

(supplementary material to lecture slides)

Proof of (1)

• Suppose $A = (A_{ij})_{i,j \in \{1, \dots, n\}}$ where

$$A_{ij} = \Pr[V_{t+1} = j \mid V_t = i], \quad \forall i, j, \forall t.$$

location at time $t+1$
location at time t

• Claim: $\chi_{t+1} = \chi_t \cdot A$

distn' of V_{t+1}
distn' of V_t

• Proof:

$$\begin{aligned}
 \forall j: \chi_{t+1}(j) &= \Pr[V_{t+1} = j] = \sum_{i=1}^n \underbrace{\Pr[V_t = i]}_{\chi_t(i)} \cdot \underbrace{\Pr[V_{t+1} = j \mid V_t = i]}_{A_{ij}} \\
 &= \chi_t \cdot A \quad \square
 \end{aligned}$$

Review: Eigenvalues/Eigenvectors:

- Let A be a square matrix
- Def: λ is an **eigenvalue** of A if $\underline{x} \cdot A = \lambda \cdot \underline{x}$ for some vector $\underline{0} \neq \underline{x} \in \mathbb{R}^n$.
- \underline{x} is called a **left-eigenvector** of A corresponding to eigenvalue λ

Lazy Random Walk on a 4-Cycle:

$$A = \begin{pmatrix} 1/3 & 1/3 & 0 & 1/3 \\ 1/3 & 1/3 & 1/3 & 0 \\ 0 & 1/3 & 1/3 & 1/3 \\ 1/3 & 0 & 1/3 & 1/3 \end{pmatrix}$$

eigenvalues $\lambda_1 = 1$, $\lambda_2 = \lambda_3 = 1/3$, $\lambda_4 = -1/3$

Claim: No matter what x_0 is, $x_t \rightarrow (1/4, 1/4, 1/4, 1/4)$ as $t \rightarrow \infty$.

Proof: • $e_1 = (1/4, 1/4, 1/4, 1/4)$ is a left-eigenvector corr. to λ_1
• By the spectral theorem, because A is symmetric there exist eigenvectors e_2, e_3, e_4 corresponding to $\lambda_2, \lambda_3, \lambda_4$ respectively, so that $\{e_1, e_2, e_3, e_4\}$

forms a basis for \mathbb{R}^4 .

- Hence x_0 can be expressed as a linear combination of e_1, e_2, e_3, e_4 . i.e.

$$x_0 = a_1 \cdot e_1 + a_2 \cdot e_2 + a_3 \cdot e_3 + a_4 \cdot e_4 \quad \text{for some scalars } a_1, a_2, a_3, a_4$$

- $$\begin{aligned} x_t &= x_0 \cdot A^t = \\ &= (a_1 \cdot e_1 + a_2 \cdot e_2 + a_3 \cdot e_3 + a_4 \cdot e_4) \cdot A^t = \\ &= a_1 \cdot e_1 A^t + a_2 \cdot e_2 A^t + a_3 \cdot e_3 A^t + a_4 \cdot e_4 A^t \\ &= a_1 \cdot e_1 A A^{t-1} + a_2 \cdot e_2 A A^{t-1} + \dots \\ &= a_1 \cdot \lambda_1 \cdot e_1 A^{t-1} + a_2 \cdot \lambda_2 \cdot e_2 A^{t-1} + \dots \\ &= \dots \\ &= a_1 \cdot \lambda_1^t \cdot e_1 + a_2 \lambda_2^t \cdot e_2 + a_3 \lambda_3^t \cdot e_3 + a_4 \lambda_4^t \cdot e_4 \\ &\rightarrow a_1 \cdot e_1, \text{ as } t \rightarrow \infty \\ &\quad \text{(because } |\lambda_2|, |\lambda_3|, |\lambda_4| < 1) \end{aligned}$$

- So $x_t \rightarrow \left(\frac{a_1}{4}, \frac{a_1}{4}, \frac{a_1}{4}, \frac{a_1}{4}\right)$ as $t \rightarrow \infty$

- Q: what may a_1 be?

A: The limit must be a distribution. So $a_1 = 1$.

- hence $x_t \rightarrow \left(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}\right)$ as $t \rightarrow \infty$

✗

Note: Above proof works for all lazy random walks on undirected, connected graphs.

Namely:

- i. transition matrix A is symmetric
- ii. it has n real eigenvalues of which $\lambda_1 = 1$ rest satisfy $|\lambda_2|, |\lambda_3|, \dots, |\lambda_n| < 1$.

- iii. because A is symmetric, eigenvectors e_1, e_2, \dots, e_n corresponding to these eigenvalues form a basis
- iv. hence starting distn' x_0 can be written as:

$$x_0 = \sum_{i=1}^n a_i \cdot e_i$$

v. So

$$x_t = x_0 A^t = \sum_{i=1}^n a_i (e_i A^t)$$
$$= \sum_i a_i \lambda_i^t e_i$$

\Rightarrow as $t \rightarrow \infty$, $x_t \rightarrow a_1 \cdot e_1$ (because $\lambda_1 = 1$ and $|\lambda_2|, |\lambda_3|, \dots, |\lambda_n|$ are < 1)

vi. So $x_\infty = a_1 \cdot e_1$ for whatever a_1 makes this a distn'.

Graph Coloring

Input: $G = (V, E)$ graph; $Q = \{1, 2, \dots, q\}$ set of colors

Goal: Sample a u.d.r. legal coloring of G

↳ assign a color to each vertex so that no two adjacent vertices get the same color

Denote by Δ : the maximum degree of G .

Remarks:

- if $q \geq \Delta + 1$: a legal coloring guaranteed to exist

- if $q = \Delta$: Brook's theorem:

◦ if $\Delta \geq 2$ the graph has a Δ -coloring iff it does not contain a $(\Delta + 1)$ -clique

◦ if $\Delta = 2$:

2-coloring exists \Leftrightarrow no odd cycle

- if $q < \Delta$: NP-hard to tell if Δ -coloring exists

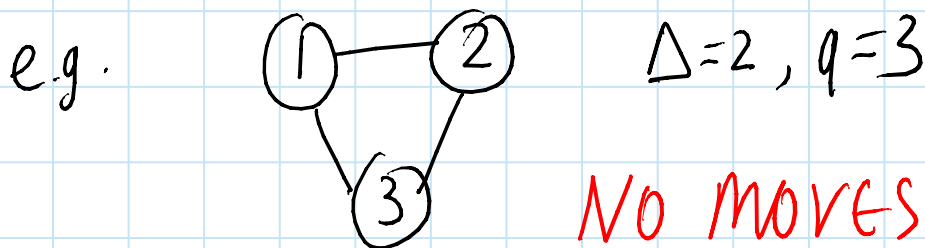
Working Regime for us: $q \geq \Delta + 1$

Markov Chain:

- start at arbitrary legal coloring (can construct it greedily)
- pick vertex $v \in V$ u.a.r. and color $c \in Q$ u.a.r.
- recolor v with c if this gives legal coloring otherwise do nothing
- repeat

Claim: chain is lazy

Claim 2: not necessarily strongly connected



NO MOVES AVAILABLE!

Claim 3: If $q \geq \Delta + 2$ it is strongly connected.

left as exercise

Claim 4: If $q \geq \Delta + 2$, the Markov chain converges to the uniform distn' over colorings.

Proof: • lazyness + strong connectivity \Rightarrow convergence to stationary distn'

• stationary distn' is left-eigenvector of transition matrix corresponding to eigenvalue 1.

• let's inspect A :

$$A = \begin{matrix} & \text{legal colorings} \\ \text{legal colorings} & \begin{pmatrix} & & & \\ & & & \\ \vdots & & & \vdots \\ i & \text{---} & & j \\ & & & \vdots \\ & & & \end{pmatrix} \end{matrix}$$

entry i, j : probability of transitioning from coloring i to coloring j

lemma: $A_{ij} = A_{ji}$

Proof: • If $A_{ij} \neq 0$:

- Suppose $i \rightarrow j$ by changing color of node v from c to c'

happens w/ prob $\frac{1}{n} \times \frac{1}{q}$

choose
node v

choose color c'

- then we can also move from j to i by changing v 's color from c' to c

happens w/ prob $\frac{1}{n} \times \frac{1}{q}$

hence $A_{ij} = A_{ji}$.

• If $A_{ij} = 0$, then can't go from col. i to col. j . So we can't go from col. j to col. i either. Hence $A_{ji} = 0$

□

Lemma \Rightarrow all columns of A sum to 1

↑
follows from the fact that all rows sum to 1

\Rightarrow If x is uniform distn' over legal colorings, then

$$x \cdot A = x$$

□

Thm: If $q \geq 2\Delta + 1$, $\tau_{\text{mix}} = O(n \cdot \log n)$

Conjecture: same is true for $q \geq \Delta + 2$.