

"Markov Processes", Problem Sheet 9

Please hand in your solutions before 12 noon on Monday, December 9.

1. (Total variation distances).

- a) Let $\nu = \bigotimes_{i=1}^{d} \nu_i$ and $\mu = \bigotimes_{i=1}^{d} \mu_i$ be two product probability measures on S^d . Show in at least two different ways that $\|\nu \mu\|_{TV} \leq \sum_{i=1}^{d} \|\nu_i \mu_i\|_{TV}$.
- b) Show that the total variation distance of the law of a Markov chain to an invariant probability measure is a non-increasing function of time.
- c) Do similar statements as in a) and b) hold when the total variation distance is replaced by a general transportation metric \mathcal{W}^1 ?

Definition 1 Let $\epsilon \in (0, 1)$. The ϵ -mixing time of a Markov semigroup (p_t) with invariant probability measure μ is defined as $t_{mix}(\epsilon) := \inf \{t \ge 0 : \|p_t(x, \cdot) - \mu\|_{TV} \le \epsilon \forall x\}.$

2. (Hard core model). Consider a finite graph (V, E) with *n* vertices of maximal degree Δ . The corresponding hard core model with fugacity $\lambda > 0$ is the probability measure μ_{λ} on $\{0, 1\}^V$ with mass function

$$\mu_{\lambda}(\eta) = \frac{1}{Z(\lambda)} \lambda^{\sum_{x \in V} \eta(x)} \quad \text{if } \eta(x) \cdot \eta(y) = 0 \text{ for any } (x, y) \in E, \quad \mu_{\lambda}(\eta) = 0 \quad \text{otherwise,}$$

where $Z(\lambda)$ is a normalization constant.

- a) Describe the transition rule for the Gibbs sampler with equilibrium μ_{λ} , and determine the transition kernel π .
- b) Prove that for $\lambda < (\Delta 1)^{-1}$ and $t \in \mathbb{N}$,

$$\mathcal{W}^{1}(\nu\pi^{t},\mu) \leq \alpha(n,\Delta)^{t} \mathcal{W}^{1}(\nu,\mu) \leq \exp\left(-\frac{t}{n}\left(\frac{1-\lambda(\Delta-1)}{1+\lambda}\right)\right) \mathcal{W}^{1}(\nu,\mu),$$

where $\alpha(n, \Delta) = 1 - \frac{1}{n} \left(\frac{1 - \lambda(\Delta - 1)}{1 + \lambda} \right)$, and \mathcal{W}^1 is the transportation metric based on the Hamming distance on $\{0, 1\}^V$.

c) Show that in this case, the ϵ -mixing time is of order $O(n \log n)$ for any $\epsilon \in (0, 1)$.

3. (Bounds for ergodic averages in the non-stationary case). Let (X_n, P_x) be a Markov chain with transition kernel π and invariant probability measure μ , and let

$$A_{b,n}f = \frac{1}{n} \sum_{i=b}^{b+n-1} f(X_i).$$

Assume that there are a distance d on the state space S, and constants $\alpha \in (0,1)$ and $\bar{\sigma} \in \mathbb{R}_+$ such that

- (A1) $\mathcal{W}_d^1(\nu\pi, \tilde{\nu}\pi) \leq \alpha \mathcal{W}_d^1(\nu, \tilde{\nu}) \quad \forall \nu, \tilde{\nu} \in \mathcal{P}(S), \text{ and}$
- (A2) $\operatorname{Var}_{\pi(x,\cdot)}(f) \leq \bar{\sigma}^2 \|f\|_{Lip(d)}^2 \quad \forall x \in S, \ f: S \to \mathbb{R}$ Lipschitz.

Prove that under these assumptions the following bounds hold for any $b, n, k \ge 0, x \in S$, and for any Lipschitz continuous function $f: S \to \mathbb{R}$:

- a) $\operatorname{Var}_{x} [f(X_{n})] \leq \sum_{k=0}^{n-1} \alpha^{2k} \bar{\sigma}^{2} ||f||^{2}_{Lip(d)}$.
- b) $|\operatorname{Cov}_x[f(X_n), f(X_{n+k})]| = |\operatorname{Cov}_x[f(X_n), (\pi^k f)(X_n)]| \le \frac{\alpha^k}{1-\alpha^2}\bar{\sigma}^2||f||^2_{Lip(d)}.$
- c) $\operatorname{Var}_{x}[A_{b,n}f] \leq \frac{1}{n} \frac{\bar{\sigma}^{2}}{(1-\alpha)^{2}} \|f\|_{Lip(d)}^{2}.$
- d) $|E_x[A_{b,n}f] \mu(f)| \leq \frac{1}{n} \frac{\alpha^b}{1-\alpha} \int d(x,y) \, \mu(dy) \, ||f||_{Lip(d)}.$
- e) $E_x \left[|A_{b,n}f \mu(f)|^2 \right] \leq \frac{1}{n} \frac{1}{(1-\alpha)^2} \left(\bar{\sigma}^2 + \frac{1}{n} \alpha^{2b} (\int d(x,y) \, \mu(dy))^2 \right) \|f\|_{Lip(d)}^2.$