

Large Deviation Theory

The theory of large deviation deals with the proba. of rare events that're exponentially small as function of some parameter.

(1) For i.i.d. seq.

Consider on $(\mathbb{R}, \mathcal{B}, \mathbb{P})$, $(X_n)_{n \geq 1}$ i.i.d.

st. $\forall \lambda \in \mathbb{R}'$. $\mathbb{E}(e^{\lambda X_1}) < \infty$

Rmk: $\forall k \in \mathbb{R}'$. $\mathbb{E}(|X_1|^k) < \infty$. Since we

$$\text{have } |x|^k \leq \sum_k e^{-x} + e^x.$$

$$\text{By CLT: } \lim_{n \rightarrow \infty} \mathbb{P}\left(\frac{\sum_{k=1}^n X_k}{n} \geq c\right) = (2\pi)^{-\frac{1}{2}} \int_c^\infty e^{-\frac{x^2}{2}} dx.$$

$$\underline{Q}: \text{ Does } \mathbb{P}\left(\frac{\sum_{k=1}^n X_k}{n} \geq c\right) / (2\pi)^{-\frac{1}{2}} \int_c^\infty e^{-\frac{x^2}{2}} dx \xrightarrow{c \rightarrow \infty} 1 ?$$

A: It holds under some suitable conditions when $c \ll n^{\frac{1}{2}}$. But when $c \sim \sqrt{n}$ it won't hold (e.g. $|X_1| \leq C$. $c > Cn^{\frac{1}{2}}$)

$$\underline{\text{Thm.}} \text{ For } A \in \mathcal{B}, \quad \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}(\bar{X}_n \in A)$$

$$= - \inf_A h(x), \quad h(x) = \sup_{\theta} (\theta x - \psi(\theta)).$$

$$\psi(\theta) = \log M(\theta), \quad M(\theta) = \mathbb{E}(e^{\theta X_1}).$$

① Upper bound:

First consider $A = (t, \infty)$. $t > \mathbb{E}(X_1) \triangleq m$.

$$P(\bar{X}_n \geq t) \stackrel{\text{Cheb}}{\leq} e^{-\theta n t} \mathbb{E}(e^{\theta S_n})$$

$$= e^{-\theta n t} M(\theta)^n \quad \text{with } \theta \geq 0$$

$$\Rightarrow \frac{1}{n} \log P(\bar{X}_n \geq t) \leq - \sup_{\theta \geq 0} (\theta t - \psi(\theta))$$

Remark: For $\theta < 0$. By Jensen inequality:

$$\psi(\theta) \geq m\theta \geq t\theta. \quad \text{But } 0 \text{ is trivial}$$

lower bdd $\Rightarrow$ Replace RHS by $-\psi(\theta)$.

About $\psi(\theta)$:

Def: i) Legendre - Fenchel transform L is defined on functions $\psi(x)$.

$$\text{by } L(\psi(t)) =: \sup_{\theta} (t\theta - \psi(\theta)).$$

ii) P_λ is p.m. on $\mathcal{X}$. $P_\lambda(A) =:$

$$\mathbb{E}(I_A e^{\lambda X_1}) / \mathbb{E}(e^{\lambda X_1}).$$

prop. i) ψ is convex $\Rightarrow L\psi$ is convex.

ii) $L^2 = \text{id}$ on convex functions.

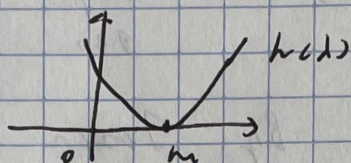
Note that $M(\lambda)$ is convex:

$$\psi'(\theta) = \mathbb{E}_{\theta}(X_1), \quad \psi''(\theta) = \mathbb{E}_{\theta}(X_1^2) - \mathbb{E}_{\theta}(X_1)^2 \geq 0$$

So $h(\lambda)$ is convex. $h(\lambda) \geq 0$.

Remark: Note $h(m) = 0$. $h'(m) = 0$. m is the point where h attains its min

convex
 $\Rightarrow$



$h \uparrow$ on $\lambda \geq m$

$h \downarrow$ on $\lambda \leq m$

By Remark, we can also extend to set

$A = (-\infty, a]$. For general set A , we

can set $(-\infty, a] \cup [b, \infty)$ contains A .

and $a, b \in \bar{A}$ with conti of h (by convex)

$$\Rightarrow \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}(\bar{X}_n \in A) \leq \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}(\bar{X}_n \leq a \text{ or } \geq b)$$

$$\leq -h(a) \wedge h(b)$$

$$\stackrel{\text{conti}}{=} -\inf_{(-\infty, a] \cup [b, \infty)} h(\lambda) \stackrel{\text{mono.}}{=} -\inf_{\lambda} h(\lambda)$$

Remark: We always assume A is closed for the upper bound.

② Lower bound:

$$\text{Consider } M_{\lambda}^N(A) =: \mathbb{E}(\mathbb{I}(X_1, \dots, X_N) \in A) \leq \sum_{i=1}^N (\lambda X_i - \psi(\lambda))$$

For $A = \mathbb{R}$. open set of $\mathbb{R}$.

$$\text{Set } J_\theta(\lambda) = \lambda\theta - \psi(\lambda), \quad J'_\theta(\lambda) = 0 \Rightarrow \theta = \psi'(\lambda)$$

Assumption: $\lim_{\lambda \rightarrow \infty} \psi'(\lambda) = +\infty$.

Remark: By monotonicity, $\lim_{\lambda \rightarrow \infty} \psi'(\lambda) = p$ exists, if $p < \infty$. then:
for $\lambda > p$, $h(\lambda) = \infty$.

$$\Rightarrow h(\theta) = \theta(\psi')^{-1}(\theta) - \psi((\psi')^{-1}(\theta))$$

$$h'(\theta) = (\psi')^{-1}(\theta).$$

For $\forall \alpha \in \mathbb{R}$. $B(\alpha, \delta)$ is nbd of α in $\mathbb{R}$.

$$P(\bar{X}_n \in \mathbb{R}) \geq P(\bar{X}_n \in B(\alpha, \delta))$$

$$\geq e^{-n(\lambda(\alpha+\delta) - \psi(\lambda))} M_n^\lambda(B(\alpha, \delta))$$

Choose $\hat{\lambda} = (\psi')^{-1}(\alpha)$. with CLT: we have,

$$M_n^{\hat{\lambda}}(B(\alpha, \delta)) \xrightarrow{n \rightarrow \infty} 1. \quad \text{by } E_{\hat{\lambda}}(\bar{X}_n) = \psi'(\hat{\lambda}) = \alpha$$

$$\Rightarrow \frac{1}{n} \log P(\bar{X}_n \in \mathbb{R}) \geq -(\psi')^{-1}(\alpha)(\alpha+\delta) - \psi((\psi')^{-1}(\alpha))$$
$$\xrightarrow{\delta \rightarrow 0} -h(\alpha). \quad \forall \alpha \in \mathbb{R}.$$

For general set A . consider $\bar{A}$ not int A .

with conti. of $h(\lambda)$.