

# Useful concrete LDPs

Large Deviations and Queues—Damon Wischik

## 1 A trivial LDP

Let  $X \sim \text{Exp}(\lambda)$ . Then  $(X/l, l \in \mathbb{R}_+)$  satisfies an LDP in  $\mathbb{R}_+$  with good rate function

$$I(x) = \lambda x.$$

## 2 Cramér's Theorem

Let  $X^L$  be the average of  $L$  independent copies of a real-valued random variable  $X$ . Then  $X^L$  satisfies an LDP in  $\mathbb{R}$  with convex rate function

$$\Lambda^*(x) = \sup_{\theta \in \mathbb{R}} \theta x - \Lambda(\theta)$$

where the log moment generating function  $\Lambda(\cdot)$  is

$$\Lambda(\theta) = \log \mathbb{E} \exp(\theta X).$$

If  $\Lambda(\cdot)$  is finite in a neighbourhood of the origin then  $\Lambda^*$  is a good rate function.

## 3 Gärtner-Ellis Theorem

The *effective domain* of a function  $\Lambda(\cdot)$  taking values in  $\mathbb{R} \cup \{\infty\}$  is  $\mathcal{D}\Lambda = \{\theta : \Lambda(\theta) \in \mathbb{R}\}$ .

A convex function  $\Lambda : \mathbb{R}^d \rightarrow \mathbb{R} \cup \{\infty\}$  is *essentially smooth* if

- i.  $(\mathcal{D}\Lambda)^\circ$  is non-empty
- ii.  $\Lambda(\cdot)$  is differentiable throughout  $(\mathcal{D}\Lambda)^\circ$
- iii.  $\Lambda(\cdot)$  is steep, namely,  $|\nabla \Lambda(\theta_n)| \rightarrow \infty$  whenever  $(\theta_n)$  is a sequence in  $(\mathcal{D}\Lambda)^\circ$  converging to a point on the boundary of  $\mathcal{D}\Lambda$ .

Let  $(X^L, L \in \mathbb{N})$  be a sequence of random vectors in  $\mathbb{R}^d$ , and let

$$\Lambda^L(\theta) = \frac{1}{L} \log \mathbb{E} \exp(L\theta \cdot X^L)$$

for  $\theta \in \mathbb{R}^d$ . Assume that for each  $\theta$  the limit

$$\Lambda(\theta) = \lim_{L \rightarrow \infty} \Lambda^L(\theta)$$

exists in  $\mathbb{R} \cup \{\infty\}$ .

Assume further that  $0 \in (\mathcal{D}\Lambda)^\circ$  and that  $\Lambda$  is essentially smooth and lower-semicontinuous. Then  $(X^L, L \in \mathbb{N})$  satisfies an LDP in  $\mathbb{R}^d$  with good rate function

$$\Lambda^*(x) = \sup_{\theta \in \mathbb{R}^d} \theta \cdot x - \Lambda(\theta).$$

Note that  $\Lambda(\cdot)$ , being the pointwise limit of convex functions, is itself convex. Note that  $\Lambda^*(\cdot)$ , being a convex conjugate, is convex and lower-semicontinuous. Since  $\Lambda(\cdot)$  is convex, it satisfies the following: if  $\Lambda(\cdot)$  is differentiable at  $\theta$  and  $\nabla \Lambda(\theta) = x$  then  $\Lambda^*(x) = \theta \cdot x - \Lambda(\theta)$ .

## 4 Schilder's Theorem

Let  $\mathcal{C}[0, 1]$  be the space of continuous functions  $f : [0, 1] \rightarrow \mathbb{R}$ , equipped with the supremum norm

$$\|f\| = \sup_{0 \leq t \leq 1} |f(t)|.$$

Say that a function  $f : [0, 1] \rightarrow \mathbb{R}$  is *absolutely continuous* (AC) if for all  $\varepsilon > 0$  there exists a  $\delta > 0$  such for every finite collection of non-overlapping intervals  $\{[s_i, t_i], 1 \leq i \leq N\}$ ,

$$\sum_{1 \leq i \leq N} (t_i - s_i) < \delta \implies \sum_{1 \leq i \leq N} |f(t_i) - f(s_i)| < \varepsilon.$$

Absolutely continuous functions are dense in the space of continuous functions. An absolutely continuous function is differentiable almost everywhere.

Let  $(B(t), t \in [0, 1])$  be a standard Brownian motion, taking values in  $\mathcal{C}[0, 1]$ . Let

$$B^l(t) = \frac{1}{\sqrt{l}} B(t).$$

Then  $(B^l(\cdot), l \in \mathbb{R}_+)$  satisfies an LDP in  $\mathcal{C}[0, 1]$  with good rate function

$$I(f) = \begin{cases} \int_{t=0}^1 \frac{1}{2} \dot{f}(t)^2 dt & \text{if } f \text{ is AC and } f(0) = 0 \\ \infty & \text{otherwise.} \end{cases}$$

## 5 Mogulskii's theorem

Let  $X_1, X_2, \dots$  be independent copies of a real-valued random variable  $X$  with log moment generating function  $\Lambda(\cdot)$ . Define the partial sums process

$$S_n = X_1 + \dots + X_n, \quad \text{for } n \in \mathbb{N};$$

and the polygonalized partial sums process

$$\tilde{S}_t = (\lceil t \rceil - t)S_{\lceil t \rceil} + (t - \lfloor t \rfloor)S_{\lfloor t \rfloor}, \quad \text{for } t \in \mathbb{R}_+;$$

and the scaled polygonalized partial sums process

$$Y^L(t) = \frac{1}{L} \tilde{S}_{Lt}, \quad \text{for } 0 \leq t \leq 1.$$

Then  $(Y^L(\cdot), L \in \mathbb{N})$  satisfies an LDP in  $\mathcal{C}[0, 1]$  with good rate function

$$I(f) = \begin{cases} \int_{t=0}^1 \Lambda^*(\dot{f}(t)) dt & \text{if } f \text{ is AC and } f(0) = 0 \\ \infty & \text{otherwise.} \end{cases}$$

This has been extended, along the lines of the Gärtner-Ellis theorem, by Dembo and Zajic (1995), *From empirical mean and measure to partial sums processes*, Stochastic Processes and their Applications.