

Rapid mixing from spectral independence beyond the Boolean domain

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SODA 2021, Online

Glauber dynamics

Sampling from joint distribution

| | |
|---------------------------|--|
| Set of variables | V |
| Finite domain | $[q] = \{1, 2, \dots, q\}$ for $q \geq 2$ |
| Joint distribution | μ over $\Omega = \text{supp}(\mu) \subseteq [q]^V$ |
| Problem | draw random samples from μ |

Fundamental MCMC: Glauber dynamics

Start from an arbitrary feasible configuration $X \in \Omega$;

For each t **from** 1 **to** T **do**

- pick a variable $v \in V$ uniformly at random;
- resample $X_v \sim \mu_v(\cdot | X_{V \setminus \{v\}})$;

Return X ;

Example: proper q -coloring

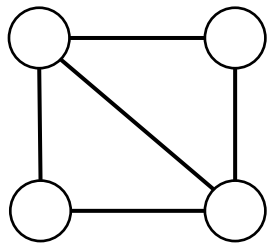
Uniform proper q -coloring

Undirected graph $G = (V, E)$

Finite set of colors $[q] = \{1, 2, \dots, q\}$

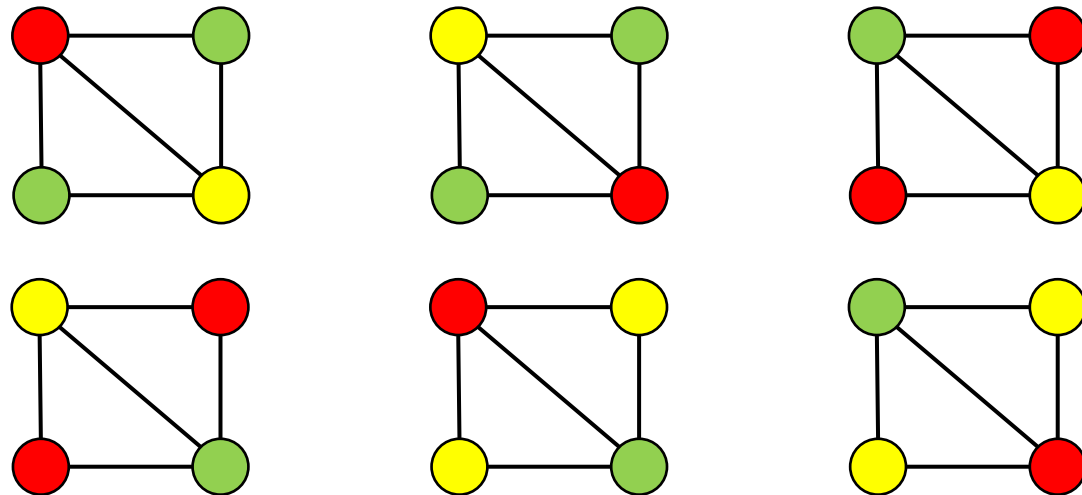
Gibbs distribution μ **uniform distribution** over Ω

$$\Omega = \{X \in [q]^V \mid X \text{ is a proper coloring}\}$$



graph $G = (V, E)$

colors $[q] = \{ \text{red} \quad \text{green} \quad \text{yellow} \}$



Ω : set of all proper colors

Example: proper q -coloring

Uniform proper q -coloring

Undirected graph $G = (V, E)$

Finite set of colors $[q] = \{1, 2, \dots, q\}$

Gibbs distribution μ **uniform distribution** over $\Omega = \{X \in [q]^V \mid X \text{ is a proper coloring}\}$

Problem sample proper coloring u.a.r.

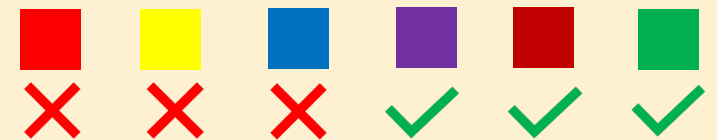
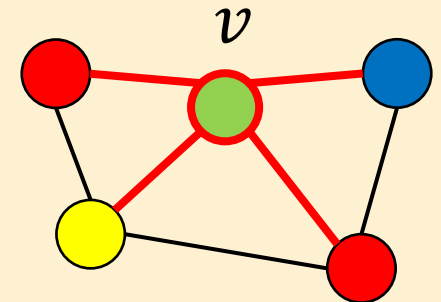
Glauber dynamics for proper q -coloring

Start from an arbitrary proper coloring $X \in \Omega$;

For each t from 1 to T **do**

- pick a vertex $v \in V$ uniformly at random;
- resample X_v from $[q] \setminus \{X_u \mid u \in \Gamma(v)\}$ uniformly at random;

Return X ;



Convergence

Glauber dynamics: **Markov chain** over Ω

Transition Matrix $P \in \mathbb{R}^{\Omega \times \Omega}$

Glauber dynamics is **reversible**

detailed balance equation with respect to μ

$$\forall X, Y \in \Omega, \mu(X)P(X, Y) = \mu(Y)P(Y, X)$$

 **Stationary distribution** $\mu P = \mu$

move among any states with positive probability

Proposition (convergence)

If Glauber dynamics is **connected**, it converges to **unique** stationary distribution μ .

If $q \geq \Delta + 2$, Glauber dynamics converges to uniform distribution over q -colorings.

Mixing time

How fast does the Glauber dynamics converge to stationary distribution μ ?

Glauber dynamics X_0, X_1, X_2, \dots where each $X_i \in \Omega \subseteq [q]^V$

Mixing time

$$T_{\text{mix}} = \max_{X_0 \in \Omega} \min \left\{ t \mid d_{TV}(X_t, \mu) \leq \frac{1}{4e} \right\},$$

$d_{TV}(X_t, \mu)$: the **total variation distance** between X_t and μ .

The Glauber dynamics is **rapid mixing** if

$$T_{\text{mix}} = \text{Poly}(n)$$

$$n = |V| = \#\{\text{variables}\}$$

- ✓ Sample from an **exponential space** $|\Omega| = \exp(O(n))$ within **polynomial steps** $T_{\text{mix}} = \text{poly}(n)$.

Open problems

Under what **condition** of the distribution μ
the Glauber dynamics for μ rapid mixing ?

Under what **relation** between q and max degree Δ
the Glauber dynamics for coloring rapid mixing ?

Previous works

Glauber dynamics for graph coloring

General graphs [Jer95, Vig00, SS97, CDMPP19]

current best result $q \geq \left(\frac{11}{6} - \epsilon_0\right) \Delta$ [CDMPP19]

Special graphs [DF01, Hay03, HV03, GMP05, HV06, Mol04, Hay13, DFHV13]

High-dimensional expansion (HDX)

Strongly log-concave distribution [ALOV19, CGM19]

Spectral independence with **Boolean domain $\{0, 1\}^V$** [ALO20]

**Spectral
Independence**



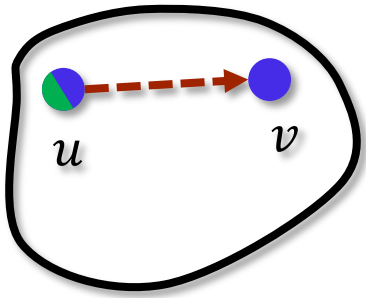
Mixing up to uniqueness for **hardcore model** [ALO20]
anti-ferro 2-spin systems [CLV20]

Our results

- A **spectral independence** condition for **general distribution**.
- **Rapid mixing** of Glauber dynamics from spectral independence.
 - combinatorial proof: **coupling**;
 - algebraic proof for Boolean variables [ALO20].
- Application: a new rapid mixing regime for **graph coloring**.
 - relate spectral independence with **correlation decay**;
 - a refined **recursive coupling** [GMP05] argument.

Our results

Result (I). A **spectral independence** condition **beyond the Boolean domain**.

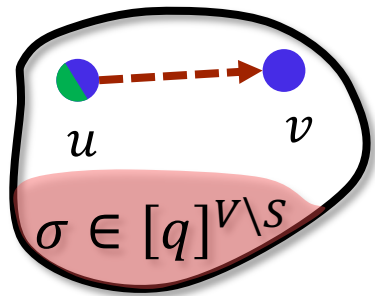


maximum influence on v caused
by a **disagreement** on u

μ : a distribution over $\Omega \subseteq [q]^V$

$|V| \times |V|$ **influence matrix** $\Psi \in \mathbb{R}^{V \times V}$ with $\Psi(u, u) = 0$ and

$$\Psi(u, v) = \max_{i, j \in [q]} d_{TV}(\mu_v(\cdot | u \leftarrow i), \mu_v(\cdot | u \leftarrow j))$$



influence matrix
for **conditional distribution**

For any subset $S \subseteq V$, any feasible $\sigma \in [q]^{V \setminus S}$

μ_S^σ distribution on S conditional on σ

influence matrix $\Psi_S^\sigma \in \mathbb{R}^{S \times S}$ for **conditional distribution**

$$\Psi_S^\sigma(u, v) = \max_{i, j \in [q]} d_{TV}(\mu_v^\sigma(\cdot | u \leftarrow i), \mu_v^\sigma(\cdot | u \leftarrow j))$$

Our results

Result (I). A **spectral independence** condition **beyond the Boolean domain**

Spectral independence [This work]

There is a constant $C > 0$ such that

for **all** conditional distributions μ_S^σ ,

spectral radius of influence matrices $\rho(\Psi_S^\sigma) \leq C$.

Spectral independence for Boolean variables [Anari, Liu, Oveis Gharan 20]

Distribution over **Boolean domain** $\{0,1\}^V$

signed influence matrix: $I_S^\sigma(u, v) = \mu_v^\sigma(1 \mid u \leftarrow 1) - \mu_v^\sigma(1 \mid u \leftarrow 0)$.

Relation: $\Psi_S^\sigma(u, v) = |I_S^\sigma(u, v)|$.

Spectral independence: for all influence matrices, max eigenvalue $\lambda_{\max}(I_S^\sigma) \leq C$.

Our results

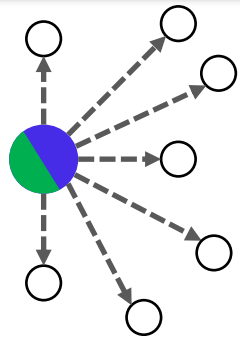
Result (II). **Rapid mixing** of Glauber dynamics from **spectral independence**

Theorem [This work]

μ is spectrally independent with constant \mathcal{C}

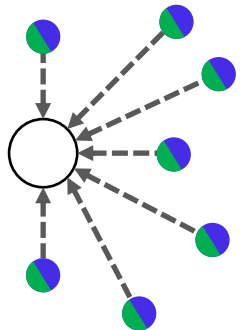
$$\longrightarrow T_{\text{mix}} = O\left(n^{1+2\mathcal{C}} \log\left(\frac{1}{\mu_{\min}}\right)\right),$$

where $\mu_{\min} = \min_{X \in \Omega} \mu(X)$.



Bounded one-to-all influence

$$\text{spectral radius} \leq \sum_{v \in S} \Psi_S^\sigma(u, v) \leq \mathcal{C}$$



Bounded all-to-one influence

$$\text{spectral radius} \leq \sum_{u \in S} \Psi_S^\sigma(u, v) \leq \mathcal{C}$$

Spectral Independence

Rapid Mixing

Our results

Result (III). Rapid mixing for q -coloring on triangle-free graph with $q > 1.763\Delta$

Theorem [This work]

Triangle free graph and $q \geq (\alpha + \delta)\Delta$ where $\alpha \approx 1.763$ s.t. $\alpha = \exp(1/\alpha)$

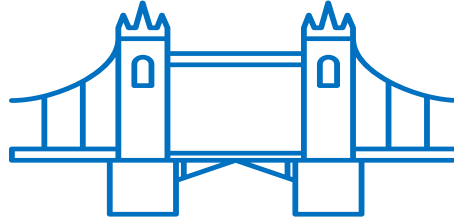
$\rightarrow T_{\text{mix}} \leq n^{2+O(1/\delta)} \log q.$

 constant

| Work | Regime | Girth | Addition condition | Mixing time |
|-----------|----------------------------------|----------|----------------------------------|----------------------------|
| [GMP05] | $q > \alpha\Delta$ | ≥ 4 | $\Delta = O(1)$ + amenable graph | $O(n^2)$ |
| [HV06] | $q \geq (\alpha + \delta)\Delta$ | ≥ 4 | $\Delta = \Omega(\log n)$ | $O(n \log n)$ |
| [DFHV13] | $q \geq (\alpha + \delta)\Delta$ | ≥ 5 | $\Delta \geq \Delta_0(\delta)$ | $O(n \log n)$ |
| This work | $q \geq (\alpha + \delta)\Delta$ | ≥ 4 | -- | $n^{2+O(1/\delta)} \log q$ |

Proof outline

Spectral
Independence



Bridge: HDX

Rapid mixing of
Glauber dynamics

Boolean domain [ALO20]
General domain [this work]

Rapid mixing of
local walks

local-to-global [AL20]

Rapid mixing of
global walk



Graph
Coloring

Decay analysis [this work]

Based on recursive coupling
[GMP05]

Spectral
Independence

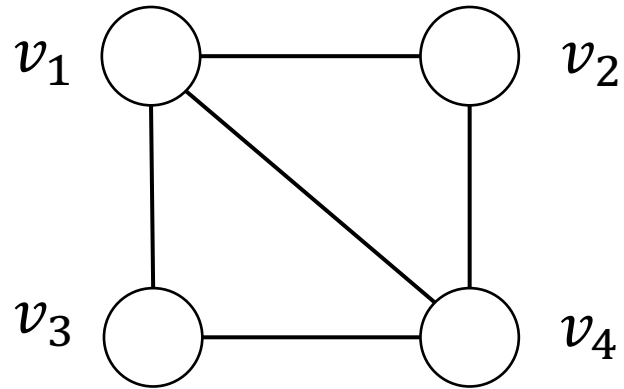
Rapid mixing of
Glauber dynamics

Lazy local random walk

State space $U = \{(v, i) \mid v \in V, i \in [q]\}$

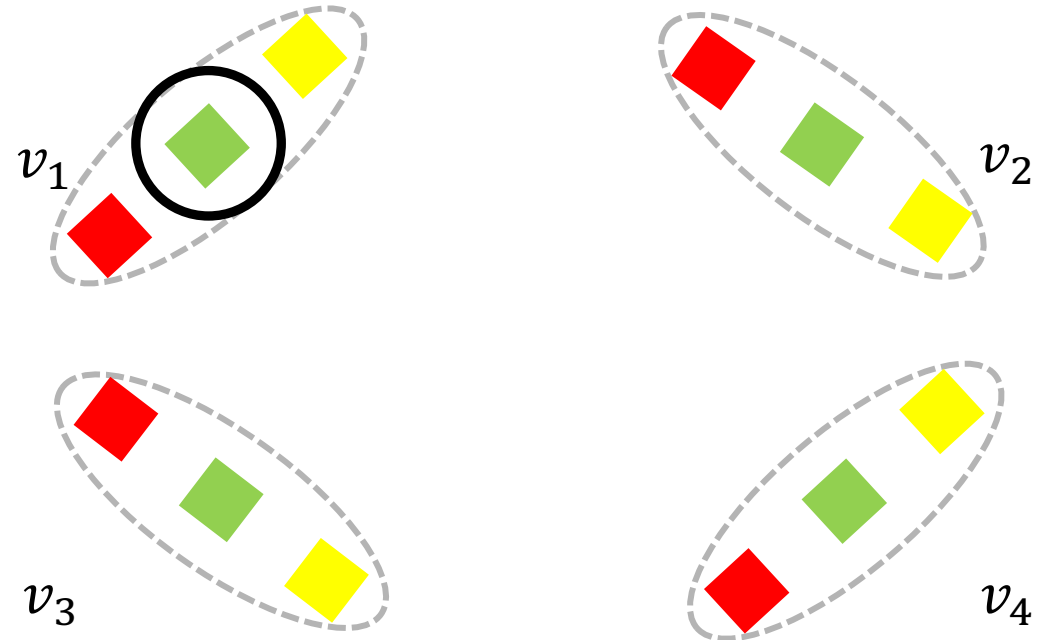
Current state $(v, i) \in U$. Transition $(v, i) \rightarrow (u, j)$

- pick a vertex $u \in V$ uniformly at random;
- sample $j \sim \mu_u(\cdot \mid v \leftarrow i)$.



graph $G = (V, E)$

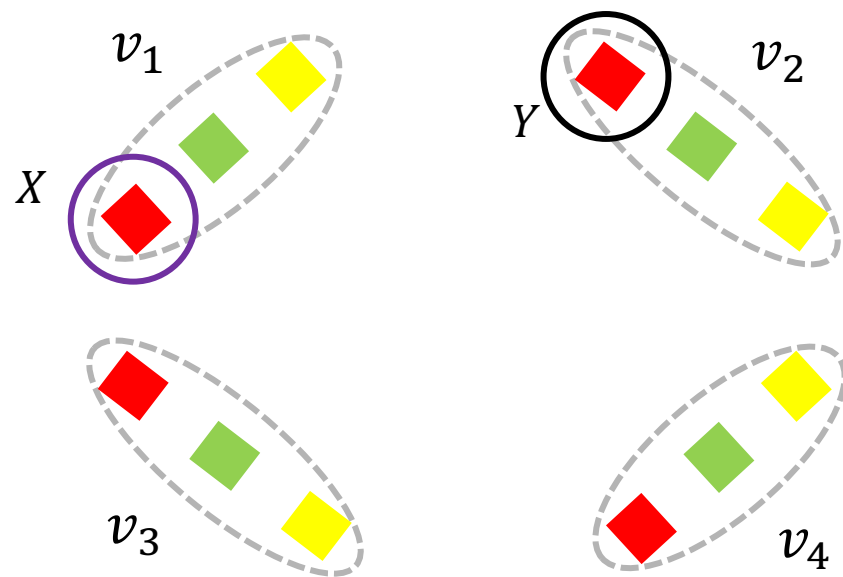
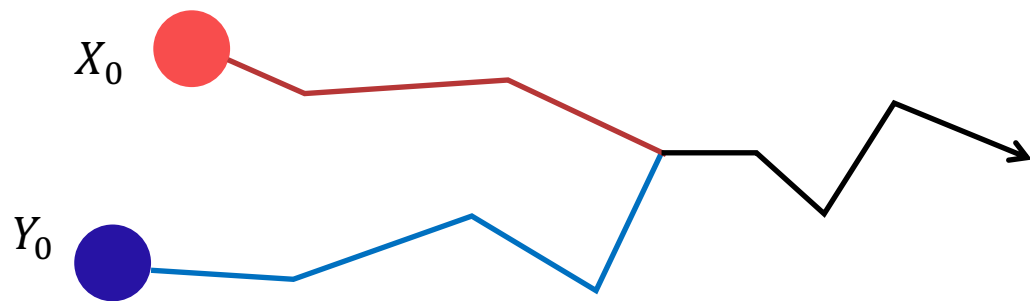
colors $[q] = \{ \color{red}\blacksquare \quad \color{green}\blacksquare \quad \color{yellow}\blacksquare \}$



Our technique: coupling

Coupling $(X_t, Y_t)_{t \geq 0}$ of local walk

- start from two states $X_0, Y_0 \in U$
- two chains $(X_t)_{t \geq 0}$ and $(Y_t)_{t \geq 0}$ follow local walk



Coupling

Current state $X_t = (u, i)$ and $Y_t = (v, j)$

Next state $X_{t+1} = (u', i')$ and $Y_{t+1} = (v', j')$

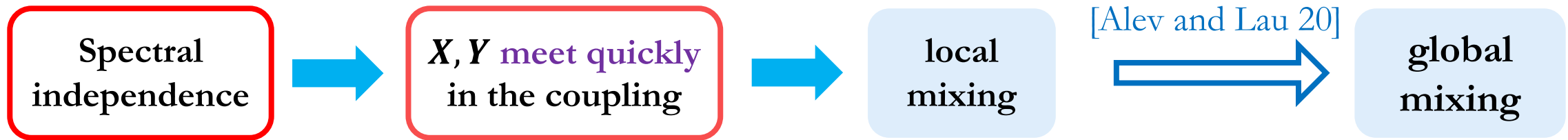
- Pick the **same** $u' = v' \in V$ uniformly at random;
- Sample (i', j') from the **optimal coupling** between $\mu_{u'}(\cdot | u \leftarrow i)$ and $\mu_{v'}(\cdot | v \leftarrow j)$.

Observation: for any $t \geq 1$, X_t and Y_t must be on the **same vertex**.

$$X_t = (v, i) \text{ and } Y_t = (v, j) \text{ (same vertex, different color)}$$

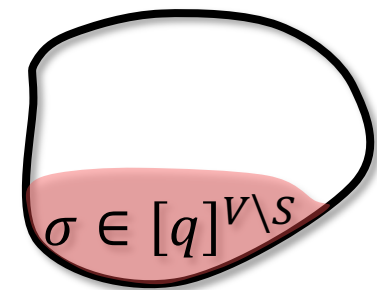
- Pick the same vertex $u \in V$ uniformly at random.
- Couple the colors on u **optimally**, the coupling fails with probability

$$d_{TV}(\mu_u(\cdot | v \leftarrow i), \mu_u(\cdot | v \leftarrow j)) \leq \Psi(v, u). \text{ (Influence } v \rightarrow u)$$



Remark (conditional distributions)

- [AL20] requires local mixing on all **conditional distributions**.
- Our coupling also works for all **conditional distributions**.



Spectral independence for coloring

List coloring instance

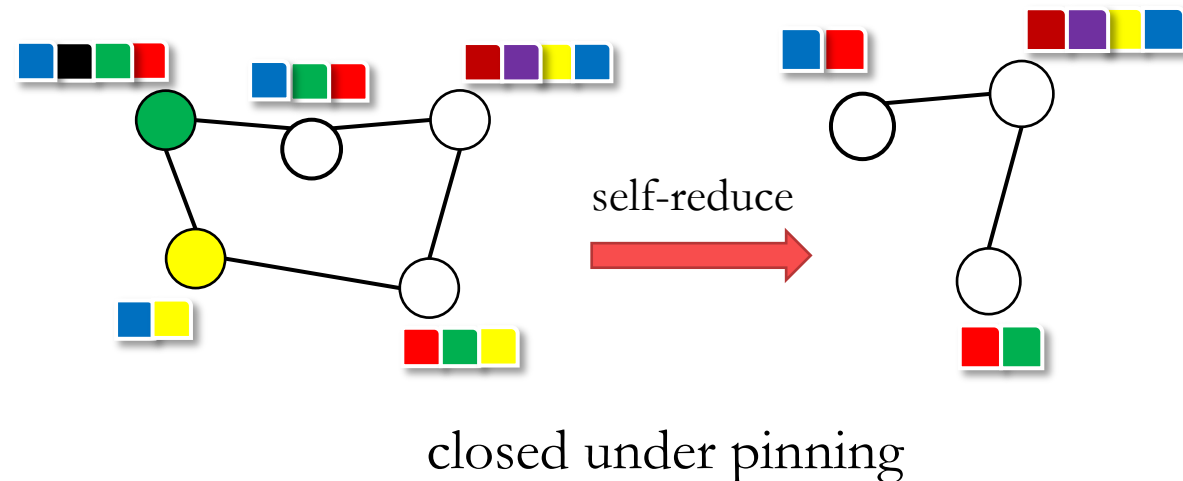
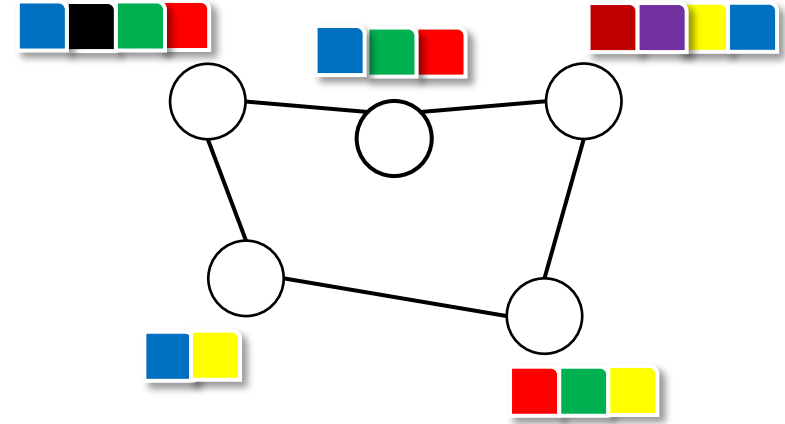
- graph $G = (V, E)$ with max degree Δ ;
- each vertex $v \in V$ has a color list L_v .

Proper list coloring X

- $X_v \in L_v$ for all $v \in V$;
- $X_u \neq X_v$ for all $\{u, v\} \in E$.

Gibbs distribution μ

- uniform distribution over all proper list colorings.



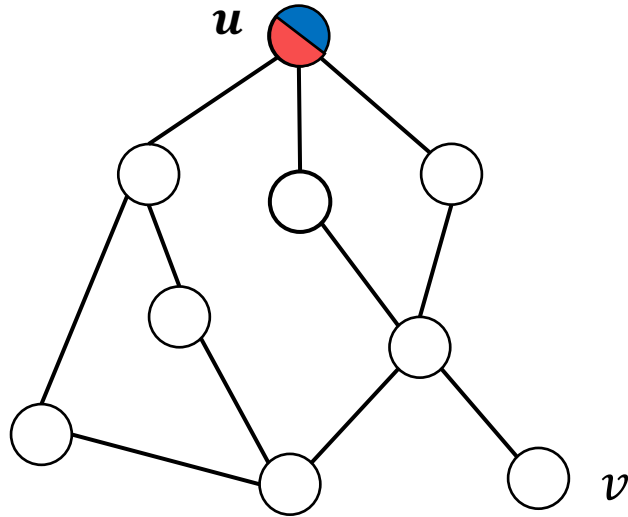
Theorem [this work]

In **triangle-free graph**, if for all $v \in V$,
 $|L(v)| \geq (\alpha + \delta)\Delta \approx (1.763 + \delta)\Delta$,
 then under any pinning,

$$\text{one-to-all influence} = o\left(\frac{1}{\delta}\right),$$

→ μ is **spectrally independent** with $C = o\left(\frac{1}{\delta}\right)$.

Recursive coupling



Influence from u to v

$$\text{Inf}(u \rightarrow v) = \max_{i, j \in L(u)} d_{TV}(\mu_v(\cdot | u \leftarrow i), \mu_v(\cdot | u \leftarrow j))$$

One-to-all influence

$$\sum_{v \in V \setminus \{u\}} \text{Inf}(u \rightarrow v)$$

Proof sketch

by **recursive coupling** [Goldberg, Martin, Paterson 05]

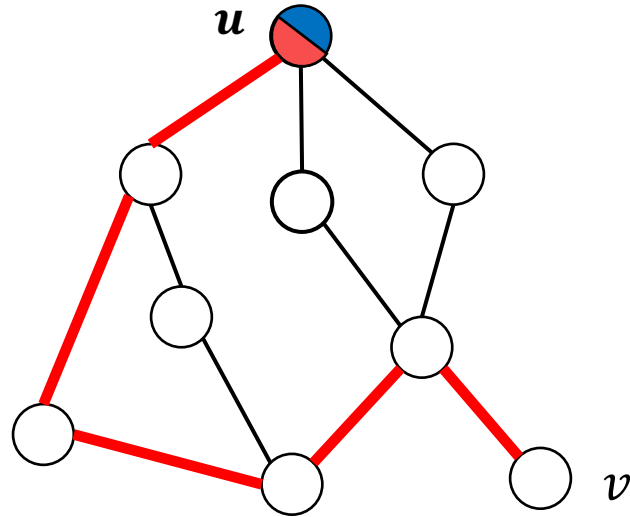
Construct a **coupling** (c_v, c'_v) between $\mu_v(\cdot | u \leftarrow i)$ and $\mu_v(\cdot | u \leftarrow j)$

$$d_{TV}(\mu_v(\cdot | u \leftarrow i), \mu_v(\cdot | u \leftarrow j)) \leq \Pr[c_v \neq c'_v]$$

Bound **one-to-all influence** by **coupling inequality**

$$\sum_{v \in V \setminus \{u\}} \text{Inf}(u \rightarrow v) \leq \sum_{v \in V \setminus \{u\}} \Pr[c_v \neq c'_v]$$

Recursive coupling



- Starting from the “**disagreement vertex**” u .
- Coupling vertex one by one in a “**DFS-manner**”.
- If the coupling on v fails (i.e. $c_v \neq c'_v$)
then **there is a path** \mathcal{P} from u to v
ALL vertices in \mathcal{P} **FAIL** in coupling.

Bound **one-to-all influence** by **enumerating all the paths from u**

$$\sum_{u \neq v} \text{Inf}(u \rightarrow v) \leq \sum_{\text{all paths } P \text{ from } u} \text{Influence along the path } P \leq O\left(\frac{1}{\delta}\right)$$

Triangle-free
Many colors



Coupling succeeds
with high prob.



Bounded total influence

Independent work

arXiv.org > cs > arXiv:2007.08058

Computer Science > Data Structures and Algorithms

[Submitted on 16 Jul 2020]

Rapid Mixing for Colorings via Spectral Independence

Zongchen Chen, Andreas Galanis, Daniel Štefankovič, Eric Vigoda

arXiv.org > cs > arXiv:2007.08091

Computer Science > Data Structures and Algorithms

[Submitted on 16 Jul 2020]

Rapid mixing from spectral independence beyond the Boolean domain

Weiming Feng, Heng Guo, Yitong Yin, Chihao Zhang

Theorem [Chen, Galanis, Štefankovič, Vigoda 20]

The Glauber dynamics for coloring is rapid mixing if $q \geq \alpha\Delta + 1$

- **different definition** of spectral independence
- **different method** to prove spectral independence for coloring

Summary

- A definition a **spectral independence** for general distribution (generalize def. in [ALO20])
- **Rapid mixing** of **Glauber dynamics** from spectral independence
- Application: sampling uniform q -coloring on **triangle-free graph** when
$$q \geq (\alpha + \delta)\Delta \approx (1.763 + \delta)\Delta.$$

Future work

- Improve the $n^{O(C)}$ mixing time for **general distribution**.
- Improve the $n^{O(1/\delta)}$ mixing time for **spin systems** (including coloring)
 - $O(n \log n)$ optimal mixing for spin systems with
 - spectral independence and $\Delta = O(1)$ [CLV20, arXiv:2011.02075].
- **Better condition** for spectral independence
 - example: prove spectral independence for graph coloring with **fewer colors**.

Thank you