

4 More on the basic notions

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4a Every LD-convergent sequence satisfies LDP

Here we prove Prop. 2a11 for every seminorm $\|\cdot\|$ on $C(K)$ that satisfies (2a1), (2a2) and (2a6).

First, let $f \in C(K)$ satisfy $\|f\| \leq 1$; we have to prove that $\max(|f|\Pi) \leq 1$, that is, $|f(x)|\Pi(x) \leq 1$ for all x . However, $1/\Pi(x) \geq |f(x)|$ by (2a7).

Second, let $f \in C(K)$ satisfy $\max(|f|\Pi) \leq 1$; we have to prove that $\|f\| \leq 1$. By (2a7),

$$|f(x)| \leq \frac{1}{\Pi(x)} = \sup\{g(x) : \|g\| \leq 1\}$$

for every $x \in K$.

Let $\varepsilon > 0$ be given. For every $x \in K$ there exists $g \in C(K)$ such that $\|g\| \leq 1$ and $g(x) > |f(x)| - \varepsilon$. The inequality still holds on some neighborhood of x . By compactness we may cover K by a finite number of such neighborhoods. In other words, we have $g_1, \dots, g_n \in C(K)$ such that $\|g_1\| \leq 1, \dots, \|g_n\| \leq 1$ and $g_1 \vee \dots \vee g_n > |f| - \varepsilon$ on K . By (2a6), $\|g_1 \vee \dots \vee g_n\| \leq \|g_1\| \vee \dots \vee \|g_n\| \leq 1$. By (2a1) and (2a2), $\|f\| \leq 1 + \varepsilon$ for every $\varepsilon > 0$, which completes the proof.

4b The probability decay rate

We deal with a compact metrizable space K and probability measures μ_n on K satisfying LDP with a rate function I . However, our first lemma does not use μ_n (and its first item does not use the topology of K).

4b1 Lemma. (a) Let $\varphi_n, \varphi : K \rightarrow \mathbb{R}$, $\varphi_n \uparrow \varphi$ pointwise; then $(\sup_K \varphi_n) \uparrow (\sup_K \varphi)$ as $n \rightarrow \infty$.

(b) Let $\varphi_n, \varphi : K \rightarrow \mathbb{R}$ be upper semicontinuous, and $\varphi_n \downarrow \varphi$ pointwise; then $(\max_K \varphi_n) \downarrow (\max_K \varphi)$ as $n \rightarrow \infty$.

Proof. (a) For every $\varepsilon > 0$ we take $x \in K$ such that $\varphi(x) > (\sup_K \varphi) - \varepsilon$ and n such that $\varphi_n(x) > (\sup_K \varphi) - \varepsilon$; then $(\sup_K \varphi_n) > (\sup_K \varphi) - \varepsilon$.

(b) We have $(\max_K \varphi_n) \downarrow c$ for some $c \in \mathbb{R}$. For every $\varepsilon > 0$ the sets $\{x \in K : \varphi_n(x) \geq c - \varepsilon\}$ are a decreasing sequence of nonempty closed sets. By compactness, some x belongs to all these sets. Thus, $\varphi(x) = \lim_n \varphi_n(x) \geq c - \varepsilon$ and $\max_K \varphi \geq c - \varepsilon$. \square

4b2 Exercise. Without the semicontinuity 4b1(b) need not hold.

Find a counterexample.

4b3 Lemma. Let $f : K \rightarrow \mathbb{R}$.

(a) If $|f|$ is lower semicontinuous then

$$\liminf_n \|f\|_{L_n(\mu_n)} \geq \sup_K (|f|e^{-I});$$

(b) if $|f|$ is upper semicontinuous then

$$\limsup_n \|f\|_{L_n(\mu_n)} \leq \max_K (|f|e^{-I}).$$

Proof. (a) We take $f_n \in C(K)$ such that $0 \leq f_n \uparrow |f|$. For every j ,

$$\liminf_n \|f\|_{L_n(\mu_n)} \geq \liminf_n \|f_j\|_{L_n(\mu_n)} = \max_K (f_j e^{-I});$$

however,

$$\max_K (f_j e^{-I}) \uparrow \sup_K (|f|e^{-I}) \quad \text{as } j \rightarrow \infty$$

by 4b1(a).

(b): similar (but using semicontinuity). \square

4b4 Corollary.

(a) $\liminf_n (\mu_n(G))^{1/n} \geq \exp(-\inf_G I)$ for every open $G \subset K$,

(b) $\limsup_n (\mu_n(F))^{1/n} \leq \exp(-\min_F I)$ for every closed $F \subset K$.

4b5 Exercise. Reconsider 2a18 and 2a20 in the light of 4b4.

4b6 Corollary. If an open set $G \subset K$ satisfies

(a) $\inf_G I = \min_{\overline{G}} I$

then

(b) $\lim_n (\mu_n(G))^{1/n} = \lim_n (\mu_n(\overline{G}))^{1/n} = \exp(-\inf_G I) = \exp(-\min_{\overline{G}} I)$,

that is,

(c) $\lim_n \frac{1}{n} \ln \mu_n(G) = \lim_n \frac{1}{n} \ln \mu_n(\overline{G}) = -\inf_G I = -\min_{\overline{G}} I$.

4b7 Exercise. 4b6(a) does not imply $\mu_n(\overline{G}) \sim \mu_n(G)$ as $n \rightarrow \infty$.

Find a counterexample.

Hint: try $K = [0, 1]$, $G = (0, 1]$, combine μ_n from Lebesgue measure and an atom at 0, and find appropriate coefficients.

Continuity of I is, of course, sufficient for 4b6(a). Here is a weaker sufficient condition:

$$(4b8) \quad \limsup_{y \rightarrow x, y \in G} I(y) \leq I(x) \quad \text{for all } x \in \partial G.$$

We may also consider $\mu_n(A_n)$ assuming that A_n converge to G in an appropriate sense. To this end we choose a metric on K and, given a set $A \subset K$, we introduce (for any $\varepsilon > 0$)

$$(4b9) \quad A_{+\varepsilon} = \{x \in K : \text{dist}(x, A) \leq \varepsilon\},$$

$$(4b10) \quad A_{-\varepsilon} = \{x \in K : \text{dist}(x, \mathcal{C}A) > \varepsilon\};$$

here $\text{dist}(x, A) = \inf_{y \in A} \text{dist}(x, y)$, and $\mathcal{C}A = \{x \in K : x \notin A\}$. Note that $A_{+\varepsilon}$ is closed, $A_{-\varepsilon}$ is open, $\bigcap_{\varepsilon} A_{+\varepsilon} = \overline{A}$ is the closure of A , and $\bigcup_{\varepsilon} A_{-\varepsilon} = A^\circ$ is the interior of A .

4b11 Exercise. Let $A_n \subset K$ be such that A_n is μ_n -measurable.

(a) Let $G \subset K$ be an open set, and

$$A_n \supset G_{-\varepsilon_n} \quad \text{for some } \varepsilon_n \downarrow 0.$$

Then

$$\liminf_n (\mu_n(A_n))^{1/n} \geq \exp(-\inf_G I).$$

(b) Let $F \subset K$ be a closed set, and

$$A_n \subset F_{+\varepsilon_n} \quad \text{for some } \varepsilon_n \downarrow 0.$$

Then

$$\limsup_n (\mu_n(A_n))^{1/n} \leq \exp(-\min_F I).$$

(c) Let $G \subset K$ be an open set such that $\inf_G I = \min_{\overline{G}} I$, and

$$G_{-\varepsilon_n} \subset A_n \subset G_{+\varepsilon_n} \quad \text{for some } \varepsilon_n \downarrow 0.$$

Then

$$\lim_n (\mu_n(A_n))^{1/n} = \exp(-\inf_G I) = \exp(-\min_{\overline{G}} I).$$

Prove it.

Hint: (a) the argument of the proof of 4b3(a) works for appropriate f_n , say, $f_n(x) = (1/\varepsilon_n) \text{dist}(x, \mathcal{C}G) - 1$ if this number lies on $[0, 1]$, otherwise 0 (if the number is negative) or 1 (if it exceeds 1); (b) similar (but using semicontinuity), (c) follows from (a), (b).

We can also describe the value $I(x)$ of the rate function at a given point x in terms of probabilities. To this end we choose (once again) a metric on K and use open and closed balls,

$$B(x, r-) = \{y \in K : \text{dist}(x, y) < r\}, \quad B(x, r+) = \{y \in K : \text{dist}(x, y) \leq r\}.$$

4b12 Proposition. For every $x \in K$ there exists a function $(0, 1) \rightarrow \{1, 2, \dots\}$, $r \mapsto n_r$, such that

$$\frac{1}{n} \ln \mu_n(B(x, r\pm)) \rightarrow -I(x) \quad \text{as } r \rightarrow 0+$$

uniformly in $n \geq n_r$. (Here ‘ \pm ’ means that the claim holds for closed and open balls.)

Proof. By 4b4,

$$\begin{aligned} \liminf_n (\mu_n(B(x, r-)))^{1/n} &\geq \exp\left(-\inf_{B(x, r-)} I\right), \\ \limsup_n (\mu_n(B(x, r+)))^{1/n} &\leq \exp\left(-\min_{B(x, r+)} I\right). \end{aligned}$$

We choose n_r such that

$$\begin{aligned} (\mu_n(B(x, r-)))^{1/n} &\geq \exp\left(-\inf_{B(x, r-)} I\right) - r, \\ (\mu_n(B(x, r+)))^{1/n} &\leq \exp\left(-\min_{B(x, r+)} I\right) + r \end{aligned}$$

for all $n \geq n_r$. By lower semicontinuity of I ,

$$\inf_{B(x, r\pm)} I \uparrow I(x) \quad \text{as } r \rightarrow 0+.$$

We have

$$\begin{array}{ccc} \exp\left(-\inf_{B(x, r-)} I\right) - r & \leq & (\mu_n(B(x, r-)))^{1/n} \leq (\mu_n(B(x, r+)))^{1/n} \leq \exp\left(-\min_{B(x, r+)} I\right) + r \\ & \searrow & \swarrow \\ & e^{-I(x)} & \end{array}$$

therefore

$$(\mu_n(B(x, r\pm)))^{1/n} \rightarrow e^{-I(x)} \quad \text{as } r \rightarrow 0+$$

uniformly in $n \geq n_r$. □

In fact, the (necessary) condition

(4b13)

$$\lim_{r \rightarrow 0+} \liminf_n (\mu_n(B(x, r-)))^{1/n} = \lim_{r \rightarrow 0+} \limsup_n (\mu_n(B(x, r+)))^{1/n} \quad \text{for } x \in K$$

is also sufficient for LD-convergence of $(\mu_n)_n$. I give no proof. (See also [3, Sect. 3.1, Remark 3.1(c)] and [1, Th. 4.1.11].)

4c Restriction and conditioning

Any large deviation is done in the least unlikely of all the unlikely ways!

den Hollander [4, p. 10]

Let probability measures μ_n on a compact metrizable space K satisfy LDP with a rate function I . Assume that an open set $G \subset K$ satisfies (4b8) (which always holds if I is continuous).

4c1 Proposition. For every $f \in C(K)$,

$$\lim_n \left(\int_G |f|^n d\mu_n \right)^{1/n} = \lim_n \left(\int_{\overline{G}} |f|^n d\mu_n \right)^{1/n} = \sup_G (|f|e^{-I}) = \max_{\overline{G}} (|f|e^{-I}).$$

Proof. We may assume that $f(\cdot) \geq 0$, since only $|f|$ is relevant. Moreover, we may assume that $f(\cdot) > 0$, since strictly positive functions are dense among weakly positive functions. Thus, we assume that $f = e^{-h}$, $h \in C(K)$.

We define probability measures ν_n on K by

$$\frac{d\nu_n}{d\mu_n} = c_n e^{-nh}$$

and apply Theorem 2c1 (change of measure): $(\nu_n)_n$ satisfies LDP with the rate function $J = I + h - a$, $a = \lim_n \frac{1}{n} \ln c_n$. Condition (4b8) is satisfied also by J , thus 4b6 can be applied to $(\nu_n)_n$ giving

$$\lim_n (\nu_n(G))^{1/n} = \lim_n (\nu_n(\overline{G}))^{1/n} = \exp(-\inf_G J) = \exp(-\min_{\overline{G}} J).$$

However,

$$\left(\int_G f^n d\mu_n \right)^{1/n} = \left(\int_G e^{-nh} d\mu_n \right)^{1/n} = (c_n^{-1} \nu_n(G))^{1/n} \rightarrow e^{-a} \lim_n (\nu_n(G))^{1/n}$$

as $n \rightarrow \infty$; the same holds for \overline{G} . Also,

$$\sup_G (fe^{-I}) = \sup_G e^{-(h+I)} = \sup_G e^{-(J+a)} = e^{-a} \exp(-\inf_G J);$$

the same holds for \overline{G} . □

It may happen that $I(x) = +\infty$ for all $x \in \overline{G}$. Let us exclude this case. Then $\mu_n(\overline{G}) \neq 0$ for all n large enough, and we may introduce *conditional measures*, — probability measures ν_n on \overline{G} such that

$$(4c2) \quad \int f d\nu_n = \frac{1}{\mu_n(\overline{G})} \int_{\overline{G}} f d\mu_n$$

for all bounded Borel functions $f : \overline{G} \rightarrow \mathbb{R}$. The set \overline{G} is another compact metrizable space, and we may consider LDP on this space.

4c3 Proposition. Let $\min_{\overline{G}} I \neq +\infty$, then the sequence $(\nu_n)_n$ of conditional measures on \overline{G} satisfies LDP with the rate function $J : \overline{G} \rightarrow [0, \infty]$,

$$J(x) = I(x) - \min_{\overline{G}} I \quad \text{for } x \in \overline{G}.$$

Proof. Let $f \in C(\overline{G})$; we have to prove that $(\int_{\overline{G}} |f|^n d\nu_n)^{1/n} \rightarrow \max_{\overline{G}}(|f|e^{-J})$. By 4c1 (applied to any continuous extension of f), $(\int_{\overline{G}} |f|^n d\mu_n)^{1/n} \rightarrow \max_{\overline{G}}(|f|e^{-I})$. Therefore

$$\left(\int_{\overline{G}} |f|^n d\nu_n \right)^{1/n} = \left(\frac{\int_{\overline{G}} |f|^n d\mu_n}{\mu_n(\overline{G})} \right)^{1/n} \rightarrow \frac{\max_{\overline{G}}(|f|e^{-I})}{\max_{\overline{G}}(e^{-I})} = \max_{\overline{G}}(|f|e^{-J}).$$

□

4c4 Exercise. Let I be continuous and $\min_{\overline{G}} I \neq +\infty$, then

$$\frac{\mu_n(\{x \in \overline{G} : I(x) \leq \min_{\overline{G}} I + \varepsilon\})}{\mu_n(\overline{G})} \rightarrow 1 \quad \text{as } n \rightarrow \infty$$

for all $\varepsilon > 0$.

Prove it.

Hint: use 4c3, and apply 2a20(a) to ν_n .

4c5 Exercise. A fair coin is tossed n times, giving S_n ‘heads’. Prove that

$$\mathbb{P}(S_n \leq 0.71n \mid S_n \geq 0.7n) \rightarrow 1 \quad \text{as } n \rightarrow \infty.$$

Hint: 3a4 and 4c4.

4c6 Exercise. A fair die is thrown n times, giving the outcomes $1, \dots, 6$ respectively $S_n^{(1)}, \dots, S_n^{(6)}$ times. Prove that

$$\mathbb{P}(0.15n \leq S_n^{(2)}, \dots, S_n^{(6)} \leq 0.17n \mid S_n^{(1)} \geq 0.2n) \rightarrow 1 \quad \text{as } n \rightarrow \infty.$$

Hint: 3b3 and 4c4.

4c7 Exercise. Let X_1, \dots, X_n be independent, identically distributed random variables, each taking on the three values $-1, 0, 1$ with equal probabilities ($1/3$). Prove that

$$\mathbb{P}\left(\frac{5}{7} - \varepsilon \leq \frac{X_1^2 + \dots + X_n^2}{n} \leq \frac{5}{7} + \varepsilon \mid \frac{X_1 + \dots + X_n}{n} \geq \frac{3}{7}\right) \rightarrow 1 \quad \text{as } n \rightarrow \infty$$

for every $\varepsilon > 0$.

Hint: 3c, and 4c4.

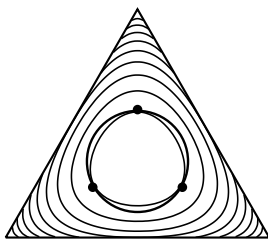
You may also think about the conditional distribution of the frequencies $\frac{k_-}{n}, \frac{k_0}{n}, \frac{k_+}{n}$ (and the mean $\frac{1}{n}(X_1 + \dots + X_n) = \frac{k_+}{n} - \frac{k_-}{n}$), where

$$k_- = \#\{j : X_j = -1\}, \quad k_0 = \#\{j : X_j = 0\}, \quad k_+ = \#\{j : X_j = 1\}$$

(recall 3c), the condition being a large deviation of the frequencies from the probabilities in the sense that

$$\left| \frac{k_-}{n} - \frac{1}{3} \right|^2 + \left| \frac{k_0}{n} - \frac{1}{3} \right|^2 + \left| \frac{k_+}{n} - \frac{1}{3} \right|^2 \geq c.$$

In terms of the so-called χ^2 statistics, $\chi^2 = \frac{3}{n} \left((k_- - \frac{n}{3})^2 + (k_0 - \frac{n}{3})^2 + (k_+ - \frac{n}{3})^2 \right)$, it means $\chi^2 \geq 3cn$.



4c8 Exercise. Generalize 4c1, 4c3 and 4c4, replacing the single set \overline{G} with a sequence of sets A_n such that A_n is μ_n -measurable, and

$$G_{-\varepsilon_n} \subset A_n \subset G_{+\varepsilon_n} \quad \text{for some } \varepsilon_n \downarrow 0.$$

Hint: use 4b11(c).

See also [2, Sect. 4] and [1, Sect. 3.3].

4d LDP for product measures

Let K_1, K_2 be compact metrizable spaces, then their product $K = K_1 \times K_2$ is also a compact metrizable space.

Let $\mu_n^{(1)}$ be probability measures on K_1 , and $\mu_n^{(2)}$ — on K_2 , then their products $\mu_n = \mu_n^{(1)} \times \mu_n^{(2)}$ are probability measures on K .

4d1 Proposition. If $(\mu_n^{(1)})_n$ satisfies LDP with a rate function I_1 and $(\mu_n^{(2)})_n$ — with I_2 , then $(\mu_n)_n$ satisfies LDP with the rate function I defined by

$$I(x, y) = I_1(x) + I_2(y) \quad \text{for } x \in K_1, y \in K_2.$$

Proof. Given $f \in C(K)$, we define $g, g_1, g_2, \dots : K_1 \rightarrow \mathbb{R}$ by

$$g_n(x) = \|f(x, \cdot)\|_{L_n(\mu_n^{(2)})} = \left(\int |f(x, y)|^n \mu_n^{(2)}(dy) \right)^{1/n},$$

$$g(x) = \sup_{K_2} (|f(x, \cdot)|e^{-I_2}) = \sup_{y \in K_2} (|f(x, y)|e^{-I_2(y)}).$$

Clearly, $g_n \rightarrow g$ pointwise. But moreover, $g_n \rightarrow g$ uniformly, due to uniform continuity:

$$\begin{aligned} |g_n(x_1) - g_n(x_2)| &\leq \sup_{K_2} |f(x_1, \cdot) - f(x_2, \cdot)|, \\ |g(x_1) - g(x_2)| &\leq \sup_{K_2} |f(x_1, \cdot) - f(x_2, \cdot)|. \end{aligned}$$

We note that $\|g_n\|_{L_n(\mu_n^{(1)})} = \|f\|_{L_n(\mu_n)}$, $\|g_n - g\|_{L_n(\mu_n^{(1)})} \leq \sup_{K_1} |g_n - g| \rightarrow 0$ and $\|g\|_{L_n(\mu_n^{(1)})} \rightarrow \sup_{K_1} (|g|e^{-I_1})$, therefore

$$\begin{aligned} \|f\|_{L_n(\mu_n)} &\rightarrow \sup_{K_1} (|g|e^{-I_1}) = \\ &= \sup_{x \in K_1} \left(e^{-I_1(x)} \sup_{y \in K_2} (e^{-I_2(y)} |f(x, y)|) \right) = \sup_K (|f|e^{-I}). \end{aligned}$$

□

4d2 Exercise. $(\mu_n)_n$ is LD-convergent if and only if both $(\mu_n^{(1)})_n$ and $(\mu_n^{(2)})_n$ are LD-convergent.

Prove it.

Hint: use the contraction principle for ‘only if’.

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