

Proof: Let $X = (Y - \mu)^2$. X is nonnegative, so

$$P(|Y - \mu| \geq \alpha) = P((Y - \mu)^2 \geq \alpha^2)$$

$$= P(X \geq \alpha^2) \leq E[X] / \alpha^2 \quad (\text{Markov})$$

$$= E[(Y - \mu)^2] / \alpha^2 = \text{Var}(Y) / \alpha^2$$

Ex: If Y is daily business cost, with $E[Y] = 1500 = \mu$ and $\sigma = \sqrt{\text{Var}(Y)} = 200$.

$$P(Y \geq 2500) = P(Y - \mu \geq 1000) \leq P(|Y - \mu| \geq 1000)$$

$$\leq \frac{\text{Var}(Y)}{1000^2} = \frac{(200)^2}{1000^2} = \frac{1}{25}$$

Cantelli's Inequality: (one-sided Chebyshev)

$$P(Y - \mu \geq \alpha) \leq \frac{\text{Var}(Y)}{\text{Var}(Y) + \alpha^2}$$

Ex: Same example:

$$P(Y \geq 2500) = P(Y - \mu \geq 1000) \leq \frac{200^2}{200^2 + 1000^2}$$

$$= \frac{1}{1 + \left(\frac{1000}{200}\right)^2} = \frac{1}{26}$$

Chernoff Bound:

Theorem: Suppose $X \sim \text{Bin}(n, p)$ and $\mu = E[X]$.

For any $0 < \delta < 1$,

$$P(X \geq (1 + \delta)\mu) \leq e^{-\frac{1}{3}\delta^2\mu} \quad \text{and}$$

$$P(X \leq (1 - \delta)\mu) \leq e^{-\frac{1}{2}\delta^2\mu}$$

Ex: $X \sim \text{Bin}(10, 1/2)$, $\delta = 0.4$, $\mu = 10 \cdot \frac{1}{2} = 5$.

$$P(X \geq 1.4\mu) = P(X \geq 7) \leq e^{-\frac{1}{3}(0.4)^2 \cdot 5}$$

Law of Large Numbers

Consider iid r.v.'s X_1, X_2, X_3, \dots , where $E[X_i] = \mu < \infty$ and $\text{Var}(X_i) = \sigma^2 < \infty$.

Define sample mean $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$.

$$\begin{aligned} E[\bar{X}_n] &= E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] = \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \sum_{i=1}^n \mu \\ &= \frac{1}{n} \cdot n\mu = \mu, \end{aligned}$$

$$\begin{aligned} \text{Var}(\bar{X}_n) &= \text{Var}\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} \text{Var}\left(\sum_{i=1}^n X_i\right) \\ &= \frac{1}{n^2} \sum_{i=1}^n \text{Var}(X_i) = \frac{1}{n^2} \sum_{i=1}^n \sigma^2 = \frac{1}{n^2} \cdot n\sigma^2 = \frac{\sigma^2}{n} \end{aligned}$$

As n increases, \bar{X}_n is more likely to be close to μ .

Theorem (Weak Law of Large Numbers):

For any $\varepsilon > 0$, as $n \rightarrow \infty$,

$$P(|\bar{X}_n - \mu| > \varepsilon) \rightarrow 0.$$

Proof: By Chebyshev's inequality,

$$P(|\bar{X}_n - \mu| > \varepsilon) \leq \frac{\sigma^2}{n\varepsilon^2} \rightarrow 0 \text{ as } n \rightarrow \infty$$