

Lecture-18

Probability & Statistics.

Moment generating function.
(MGF)

$$M(t) = E[e^{tX}] .$$

Markov's Inequality.

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

$$\begin{array}{c} X \geq 0 \\ a \geq 0 \end{array}$$

Chebyshov's inequality.

$$E[X] = \mu$$

$$P(|X-\mu| \geq a) \leq \frac{\sigma^2}{a^2}, \quad a > 0.$$

$$\text{var}(X) = \sigma^2$$

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

a) What can be said about the probability 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?

Ans. a) $\frac{2}{3}$

b)

$$-⑤ \quad P\{(X-50) > 10\} \leq \frac{\sigma^2}{a^2} = \frac{25}{10^2} = \frac{1}{4}$$

$$\begin{aligned} P\{|X-50| \leq 10\} \\ = P\{40 \leq X \leq 60\} \end{aligned}$$

$$\geq 1 - \frac{1}{4} = \frac{3}{4}.$$

Law of Large numbers.

Weak law of large numbers.

It states that the probability that the average of the first n terms in a seq: of independent & identically distributed random variables differs by its mean by more than ϵ goes to 0 as $n \rightarrow \infty$.

$$P\left\{ \left| \frac{X_1 + X_2 + \dots + X_n}{n} - \mu \right| > \epsilon \right\} \rightarrow 0 \text{ as } n \rightarrow \infty$$

$$E[X_i] = \mu, \quad \text{var}(X_i) = \sigma^2$$

$$E\left[\frac{X_1 + X_2 + \dots + X_n}{n}\right] = \mu$$

$$\text{Var} \left[\frac{X_1 + X_2 + \dots + X_n}{n} \right] = \frac{\sigma^2}{n}$$

$$P\left\{ \left| \frac{X_1 + X_2 + \dots + X_n}{n} - \mu \right| > \epsilon \right\} \leq \frac{\sigma^2}{n\epsilon^2}$$

↙
by Chebyshev's
inequality

as $n \rightarrow \infty$
↓
get. 0

Strong law of large numbers.

Thm. let X_1, X_2, \dots, X_n be a seqe of
independent random variables having
a common distribution, and let
 $E[X_i] = \mu$. Then with prob. 1

$$\frac{X_1 + X_2 + \dots + X_n}{n} \rightarrow \mu \text{ as } n \rightarrow \infty$$

As an example, suppose a seq: of independent trials is performed.
 Let E be a fixed event and denote
 $P(E)$ the prob. that E occurs
 on any particular trial.

Let $X_i = \begin{cases} 1 & \text{if } E \text{ occurs on the } i\text{th trial} \\ 0 & \text{if } E \text{ does not occur on the } i\text{th trial.} \end{cases}$

By Strong law of large numbers

$$\frac{X_1 + X_2 + \dots + X_n}{n} \rightarrow E[X_i] = P(E)$$

when n is large -

Central Limit Theorem.



The number one theorem
in probability.

Let X_1, X_2, \dots be a seqⁿ of
independent, identically distributed
random variables, each with
mean μ and variance σ^2 .

Then the distribution of

$$\frac{X_1 + X_2 + \dots + X_n - n\mu}{\sigma\sqrt{n}}$$

tends to the standard normal
as $n \rightarrow \infty$.

$$P \left\{ \frac{X_1 + X_2 + \dots + X_n - n\mu}{\sigma\sqrt{n}} \leq a \right\} \rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-x^2/2} dx$$

Let $X_i, i = 1, 2, \dots, 10$ be independent random variables, each being uniformly distributed over $(0, 1)$.

Estimate $P\left\{\sum_{i=1}^{10} X_i > 7\right\}$,

$$\mu = \frac{1}{2}, \sigma^2 = \frac{1}{12}$$

$$P\left\{\sum_{i=1}^{10} X_i > 7\right\}$$

$$P\left\{\frac{\sum_{i=1}^{10} X_i - 5}{\sqrt{10 \cdot \frac{1}{12}}} > \frac{7-5}{\sqrt{10 \cdot \frac{1}{12}}}\right\}$$

$$P\left\{\downarrow > \right\}$$

$$1 - \Phi(2.2)$$

$$= 0.0139.$$