# Report of MCMC Mixing on Treelike Graphs

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#### Outline

Preliminaries of MCMC

Randomly coloring graphs of bounded treewidth

Counting independent sets in graphs with bounded bipartite pathwidth

#### Preliminaries of MCMC

- Irreducible: state space is connected
- Aperiodic: state space is not bipartite
- Irreducible & aperiodic -> converge to a unique stationary distribution
  - Verifying local condition, for all x,y,  $\pi(x)p(x,y)=\pi(y)p(y,x)$
- This verifies the correctness of algorithm

#### Preliminaries of MCMC

- We also want fast mixing in polynomial time
- It is related to the largest eigenvalue of Markov matrix (lazy chain)
- Exponential number of states: impossible to do eigen decomposition

Idea: verifying local condition to prove global property

#### Preliminaries of MCMC

- If there is a 'bottleneck' in state graph: hard to converge
- Rapid convergence: no 'bottleneck', hard to prove all partition

- Dual problem: find a multi-commodity flow with low congestion!
- Only need to construct one and verify!

We define the congestion on an edge  $(\sigma, \sigma')$  with respect to a flow f by

$$\rho_f(\sigma, \sigma') = \frac{1}{q(\sigma, \sigma')} \sum_{\alpha, \beta \in \Omega} \sum_{p:(\sigma, \sigma') \in p \in \mathcal{P}_{\alpha, \beta}} f(p)|p|,$$

and the congestion of f by

$$\rho_f = \max_{(\sigma, \sigma') \in F} \rho_f(\sigma, \sigma').$$

#### 1.1 Results

Our main result is an algorithm that efficiently samples a  $((1+\epsilon)\Delta)$ -coloring (almost) uniformly at random if the input graph has logarithmically bounded pathwidth, for any  $\epsilon > 0$ .

**Theorem 1.1.** (Informal) Let  $\epsilon > 0$  and G be a graph with maximal degree  $\Delta$  and pathwidth bounded by  $O(\log n)$ . There exists a polynomial time algorithm for sampling a  $((1+\epsilon)\Delta)$ -proper coloring of G (almost) uniformly at random.

Using the fact that the pathwidth of a graph is at most  $O(\log n)$  times its treewidth [35], we have the following corollary.

Corollary 1.2. (Informal) Let  $\epsilon > 0$  and G be a graph with maximal degree  $\Delta$  and treewidth bounded by a constant. There exists a polynomial time algorithm for sampling a  $((1+\epsilon)\Delta)$ -proper coloring of G (almost) uniformly at random.

- Idea for sampling matchings
  - Construct a mutli-commodity flow
  - Find an injective function from paths going through the edge to state space
- It doesn't work for sampling proper colorings!
  - Unlike matchings, we cannot guarantee the following:
    - part of starting state + part of destination = a proper state
  - Changing one vertex may lead to illegal states, and cases vary!
    - This method allows single-flaw state

- Allowing single-flaw state
  - Once entering, we can fix it quickly
  - Not too many single-flaw state, so # of flows are polynomial times of sth.
- Could we eliminate single-flaw state?
  - Using a deterministic method to move to proper state?
  - Failed! A counter-example: star graph
  - Reason: deterministic method leads to large congestion!
- So we use randomness to split the flow!

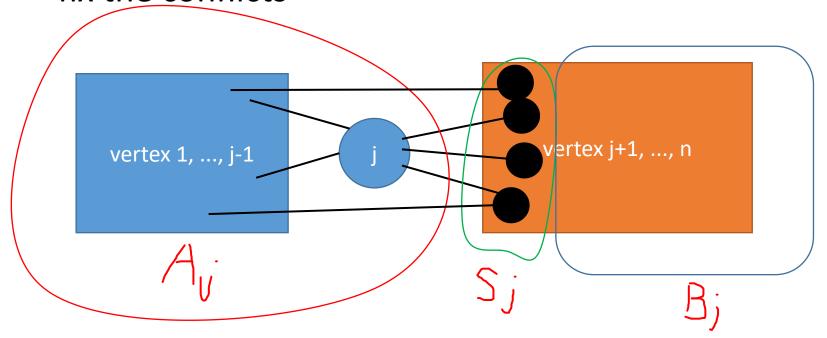
Let G = (V, E) be a graph with maximal degree  $\Delta$  and  $\epsilon > 0$  such that  $(1 + \epsilon)\Delta \geq \Delta + 2$ . The state space  $\Omega$  of Markov chain  $\mathcal{MC}(G, \epsilon)$  (or simply  $\mathcal{MC}$ ) is the set of all proper and singly-flawed k-colorings of G, for  $k = \lceil (1 + \epsilon)\Delta \rceil$ :  $\Omega = \mathcal{C}_p(G, k) \cup \mathcal{C}_{sf}(G, k)$ . For simplicity, we henceforth assume that  $\epsilon \Delta$ ,  $(1 + \epsilon)\Delta$  and  $(1 + \epsilon)\epsilon^{-1}$  are integers. It is easy to generalize the results to real values thereof. For  $\sigma \in \Omega$ , the transitions  $\sigma \to \sigma'$  of  $\mathcal{MC}$  are the following

- Let  $\sigma' = \sigma$ .
- With probability 1/2, do nothing (laziness).
- Otherwise, choose a vertex v and color c uniformly at random from V and [k] respectively. Tentatively, set  $\sigma'(v) = c$
- If  $\sigma' \notin \Omega$ , set  $\sigma'(v) = \sigma(v)$ .

- Technique: vertex separator, if there are not too many undeterministic vertices, i.e. vertex separator size is O(log n)
- The whole mixing time will be polynomial

- VSN = pathwidth, so pw = O(log n)
- tw = O(1) -> pw = O(log n)

- Canonical paths: n phases, each phase j with |S<sub>i</sub>|+1 steps
- Order vertices from 1 to n, using minimal order (i.e., reaching VSN)
- For any a, b  $\subseteq \Omega$ , phase j will change color of vertex j from a to b, and fix the conflicts



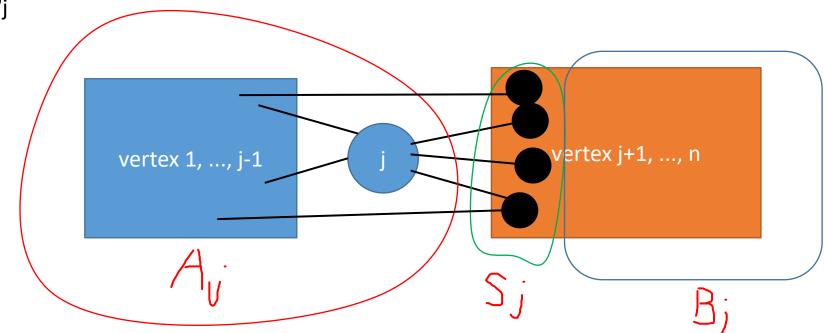
• First, assume a,b are proper colors

• Phase j step 1: change j to its color in b

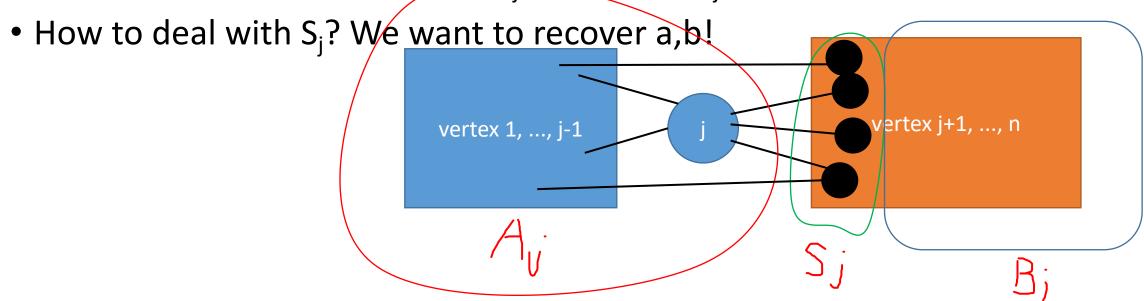
• Phase j, the next  $|S_i|$  steps: randomly choose one of feasible colors

for each vertex in S<sub>j</sub>

• VSN is O(log n)



- Similarly, we want an injective function from (a,b) to something
- In order to bound the number of flows
- From a to b, the current state has same A<sub>j</sub> as b, the same B<sub>j</sub> as a
- Get a proper coloring X, same A<sub>i</sub> as a, same B<sub>i</sub> as b!

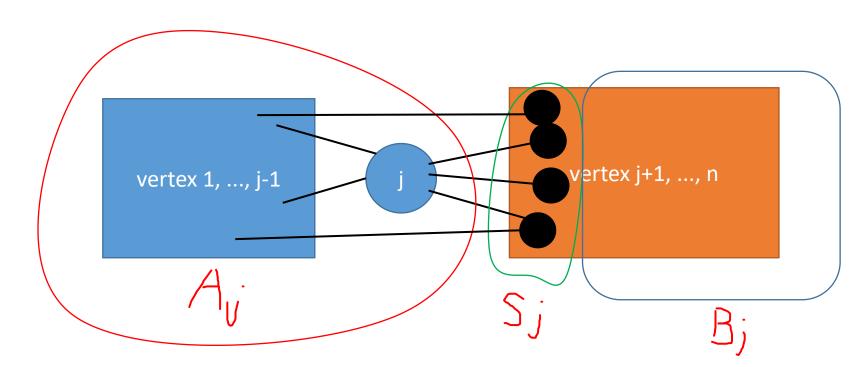


- Y is  $[(1+\epsilon)\epsilon^{-1}]^{|S_j|}$ , allowing us to pinpoint  $\beta(v)$  for every  $v \in S_j$ . tell the random choice of  $S_j$  by b
- Finally, Z is simply all possible colorings of the vertices of QS(j,t) under  $\alpha$ . tell the random choice of (part of)  $S_j$  (needs randomness) by a

Clearly x, y, z, j and t allow us to recover  $\alpha$  and  $\beta$ . The size of the co-domain is at most

$$|\mathcal{C}_p| \cdot ((1+\epsilon)\epsilon^{-1})^{|S_j|} \cdot k^{|QS(j,t)|}.$$

t tells which step we are in phase j



# Why Y is this way?

- Y is  $[(1+\epsilon)\epsilon^{-1}]^{|S_j|}$ , allowing us to pinpoint  $\beta(v)$  for every  $v \in S_j$ .
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Clearly x, y, z, j and t allow us to recover  $\alpha$  and  $\beta$ . The size of the co-domain is at most

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In canonical path, we 'split' the flow evenly, but not randomly. Evenly is enough!

For its previous color, we project to a feasible color evenly We only use  $\epsilon^{\triangle}$  feasible colors (even if we can use more for some vertex), so we can recover the info in  $(1+\epsilon)/\epsilon$  size set! Because in flow size, we only split  $\epsilon^{\triangle}$  times for each 'unknown' state!

• Each flow will split in the undeterministic vertices, so

Claim 3.3. The flow routed from  $\alpha$  to  $\beta$  through any  $t = (\sigma, \sigma') \in F$  in any phase  $j \in [n]$  is at most  $f_{j,t,\alpha,\beta} \leq \frac{\pi(\alpha)\pi(\beta)}{(\epsilon\Delta)^{|QS(j,t)|}}.$ 

- Combining together, we'll get a polynomial mixing time
- What does undeterministic give us? (my view)
  - Each flow becomes smaller
  - But there are many more flows
  - Leading to  $|S_i|$  in the exponential part, so we have to bound it in  $O(\log n)$
- Another question: it seems that k=△+1 also works
  - Since we have single-flaw states, MC is no longer reducible

- The last problem: flows may start / end at a single-flaw state
- There are not too many, each proper state may have kn times more neighbors, # of flows multiply at most 4k²n²

mixing time is still polynomial

Corollary 2.2. For any G = (V, E) such that |V| = n and  $k \ge \Delta + 2$ ,

$$|\mathcal{C}_{sf}(G,k)| \leq kn|\mathcal{C}_p(G,k)|.$$

*Proof.* Immediate from the surjectivity of the function g in Lemma 2.1.

Corollary 2.3. For any G = (V, E) such that |V| = n and  $k \ge \Delta + 2$ , there exists a function

$$g': \mathcal{C}_{sf}(G,k) \to \mathcal{C}_p(G,k),$$

for which each element in the co-domain has at most kn pre-images in the domain.

*Proof.* For every  $\sigma' \in \mathcal{C}_{sf}$ , arbitrarily select one pre-image  $(\sigma, v, c)$  w.r.t. g, and set  $\sigma$  as the image for  $\sigma'$  under g'.

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