

Approximate Counting via Correlation Decay

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Outline

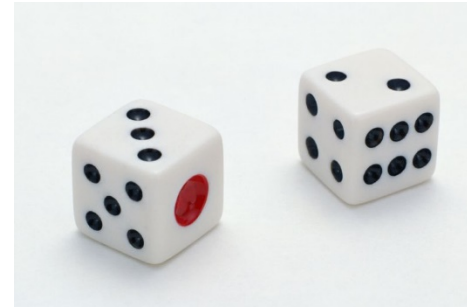
- Counting and probability distribution
- Two-state spin systems
- Multi-spin and Multi-part systems

Counting Problems

- SAT: Is there a satisfying assignment for a given a CNF formula?
- Counting SAT: How many?
- Counting Colorings of a graph
- Counting Independent sets of a graph
- Counting perfect matchings of a bipartite graph (Permanent)
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Counting Problems

- Probability



Blackjack Card
Counting
Learn How to Count
Cards- An Interactive
Games Quiz Book

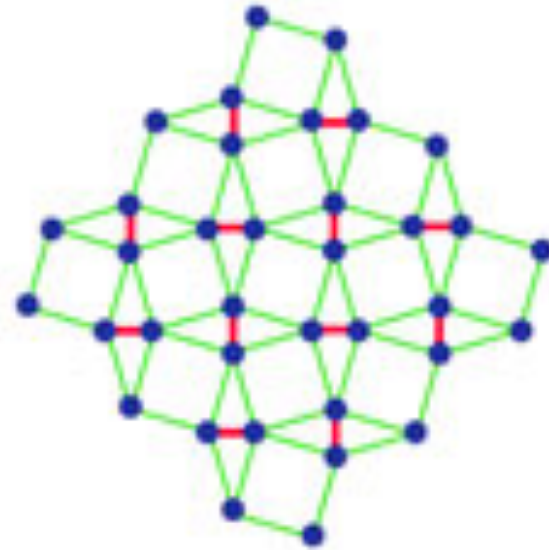
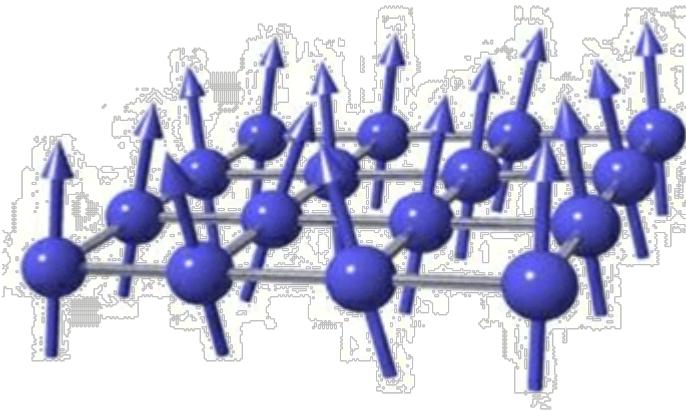


Interactive Games



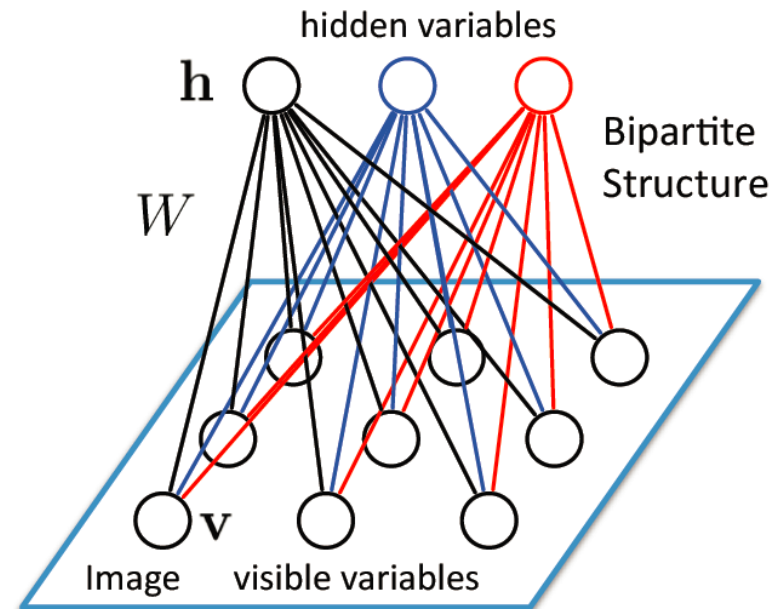
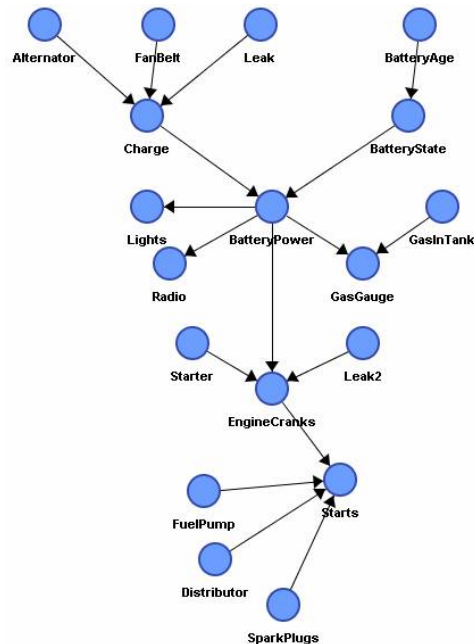
Counting Problems

- Probability
- Partition function on statistical physics



Counting Problems

- Probability
- Partition function on statistical physics
- Inference on Graphical Models



Counting Problems

- Probability
- Partition function on statistical physics
- Inference on Graphical Models
- Query on probabilistic database
- Optimization on stochastic model
-

Approximate Counting

- Let $\epsilon > 0$ be an approximation parameter and Z be the correct counting number of the instance, the algorithm returns a number Z^\uparrow such that $(1 - \epsilon)Z \leq Z^\uparrow \leq (1 + \epsilon)Z$, in time $\text{poly}(n, 1/\epsilon)$.
- Fully polynomial-time approximation scheme (FPTAS).
- Fully polynomial-time **randomized** approximation scheme (FPRAS) is its randomized version.

Counting vs Distribution

- $IS(G)$: the set of independent sets of graph G
- X is chosen from $IS(G)$ uniformly at random
- $P \downarrow G (v)$: the probability that v is not in X
- $\Pr(X = \emptyset) = 1 / |IS(G)|$
- $\Pr(X = \emptyset) = P \downarrow G \downarrow 1 (v \downarrow 1) P \downarrow G \downarrow 2 (v \downarrow 2) \dots P \downarrow G \downarrow n (v \downarrow n)$, where $G \downarrow 1 = G$, $G \downarrow i+1 = G \downarrow i - v \downarrow i$

Counting vs Distribution

- $1/|IS(G)| = P \downarrow G \downarrow 1 (v \downarrow 1) P \downarrow G \downarrow 2 (v \downarrow 2) \dots P \downarrow G \downarrow n (v \downarrow n)$
- If we can compute (estimate) $P \downarrow G (v)$, we can (approximately) compute $|IS(G)|$.
- FPRAS: Estimate $P \downarrow G (v)$ by sampling
- **FPTAS: Approximately compute $P \downarrow G (v)$ directly and deterministically**

Outline

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Spin Systems

- System $G=(V,E)$ and spin states $[q]$
- Configuration $\sigma:V\rightarrow[q]$
- Edge function $A:[q]\times[q]\rightarrow R^{\uparrow+}$
- Vertex function $b:[q]\rightarrow R^{\uparrow+}$
- Weight of a configuration

$$w(\sigma) = \left(\prod_{(u,v) \in E} A(\sigma(u), \sigma(v)) \right) \left(\prod_{v \in V} b(\sigma(v)) \right)$$

- Partition function: $Z_A(G) = \sum_{\sigma} w(\sigma)$

Gibbs Measure

- $\rho(\sigma) = w(\sigma) / Z_A(G)$ is a distribution over all configurations.
- We can define the marginal distribution of spins on a vertex $p_{\downarrow v}$.
- We can also fix the configuration of a subset Λ of the vertices as $\sigma_{\downarrow \Lambda}$, and define the conditional distribution of other vertex as $p_{\downarrow v \uparrow \sigma_{\downarrow \Lambda}}$.

Spin Systems

- A model in Statistical Physics and Applied Probability
- A framework of many combinatorial counting problems
- Have applications in AI, coding theory and so on.

Constraint Satisfaction Problems

- Graph coloring

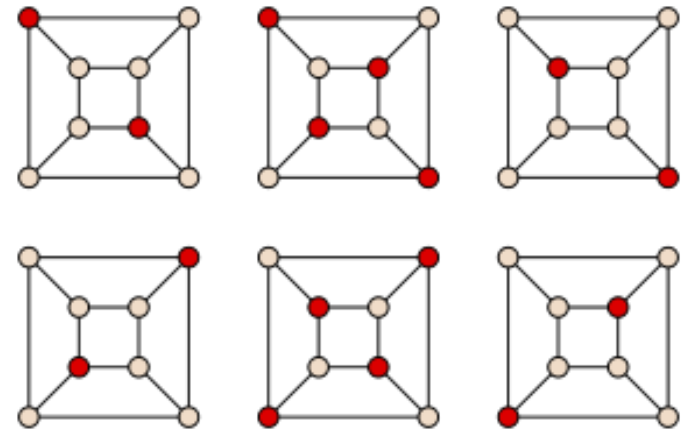
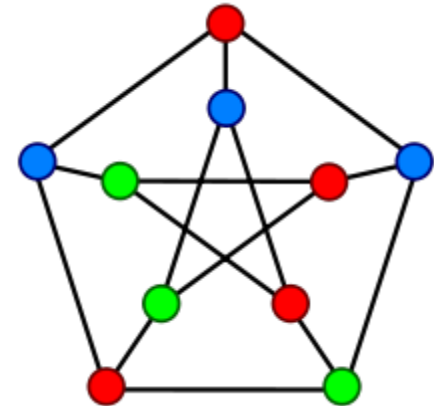
$$A = [\blacksquare 0 \& 1 \& 1 @ 1 \& 0 \& 1 @ 1 \& 1 \& 0]$$

- Independent set

$$A = [\blacksquare 1 \& 1 @ 1 \& 0]$$

Partition function:

Counting the number of solutions



Some Tasks

- Computing (conditional) marginal
- Sampling (wrt. the distribution)
- Computing partition function
- They are all related

Belief Propagation

- Also called **message passing algorithms**
- Widely used in statistical physics, machine learning and other applications
- Behave well in certain applications but without much theoretical justification unless the graph is a tree or very special

Belief Propagation

- Start with some initial guess
- Recursively refine the guess according to its local neighborhood
- Stop and output the current state after a fixed number of recursions
- Convergence and correctness?

Weak Correlation Decay

A spin system on a family of graphs is said to have exponential correlation decay if for any graph $G=(V,E)$ in the family, any $v \in V, \Lambda \subset V$ and $\sigma \downarrow \Lambda, \tau \downarrow \Lambda \in \{0,1\}^{\uparrow \Lambda}$,

$$|p \downarrow v \uparrow \sigma \downarrow \Lambda - p \downarrow v \uparrow \tau \downarrow \Lambda| \leq \exp(-\Omega(d(v,\Lambda))).$$

Correlation Decay

A spin system on a family of graphs is said to have exponential correlation decay if for any graph $G=(V,E)$ in the family, any $v \in V, \Lambda \subset V$ and $\sigma \downarrow \Lambda, \tau \downarrow \Lambda \in \{0,1\}^{\uparrow \Lambda}$,

$$|p \downarrow v \uparrow \sigma \downarrow \Lambda - p \downarrow v \uparrow \tau \downarrow \Lambda| \leq \exp(-\Omega(d(v,S))),$$

where $S \subset \Lambda$ is the subset on which $\sigma \downarrow \Lambda$ and $\tau \downarrow \Lambda$ differ.

Correlation Decay

- Significance of long distance interaction
- Effect of boundary condition
- Effect of the initial guess in Belief Propagation
- The big question: **Characterize systems with correlation decay**

Two-State Spin System

- After normalization: $A = \begin{bmatrix} \beta & 1 \\ 1 & \gamma \end{bmatrix}$,
 $b = \begin{bmatrix} \lambda \\ 1 \end{bmatrix}$
- Anti-ferromagnetic system: $\beta\gamma < 1$
- Hardcore model : $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$
- Ising model: $A = \begin{bmatrix} \beta & 1 \\ 1 & \beta \end{bmatrix}$

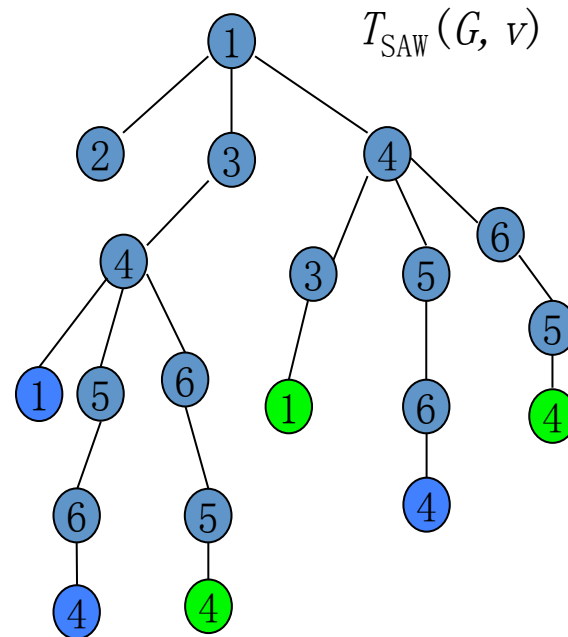
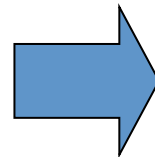
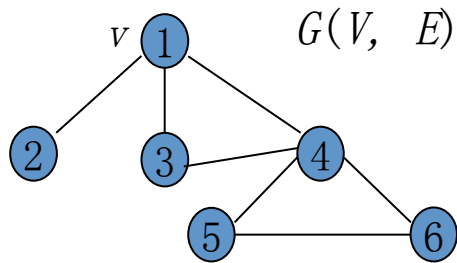
Uniqueness Condition

- $f(x) = \lambda(\beta x + 1/x + \gamma)$
- Let $x = f(x)$ be the fixed point of f .
- The system is called d -uniqueness if $|f'(x)| < 1$
- This can be numerically tested, but there is no closed form in general.

Hardcore Model [Weitz 2006]

- Strong correlation decay holds on all graphs with maximum degree at most Δ iff the uniqueness condition holds on infinite Δ -regular tree.
- Self Avoiding walk(SAW) tree: transform a general graph to a tree and then keep the marginal distribution for the root.
- Monotonicity: any tree with degree at most Δ decays at least as fast as the complete Δ -regular tree.

Self Avoiding Walk Tree



Self Avoiding Walk Tree

- It is enough to prove correlation decay on trees.
- The entire SAW tree is of exponential size comparing to the original graph.
- It only works for two-state spin systems.

Ising Model

[Sinclair, Srivastava, and Thurley 2011]

- Strong correlation decay holds on all graphs with maximum degree at most Δ iff the uniqueness condition holds on infinite Δ -regular tree.
- Have the same monotonicity property as hardcore model.

Non-Monotonicity

- The monotonicity does not hold for general two-state spin systems
- We need to prove correlation decay for general trees (with degrees up to Δ) rather than regular trees
- The previous techniques cannot be used

Our Results [Li, L., Yin 2012,13]

- The system is of correlation decay on all the graphs with maximum degree Δ iff the system exhibits uniqueness on **all the infinite regular trees up to degree Δ** .
- In particular, if the system exhibits uniqueness on infinite regular trees of all degrees, then the system is of correlation decay on all graphs.

Our Results [Li, L., Yin 2012,13]

- We obtain a FPTAS as long as the system satisfies the uniqueness condition.
- There is a matching hardness result [Sly,Sun 2012]: It is NP-hard if the system does not satisfy the uniqueness condition.

From correlation decay to FPTAS

- Marginal distribution $p \downarrow v \uparrow \sigma \downarrow \Lambda$ \rightarrow partition function

$$w(0)/z = p \downarrow v \downarrow 1 \quad p \downarrow v \downarrow 2 \quad \uparrow \sigma(1) = 0 \quad \dots \quad p \downarrow v \downarrow n$$
$$\uparrow \sigma(i) = 0, i = 1, 2, \dots, n-1$$

- Correlation decay \rightarrow estimate $p \downarrow v \uparrow \sigma \downarrow \Lambda$ by a local neighborhood: $O(\log n)$ depth of the SAW tree.
- How about unbounded degree?

Computational Efficient Correlation Decay

- M-based depth:
 - $L \downarrow M(\text{root}) = 0$;
 - $L \downarrow M(u) = L \downarrow M(v) + \lceil \log \downarrow M(d+1) \rceil$, if u is one of the d children of v .
- Exponential correlation decay with respect to M-based depth.
- Computational efficient correlation decay supports FPTAS for general graph.

Proof Sketch for Correlation Decay

- Self avoiding walk tree and recursion relation on tree: $p \downarrow v = f(p \downarrow v \downarrow 1, p \downarrow v \downarrow 2, \dots, p \downarrow v \downarrow d)$

- Estimate the error for one recursive step:

$$\epsilon \downarrow v = \frac{\partial f}{\partial p \downarrow v \downarrow 1} \epsilon \downarrow v \downarrow 1 + \frac{\partial f}{\partial p \downarrow v \downarrow 2} \epsilon \downarrow v \downarrow 2 + \dots + \frac{\partial f}{\partial p \downarrow v \downarrow d} \epsilon \downarrow v \downarrow d$$

$$|\epsilon \downarrow v| \leq (|\frac{\partial f}{\partial p \downarrow v \downarrow 1}| + |\frac{\partial f}{\partial p \downarrow v \downarrow 2}| + \dots + |\frac{\partial f}{\partial p \downarrow v \downarrow d}|) \max(|\epsilon \downarrow v \downarrow i|)$$

- $(|\frac{\partial f}{\partial p \downarrow v \downarrow 1}| + |\frac{\partial f}{\partial p \downarrow v \downarrow 2}| + \dots + |\frac{\partial f}{\partial p \downarrow v \downarrow d}|) < 1$

Potential Function

- This may not be correct stepwise. We use a potential function to amortize it.
- Let $\phi: R^+ \rightarrow R^+$ be a bijective function. $q \downarrow v = \phi(p \downarrow v)$, $q \downarrow v \downarrow i = \phi(p \downarrow v \downarrow i)$.
- $q \downarrow v = \phi(f(\phi^{\uparrow-1}(q \downarrow v \downarrow 1), \phi^{\uparrow-1}(q \downarrow v \downarrow 2), \dots, \phi^{\uparrow-1}(q \downarrow v \downarrow d)))$.
- Then we show that the error for q is stepwise decreased by a constant factor.
- The main difficulty is to find the potential function ϕ .

A mathematical problem

- $f: \mathbb{R}^d \rightarrow \mathbb{R}^d$
- Contraction: $\forall x, \delta \in \mathbb{R}^d$, we have $\|Jf(x)\delta\| \leq c \|\delta\|$ for some constant $c < 1$
- For some bijective mapping $\phi: \mathbb{R}^d \rightarrow \mathbb{R}^d$, such that $\phi \circ f \circ \phi^{-1}$ is contracted
- The effect of ϕ can be viewed as a local Riemann metric

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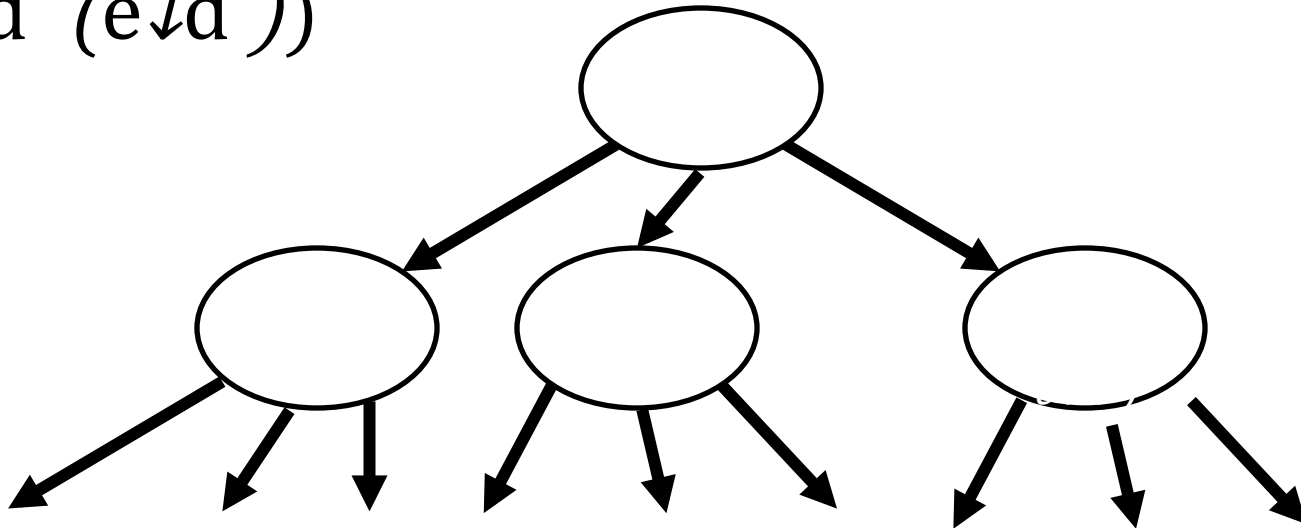
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Multi-spin and Multi-part systems

- Multi-spin: the domain of the variables is larger than two
- Multi-part: each constraint involves more than two variables
- Obstacle: SAW tree does not work

Computation Tree

- Relate the probability $P \downarrow G (e)$ to these of its neighbors in smaller instances.
- $P \downarrow G (e) = f(P \downarrow G \downarrow 1 (e \downarrow 1), P \downarrow G \downarrow 2 (e \downarrow 2), \dots, P \downarrow G \downarrow d (e \downarrow d))$



Correlation Decay

Truncate the computation tree at depth L , we can compute an estimation $P \downarrow G \uparrow L (e)$.

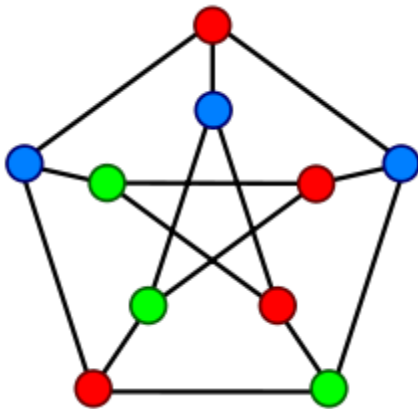
The system is called of exponential correlation decay if

$$|P \downarrow G \uparrow L (e) - P \downarrow G (e)| \leq \exp(-L).$$

We can estimate $P \downarrow G (e)$ by set $L = O(\log n + \log 1/\epsilon)$

Coloring

- It is NP-hard if $q < \Delta$ and there is always a solution if $q \geq \Delta$.
- FPTAS if $q > 2.8432\Delta + C$ [Gamarnik, Katz 07]
- **FPTAS if $q > 2.581\Delta + 1$ [L., Yin 2013]**



q : number of colors

Δ : maximum degree of the graph

Multi-spin System

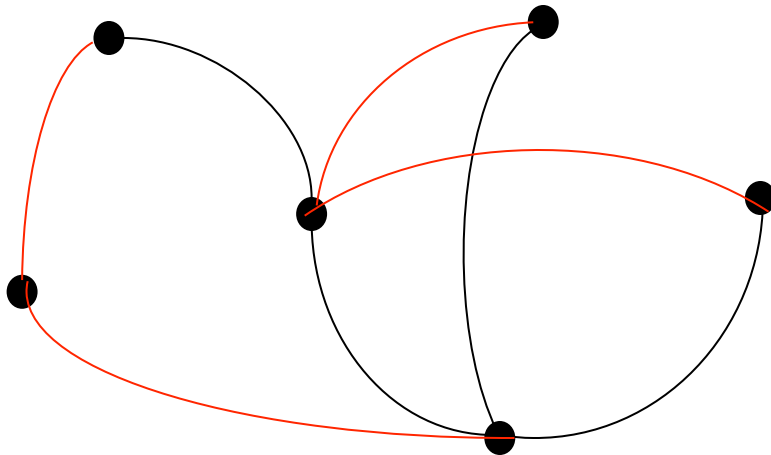
- Also called **(Pairwise) Markov Random Field**
- $c \downarrow A = \max A(x, y) / A(z, w)$
- We obtain an FPTAS when $3\Delta(c \downarrow A - 1) < 1$
- Potts model: inverse temperature $|\beta| = O(1/\Delta)$

Previously

- $(c \downarrow A \uparrow \Delta - c \downarrow A \uparrow - \Delta) \Delta q \uparrow \Delta < 1$
- $|\beta| = O(1/\Delta q \uparrow \Delta)$

Counting Edge Covers

- A set of edges such that every vertex has at least one adjacent edge in it



Counting Edge Covers

- A set of edges such that every vertex has at least one adjacent edge in it
- FPRAS for 3-regular graphs based on Markov Chain Monte Carlo [Bezakova, Rummel 2009].
- **FPTAS for general graph.** [Lin, Liu, L. 2014]

Counting Monotone CNF [Liu, L. 2015]

- A CNF formula is monotone if each variable appears positively.
- We give a FPTAS to count the number of solutions for a monotone CNF when each variable appears at most five times.
- It is NP-hard to approximately count if we allow a variable to appear six times.

Taking home messages

- Counting is in the heart of many computational problems, especially these related to probabilistic distribution.
- Correlation decay offers a new promising approach to design approximate counting algorithms.