perfect matchings** take **polynomial** fraction.

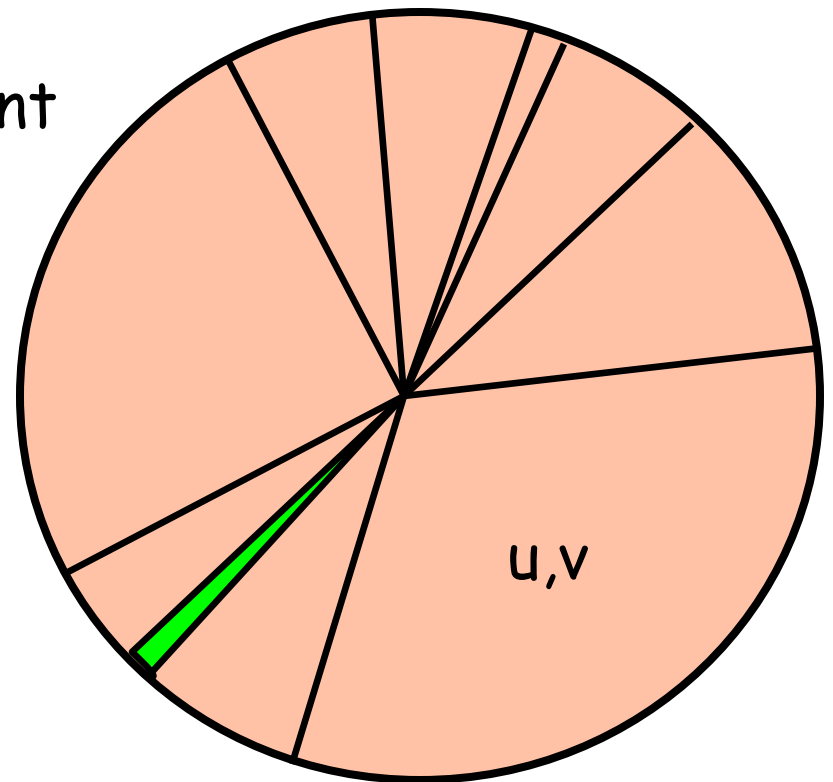


Broder Chain

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n^2+1 regions,
very different
size

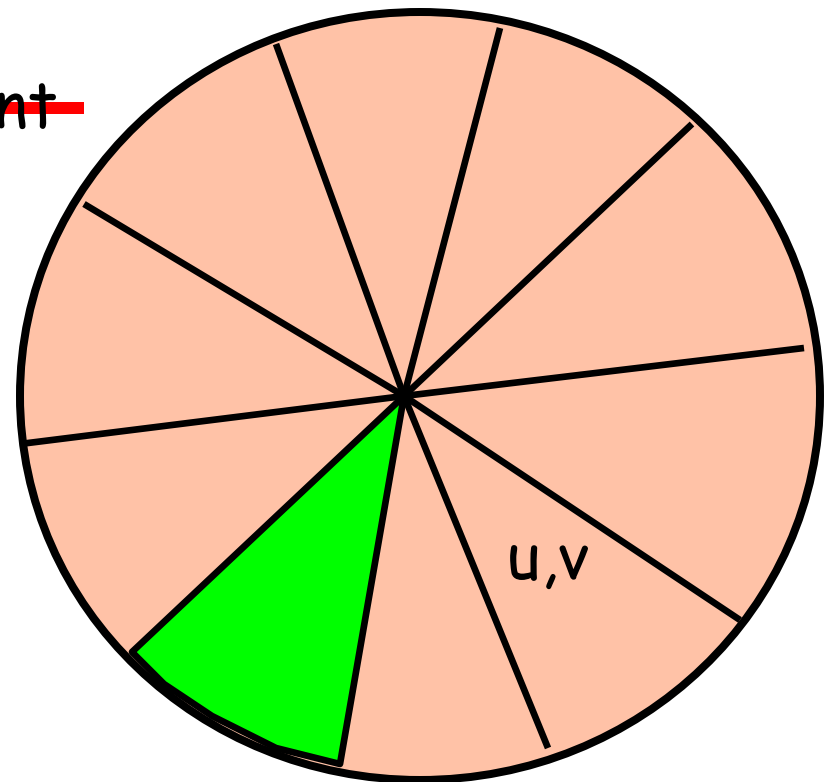


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Change the **weight** so that **perfect matchings** take **polynomial** fraction.

n^2+1 regions,
~~very different~~
the same size



Broder Chain

Idea [Jerrum-Sinclair-Vigoda '01]:

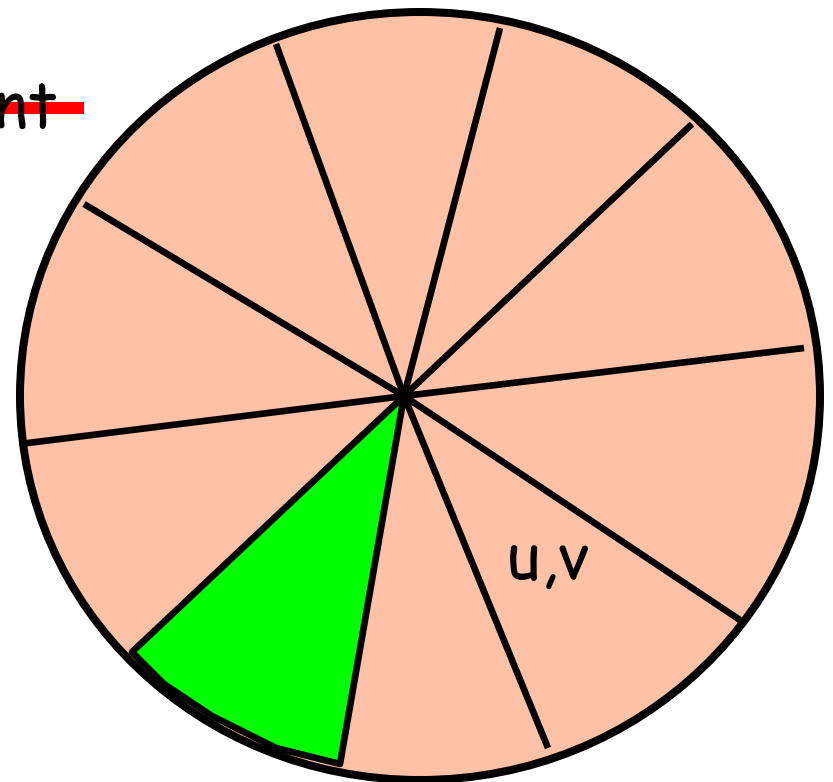
Change the **weight** so that **perfect matchings** take **polynomial** fraction.

n^2+1 regions,
~~very different~~
the same size

Ideal weights

(for a matching with holes u,v):

$$\frac{(\# \text{ perfects})}{(\# \text{ nears with holes } u,v)}$$



Broder Chain

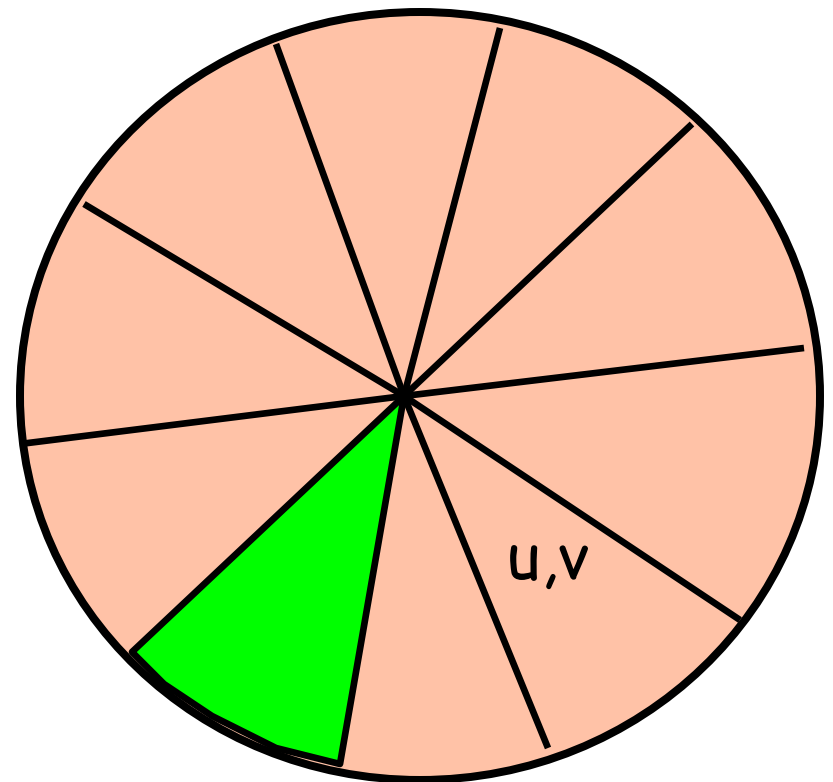
Good: A perfect matching sampled with prob. $1/(n^2+1)$

Bad: Computing ideal weights as hard as original problem ?

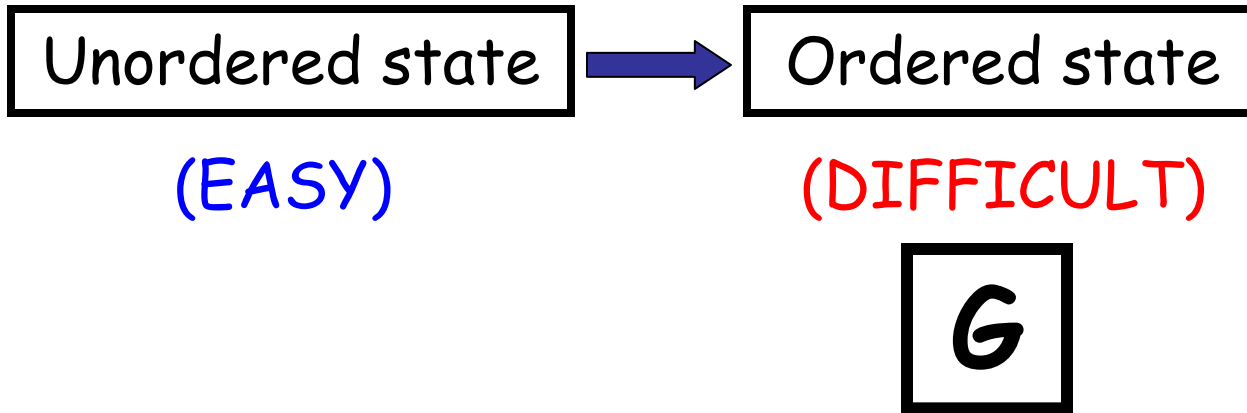
Solution: **Approximate**

Ideal weights
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Simulated Annealing for Permanent

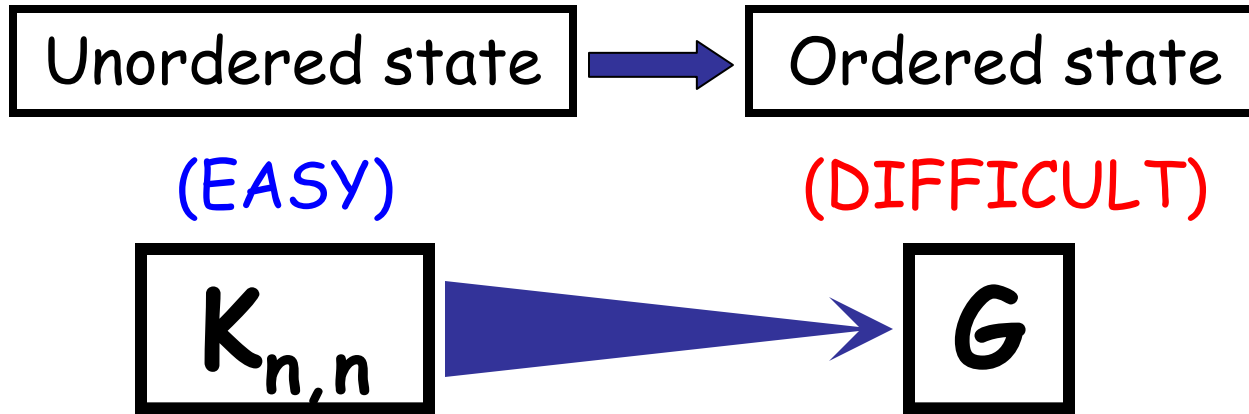


Solution: **Approximate**

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Simulated Annealing for Permanent

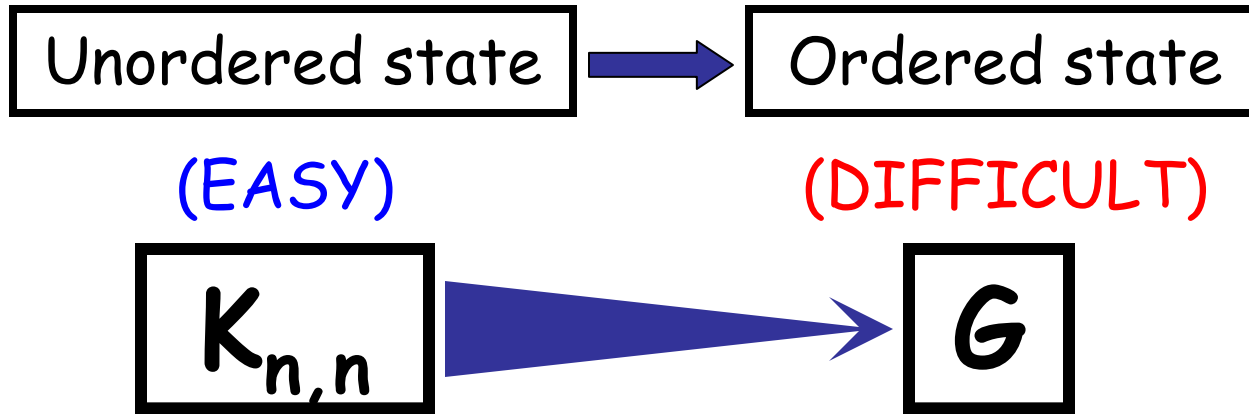


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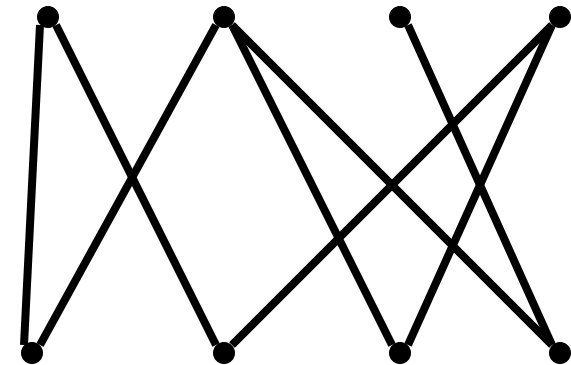
Simulated Annealing for Permanent



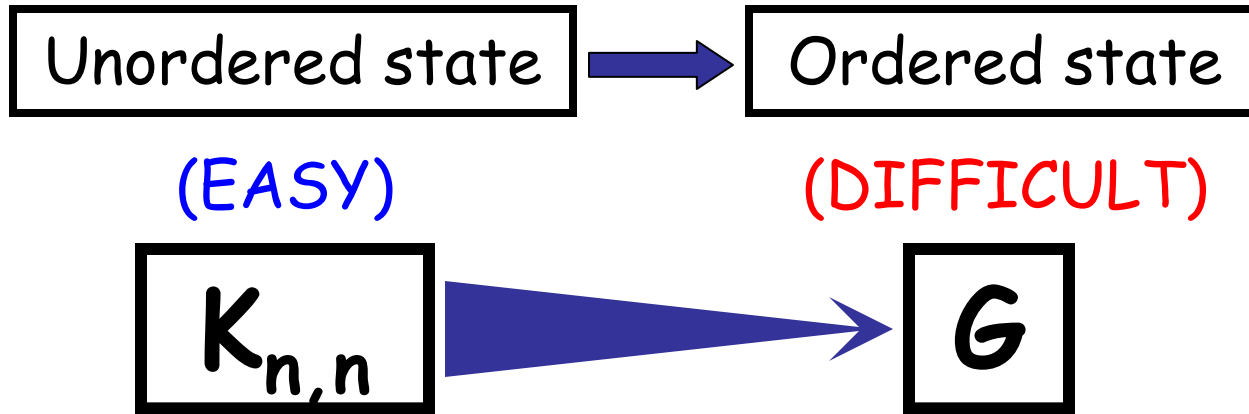
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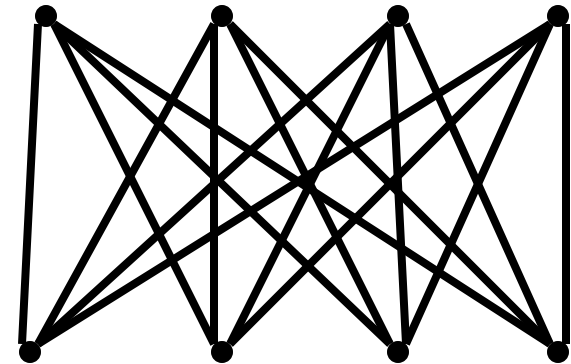
Simulated Annealing for Permanent



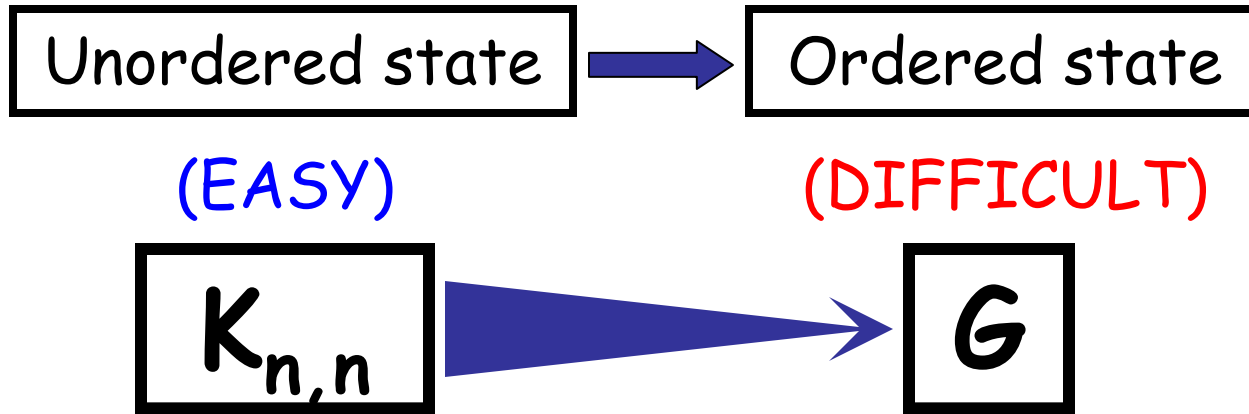
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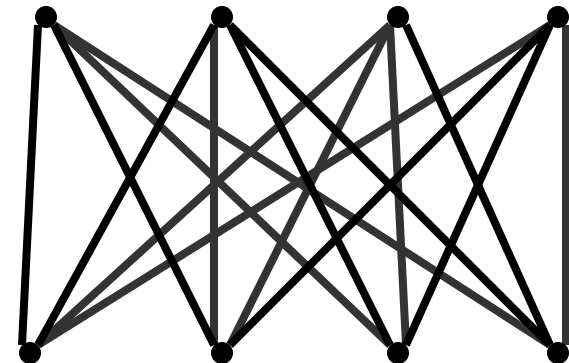
Simulated Annealing for Permanent



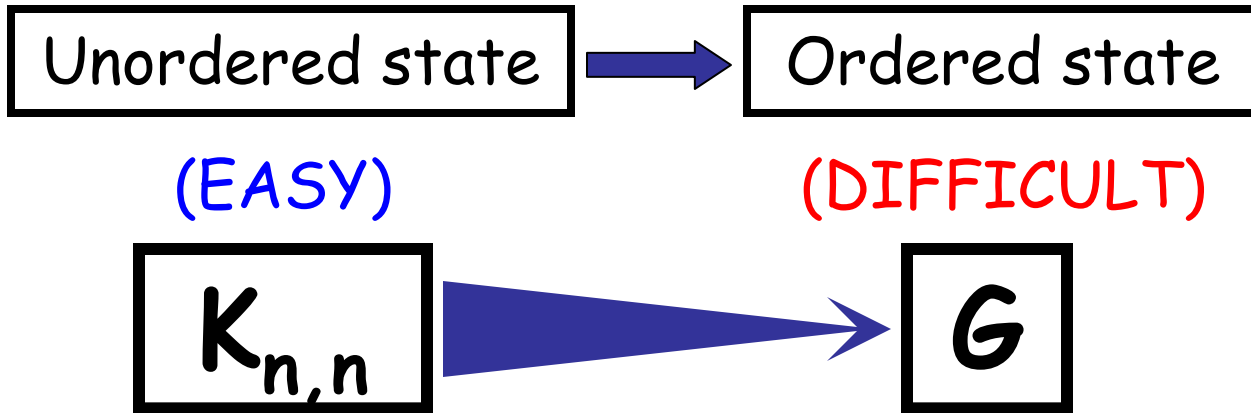
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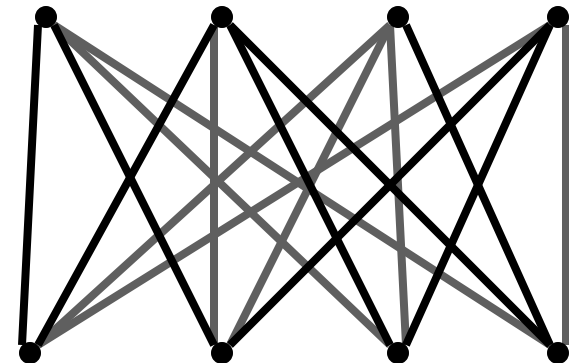
Simulated Annealing for Permanent



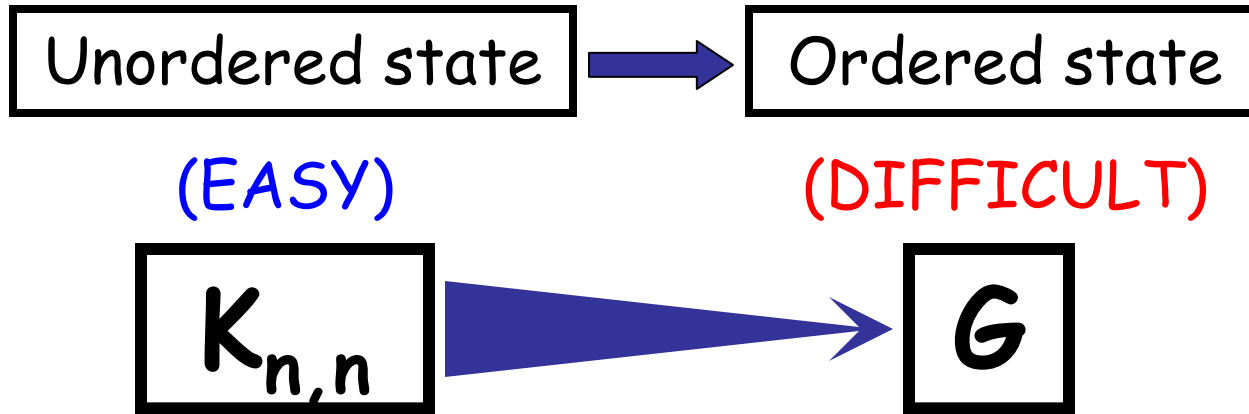
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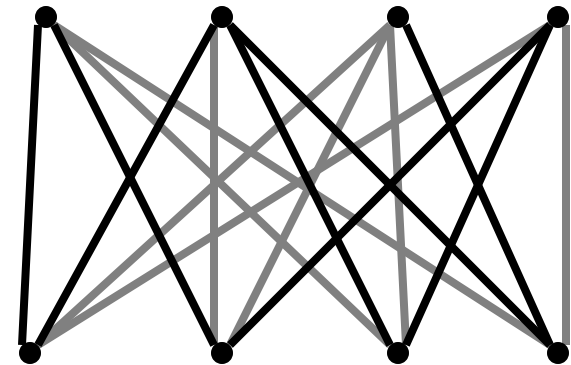
Simulated Annealing for Permanent



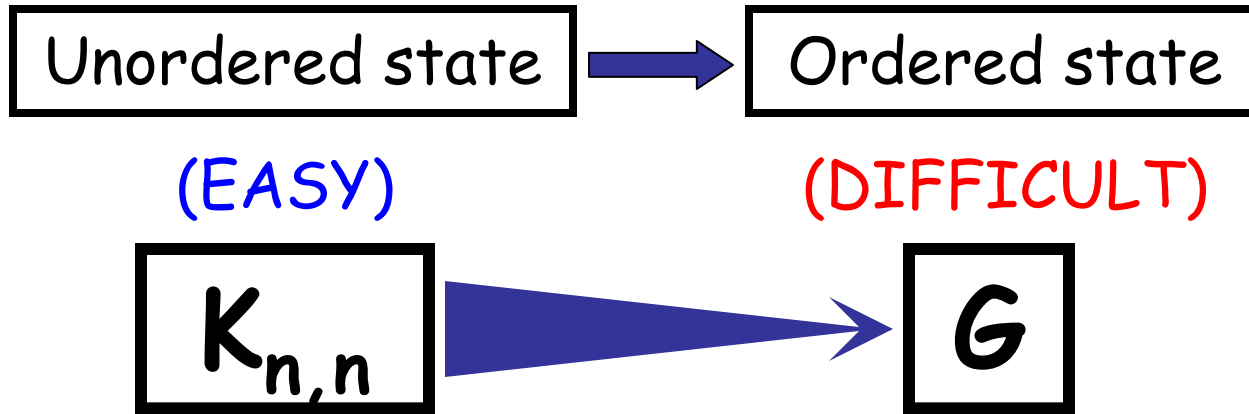
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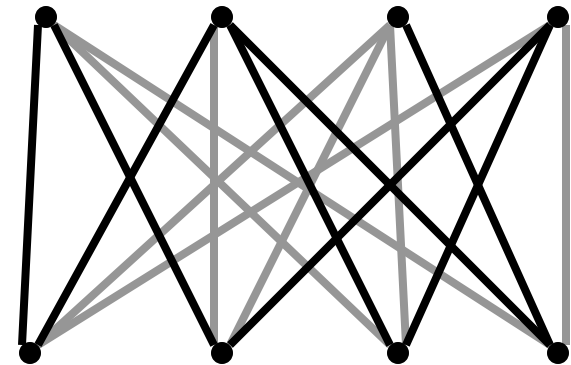
Simulated Annealing for Permanent



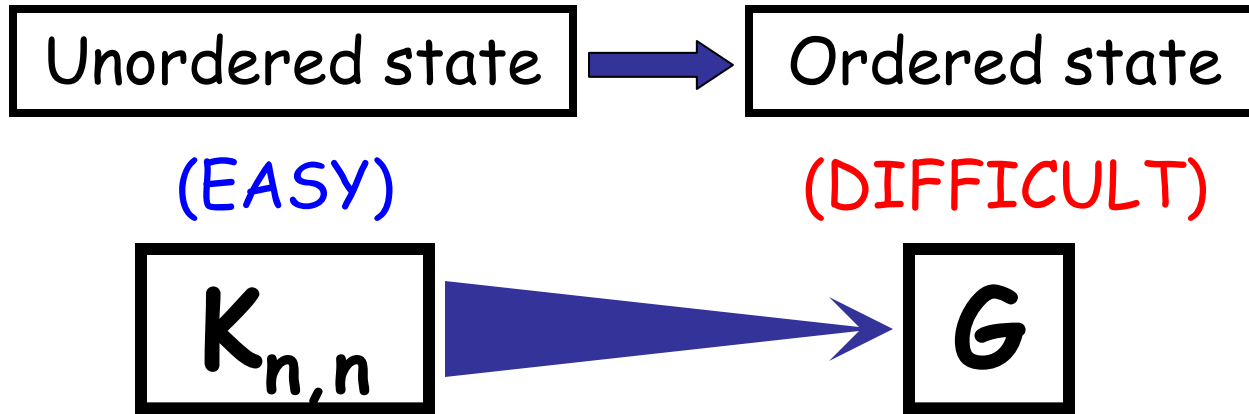
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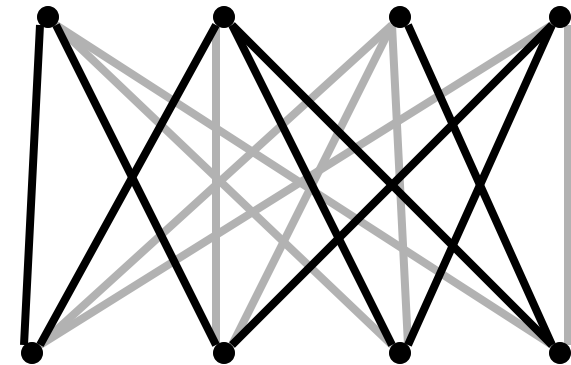
Simulated Annealing for Permanent



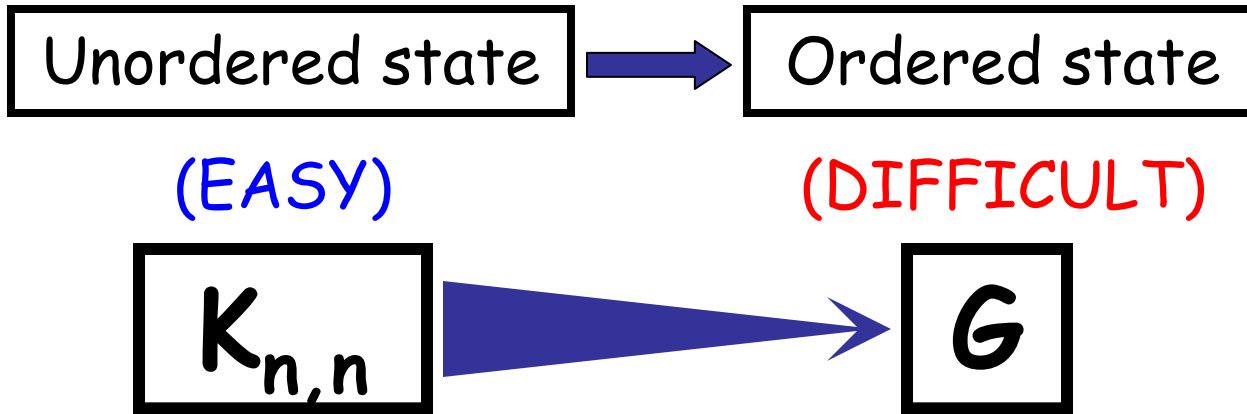
Solution: **Approximate**

Ideal weights
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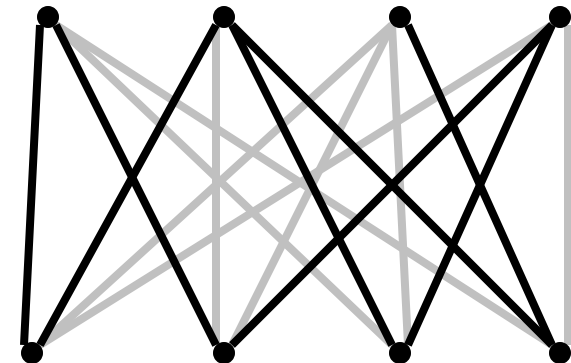
Simulated Annealing for Permanent



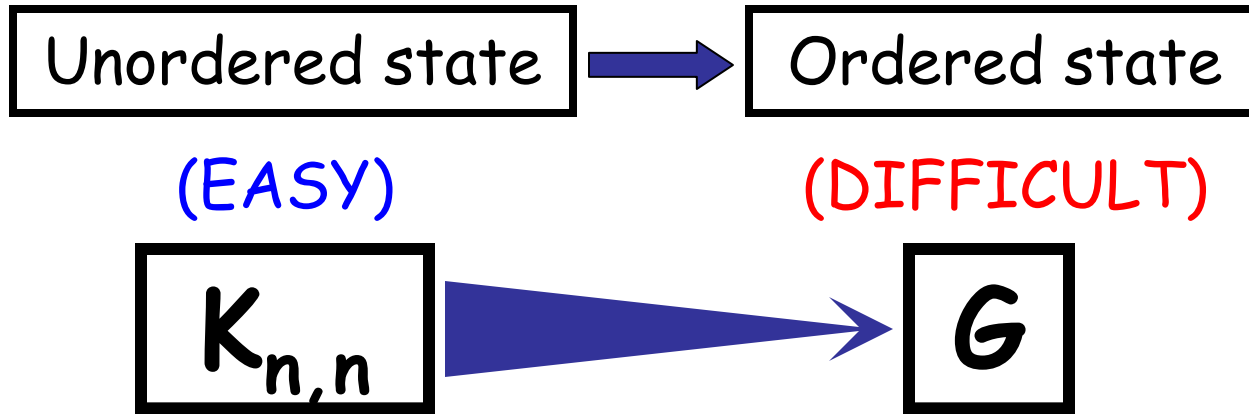
Solution: **Approximate**

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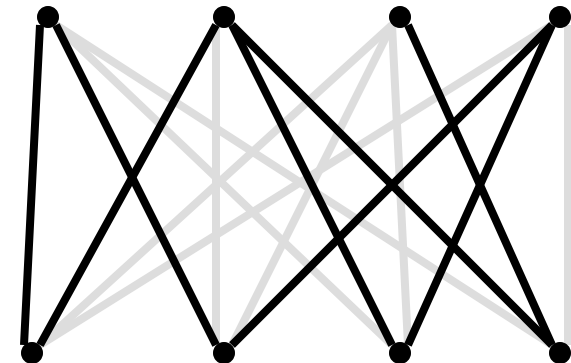
Simulated Annealing for Permanent



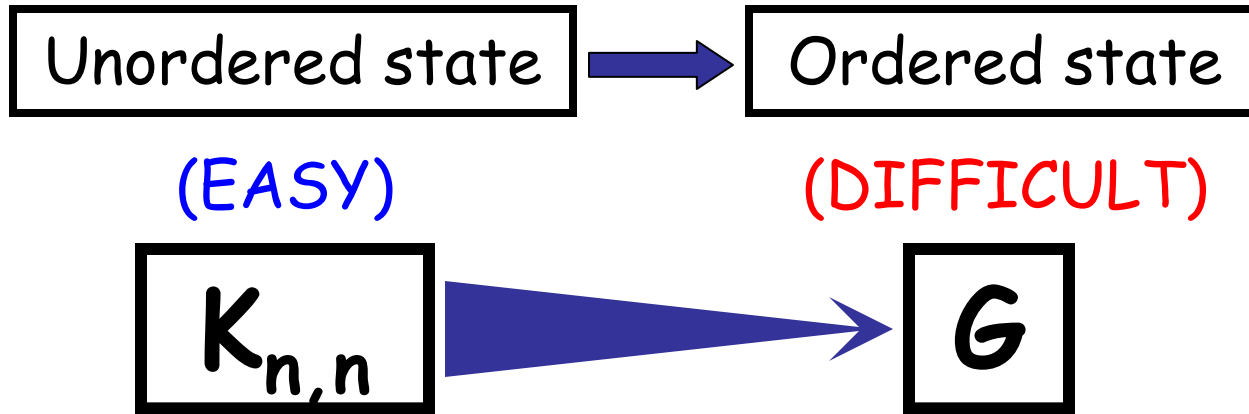
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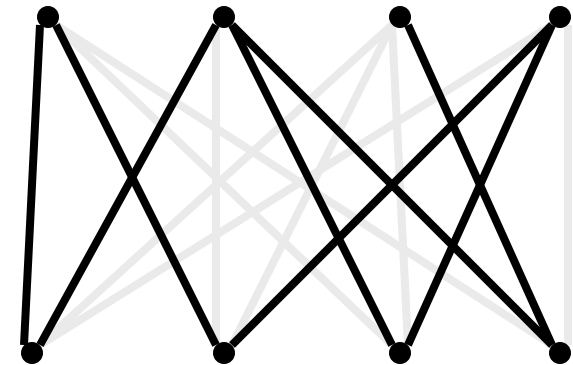
Simulated Annealing for Permanent



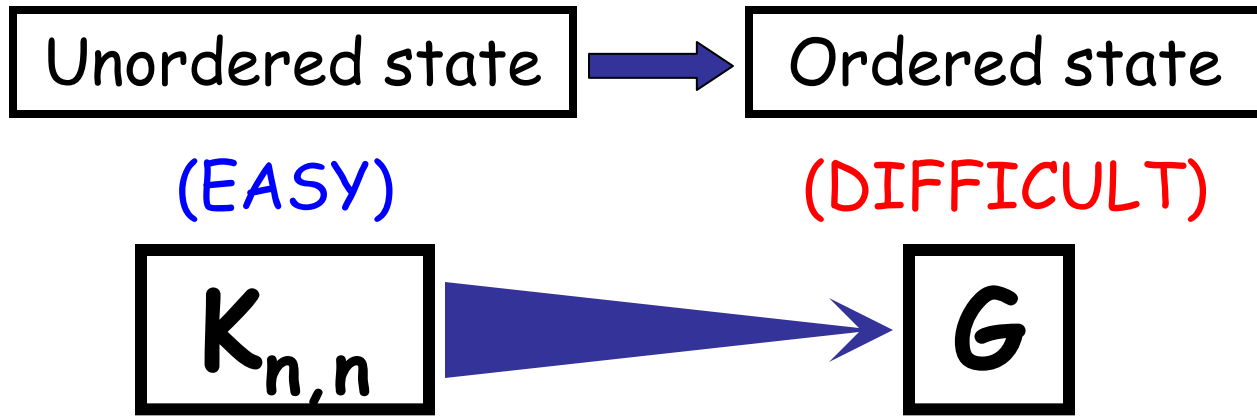
Solution: **Approximate**

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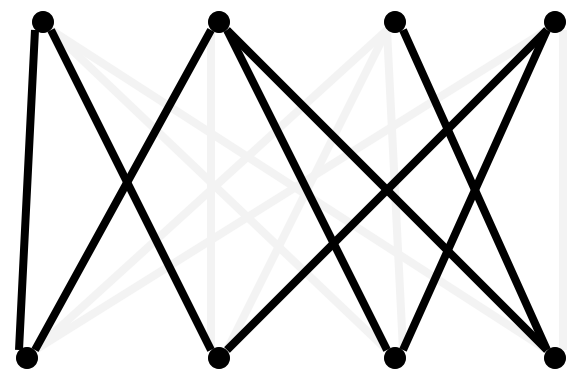
Simulated Annealing for Permanent



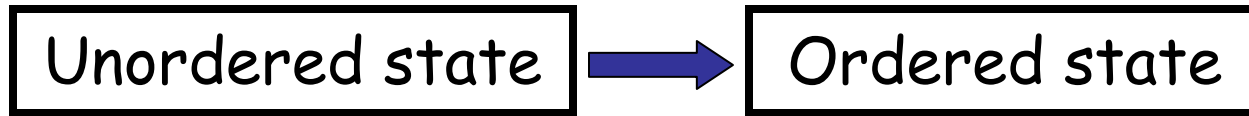
Solution: **Approximate**

Ideal weights
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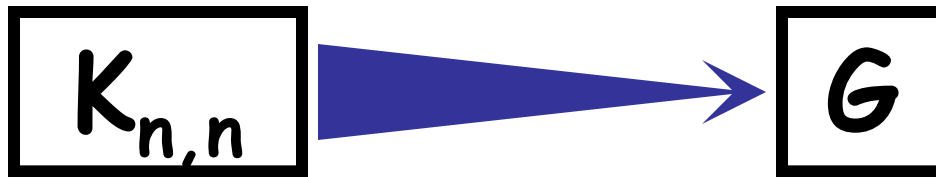


Simulated Annealing for Permanent



(EASY)

(DIFFICULT)



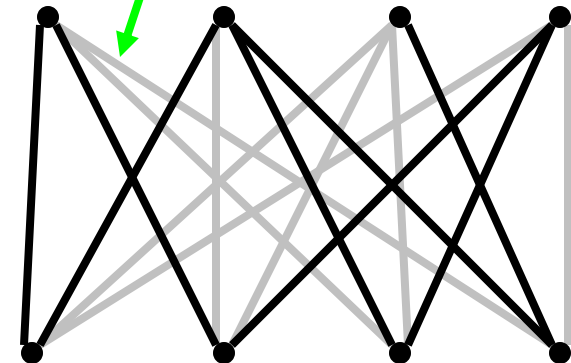
Solution: **Approximate**

Ideal weights

(for a matching with holes u,v):

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$\lambda = 1 \dots \sim 0$



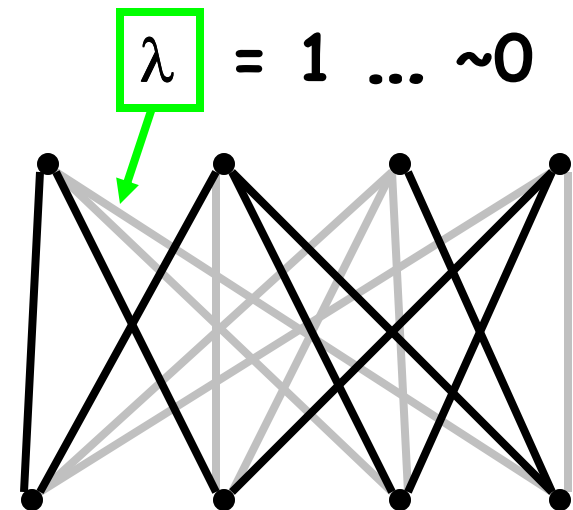
New Cooling Schedule

[Bezáková-Štefankovič-
Vazirani-Vigoda SODA '06]

$$\frac{(\# \text{ perfects})}{(\# \text{ nears with holes } u,v)}$$

← ratio of polynomials in λ
(don't know coefficients !)

Want: decrease λ so that **polynomials drop by ≤ 2 factor**



New Cooling Schedule

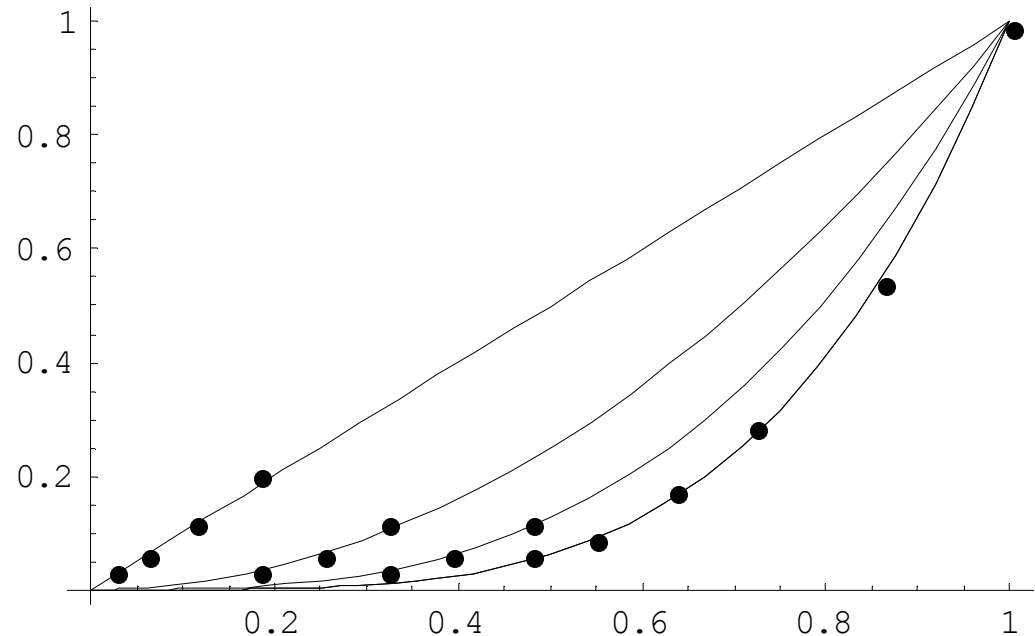
[Bezáková-Štefankovič-
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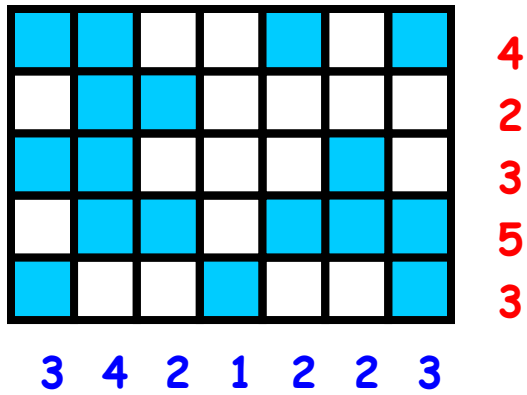
Want: decrease λ so that **polynomials drop by ≤ 2 factor**

Thm [BŠVV '06]:
Permanent of
nonnegative matrix
in time $O^*(n^7)$.

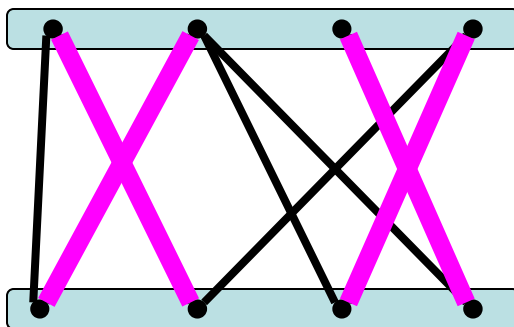


Problems

Binary contingency tables

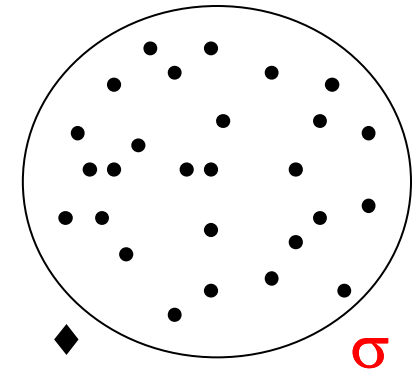


Permanent

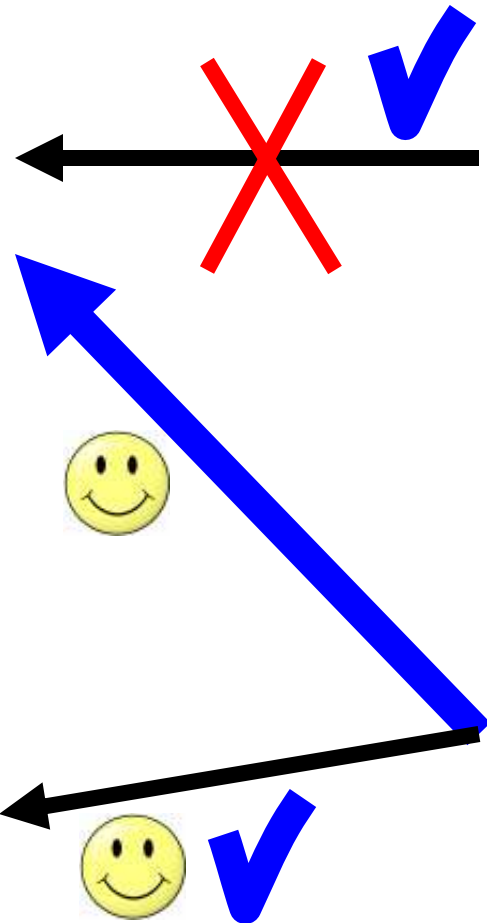


Heuristics

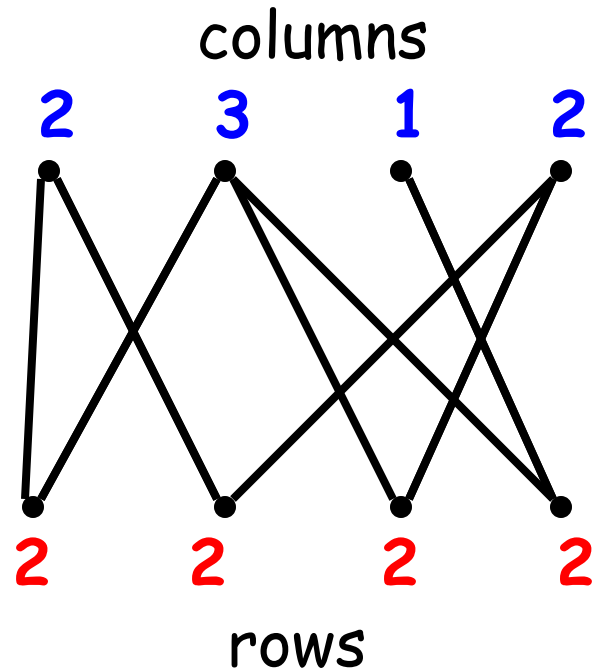
Importance sampling



Simulated annealing



BCT: Bipartite Graphs with Given Degrees

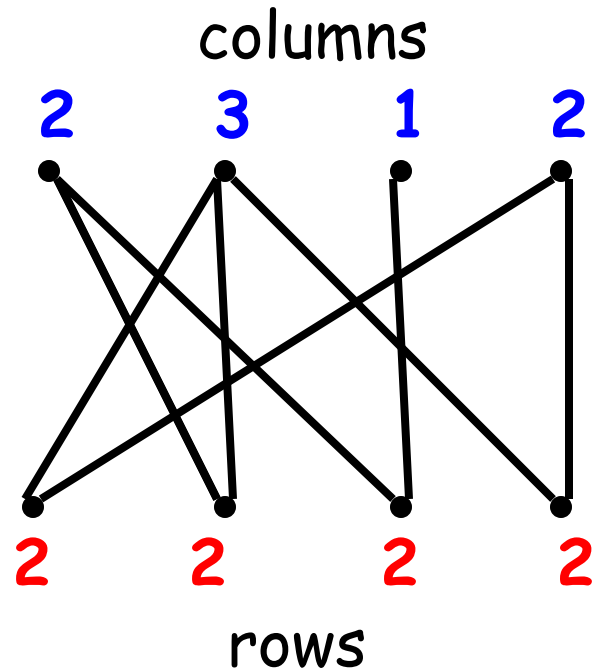


columns

1	1	0	0	2
1	0	0	1	2
0	1	0	1	2
0	1	1	0	2
2	3	1	2	

rows

BCT: Bipartite Graphs with Given Degrees



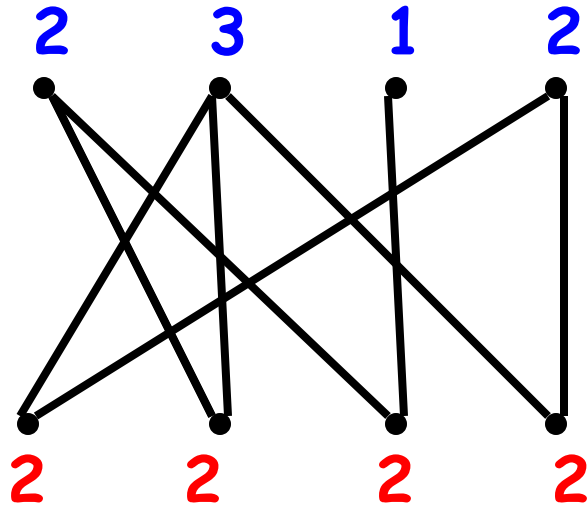
columns

0	1	0	1	2
1	1	0	0	2
1	0	1	0	2
0	1	0	1	2
	2	3	1	2

rows

Detailed description: A 4x4 bipartite graph adjacency matrix. The columns are labeled with degrees 2, 3, 1, 2. The rows are labeled with degrees 2, 2, 2, 2. The matrix is symmetric. The top row is (0, 1, 0, 1) with a red 2 to its right. The second row is (1, 1, 0, 0) with a red 2 to its right. The third row is (1, 0, 1, 0) with a red 2 to its right. The bottom row is (0, 1, 0, 1) with a red 2 to its right. The bottom labels are 2, 3, 1, 2.

BCT: Bipartite Graphs with Given Degrees



columns

0	1	0	1	2
1	1	0	0	2
1	0	1	0	2
0	1	0	1	2
	2	3	1	2

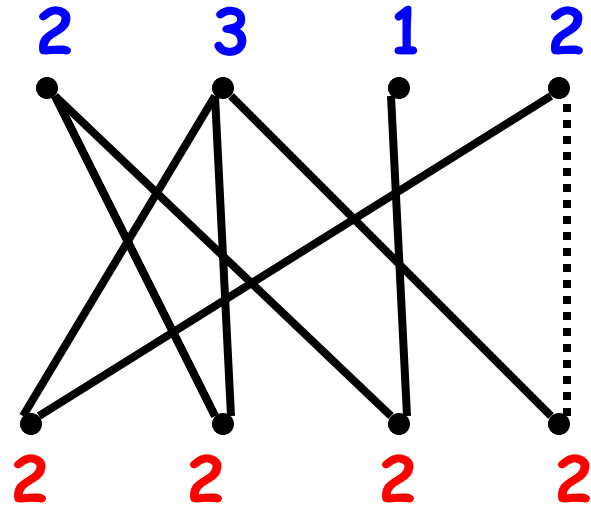
rows

"Sliding" Markov Chain on perfect and near tables

Perfect: remove a random edge

Near: slide edges or match

BCT: Bipartite Graphs with Given Degrees



columns

0	1	0	1	2
1	1	0	0	2
1	0	1	0	2
0	1	0	1	2
	2	3	1	2

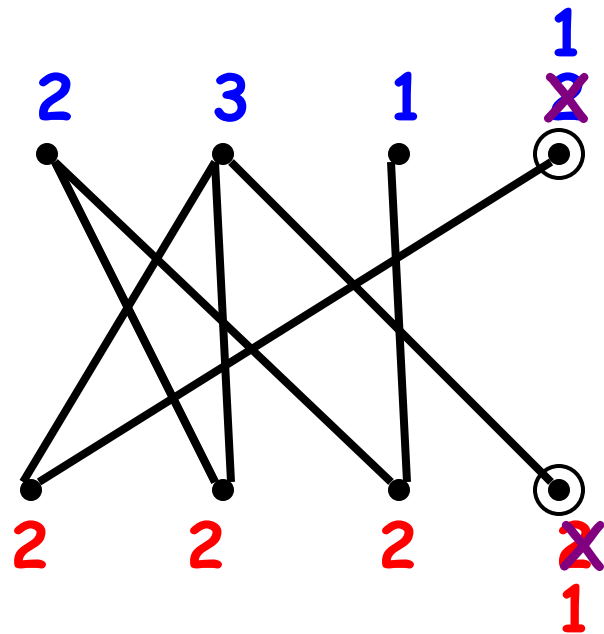
rows

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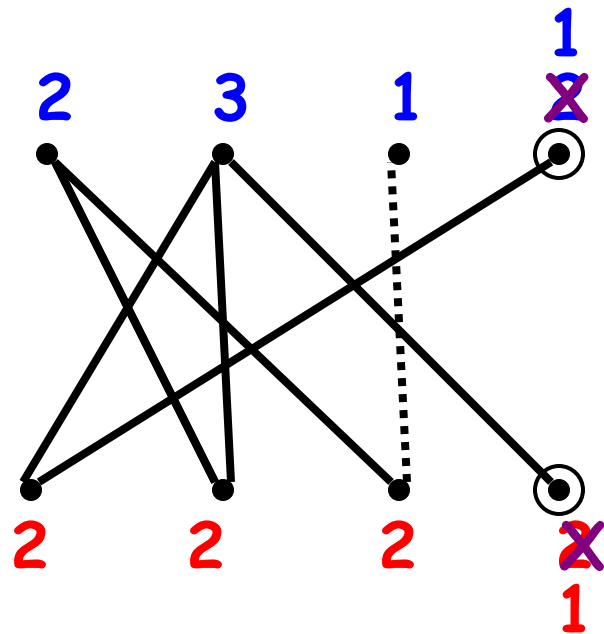
	columns				
	0	1	0	1	2
ROWS	1	1	0	0	2
	1	0	1	0	2
	0	1	0	0	2 1
		2	3	1	1 1

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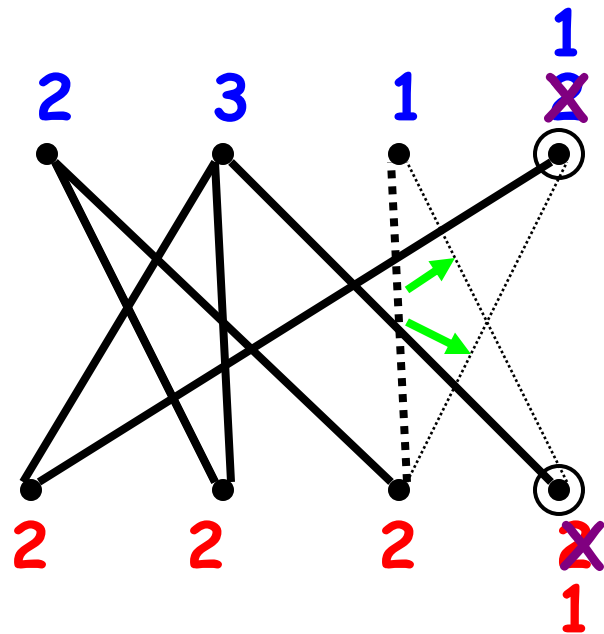
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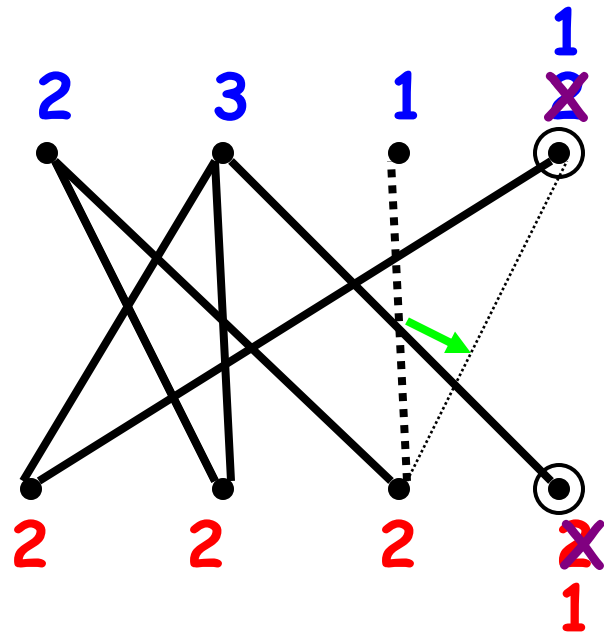
	columns				
	0	1	0	1	2
rows	1	1	0	0	2
	1	0	1	0	2
	0	1	0	0	2 1
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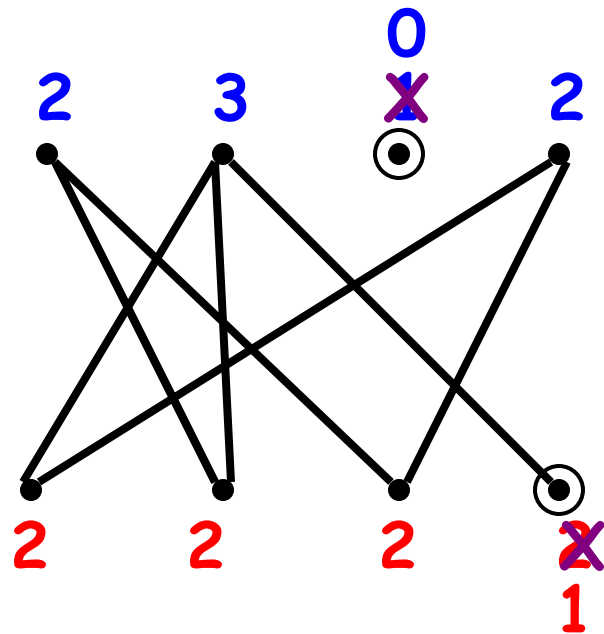
	columns				
	0	1	0	1	2
ROWS	1	1	0	0	2
	1	0	1	0	2
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	2	3	0 0	2

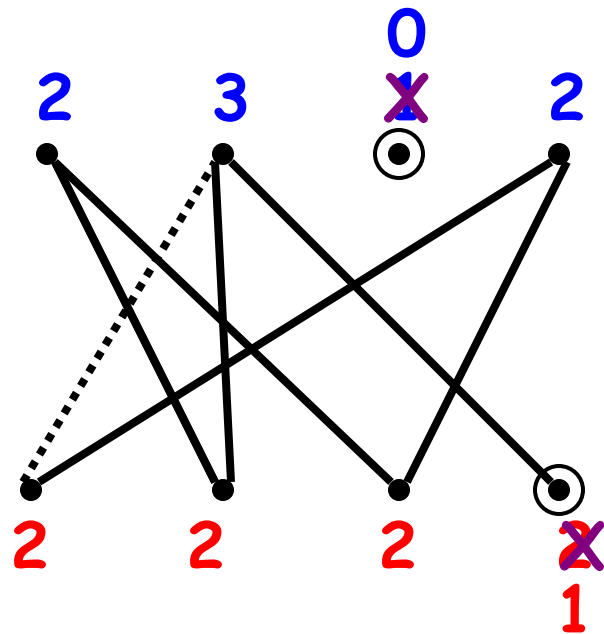
rows

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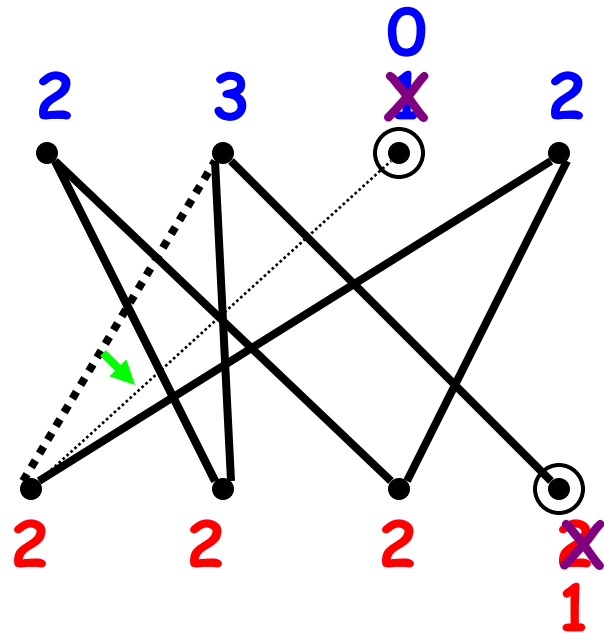
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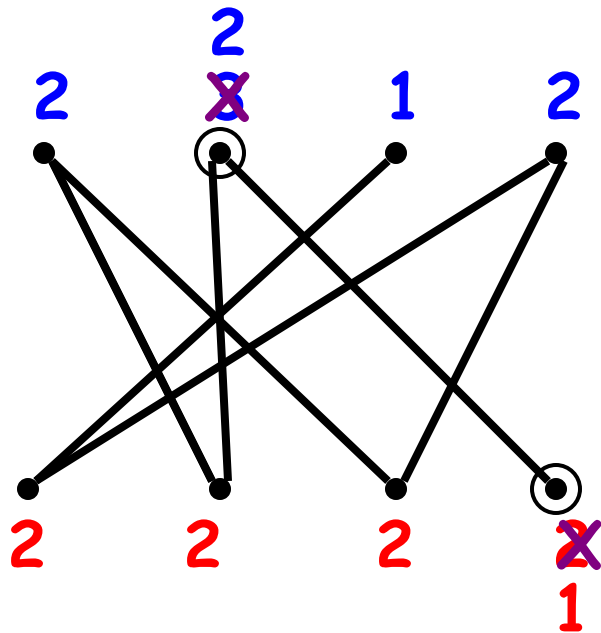
	columns				
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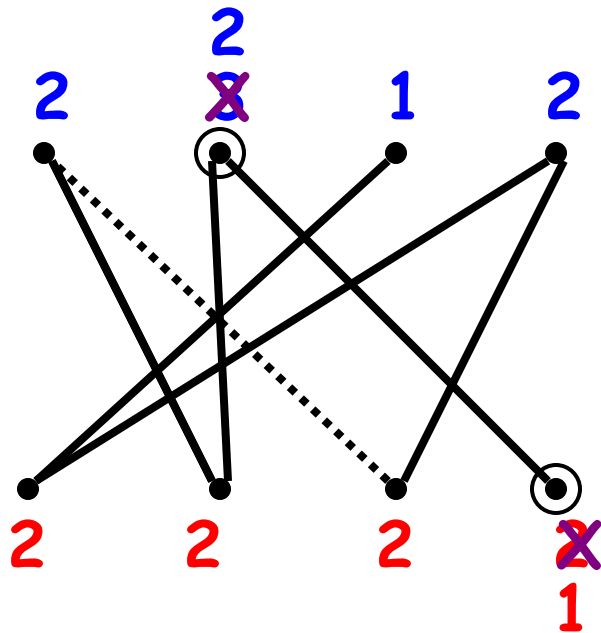
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1	0	1	0	2
1	0	0	1	2
0	1	0	0	2 1
	2	2	1	2
		2		

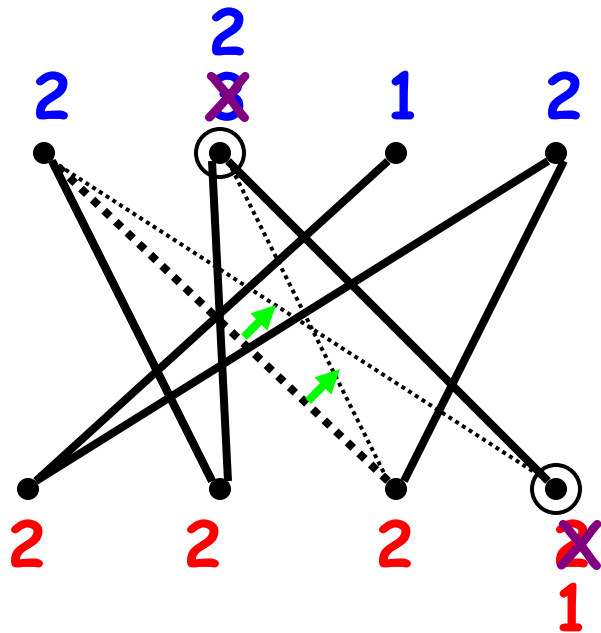
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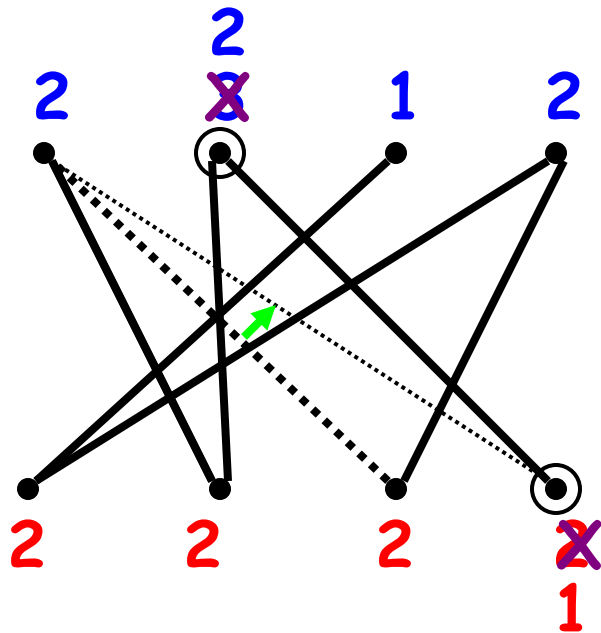
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0	1	0	0	2 1
	2	2 2	1	2

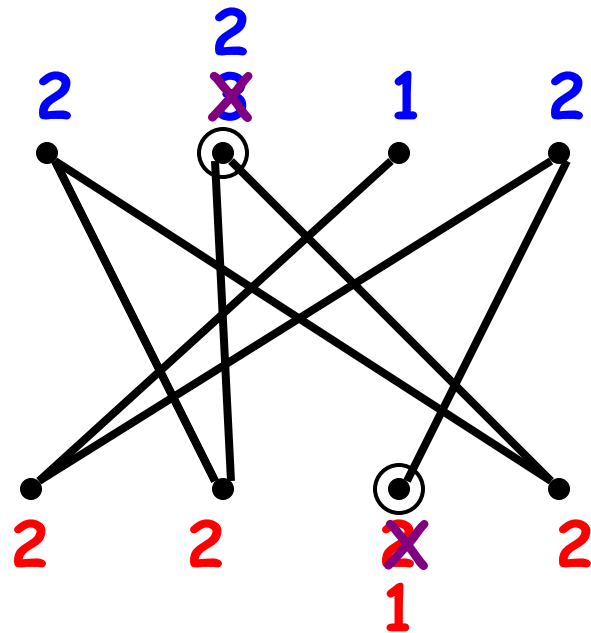
rows

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columns

0	1	0	1	2
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0	0	0	1	2 1
1	1	0	0	2
2	2 2	1	2	

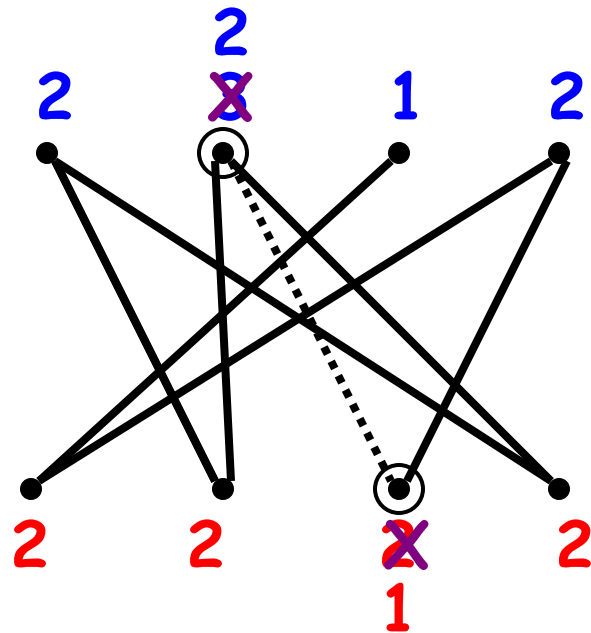
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0	1	0	1	2
1	0	1	0	2
0	0	0	1	2 1
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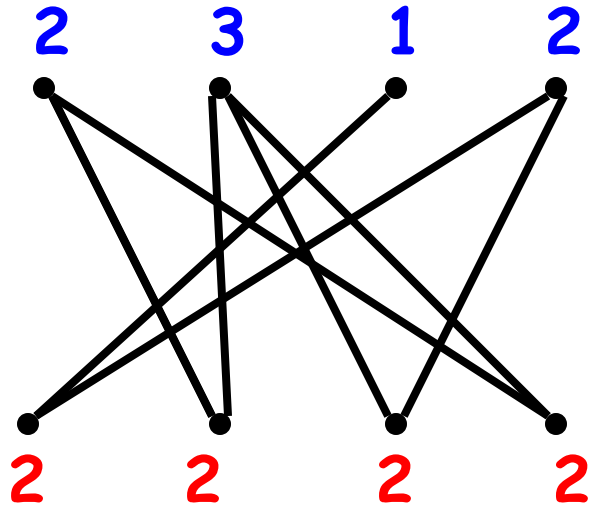
rows

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columns

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1	0	1	0	2
0	1	0	1	2
1	1	0	0	2
2	3	1	2	

rows

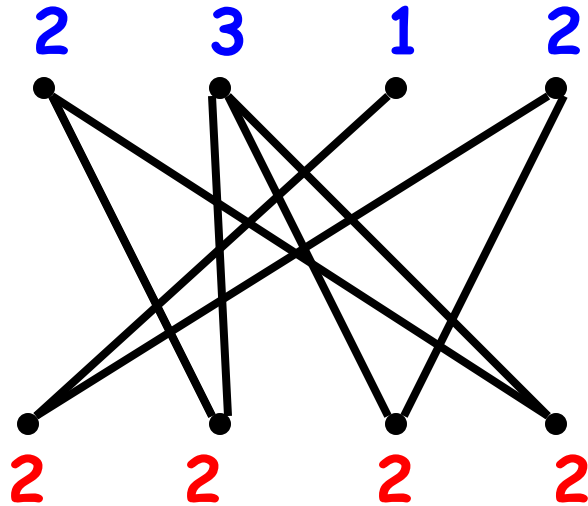
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Simulated Annealing for BCT ?

[Bezáková-
Bhatnagar-
Vigoda
SODA '06]



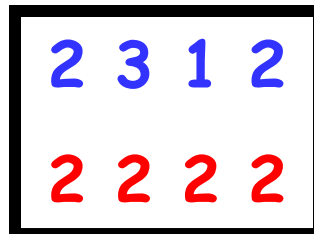
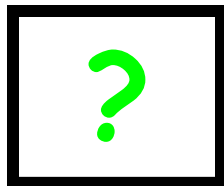
Ideal weights

$$\frac{(\# \text{ perfects})}{(\# \text{ nears with holes } u,v)}$$



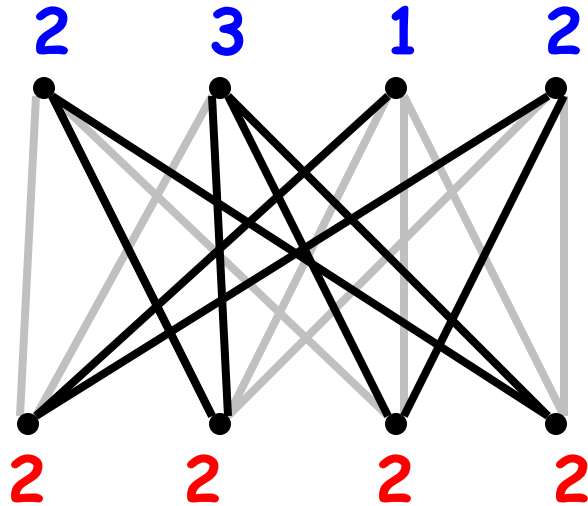
(EASY)

(DIFFICULT)



Simulated Annealing for BCT ?

Bezáková-
Bhatnagar-
Vigoda
SODA '06

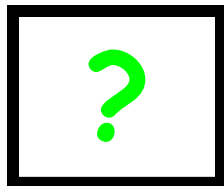


Ideal weights

$$\frac{(\# \text{ perfects})}{(\# \text{ nears with holes } u,v)}$$

Unordered state

(EASY)



Ordered state

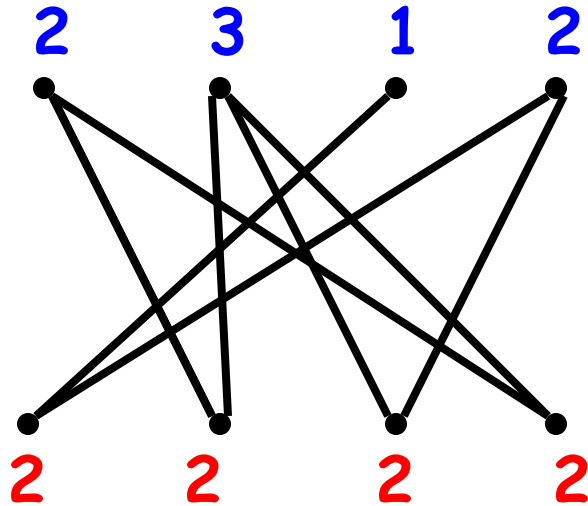
(DIFFICULT)

2 3 1 2
2 2 2 2 on $K_{n,n}$



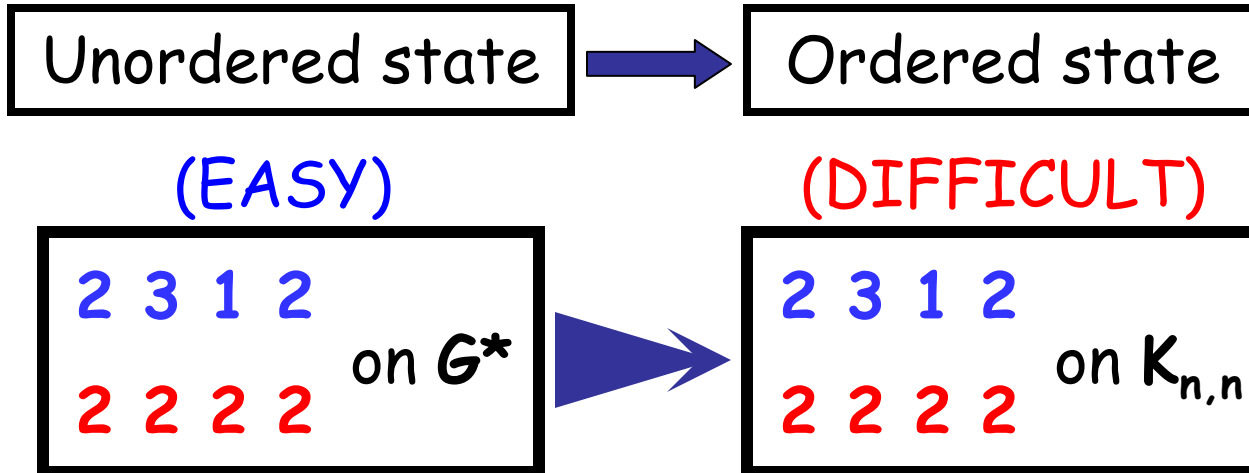
Simulated Annealing for BCT ?

[Bezáková-
Bhatnagar-
Vigoda
SODA '06]



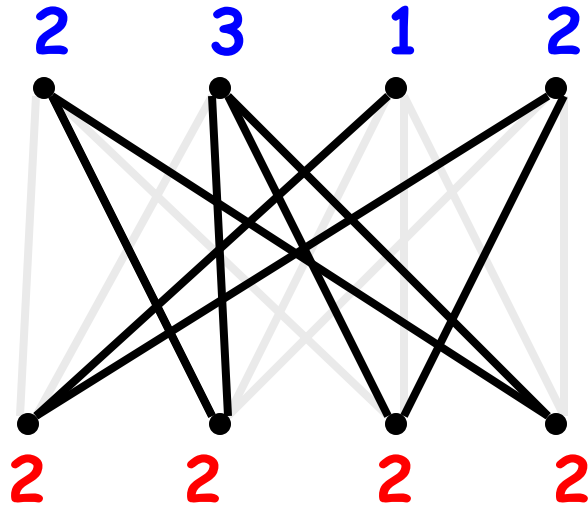
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Simulated Annealing for BCT ?

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SODA '06



Thm [BBV '06]:

G^* s.t. between any two vertices exists an "alternating" path of length ≤ 5 .

Unordered state

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2 3 1 2

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on G^*

Ordered state

(DIFFICULT)

2 3 1 2

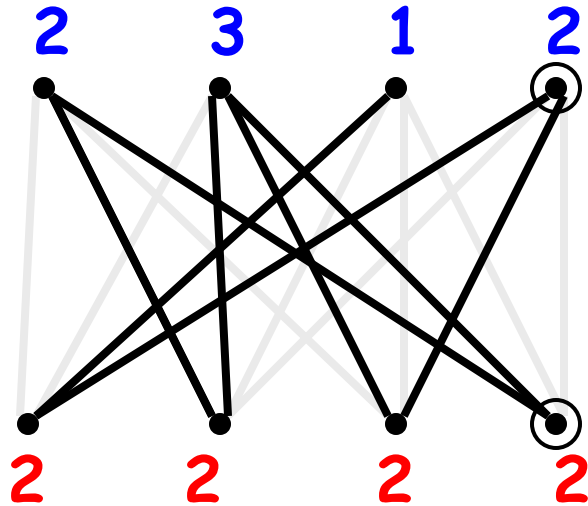
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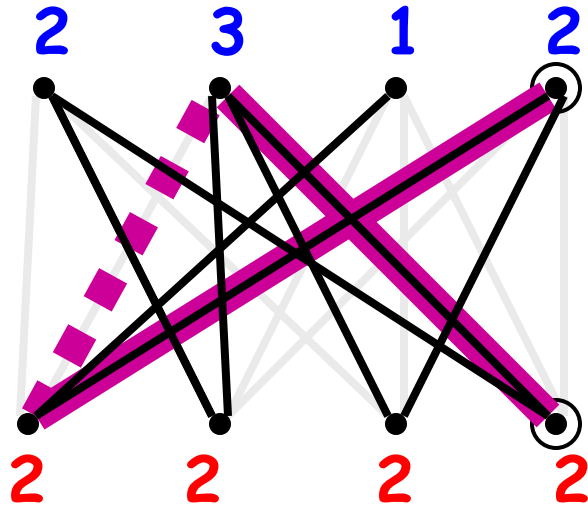
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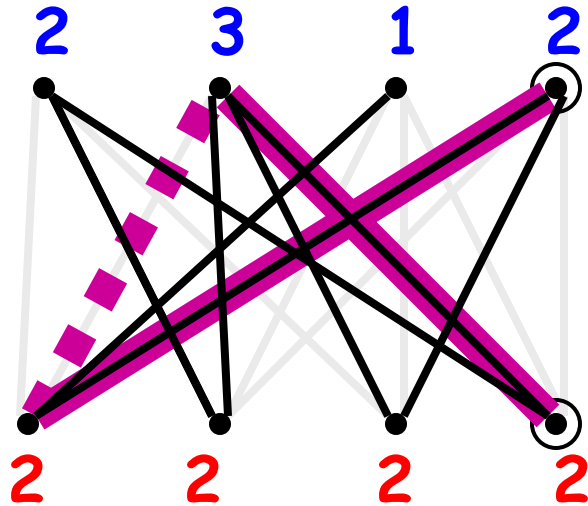
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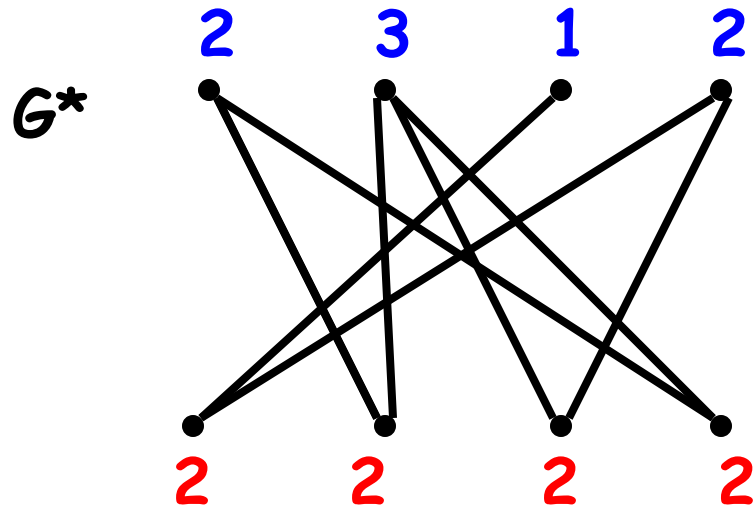
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Thm [BBV '06]:

Counting **subgraphs** (of a bipartite graph) **with given degrees** in poly-time.

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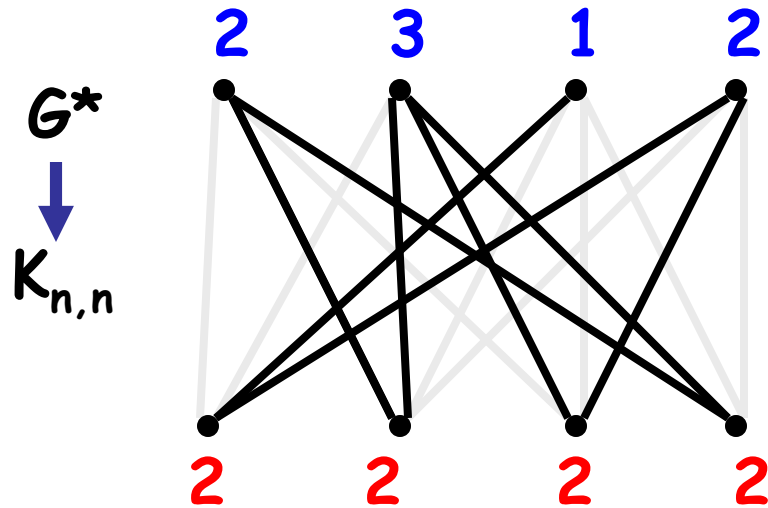
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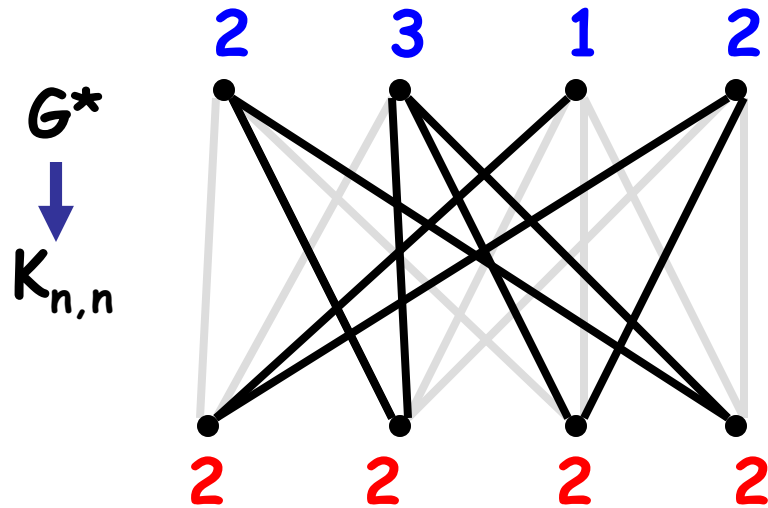
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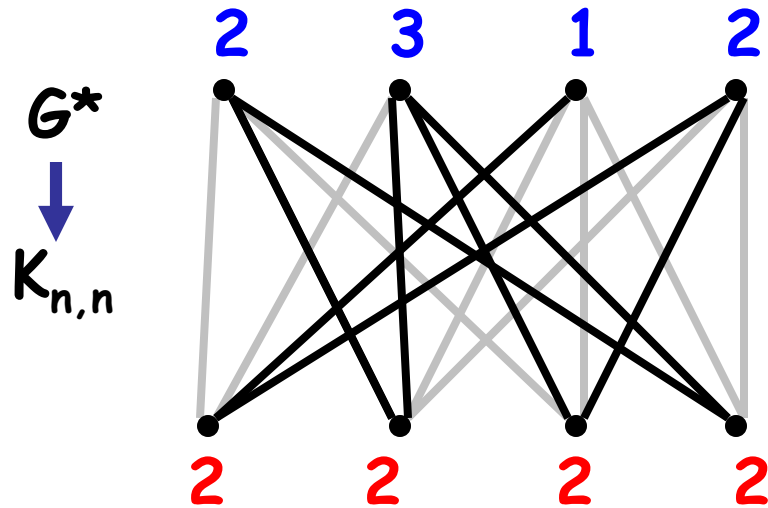
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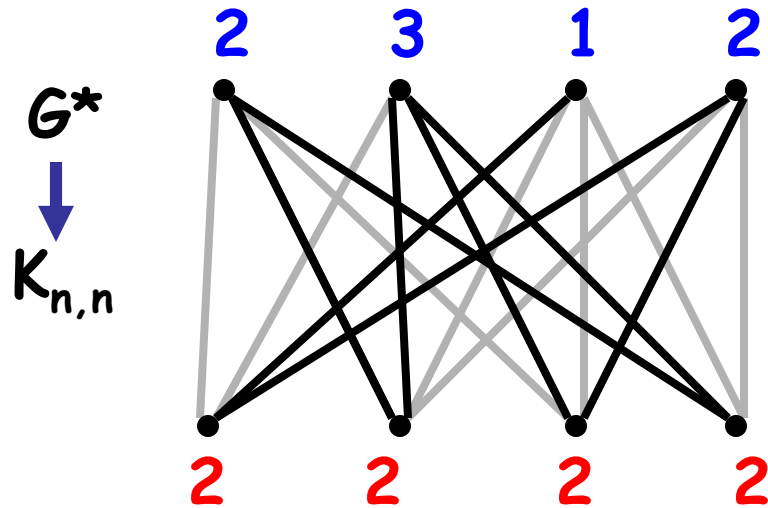
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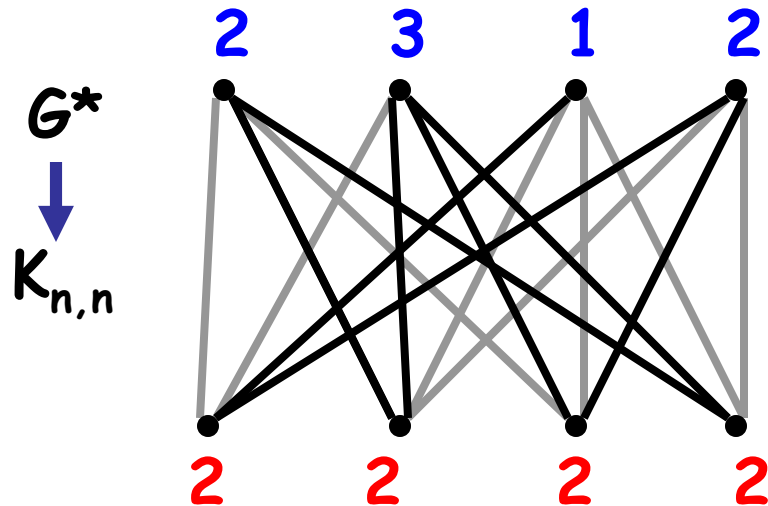
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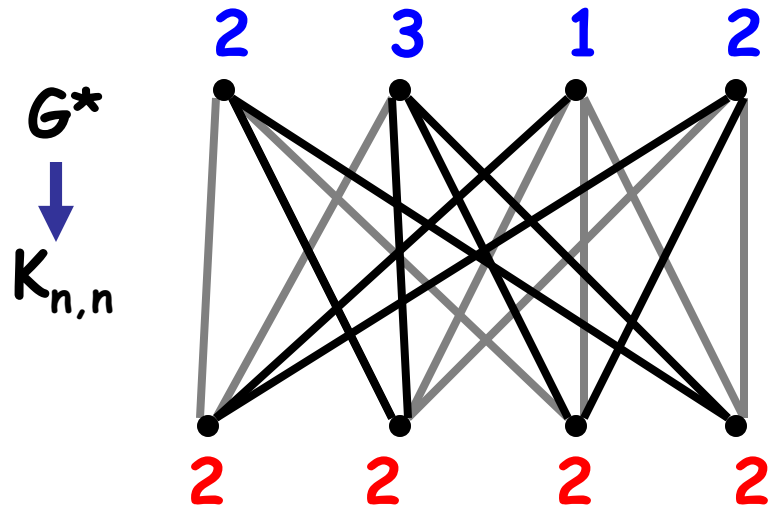
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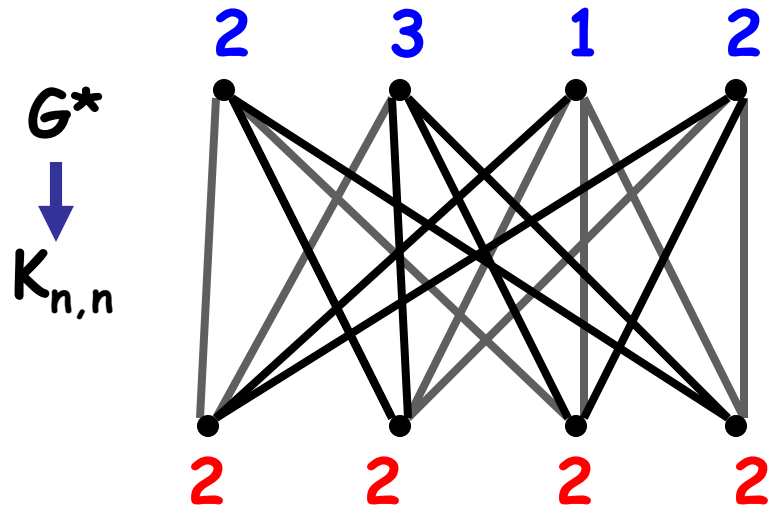
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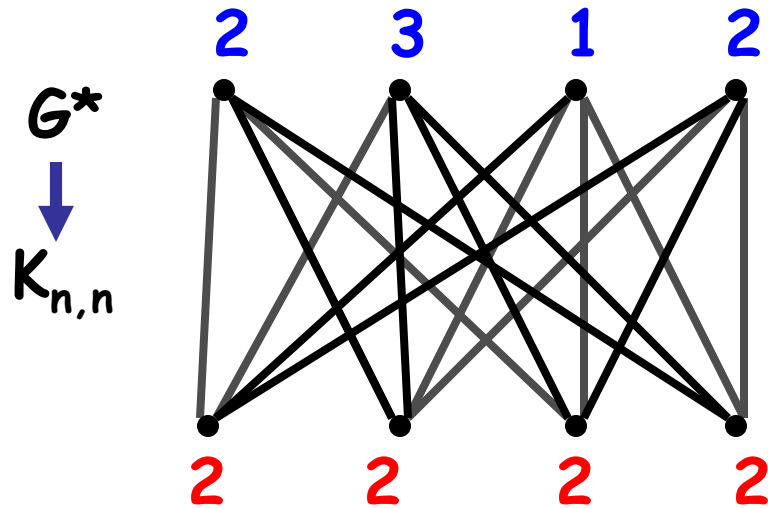
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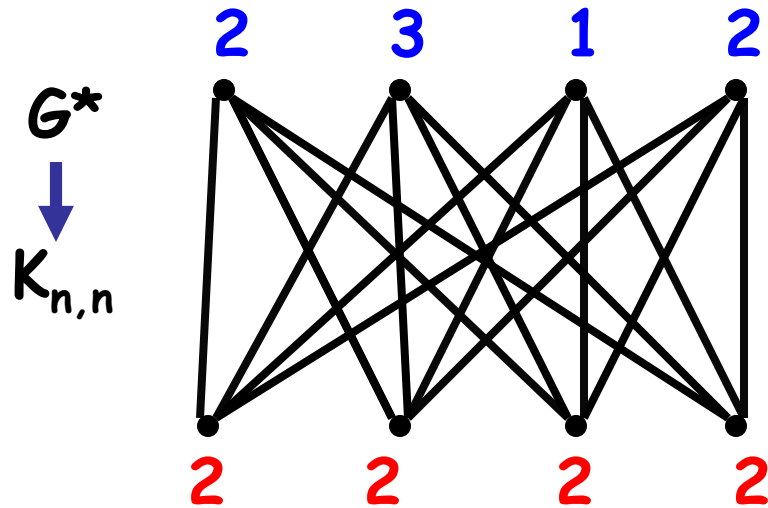
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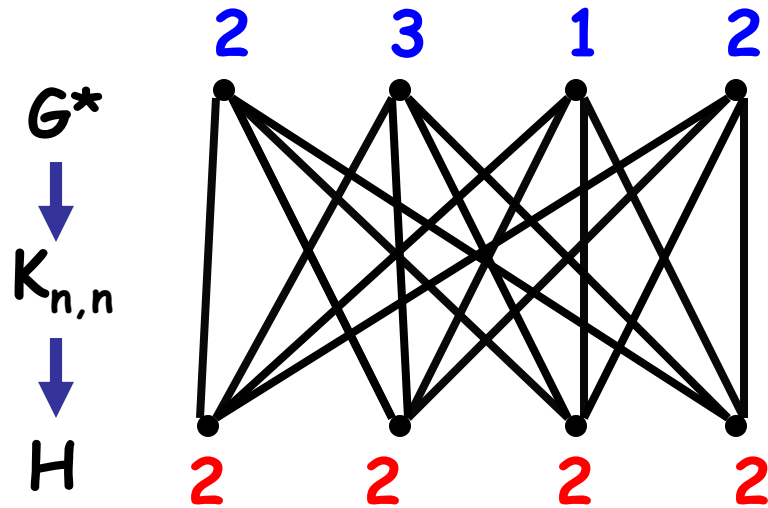
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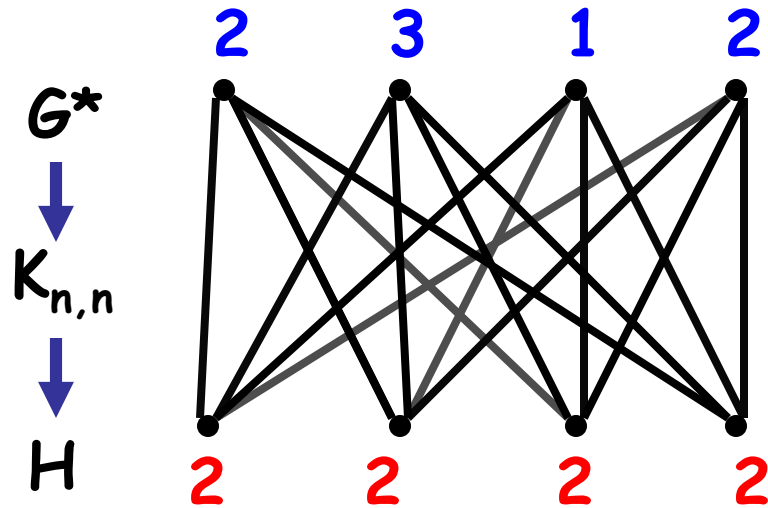
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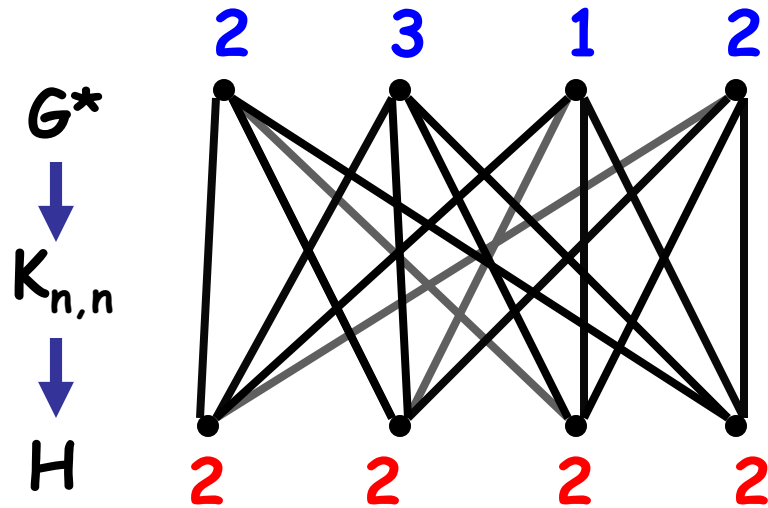
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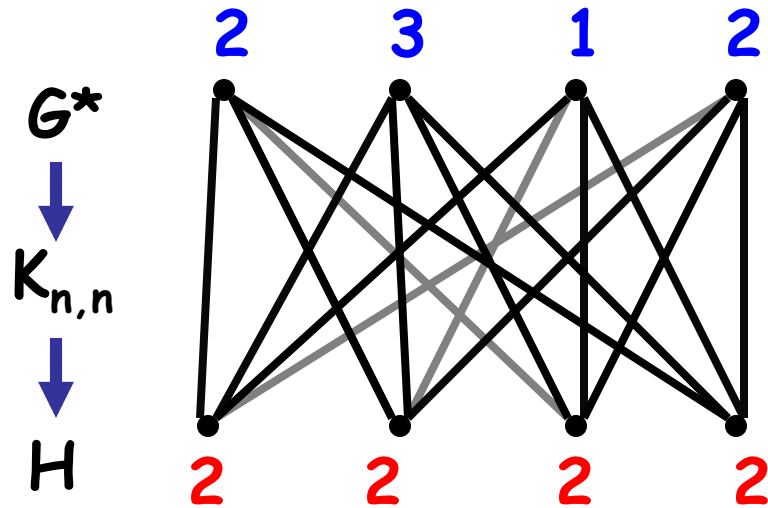
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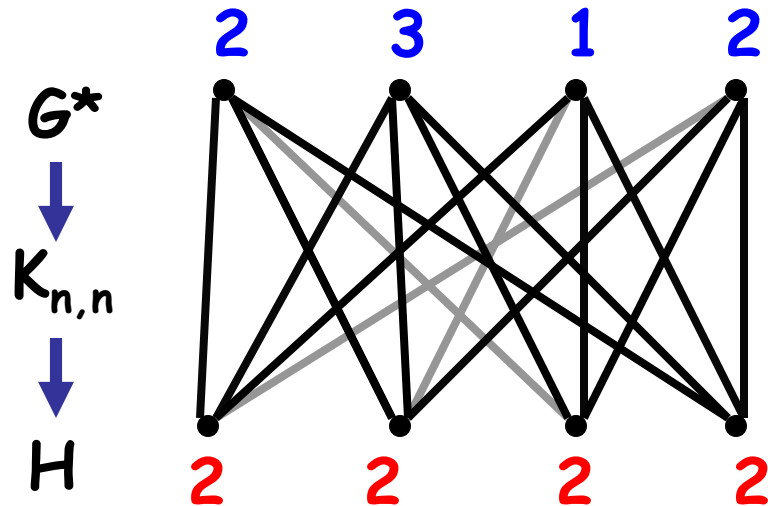
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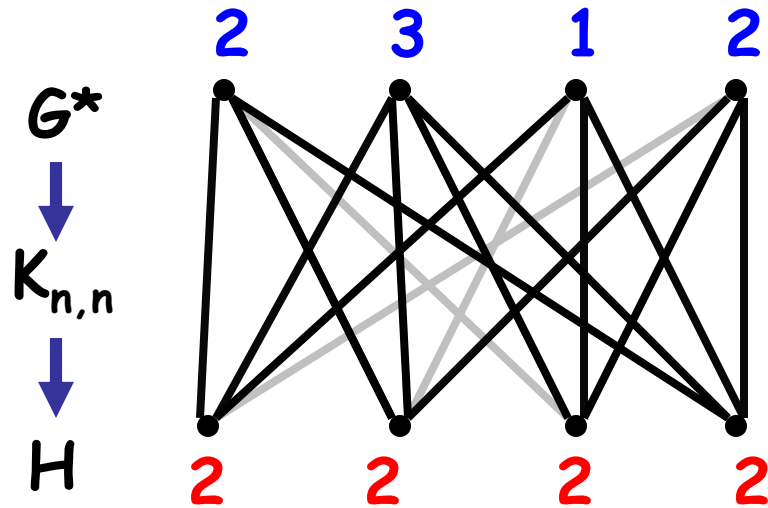
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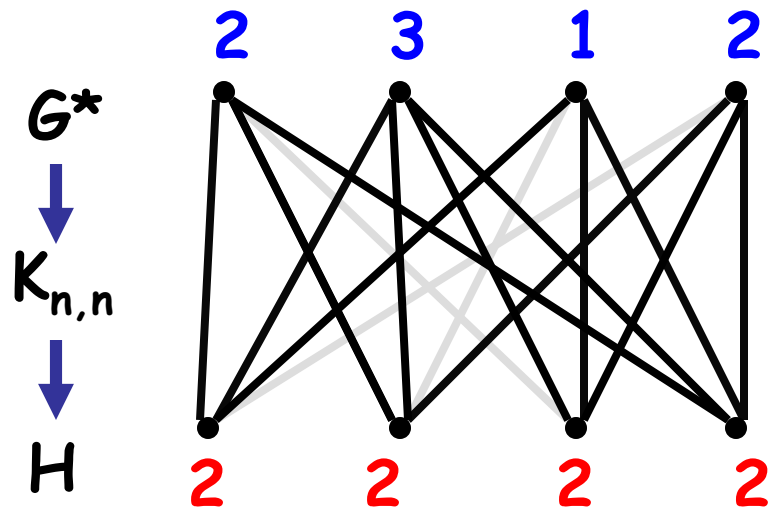
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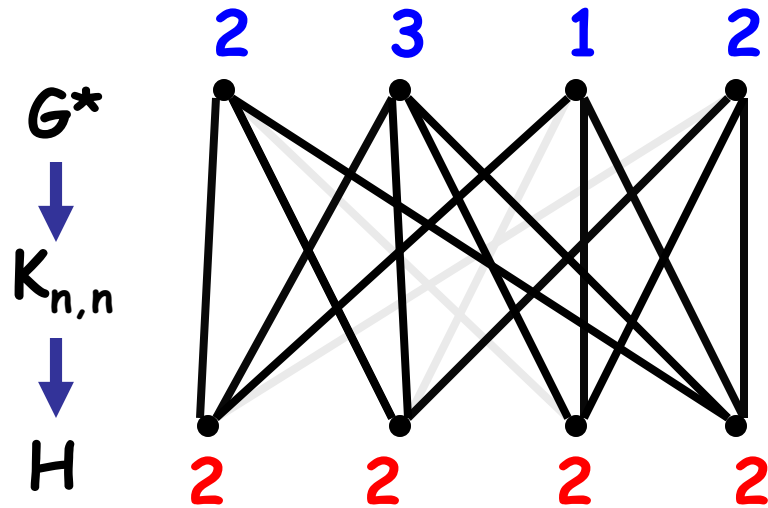
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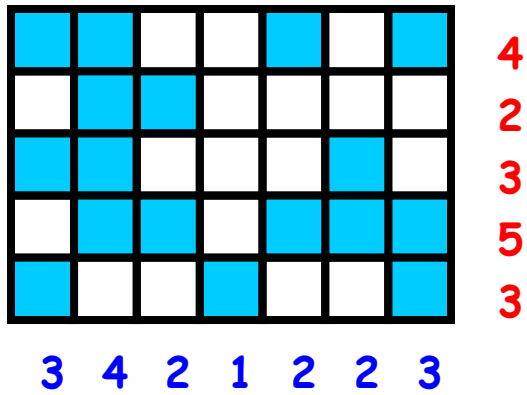
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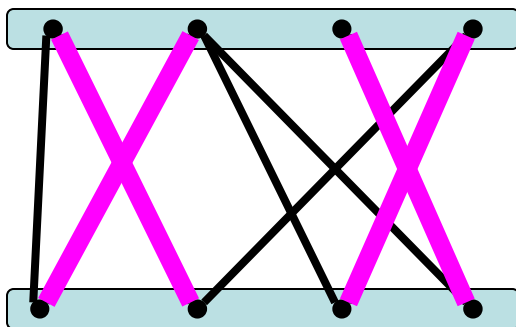


Problems

Binary contingency tables

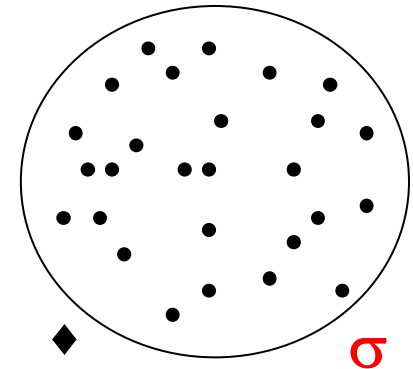


Permanent



Heuristics

Importance sampling



Simulated annealing

