

▣ Markov's Inequality

If X is a random variable that takes only non-negative values, then, for any value $a > 0$,

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

▣ Chebyshev's Inequality

If X is a random variable with finite mean μ & variance σ^2 , then, for any value $k > 0$

$$P\{|X - \mu| \geq k\} \leq \frac{\sigma^2}{k^2}$$

Example:

Suppose that it is known that the number of items produced in a factory during a week is a random variable with mean 50.

(a) What can be said about the prob. that this week's production will exceed 75.

(b) If the variance of a week's production is known to equal 25, then what can be said about the probability that this week's production will be between 40 & 60?

⇒ (a) By Markov's inequality,

$$P\{X > 75\} \leq \frac{E(X)}{75} = \frac{50}{75} = \frac{2}{3}$$

(b) By Chebyshev's inequality

$$P\{|X - 50| \geq 10\} \leq \frac{\sigma^2}{10^2} = \frac{1}{4}$$

$$\text{hence, } P\{|X - 50| < 10\} = 1 - \frac{1}{4} = \frac{3}{4}$$

Which is the reqd. prob. that week's production will be
b/w 40 260

■ The weak law of large numbers

Let X_1, \dots, X_n be a seqn. of independent & identically distributed random variables, (i.e. if each random variable has the same probability distribution as others and all are mutually independent). each having finite mean $E(X_i) = \mu$, then for each $\epsilon > 0$,

$$P \left\{ \left| \frac{X_1 + \dots + X_n}{n} - \mu \right| \geq \epsilon \right\} \rightarrow 0 \text{ as } n \rightarrow \infty$$

■ Central limit theorem

Let X_1, X_2, \dots be a seqn. of independent & identically distributed random variables, each having mean μ & variance σ^2 . Then the distribution of

$$\frac{X_1 + \dots + X_n - n\mu}{\sigma\sqrt{n}}$$

tends to standard normal as $n \rightarrow \infty$. That is for

$$-\infty < a < \infty, \quad P \left\{ \frac{X_1 + \dots + X_n - n\mu}{\sigma\sqrt{n}} \leq a \right\} \rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-\frac{z^2}{2}} dz \text{ as } n \rightarrow \infty.$$