

“Markov Processes”, Problem Sheet 10.

Hand in solutions before Monday 9.1., 2 pm

*We wish you a merry Christmas
and a happy new year!*



1. (Gibbs sampler for the Ising model). Consider a finite graph (V, E) with n vertices of maximal degree Δ . The Ising model with inverse temperature $\beta \geq 0$ is the probability measure μ_β on $\{-1, 1\}^V$ with mass function

$$\mu_\beta(\eta) = \frac{1}{Z(\beta)} \exp \left(\beta \sum_{\{x,y\} \in E} \eta(x)\eta(y) \right),$$

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

- Show that given $\eta(y)$ for $y \neq x$, $\eta(x) = \pm 1$ with probability $(1 \pm \tanh(\beta m(x, \eta)))/2$, where $m(x, \eta) := \sum_{y \sim x} \eta(y)$ is the local magnetization in the neighbourhood of x . Hence determine the transition kernel π for the Gibbs sampler with equilibrium μ_β .
- Prove that for any $t \in \mathbb{N}$,

$$\mathcal{W}^1(\nu \pi^t, \mu_\beta) \leq \alpha(n, \beta, \Delta)^t \mathcal{W}^1(\nu, \mu_\beta) \leq \exp \left(-\frac{t}{n} (1 - \Delta \tanh(\beta)) \right) \mathcal{W}^1(\nu, \mu_\beta),$$

where $\alpha(n, \beta, \Delta) = 1 - (1 - \Delta \tanh(\beta))/n$, and \mathcal{W}^1 is the transportation metric based on the Hamming distance on $\{-1, 1\}^V$. Conclude that for $\Delta \tanh \beta < 1$, the Gibbs sampler is geometrically ergodic with a rate of order $\Omega(1/n)$. *Hint: You may use the inequality*

$$|\tanh(y + \beta) - \tanh(y - \beta)| \leq 2 \tanh(\beta) \quad \text{for any } \beta \geq 0 \text{ and } y \in \mathbb{R}.$$

- The *mean-field Ising model* with parameter $\alpha \geq 0$ is the Ising model on the complete graph over $V = \{1, \dots, n\}$ with inverse temperature $\beta = \alpha/n$. Show that for $\alpha < 1$, the ϵ -mixing time for the Gibbs sampler on the mean field Ising model is of order $O(n \log n)$ for any $\epsilon \in (0, 1)$.

2. (Successful couplings and TV-convergence to equilibrium). Consider a time-homogeneous Markov process with transition function (p_t) and invariant probability measure μ . A coupling (X_t, Y_t) of versions of the process with initial distributions ν and μ , respectively, is called *successful* if the coupling time

$$T = \inf \{t \geq 0 : X_s = Y_s \text{ for any } s \geq t\}$$

is almost surely finite. Show that a successful coupling exists if and only if $\|\nu p_t - \mu\|_{TV} \rightarrow 0$ as $t \rightarrow \infty$.

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

$$(A1) \quad \mathcal{W}_d^1(\nu\pi, \tilde{\nu}\pi) \leq \alpha \mathcal{W}_d^1(\nu, \tilde{\nu}) \quad \forall \nu, \tilde{\nu} \in \mathcal{P}(S), \text{ and}$$

$$(A2) \quad \text{Var}_{\pi(x, \cdot)}(f) \leq \bar{\sigma}^2 \|f\|_{Lip(d)}^2 \quad \forall x \in S, f : S \rightarrow \mathbb{R} \text{ Lipschitz.}$$

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

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

4. (Equivalent descriptions for weighted total variation norms). Let $V : S \rightarrow (0, \infty)$ be a measurable function, and let $d_V(x, y) := (V(x) + V(y))1_{x \neq y}$. Show that the following identities hold for probability measures μ, ν on (S, \mathcal{B}) :

$$\begin{aligned} \|\mu - \nu\|_V &= \left\| \frac{d\mu}{d\lambda} - \frac{d\nu}{d\lambda} \right\|_{L^1(V \cdot \lambda)} \\ &= \sup \{ |\mu(f) - \nu(f)| : f \in \mathcal{F}(S) \text{ s.t. } |f| \leq V \} \\ &= \sup \{ |\mu(f) - \nu(f)| : f \in \mathcal{F}(S) \text{ s.t. } |f(x) - f(y)| \leq d_V(x, y) \forall x, y \} \\ &= \inf \{ E[d_V(X, Y)] : X \sim \mu, Y \sim \nu \} \end{aligned}$$