

## 1 Convergence in Probability

**Def 1.1** (Convergence in Probability).  $Y_n \xrightarrow{p} c$  if for every  $\epsilon > 0$  and  $\delta > 0$ ,  $\exists n_0(\epsilon, \delta)$  such that

$$P(|Y_n - c| > \epsilon) < \delta, \forall n > n_0(\epsilon, \delta)$$

**Thm 1.1** (Chebyshev Inequality). For random variable,  $Y$ ,  $a > 0$ , and  $c$ ,

$$P(|Y - c| \geq a) \leq \frac{\mathbf{E}(Y - c)^2}{a^2}$$

**Def 1.2** (Markov Inequality). If  $X$  is a non-negative random variable and  $a > 0$  then

$$P(X \geq a) \leq \frac{\mathbf{E}X}{a}$$

**Thm 1.2.** If  $\mathbf{E}(Y - c)^2 \rightarrow 0$ , then  $Y_n \xrightarrow{p} c$ .

**Thm 1.3.** If  $X_1, \dots, X_n$  iid,  $\mathbf{E}X_i = \mu$ ,  $\mathbf{Var}X_i = \sigma^2 < \infty$ , then

$$\bar{X} \xrightarrow{p} \mu$$

**Thm 1.4.** If  $A_n \xrightarrow{p} a$  and  $B_n \xrightarrow{p} b$ , then

1.  $A_n \pm B_n \xrightarrow{p} a \pm b$ ,
2.  $A_n \cdot B_n \xrightarrow{p} a \cdot b$ .

**Thm 1.5.** If  $Y_n \xrightarrow{p} c$  and  $f$  is continuous at  $c$ , then  $f(Y_n) \xrightarrow{p} f(c)$ .

**Def 1.3.** A sequence of estimators  $\delta_n$  of  $g(\theta)$  is *consistent* if

$$\delta_n \xrightarrow{p} g(\theta)$$

**Thm 1.6.** If bias and variance of  $\delta_n \rightarrow 0$  as  $n \rightarrow \infty$ ,  $\delta_n$  is consistent.

**Def 1.4.**  $A_n = o_p(B_n)$  if  $\frac{A_n}{B_n} \xrightarrow{p} 0$ .

## 2 Convergence in Law

**Def 2.1.**  $Y_n$  is *bounded in probability* if for any  $\epsilon > 0 \exists K$  and  $n_0(\epsilon)$  such that

$$P(|Y_n| > K) < \epsilon \text{ for all } n > n_0(\epsilon)$$

**Def 2.2.** Convergence in Law/Distribution  $Y_n \xrightarrow{L} Y$  if  $H_n(x) \rightarrow H(x)$  for all continuity points  $x$ .