

Q1 Acc. to convergence in probability:-  $X_n \xrightarrow{p} X$  as  $n \rightarrow \infty$  if

$$P(|X_n - X| \geq \epsilon) = 0 \text{ as } n \rightarrow \infty \quad \forall \epsilon > 0$$

$$\text{we know } \bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

where each  $X_i$  is iid  $U(a, b)$ .

$$\therefore E(X_i) = \frac{a+b}{2} \quad \Delta \quad V(X_i) = \frac{(b-a)^2}{12}$$

Now, Acc. to Chebyshev's Inequality.

$$P(|\bar{X}_n - E(\bar{X}_n)| \geq \epsilon) \leq \frac{V(\bar{X}_n)}{\epsilon^2}$$

$$\Rightarrow P\left(\left|\bar{X}_n - \frac{a+b}{2}\right| \geq \epsilon\right) \leq \frac{(b-a)^2}{12n\epsilon^2}$$

$$\text{Now, as } n \rightarrow \infty \Rightarrow \frac{(b-a)^2}{12n\epsilon^2} = 0$$

$$\Rightarrow P\left(\left|\bar{X}_n - \frac{a+b}{2}\right| \geq \epsilon\right) \leq 0$$

$$\text{Sat as } P() \geq 0$$

$$\Rightarrow P\left(\left|\bar{X}_n - \frac{a+b}{2}\right| \geq \epsilon\right) = 0$$

$\therefore$  Acc to convergence in probability:-  $\bar{X}_n \xrightarrow{p} \frac{a+b}{2}$  as  $n \rightarrow \infty$

$\therefore$  The required value of  $C = \frac{a+b}{2}$ .

Q2 Given  $\bar{X}_n \xrightarrow{p} 0 \Rightarrow E(X) = 0$

$$\text{as } X_i \text{ is iid } U(a, b) \Rightarrow E(X_i) = \frac{a+b}{2}$$

$$\Rightarrow \frac{a+b}{2} = 0 \Rightarrow \boxed{a = -b}$$

$$\begin{aligned} \therefore E(\bar{X}_n) &= E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] \\ &= \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \cdot n \left(\frac{a+b}{2}\right) \\ \text{Also, } V(\bar{X}_n) &= \frac{\text{Var}(X)}{n} \end{aligned}$$

Q.3: Given  $\bar{X}_n \xrightarrow{P} 0 \Rightarrow E(X) = 0$   
 as  $X_i$  is iid  $U(-2, b) \Rightarrow E(X_i) = \frac{-2+b}{2}$   
 $\Rightarrow \frac{-2+b}{2} = 0 \Rightarrow \boxed{b = +2}$

Q.4 Given  $\bar{X}_n \xrightarrow{P} 1 \Rightarrow E(X) = 1$   
 as  $X_i$  is iid  $U(-2, b) \Rightarrow E(X_i) = \frac{-2+b}{2}$   
 $\Rightarrow \frac{-2+b}{2} = 1 \Rightarrow \boxed{b = 4}$

Q.5: Given  $\bar{X}_n \xrightarrow{P} 0$  as  $n \rightarrow \infty \Rightarrow E(X) = 0$   
 as  $X_i$  is iid  $U(a, 1) \Rightarrow E(X) = \frac{a+1}{2}$   
 $\Rightarrow \frac{a+1}{2} = 0 \Rightarrow \boxed{a = -1}$

Q.6 Given  $X_i$  are iid  $N(0, 1) \Rightarrow E(X_i) = 0$  &  $V(X_i) = 1$   
 $\Rightarrow E(\bar{X}_n) = E\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \frac{1}{n} \cdot n \cdot (0) = 0$

$\downarrow V(\bar{X}_n) = V(X)/n = 1/n$

Acc. to Chebyshev's Inequality

$$P(|\bar{X}_n - E(\bar{X}_n)| \geq \varepsilon) \leq \frac{V(\bar{X}_n)}{\varepsilon^2}$$

$$\Rightarrow P(|\bar{X}_n - 0| \geq \varepsilon) \leq \frac{1}{n\varepsilon^2}$$

as  $n \rightarrow \infty$

$$\Rightarrow P(|\bar{X}_n - 0| \geq \varepsilon) = 0$$

$\therefore$  Acc. to convergence in prob.

$$\bar{X}_n \xrightarrow{P} 0 \quad \therefore \boxed{C = 0}$$

Q.7 We know K-WLLN,

$\bar{X}_n \xrightarrow{P} E(\bar{X}_n)$  as  $n \rightarrow \infty$   
 $\star E(\bar{X}_n) = \mu$  which is in  $N(\mu, \sigma^2)$

$$\therefore E(\bar{X}_n) = E\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \frac{1}{n} \cdot n \cdot E(X_i)$$

As each  $X_i$  is iid  $N(\mu, \sigma^2)$



$$\Rightarrow E(\bar{X}_n) = \frac{1}{n} \cdot n \cdot \mu = \mu$$

Given  $\bar{X}_n \xrightarrow{P} 0$  as  $n \rightarrow \infty$

$$\therefore E(\bar{X}_n) = \boxed{\mu = 0}$$

Q.8 Also,  $\bar{X}_n \xrightarrow{P} 5$  as  $n \rightarrow \infty$  & from WLLN

$$\bar{X}_n \xrightarrow{P} E(\bar{X}_n) \text{ as } n \rightarrow \infty, \text{ \& we know}$$

$$E(\bar{X}_n) = E(X) = \mu$$

$$\therefore E(\bar{X}_n) = \boxed{\mu = 5}$$

Q.9 Given  $X_i$  iid  $B(2, 1/2)$

$$\Rightarrow E(X_i) = np = 2 \times 1/2 = 1$$

$$\& V(X_i) = np(1-p) = 2 \times 1/2 \times 1/2 = 1/2$$

$$\text{Also } E(\bar{X}_n) = E\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \frac{1}{n} \cdot n \cdot 1 = 1$$

$$\& V(\bar{X}_n) = V\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} \sum_{i=1}^n V(X_i) = \frac{1}{n^2} \cdot n \cdot V(X_i)$$

$$= \frac{V(X_i)}{n} = \frac{1}{2n}$$

Acc. to Chebyshev's Inequality,

$$P(|\bar{X}_n - E(\bar{X}_n)| \geq \epsilon) \leq \frac{V(\bar{X}_n)}{\epsilon^2}$$

$$\Rightarrow P(|\bar{X}_n - 1| \geq \epsilon) \leq \frac{1}{2n\epsilon^2}$$

$$\text{as } n \rightarrow \infty$$

$$\Rightarrow P(|\bar{X}_n - 1| \geq \epsilon) \leq 0 \text{ \& as } P(\cdot) \geq 0$$

$$\Rightarrow P(|\bar{X}_n - 1| \geq \epsilon) = 0$$

$\therefore$  Acc. to convergence in prob  $\Rightarrow \bar{X}_n \xrightarrow{P} 1$

$$\therefore \boxed{C=1}$$

Q.10. As  $X_i$  iid  $B(a, 1/2)$  & from K-WLLN

$$\bar{X}_n \xrightarrow{p} 3 \text{ as } n \rightarrow \infty \Rightarrow E(\bar{X}_n) = 3$$

Also  $E(\bar{X}_n) = E(X) = np = a/2$

$$\Rightarrow 3 = a/2 \Rightarrow \boxed{a=6}$$

Q.11. As  $X_i$  iid  $B(a, 1/3)$  & from K-WLLN

$$\bar{X}_n \xrightarrow{p} 5 \text{ as } n \rightarrow \infty \Rightarrow E(\bar{X}_n) = 5$$

Also  $E(\bar{X}_n) = E(X) = np = a/3$

$$\Rightarrow 5 = \frac{a}{3} \Rightarrow \boxed{a=15}$$

Q.12. As  $X_i$  iid  $B(10, p)$  &  $E(X_n) = 2$  (from K-WLLN)

Also  $E(\bar{X}_n) = E(X) = np = 10p$

$$\therefore 2 = 10p \Rightarrow \boxed{p = \frac{1}{5}}$$