

MIXING TIME OF SWITCH MARKOV CHAINS AND *P*-STABILITY OF DEGREE SEQUENCES

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- *Problem*: given a non-negative integer sequence d of even sum, generate a graph $G \in \mathcal{G}(d)$ with degree sequence d , uniformly at random (labeled vertices)
- *Motivation*:
 - network science: hypothesis testing
 - there is usually only one observed network, so experiments cannot be repeated
 - null model: structure of network can be explained by the properties of the degree sequence
 - via sampling, statistical parameters of the null model can be measured
 - testing software, algorithms
 - simulations

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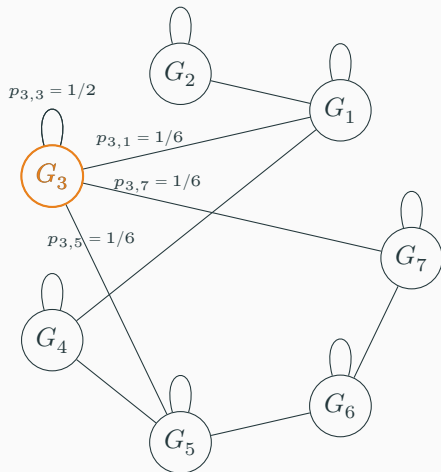
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- **Monte Carlo Markov Chain (MCMC) methods \Rightarrow**

MARKOV CHAINS - A REMINDER

Our chains transition from state i to state j with some probability $p_{i,j} = p_{j,i}$ (symmetric), independently of time and previous steps

$$\forall i \sum_j p_{i,j} = 1$$

$t = 1$

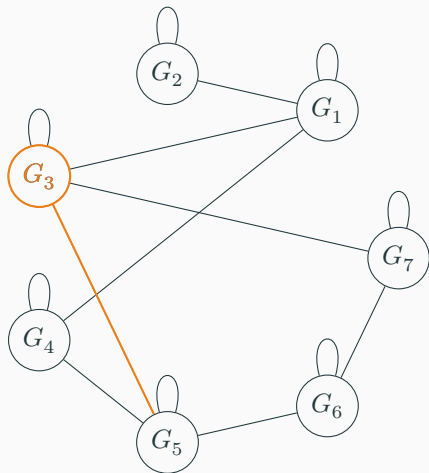


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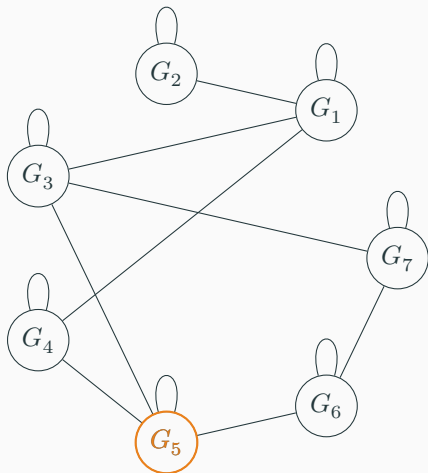


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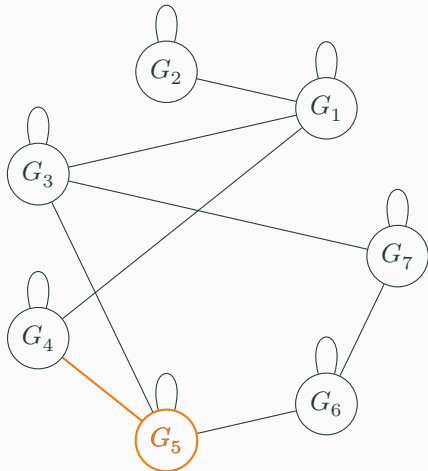


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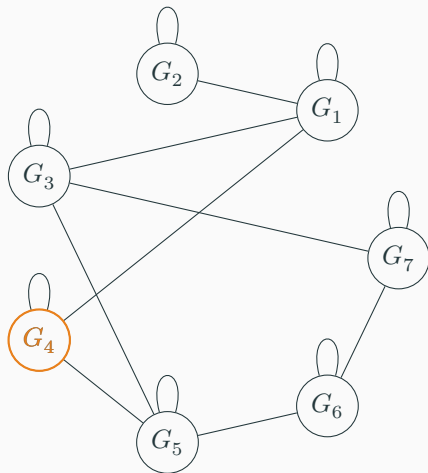


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$t = 3$

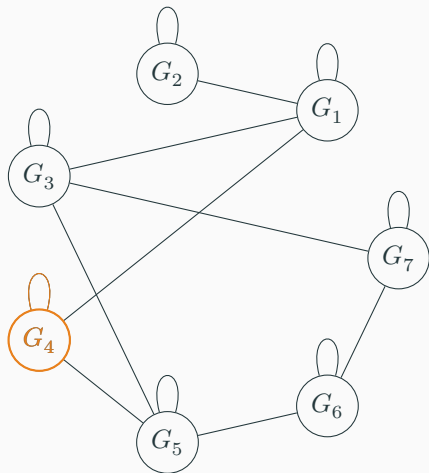


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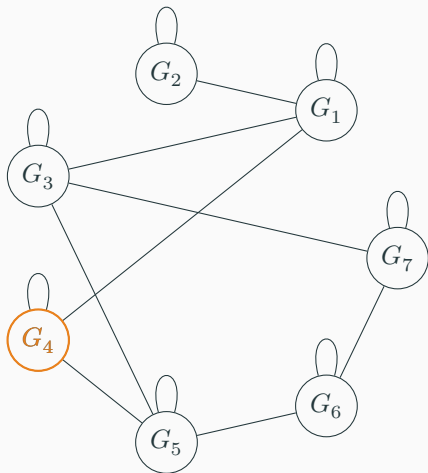


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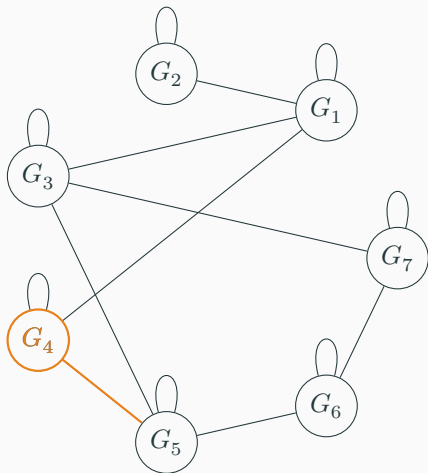


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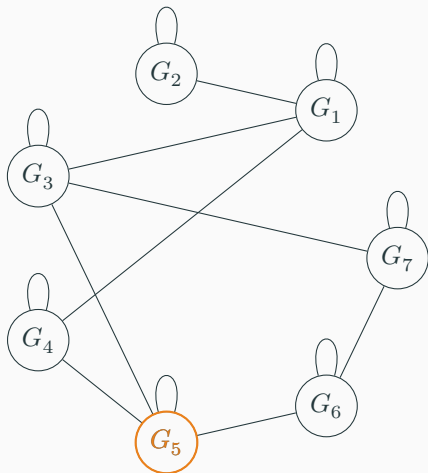


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$t = 5$

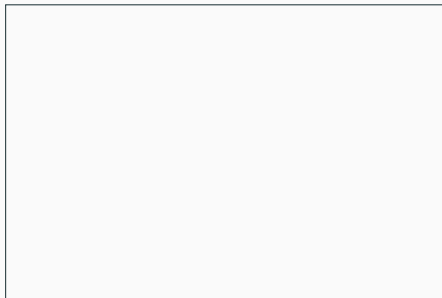
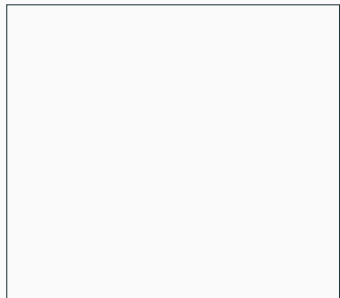


- If the Markov-chain is irreducible, symmetric, and aperiodic then the MC converges to the uniform distribution
- Instead of exact, only require *approximate* sampling: the sampled distribution is ε close to the uniform distribution in variation (ℓ_1 -)distance in $\text{poly}(n) \cdot \log \varepsilon^{-1}$ steps (rapidly mixing)

JERRUM-SINCLAIR CHAIN

State space: $\mathcal{G}(d) \cup \bigcup_{i,j \in V} \mathcal{G}(d - \mathbf{1}_i - \mathbf{1}_j)$.

Transitions: u.a.r. choose $a, b \in V(G)$, then

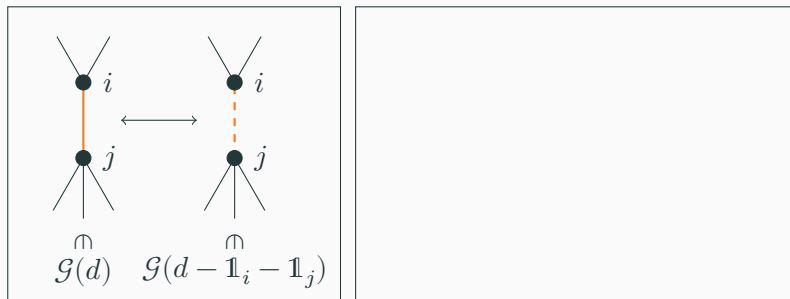


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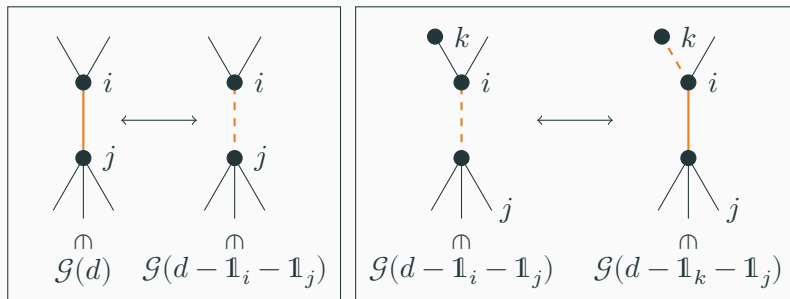


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- if $G \in \mathcal{G}(d)$, delete $ab \in E(G)$ if it exists.
- if $G \in \mathcal{G}(d - \mathbf{1}_i - \mathbf{1}_j)$ and $\deg_G(a) < d(a)$, try to add ab to $E(G)$. If $\deg_{G+ab}(b) > d(b)$, then delete u.a.r. an edge of b .



The state space of JS chain: $\mathcal{G}(d) \cup \bigcup_{i,j \in V} \mathcal{G}(d - \mathbf{1}_i - \mathbf{1}_j)$

To get a sample from $\mathcal{G}(d)$ in reasonable time, we must have (where $n = \dim(d)$)

$$\frac{\sum_{i,j} |\mathcal{G}(d - \mathbf{1}_i - \mathbf{1}_j)|}{|\mathcal{G}(d)|} \leq \text{poly}(n) \quad \forall d \in \mathcal{D}.$$

In this case, we call \mathcal{D} a *P-stable* class of degree sequences.

Theorem (Jerrum and Sinclair 1990)

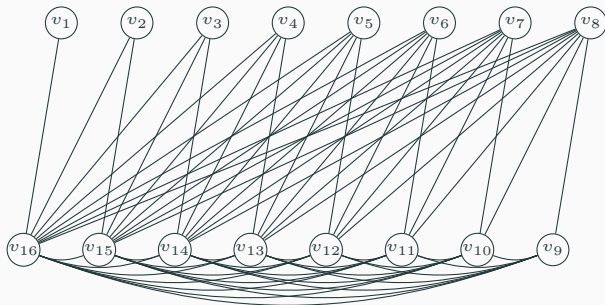
The JS chain is rapidly mixing on degree sequences from a P-stable class.

Theorem (Jerrum and Sinclair 1992)

The class of degree sequences satisfying $(\Delta - \delta + 1)^2 \leq 4\delta(n - \Delta + 1)$ is P -stable.

A (SEEMINGLY PATHOLOGICAL) OBSTACLE

$n = 8$



$$h_0(n) = (1, 2, \dots, n-1, n, n, n+1, \dots, 2n-1)$$

has a unique realization

$\{h_0(n) \mid n \in \mathbb{N}\}$ not P -stable: $|\mathcal{G}(h_0(n) - \mathbf{1}_n - \mathbf{1}_{2n})| \approx \left(\frac{3+\sqrt{5}}{2}\right)^n$

Can be blown up to a non-pathological non- P -stable class.

APPLICABILITY REMARKS

The cardinality of the state space of the JS chain can easily be a factor of n^8 larger than $\mathcal{G}(d)$.

$\mathcal{G}(7, 4, 1, \dots)$

$\mathcal{G}(6, 4, 1, \dots)$

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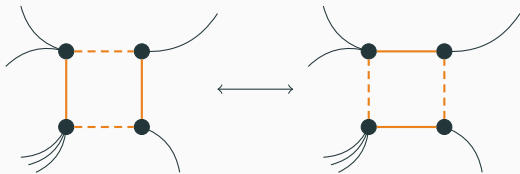
THE SWITCH MARKOV-CHAIN

Proposed by Kannan, Tetali, Vempala (1997)

State space: only the set of realizations $\mathcal{G}(d)$ of a deg. seq. d

Transitions: exchange edges with non-edges along a randomly chosen alternating C_4 (least perturbation)

switch:



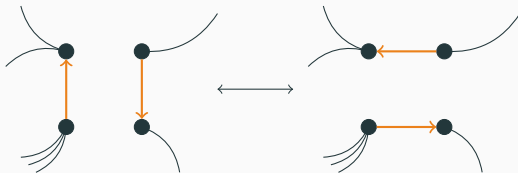
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directed \triangle :



PREVIOUS RESULTS ON THE SWITCH CHAIN

Rapid mixing of the Switch Markov chain shown by

	<i>simple</i>	<i>bipartite</i>	<i>directed</i>
regular	Cooper, Dyer, Greenhill 2007	Erdős et al. 2013	Greenhill 2011
$\Delta \leq c\sqrt{m}$	Greenhill and Sfragara 2018	Erdős, Miklós, M, Soltész 2018	
Interval	—	Erdős, Miklós, M, Soltész 2018 $(\Delta - \delta)^2 \leq \delta(n - \Delta)$	similar
strongly stable	Amanatidis and Kleer 2019		—

Theorem (Greenhill, Erdős, Miklós, M, Soltész, Soukup 2019+)

The switch Markov-chain is rapidly mixing on P -stable degree sequences (unconstrained, bipartite, directed)

- Proof: complex (based on the Jerrum-Sinclair method)
- Every previously known rapidly mixing region is P -stable
- Gao and Wormald (2016) describe several P -stable regions, including power-law distribution-bounded degree sequences for $\gamma > 1 + \sqrt{3}$
- Power-law degree sequences with $\gamma > 2$ are also conjectured to be P -stable

BEYOND P -STABILITY...?

P -STABILITY IS NOT NECESSARY FOR RAPID MIXING

For all $n, k \in \mathbb{Z}^+$, let us define the bipartite degree sequence

$$h_k(n) := \begin{pmatrix} 1 & 2 & 3 & \dots & n-2 & n-1 & n-k \\ n-k & n-1 & n-2 & \dots & 3 & 2 & 1 \end{pmatrix}$$

Theorem (Erdős, Györi, M, Miklós, Soltész 2019+)

For any $k \in \mathbb{Z}^+$, the switch Markov chain is rapidly mixing on

$$\mathcal{H}_k := \left\{ h_k(n) : n \geq k \right\},$$

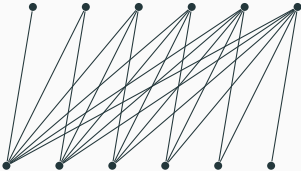
even though the class is not P -stable:

$$\frac{|\mathcal{G}(h_{k+1}(n))|}{|\mathcal{G}(h_k(n))|} = e^{\Omega_k(n)}$$

Remark: the proof works up to $k \leq c\sqrt{\log n}$ for some c .

THE SIMPLE AND DIRECTED ANALOGUES FOLLOW IMMEDIATELY

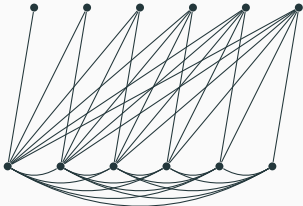
bipartite



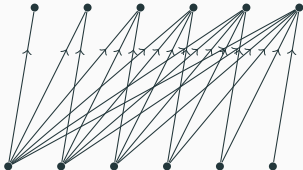
add a clique



orient upwards



simple



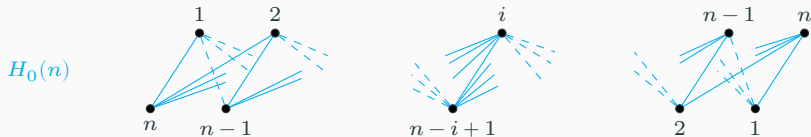
directed

PROOF OF RAPID MIXING FOR $k = 1$; GEOMETRIC REPRESENTATION

$$h_0(n) := \begin{pmatrix} 1 & 2 & 3 & \dots & n-2 & n-1 & n \\ n & n-1 & n-2 & \dots & 3 & 2 & 1 \end{pmatrix}$$

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Suppose $G \in \mathcal{G}(h_1(n))$. What does $H_0(n) \Delta G$ look like?

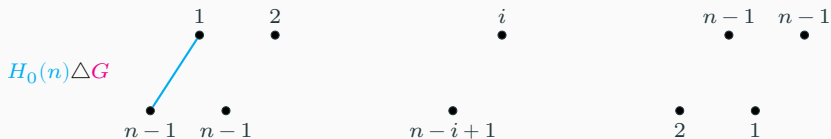


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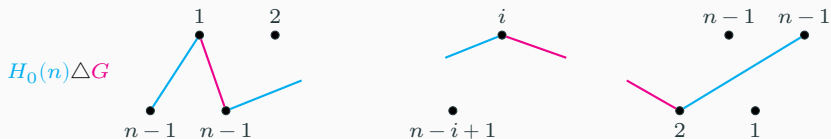


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$H_0(n) \Delta G$ is an x -monotone path!

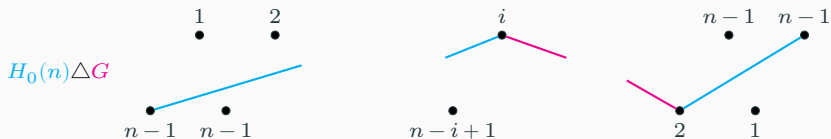
Switch in this representation: moves a vertex of the path or deletes/inserts a pair of adjacent vertices

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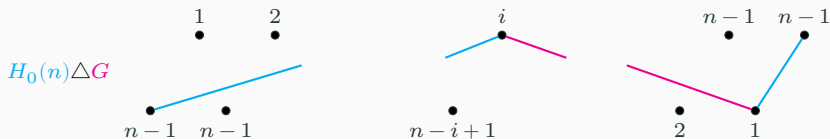
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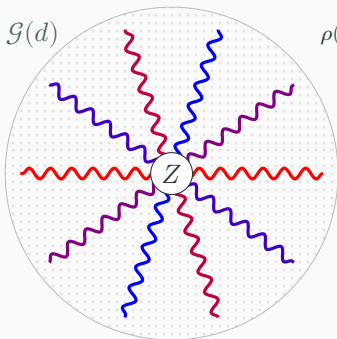
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THE JERRUM-SINCLAIR METHOD

Let Γ contain a switch sequence

$X = Z_0^{X,Y}, Z_1^{X,Y}, Z_2^{X,Y}, \dots, Z_\ell^{X,Y} = Y$ for each pair of realizations
 $X, Y \in \mathcal{G}(d)$.



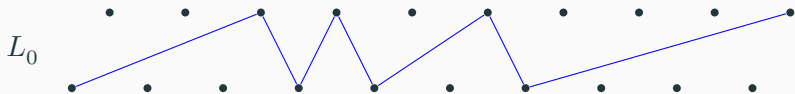
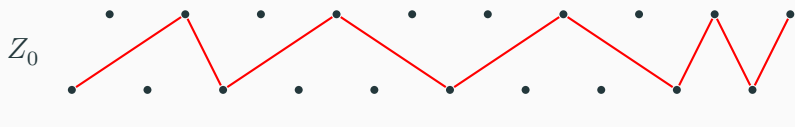
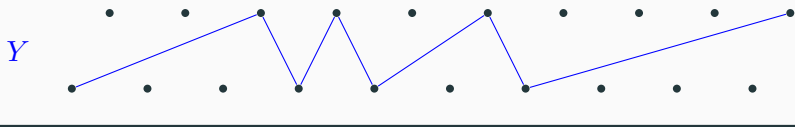
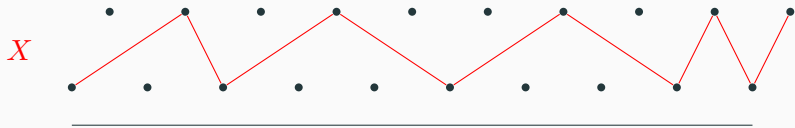
$\rho(\Gamma)$: max number of sequences through a Z

$\ell(\Gamma)$: max length of a sequence in Γ

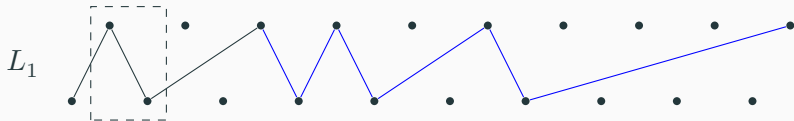
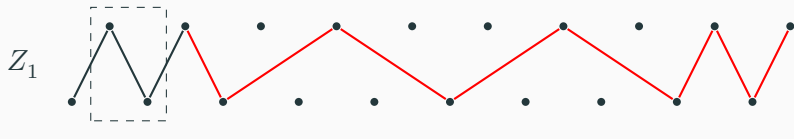
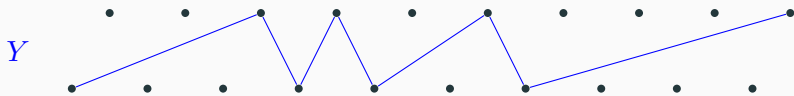
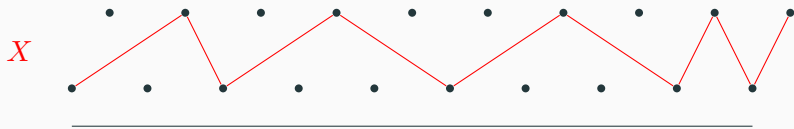
Theorem (follows from Jerrum and Sinclair 1990)

$$\tau_{\text{switch}}(\varepsilon) \leq \text{poly}(n) \cdot \frac{\rho(\Gamma)}{|\mathcal{G}(d)|} \cdot \ell(\Gamma) \cdot (\log(|\mathcal{G}(d)|) + \log(\varepsilon^{-1}))$$

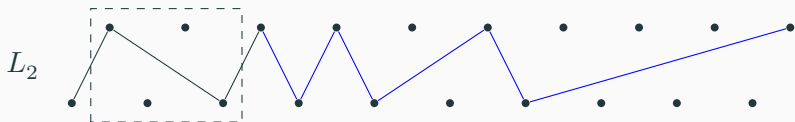
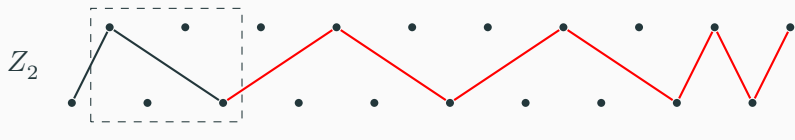
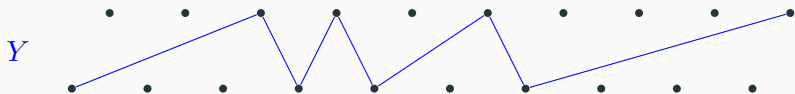
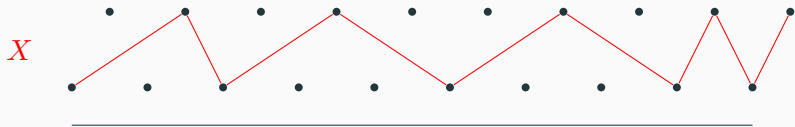
SWITCH SEQUENCE BETWEEN $X, Y \in \mathcal{G}(h_1(n))$



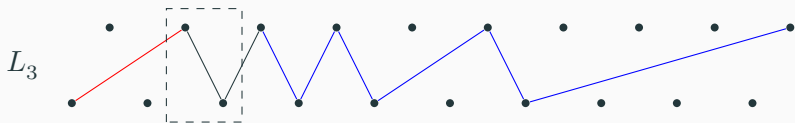
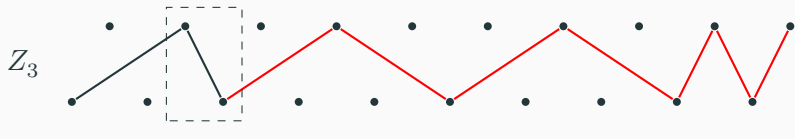
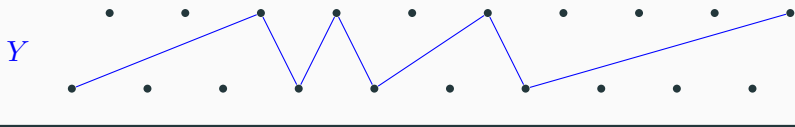
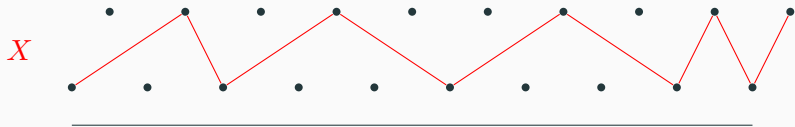
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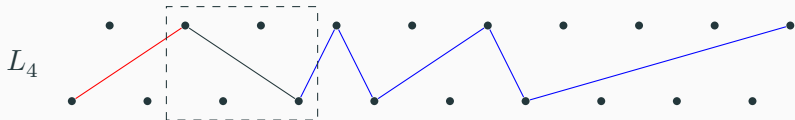
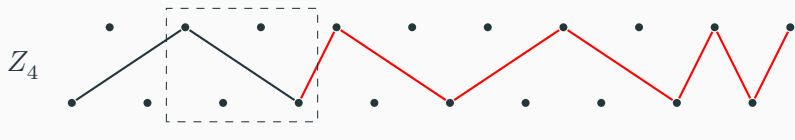
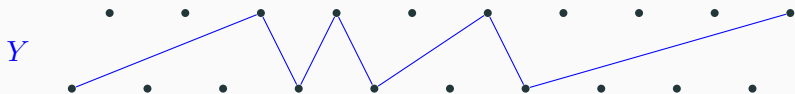
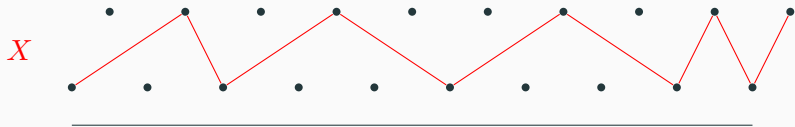
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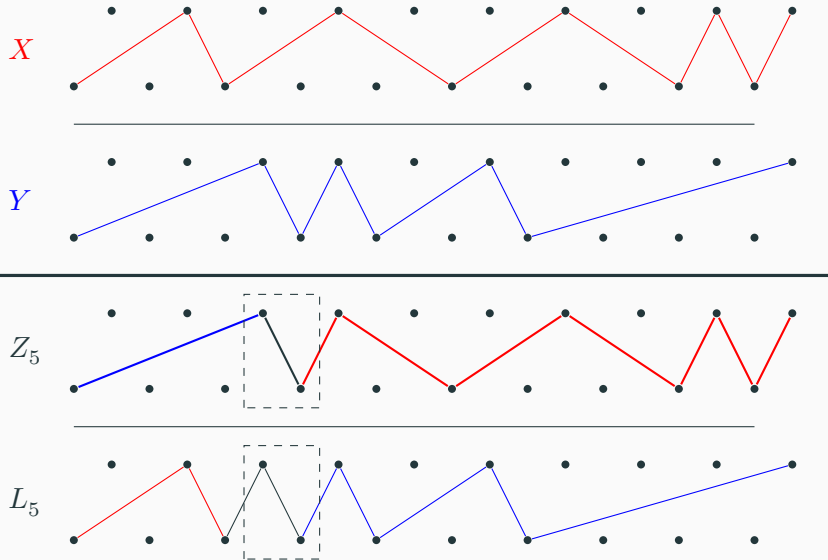
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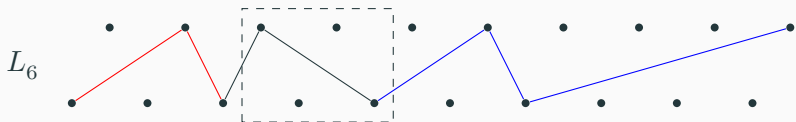
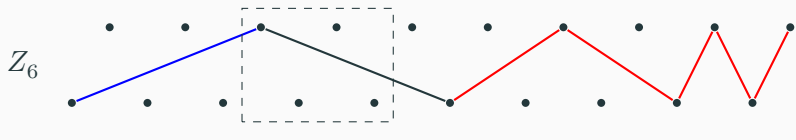
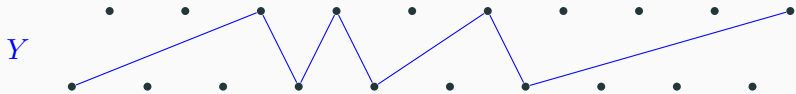
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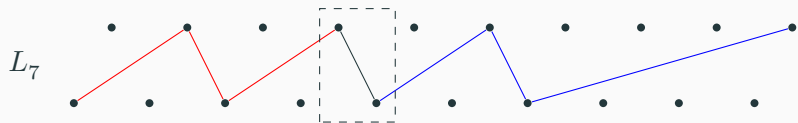
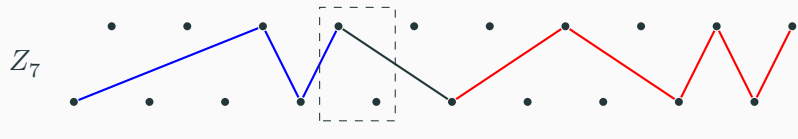
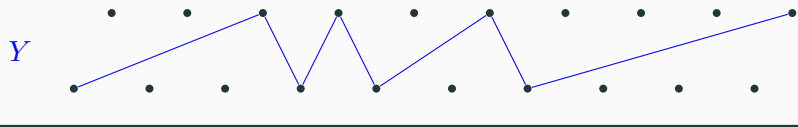
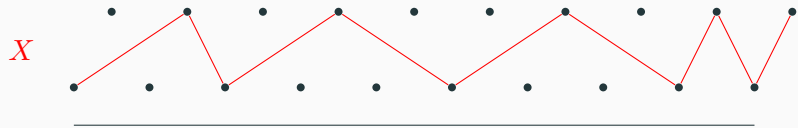
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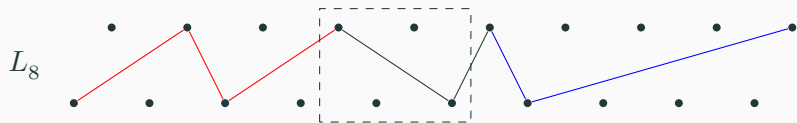
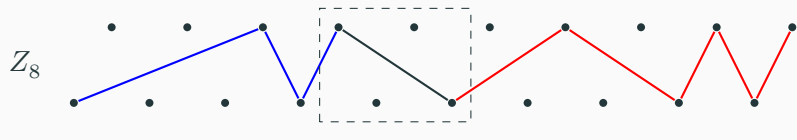
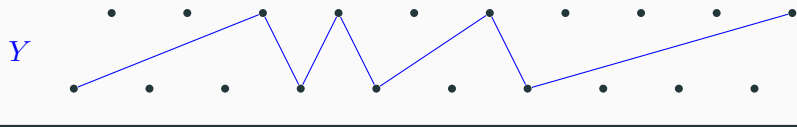
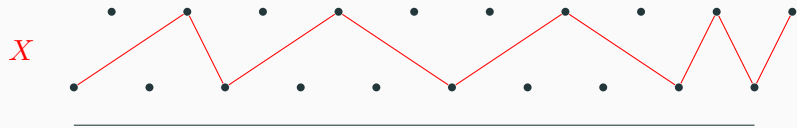
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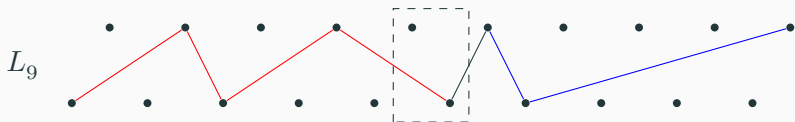
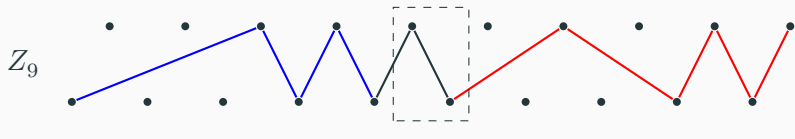
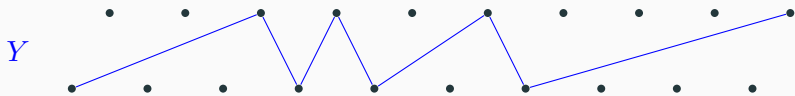
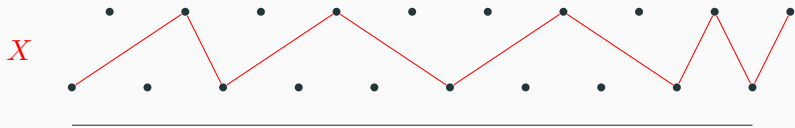
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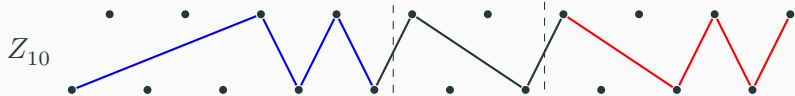
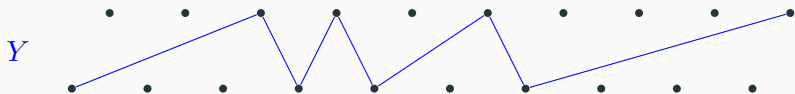
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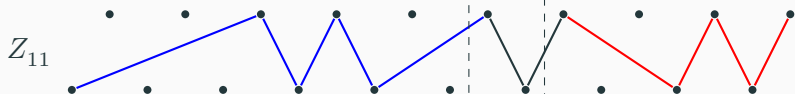
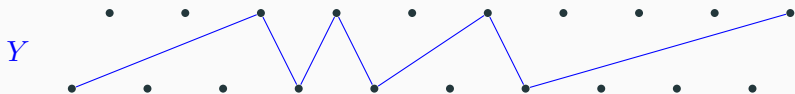
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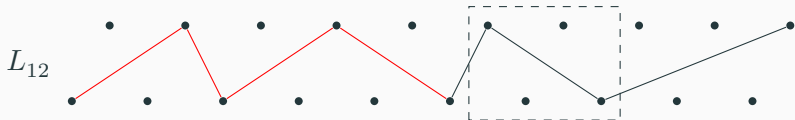
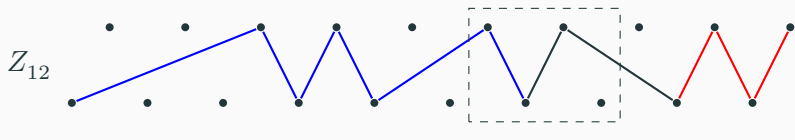
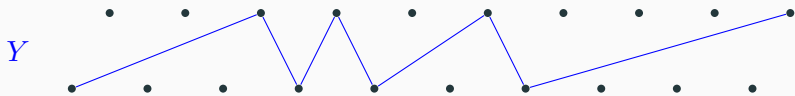
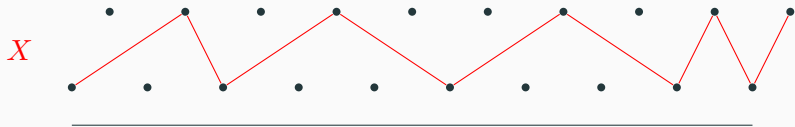
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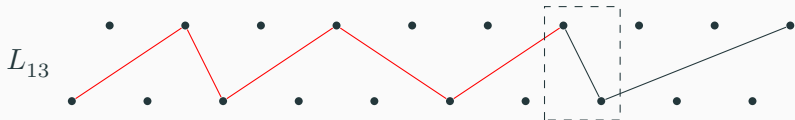
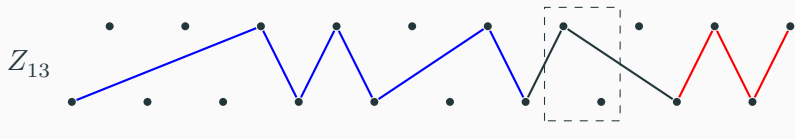
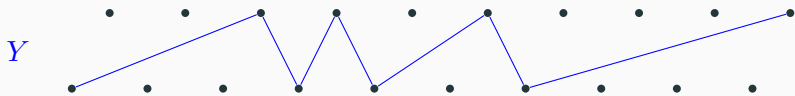
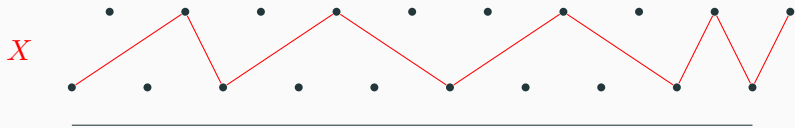
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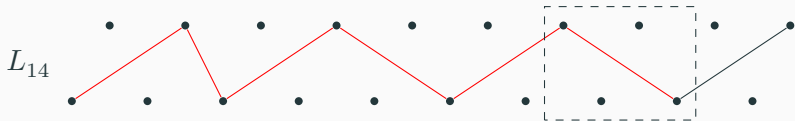
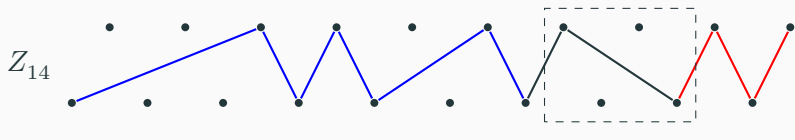
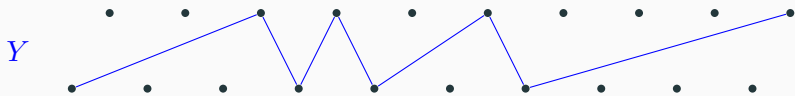
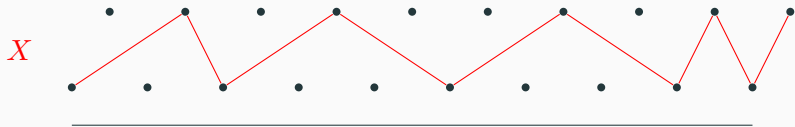
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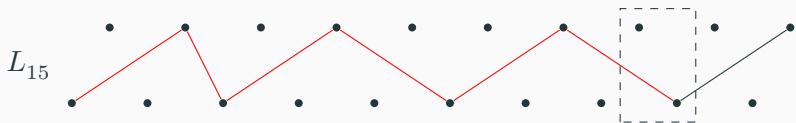
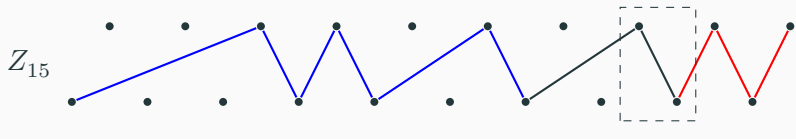
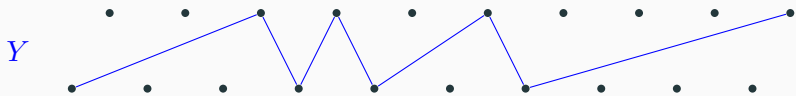
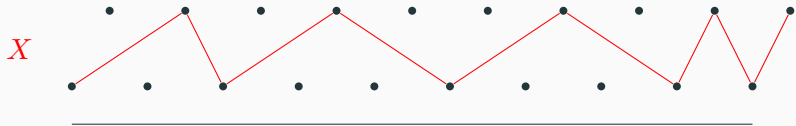
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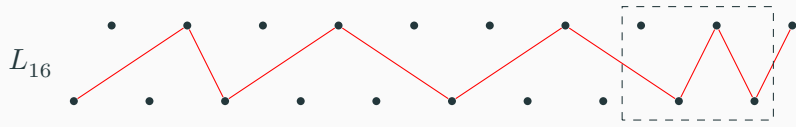
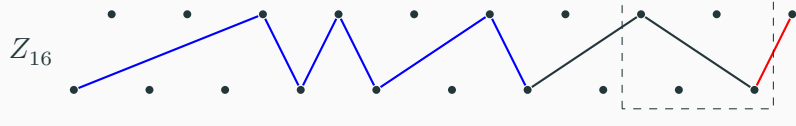
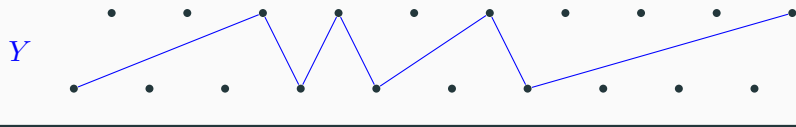
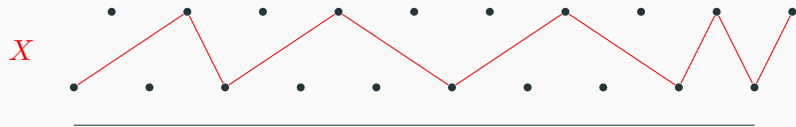
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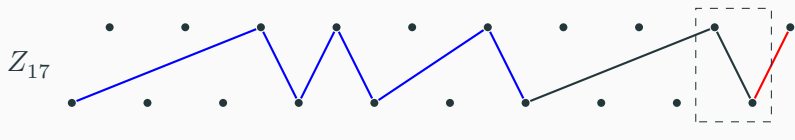
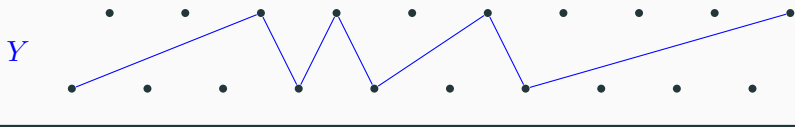
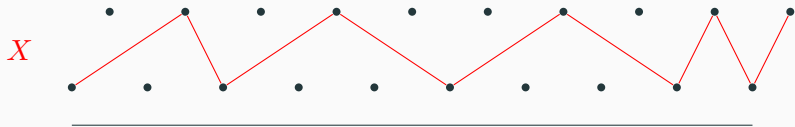
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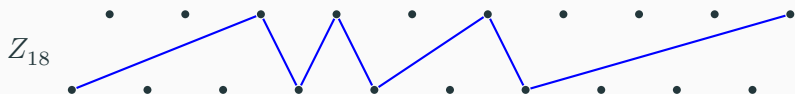
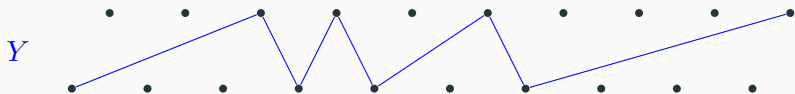
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Clearly, $\ell(\Gamma) = \mathcal{O}(n)$.

From Z_i and L_i and a $\mathcal{O}(\log n)$ bits we can recover X and Y !

\implies For a fix $Z \in \mathcal{G}(d)$, the number of switch sequences of Γ passing through Z is at most the number of possible L_i times $\text{poly}(n)$!

$\implies \rho(\Gamma) = \text{poly}(n) \cdot |\mathcal{G}(h_1(n))| \stackrel{\text{Jerrum-Sinclair}}{\implies} \text{Switch MC is rapidly mixing on } \mathcal{G}(h_1(n))!$

THANK YOU FOR LISTENING TO MY PRESENTATION!

HOME PAGE: <https://trm.hu>

FULL PAPERS

UNIFIED APPROACH:

<https://arxiv.org/abs/1903.06600>

BEYOND P -STABILITY:

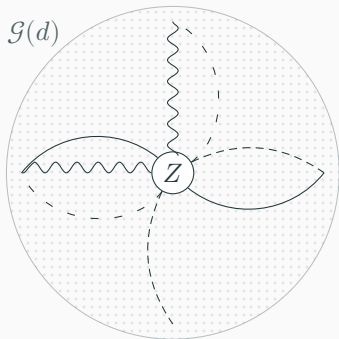
<https://arxiv.org/abs/1909.02308>

PROOF OUTLINE OF RAPID MIXING ON P -STABLE DEGREE SEQUENCES

- Jerrum-Sinclair for multicommodity-flows
- Determining a flow between any two $X, Y \in \mathcal{G}(d)$
 - Decomposing $E(X) \Delta E(Y)$ into red/blue alternating circuits
 - Decomposing alternating circuits into elementary circuits
 - Processing elementary circuits
- Estimating the load of the flow
 - Coding
 - Reconstruction

THE JERRUM-SINCLAIR METHOD

Let f be a multicommodity-flow that sends 1 quantity of commodity between each two realizations in the switch graph on $\mathcal{G}(d)$.

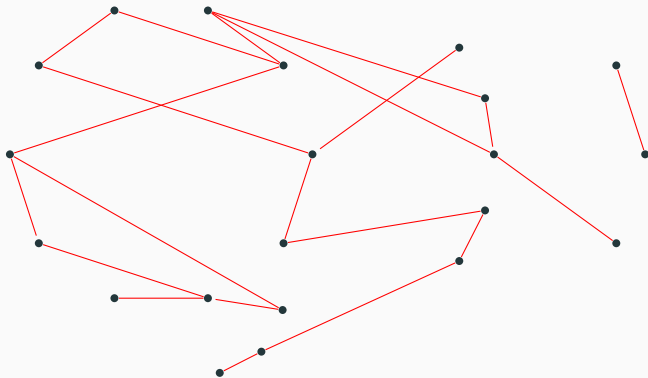


Theorem (follows from Jerrum and Sinclair 1990)

$$\tau_{\text{switch}}(\varepsilon) \leq n^4 \cdot \max_{G \in \mathcal{G}(d)} \sum_{G \in \gamma} \frac{f(\gamma)|\gamma|}{|\mathcal{G}(d)|} \cdot (\log(|\mathcal{G}(d)|) + \log(\varepsilon^{-1}))$$

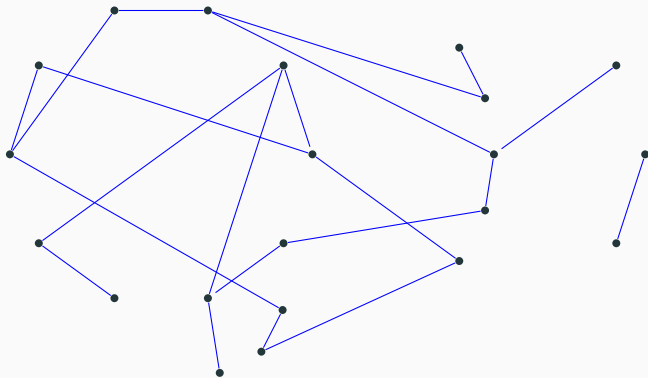
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- Let s be a complete matching between the red and blue edges at each vertex
- $s \xleftrightarrow{\text{bij.}}$ an alternating-circuit decomposition
- Thus $E(X)\Delta E(Y) = W_1 \uplus \dots \uplus W_k$, where each W_i is an alternating-circuit
- Traverse each circuit W_i (from v_1) and cut off an alternating-circuit whenever a node is visited twice with the same parity (elementary alternating circuit)
- “Process” each elementary alternating circuit when found, while maintaining the matching s



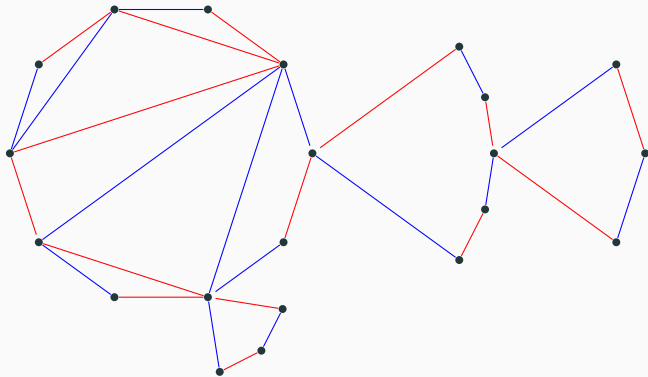
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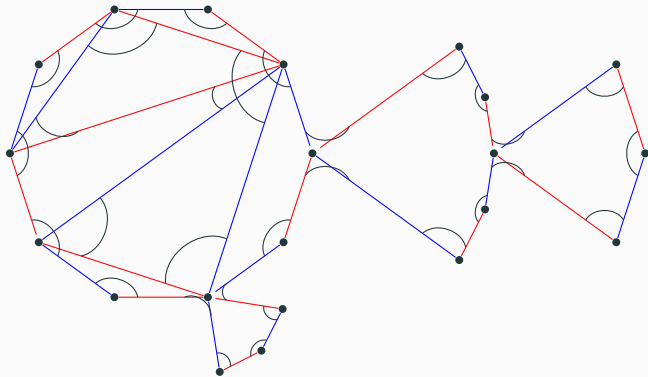
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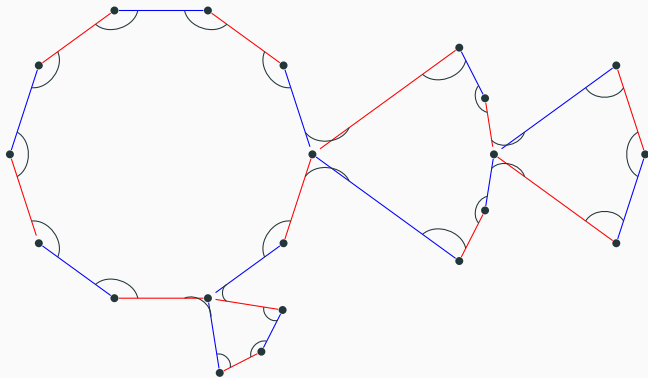
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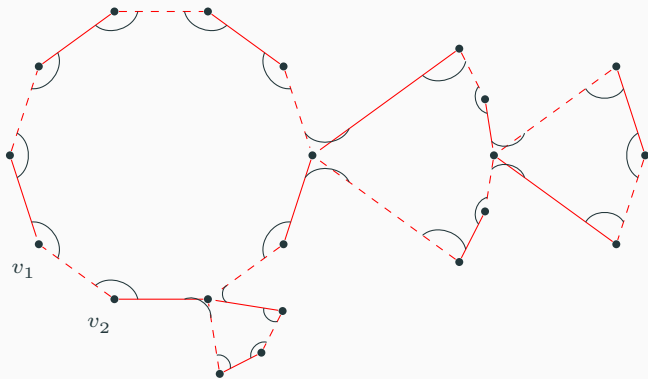
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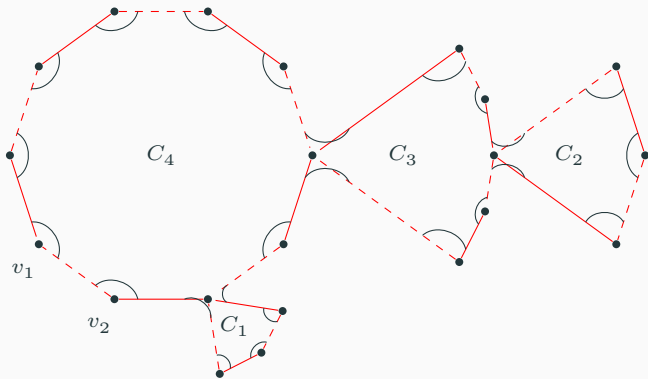
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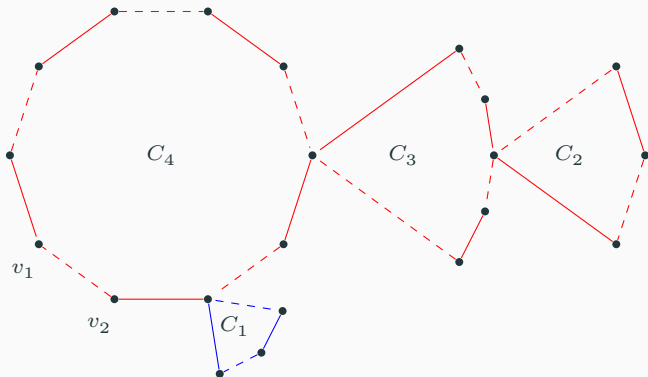
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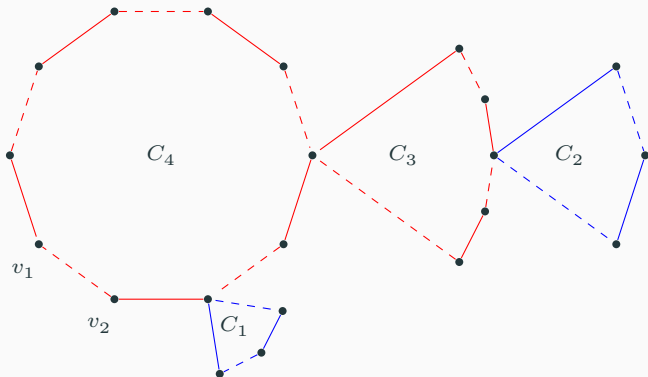
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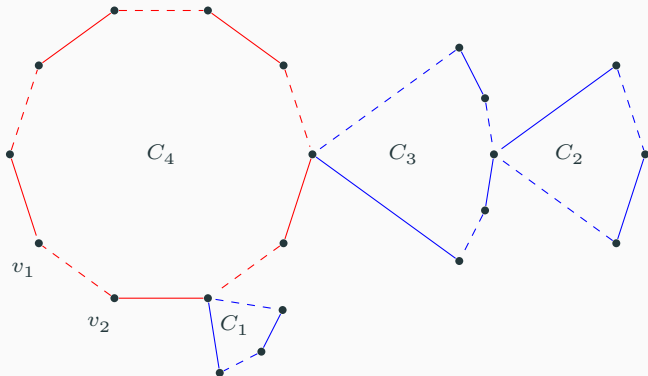
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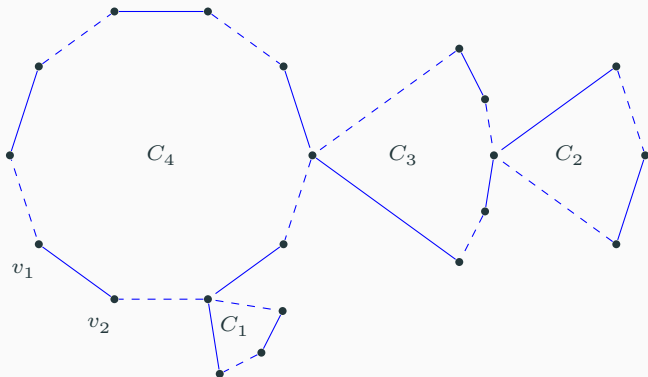
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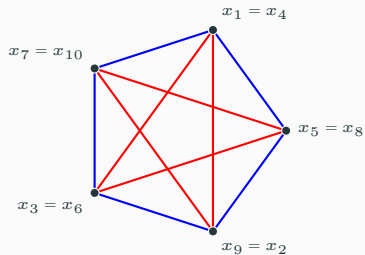
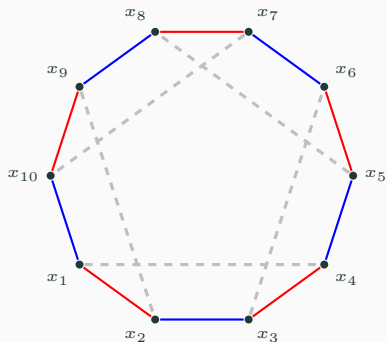
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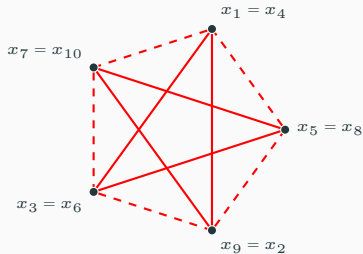
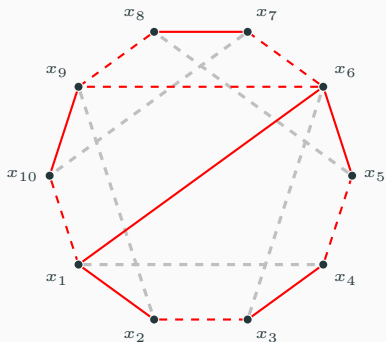
SWEEPING ELEMENTARY ALTERNATING CIRCUITS

Elementary alternating circuit: each vertex on the trail is either visited once, or twice but with different parity



SWEEPING ELEMENTARY ALTERNATING CIRCUITS

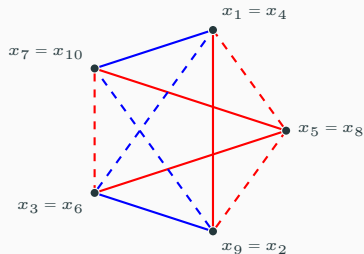
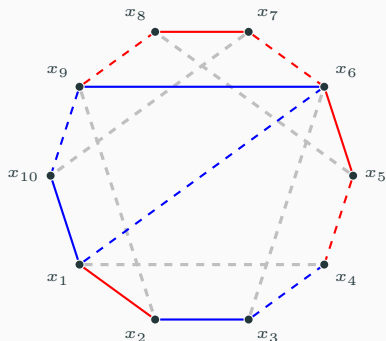
Elementary alternating circuit: each vertex on the trail is either visited once, or twice but with different parity



Sweep: the edges are exchanged with non-edges and vica versa from x_1 in counter-clockwise order

SWEEPING ELEMENTARY ALTERNATING CIRCUITS

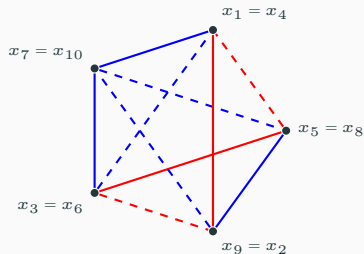
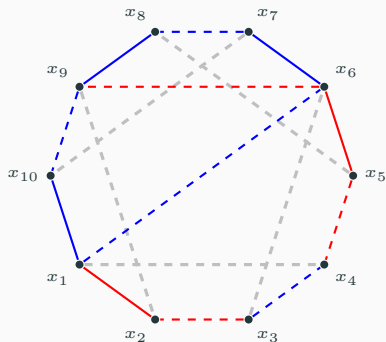
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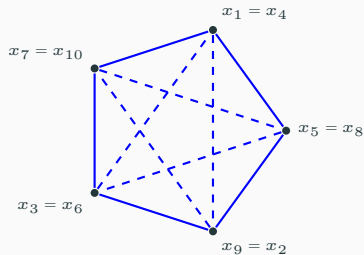
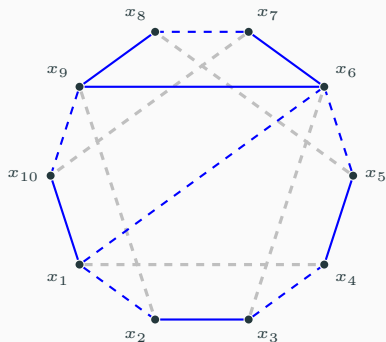
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- Encoding of a realization Z along the $X \xrightarrow{s} Y$ path:
$$L = A_X + A_Y - A_Z$$
- From Z and L , we can reconstruct $X\Delta Y$, but cannot immediately tell which edges belong to X and Y
- Via a matching of edges and non-edges of Z supported by $X\Delta Y$ and a tricky theorem, we can reconstruct which edges belong to X and Y