

PROBLEM SESSION 3

The following two exercises produce parts of a proof of Varadhan-Bryc theorem.

**Exercise 1.** Prove that if a sequence of probability measures  $(\mu)_{n \in \mathbb{N}}$  on  $(\mathbb{X}, \mathcal{X})$  satisfies a LD lower bound with a rate function  $\mathcal{I}$  and  $F$  is a lower-semicontinuous function on  $\mathbb{X}$ , then

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \int_{\mathbb{X}} e^{nF} d\mu_n \geq \sup_{x \in \mathbb{X}} (F(x) - \mathcal{I}(x)).$$

Hint: for each  $x \in \mathbb{X}$  and  $\epsilon > 0$  consider an open set  $O_{x,\epsilon} := \{y \in \mathbb{X} : F(y) > F(x) - \epsilon\}$ , restrict the integral to this set to set a lower bound. Then let  $\epsilon \rightarrow 0$ .

**Exercise 2.** Given a rate function  $\mathcal{I}$  and a sequence of probability measures  $(\mu)_{n \in \mathbb{N}}$  on  $(\mathbb{X}, \mathcal{X})$ , show that if for every lower-semicontinuous function  $F$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \int_{\mathbb{X}} e^{nF} d\mu_n \geq \sup_{x \in \mathbb{X}} (F(x) - \mathcal{I}(x)). \quad (1)$$

then  $(\mu_n)_{n \in \mathbb{N}}$  satisfies a LD lower bound with a rate function  $\mathcal{I}$ .

Hint: given an open set  $O$ , for  $x \in O$ , a small  $\delta > 0$ , and large  $N > 0$  define  $F_{x,\delta,N}(y) = -N \min \left\{ \frac{d(x,y)}{\delta}, 1 \right\}$ . For  $F_{x,\delta,N}$  estimate the left hand side of (1) above and below, and then let  $N \rightarrow \infty$ . Notation:  $d(x,y)$  is the distance between  $x$  and  $y$  in  $\mathbb{X}$ .

$Y$  is said to have a *lognormal* distribution with parameters  $\mu \in \mathbb{R}$  and  $\sigma > 0$  if  $\ln Y$  is normal with mean  $\mu$  and variance  $\sigma^2$ .

**Exercise 3.** Let  $(Y_i)_{i \in \mathbb{N}}$  be an i.i.d. sequence of lognormal random variables with parameters  $\mu$  and  $\sigma$  and  $\nu_n$  be the distribution of geometric means,  $\tilde{Y}_n = (\prod_{i=1}^n Y_i)^{1/n}$ . Is there a LDP for  $\nu_n$ ? If yes, then what is the rate function?

**Exercise 4.** Use the contraction principle to derive Cramér theorem on a finite probability space from Sanov theorem (see the second lecture).

**Exercise 5.** Let  $(\mu_n)_{n \in \mathbb{N}}$  be the distribution of empirical means of a sequence of i.i.d. Bernoulli variables with parameter  $1/2$ . Use exponential tilting to obtain a LDP for the distributions  $(\nu_n)_{n \in \mathbb{N}}$  of empirical means of a sequence of i.i.d. Bernoulli variables with parameter  $p \neq 1/2$  directly from the LDP for  $(\mu_n)_{n \in \mathbb{N}}$  (see the first lecture).

In preparation for the last lecture, solve the following exercise.

**Exercise 6.** Let  $\mu$  be a probability measure on  $(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$  with a density proportional to  $e^{-|x|}/(1 + |x|^{d+2})$  and  $\Lambda$  be its logarithmic MGF.

(a) Show that  $D_\Lambda = \{x \in \mathbb{R}^d : \|x\| \leq 1\}$ .

(b) Show that  $|\nabla \Lambda(x)| \not\rightarrow \infty$  as  $x \in D_\Lambda$  approaches the boundary of  $D_\Lambda$ .

This is an example of non-steep logarithmic MGF.

**Exercise 7.** Study the Curie-Weiss model discussed in lecture notes in the case when  $h \neq 0$ .