

Today

1) Wrap up Characteristic Function (or MGF)

2) Two Random Variables

- Joint CDF & Joint PDF
- Marginal CDF & Marginal PDF
- Independent Random Variables.

Characteristic Function

$$\Phi_x(w) = E[e^{jwX}]$$

w is real
 $\Phi_x(w)$ is complex valued

if X is discrete

$$\Phi_x(w) = \sum_{x_i \in X} P(X=x_i) e^{jw x_i}$$

if X is continuous

$$\Phi_x(w) = \int_{-\infty}^{\infty} e^{jw x} f_x(x) dx$$

how to compute the characteristic function.

Moment Generating Function

$$M_X(s) = E[e^{sx}]$$

s is real
 $M_X(s)$ is real valued

$$\Phi_x(s) / M_X(t)$$

Computing the n^{th} moment of X from either $\Phi_x(w)$ or $M_X(s)$.

$$\rightarrow \frac{d}{dw} \Phi_x(w) = \frac{d}{dw} \left[\sum_k P(X=k) e^{jw k} \right] \quad \text{E}[e^{jwX}]$$

$$= \sum_k P(X=k) \cdot \frac{d(e^{jw k})}{dw}$$

$$= \sum_k P(X=k) \cdot (jk) \cdot e^{jw k}$$

$$\frac{1}{j} \frac{d}{dw} \Phi_x(w) = \sum_k P(X=k) \cdot k \cdot e^{jw k}$$

$$E[X] = \sum_k P(X=k) \cdot k$$

Evaluating @ $w = 0$

$$\left. \frac{d}{j} \frac{d \Phi_x(w)}{dw} \right|_{w=0} = \sum_k k \times P(x=k) = E[x].$$

$$\frac{d^2 \Phi_x(w)}{dw^2} = \sum_k P(x=k) (jk)^2 e^{jwk}.$$

$$\left. \frac{1}{j^2} \frac{d^2 \Phi_x(w)}{dw^2} \right|_{w=0} = \sum_k P(x=k) k^2 e^{jwk}.$$

$$E[X^2] = \left. \frac{1}{j^2} \frac{d^2 \Phi_x(w)}{dw^2} \right|_{w=0},$$

$$E[X^n] = \left. \frac{1}{j^n} \frac{d^n \Phi_x(w)}{dw^n} \right|_{w=0}.$$

nth moment of X

Eg Expected Value of a Binomial r.v.

$$\Phi_x(w) = \sum_{k=0}^n P(x=k) \cdot e^{jwk}$$

$P(x=k) = \binom{n}{k} p^k (1-p)^{n-k}$
 $k = 0, 1, 2, \dots, n$

$$= \sum \binom{n}{k} p^k (1-p)^{n-k} e^{jwk}$$

$$= \sum_{k=0}^n \binom{n}{k} (pe^{jw})^k \cdot (1-p)^{n-k}.$$

$$= (pe^{jw} + (1-p))^n$$

$$\sum_{k=0}^n \binom{n}{k} a^k b^{n-k} = (a+b)^n$$

$$E[X] = \frac{1}{j} \left. \frac{d \Phi_X(\omega)}{d\omega} \right|_{\omega=0}$$

$$= \frac{1}{j} n \left(p e^{j\omega} + (1-p) \right)^{n-1} \times p \times j \cdot e^{j\omega} \Big|_{\omega=0}$$

$$= np \times (p + (1-p))^{n-1}$$

$$E[X] = \boxed{np}$$

$$E[X^2] = \dots$$

$$\boxed{\text{Var}(X) \stackrel{?}{=} E[X^2] - (E[X])^2} = np(1-p)$$

try this yourself.

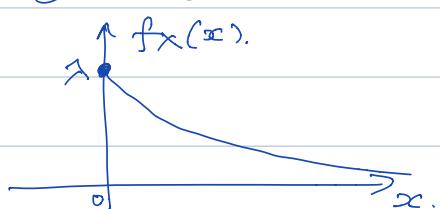
Eg 2 n^{th} moment of an Exp(λ)
 Exponential r.v.
 w/ parameter λ .

$$\Phi_X(\omega) = E[e^{j\omega X}]$$

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{otherwise.} \end{cases}$$

$$= \int_0^\infty e^{j\omega x} \times \lambda e^{-\lambda x} dx$$

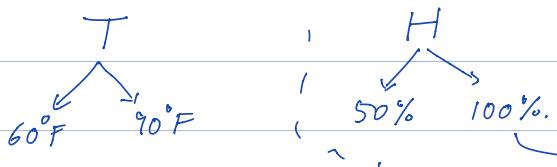
$$= \lambda \int_0^\infty (-\lambda + j\omega)x dx$$



$$\boxed{\Phi_X(\omega) = \frac{\lambda}{\lambda - j\omega}} \quad (\text{See my notes....})$$

Two Random Variables

Temp & Humidity.



$$P(T=60^{\circ}\text{F}) = 0.3$$

$$P(T=70^{\circ}\text{F}) = 0.7$$

$$P(H=50\%) = 0.8$$

$$P(H=100\%) = 0.2$$

For discrete r.v.'s.

Joint PMF

$$\begin{aligned} P(T=60, H=50) &=? & 0.1 \\ P(T=60, H=100) &=? & 0.2 \\ P(T=70, H=50) &=? & 0.6 \\ P(T=70, H=100) &=? & 0.1 \end{aligned}$$

(T, H)

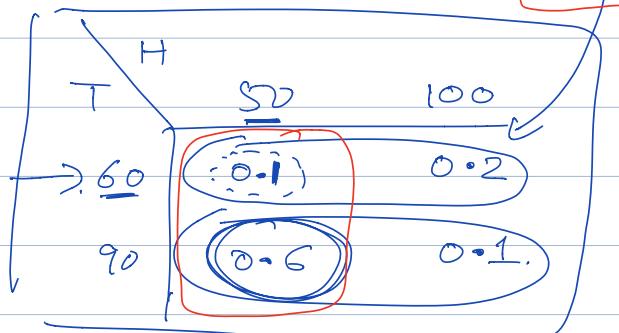
$$S = \{(60, 50), (60, 100), (70, 50), (70, 100)\}$$

T	H
60	50
60	100
70	50

mutually
exclusive
events.

$$P(S) = 1.$$

joint
PMF
=



$$P(T=70) = 0.7$$

$$P(H=50) =$$

$$\begin{aligned} P(T=60) &=? \\ &= P(T=60, H=50) + P(T=60, H=100) \\ &= 0.1 + 0.2 = 0.3. \end{aligned}$$

$$\left\{ \begin{array}{l} P(H=50) = P(H=50, T=60) + P(H=50, T=90) \\ \quad = 0.1 + 0.6 = 0.7. \\ P(H=100) = 0.3. \\ \text{``Marginal'' behavior of } H. \end{array} \right.$$

X, Y Joint PMF
(suppose X, Y are discrete valued)

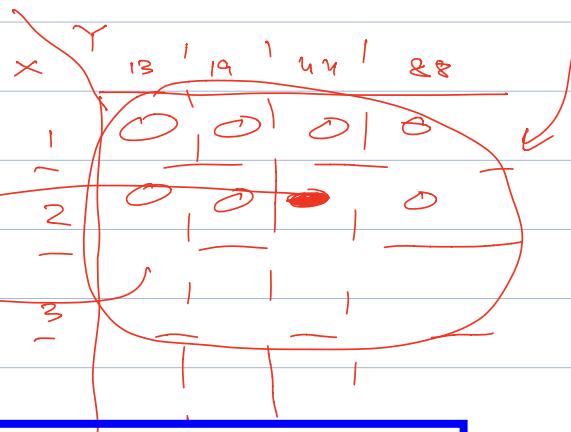
$$P_{X,Y}(x,y) = P(X=x, Y=y) \quad \text{for } x \in X, y \in Y.$$

$$X = \{1, 2, 3\}.$$

$$Y = \{13, 19, 44, 88\}.$$

$$P(X=2, Y=44).$$

$$\boxed{\text{Sum of all entries} = 1.}$$



$$\text{First property : } \sum_{y \in Y} \sum_{x \in X} P(x=x, y=y) = 1.$$

$P_X(x) \rightarrow$ Marginal PMF of X

$P_Y(y) \rightarrow$ " Marginal PMF of Y .

Computing Marginal PMF of X .

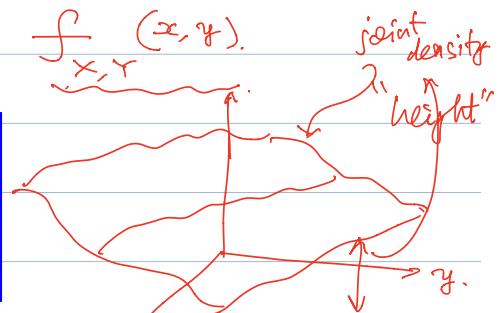
$$\underbrace{P_X(x)}_{\substack{\\ \text{Total prob. theorem.}}} = \underbrace{P(X=x)}_{\substack{\\ \uparrow y \in Y}} = \sum_{y \in Y} P(X=x, Y=y).$$

$$P_Y(y) = \underbrace{P(Y=y)}_{\substack{\\ \uparrow x \in X}} = \sum_{x \in X} P(X=x, Y=y).$$

Continuous Valued X, Y

Joint PDF of $(X, Y) \Rightarrow f_{X,Y}(x, y)$. joint density "height"

$$\int_{y=-\infty}^{\infty} \int_{x=-\infty}^{\infty} f_{X,Y}(x, y) dx dy = 1.$$



Marginal PDF of X

$$f_X(x) = \int_{y=-\infty}^{\infty} f_{X,Y}(x, y) dy$$

$$f_Y(y) = \int_{x=-\infty}^{\infty} f_{X,Y}(x, y) dx$$

Joint CDF of two random Variables

$$F_{X,Y}(x, y) = P(X \leq x, Y \leq y)$$

Comma = "and" or \cap

$$F_X(x) = P(X \leq x).$$

$$F_Y(y) = P(Y \leq y).$$

$$F_{X,Y}(\infty, \infty) = ? = 1.$$

$$F_{X,Y}(-\infty, \infty) = ? = P(X \leq -\infty, Y \leq \infty) = P(X \leq -\infty) = F_X(-\infty).$$

$$F_{X,Y}(\infty, y) = F_Y(y).$$

$$F_{X,Y}(-\infty, y) = ? = 0 \quad \text{---}$$

$$F_{X,Y}(x, -\infty) = 0.$$

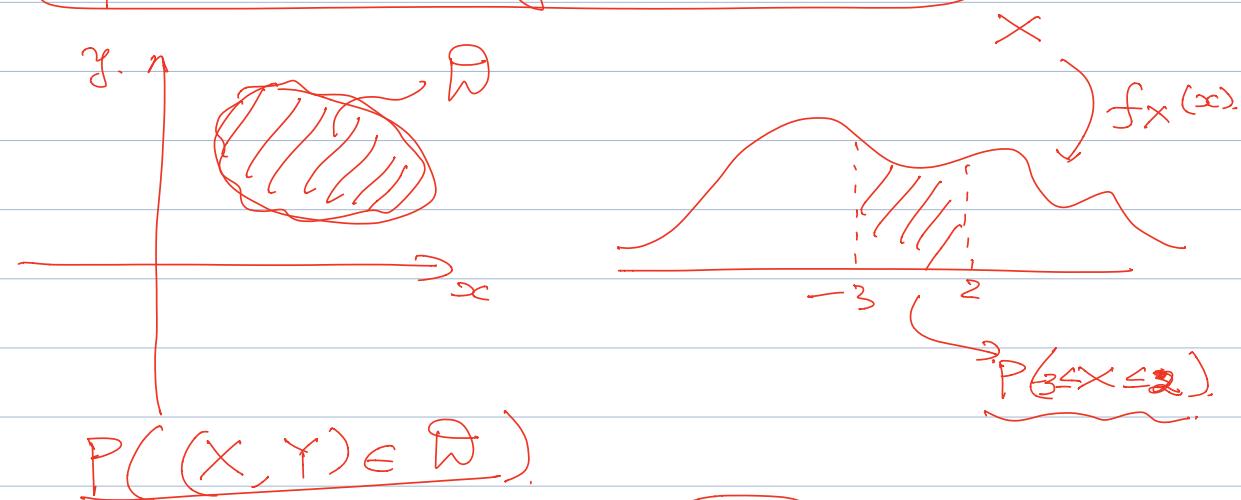
Computing Joint PDF from Joint CDF.

$$f_{X,Y}(x,y) = \frac{\partial^2 F_{X,Y}(x,y)}{\partial x \partial y}, \quad F_{X,Y}(x,y).$$

Joint CDF from joint PDF.

$$\begin{aligned} F_{X,Y}(x,y) &= P(X \leq x, Y \leq y) \\ &= \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(z,\beta) dz d\beta. \end{aligned}$$

Operational meaning of joint PDF



$$P((X,Y) \in D)$$

$$= \iint_{(x,y) \in D} f_{X,Y}(x,y) dx dy$$

$$\text{if } D = \{ -\infty \leq x \leq \infty, -\infty \leq y \leq \infty \} \rightarrow 1$$

Sample space.

Suppose we are given $f_{X,Y}(x,y)$.

Find: $P(-10 \leq X \leq 1)$

Joint PDF \rightarrow marginal PDF of X ?

PDF of X $\rightarrow f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$

$$= \int_{-\infty}^{1} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy.$$