

Homework 8 - Answers

1. Using the hint

$$\begin{aligned}
 P\left(\left|\frac{1}{n} \sum_{i=1}^n w_i - \mu\right| > \epsilon\right) &< \frac{1}{\epsilon^2} E\left\{\left(\frac{1}{n} \sum_{i=1}^n w_i - \mu\right)^2\right\} \\
 &= \frac{1}{\epsilon^2} V\left(\frac{1}{n} \sum_{i=1}^n w_i\right) \\
 &= \frac{1}{\epsilon^2} \frac{1}{n^2} n V(w_i) \\
 &= \frac{1}{\epsilon^2} \frac{1}{n} \sigma^2
 \end{aligned}$$

Taking limits as $n \rightarrow \infty$ on both sides gives

$$\lim_{n \rightarrow \infty} P\left(\left|\frac{1}{n} \sum_{i=1}^n w_i - \mu\right| > \epsilon\right) = 0$$

hence, by definition of convergence in probability

$$\frac{1}{n} \sum_{i=1}^n w_i \xrightarrow{P} \mu.$$

2 (a) Let $\{x_i^T\}_{i=1,2,\dots,n}$ be an IID sequence of vectors such that $E(x_{ik}^2) < \infty$. Then, $E(x_{ik}x_{ik'}) < \infty$.

Since, $\frac{x^T x}{n} = \frac{1}{n} \sum_{i=1}^n x_i^T x_i$, where $x_i^T x_i = [x_{ik} x_{ik'}]_{k,k'=1}^{K+1}$

By Kolmogorov's LLN $\frac{x^T x}{n} \xrightarrow{P} E(x_i^T x_i) = Q$.

(b) Similarly, if $\{u_i\}_{i=1,2,\dots,n}$ is IID with $E(u_i|x_i) = 0$

then, $\frac{x^T u}{n} = \frac{1}{n} \sum_{i=1}^n x_i^T u_i$ and $E(x_i^T u_i) = 0$.

Hence, by Kolmogorov's LLN $\frac{x^T u}{n} \xrightarrow{P} 0$. (a vector).

$$(c) \hat{\beta} = \beta + \left(\frac{x^T x}{n}\right)^{-1} \frac{x^T u}{n}$$

Since $\frac{x^T x}{n} \xrightarrow{P} Q$ and $\frac{x^T u}{n} \xrightarrow{P} 0$, then

$$\hat{\beta} \xrightarrow{P} \beta + Q^{-1} \cdot 0 = \beta$$

$$(d) \sqrt{n} (\hat{\beta} - \beta) = \left(\frac{x^T x}{n}\right)^{-1} \frac{x^T u}{\sqrt{n}}$$

By Levy's CLT, $\frac{x^T u}{\sqrt{n}} \xrightarrow{d} N(0, Q)$. Since, for

any $w_n \xrightarrow{P} w$ and $v_n \xrightarrow{d} v$ we have $w_n v_n \xrightarrow{d} w v$.

Then, $\sqrt{n} (\hat{\beta} - \beta) \xrightarrow{d} Q^{-1} z$, $z \sim N(0, Q)$.

Hence, $Q^{-1} z \sim N(0, Q^{-1} Q Q^{-1}) := N(0, \sigma^2 Q^{-1})$.

3. From Lévy's CLT

$$\frac{\frac{1}{n} \sum_{i=1}^n X_i - \mu}{\sqrt{\frac{\sigma^2}{n}}} \xrightarrow{d} Z \sim N(0,1)$$

thus $s^2 \xrightarrow{P} \sigma^2$

$$\frac{\frac{1}{n} \sum_{i=1}^n X_i - \mu}{\sqrt{\frac{s^2}{n}}} \xrightarrow{d} Z$$

Thus, under $H_0: \mu = c$

$$t_n = \frac{\frac{1}{n} \sum_{i=1}^n X_i - c}{\sqrt{\frac{s^2}{n}}} \xrightarrow{d} Z$$

Then, if $|t_n| \leq t_{1-\frac{\alpha}{2}}^*$ accept H_0 , where

$t_{1-\frac{\alpha}{2}}^*$ is the $1 - \frac{\alpha}{2}$ quantile associated with a standard normal, and $\alpha \in (0,1)$. Otherwise, reject H_0 .