

# \* Improving the coloring chain analysis

## • One-step couplings

- Recall our analysis of the graph-coloring MC

-  $(X_t, Y_t) \equiv$  Pick  $(V_t, C_t)$  u.a.r and apply to both  
node  $\swarrow$   $V_t$   $\swarrow$   $C_t$   $\swarrow$  color  $\swarrow$  if proper coloring

-  $d_t = d_t(X_t, Y_t) \equiv$  # of disagreeing vertices in  $X_t, Y_t$

- Now suppose we analyze  $d_t$  backwards from the point of coupling. Let  $x, y \in \Omega (= \mathcal{Q}^{|\Omega|})$  st  $d(x, y) = 1$  (i.e., differ in 1 vertex).

- Let  $X_1^x, Y_1^y$  be one step of the coupling

-  $\mathbb{P}[d(X_1^x, Y_1^y) = 0] = \mathbb{P}_{x, y}[d_1 = 0] \geq \frac{q - \Delta}{nq}$  ← Since the colors of all ~~but~~ one vertex is same

(note = for general  $x, y$ , we had  $\mathbb{P}_{x, y}[d_1 = 0] \geq \frac{q - 2\Delta}{nq}$ )

$\mathbb{P}_{x, y}[d_1 = 2] \leq \frac{2\Delta}{nq}$  (similar to  $\frac{2 \cdot \Delta dt}{nq}$ )

$\Rightarrow \mathbb{E}_{x, y}[d_1 - 1] \leq \frac{3\Delta - q}{nq} \Rightarrow \mathbb{E}_{x, y}[d_1] \leq 1 - \frac{(q - 3\Delta)}{nq}$   
 $d(x, y) = 1$

- Now for any  $x, y$  with  $d(x, y) = r$ , there is some sequence of ~~vertices~~ <sup>colorings</sup>  $x_0 = x, x_1, x_2, \dots, x_r = y$  st  $d(x_k, x_{k+1}) = 1$

(2)  
 - Since  $d(\cdot, \cdot)$  is a metric, then by  $\Delta$  inequality, for any  $x, y$  with  $d(x, y) = \delta$

$$\mathbb{E}[d(X_t^x, Y_t^y)] \leq \sum_{k=1}^t \mathbb{E}[d(X_k^{x_{k-1}}, X_k^{y_{k-1}})]$$

$$\leq d(x, y) \left(1 - \frac{q-3\Delta}{nq}\right)^t$$

- Now for our coupling  $(X_t, Y_t)$ , given any starting states  $x, y$ , let  $X_{t-1} = x_{t-1}, Y_{t-1} = y_{t-1}$ . Then  $(X_t, Y_t)$  (starting from  $x, y$ ) has the same distn<sup>n</sup> as  $(X_t^{x_{t-1}}, Y_t^{y_{t-1}})$

$$\Rightarrow \mathbb{E}[d(X_t^x, Y_t^y) | X_{t-1}^x = x_{t-1}, Y_{t-1}^y = y_{t-1}] = \mathbb{E}[d(X_t^{x_{t-1}}, Y_t^{y_{t-1}})]$$

$$\leq d(x_{t-1}, y_{t-1}) \left(1 - \frac{q-3\Delta}{nq}\right)$$

$$\Rightarrow \mathbb{E}[d(X_t^x, Y_t^y)] \leq \mathbb{E}[d(X_{t-1}^x, Y_{t-1}^y)] \left(1 - \frac{q-3\Delta}{nq}\right)$$

$$\leq d(x, y) \left(1 - \frac{q-3\Delta}{nq}\right)^t \leq n \exp\left(-\frac{(q-3\Delta)t}{nq}\right)$$

$$\Rightarrow \mathbb{P}[d(X_t^x, Y_t^y) \geq 1] \leq n e^{-t \left(\frac{q-3\Delta}{nq}\right)}$$

$$\Rightarrow t_{\text{mix}}(1/2e) \leq \left(\frac{n}{q-3\Delta}\right) n(\ln(n) + \ln(2e))$$

\* Idea for improving to  $q \geq 2\Delta + 1$

- In the previous analysis, we lost a  $2\Delta$  factor in analyzing the 'bad move', i.e.,  $P[d(x_t^x, y_t^y) = 2] \leq \frac{2\Delta}{nq}$
- Consider an alternate coupling for all ~~states~~
  - $X_t, Y_t$  s.t.  $d(x_t, y_t) = 1$  (Let  $v_0 \equiv$  disagreeing vertex)
    - Pick  $(V_t, C_t)$  u.a.r. If  $V_t = v_0$  or  $V_t \notin N(v_0)$ , <sup>neighborhood</sup>
      - ~~assigning~~ color with  $C_t$  if possible
    - If  $V_t \in N(v_0)$ , then color  $X_t$  with  $C_t$ , and  $Y_t$  with  $C'_t$  s.t.
 
$$C'_t = \begin{cases} C_x & \text{if } C_t = C_y \\ C_y & \text{if } C_t = C_x \\ C_t & \text{if } C_t \notin \{C_x, C_y\} \end{cases}$$

$(C_x \equiv \text{color of } v_0 \text{ in } X_t)$

$(C_y \equiv \text{color of } v_0 \text{ in } Y_t)$
- Check this is still a valid coupling for  $(X_t, Y_t)$  when  $d(x_t, y_t) = 1$ .
- Now let's try to redo our previous proof with this coupling

- As before, for good moves, we have (for  $d(x,y)=1$ ) (4)

$$P[d(x^x, y^y) = 0] \geq \frac{q - \Delta}{nq}$$

- For bad moves, now we have an error only if  $C_t = C_y$

$$\Rightarrow P[d(x^x, y^y) = 2] \leq \frac{\Delta}{nq}$$

$\leftarrow P[\forall v \in N(v_0)]$ , only one option for  $C_t$  as if  $C_t = C_y$ , we can not color in either chain!

- Now we get  ~~$P_{xy}$~~   $E_{xy}[d_t] \leq 1 - \left(\frac{q - 2\Delta}{nq}\right)!$

However, we only defined the coupling for 'adjacent'  $X_t, Y_t$  (i.e.,  $d(x_t, y_t) = 1$ )

- If we could guarantee that  $\exists$  coupling  $(X_t, Y_t)$  that matches the one above for all  $x, y$  s.t.  $d(x, y) = 1$ , then we are done (extension thm)

- this is what we get from path coupling

- Suppose we are given a <sup>(connected, undirected)</sup> graph  $H(\Omega, E_0)$  on the state space  $\Omega$  of a MC, with edge lengths  $l(x,y)$  (s.t.  $l(x,y) \geq 1 \forall x,y \in E_0$ ). ~~Assume~~ The path metric on  $\Omega$  is

(Assume  $H$  is a pre-metric, i.e.  $l(x,y) = l(y,x) \forall (x,y) \in E_0$ )

$$P(x,y) = \min \{ \text{length of } \xi \mid \xi \text{ path from } x \rightarrow y \text{ in } H \}$$

- Note -  $H$  may be different from the graph  $G$  associated with MC  $(\Omega, P)$

- Thm . Suppose for every edge  $(x, y) \in E_0$ ,  $\exists$  a coupling  $(X_1, Y_1)$  for  $P(x, \cdot), P(y, \cdot)$  s.t.

$$E_{x,y} [f(X_1, Y_1)] \leq f(x, y) e^{-\alpha} = l(x, y) e^{-\alpha}$$

Then for any  $(x, y) \in \Omega$ , we have

$$E_{x,y} [f(X_1, Y_1)] \leq e^{-\alpha} f(x, y)$$

Pf 1 - Direct construction of coupling

- Given  $x, y$  arbitrary states in  $\Omega$ , and a pre-metric, let  $x = z_0 \rightarrow z_1 \rightarrow \dots \rightarrow z_n = y$  be a shortest path.
- We are given couplings  $\mathbb{P}[X_1^{z_k}, Y_1^{z_{k+1}}]$  for all  $z_k, z_{k+1}$  (since  $(z_k, z_{k+1}) = \text{edge in pre-metric}$ )
- - Sample  $(\hat{X}_1^{z_0}, \hat{X}_1^{z_1})$  from coupling  $\mathbb{P}$  for  $P(z_0, \cdot), P(z_1, \cdot)$
- Conditioned on  $\hat{X}_1^{z_1}$ , sample  $(\hat{X}_1^{z_1}, \hat{X}_1^{z_2})$  from next coupling
- Repeat to get  $\hat{X}_1^{z_2}, \hat{X}_1^{z_3}, \dots, \hat{X}_1^{z_n} = \hat{X}_1^y$
- $(\hat{X}_1^{z_0}, \hat{X}_1^{z_n}) = (X_1^x, X_1^y)$  is a coupling for  $P(x, \cdot), P(y, \cdot)$

• To show this is a valid coupling, we show

by induction that  $\mathbb{P}[\hat{X}_1^{z_k} = w] = P(z_k, w)$

- For  $X_1^{z_0}, X_1^{z_1}$ , this is true (since it's a coupling)

- Assume true for  $\hat{X}_1^{z_k}$ . Then

$$\mathbb{P}[\hat{X}_1^{z_{k+1}} = w] = \sum_{w' \in \Omega} \mathbb{P}[\hat{X}_1^{z_k} = w'] \cdot \frac{\mathbb{P}[\hat{X}_1^{z_k} = w', \hat{X}_1^{z_{k+1}} = w]}{\sum_{s \in \Omega} \mathbb{P}[\hat{X}_1^{z_k} = w', \hat{X}_1^{z_{k+1}} = s]}$$

$$= \sum_{w' \in \Omega} \underbrace{P(z_k, w')}_{\text{induction}} \cdot \frac{P(\hat{X}_1^{z_k} = w', \hat{X}_1^{z_{k+1}} = w)}{\underbrace{P(z_k, w')}_{\text{valid coupling}}}$$

$$= \sum_{w' \in \Omega} \mathbb{P}[\hat{X}_1^{z_k} = w', \hat{X}_1^{z_{k+1}} = w] = P(z_{k+1}, w)$$

$$\begin{aligned} \mathbb{E}[f(X_1^x, X_1^y)] &\leq \mathbb{E} \left[ \sum_{k=0}^{\eta-1} f(X_1^{z_k}, X_1^{z_{k+1}}) \right] \\ &\leq \sum_{k=0}^{\eta-1} \mathbb{E} [f(X_1^{z_k}, X_1^{z_{k+1}})] \leq \sum_{k=0}^{\eta-1} e^{-\alpha} f(z_k, z_{k+1}) \\ &= e^{-\alpha} \sum_{k=0}^{\eta-1} f(z_k, z_{k+1}) = e^{-\alpha} f(x, y) \end{aligned}$$

Corollary - If we have a path coupling as above, and then  
 (Note -  $\frac{1}{\alpha}$  for nbrs of  $y$ )  $t_{\text{mix}}(\epsilon) \leq \left\lceil \frac{\ln(D) - \ln(\epsilon)}{\alpha} \right\rceil$ , where  $D \equiv$  diameter of  $\Omega$  under metric

Pf 2 (Via the Wasserstein distance)

- For any given metric  $\rho(\cdot, \cdot)$  over  $\Omega$ , the following quantity is called the Wasserstein metric (or Kantorovich metric) or transportation metric over distributions on  $\Omega$

$$d_\rho(\mu, \nu) = \inf \{ \mathbb{E}[\rho(x, y)] \mid (x, y) \text{ is a coupling of } \mu, \nu \}$$

- Remarks on  $d_\rho(\cdot, \cdot)$

- If  $\rho(x, y) = \mathbb{1}_{\{x \neq y\}}$ , then  $d_\rho(\mu, \nu) = d_{TV}(\mu, \nu)$

- For any  $X, Y, Z$ , we have  $d_\rho(X, Z) \leq d_\rho(X, Y) + d_\rho(Y, Z)$   
( $\Delta$ -inequality for the Wasserstein metric)

- The set of distn  $(\mu, \nu)$  on  $\Omega \times \Omega$  is a compact subset of  $\mathbb{R}^{|\Omega|^2}$  (i.e., the  $|\Omega|^2$  simplex)

The set of distn in  $\Omega_1 \times \Omega_2$  st the projection on  $\Omega_1 = \mu$ , on  $\Omega_2 = \nu$  is a compact subset of the  $|\Omega|^2$  simplex

$$d_\rho(\mu, \nu) = \inf \left\{ \sum_{(x, y) \in \Omega \times \Omega} \rho(x, y) q(x, y) \mid \underbrace{q(\cdot; \Omega) = \mu}_{\text{marginals}}, \underbrace{q(\Omega, \cdot) = \nu}_{\text{marginals}} \right\}$$

Note - the fn  $q \mapsto \sum_{(x, y) \in \Omega \times \Omega} \rho(x, y) q(x, y)$  is continuous on set of couplings

⇒ There exists  $q^*$  s.t.  $\sum_{(x,y) \in \Omega \times \Omega} p(x,y) q^*(x,y) = d_p(\mu, \nu)$

This is called the optimal coupling  $(X_p^*, Y_p^*)$  for  $d_p(\mu, \nu)$ . Note  $\mathbb{E}[p(X_p^*, Y_p^*)] = d_p(\mu, \nu)$

• Returning to path coupling, we are given a ~~and~~ metric  $p$  over  $\Omega$ . Now fix  $x, y \in \Omega$  and let  $(x=z_0, z_1, \dots, z_n=y)$  be a shortest path.

- by  $\Delta$  inequality for  $d_p$ , we have

$$d_p(P(x, \cdot), P(y, \cdot)) \leq \sum_{k=0}^{n-1} d_p(P(x_k, \cdot), P(x_{k+1}, \cdot))$$

- Now if for any edge  $(a,b)$ , we have  $l(a,b) \leq p(a,b) \leq 1$

and by assumption,  $d_p(P(a, \cdot), P(b, \cdot)) \leq e^{-\alpha} p(a,b)$

$$\Rightarrow d_p(P(x, \cdot), P(y, \cdot)) \leq e^{-\alpha} \sum_{k=0}^{n-1} p(z_k, z_{k+1}) \leq e^{-\alpha} p(x,y)$$

- Moreover, we know  $\exists$  coupling  $(X_i^z, X_i^y)$  s.t.

$$\mathbb{E}[d_p(X_i^z, X_i^y)] = \mathbb{E}[p(X_i^z, X_i^y)] \leq e^{-\alpha} p(x,y)$$

$$\Rightarrow \mathbb{E}[p(X_i^z, X_i^y)] \leq e^{-\alpha} p(x,y)$$



Balanced coupling - If we have a pre-metric  $\rho(\cdot, \cdot)$  on  $\Omega$ , and a coupling  $(X, Y)$  on adjacent states  $x, y$  s.t

$$\mathbb{E}[\rho(X^x, X^y)] \leq (1-\alpha) \rho(x, y)$$

- If  $\alpha < 1$ ,  $t_{\text{mix}}(\epsilon) = O(\alpha^{-1} \ln(D))$  where  $D = \max_{x,y} \rho(x,y)$

- If  $\alpha = 1$ , then  $t_{\text{mix}}(\epsilon) = O(D^2/\beta)$ , where

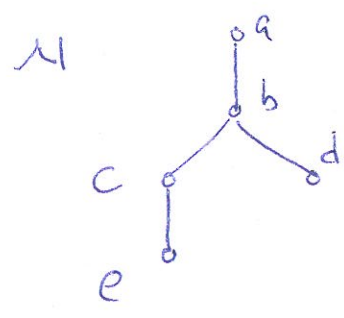
$$\beta = \min_{x,y \in \Omega} \mathbb{E}[(\rho(X^x, X^y) - \rho(x,y))^2]$$

(Note -  $\beta \geq \min_{x,y \in \Omega} \mathbb{P}[|\rho(X^x, X^y) - \rho(x,y)| \geq 1]$ )

Pf via Martingale Optional Stopping Thm (later in course)

Application - Sampling linear extensions of a partial order

- Input - a partial order  $\leq$  on  $V = \{1, 2, \dots, n\}$
- Output - a linear extension (ie, a total order  $\leq$  on  $V$  that respects  $\leq$  in the sense  $x \leq y \Rightarrow x \leq y$ ) u.a.s



- $a \leq b \leq c \leq d \leq e$
- $a \leq b \leq c \leq e \leq d$
- $a \leq b \leq d \leq c \leq e$

linear extension  $\Omega \subseteq \text{Set of permutations } \sigma(\cdot) \text{ on } V$

- Applications in combinatorics, ranking, <sup>approx</sup> sorting, decision theory, etc.

- Markov chain - wp  $\frac{1}{2}$ , do nothing
- wp  $\frac{1}{2}$ , pick random  $p \in \{1, 2, \dots, n-1\}$  and exchange  $\sigma(p), \sigma(p+1)$  if possible (ie, if new order is a valid extension)

- MC is aperiodic, symmetric (so  $\pi$  is uniform). Check that it is irreducible.

- Analysis using path coupling.

- pre-metric:  $x, y \in \Omega$  are adjacent iff  ~~$\sigma(x) \neq \sigma(y)$~~  differ in 2 posns  $1 \leq i < j \leq n$ 
  - distance between adjacent pairs =  $j-i$
- natural extension to metric (may be complicated)

- Coupling for adjacent pairs - Given adjacent  $x, y$  differing in  $i, j$

- Case 1 - If  ~~$j \neq i+1$~~   $j \neq i+1$ : Same as MC

Case 2 - If  $j = i+1$ : i) wp  $\frac{1}{2(n-1)}$  - do nothing in  $X$  swap  $i, i+1$  in  $Y$

ii) wp  $\frac{1}{2(n-1)}$  - do nothing in  $Y$  swap  $i, i+1$  in  $X$

- iii) w.p.  $\frac{n-2}{2(n-1)}$  - do nothing in both (11)
- iv) w.p.  $\frac{1}{2}$ , choose  $p$  in  $\{1, \dots, n-1\} \setminus \{i\}$  and swap  $\sigma(p), \sigma(p+1)$  if possible.

## Analysis of coupling

- If  $p \notin \{i-1, i, j-1, j\} \Rightarrow d(x_i^x, x_i^y) = d(x, y) = j-i$
- If  $p = i-1$  or  $j \Rightarrow d(x_i^x, x_i^y) \leq d(x, y) + 1 = j-i+1$ 
  - Happens with prob  $(\frac{1}{n-1} \cdot 2) \cdot 2 = \frac{1}{n-1}$
- If  $p = i$  ~~or  $p = j-1$~~  or  $p = j-1$ 
  - If  $j-i=1$ , then  $d(x_i^x, x_i^y) = 0 = d(x, y) - 1$
  - If  $j-i > 1$ : Note that  $\sigma(i), \sigma(i+1)$  and  $\sigma(j), \sigma(j+1)$  must be incomparable in  $\leq \Rightarrow$  exchange is legal!  $\leq$
  - $d(x_i^x, x_i^y) = d(x, y) - 1$  w.p.  $\frac{1}{n-1}$
- Thus  $E[d(x_i^x, x_i^y)] \leq d(x, y)$ 

adjacent  $x, y$
- By path coupling,  $t_{\text{mix}}(\epsilon) = O(\beta^{-1} D^2) = O(n^5)$ , as  $D \leq n \binom{n}{2} = O(n^2)$ , and  $\beta = \min_{x, y \in \Omega} \mathbb{P}[|H(x_i^x, x_i^y) - d(x, y)| \geq 1] \leq \frac{1}{n}$