

Ek 381 HW 9

Chase
Mandel
V18719879

(a) Markov's inequality: $P(\bar{X} \geq c) \leq \frac{E[\bar{X}]}{c}$

$$\rightarrow P(\bar{X} \geq 6) \leq 3/6 = \boxed{1/2}$$

$$P(\bar{X} \geq 6) = 1 - (P(X=0) + P(X=1) + \dots + P(X=5)) \\ = 1 - 0.916082 = \boxed{0.08}$$

The upper bound ($1/2$) > the exact value (0.08), satisfying the inequality.

b) $P(|X - E[\bar{X}]| \geq c) \leq \frac{\text{Var}(\bar{X})}{c^2}$ • Chebyshev's inequality

Using Chebyshev's inequality $\rightarrow P(|X - E[\bar{X}]| \geq 2) \leq \frac{3/2}{2^2} = \boxed{3/4}$

$$P(|X - 3| \geq 2) = P(X \geq 5) = 1 - (P(X=0) + \dots + P(X=4)) \\ = 1 - 0.815263 \\ = 0.184737 \\ = P(X \leq 1) = P(X=0) + P(X=1) \\ = 0.199$$

$$P(X \geq 5) + P(X \leq 1) = \boxed{0.383}$$

The upper bound ($3/4$) is closer to the exact value (0.383) in the case of Chebyshev's inequality compared to Markov's inequality.

c) Weak Law of Large Numbers: $P\left[\left|\frac{1}{n} \sum_{i=1}^n X_i - E[\bar{X}]\right| \geq c\right] \leq \frac{\text{Var}(\bar{X})}{nc^2}$

Weak law population mean \rightarrow sample mean

Chebyshev's inequality implies $P(|X - E[\bar{X}]| \geq c) \leq \frac{\text{Var}(\bar{X})}{c^2}$

$$\text{where } \rightarrow P(|X - E[\bar{X}]| \geq c) \leq \frac{300}{60^2} = \boxed{0.0833}$$

may serve as the upper bound

1d) Central Limit Theorem: $\bar{W}_n(w) \approx \phi\left(\frac{w - E(\bar{X})}{\sqrt{n \text{Var}(X)}}\right)$

where W_{100} is a Normal w/mean 300
& $\text{Var} = 100(3) = 300$

$$\begin{aligned} P[|W_{100} - 300| \geq 60] &= P[W_{100} > 360] + P[W_{100} < 240] \\ &= 0.000266 + 0.000266 \\ &= \underline{0.000532} \end{aligned}$$

2) $N_i = \#$ of flips until one head $i = 1, 2, \dots, 100$

Given $N_i \sim \text{Geometric}(\frac{1}{2})$

$$E[N_i] = \sum_{i=1}^{\infty} i \cdot \frac{1}{2}^i = \sum_{i=1}^{\infty} i p^i \quad \text{where } p = \frac{1}{2}$$

$$p(1-p)^{-2} = \frac{1}{2}(1-\frac{1}{2})^{-2} = \frac{1}{2}(\frac{1}{2})^{-2} = p + 2p^2 + 3p^3 + \dots$$

$$= (\frac{1}{2})^{-1} = p(1 + 2p + 3p^2 + \dots)$$

$$\underline{E[N_i] = 2} = p(1-p)^{-2}$$

$$E[N_i^2] = \sum_{i=1}^{\infty} i^2 \cdot \frac{1}{2}^i = \sum_{i=1}^{\infty} i^2 p^i \quad \text{where } p = \frac{1}{2}$$

$$= \sum_{i=1}^{\infty} (i(i-1) + i) p^i$$

$$= \sum_{i=1}^{\infty} (i(i-1)p^i) + \sum_{i=1}^{\infty} ip^i$$

$$= p^2 \sum_{i=1}^{\infty} (i(i-1)p^{i-2}) + E[N_i]$$

$$= p^2 \sum_{i=2}^{\infty} \frac{d^2}{dp^2} p^i + 2$$

$$= p^2 \frac{d^2}{dp^2} \left(\frac{p}{1-p}\right) + 2$$

$$= \frac{2p^2}{(1-p)^3} + 2 = \underline{6}$$

By the Central Limit Theorem, $N_T \sim N(E[N_T], \text{Var}[N_T])$
 $\sim N(100, 200)$

$$\rightarrow P[N_T \leq 80] = P\left[\frac{N_T - 100}{\sqrt{200}} \leq \frac{80 - 100}{\sqrt{200}}\right]$$

$$= P\left[\frac{N_T - 100}{\sqrt{200}} \leq -1.414\right]$$

$$= \underline{\phi(-1.414) = 0.079}$$

$$\text{Var}[N_i] = E[N_i^2] - (E[N_i])^2 = 6 - 2^2 = 2$$

$$E[N_T] = E\left[\sum_{i=1}^{100} N_i\right] = 100 = 200 - 100 = 100$$

$$\text{Var}[N_T] = \text{Var}\left[\sum_{i=1}^{100} N_i - 100\right] = \sum_{i=1}^{100} \text{Var}[N_i] = 200$$

$$3a) P\left[\frac{1}{n} \left| \sum_{k=1}^n X_k - p \right| \geq c\right] \leq \frac{1}{4nc^2}$$

$$\frac{1}{4nc^2} = 0.955, n = \frac{1}{4(0.02)^2(0.955)} = \boxed{654}$$

$$b) 0.02 > Z \sqrt{\frac{0.5(1-0.5)}{n}} \Rightarrow \underline{n > 2500}$$

b)

$$4a) T_i \sim \text{exponential}(\lambda)$$

$$E[T_i] = 100$$

$$1/\lambda = 100 \Rightarrow \lambda = 1/100$$

$$\text{Var}[T_i] = 1/\lambda^2 = 1/(1/100)^2 = \boxed{10000}$$

$$b) T = \frac{1}{10000} \sum T_i$$

$$E[T] = 1/10000 \cdot E[\sum T_i]$$

$$= 1/10000 \cdot 10000 \cdot 100 = \boxed{100}$$

$$c) V[T] = \left(\frac{1}{10000}\right)^2 \cdot V[\sum T_i]$$

$$= \left(\frac{1}{10000}\right)^2 \cdot 10000 \cdot 10000 = 1$$

$$\text{By CLT, } T \sim N(E[T], V[T])$$

$$P[T > 110] = P\left[\frac{T - E[T]}{\sqrt{V[T]}} > \frac{110 - 100}{1}\right]$$

$$= P[Z > 10]$$

$$= 1 - P[Z < 10]$$

$$= \underline{1 - \Phi(10)}$$