

## TOPICS IN PROBABILITY. PART III: PHASE TRANSITION

### EXERCISE SHEET 11: SHARP PHASE TRANSITION AND HYPERCONTRACTIVITY

**Exercise 1** (Russo's almost 0 – 1 law). Consider a product measure  $\mathbb{P}_p$  on  $\{0, 1\}^n$ : each 1 is assigned weight  $p$ . Prove the following result: for any  $\varepsilon > 0$ , there exists  $\delta > 0$  such that for any increasing event  $A$  with the associated  $i$ th influence satisfying  $I_i^p(A) (= \mathbb{P}_p[\mathbf{1}_A(\bar{X}) \neq \mathbf{1}_A(\bar{X}^{(i)})]) \leq \delta$  for all  $i, p$ , there exists  $p_c$  such that for all  $p \geq p_c + \varepsilon$ ,  $\mathbb{P}_p[A] \geq 1 - \varepsilon$ , and for all  $p \leq p_c - \varepsilon$ ,  $\mathbb{P}_p[A] \leq \varepsilon$ .

*Proof.* Let  $\varepsilon > 0$ , and  $\delta = \delta(\varepsilon) > 0$  (to be specified later). Let  $A$  be an arbitrary increasing event such that  $I_i^p(A) \leq \delta$  for all  $i, p$ . We choose  $p_c$  (depending on  $A$ ) such that  $\mathbb{P}_{p_c}[A] = 1/2$ . Note that the latter  $p_c$  indeed exists and is in  $(0, 1)$  since  $p \mapsto \mathbb{P}_p[A]$  is increasing with  $\{\mathbb{P}_0[A], \mathbb{P}_1[A]\} = \{0, 1\}$ . Consider  $\varepsilon_0 := \min(1 - p_c, p_c)/2$ . By monotonicity, it suffices to prove that  $\mathbb{P}_{p_c + \varepsilon \wedge \varepsilon_0}[A] \geq 1 - \varepsilon \wedge \varepsilon_0$  and  $\mathbb{P}_{p_c - \varepsilon \wedge \varepsilon_0}[A] \leq \varepsilon \wedge \varepsilon_0$ . By Talagrand's inequality, for an appropriate absolute constant  $c > 0$ ,

$$\begin{aligned} \text{Var}_p[\mathbf{1}_A] &\leq c \log\left(\frac{1}{p(1-p)}\right) \sum_{i=1}^n \frac{I_i^p(A)}{\log(1/I_i^p(A))} \\ &\leq c \log\left(\frac{1}{p(1-p)}\right) \frac{1}{\log(1/\delta)} \sum_{i=1}^n I_i^p(A) \leq c(p_c) \frac{1}{\log(1/\delta)} \sum_{i=1}^n I_i^p(A) \end{aligned}$$

for all  $p \in [p_c - \varepsilon_0, p_c + \varepsilon_0]$ . By Lemma 2.3,

$$\mathbb{P}_{p_c + \varepsilon \wedge \varepsilon_0}[A] \geq 1 - \exp\left(-\log(1/\delta) \frac{\varepsilon \wedge \varepsilon_0}{c(p_c)}\right) = 1 - \delta^{\frac{\varepsilon \wedge \varepsilon_0}{c(p_c)}};$$

$$\mathbb{P}_{p_c - \varepsilon \wedge \varepsilon_0}[A] \leq \delta^{\frac{\varepsilon \wedge \varepsilon_0}{c(p_c)}}.$$

Choose  $\delta = (\varepsilon \wedge \varepsilon_0)^{c(p_c)/\varepsilon \wedge \varepsilon_0}$ , then as desired for all  $p \geq p_c + \varepsilon$ ,  $\mathbb{P}_p[A] \geq 1 - \varepsilon \wedge \varepsilon_0$  and for all  $p \leq p_c - \varepsilon$ ,  $\mathbb{P}_p[A] \leq \varepsilon \wedge \varepsilon_0$ .  $\square$

**Exercise 2** (Heat semigroup on the hypercube).

Let  $\mu = \mu_1 \otimes \dots \otimes \mu_n := (\frac{1}{2}\delta_{-1} + \frac{1}{2}\delta_1)^{\otimes n}$  be the uniform measure on the hypercube  $\{-1, 1\}^n$ . Define the process  $X_t = (X_t^1, \dots, X_t^n)_{t \geq 0}$  as follows. To each coordinate  $i$ , we attach an independent Poisson process  $(N_t^i)_{t \geq 0}$  of intensity one<sup>1</sup>. Then,

- draw  $X_0 \sim \mu$  independently from the PP  $N = (N^1, \dots, N^n)$ ;
- each time  $N_t^i$  jumps for some  $i$ , replace the current value of  $X_t^i$  by an independent sample from  $\mu_i$  while keeping the remaining coordinates fixed.

We define  $P_t f(x) := \mathbb{E}[f(X_t) | X_0 = x]$ . Prove the following properties of  $(X_t)_t$  and its associated semigroup  $(P_t)_t$ :

<sup>1</sup>Poisson process  $(N_t)_{t \geq 0}$  of intensity  $\lambda > 0$  on  $\mathbb{R}_+$  is the counting process that satisfies  $N_0 = 0$ , has independent increments and  $N_t - N_s \sim \text{Poisson}(\lambda(t-s))$  for any  $t \geq s \geq 0$ . Recall that such a process can be constructed, for instance, in the following way: let  $(T_i)_i$  be i.i.d. exponential random variables of intensity  $\lambda$ , then  $N_t := \#\{n \in \mathbb{N}_0 : \sum_{i=1}^n T_i \leq t\}$  is the Poisson process of intensity  $\lambda$ .

- (1)  $(X_t)_t$  satisfies the Markov property;  
(2)  $\mu$  is its stationary measure, i.e., that  $\int \mathbb{E}[f(X_t)|X_0 = x]\mu(dx) = \int f\mu(dx)$ ;  
(3)  $P_t f(x) = \sum_{S \subset \{1, \dots, n\}} (1 - e^{-t})^{|S|} e^{-t(n-|S|)} \int f(x_1, \dots, x_n) \prod_{i \in S} \mu_i(dx_i)$ ;  
(4) Each  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$  can be written as  $f = \sum_{S \subset \{1, \dots, n\}} \hat{f}(S) u_S$  for appropriate coefficients  $\hat{f}(S)$  and  $u_S(x) = \prod_{i \in S} x_i$ <sup>2</sup>. Recall that using this representation, you defined in class an operator  $T_{e^{-t}} f(x) = \sum_{S \subset \{1, \dots, n\}} e^{-t|S|} \hat{f}(S) u_S$  for any  $t \geq 0$ . Show that  $T_{e^{-t}} = P_t$ .  
(5)

$$\mathcal{L}f := \left. \frac{d}{dt} \right|_{t=0+} P_t f(x) := \lim_{t \downarrow 0} \frac{P_t f - f}{t} = - \sum_{i=1}^n \delta_i f,$$

where  $\delta_i f(x) = f(x) - \int f(x_1, \dots, y_i, \dots, x_n) \mu_i(dy_i) = f(x) - \frac{f(x_1, \dots, 1, \dots, x_n) + f(x_1, \dots, -1, \dots, x_n)}{2}$ . Note that  $\delta_i \delta_i = \delta_i$  and  $\delta_i \delta_j = \delta_j \delta_i$ , hence,  $-\sum_i \delta_i$  is a discrete Laplacian. Thus, let us write  $\Delta$  instead of  $\mathcal{L}$ .

- (6) Conclude that  $\frac{d}{dt} P_t f = \Delta P_t f$ ;  
(7)  $\delta_i P_t f = P_t \delta_i f$ .

*Proof.* Let  $(X_m^i)_{i \leq n, m \in \mathbb{N}}$  be iid Rademacher variables independent of  $N$ . Then  $(X_t)_t$  has the same law as  $(X_{N_t}^1, \dots, X_{N_t}^n)_t$ . Using this observation, the Markov property follows directly from the independence of increments of the Poisson process, and stationarity of  $\mu$  from the independence of  $X_m^i$ 's and  $N$  and identical distribution of  $(X_m^i)_m$ 's (use tower property). Let us determine the transition semigroup and the generator of  $(X_t)_t$ ,

$$\begin{aligned} P_t f(x) &= \mathbb{E}[f(X_t)|X_0 = x] = \sum_{S \subset \{1, \dots, n\}} \mathbb{E}_x[f(X_t)(\mathbf{1}_{N_t^i > 0 \text{ for } i \in S} \mathbf{1}_{N_t^i = 0 \text{ for } i \notin S})] \\ &= \sum_{S \subset \{1, \dots, n\}} (1 - e^{-t})^{|S|} e^{-t(n-|S|)} \int f(x_1, \dots, x_n) \prod_{i \in S} \mu_i(dx_i) \\ &= \sum_{S \subset \{1, \dots, n\}} \sum_{S' \subset \{1, \dots, n\}} \hat{f}(S') (1 - e^{-t})^{|S|} e^{-t(n-|S|)} \int \prod_{i \in S'} x_i \prod_{i \in S} \mu_i(dx_i) \\ &= \sum_{S \subset \{1, \dots, n\}} \sum_{S' \subset \{1, \dots, n\}} \hat{f}(S') (1 - e^{-t})^{|S|} e^{-t(n-|S|)} \prod_{i \in S' \setminus S} x_i \underbrace{\prod_{j \in S \cap S'} \int x_j \mu_j(dx_j)}_{= \mathbf{1}_{\{S \cap S' = \emptyset\}}} \\ &= \sum_{S' \subset \{1, \dots, n\}} \hat{f}(S') \prod_{i \in S'} x_i \underbrace{\sum_{S \subset \{1, \dots, n\} \setminus S'} (1 - e^{-t})^{|S|} e^{-t(n-|S|)}}_{= e^{-t|S'|} \sum_{k=0}^{n-|S'|} \binom{n-|S'|}{k} (1 - e^{-t})^k (e^{-t})^{n-|S'|-k} = e^{-t|S'|}} \\ &= \sum_{S' \subset \{1, \dots, n\}} e^{-t|S'|} \hat{f}(S') u_{S'} = T_{e^{-t}} f; \\ \mathcal{L}f(x) &= -nf(x) + \sum_{k=1}^n \int f(x_1, \dots, x_n) \mu_k(dx_k) = - \sum_{k=1}^n \delta_k f(x). \end{aligned}$$

<sup>2</sup>this definition of  $u_S$  might differ from the one used in class by a factor of  $(-1)^{|S|}$

By Markov property (and time homogeneity), we further get that  $P_{t+s}f(x) = \mathbb{E}[\mathbb{E}[f(X_{t+s})|(X_u)_{u \leq t}]|X_0 = x] = \mathbb{E}[\tilde{\mathbb{E}}[f(\tilde{X}_s)|\tilde{X}_0 = X_t]|X_0 = x] = \mathbb{E}[P_s f(X_t)|X_0 = x] = P_t(P_s f)(x)$ . Thus,  $\frac{d}{dt}P_t f = \lim_{u \downarrow 0} \frac{P_{t+u}f - P_t f}{u} = \lim_{u \downarrow 0} \frac{P_u(P_t f) - P_t f}{u} = \mathcal{L}(P_t f) = \Delta P_t f$ . Since  $P_0 = \text{Id}$ , this in turn implies that  $P_t = e^{t\Delta}$ , but since all  $\delta_j$ 's commute with one another, it follows that  $P_t = \prod_{j=1}^n e^{-t\delta_j}$ . Clearly,  $P_t$  and  $\delta_i$  commute. <sup>3</sup>  $\square$

★ **For fun.**

**Exercise 3** (Hypercontractivity and log-Sobolev on the hypercube). *Let  $P_t, \mu$  be as in the previous exercise. Prove that the following are equivalent:*

- (1) (log-Sobolev)  $\text{Ent}_\mu[f^2] \leq \frac{1}{2}\mathbb{E}_\mu[\sum_{i=1}^n (2\delta_i f)^2]$
- (2) (Hypercontractivity)  $\|P_t f\|_{L^{q(t)}(\mu)} \leq \|f\|_{L^q(\mu)}$  for all  $q \geq 1, t \geq 0, f \in L^q(\mu)$  and  $q(t) = 1 + (q-1)e^{2t}$ .

To show the implication (1)  $\Rightarrow$  (2) you may proceed as follows:

- Reduce to the case  $f \geq 0$ .
- Consider  $\log \|P_t f\|_{L^{q(t)}}$  and compute its derivative.
- Show that the derivative is non-positive by reducing to the following problem: for a function  $g \geq 0, p = q(t)$  and  $p' = q'(t)$ ,

$$(0.1) \quad \sum_{i=1}^n \left( \mathbb{E} \left[ \left( g^{p/2} - g_{(i)}^{p/2} \right)^2 \right] - \frac{p^2}{p'} \mathbb{E}[g^{p-1} \delta_i g] \right) \leq 0.$$

*Hint: apply log-Sobolev in this step.*

- Prove (0.2) using  $\left( \frac{b^{p/2} - a^{p/2}}{b-a} \right)^2 \leq \frac{p^2}{4(p-1)} \frac{b^{p-1} - a^{p-1}}{b-a}$  (check this) and the previous exercise.

To show the converse implication you may want to re-use the above second step.

*Proof.* Let  $q \geq 1$  and  $q(t) = 1 + (q-1)e^{2t}$ . Note that  $|P_t f| \leq P_t |f|$ , therefore, we can restrict to the case  $f \geq 0$ . Note that  $\log \|P_t f\|_{L^{q(t)}} = \frac{1}{q(t)} \log \mathbb{E}[(P_t f)^{q(t)}]$  and so,

$$\begin{aligned} & \frac{d}{dt} \left( \frac{1}{q(t)} \log \mathbb{E}[(P_t f)^{q(t)}] \right) \\ &= \frac{-q'(t)}{q^2(t)} \log \mathbb{E}[(P_t f)^{q(t)}] + \frac{1}{q(t)} \frac{1}{\mathbb{E}[(P_t f)^{q(t)}]} \mathbb{E} \left[ (P_t f)^{q(t)} \left( q'(t) \log(P_t f) + q(t) \frac{1}{P_t f} \frac{d}{dt} (P_t f) \right) \right] \\ &= \frac{q'(t)}{q(t)^2 \mathbb{E}[(P_t f)^{q(t)}]} \left( \text{Ent}[(P_t f)^{q(t)}] + \frac{q(t)^2}{q'(t)} \mathbb{E}[(P_t f)^{q(t)-1} \mathcal{L}(P_t f)] \right) \\ &\stackrel{\text{log-Sobolev}}{\leq} \frac{q'(t)}{q(t)^2 \mathbb{E}[(P_t f)^{q(t)}]} \sum_{i=1}^n \left( \mathbb{E} \left[ \left( (P_t f)^{q(t)/2} - (P_t f)_{(i)}^{q(t)/2} \right)^2 \right] - \frac{q(t)^2}{q'(t)} \mathbb{E}[(P_t f)^{q(t)-1} \delta_i (P_t f)] \right), \end{aligned}$$

where  $f_{(i)}(Y) = f(Y^1, \dots, Y^{i-1}, \tilde{Y}^i, Y^{i+1}, \dots, Y^n)$ , where  $\tilde{Y}_i$  is an independent copy of  $Y_i$ . Note that the coefficient in front of the sum is positive. Thus, to prove that the derivative

<sup>3</sup>Note that to prove the latter two bullet-points of the exercise, one may also proceed differently and use an explicit form obtained in (4).

is non-positive, it suffices to show that for  $g \geq 0$ ,  $p = q(t)$  and  $p' = q'(t)$ ,

$$(0.2) \quad \sum_{i=1}^n \left( \mathbb{E} \left[ \left( g^{p/2} - g_{(i)}^{p/2} \right)^2 \right] - \frac{p^2}{p'} \mathbb{E}[g^{p-1} \delta_i g] \right) \leq 0.$$

Note that for  $a, b \geq 0$ , by Cauchy-Schwarz,  $(b^{p/2} - a^{p/2})^2 = \left( \frac{p}{2} \int_a^b u^{p/2-1} du \right)^2 \leq \frac{p^2(b-a)}{4} \int_a^b u^{p-2} = \frac{p^2(b-a)(b^{p-1} - a^{p-1})}{4(p-1)}$ . Hence, by (0.2) it suffices to show that

$$\begin{aligned} \sum_{i=1}^n \frac{p^2}{2(p-1)} & \left( \underbrace{\frac{1}{2} \mathbb{E} \left[ (g - g_{(i)})(g^{p-1} - g_{(i)}^{p-1}) \right]}_{=\mathbb{E}[g(g^{p-1} - g_{(i)}^{p-1})]} - \frac{2(p-1)}{p'} \underbrace{\mathbb{E}[g^{p-1} \delta_i g]}_{=\mathbb{E}[g(g^{p-1} - g_{(i)}^{p-1})]} \right) \\ & = \sum_{i=1}^n \frac{p^2}{2(p-1)} \mathbb{E} \left[ g(g^{p-1} - g_{(i)}^{p-1}) \right] \left( 1 - \frac{2(p-1)}{p'} \right) \leq 0. \end{aligned}$$

Note that  $1 - 2(p-1)/p' = 1 - 2(q-1)e^{2t}/(2e^{2t}(q-1)) = 0$ . All together, we have shown that  $\log \|P_t f\|_{L^{q(t)}}$  is non-increasing;  $q(0) = q$ . Therefore,  $\|P_t f\|_{L^{q(t)}} \leq \|f\|_{L^q}$  as desired.

For the reverse implication, note that by (2),  $\frac{d}{dt} \big|_{t=0} \log \|P_t f\|_{L^{q(t)}} \leq 0$ . And hence by the above and the previous exercise with  $q = 2$  (hence,  $q(0) = 2, q'(0) = 2$ ), we get that

$$0 \geq \frac{1}{2\mathbb{E}[f^2]} (\text{Ent}[f^2] + 2\mathbb{E}[f\mathcal{L}f]) = \frac{1}{2\mathbb{E}[f^2]} \left( \text{Ent}[f^2] - 2 \sum_i \mathbb{E}[f\delta_i f] \right).$$

We may yield the desired result since  $2\mathbb{E}[f\delta_i f] = \mathbb{E}[f\delta_i f] - \mathbb{E}[f^i \delta_i f] = 2\mathbb{E}[(\delta_i f)^2]$ , where  $f^i$  stands for  $f$  with flipped  $i$ th coordinate.  $\square$