

Lecture 6

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- Overview:
- Sampling Graph Colorings
 - Path Coupling

- Graph Colorings:
- Input: $G = (V, E)$ graph; $Q = \{1, \dots, q\}$ set of colors
 - Goal: Sample a u.a.r. legal coloring of G
 - ↳ assign a color to each vertex so that no two adjacent nodes get same color
 - Denote by Δ the maximum degree of G

Decision Problem:

- $q \geq \Delta + 1$: a legal coloring guaranteed to exist
- $q = \Delta$: Brook's theorem:
 - if $\Delta \geq 2$, the graph has a Δ -coloring iff it does not contain a $(\Delta + 1)$ -clique
 - if $\Delta = 2$, Δ -coloring exists \Leftrightarrow no $(\Delta + 1)$ -clique & no odd cycle
- $q < \Delta$: NP-hard to decide

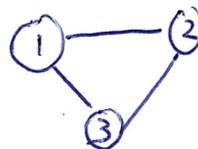
- Working Regime: $q \geq \Delta + 1$ \rightarrow since $q < \Delta$, NP-hard \rightarrow $q = \Delta$, non-trivial

Natural MC:

- start at arbitrary legal coloring
- pick vertex v u.a.r. and color c u.a.r.
- recolor v with c if this gives legal coloring, o.w. do nothing

Clearly symmetric, aperiodic;

Not necessarily irreducible if $q < \Delta + 1$; e.g.



$\Delta = 2$
 $q = 3$
no moves

- Exercise (1 pts) Chain is irreducible if $q \geq \Delta + 2$.

- Conjectures:

(1). (10 pts) Random sampling of legal colorings can be done in polynomial time if $q \geq \Delta + 1$.

(2). (10 pts) The MC given above has mixing time $O(n \log n)$ if $q \geq \Delta + 2$

- This lecture: Show (2) for $q \geq 2\Delta + 1$.

- Warming Up: ^{Thm:} The mixing time of MC given earlier is $O(n \log n)$ if $q \geq 4\Delta + 1$.

Proof: \rightarrow Use coupling; X_0, Y_0 are arbitrary; couple $(X_t)_t, (Y_t)_t$ as follows: pick same v, c at all times t

\rightarrow Denote: $d_t(X_t, Y_t)$ = number of colors where X_t, Y_t differ

\rightarrow Accounting: ~~for~~ an arbitrary pair X_t, Y_t , out of all $n \cdot q$ choices of vertex, color:

good moves ($d_{t+1} = d_t - 1$): choose disagreeing vertex v and choose color c that does not appear in neighborhood of v in neither of the colorings X_t, Y_t and hence will be accepted at least $(q - 2\Delta) \times d_t$ good moves

bad moves ($d_{t+1} = d_t + 1$): choose agreeing vertex v in the neighborhood of disagreeing vertex v_0 and color c that is acceptable in X_t & not in Y_t or vice versa
at most $\underbrace{2}_{\substack{\text{per} \\ \text{disagreeing} \\ \text{vertex } v_0}} \times \underbrace{\Delta d_t}_{\substack{\text{vertices in} \\ \text{neighborhood of disagreeing vertices}}} \text{ bad moves}$
~~at most~~ $2 \times \Delta d_t$ bad moves
 $C^X(v_0) \& C^Y(v_0)$

neutral moves ($d_{t+1} = d_t$): the rest

good moves - bad moves $\geq d_t \times (q - 4\Delta)$

$$\mathbb{E}[d_{t+1} | X_t, Y_t] \leq d_t - 1 \times \frac{d_t(q-2\Delta)}{qn} + 1 \times \frac{d_t 2\Delta}{qn} = d_t \left(1 - \frac{q-4\Delta}{qn}\right)$$

hence
$$\mathbb{E}[d_{t+1} | X_0, Y_0] \leq d_0 \left(1 - \frac{q-4\Delta}{qn}\right)^t$$

$$\leq n \cdot e^{-t \cdot \frac{q-4\Delta}{qn}} \leq \varepsilon$$

$$\hookrightarrow t \geq \frac{q}{q-4\Delta} \cdot n \cdot (\log n + \log \frac{1}{\varepsilon})$$

$$\Rightarrow \Pr[X_t \neq Y_t | X_0, Y_0] = \Pr[d_t > 1 | X_0, Y_0] \leq \mathbb{E}[d_t | X_0, Y_0] \leq \varepsilon$$

$$\hookrightarrow t \geq \frac{q}{q-4\Delta} n (\log n + \log \frac{1}{\varepsilon})$$

hence: $\tau(\varepsilon) = \frac{q}{q-4\Delta} n (\log n + \log \frac{1}{\varepsilon})$

$$\tau_{\text{mix}} = O\left(\frac{q}{q-4\Delta} n \log n\right)$$

[sanity check:
the larger q is
the faster it mixes]

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Part II: Path Coupling

(4)

Bubley & Dyer '97

- Def: A pre-metric on Ω is a connected, undirected graph with positive weights s.t. every edge is a shortest path

\rightarrow i.e. if $x, y \in \Omega$ are adjacent in pre-metric, the weight of edge (x, y) should be = to shortest path weight

- Can extend pre-metric to metric by considering shortest path distances

- Idea of Path Coupling: define coupling only for pairs of states that are adjacent in pre-metric

(Recall coupling of a mc needs to specify for all pairs of states z_1, z_2 : $\Pr[X_{t+1}, Y_{t+1} | X_t = z_1, Y_t = z_2]$; path coupling requires that we only specify such distn's for z_1, z_2 that are adjacent in premetric.)

Theorem: Suppose exists coupling $(X, Y) \xrightarrow{\Pr} (X', Y')$ defined only for pairs (x, y) of states that are adjacent in pre-metric, & such that:

$$\mathbb{E} [d(X', Y') | x, y] \leq (1 - \alpha) \cdot d(x, y) \quad (*)$$

for some $\alpha \in [0, 1]$, where d is the metric extending the pre-metric

• Then can extend coupling to full coupling so that $(*)$ is satisfied for all $x, y \in \Omega \times \Omega$.

Proof: Let x, y be arbitrary states that are non-adjacent in premetric.

• fix arbitrary shortest path between these states:

$$x \equiv z_0 \rightarrow z_1 \rightarrow \dots \rightarrow z_{k-1} \rightarrow z_k \equiv y$$

• Need to specify: $\Pr[Z'_0, Z'_k | z_0, z_k]$ that is a valid mc coupling from (z_0, z_k) .

• We construct this from $\Pr[Z'_0, Z'_1 | z_0, z_1], \dots, \Pr[Z'_{k-1}, Z'_k | z_{k-1}, z_k]$ which are known and given.

Pr [Z'_0, Z'_k | z_0, z_k] is sampled via the following randomized procedure:

- (1) sample Z'_0, Z'_1 from Pr [Z'_0, Z'_1 | z_0, z_1]
- (2) conditioning on sampled value of Z'_1 sample Z'_2 from distribution Pr [Z'_1, Z'_2 | z_1, z_2]
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- (k) conditioning on sampled value of Z'_{k-1} sample Z'_k from distribution Pr [Z'_{k-1}, Z'_k | z_{k-1}, z_k]

Claim 1: The above procedure:

- i. is well-defined
- ii. is a valid coupling

mc matrix

Proof: We show by induction that Pr [Z'_i = w] = P(z_i, w), ∀ i, w ∈ Ω. i.e. we show that the step z_i → Z'_i is a valid Markov Chain step from z

- For Z'_0, Z'_1 obvious, since Pr [Z'_0, Z'_1 | z_0, z_1] defines valid coupling.

- Suppose claim true for Z'_i; will show true for Z'_{i+1}

Since Pr [Z'_i, Z'_{i+1} | z_i, z_{i+1}] is valid coupling } & claim is true for Z'_i

we have

$$Pr [Z'_{i+1} = w] = \sum_{z'_i} P(z_i, z'_i) \cdot \frac{Pr [Z'_i = z'_i; Z'_{i+1} = w | z_i, z_{i+1}]}{\sum_w Pr [Z'_i = z'_i; Z'_{i+1} = w | z_i, z_{i+1}]} \quad (*)$$

↳ value sampled for Z'_i in step i of procedure

Pr (z_i, z'_i) since Pr [Z'_i, Z'_{i+1} | z_i, z_{i+1}] is valid coupling

$$\Rightarrow Pr [Z'_{i+1} = w] = \sum_{z'_i} Pr [Z'_i = z'_i; Z'_{i+1} = w | z_i, z_{i+1}]$$

Pr (z_{i+1}, w) because Pr [Z'_i, Z'_{i+1} | z_i, z_{i+1}] is valid coupling. probability that z_i is sampled in previous step; by induction hypothesis this is equal to P(z_i, z'_i)

$$\Rightarrow \Pr[Z'_{i+1} = w] \equiv P(z_{i+1}, w), \forall i, w \in \Omega$$

→ this readily establishes (ii)

→ regarding (i), we have already implicitly shown it:

- notice that the highlighted ratio in (*) is the conditional probability from which we sample Z'_{i+1} conditioning on the value z'_i that was sampled for Z'_i in previous step; i.e.
- the above analysis shows that we only compute this ratio for values z'_i in the support of $P(z_i, z'_i)$ and hence the denominator is never 0. ✗

Claim 2: The distribution of (Z'_i, Z'_{i+1}) sampled from the procedure is identical to $\Pr[Z'_i, Z'_{i+1} | z_i, z_{i+1}]$, for all i .

Proof: follows easily from claim 1.

Now:

$$\begin{aligned} \mathbb{E}[d(Z'_0, Z'_k) | z_0, z_k] &\stackrel{\text{triangle inequality}}{\leq} \mathbb{E}\left[\sum_{i=0}^{k-1} d(Z'_i, Z'_{i+1}) | z_0, z_k\right] \\ &\stackrel{\text{linearity of expectation}}{\leq} \sum_{i=0}^{k-1} \mathbb{E}[d(Z'_i, Z'_{i+1}) | z_0, z_k] \\ &\stackrel{\text{using claim 2}}{\leq} \sum_{i=0}^{k-1} (1-d) d(z_i, z_{i+1}) \\ &= (1-d) \cdot \underbrace{\sum_{i=0}^{k-1} d(z_i, z_{i+1})}_{\parallel d(z_0, z_k)} \end{aligned}$$

because $z_0 \rightarrow z_1 \rightarrow \dots \rightarrow z_k$ was a shortest path. ✗