

Consider  $X$  an irreducible Markov chain on a finite state space  $E$  of size  $n$ , and denote by  $\pi$  the unique invariant probability of  $X$ .

## Definition (3.2)

For  $t \in \mathbb{N}$  define

$$d(t) := \max_{x \in E} \|P^t(x, \cdot) - \pi\|_{TV}$$

$$\bar{d}(t) := \max_{x, y \in E} \|P^t(x, \cdot) - P^t(y, \cdot)\|_{TV}.$$

The mixing time at level  $\varepsilon$  is defined as

$$t_{\text{mix}}(\varepsilon) = \inf\{t \in \mathbb{N} : d(t) \leq \varepsilon\}.$$

We'll have a particular interest in

$$t_{\text{mix}} := t_{\text{mix}}(1/4).$$

### Theorem (3.2.1)

$$d(t) \leq \bar{d}(t) \leq 2d(t).$$

Proof : The second inequality follows trivially from the triangle inequality. As for the first, since  $\pi$  is invariant,

$$\begin{aligned} d(t) &= \max_{x \in E} \max_{A \subseteq E} |P^t(x, A) - \pi(A)| \\ &= \max_{x \in E} \max_{A \subseteq E} \left| \sum_{y \in E} \pi(y) (P^t(x, A) - P^t(y, A)) \right| \\ &\leq \max_{x \in E} \max_{A \subseteq E} \sum_{y \in E} \pi(y) |P^t(x, A) - P^t(y, A)| \leq \bar{d}(t) \end{aligned}$$

### Theorem (3.2.2)

$\bar{d}$  is submultiplicative, that is, for any  $s, t \in \mathbb{N}$ ,

$$\bar{d}(t+s) \leq \bar{d}(t)\bar{d}(s).$$

Proof : Fix  $x, y \in E$  and assume  $(X_s, Y_s)$  realize an optimal coupling of  $P^s(x, \cdot)$ ,  $P^s(y, \cdot)$  and denote by  $\mathbb{P}_{x,y}$  their joint law (so under  $\mathbb{P}_{x,y}$ ,  $P^t(X_s, \cdot)$  has law  $P^{t+s}(x, \cdot)$ ,  $P^t(Y_s, \cdot)$  has law  $P^{t+s}(y, \cdot)$ , and  $\mathbb{P}_{x,y}(X_s \neq Y_s) = \bar{d}(s)$ ). Then

$$\begin{aligned} & \|P^{t+s}(x, \cdot) - P^{t+s}(y, \cdot)\|_{TV} \\ &= \frac{1}{2} \sum_{w \in E} |\mathbb{E}_{x,y} [P^t(X_s, w) - P^t(Y_s, w)]| \\ &\leq \frac{1}{2} \sum_{w \in E} \mathbb{E}_{x,y} [ |P^t(X_s, w) - P^t(Y_s, w)| \mathbf{1}_{\{X_s \neq Y_s\}} ] \\ &\leq \bar{d}(t) \mathbb{P}_{x,y}(X_s \neq Y_s) = \bar{d}(t)\bar{d}(s), \end{aligned}$$

where for the inequality at the last line, we used from its definition that  $\bar{d}(t) \geq \frac{1}{2} \sum_{w \in E} |P^t(X_s, w) - P^t(Y_s, w)|$ .

### Theorem (3.2.3)

For any  $\varepsilon \leq 1/2$ , we have

$$t_{\text{mix}}(\varepsilon) \leq \lceil \log_2 \left( \frac{1}{\varepsilon} \right) \rceil t_{\text{mix}}.$$

Proof : By the above results, for  $\ell \in \mathbb{N}^*$ ,

$$\begin{aligned} d(\ell t_{\text{mix}}) &\leq \bar{d}(\ell t_{\text{mix}}) \\ &\leq \bar{d}(t_{\text{mix}})^\ell \\ &\leq (2d(t_{\text{mix}}))^\ell \leq \frac{1}{2^\ell}, \end{aligned}$$

thus if  $\ell \geq \log_2 \left( \frac{1}{\varepsilon} \right)$ ,  $d(\ell t_{\text{mix}}) \leq \varepsilon$ , as required.