

1 Review

Uniform Weak LLN

$$W_1, W_2, \dots \text{ iid } \in C(\underbrace{K}_{\text{compact}}) \quad (1.1)$$

$$\mathbb{E} \|W_i\|_\infty < \infty \quad (1.2)$$

$$\implies \bar{W}_n = \frac{1}{n} \sum_{i=1}^n W_i \xrightarrow{\text{unif } P} \mathbb{E} W_i \quad (1.3)$$

$$\text{i.e. } \|\bar{W}_n - \mathbb{E} W_1\|_\infty \xrightarrow{P} 0 \quad (1.4)$$

Theorem 1.1 (Keener 9.4). Let $G_n, n \geq 1$ be random functions in $C(K)$, K compact, and for some fixed $g \in C(K)$

$$\|G_n - g\|_\infty \xrightarrow{P} 0 \quad (1.5)$$

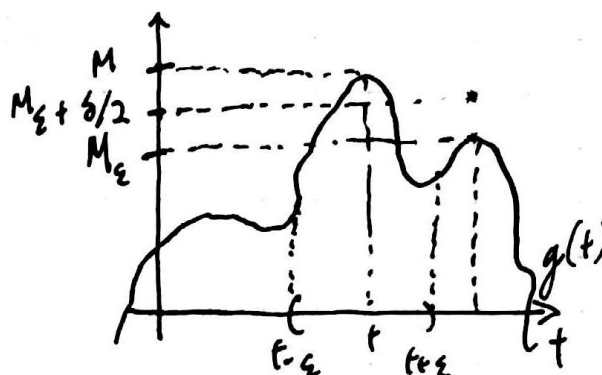
("G_n converging uniformly to g"), then

(a) If $t_n \xrightarrow{P} t^*$ (fixed), then $G_n(t_n) \xrightarrow{P} g(t^*)$

(b) If g has a unique maximizer $t^* \in K$, and $G_n(t_n) = \sup_{t \in K} G_n(t)$, then $t_n \xrightarrow{P} t^*$.

Proof. Item (a): Exercise

Item (b): Fix $\varepsilon > 0$, let $K_\varepsilon = K \setminus B_\varepsilon(t^*)$. Define $M = g(t^*) = \sup_{t \in K} g(t)$ and $M_\varepsilon = \sup_{t \in K_\varepsilon} g(t)$. Then $\delta = M - M_\varepsilon > 0$.



If $\|G_n - g\|_\infty \leq \delta/2$ (by \xrightarrow{P} , highly probable for suff large n), then

$$\sup_{t \in K_\varepsilon} G_n(t) < M_\varepsilon + \frac{\delta}{2} \quad (1.6)$$

$$\sup_{t \in K} G_n(t) \geq G_n(t^*) > M - \frac{\delta}{2} = M_\varepsilon + \delta/2 \quad (1.7)$$

$$P(t_n \in B_\varepsilon(t^*)) \geq P(\|G_n - g\|_\infty < \frac{\delta}{2}) \rightarrow 1 \quad (1.8)$$

□

Setup:

- $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} p_{\theta_0}(x)$
- $\theta_0 \in \Theta$
- $l_n(\theta; X) = \sum_{i=1}^n \log p_\theta(x_i)$
- Assume p_θ continuous in Θ , all p_θ distinct (model is *identifiable*)
- $\bar{W}_n = \frac{1}{n} l(\theta; X) - \frac{1}{n} l(\theta_0; X)$

Definition 1.2. The *Kullbeck-Leibler Divergence (KL-Divergence)* is

$$D_{KL}(\theta_0 \parallel \theta) = \mathbb{E}_{\theta_0} \log \frac{p_{\theta_0}(X)}{p_\theta(x)} = \mathbb{E}_{\theta_0}[-\bar{W}_n] \quad (1.9)$$

Lemma 1.3. If $P_\theta \neq P_{\theta_0}$ then $D_{KL}(\theta_0 \parallel \theta) > 0$

Proof. By Jensen's inequality

$$-D_{KL}(\theta_0 \parallel \theta) \leq \log \mathbb{E}_{\theta_0} \frac{p_\theta(X)}{p_{\theta_0}(X)} = \log \int_{x: p_{\theta_0}(x) > 0} \frac{p_\theta(x)}{p_{\theta_0}(x)} p_{\theta_0}(x) dx \leq \log 1 = 0 \quad (1.10)$$

When $p_\theta \neq p_{\theta_0}$, Jensen's inequality is tight and the two have common support so the second inequality is also tight. □

Theorem 1.4 (Consistency of MLE if Θ compact). Let $W_i(\theta) = \log p_\theta(X_i) - \log p_{\theta_0}(X_i)$. If

- Θ compact
- $\mathbb{E}_{\theta_0} \|W_1\|_\infty < \infty$
- $p_\theta(x)$ cts in Θ (for a.e. x)
- $p_\theta \neq p_{\theta_0}$ for all $\theta \neq \theta_0$ (Identifiable)

Then

$$\hat{\theta}_{MLE} \xrightarrow{P} \theta_0 \quad (1.11)$$

Proof. By definition $W_i(\theta) = \log \frac{p_\theta(X_i)}{p_{\theta_0}(X_i)}$. By assumption p_θ continuous, we have $W_i \in C(\Theta)$ so

$$\|\bar{W}_n + D_{KL}(\theta_0 \parallel \theta)\|_\infty \xrightarrow{P} 0 \quad (1.12)$$

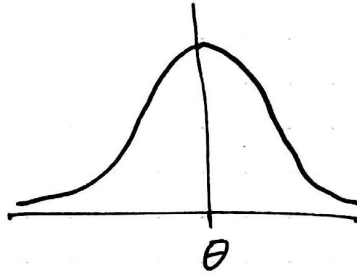
and we have uniform convergence. Also

$$\mathbb{E}_{\theta_0} W_i(\theta) = \underbrace{-D_{KL}(\theta_0 \parallel \theta)}_{\text{unique max at } \theta_0} \quad (1.13)$$

Applying theorem 1.1 yields $\hat{\theta}_{MLE} \xrightarrow{P} \theta_0$ □

Keener gives a weaker sufficient condition

$$\mathbb{E} \left[\sup_{\theta \in \Theta} \log \frac{p_\theta(X)}{p_{\theta_0}(X)} \right] < \infty \quad (1.14)$$



What if Θ not compact?

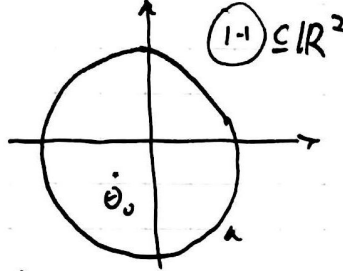
Theorem 1.5. Suppose

- $\Theta = \mathbb{R}^p$
- $p(\theta(x))$ cts in θ (a.e. x)
- $p_\theta \neq p_{\theta_0} \forall \theta \neq \theta_0$ (identifiable)
- $p_\theta(x) \rightarrow 0$ as $\theta \rightarrow \infty$

If

- (a) $\mathbb{E}_{\theta_0} \|1_K W_1\|_\infty < \infty, \forall \text{ compact } K \subset \Theta$
- (b) $\mathbb{E}_{\theta_0} \sup_{\|\theta\| > a} W_1(\theta) < \infty$ for some $a > 0$

Then $\hat{\theta}_n \xrightarrow{P} \theta_0$



Proof.

$$p_\theta(x) \xrightarrow{\text{a.s.}} 0 \quad \text{as } \theta \rightarrow \infty \quad (1.15)$$

$$\implies \sup_{\|\theta\| > b} W_i(\theta) \rightarrow -\infty \quad \text{as } b \rightarrow \infty \quad (1.16)$$

By dominated convergence

$$\mathbb{E}_{\theta_0} \left[\sup_{\|\theta\| > b} W_1(\theta) \right] \rightarrow -\infty \quad (1.17)$$

Choose $\delta > 0$ sufficiently large such that

$$\mathbb{E}_{\theta_0} \left[\sup_{\|\theta\| > b} W_1(\theta) \right] < -\delta < 0 \quad (1.18)$$

So

$$P_{\theta_0} \left(\sup_{\|\theta\| > b} \bar{W}_n(\theta) \geq -\delta \right) \rightarrow 0 \quad (1.19)$$

$$\sup_{\|\theta\| > b} \bar{W}_n(\theta) = \frac{1}{n} \sup_{\|\theta\| > b} \sum_{i=1}^n W_i(\theta) \quad (1.20)$$

$$= \frac{1}{n} \sum_{i=1}^n \sup_{\|\theta\| > b} W_i(\theta) \quad (1.21)$$

$$\rightarrow \mathbb{E}_{\theta_0} [\sup \dots] < \delta \quad (1.22)$$

Let $\tilde{\theta}_n = \arg \max_{\|\theta\| \leq b} \bar{W}_n(\theta)$, which is consistent ($\xrightarrow{P} \theta_0$) because $\|\theta\| \leq b$ is a compact (closed-bounded) set.

$$P_{\theta_0}(\tilde{\theta}_n \neq \hat{\theta}_n) = P_{\theta_0} \left(\sup_{\|\theta\| > b} \bar{W}_n(\theta) \geq \bar{W}_n(\theta_0) \right) \quad (1.23)$$

$$\leq P_{\theta_0} \left(\sup_{\|\theta\| > b} \bar{W}_n(\theta) \geq -\delta \right) + P_{\theta_0}(\bar{W}_n(\theta_0) \leq -\delta) \quad (1.24)$$

$$\rightarrow 0 \quad \text{both by LLN} \quad (1.25)$$

□

