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consistent estimator

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Owner	CWoo (3771)
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Author	CWoo (3771)
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Defines	consistent sequence of estimators

Given a set of samples  $X_1, \dots, X_n$  from a given probability distribution  $f$  with an unknown parameter  $\theta \in \Theta$ , where  $\Theta$  is the parameter space that is a subset of  $\mathbb{R}^m$ . Let  $U(= U(X_1, \dots, X_n))$  be an estimator of  $\theta$ . Allowing the sample size  $n$  to vary, we get a sequence of estimators for  $\theta$ :

$$\begin{aligned} U_1 &= U(X_1), \\ &\vdots \\ U_n &= U(X_1, \dots, X_n), \\ &\vdots \end{aligned}$$

We say that the sequence of estimators  $\{U_n\}$  *consistent* (or that  $U$  is a *consistent estimator* of  $\theta$ ), if  $U_i$  converges in probability to  $\theta$  for *every*  $\theta \in \Theta$ . That is, for every  $\varepsilon > 0$ ,

$$\lim_{n \rightarrow \infty} P(|h_n - \theta| \geq \varepsilon) = 0$$

for all  $\theta \in \Theta$ .

**Remark.** Suppose  $U$  is an estimator of  $\theta$  such that the sequence  $\{U_n\}$  is consistent. If  $\alpha_n \rightarrow \alpha \in \mathbb{R}$  and  $\beta_n \rightarrow \beta \in \mathbb{R}^m$  are two convergent sequences of constants with  $0 < |\alpha| < \infty$  and  $|\beta| < \infty$ , then the sequence  $\{V_n\}$ , defined by  $V_n := \alpha_n U_n + \beta_n$ , is consistent,  $V$  is an estimator of  $\alpha\theta + \beta$ .

*Proof.* First, observe that

$$\begin{aligned} |V_n - (\alpha\theta + \beta)| &= |\alpha_n U_n + \beta_n - \alpha\theta - \beta| \\ &\leq |\alpha_n U_n - \alpha\theta| + |\beta_n - \beta| \\ &= |\alpha_n U_n - \alpha_n \theta + \alpha_n \theta - \alpha\theta| + |\beta_n - \beta| \\ &\leq |\alpha_n U_n - \alpha_n \theta| + |\alpha_n \theta - \alpha\theta| + |\beta_n - \beta| \\ &= |\alpha_n| |U_n - \theta| + |\alpha_n - \alpha| |\theta| + |\beta_n - \beta|. \end{aligned}$$

This implies

$$\begin{aligned} &P(|V_n - (\alpha\theta + \beta)| \geq \varepsilon) \\ &\leq P(|\alpha_n| |U_n - \theta| + |\alpha_n - \alpha| |\theta| + |\beta_n - \beta| \geq \varepsilon) \\ &= P(|U_n - \theta| \geq \frac{\varepsilon - |\beta_n - \beta| - |\alpha_n - \alpha| |\theta|}{|\alpha_n|}). \end{aligned}$$

As  $n \rightarrow \infty$ ,  $|\beta_n - \beta| \rightarrow 0$ ,  $|\alpha_n - \alpha||\theta| \rightarrow 0$ , and  $|\alpha_n| \rightarrow |\alpha| \neq 0$ . So the last expression goes to 0 as  $n \rightarrow \infty$ . Therefore,

$$\lim_{n \rightarrow \infty} P(|V_n - (\alpha\theta + \beta)| \geq \varepsilon) = 0,$$

and thus  $\{V_n\}$  is a consistent sequence of estimators of  $\alpha\theta + \beta$ .  $\square$