

Random Variable

- Discrete, continuous and mixed random variable
- probability mass, probability density and cumulative distribution function
- mathematical expectation, moments, moment generating function
- median and quantiles
- Chebyshev's inequality
- Problems

4 lectures

Example 1 (Tossing a coin)

$$\Omega = \{H, T\}, \quad X(H) = 1 \quad X(T) = 0$$

P assigns equal masses to $\{H\}$ & $\{T\}$

$$P(X=0) = \frac{1}{2} = P(X=1)$$

$$F(x) = Q(-\infty, x] = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1}{2} & \text{if } 0 \leq x < 1 \\ 1 & \text{if } 1 \leq x. \end{cases}$$

Example 2 (Tossing two dice)

$$F(x) = Q(-\infty, x] = P(X \leq x) = \begin{cases} 0 & \text{if } x < 2 \\ \frac{1}{36} & \text{if } 2 \leq x < 3 \\ \frac{3}{36} & \text{if } 3 \leq x < 4 \\ \frac{6}{36} & \text{if } 4 \leq x < 5 \\ \vdots & \vdots \\ \frac{35}{36} & \text{if } 11 \leq x < 12 \\ 1 & \text{if } 12 \leq x \end{cases}$$

Example 3 Rolling two dice, yielding X and Y . Note combined score $Z = X + Y$.

$$2 \leq Z(\omega) \leq 12.$$

$$\{ \omega : Z(\omega) \leq 12 \} = \begin{cases} \emptyset & Z(\omega) < 2 \\ \{(1,1)\} & 2 \leq Z(\omega) < 3 \\ \{(1,2), (2,1)\} & 3 \leq Z(\omega) < 4 \\ \vdots & \vdots \\ \{(6,6)\} & Z(\omega) = 12 \end{cases}$$

Example 4 Devise experiment that selects a pt. P randomly from the interval $[0, 2]$, where any pt. may be chosen. Let us write formally,

$$\Omega = \{ \omega : \omega \in [0, 2] \}$$

Define X and Y

$$X(\omega) = \begin{cases} 0 & 0 \leq \omega \leq 1 \\ 1 & 1 < \omega \leq 2 \end{cases}$$

and $Y(\omega) = \omega^2$.

X can take only 2 values \rightarrow discrete

Y can take uncountable no. in $[0, 4]$ \rightarrow continuous.

Example 5 . Let $\Omega = [0, 1]$

Define X on Ω as follows:

$$X(\omega) = \begin{cases} \omega & \text{if } 0 \leq \omega \leq \frac{1}{2} \\ \omega - \frac{1}{2} & \text{if } \frac{1}{2} < \omega \leq 1 \end{cases}$$

Define $Y : \left\{ \omega : X(\omega) \in \left(\frac{1}{4}, \frac{1}{2} \right) \right\}$

then $Y(\omega)$ takes values $\left(\frac{1}{4}, \frac{1}{2} \right) \cup \left(\frac{3}{4}, 1 \right)$

Example 6 You ask people whether they approve the present government. The sample space could be

$$\Omega = \{ \text{approve strongly, approve, indifferent, disapprove, disapprove strongly} \}$$

You may put in numerical scale

$$X = \{ -2, -1, 0, 1, 2 \}$$

or

$$Y = \{ 0, 1, 2, 3, 4 \}$$

X, Y etc. are Random Variable.

Introduction to Random Variable.

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Experimental outcomes are not always numerical

↓

Assign a number to every outcome $\omega \in \Omega$
and denote by $X(\omega)$

(Advantage: convenient to work)

Example 1 Roll a fair die. $\Omega = \{1, 2, 3, 4, 5, 6\}$

X : Number shown

$$X(\omega) = \omega, 1 \leq \omega \leq 6$$

Y : Number of sixes

$$Y(\omega) = \begin{cases} 1 & \text{if } \omega = 6 \\ 0 & \text{otherwise} \end{cases}$$

Example 2 Flip three coins.

$$\Omega = \{HHH, HHT, HTH, HTT, TTH, THT, TTT\}$$

X : Number of heads

$$X(\omega) \text{ takes } \{0, 1, 2, 3\}$$

$$\text{as } X(HHH) = 3 \quad X(TTT) = 0$$

Y : Signed difference betⁿ no. of heads & tails

$$Y(\omega) \text{ takes } \{-3, -2, -1, 0, 1, 2, 3\}$$

$$\text{e.g. } Y(TTH) = -1$$

Definition

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A RV is a variable whose possible values are numerical outcomes of a random phenomenon.

Let (Ω, β, P) be a probability space.
A function X from Ω to \mathbb{R} is a RV if it is measurable.

Let \mathcal{C} be a class of subsets of \mathbb{R} so that \mathcal{C} be a σ -field

i.e. for any $B \in \mathcal{C}$, $X^{-1}(B) \in \beta$

Probability distⁿ of RV

The RV X defined on the probability space (Ω, β, P) induces a probability space $(\mathbb{R}, \mathcal{C}, Q)$ s.t.

$$Q(B) = P\{X^{-1}(B)\} = P\{\omega : X(\omega) \in B, \omega \in \Omega, B \in \mathcal{C}\}$$

Then $Q = PX^{-1}$ is the probability distⁿ of X .

* Prove that Q or PX^{-1} is a probability measure

Proof $P_1 : Q(B) \geq 0 \quad \forall B \in \mathcal{C}$

$$P_2 : Q(\mathbb{R}) = P(\Omega) = 1$$

$P_3 : \text{Let } B_i \in \mathcal{C} \text{ are pairwise disjoint}$

$$Q\left(\bigcup_{i=1}^{\infty} B_i\right) = P\left\{X^{-1}\left(\bigcup_{i=1}^{\infty} B_i\right)\right\}$$

$$= \sum_{i=1}^{\infty} P X^{-1}(B_i)$$

$$= \sum_{i=1}^{\infty} Q(B_i)$$

$\Rightarrow (\mathbb{R}, \mathcal{C}, Q)$ is a probability space.

Cumulative distribution function

$F_X(x)$ is called cumulative distⁿ fⁿ of RV X
 if $F_X(x) = P(X \leq x)$

Where $P(X \leq x) = P\{\omega : X(\omega) \leq x\}$

Properties

- $\lim_{x \rightarrow -\infty} F_X(x) = 0$

- $\lim_{x \rightarrow +\infty} F_X(x) = 1$

- If $x_1 < x_2$ then $F_X(x_1) < F_X(x_2)$
 (i.e. F is nondecreasing)

- $\lim_{h \rightarrow 0} F_X(x+h) = F_X(x)$
 (F is right continuous at every pt.)

1. Let $\{x_n\}$ be a decreasing sequence s.t. $\lim x_n = -\infty$

$$\text{Let } A_n = \{\omega : x(\omega) \leq x_n\}$$

$$\Rightarrow \lim_{n \rightarrow \infty} A_n = \emptyset$$

$$\Rightarrow \lim_{n \rightarrow \infty} P(A_n) = P(\emptyset) \Rightarrow \lim_{n \rightarrow \infty} F(x_n) = 0$$

Again,

$$P\left(\lim_{n \rightarrow \infty} A_n\right)$$

$$\text{or, } \lim_{x_n \rightarrow -\infty} F(x_n) = 0$$

2. Let $\{x_n\}$ be an increasing sequence such that $\lim x_n = +\infty$

Similarly, we can prove that

$$\lim P(A_n) = \lim P(\Omega) = 1.$$

So, Property 2 follows.

3. Given $x_1 < x_2$

$$\Rightarrow \{\omega : x(\omega) \leq x_1\} \subset \{\omega : x(\omega) \leq x_2\}$$

$$\Rightarrow P\{\omega : x(\omega) \leq x_1\} \leq P\{\omega : x(\omega) \leq x_2\}$$

$$\Rightarrow F(x_1) \leq F(x_2)$$

$\Rightarrow F$ is non-decreasing.

4. Let $\{x_n\}$ be a decreasing sequence such that $\lim_{n \rightarrow \infty} x_n = x$.

Take $A_n = \{\omega : x < x(\omega) \leq x_n\}$

$\Rightarrow A_n$ is decreasing sequence.

$$\Rightarrow \lim_{n \rightarrow \infty} A_n = \bigcap_{n=1}^{\infty} A_n = \emptyset$$

$$\begin{aligned} \Rightarrow \lim_{n \rightarrow \infty} P(A_n) &= P(\lim_{n \rightarrow \infty} A_n) \\ &= \lim_{n \rightarrow \infty} [F(x_n) - F(x)] = 0 \end{aligned}$$

$$\Rightarrow \lim_{n \rightarrow \infty} F(x_n) = F(x)$$

$$\Rightarrow \lim_{h \rightarrow 0} F(x+h) = F(x).$$

Discrete and continuous RV

A RV X defined on a probability space (Ω, β, P) is said to be discrete if \exists a countable set $E \subset \mathbb{R}$ s.t. $P(X \in E) = 1$.

\Rightarrow Range of X is countable

The points of E which have positive mass are called jump points and size of jump is the probability of RV taking at that point

Probability Mass fcn (PMF) of discrete RV

$$P_X(x_i) = P(X = x_i), \quad x_i \in E$$

Where,

$$(i) \quad P_X(x_i) > 0$$

$$(ii) \quad \sum_{x_i \in E} P_X(x_i) = 1$$

Relation betⁿ PMF & CDF

$$F_X(x) = \sum_{x_i \leq x} P_X(x_i)$$

$$\begin{aligned} \text{Thus, } P_X(x_i) &= P_X(x_i \leq x_i) - P_X(x < x_{i-1}) \\ &= F_X(x_i) - F_X(x_{i-1}). \end{aligned}$$

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 A computer store contains 10 computers of which 3 are defectives. A customer buys 2 at random. Find PMF and CDF.

Let the RV X be no. of defectives in purchase

$$\Rightarrow X \in \{0, 1, 2\}$$

$$P_X(X=0) = \frac{{}^7C_2}{{}^{10}C_2} = \frac{21}{45}$$

$$P_X(X=1) = \frac{{}^7C_1 {}^3C_1}{{}^{10}C_2} = \frac{21}{45}$$

$$P_X(X=2) = \frac{{}^3C_2}{{}^{10}C_2} = \frac{3}{45}$$

↑
PMF

$$F_X(X \leq 0) = P(X \leq 0) = 0$$

$$F_X(X \leq 0) = P(X \leq 0) = \frac{21}{45}$$

$$F_X(X \leq 1) = P(X \leq 0) + P(0 < X \leq 1) = \frac{44}{45}$$

$$F_X(X \leq 2) = P(X \leq 1) + P(1 < X \leq 2)$$

↑ = 1
CDF

Let X be the number of heads in three tosses of a coin. What is Ω ? What are the ~~the~~ probability of events $(X \leq 2.75)$ and $(0.5 \leq X \leq 1.72)$.

Write PMF and CDF of RV X .

x	0	1	2	3
$P(X=x)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

PMF

$$F(x) = \begin{cases} 0 & x < 0 \\ \frac{1}{8} & 0 \leq x < 1 \\ \frac{1}{2} & 1 \leq x < 2 \\ \frac{7}{8} & 2 \leq x < 3 \\ 1 & x \geq 3 \end{cases}$$

 Find whether the following F are CDF or not.

$$(a) F(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x < \frac{1}{2} \\ 1 & x \geq \frac{1}{2} \end{cases}$$

$$(b) F(x) = \frac{1}{\pi} \tan^{-1} x$$

$$-\infty < x < \infty$$

$$(c) F(x) = \begin{cases} 0 & x \leq 1 \\ 1 - \frac{1}{x} & x > 1 \end{cases}$$

$$(d) F(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq \frac{1}{2} \\ 1 & x > \frac{1}{2} \end{cases}$$

Continuous RV

RVs associated with distribution f_X that have no jump points.

Probability Density f_X (PDF)

Let X be a RV defined on (Ω, \mathcal{B}, P) with distribution f_X . Then X is said to be continuous type if F is absolutely continuous, i.e.

if \exists non-negative f_X s.t.

$$F_X(x) = \int_{-\infty}^x f_X(x) dx \quad \forall \text{ real } x.$$

Then f_X is called prob. density f_X of RV X

Properties

$$\bullet P(a \leq x \leq b) = F_X(b) - F_X(a) = \int_a^b f_X(x) dx.$$

$$\bullet \int_{-\infty}^{\infty} f_X(x) dx = 1.$$

$$\bullet \text{ If } F \text{ is absolutely cont \& } f \text{ is cont at } x$$


$$\frac{d}{dx} F_X(x) = f_X(x).$$

Note


For discrete RV $P(X=a) \Rightarrow$ prob that X takes a

In cont. RV $f(a) \neq$ prob that X takes a

Indeed, if X is cont, it assumes every value with probability 0.

 CDF given $F_X(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq 1 \\ 1 & x > 1 \end{cases}$

$$F'_X(x) = f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

 PDF given $f_X(x) = \begin{cases} \frac{10}{x^2} & x > 10 \\ 0 & x \leq 10 \end{cases}$

$$F_X(x) = \int_{-\infty}^x f_X(t) dt = \begin{cases} 0 & x < 10 \\ \int_{10}^x \frac{10}{t^2} dt = \left(1 - \frac{10}{x}\right) & x \geq 10 \end{cases}$$

$$P(15 < X < 20) = \int_{15}^{20} \frac{10}{x^2} dx = \frac{1}{6}.$$

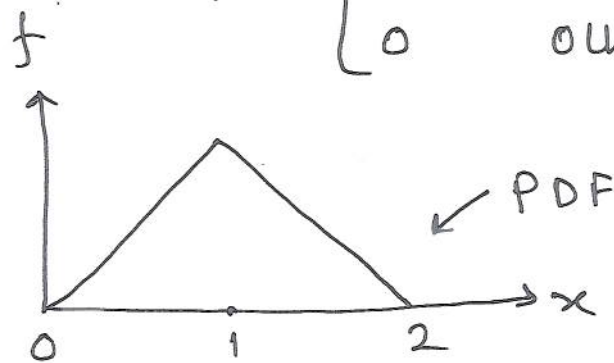
$$= F_X(20) - F_X(15)$$

check!!



Given
PDF

$$f_x(x) = \begin{cases} x & 0 < x \leq 1 \\ 2-x & 1 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$



$$\underline{\underline{F_x(x)}}$$

$$= \begin{cases} 0 & x \leq 0 \\ \int_0^x t dt = \frac{x^2}{2} & 0 < x \leq 1 \\ \int_0^1 t dt + \int_1^x (2-t) dt & 1 < x \leq 2 \\ 1 & x \geq 2 \end{cases}$$

Find. $P(-3 \leq x \leq 1.5)$

$$= .83.$$



Given $f_x(x) = \begin{cases} \frac{x}{2} & 0 \leq x \leq 1 \\ \frac{1}{2} & 1 < x \leq 2 \\ \frac{3-x}{2} & 2 < x \leq 3 \end{cases}$

Find $F_x(x)$. Hence find $P\left(\frac{1}{2} \leq x \leq \frac{5}{2}\right)$.

$$F_x(x) = \begin{cases} \frac{x^2}{4} & 0 \leq x < 1 \\ \frac{1}{4} + \frac{x-1}{2} & 1 \leq x < 2 \\ 1 - \frac{(3-x)^2}{4} & 2 \leq x < 3 \\ 1 & x \geq 3 \end{cases}$$

■ The diameter of an electric cable, say X , is assumed to be a continuous RV with PDF

$$f(x) = 6x(1-x) \quad 0 \leq x \leq 1.$$

(i) Check that $f(x)$ is PDF

(ii) Determine a number b such that $P(x < b) = P(x > b)$
(Ans. $b = \frac{1}{2}$)



Let x be a continuous RV with following pdf

$$f(x) = \begin{cases} ax & 0 \leq x < 1 \\ a & 1 \leq x \leq 2 \\ -ax + 3a & 2 \leq x \leq 3 \\ 0 & \text{elsewhere} \end{cases}$$

Determine a . Compute $P(X \leq 1.5)$

(Ans. $a = \frac{1}{2}$ & $P = \frac{1}{2}$)

Symmetric Distribution

A RV X is symmetric about a point α if

$$P(X \geq \alpha + x) = P(X \leq \alpha - x) \quad \forall x$$

or

$$F(\alpha - x) = 1 - F(\alpha + x) + P(X = \alpha + x)$$

If $\alpha = 0$


$$F(-x) = 1 - F(x) + P(X = x)$$

If X is continuous,

$$F(-x) = 1 - F(x)$$

$$\text{or } f(-x) = f(x)$$

$$\text{or, } f(\alpha - x) = f(\alpha + x)$$

 $P(X = -1) = \frac{1}{4}$, $P(X = 0) = \frac{1}{2}$, $P(X = 1) = \frac{1}{4}$.

Symmetric about 0.

Moments

Mathematical Expectation

Let X be discrete RV with pmf $P_X(x_i)$, $x_i \in \Omega$.
We define expected value of X as

$$E(X) = \sum_{x_i} x_i P_X(x_i)$$

provided the series is absolutely convergent.

$$\begin{aligned} E(X) &= 0 \cdot \frac{21}{45} + 1 \cdot \frac{21}{45} + 2 \cdot \frac{3}{45} \\ &= \frac{27}{45} \end{aligned}$$

$$P_X(0) = \frac{21}{45}$$

$$P_X(1) = \frac{21}{45}$$

$$P_X(2) = \frac{3}{45}$$

$$\begin{aligned} P_X(X=1) &= P_X(X=2) = \frac{1}{4}, \quad P_X(X=3) = \frac{1}{2} \\ E(X) &= 1 \cdot \frac{1}{4} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{2} = \frac{9}{4} \end{aligned}$$

If X is continuous RV with pdf $f_X(x)$

$$E(X) = \int_{-\infty}^{\infty} x f_X(x) dx$$

provided the integral is absolutely convergent

$$\boxed{\text{■}} \quad f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$E(X) = \int_0^1 x dx = \frac{1}{2}$$

$$\boxed{\text{■}} \quad f_X(x) = \frac{10}{x^2} \quad x > 10$$

$$E(X) = \int_{10}^{\infty} x \cdot \frac{10}{x^2} dx = 10 \ln x \Big|_{10}^{\infty} \rightarrow \text{divergent}$$

$$\boxed{\text{■}} \quad f_X(x) = \frac{1}{\pi} \frac{1}{1+x^2} \quad -\infty < x < \infty$$

But $E(X)$ is divergent.

Properties

$$= \text{If } Y = aX + b, \text{ then } E(Y) = aE(X) + b.$$

$$= \text{If } Y = g(x), E(Y) = E(g(x))$$

$$= \begin{cases} \sum x_i g(x_i) P_X(x_i) \\ \int_{-\infty}^{\infty} g(x) f_X(x) dx \end{cases}$$

$$= \text{Variance} = E(X - \mu)^2 = E(X^2) - \{E(X)\}^2$$

Moments

$$\boxed{\mu'_k = E(X^k)} \rightarrow k^{\text{th}} \text{ moment about origin or non-central moment}$$

$$\boxed{\mu_k = E(X - \mu)^k} \rightarrow k^{\text{th}} \text{ central moment}$$

$$\mu_2 = E(X - \mu)^2 = \text{variance of } X.$$

$$\begin{aligned}
 \mu_k &= E(x - \mu)^k \\
 &= E \left[x^k - {}^k C_1 x^{k-1} \mu + {}^k C_2 x^{k-2} \mu^2 + \dots + (-1)^k \mu^k \right] \\
 &= \mu_k' - {}^k C_1 \mu_{k-1}' \mu + {}^k C_2 \mu_{k-2}' \mu^2 + \dots + (-1)^k \mu^k
 \end{aligned}$$

\Rightarrow

$$\begin{aligned}
 \mu_2 &= \mu_2' - 2\mu_1' \mu + \mu^2 \\
 &= \mu_2' - \tilde{\mu}^2 = E(x^2) - \{E(x)\}^2
 \end{aligned}$$

\Rightarrow As $\mu_2 \geq 0$, $E(x^2) \geq \{E(x)\}^2$ Note

Absolute Moment

$$\begin{aligned}
 \beta_k' &= E|x|^k \\
 \beta_k &= E|x - \mu|^k
 \end{aligned}$$



X has uniform distribution of first N natural numbers

$$P(X = k) = \frac{1}{N}, \quad k = 1, 2, \dots, N$$

Find mean and variance

$$\text{Hint: } E(X) = \sum_{k=1}^N k \cdot \frac{1}{N} = \frac{N+1}{2}$$

$$E(X^2) = \sum_{k=1}^N k^2 \cdot \frac{1}{N} = \frac{(N+1)(2N+1)}{6}$$

Find mean and variance of the following

$$f_X(x) = \begin{cases} \frac{2}{x^3} & x \geq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Hint: } E(X) = \int_1^{\infty} x \cdot \frac{2}{x^3} dx = 2$$

$$E(X^2) = \int_1^{\infty} x^2 \cdot \frac{2}{x^3} dx \rightarrow \text{divergent}$$

\Rightarrow Lower order moment may exist.

Properties

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= If X and Y are two RVs then

$$E(X \pm Y) = E(X) \pm E(Y)$$

(Addition Rule)

= If X and Y are RVs
(Multiplication Rule)

$$E(XY) = E(X)E(Y)$$

if X & Y are ind.

$$= E(X)E(Y) + \text{Cov}(X, Y)$$

if X & Y are dep.

Covariance (Defn)

$$\begin{aligned}\text{Cov}(X, Y) &= E[(X - E(X))(Y - E(Y))] \\ &= E(XY - XE(Y) - YE(X) + E(X)E(Y)) \\ &= E(XY) - E(X)E(Y)\end{aligned}$$

$\text{Cov}(X, Y) > 0 \Rightarrow Y$ increases as X increases

$\text{Cov}(X, Y) < 0 \Rightarrow Y$ decreases as X increases

Moment Generating Function (mgf)

The mgf about origin is defined as

$$M_X(t) = E(e^{tx}) = \begin{cases} \sum_{x_i} e^{tx_i} p_X(x_i) & X \text{ is discrete} \\ \int_{-\infty}^{\infty} e^{tx} f_X(x) dx & X \text{ is continuous} \end{cases}$$

$$\begin{aligned} M_X(t) &= E \left[1 + tx + \frac{t^2}{2!} X^2 + \dots + \frac{t^r}{r!} X^r + \dots \right] \\ &= 1 + t E(X) + \frac{t^2}{2!} E(X^2) + \dots + \frac{t^r}{r!} E(X^r) + \dots \\ &= 1 + t \mu_1' + \frac{t^2}{2!} \mu_2' + \dots + \frac{t^r}{r!} \mu_r' + \dots \end{aligned}$$

Thus,

$$\mu_r' = \left. \frac{d^r}{dt^r} M_X(t) \right|_{t=0}$$

Find the mgf of the exponential distⁿ.

$$f(x) = a e^{-ax}, \quad 0 \leq x < \infty, \quad a > 0$$

Hence find its means and SD.

By defⁿ:

$$M_X(t) = E(e^{tx}) = \int_0^{\infty} a e^{(t-a)x} dx$$

$$= \left(1 - \frac{t}{a}\right)^{-1}$$

$$= 1 + \frac{t}{a} + \frac{t^2}{a^2} + \frac{t^3}{a^3} + \dots$$

$$\mu_1' = \left. \frac{d}{dt} M_X(t) \right|_{t=0} = \frac{1}{a}$$

$$\mu_2' = \left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0} = \frac{2}{a^2}$$

$$\Rightarrow \text{Mean} = \frac{1}{a}, \quad \text{SD} = \sqrt{\mu_2' - (\mu_1')^2} = \frac{1}{a}.$$

Find mgf of a Bernoulli Variable X with pmf

$$P(X=1)=p \quad P(X=0)=q \quad , \quad p+q=1.$$

Hence find its mean and variance.

$$\begin{aligned} M_X(t) &= E(e^{tx}) = \sum e^{tx_i} P_X(x_i) \\ &= pe^t + q \end{aligned}$$

$$\Rightarrow \mu_1' = p \quad \text{and} \quad \mu_2' = q$$

$$\Rightarrow \text{Mean} = p$$

$$\text{Variance} = \mu_2' - (\mu_1')^2 = p(1-p) = pq.$$

Distribution	Moment Generating f ⁿ .
Binomial (n, p)	$(1-p+pet)^n$
Poisson (λ)	$e^\lambda(e^t-1)$
Uniform (a, b)	$\frac{e^{tb} - e^{ta}}{t(b-a)}$

Chebyshev's Inequality

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If X is RV with mean μ and variance σ^2 then for any value $k > 0$

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

Proof.

$$\sigma^2 = \text{var}(X) = E(X - \mu)^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f_X(x) dx$$

$$\geq \int_{|x - \mu| \geq k} (x - \mu)^2 f_X(x) dx$$

$$\geq k^2 \int_{|x - \mu| \geq k} f_X(x) dx$$

$$\geq k^2 P(|X - \mu| \geq k)$$

Replacing k by $k\sigma$

$$P(|X - \mu| \geq k\sigma) \leq \frac{1}{k^2}$$

Another form

$$P(|X - \mu| < k\sigma) \geq 1 - \frac{1}{k^2}$$

Markov Inequality

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

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

Proof.

$$\begin{aligned} E(X) &= \int_0^{\infty} x f(x) dx = \int_0^a x f(x) dx + \int_a^{\infty} x f(x) dx \\ &\geq \int_a^{\infty} x f(x) dx \\ &\geq a \int_a^{\infty} f(x) dx \\ &= a P(X \geq a). \end{aligned}$$

▨ Suppose that it is known that the number of items produced in a factory during a week is RV 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 and 60?

Ans.

$$(a) \quad P(X \geq 75) \leq \frac{E(X)}{75} = \frac{50}{75} = \frac{2}{3}.$$

using Markov's inequality

$$(b) \quad P(|X - 50| < 10) \geq 1 - \frac{1}{4} = \frac{3}{4}$$

as

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

using Chebyshev's inequality

Weak Law of Large Numbers

Let x_1, x_2, \dots, x_n be a sequence of iid RVs, each having mean μ (i.e. $E(x_i) = \mu$). Then for any $\epsilon > 0$

$$P \left\{ \left| \frac{x_1 + x_2 + \dots + x_n}{n} - \mu \right| > \epsilon \right\} \rightarrow 0 \text{ as } n \rightarrow \infty$$

$$E \left(\frac{x_1 + x_2 + \dots + x_n}{n} \right) = \mu \quad \text{Var} \left(\frac{x_1 + x_2 + \dots + x_n}{n} \right) = \frac{\sigma^2}{n}$$

By Chebyshev's inequality

$$P \left\{ \left| \frac{x_1 + x_2 + \dots + x_n}{n} - \mu \right| > \epsilon \right\} \leq \frac{\sigma^2}{n \epsilon^2}$$

$$\Rightarrow \text{as } n \rightarrow \infty, \text{ LHS} \rightarrow 0.$$