

# Spectral Gap & Mixing TIMES

- $M = (P, \pi, \Omega)$  ergodic, reversible Markov chain.
- Eigenvalues of  $P$ :  $1 = \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_{|\Omega|} \geq -1$ .
- Spectral Gap:  $\lambda(P) = 1 - \max_{i \geq 2} |\lambda_i|$
- If all eigenvalues non-negative, then  $\lambda(P) = 1 - \lambda_2$ .

## Thm 1

$$(\lambda(P)^{-1} - 1) \log 2 \leq T_{\text{mix}}(M) \leq \lambda(P)^{-1} \cdot \log \left( \frac{4}{\pi_{\min}} \right)$$

$$\pi_{\min} = \min_{x \in \Omega} \pi(x).$$

For simplicity, we shall assume that  $M$  is lazy,

$$\text{i.e., } P = \frac{I + \hat{P}}{2}$$

and show instead:

Thm 1.1  $T_{\text{mix}}(M) \leq \lambda(P)^{-1} \log \left( \frac{16}{\pi_{\min}} \right)$

\* Recall if  $P$  is lazy, then all its eigenvalues are non-negative.

## Some definitions:

Take  $\pi \in \mathbb{R}^{|\mathcal{S}|}$  [or  $f: \mathcal{U} \rightarrow \mathbb{R}$ ].

$$\Rightarrow E_\pi[f] = \sum_{x \in \mathcal{S}} \pi(x) \cdot f(x).$$

$$\Rightarrow \text{Var}_\pi[f] = E_\pi[(f - E_\pi[f])^2]$$

$$= \frac{1}{2} \sum_{x,y \in \mathcal{S}} \pi(x)\pi(y) (f(x) - f(y))^2 \quad [\text{Ex. check!}]$$

The Dirichlet form: ("local" variance).

$$\mathcal{E}_\pi(f, f) = \frac{1}{2} \sum_{x,y \in \mathcal{S}} \pi(x) \cdot P(x,y) (f(x) - f(y))^2.$$

Lemma 2 If  $\exists \alpha > 0$  s.t.

$$\forall f \in \mathbb{R}^{|\mathcal{S}|} \quad \alpha \cdot \mathcal{E}_\pi(f, f) \geq \text{Var}_\pi(f),$$

then:

$$T_{\min} \leq \alpha \cdot \log \frac{16}{T_{\min}}.$$

What is the optimal  $\lambda$ ?

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Lemma 3 (Courant-Fischer Thm)

P ergodic and reversible w.r.t  $\pi$ . Then

$$1 - \lambda_2 = \min_{f: \text{Var}_\pi(f) \neq 0} \frac{\mathbb{E}_\pi(f, f)}{\text{Var}_\pi(f)}.$$

- Since P is lazy, all its eigenvalues are positive and so  $\lambda(P) = 1 - \lambda_2$ .
- Then  $\lambda_{\text{optimal}} = \lambda(P)^{-1}$ , and

$$T_{\text{mix}} \leq \lambda(P)^{-1} \cdot \log\left(\frac{16}{\epsilon}\right),$$

as claimed

Note: If we consider the inner product

$$\langle f, g \rangle_\pi = \sum_{x \in \Omega} f(x) \cdot g(x) \cdot \pi(x).$$

Then, assuming  $E\pi(f) = 0$  w.l.o.g, we get:

$$\frac{\mathbb{E}_\pi(f, f)}{\text{Var}_\pi(f)} = \frac{\langle f, (\mathbb{I} - P)f \rangle_\pi}{\langle f, f \rangle_\pi} \quad [\text{check!}]$$

Proof of Lemma 2:

$$\text{Var}_{\pi}(P_f) \leq \text{Var}_{\pi}(f) - \varepsilon_{\pi}(f, f) \quad (*)$$

Before proving  $(*)$ , let us use it to prove Lemma 2.

Since,  $\varepsilon_{\pi}(f, f) \geq \frac{1}{\alpha} \cdot \text{Var}_{\pi}(f)$  by assumption,

$$\begin{aligned} \text{Var}_{\pi}(P_f) &\leq \text{Var}_{\pi}(f) - \frac{1}{\alpha} \text{Var}_{\pi}(f) \\ &\leq \left(1 - \frac{1}{\alpha}\right) \text{Var}_{\pi}(f) \end{aligned}$$

Inductively:

$$\text{Var}_{\pi}(P^T f) \leq \left(1 - \frac{1}{\alpha}\right)^T \cdot \text{Var}_{\pi}(f) \quad \forall f \in \mathbb{R}^{|\Omega|}.$$

Now, let

$$f_A(x) = \begin{cases} 1 & x \in A \\ 0 & \text{o.w.} \end{cases}$$

↳ indicator for the set  $A$ .

Then:  $\text{Var}_{\pi}(f_A) \leq E_{\pi}[f_A^2] \leq 1$ , and.

$$\text{Var}(P^T f_A) \leq \left(1 - \frac{1}{\alpha}\right)^T \leq e^{-T/\alpha}.$$

Suppose  $x \in \Omega$  is the initial state of the chain  
and take  $T = \alpha \log\left(\frac{16}{\pi(x)}\right)$ .

Then,

$$\frac{\pi(x)}{16} \geq \text{Var}_{\pi}(P^T f_A) \geq \pi(x) \left( P^T f_A(x) - E_{\pi}[P^T f_A] \right)^2$$

$$\frac{1}{4} \geq |P^T f_A(x) - E_{\pi}[P^T f_A]| \quad \textcircled{*} \textcircled{*}$$

Now, using the fact that  $f_A$  is the indicator for set  $A$   
we get:

$$(a) P^T f_A(x) = \sum_{y \in \mathcal{L}} P^T(x, y) \cdot f_A(y) = \sum_{y \in A} P^T(x, y) = P^T(x, A),$$

$$(b) E_{\pi}[P^T f_A] = E_{\pi}[f_A] = \sum_{x \in \mathcal{L}} \pi(x) \cdot f_A(x) = \pi(A).$$

$\hookrightarrow [\text{Ex. check!}]$ .

Plugging (a) and (b) into  $\textcircled{*} \textcircled{*}$ , we get

$$\frac{1}{4} \geq |P^T(x, A) - \pi(A)|.$$

Since this holds  $\forall x \in \mathcal{L}$  and  $A \subseteq \mathcal{L}$ ,

$$\frac{1}{4} \geq \|P^T(\cdot, \cdot) - \pi(\cdot)\|_{\text{TV}}$$

and so

$$T_{\min} \leq t = \alpha \cdot \log \left( \frac{16}{\pi_{\min}} \right).$$

We still need to show inequality  $\circledast$ :

$$\text{Var}_{\pi}(P_f) \leq \text{Var}_{\pi}(f) - \mathbb{E}_{\pi}(f_f f).$$

• Recall that  $P$  is lazy: Namely,  $P = \frac{I + \hat{P}}{2}$ .

Let us assume w.l.o.g. that  $\mathbb{E}_{\pi}[f] = 0$ .

(Note that adding a constant to  $f$  does not change any of the terms in  $\circledast$ ).

Then,

$$\begin{aligned} \text{Var}_{\pi}(P_f) &= \sum_{x \in \Omega} [P_f(x)]^2 \pi(x) = \sum_{x \in \Omega} \pi(x) \left[ \sum_{y \in \Omega} P(x,y) \cdot f(y) \right]^2 \\ &= \sum_{x \in \Omega} \pi(x) \left[ \sum_{y \in \Omega} \frac{I + \hat{P}}{2}(x,y) \cdot f(y) \right]^2 \\ &= \frac{1}{4} \sum_{x \in \Omega} \pi(x) \left[ f(x) + \sum_{y \in \Omega} P(x,y) \cdot f(y) \right]^2 \\ &= \frac{1}{4} \sum_{x \in \Omega} \pi(x) \left[ \cancel{\sum_{z \in \Omega}} \sum_{y \in \Omega} P(x,y) (f(x) + f(y)) \right]^2. \end{aligned}$$

$$\text{Var}_{\pi}(P_f) \leq \frac{1}{4} \sum_{x \in \Omega} \pi(x) \left( \sum_{z \in \Omega} P(x,z) \right) \left( \sum_{y \in \Omega} P(x,y) (f(x) + f(y))^2 \right)$$

↳ (Cauchy-Schwarz inequality).

$$\text{Var}_{\pi}(P_f) \leq \frac{1}{4} \sum_{x,y \in \Omega} \pi(x) \cdot P(x,y) (f(x) + f(y))^2 \quad (1)$$

Now,

$$\text{Var}_{\pi}(f) = \sum_{x \in \Omega} \pi(x) f(x)^2 = \frac{1}{2} \sum_{x \in \Omega} \pi(x) \cdot f(x)^2 + \frac{1}{2} \sum_{y \in \Omega} \pi(y) \cdot f(y)^2$$

$$= \frac{1}{2} \sum_{x,y \in \Omega} \pi(x) \cdot f(x)^2 \hat{P}(x,y) + \frac{1}{2} \sum_{x,y \in \Omega} \pi(y) \cdot f(y)^2 \hat{P}(y,x)$$

$$= \frac{1}{2} \sum_{x,y \in \Omega} \pi(x) \hat{P}(x,y) (f(x)^2 + f(y)^2). \quad (2)$$

Then, (2) - (1) gives:

$$\begin{aligned} \text{Var}_{\pi}(f) - \text{Var}_{\pi}(P_f) &\geq \frac{1}{4} \sum_{x,y \in \Omega} \pi(x) \hat{P}(x,y) [2(f(x)^2 + f(y)^2) - (f(x) + f(y))^2] \\ &\geq \frac{1}{4} \sum_{x,y \in \Omega} \pi(x) \hat{P}(x,y) (f(x) - f(y))^2 \\ &= \frac{1}{4} \sum_{x,y \in \Omega} \pi(x) (2P-I)(x,y) (f(x) - f(y))^2 \\ &= \frac{1}{2} \sum_{x,y \in \Omega} \pi(x) P(x,y) (f(x) - f(y))^2 - \frac{1}{4} \sum_{x,y \in \Omega} \pi(x) I(x,y) (f(x) - f(y))^2 \\ &= \mathcal{E}_{\pi}(f, f) \end{aligned}$$

as desired

## Proof of Lower Bound :

- Let  $\{f_1, \dots, f_{n+1}\}$  be the eigenvectors of  $\hat{P}$ .
- Consider the inner product:

$$\langle f, g \rangle_{\Pi} = \sum_{x \in \mathcal{X}} f(x) \cdot g(x) \cdot \Pi(x) \quad \forall f, g \in \mathbb{R}^{n+1}$$

- Consider the inner product space.

$$L_2(\Pi) = (\mathbb{R}^{n+1}, \langle \cdot, \cdot \rangle_{\Pi})$$

Claim 21  $\{f_j\}_{j=1}^{n+1}$  is an orthonormal basis of  $L_2(\Pi)$ .

→ [Direct consequences of spectral thm for symmetric matrices. Check!!]

Orthonormal basis:

$$\cdot \langle f_j, f_i \rangle = 0 \quad \forall j \neq i \quad [\text{Orthogonality}]$$

$$\cdot \langle f_i, f_i \rangle = 1 \quad \forall i \quad [\text{Normality}]$$

$$\cdot \forall g \in \mathbb{R}^{n+1}: g = \sum_{j=1}^{n+1} a_j \cdot f_j \quad [\text{Basis}]$$

$$a_j = \langle g, f_j \rangle_{\Pi}.$$

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Lower bound :

$$(\lambda(\hat{P})^{-1} - 1) \log 2 \leq T_{\text{mix}}.$$

Note here:  $\sum_q \pi(q) \cdot f_J^T(q) = \langle \vec{f}_J^T, \vec{1} \rangle_\pi = 0. \quad \begin{cases} \text{Since } f_J \\ \text{orthogonal} \\ \text{to } \vec{1} \end{cases}$

Then,

$$\begin{aligned} |\lambda_J^T \cdot f_J(x)| &= |\hat{P}^T f_J(x)| = \left| \sum_y P^T(y, x) \cdot f_J^T(y) \right| \\ &= \left| \sum_y P^T(y, x) \cdot f_J^T(y) - \sum_y \pi(y) \cdot f_J^T(y) \right| \\ &\leq \sum_y |P^T(y, x) - \pi(y)| \cdot \max_{y' \in S} |f_J^T(y')|. \\ &\geq 2 \|P^T(x, \cdot) - \pi(\cdot)\|_{TV} \cdot \|f_J\|_\infty \end{aligned}$$

Taking the  $x$  that maximizes L.H.S.

$$|\lambda_J|^T \leq 2 \|P^T(x, \cdot) - \pi(\cdot)\|_{TV}$$

$$|\lambda_J|^{T_{\text{mix}}} \leq \frac{1}{2} \Rightarrow T_{\text{mix}} \log \left( \frac{1}{|\lambda_J|} \right) \geq \log 2.$$

$$T_{\text{mix}} \left( \frac{1}{|\lambda_J|} - 1 \right) \geq T_{\text{mix}} \log \frac{1}{|\lambda_J|} \geq \log 2.$$

$$T_{\text{mix}} \left[ \frac{\lambda(P)}{1-\lambda(P)} \right] \geq \log 2$$

$$\begin{aligned} T_{\text{mix}} &\geq \lambda(P)^{-1} (1-\lambda(P)) \log 2. \\ &\geq (\lambda(P)^{-1} - 1) \log 2 \end{aligned}$$