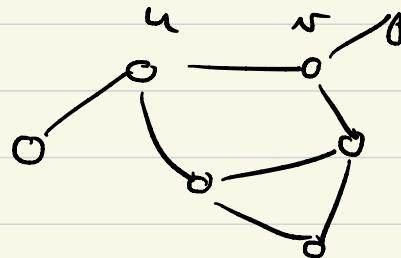


Coloring problem : Analysis of the Metropolis chain.

Recap the problem.

$$G = (V, E)$$



sample from the space of proper colorings.

i.e. s.t. two neighboring vertices $(u, v) \in E$

don't have same color.

Notation: $\{1, 2, 3, \dots, q\}$ = set of colors.

$\Delta = \max$ degree of a vertex $v \in V$.

$$\underline{x} = (x_1, x_2, \dots, x_N) \quad N = |V|.$$

and x_v color assigned to $v \in V$.

$$\pi(\underline{x}) = \frac{\mathbb{I}(\underline{x} \text{ is proper})}{Z} \quad Z \leftarrow \text{counts total # of proper cols}$$

$$N = |V|$$

Recap the proposed algorithm.

- Start from an initial proper coloring.
- Select $v \in V$ uniformly at random
- Select a color $c \in \{1, 2, \dots, q\}$ uniform random
- Recolor vertex v if c is an allowed color.
(otherwise we do nothing).

Recap the theorem.

Theorem: Assume $\underline{q} > 3\Delta$ then for any

initial proper coloring \underline{x}

$$\|\underline{P}_{\underline{x}}^m - \pi\|_{TV} \leq N e^{-\frac{m}{N}\left(1 - \frac{3\Delta}{q}\right)}.$$

distn with initial cond \underline{x}
after m iterations.

and the mixing time $T_\epsilon = \inf \{m \geq 1 : \max_{\substack{\underline{x} \text{ prop} \\ \text{wl}}} \|\underline{P}_{\underline{x}}^m - \pi\| \leq \epsilon\}$

satisfies

$$T_\epsilon \leq \left(1 - \frac{3\Delta}{q}\right)^{-1} \left\{ \underline{N} \log \underline{N} + \log \frac{1}{\epsilon} \right\}.$$

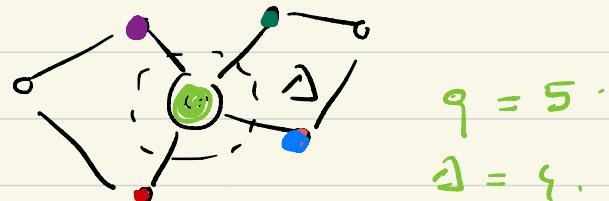
Remark: The theorem with a similar but more advanced proof holds for $q > 2\Delta$.

q smaller: out of our scope here and the problem becomes much harder.

(q too small: then breaks down.)

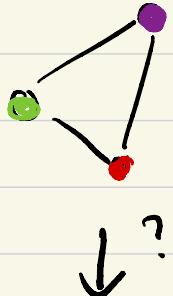
Remark:

- $q = \Delta + 1$ \Rightarrow certainly we can color the graph.

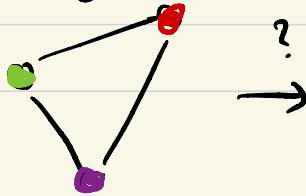


- but the chain of the algo might not be irreducible.
for example

$$\Delta = 2 \text{ and } q = 3$$



Can you move out of this configuration by following the algo steps ?



... .

• It is a fact that for $\underline{q > 3\Delta}$ the chain is
irreducible.

But also for $\underline{q \geq \Delta + 2}$ as one can show.



follow up this during the quiz session.

for a little proof.

Remark: The proof of the thm. will use this
fact that for $q > 3\Delta$ the chain is irreducible.

The proof will proceed by a coupling
argument:

$\underbrace{X_n}_{\text{of interest}}$ and $\underbrace{Y_n}_{\text{another chain}} \leftarrow$ coupled chains.

Recap the property: $\| P_x^m - Q_y^m \|_{TV} = \inf_{\{\text{couplings}\}} \Pr(X_n \neq Y_n)$

Proof of Theorem.

Coupled chains.

- $(X_m, m \geq 0)$ the chain starting at $X_0 = \underline{x}$
the proper initial coloring. and follows the steps of
the algo : select $v \in V$ at random, select $c \in \{1, \dots, p\}$
at random, recolor if c is allowed.
- $(Y_m, m \geq 0)$ starts at $\underline{Y}_0 \sim \underline{\pi}$ and then
evolves with the same steps than $(X_m, m \geq 0)$ with
the same $v \in V$ and $c \in \{1, \dots, p\}$ at each time step.

↑
so the chains are coupled.

Hamming distance between chains at each time step.

$$d(\underline{x}_m, \underline{y}_m) \equiv \sum_{v \in V} I(X_v^{(m)} \neq Y_v^{(m)}).$$

Since this is same coupling between chain:

$$\|\underline{P}_x^n - \pi\| \leq \mathbb{P}(X_n \neq Y_n)$$

↓
 at all times
 dist of chain $(Y_m, m \geq 0)$
 is π because π
 is stat distr.

$\mathbb{P}(d(X_n, Y_n) \geq 1)$

Markov
inequality: \leq

$$\mathbb{E}(d(X_n, Y_n))$$

We have a nice inequality to start with:

$$\|\underline{P}_x^n - \pi\|_{TV} \leq \mathbb{E}(d(X_n, Y_n)).$$



} we are going to bound this
 expectation now. To prove
 $\leq N \exp\left(-\frac{m}{N}\left(1 - \frac{3\Delta}{\rho}\right)\right).$

Proceed by induction:

- First we assume that $d(x_0, \gamma_0) = 1$
prove the bound.
- Then we generalize to $d(x_0, \gamma_0) = r \geq 1$.
- Finally we conclude.
#.

Assume at time at $n=0$ $d(x_0, \gamma_0) = 1$.

$\exists w \in V$ s.t. $x_w \neq \gamma_w$ (at time zero)

and $x_w = \gamma_w \nexists w \neq w$.

$\Rightarrow d(x_1, \gamma_1) \in \{0, 1, 2\}$.

$$\Rightarrow \underline{\mathbb{E}(d(x_1, \gamma_1))} = 0 \cdot P(d(x_1, \gamma_1) = 0) + 1 \cdot P(d(x_1, \gamma_1) = 1) + 2 \cdot P(d(x_1, \gamma_1) = 2)$$

$$+ 2 \cdot P(d(x_1, \gamma_1) = 2)$$

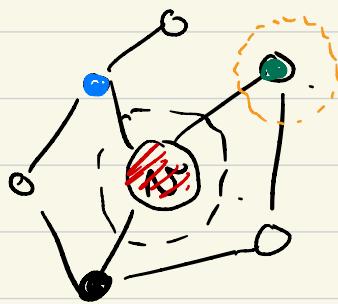
$$= \underline{\{1 - P(d(x_1, \gamma_1) = 0)\}} + \underline{P(d(x_1, \gamma_1) = 2)}$$

at time 0 if $d(x_0, t_0) = 1$

$$E(d(x_i, t_i)) = 1 - P(d(x_i, t_i) = 0) + P(d(x_i, t_i) = 2).$$

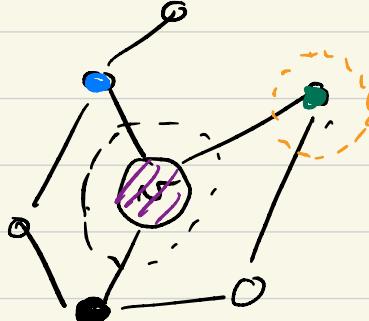
A

$$P(d(x_i, t_i) = 0) = \frac{1}{N} \frac{\# \text{ allowed colorator}}{q} \geq \frac{1}{N} \frac{q-1}{q}.$$



x_0

$\Delta = 3$



t_i

$$P(d(x_i, t_i) = 2) \leq \frac{\Delta}{N} \frac{2}{q}.$$

↑

- if the selected vertex w is not a neighbor of x you will do the same recoloring or non-recoloring of w in both chains and $d(x_i, t_i)$ remains = 1.

- The event $d(x_i, t_i) = 2$ happens only if w is a neighbor \rightarrow It should be that you recolor w is X and not in T OR recolor w in T and w in X .

$$\boxed{E(d(x_i, \tau_i)) \leq 1 - \frac{1}{n} \left(1 - \frac{3\gamma}{9}\right)} \quad \star$$

Generalize to case $d(x_0, \tau_0) = r$:

Claim by irreducibility (for $\gamma > 3\Delta$ or in fact for $\gamma \geq \Delta + 2$):

\exists path between assignments:

$$x_0 \rightarrow z_0^{(0)} \rightarrow z_0^{(1)} \rightarrow \dots z_0^{(k)} \rightarrow \dots z_0^{(r-1)} \rightarrow \tau_0. \quad \checkmark$$

s.t. $d(z_0^{(0)}, z_0^{(1)}) = 1$. all dist between assignments are equal to 1.

Let evolve the chains by one time unit:

$$x_1 \rightarrow z_1^{(0)} \rightarrow z_1^{(1)} \rightarrow \dots z_1^{(k)} \rightarrow \dots z_1^{(r-1)} \rightarrow \tau_1.$$

By triangle inequality

$$d(x_i, \tau_i) \leq d(x_i, z_i^{(0)}) + d(z_i^{(0)}, z_i^{(1)}) + \dots + d(z_i^{(r-1)}, \tau_i).$$

Take the expectation, use linearity, use result (*) under \checkmark

\Rightarrow

$$\boxed{E(d(x_i, \tau_i)) \leq r \left\{ 1 - \frac{1}{n} \left(1 - \frac{3\gamma}{9}\right) \right\}}. \quad \star$$

Conclusion of proof:

Remark by homogeneity of the Markov chain:

$$\mathbb{E}(d(X_{m+1}, Y_{m+1}) \mid d(X_m, Y_m) = r)$$

$$\leq \underline{r} \left(1 - \frac{1}{N} \left(1 - \frac{3\Delta}{9} \right) \right).$$

Average over r :

$$\sum_r \mathbb{E}(d(X_{m+1}, Y_{m+1}) \mid d(X_m, Y_m) = r) \mathbb{P}(d(X_m, Y_m) = r)$$

$$= \mathbb{E}(d(X_{m+1}, Y_{m+1})) .$$

$$\leq \left(1 - \frac{1}{N} \left(1 - \frac{3\Delta}{9} \right) \right) \underbrace{\sum_r r \mathbb{P}(d(X_m, Y_m) = r)}_{\mathbb{E}(d(X_m, Y_m))}.$$

\Rightarrow

We found:

$$\mathbb{E}(d(x_{m+1}, t_{m+1})) \leq \mathbb{E}(d(x_m, t_m)) \left(1 - \frac{1}{\alpha} \left(1 - \frac{3\beta}{\gamma}\right)\right).$$

:

$$\leq \underbrace{\mathbb{E}(d(x_0, t_0))}_{\text{total # of vertex}} \left(1 - \frac{1}{\alpha} \left(1 - \frac{3\beta}{\gamma}\right)\right)^n$$

$\leq N$

total # of vertex.

$$\Rightarrow \mathbb{E}(d(x_{m+1}, t_{m+1})) \leq N \left(1 - \frac{1}{\alpha} \left(1 - \frac{3\beta}{\gamma}\right)\right)^n.$$

use $1 - x \leq e^{-x}$



QED. \blacksquare