

Lecture: Apr 7

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1 Convergence rate

1.1 Notation

Let us define

$$M_n(\theta) = \frac{1}{n} \sum_{i=1}^n l_\theta(X_i) = \mathbb{P}_n l_\theta,$$

$$\hat{\theta}_n = \arg \min_{\theta \in \Theta} M_n(\theta),$$

$$M(\theta) = \mathbb{P} l_\theta,$$

$$\Delta_n(\theta) = (M_n(\theta) - M(\theta)) - (M_n(\theta_0) - M(\theta_0)) = (\mathbb{P}_n - \mathbb{P})(l_\theta - l_{\theta_0}).$$

Also, define the modulus of continuity

$$\mathbb{W}_n(\delta) = \sup_{d(\theta, \theta_0) \leq \delta} |\Delta_n(\theta)|.$$

Let $\mathcal{F} = \{l_\theta - l_{\theta_0} : \theta \in \Theta\}$. Then

$$\Delta_n(\theta) = (\mathbb{P}_n - \mathbb{P})f_\theta$$

where $f_\theta = l_\theta - l_{\theta_0} \in \mathcal{F}$.

1.2 Theorem

Suppose

$$M(\theta) \geq M(\theta_0) + \lambda d^2(\theta, \theta_0)$$

for θ near θ_0 for some $\lambda > 0$. Let ϕ be a function such that $\phi(\delta) \leq c\delta^\alpha$ for $\alpha \in (0, 2)$ and some constant $c > 0$. Assume,

$$\mathbb{E}[\mathbb{W}_n(\delta)] \leq \frac{\phi(\delta)}{\sqrt{n}}.$$

Let $r_n \rightarrow \infty$ such that $cr_n^{2-\alpha} \leq \sqrt{n}$. If $\hat{\theta}_n \xrightarrow{P} \theta_0$, then

$$r_n d(\hat{\theta}_n, \theta_0) = O_p(1).$$

1.3 Proof

We aim to show that $r_n d(\hat{\theta}_n, \theta_0) = O_p(1)$. That is for any $\epsilon > 0$, there exists an t such that

$$\mathbb{P}(r_n d(\hat{\theta}_n, \theta_0) > 2^t) \leq \epsilon.$$

Since $\hat{\theta}_n \xrightarrow{P} \theta_0$, for a fixed $\eta > 0$, we know that

$$\mathbb{P}[d(\hat{\theta}_n, \theta_0) > \eta] \rightarrow 0.$$

The key technique we employ here is the so-called peeling device. Define

$$S_{n,j} = \{\theta : 2^j < r_n d(\theta, \theta_0) \leq 2^{j+1}\}.$$

Then we have

$$\begin{aligned} \mathbb{P}[r_n d(\hat{\theta}_n, \theta_0) > 2^t] &\leq \mathbb{P}[r_n d(\hat{\theta}_n, \theta_0) > 2^t, d(\hat{\theta}_n, \theta_0) \leq \eta] + \mathbb{P}[d(\hat{\theta}_n, \theta_0) > \eta] \\ &\leq \mathbb{P}[\exists j \geq t \text{ such that } 2^j \leq r_n \eta \text{ and } \hat{\theta}_n \in S_{n,j}] + o(1) \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \mathbb{P}[\hat{\theta}_n \in S_{n,j}] + o(1). \end{aligned}$$

Since $\hat{\theta}_n \in S_{n,j}$, there exists $\theta \in S_{n,j}$ such that $M_n(\theta) \leq M_n(\theta_0)$. By the assumption, we have

$$\begin{aligned} M_n(\theta_0) - M(\theta_0) &\geq M_n(\theta) - M(\theta) + M(\theta) - M(\theta_0) \\ &\geq M_n(\theta) - M(\theta) + \lambda[d(\theta, \theta_0)]^2. \end{aligned}$$

which implies that

$$(M_n(\theta_0) - M(\theta_0)) - (M_n(\theta) - M(\theta)) \geq \lambda[d(\theta, \theta_0)]^2 > 0.$$

Thus we get

$$|\Delta_n(\theta)| \geq \lambda[d(\theta, \theta_0)]^2 > \lambda \frac{2^{2j}}{r_n^2}.$$

Using this fact, we have for any $\epsilon > 0$

$$\begin{aligned} \sum_{j \geq t; 2^j \leq r_n \eta} \mathbb{P}[\hat{\theta}_n \in S_{n,j}] &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \mathbb{P}\left(\exists \theta \in S_{n,j} \text{ such that } |\Delta_n(\theta)| > \lambda \frac{2^{2j}}{r_n^2}\right) \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \mathbb{P}\left(\sup_{\theta \in S_{n,j}} |\Delta_n(\theta)| > \lambda \frac{2^{2j}}{r_n^2}\right) \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \frac{\mathbb{E}\left[\sup_{\theta \in S_{n,j}} |\Delta_n(\theta)|\right]}{\lambda \frac{2^{2j}}{r_n^2}} \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \frac{r_n^2}{\lambda 4^j} \frac{\phi\left(\frac{2^j}{r_n}\right)}{\sqrt{n}} \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \frac{cr_n^2}{\lambda \sqrt{n} 4^j} \left[\frac{2^j}{r_n}\right]^\alpha \\ &\leq \sum_{j \geq t; 2^j \leq r_n \eta} \frac{1}{\lambda 2^{(2-\alpha)j-\alpha}} \\ &\leq \epsilon \quad \text{take } t \text{ large enough} \end{aligned}$$

where to get the third inequality we have used the fact that $\theta \in S_{n,j}$ implies $d(\theta, \theta_0) \leq \frac{2^{j+1}}{r_n}$ and we have used the assumption that $cr_n^{2-\alpha} \leq \sqrt{n}$ to get the second inequality from the bottom.

2 Convergence in Distribution

2.1 Weak convergence: equivalent definition

$X_n \xrightarrow{d} X$ if and only if $E[f(X_n)] \rightarrow E[f(X)]$ for all bounded and (Lipschitz) continuous function f .

2.2 Tightness

Let \mathcal{D} be a metric space. A random variable $X : \Omega \rightarrow \mathcal{D}$ is tight if there is a compact set $K \subset \mathcal{D}$ such that $P(X \in K) > 1 - \epsilon$.

2.3 Asymptotically tight

A sequence of \mathcal{D} -valued random variables $\{X_n\}$ is asymptotically tight if for all $\epsilon > 0$, there is a compact set $K \subset \mathcal{D}$ such that $\limsup_{n \rightarrow \infty} P(X_n \notin K^\delta) < \epsilon$, where

$$K^\delta = \{y \in \mathcal{D} ; d(y, K) < \delta\}$$

for $\delta > 0$ and the metric d associated with the metric space \mathcal{D} .

2.4 Prokhorov's theorem

(a) If $X_n \xrightarrow{d} X$ and X is tight, then X_n is asymptotically tight.

(b) If X_n is asymptotically tight, then there exists a subsequence X_{n_k} and tight X such that $X_{n_k} \xrightarrow{d} X$.

2.5 L^∞ class of functions

For a compact set T , define $L^\infty(T)$ as the set of functions $f : T \rightarrow \mathbb{R}$ such that

$$\|f\|_\infty = \sup_{t \in T} |f(t)| < \infty.$$

2.6 Uniformly continuous functions

Suppose we have a metric ρ on $T \times T$. Define $UC(T, \rho)$ as the set of uniformly continuous functions $f : T \rightarrow \mathbb{R}$. We note that $UC(T, \rho) \subseteq L^\infty(T)$. Consider $\sqrt{n} \sup_{t \in \mathbb{R}} |F_n(t) - F_0(t)|$. Note that

$$\sqrt{n}(F_n(t) - F_0(t)) = \frac{1}{\sqrt{n}} \sum_{i=1}^n (\mathbf{1}\{X_i \leq t\} - F_0(t)).$$

Suppose that for any t_1, t_2, \dots, t_k ,

$$\sqrt{n}(F_n(t_1) - F_0(t_1), \dots, F_n(t_k) - F_0(t_k)) \xrightarrow{d} (\mathcal{G}_{F_0}(t_1), \dots, \mathcal{G}_{F_0}(t_k)).$$

Then,

$$\text{cov}(\mathcal{G}_{F_0}(t_i), \mathcal{G}_{F_0}(t_j)) = F_0(t_i \wedge t_j) - F_0(t_i)F_0(t_j).$$

We expect the empirical process $\sqrt{n}(F_n - F_0)$ converges in distribution to a Gaussian process \mathcal{G}_{F_0} with zero mean and covariance function as in the preceding display.

2.7 F_0 -Brownian bridge

The limit process \mathcal{G}_{F_0} is known as an F_0 -Brownian bridge. In particular, if F_0 is uniform on $[0, 1]$,

$$\text{cov}(\mathcal{G}_{F_0}(t_i), \mathcal{G}_{F_0}(t_j)) = t_i \wedge t_j - t_i t_j$$

and \mathcal{G}_{F_0} is called a standard Brownian bridge. For standard Brownian bridge, we can write

$$\mathcal{G}_{F_0}(t) = B(t) - tB(1)$$

where B is a standard Brownian motion.

2.8 Asymptotically equicontinuous: ASEC

Let $X_n \in L^\infty(T)$. We say that X_n 's are asymptotically equicontinuous if for all $\epsilon, \eta > 0$, there is a finite partition T_1, \dots, T_k of T such that

$$\limsup_{n \rightarrow \infty} P \left(\max_{1 \leq i \leq k} \sup_{s, t \in T_i} |X_{n,s} - X_{n,t}| \geq \epsilon \right) \leq \eta.$$

2.9 Example

Let $Z_i \in \mathbb{R}^d$ and $Z_i \sim^{i.i.d} P$. Assume $E\|Z_1\|^2 < \infty$ and $E[Z_i] = 0$. Define

$$X_{n,t} = \frac{1}{\sqrt{n}} \sum_{i=1}^n Z_i^\top t$$

and

$$T = \{t \in \mathbb{R}^d : \|t\| \leq M\}.$$

Then for small enough δ ,

$$\begin{aligned} P \left(\sup_{s, t \in T, \|s-t\| \leq \delta} |X_{n,s} - X_{n,t}| > \epsilon \right) &\leq P \left(\left\| \frac{1}{\sqrt{n}} \sum_{i=1}^n Z_i \right\| \delta \geq \epsilon \right) \\ &\leq \frac{\delta^2}{\epsilon^2} E \left[\left\| \frac{1}{\sqrt{n}} \sum_{i=1}^n Z_i \right\|^2 \right] = \frac{\delta^2}{\epsilon^2} E[\|Z_1\|^2] < \eta. \end{aligned}$$