

Duality, Evolving Sets and Mixing

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Conductance

- For $A, B \subset \Omega$, ergodic flow from A to B :
 $Q(A, B) := \sum_{x \in A} \sum_{y \in B} Q(x, y)$ with $Q(x, y) = \pi(x)P(x, y)$.
- Conductance

$$\Phi(S) := \frac{Q(S, \bar{S})}{\pi(S)} \text{ and } \Phi_* := \min_{S \subset \Omega} \{\Phi(S) : 0 < \pi(S) \leq 1/2\}$$

Theorem (From Lecture)

$$\tau_{\text{mix}} \geq \frac{1}{4\Phi_*}$$

for any Markov chain

$$\tau_X(\epsilon) \leq \frac{2}{\Phi_*^2} \left(\log \frac{1}{\pi(X)} + \log \frac{1}{\epsilon} \right)$$

for lazy, reversible chains

Main Result [Morris and Peres '03]

Define *conductance profile* as (where $\pi^* = \min_x \pi(x)$)

$$\Phi(r) := \begin{cases} \min_{S \in \Omega} \{\Phi(S) : 0 < \pi(S) \leq r\} & r \in [\pi^*, 1/2] \\ \Phi_* & r > 1/2 \end{cases}$$

and ϵ -uniform mixing time as

$$\tau^\infty(\epsilon) := \min\{n : \forall x, y \in \Omega, \left| \frac{P^n(x, y)}{\pi(y)} - 1 \right| \leq \epsilon\}$$

Theorem (MP03)

For lazy (not necessarily reversible) MC,

$$\tau^\infty(\epsilon) \leq \int_{4\pi_*}^{4/\epsilon} \frac{4du}{u\Phi^2(u)}$$

Since $\Phi(u) \geq \Phi^*$, this is no worse than previous bound.

Profile bound can be better

- For the $n \times n$ 2-D grid, conductance profile satisfies

$$\Phi(u) \geq \frac{a}{n\sqrt{u}}.$$

- Profile bound is bounded by

$$C \log \frac{1}{\epsilon} + 4 \int_{1/n^2}^{1/2} \frac{du}{u \left(\frac{a}{n\sqrt{u}} \right)^2} = C \log \frac{1}{\epsilon} + O(n^2).$$

- Previous bound would have given

$$C' n^2 (2 \log n + \log(1/\epsilon)).$$

Various Mixing Times

- More generally, for $1 \leq p \leq \infty$, define

$$\tau_X^p(\epsilon) := \min\{n : \left\| \frac{P^n(x, \cdot)}{\pi(\cdot)} - \mathbf{1} \right\|_{L_p(\pi)} \leq \epsilon\}$$

- Note that $\tau_X^1(\epsilon) \leq \tau_X^2(\epsilon) \leq \tau_X^\infty(\epsilon)$.
- Define $\tau^p(\epsilon) = \max_{x \in \Omega} \tau_X^p(\epsilon)$.

The Time Reversed Chain

Given P **time-reversed** MC has transition matrix

$$P_R(x, y) := \frac{\pi(y)P(y, x)}{\pi(x)}.$$

It satisfies:

- $\pi(x)P_R^n(x, y) = \pi(y)P^n(y, x)$
- $\left| \frac{P^{(n+m)}(x, y) - \pi(y)}{\pi(y)} \right| \leq \left\| \frac{P^n(x, \cdot)}{\pi(\cdot)} - \mathbf{1} \right\|_{L_2(\pi)} \left\| \frac{P^m(y, \cdot)}{\pi(\cdot)} - \mathbf{1} \right\|_{L_2(\pi)}$

Corollary

$$\tau^\infty(\epsilon) \leq \tau^2(\sqrt{\epsilon}) + \tau_R^2(\sqrt{\epsilon})$$

Reversed chain has same profile

- Note that “conservation of mass” tells us

$$Q(S, \bar{S}) = Q(\bar{S}, S) .$$

- Since P_R also has stationary dist. π , we have

$$\Phi_R(S) := \frac{Q_R(S, \bar{S})}{\pi(S)} = \frac{Q(\bar{S}, S)}{\pi(S)} = \frac{Q(S, \bar{S})}{\pi(S)} =: \Phi(S)$$

- Therefore, it suffices to prove

$$\tau^2(\epsilon) \leq \int_{4\pi_*}^{4/\epsilon^2} \frac{2du}{u\Phi^2(u)} .$$

Notion of a Dual Chain

- If we have another MC with transition matrix P_D on some state space V (we consider $V = 2^\Omega$) such that

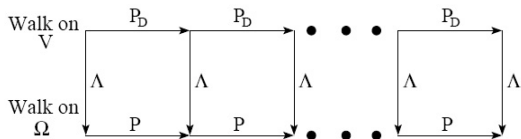
$$\Lambda \times P = P_D \times \Lambda$$

$$|V| \times |\Omega| \quad |\Omega| \times |\Omega| \quad |V| \times |V| \quad |V| \times |\Omega|$$

for some “link” matrix Λ .

- Then the evolution of the two chains will be related:

$$\forall n > 0, \Lambda P^n = P_D^n \Lambda.$$



Evolving Set

Definition

Evolving set process is a MC $\{S_n\}$ defined on 2^Ω as follows. If $S_n = A \subseteq \Omega$, pick $u \sim \text{Unif}[0, 1]$, and set $S_{n+1} = A_u$ where

$$A_u := \{y \in \Omega : Q(A, y) \geq u\pi(y)\}.$$

- Let $K(A, A') := \Pr(S_{n+1} = A' | S_n = A)$ be its transition matrix
- Simple but important observation:

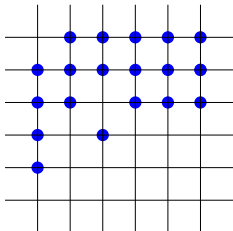
$$\Pr(y \in S_{n+1} | S_n = A) = \Pr(Q(A, y) \geq u\pi(y)) = \frac{Q(A, y)}{\pi(y)}.$$

- \emptyset and Ω are absorbing states of the evolving set process.
- Important martingale property:

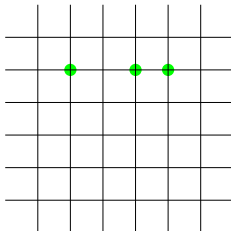
$$\mathbb{E}[\pi(S_{n+1}) | S_n = A] = \sum_{y \in \Omega} \pi(y) \Pr(y \in S_{n+1} | S_n = A) = \sum_y Q(A, y) = \pi(A).$$

Example

A



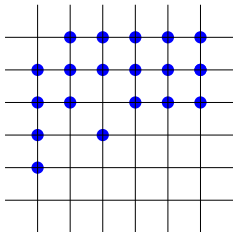
A_u



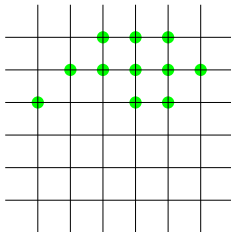
$$u \in \left(\frac{7}{8}, 1\right]$$

Example

A



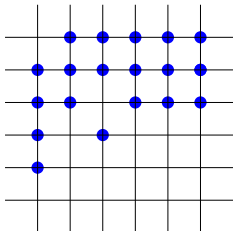
A_u



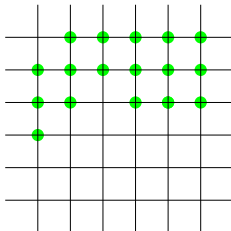
$$u \in \left(\frac{6}{8}, \frac{7}{8}\right]$$

Example

A



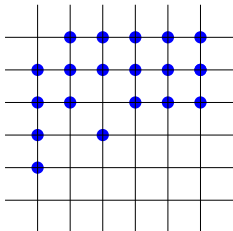
A_u



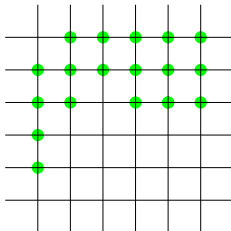
$$u \in \left(\frac{5}{8}, \frac{6}{8}\right]$$

Example

A



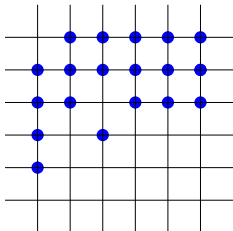
A_u



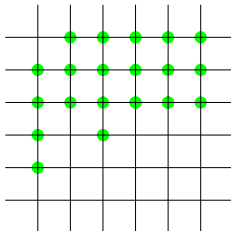
$$u \in \left(\frac{4}{8}, \frac{5}{8}\right]$$

Example

A



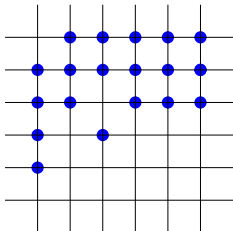
A_u



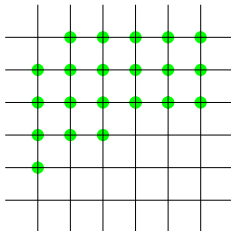
$$u \in \left(\frac{3}{8}, \frac{4}{8}\right]$$

Example

A



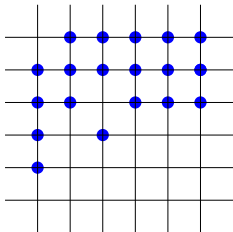
A_u



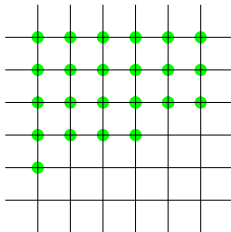
$$u \in \left(\frac{2}{8}, \frac{3}{8}\right]$$

Example

A



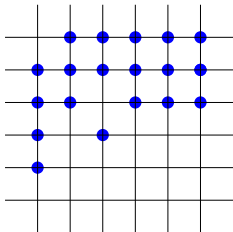
A_u



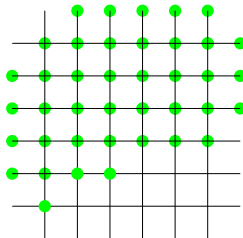
$$u \in \left(\frac{1}{8}, \frac{2}{8}\right]$$

Example

A



A_u



$$u \in (0, \frac{1}{8}]$$

Relating Original MC to evolving set process

Define another MC on sets with transition kernel

$\hat{K}(S, S') = \frac{\pi(S')}{\pi(S)} K(S, S')$. Can show

- \hat{K} is a valid transition matrix.
- $\hat{K}^n(S, S') = \frac{\pi(S')}{\pi(S)} K^n(S, S')$ and so, for any function f , if $S_0 = S$, then expectation $\hat{\mathbb{E}}^n$ under \hat{K}^n is $\hat{\mathbb{E}}_S^n[f(S_n)] = \mathbb{E}_S^n \left[\frac{\pi(S_n)}{\pi(S)} f(S_n) \right]$

Lemma

If $S \subset \Omega$, $y \in \Omega$, and $\Lambda(S, y) = \frac{\pi(y)}{\pi(S)} 1_{(y \in S)}$, $[\Lambda P](S, y) = [\hat{K}\Lambda](S, y)$.

That is, \hat{K} is dual of P .

Relating n-step transition probability to evolving set process

Lemma

If $x \in \Omega$, $S_0 = \{x\}$, $P^n(x, y) = \hat{\mathbb{E}}_{\{x\}}^n[\pi_{S_n}(y)]$ with

$$\pi_S(y) := \frac{\mathbf{1}_{S(y)}\pi(y)}{\pi(S)} = \Lambda(S, y).$$

Proof.

$$P^n(x, y) \stackrel{(1)}{=} [\Lambda P^n](\{x\}, y) \stackrel{(2)}{=} [\hat{K}^n \Lambda](\{x\}, y) \stackrel{(3)}{=} \hat{\mathbb{E}}_{\{x\}}^n(\pi_{S_n}(y))$$

- $\Lambda(\{x\}, y) = \mathbf{1}(x = y) \frac{\pi(y)}{\pi(x)} \Rightarrow (1)$
- Duality $\Rightarrow (2)$
- $[\hat{K}^n \Lambda](\{x\}, y) = \sum_{S' \subset \Omega} \hat{K}^n(\{x\}, S') \Lambda(S', y) = \sum_{S' \subset \Omega} \hat{K}^n(\{x\}, S') \pi_{S'}(y) \Rightarrow (3)$

Relating L^2 distance to evolving sets process

Theorem

If $x \in \Omega$, $S_0 = \{x\}$, then $\|P^n(x, \cdot), \pi\|_2 \leq \hat{\mathbb{E}}_{\{x\}}^n \sqrt{\frac{1 - \pi(S_n)}{\pi(S_n)}}$

Proof. $\|P^n(x, \cdot), \pi\|_2 = \left\| \hat{\mathbb{E}}_{\{x\}}^n \pi_{S_n}, \pi \right\|_2 \leq \hat{\mathbb{E}}_{\{x\}}^n \|\pi_{S_n}, \pi\|_2$

Now, $\|\pi_{S_n}, \pi\|_2 = \sqrt{\sum_{y \in \Omega} \pi(y) \left(\frac{\pi_{S_n}(y)}{\pi(y)} - 1 \right)^2} =$
 $\sqrt{\sum_{y \in \Omega} \left(\frac{\pi_{S_n}^2(y)}{\pi(y)} \right) - 1} = \sqrt{\sum_{y \in S_n} \left(\frac{\pi(y)}{\pi^2(S_n)} \right) - 1} = \sqrt{\frac{1 - \pi(S_n)}{\pi(S_n)}}$

Conductance vs. Congestion

Symmetric Conductance

$\tilde{\Phi}(r) := \min_{S \subset \Omega} \left\{ \tilde{\Phi}(S) : 0 < \pi(S) \leq r \right\}$ with $\tilde{\Phi}(S) = \frac{Q(S, \bar{S})}{\pi(S)\pi(\bar{S})}$
 and $\tilde{\Phi}^* = \min_{S \subset \Omega} \tilde{\Phi}(S)$.

$\tilde{\Phi}^*$ and Φ^* are equal up to a factor of 1/2.

Congestion

$\mathcal{C}(r) := \max_{S \subset \Omega} \left\{ \mathcal{C}(S) : 0 < \pi(S) \leq r \right\}$ with
 $\mathcal{C}(S) = \mathbb{E}_S \left[\sqrt{\frac{(1-\pi(S'))\pi(S')}{(1-\pi(S))\pi(S)}} \right]$ and $\mathcal{C}^* = \max_{S \subset \Omega} \mathcal{C}(S)$.

For lazy MC, $\mathcal{C}(S) \leq \sqrt{1 - \tilde{\Phi}(S)^2}$.

Simple Result

Lemma

$$\hat{\mathbb{E}}_{\{X\}}^{n+1} f(\pi(S_{n+1})) - \hat{\mathbb{E}}_{\{X\}}^n f(\pi(S_n)) = -\hat{\mathbb{E}}_{\{X\}}^n f(\pi(S_n))(1 - C(S_n))$$

with $f(z) = \sqrt{\frac{1-z}{z}}$.

Proof.

$$\begin{aligned} \hat{\mathbb{E}}_{\{X\}}^{n+1} f(\pi(S_{n+1})) &= \hat{\mathbb{E}}_{\{X\}}^n \sum_{S \subset \Omega} \hat{K}(S_n, S) f(\pi(S)) \\ &= \hat{\mathbb{E}}_{\{X\}}^n f(\pi(S_n)) K(S_n, S) \sqrt{\frac{\pi(S)(1-\pi(S))}{\pi(S_n)(1-\pi(S_n))}} \\ &= \hat{\mathbb{E}}_{\{X\}}^n f(\pi(S_n)) C(S_n) \end{aligned}$$

Since $C^* \geq C(S) \forall S \subset \Omega$,

$$\hat{\mathbb{E}}_{\{X\}}^{n+1} f(\pi(S_{n+1})) \leq C^* \hat{\mathbb{E}}_{\{X\}}^n f(\pi(S_n)).$$

Simple bound on mixing time

- $\|P^n(x, \cdot), \pi\|_2 \leq \hat{\mathbb{E}}_{\{x\}}^n \sqrt{\frac{1 - \pi(S_n)}{\pi(S_n)}} \leq (C^*)^n f(\pi(S_0)) = (C^*)^n f(\pi(x)).$
- Solving for when RHS drops to ϵ and using $\log C^* \leq -(1 - C^*)$ gives

$$\begin{aligned} \tau^2(\epsilon) &\leq \left\lceil \frac{1}{1 - C^*} \log \frac{f(\pi_*)}{\epsilon} \right\rceil \\ &\leq \left\lceil \frac{1}{1 - C^*} \log \frac{1}{\epsilon \sqrt{\pi_*}} \right\rceil \\ &\leq \left\lceil \frac{2}{(\tilde{\Phi}^*)^2} \log \frac{1}{\epsilon \sqrt{\pi_*}} \right\rceil \end{aligned}$$

using $C^* \leq \sqrt{1 - (\tilde{\Phi}^*)^2}$

Improved bound on mixing time

- Before we used $\mathcal{C}^* \geq \mathcal{C}(S), \forall S \subset \Omega$
- Infact, we can do better by using $\mathcal{C}(\pi(S)) \geq \mathcal{C}(S), \forall S \subset \Omega$
- Exploiting this fact and using *a bit of analysis*,

$$\begin{aligned}\tau^2(\epsilon) &\leq \left[\int_{4\pi_*}^{4/\epsilon^2} \frac{du}{u(1 - \mathcal{C}(u))} \right] \\ &\leq \left[\int_{4\pi_*}^{4/\epsilon^2} \frac{du}{u\tilde{\Phi}^2(u)} \right]\end{aligned}$$

using $\mathcal{C}(u) \leq \sqrt{1 - \tilde{\Phi}(u)^2}$

Remarks

- Evolving set L^2 bounds are at least as good as bounds discussed in class.
- Conductance bounds apply to non reversible chains.
- Conductance is inappropriate for non-lazy chains. In particular, conductance does not distinguish between periodic and aperiodic chains.