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MSE

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continuous probability distribution

$f(x) \rightarrow \text{pdf}$

iff:

(i.) $f(x) \geq 0 \quad \forall x$

(ii.) $\int_{-\infty}^{\infty} f(x) dx = 1$

(iii.) $\int_a^b f(x) dx = P(a < x < b)$

$\sum f(x) = 1 \rightarrow \text{discrete}$

→ cumulative distribⁿ funcⁿ (cdf)

$$\begin{aligned} F(x) &= P(X \leq x) \\ &= \int_{-\infty}^x f(t) dt \end{aligned}$$

Properties of $F(x)$

(i.) Domain is $(-\infty, \infty)$

Range $[0, 1]$

(ii.) $F(x)$ is non-decreasing funcⁿ of x in the right
ie., $F'(x) = f(x) \quad x \geq 0$

(iii.) $F(x)$ is continuous on the right

(iv.) $F(-\infty) = 0$ (Prob. always defined on an
 $F(\infty) = 1$ interval)

(v.) $P(a \leq x \leq b) = P(a \leq x < b) = P(a < x \leq b) = P(a < x < b)$
 $= \int_a^b f(x) dx$

in discrete $\rightarrow P(a \leq x \leq b) = F(b) - F(a)$
 $P(a < x \leq b) = F(b) - F(a-1)$

Ex. $f(x) = k e^{-3x}, x \geq 0$
0, $x < 0$

Determine k & hence, compute cdf of x

$$\int_{-\infty}^{\infty} f(x) dx = 1 \Rightarrow \int_{-\infty}^{0} + \int_{0}^{\infty} K e^{-3x} dx = 1$$

$$= K \left(\frac{e^{-3x}}{-3} \right) \Big|_0^{\infty} = K \left[\frac{+1}{3} \right] = \frac{1}{3} \Rightarrow K = \sigma + 3$$

$$F(x) = \int_0^x 3e^{-3x} dx = \frac{3}{-3} \left[e^{-3x} \right]_0^x = 1 - e^{-3x}, x > 0$$

0 ew

$$F(2) = P(X \leq 2) = 1 - e^{-3(2)} = 1 - e^{-6}$$

$$F(4) = 1 - e^{-3(4)} = 1 - e^{-12}$$

Ex. Pdf of random variable x :

$$f(x) = \begin{cases} x & 0 \leq x < 1 \\ 2-x & 1 \leq x < 2 \\ 0 & x \geq 2 \end{cases}$$

4 - 2

Compute cdf of x :

$$F(x \leq 1) = \int_0^1 x dx = \frac{x^2}{2} \Big|_0^1 = \frac{1}{2}$$

2 - 1/2

$$F(x \leq 2) = \int_0^x x dx + \int_1^x (2-x) dx = \frac{x^2}{2} + 2x - \frac{x^2}{2} \Big|_1^x$$

$$= \frac{1}{2} + 2x - \frac{x^2}{2} - \frac{3}{2} = 2x - \frac{x^2}{2} - 1$$

$$F(x \geq 2) = \int_0^1 x dx + \int_1^2 (2-x) dx = \frac{1}{2} + 2 - \frac{3}{2} = \frac{1}{2}$$

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

$$(i) P(-1 \leq X \leq 3) = F(3) - F(-1) = 1 - 0 = \boxed{1}$$

$$(ii) P(1 \leq X \leq 1.5) = 2x - \frac{x^2}{2} - 1 = 2\left(\frac{3}{2}\right) - \left(\frac{9}{2}\right) - 1$$

$$\begin{aligned} F(1.5) - F(1) &= 3 - \frac{17}{8} = \frac{7}{8} - \left(\frac{1}{2}\right) \\ &= \boxed{\frac{3}{8}} \end{aligned}$$

Expectation :

$$\mu = E(X) = \sum_{x \in A} x p(x) \quad \text{if } x \text{ is discrete}$$

$$= \int_{-\infty}^{\infty} x f(x) dx \quad \text{if } x \text{ is continuous}$$

Let $g(x)$ = func' of x , so,

$$E[g(x)] = \sum_{x \in A} g(x) p(x)$$

$$= \int_{-\infty}^{\infty} g(x) f(x) dx$$

(Always
+ive)
(units)

Variance :

$$\sigma^2 = E[(X - \mu)^2] = \sum_{x \in A} (x - \mu)^2 p(x) \quad \text{discrete}$$

$$= \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx \quad \text{cont.}$$

$$\text{(Has unit)} \quad S.D. (\sigma) = \sqrt{E(X - \mu)^2}$$

Computational formula for σ^2 :

$$\sigma^2 = \sum_{x \in A} x^2 p(x) - [\sum_{x \in A} x p(x)]^2$$

$$\boxed{\sigma^2 = E[X^2] - (E[X])^2}$$

$$\sigma^2 = E[(X - \mu)^2] = \sum_{x \in A} (x - \mu)^2 p(x)$$

$$= \sum_{x \in A} x^2 p(x) - 2\mu x p(x) + \mu^2 p(x)$$

$$\begin{aligned}
 &= \sum_{x \in A} x^2 p(x) - 2\mu \sum_{x \in A} x p(x) + \mu^2 \sum_{x \in A} p(x) \\
 &= \sum_{x \in A} x^2 p(x) - 2\mu(\mu) + \mu^2 (1) \\
 &= \sum_{x \in A} x^2 p(x) - \mu^2 \\
 &= E[X]^2 - (E[X])^2
 \end{aligned}$$

continuous: $\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$

Properties of Expectations & Variance

- (i) $E(c) = c$ where c : constant $= \sum c p(x) = c \sum p(x) = c$
- (ii) $E(cx) = c E(x)$
- (iii) $E(c+x) = c + E(x)$
- (iv) $\text{Var}(c) = 0 = E[c^2] - (E[c])^2 = c^2 - c^2 = 0$
- (v) $\text{Var}(cx) = c^2 \text{Var}(x) = E[x^2 c^2] - [E(cx)]^2$
 $= c^2 E[X^2] - [c^2 E(X)]^2 = c^2 \text{Var}(x)$

Ex $E[X] = 5$

$\text{Var}(X) = 2$

$E[X-5] = ?$ $\text{Var}(-2X) = ?$

$\text{Var}(-2X) = 4 \text{Var}(X) = 4 \times 2 = 8$

$E[X-5] = E[X] - 5 = 0$

The r th moment about origin :

$$\mu_r' = E[X^r] = \sum_{x \in A} x^r p(x) : x \text{ is discrete}$$

$$\int_{-\infty}^{\infty} x^r f(x) dx \sim \text{cont.}$$

$r=1 \rightarrow E[X] = 1^{\text{st}} \text{ moment about Origin} = \underline{\underline{\mu}}$.
 (i.e.) 's Mean.

r th moment about mean :

$$\mu_r = E[(X-\mu)^r] = \sum_{x \in A} (x-\mu)^r p(x)$$

$$\int_{-\infty}^{\infty} (x-\mu)^r f(x) dx$$

~~$M(x(t)) = E[x]$~~

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if σ^2

$$\mu_2 = E[(x-\mu)^2] = \text{Var}(x) = \sigma^2 \rightarrow \text{2nd Moment about Mean.}$$

is Variance

$$\boxed{\sigma^2 = \mu'_2 - \mu^2}$$

$$\begin{aligned} \sigma^2 &= E[(x-\mu)^2] \\ &= E[x^2] - (E(x))^2 \\ &= \mu'_2 - \mu^2 \end{aligned}$$

→ Moment Generating Func : MGF func is used to generate moments

$$\begin{aligned} M_x(t) &= E[e^{tx}] = \sum_{x=n} e^{tx} p(x) && \text{discrete} \\ &= \int_{-\infty}^{\infty} e^{tx} f(x) dx && \text{continuous} \end{aligned}$$

$$\mu'_r = \left. \frac{d^r}{dt^r} [M_x(t)] \right|_{t=0}$$

diff. r times

$$\mu = \mu'_1 = \left. \frac{d}{dt} [M_x(t)] \right|_{t=0} \quad \sigma^2 = \mu'_2 - \mu^2 \rightarrow \text{calculate from } M_x(t)$$

→ Mean Deviation about Mean:

$$M.D. = \sum_{x} |x-\mu| p(x)$$

• sometimes, moments generated are complex, but we want only real func's, so we multiply it with i :

→ Characteristic Func :

$$\phi_x(t) = E[e^{itx}]$$

Probability Distributions① Bernoulli:

Assump's:

→ 2 outcomes of each experiment

→ $P(\text{success})$ is const. (p) , $P(\text{failure}) : (1-p) = q$

$$p(x) = \begin{cases} p & x=1 \\ q & x=0 \end{cases} \quad \begin{array}{l} x=1 \rightarrow \text{success} \\ x=0 \rightarrow \text{failure} \end{array}$$

$$P(X=1) = p \rightarrow \text{success} \quad P(X=0) = q \rightarrow \text{failure}$$

$$p(x) = \begin{cases} p^x q^{1-x}, & x=0,1 \\ 0 & \text{e.w.} \end{cases}$$

$$E(X) = p, \quad \text{Var}(X) = pq$$

$$M_X(t) = q + pe^t$$

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Joint DistributionsJoint Discrete Mass Funcⁿ (discrete joint densities)

$$f_{XY}(x,y) = P(X=x \text{ and } Y=y)$$

$$\rightarrow f_{XY}(x,y) \geq 0 \quad \forall (x,y) \quad \rightarrow \text{Necessary & sufficient}$$

$$\rightarrow \sum_{x \in A} \sum_{y \in A} f_{XY}(x,y) = 1$$

* Necessary / sufficient / Necessary & sufficient

Marginal distribⁿ (individual ki prob.)

$$f_X(x) = \sum_{y \in A} f(x,y) \quad \rightarrow \text{Marginal density for } X.$$

$$f_Y(y) = \sum_{x \in A} f(x,y)$$

Continuous Joint Density

$$1. f_{xy}(x,y) \geq 0 ; x, y \in \mathbb{R}$$

$$2. \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dy dx = 1$$

$$3. P[a \leq x \leq b \text{ and } c \leq y \leq d] = \int_a^b \int_c^d f(x,y) dy dx.$$

↳ Joint density
for (x,y)

Cont. Marginal densities :

$$f_x(x) = \int_{-\infty}^{\infty} f_{xy}(x,y) dy, \quad f_y(y) = \int_{-\infty}^{\infty} f(x,y) dx$$

Independence :

$$P(A_1 \cap A_2) = P(A_1) \cdot P(A_2)$$

$$P(X=x, Y=y) = P(X=x) * P(Y=y)$$

$$\text{or } f_{xy}(x,y) = f_x(x) \cdot f_y(y)$$

Multivariate : $f(x_1, x_2, \dots, x_n) = f(x_1) \cdot f(x_2) \cdots f(x_n)$

Ex - 1

1st task \rightarrow 2 g. joiners

2nd task \rightarrow 3 bolts

$X \rightarrow$ no. of defective welds

$Y \rightarrow$ no. of improperly tightened bolts produced per car

x/y	0	1	2	3	$f_x(x)$
0	.840	.030	.020	.010	.900
1	.060	.010	.008	.002	.080
2	.010	.005	.004	.001	.020
$f_y(y)$.910	.045	.032	.013	1.000

$$f_x(x=0) = \sum_y f(0,y) \Rightarrow 1^{\text{st}} \text{ row sum}$$

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Question 2

checking independence
here are different (check all 12 values combination) \hookrightarrow that's why checking independence in case of discrete is difficult

$$(a) P(X=2, Y=1) = 0.005$$

$$(b) P(X \geq 1, Y \geq 1) = 0.045 + 0.032 + 0.013 = 0.090$$

$$(c) P(X \leq 1) = 0.980$$

$$(d) P(Y \geq 2) = 0.032 + 0.013 = 0.045$$

Ques 3

$$f_{XY}(x,y) = \frac{1}{n^2}, \quad x=1,2,\dots,n \\ y=1,2,\dots,n$$

(a) Verify $f_{XY}(x,y)$ satisfies conditional.

$$(i) f(x,y) \geq 0$$

$$n^2 \geq 0$$

$$\Rightarrow \frac{1}{n^2} \geq 0$$

$$(ii) \sum_{x \in X} \sum_{y \in Y} f(x,y) = 1$$

$$\begin{aligned} \sum_{x \in X} \sum_{y \in Y} f(x,y) &= \sum_{x=1,2,\dots,n} \sum_{y=1,2,\dots,n} \frac{1}{n^2} \\ &= \sum_{x=1,2,\dots,n} \frac{1}{n^2} + \frac{1}{n^2} + \dots \text{ n times } = \sum_{x=1,2,\dots,n} n \left(\frac{1}{n^2} \right) = \frac{1}{n} \\ &= \boxed{1} \end{aligned}$$

(b) find marginal density for $X \in \mathcal{X}$

$$f_X(x) = \sum_{y=1}^n \frac{1}{n^2} = \frac{1}{n}, \quad f_Y(y) = \frac{1}{n}, \quad y=1,2,\dots,n$$

(c) Are $X \in \mathcal{X}$ and $Y \in \mathcal{Y}$ independent?

$$f_X(x) \times f_Y(y) = f(x,y) \Rightarrow \text{Independent}$$

Ques 4

$$f_{XY}(x,y) = c(4x+2y+1), \quad 0 \leq x \leq 40; 0 \leq y \leq 2;$$

$$f(x,y) = c(4x+2y+1) \quad 0 \leq x \leq 40, \quad 0 \leq y \leq 2$$

Q5

(a) Find c

$$c \int_0^{40} \int_0^2 (4x+2y+1) dy dx$$

$$c \int_0^{40} (4xy + y^2 + y) \Big|_0^2 dx = c \int_0^{40} (8x + 4x^2) dx = 1$$

$$4 + 2 + 8x \Rightarrow c \left. 6x + 4x^2 \right|_0^{40} = 1$$

$$\therefore \boxed{c = \frac{1}{6640}}$$

$$\Rightarrow f(x,y) = \begin{cases} \frac{1}{6640} (4x+2y+1) & 0 \leq x \leq 40, 0 \leq y \leq 2 \\ 0 & \text{e.w.} \end{cases}$$

$$(b) f_x(x) = ? \quad f_y(y) = ?$$

$$f_x(x) = c \int_0^2 (4x+2y+1) dy = \frac{8x+6}{6640}, \quad 0 \leq x \leq 40$$

$$f_y(y) = c \int_0^{40} (4x+2y+1) dx = c \left[2x^2 + (2y+1)x \right] \Big|_0^{40} = \frac{80y+3240}{6640}$$

0, e.w.

$$(c) f(x > 20, y > 1) = c \int_{20}^{40} \int_1^2 (4x+2y+1) dy dx$$

$$= c \int_{20}^{40} (4xy + y^2 + y) \Big|_1^2 dx = \underbrace{\int_{20}^{40} \frac{2480}{6640} dx}_{\frac{2480}{6640}}$$

$$(d) P(y > 1) = \int_1^2 \frac{80y+3240}{6640} dy = 0.526$$

$$(e) P(x > 20) = c \int_{20}^{40} \frac{8x+6}{6640} dx = 0.741$$

$$(f) f_x(x) \cdot f_y(y) \neq f_{xy}(x,y) \rightarrow \text{Not Independent}$$

Q5. (n -dimensional discrete random variables)

$$f(x_1, x_2, \dots, x_n) = P[x_1=x_1, x_2=x_2, \dots, x_n=x_n]$$

$$\begin{aligned} P(x_1) &= 0.9 & P(x_2) &= 0.08 & P(x_3) &= 0.02 \\ \downarrow \text{non-def} & & \downarrow \text{def but} & & \downarrow \text{neither} & \\ n=20 \text{ items} & & \text{salvagable} & & & \end{aligned}$$

$$(a) P[x_1=15, x_2=3, x_3=2] = ?$$

(b) Find general formula for density for (x_1, x_2, x_3) .

$$\begin{aligned} \text{soln. } P(x_1=x_1, x_2=x_2, x_3=x_3) &= f(x_1, x_2, x_3) \\ &= \frac{n!}{x_1! x_2! x_3!} p_1^{x_1} p_2^{x_2} p_3^{x_3} \Rightarrow \boxed{\text{Multinomial distrib}} \\ &= \frac{20!}{15! 3! 2!} (0.9)^{15} (0.08)^3 (0.02)^2 \quad \downarrow \text{extension of binomial one.} \\ &= \underline{\underline{0.0065}} \end{aligned}$$

Lecture 2 :

Expectation & Covariance

$$① E[H(x, y)] = \sum_{\forall x} \sum_{\forall y} H(x, y) f_{xy} \text{ permut}$$

$$② = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H(x, y) f_{xy} dy dx$$

$$\text{Ex. } f(x, y) = \begin{cases} \frac{3}{2} (x^2 + y^2) & 0 \leq x < 1, 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$E[x^2 + y^2]$$

$$= \int_0^1 \int_0^1 (x^2 + y^2) \frac{3}{2} (x^2 + y^2) dy dx$$

$$= \frac{3}{2} \int_0^1 \left(\int_0^1 (x^4 + 2x^2y^2 + y^4) dy \right) dx =$$

$$\begin{aligned}
 & \int_0^{\frac{3}{2}} \left[x^4 y + \frac{2x^2 y^3}{3} + \frac{y^5}{5} \right] dx \\
 &= \frac{3}{2} \int_0^{\frac{3}{2}} \left(x^4 + \frac{2x^2}{3} + \frac{1}{5} \right) dx \\
 &= \frac{3}{2} \left[\frac{1}{5} + \frac{2}{3} \times \frac{1}{3} + \frac{1}{5} \right] = \frac{3}{2} \left[\frac{1}{5} + \frac{2}{9} + \frac{1}{5} \right] \\
 &= \frac{3}{2} \left(\frac{2}{5} + \frac{2}{9} \right) = 3 \left[\frac{1}{5} + \frac{1}{9} \right] = \frac{3 \times 14}{5 \times 9} = \frac{14}{15}.
 \end{aligned}$$

Univariate Average : calculate $E(x)$ from $f(x,y)$

$$E[x] = \sum_{\forall x} x \underbrace{f_x(x)}_{\substack{\downarrow \\ \text{Marginal density of } x}} = \sum_{\forall x} \sum_{\forall y} x f(x,y)$$

Covariance : → How x & y are related

$$\text{cov}(x,y) = E[(x-\mu_x)(y-\mu_y)] \quad E[X]=\mu_x, E[Y]=\mu_y$$

* $x \downarrow$ & $y \downarrow \Rightarrow (x-\mu_x)(y-\mu_y) > 0 \rightarrow$ positively correlated
 $x \uparrow$ & $y \uparrow \Rightarrow n \quad n \quad < 0$
else

Proof

if $\text{cov}(x,y) = 0 \rightarrow x$ & y are independent

$$\boxed{\text{cov}(x,y) = E[XY] - E[X]E[Y]}$$

Exercise

$$\begin{aligned}
 \text{cov}(x,y) &= E[(x-\mu_x)(y-\mu_y)] \\
 &= E[XY - YE[X] - xE[Y] + E[X]E[Y]] \\
 &= E[XY] - E[Y]E[X] - E[X]E[Y] + E[X]E[Y] \\
 &= E[XY] - E[X]E[Y]
 \end{aligned}$$

(a)

(b)

In If X & Y are independent :

$$E[XY] = E[X] E[Y] \Rightarrow \text{Cov}(X, Y) = 0$$

But not
vice versa

Proof : $E(XY) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x, y) dx dy$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_x(x) f_y(y) dx dy$$

$$= \int_{-\infty}^{\infty} x f_x(x) dx \int_{-\infty}^{\infty} y f_y(y) dy = E[X] E[Y]$$

Ex. $P(-1) = P(0) = P(1) = 1/3$

Let $Y = X^2$

$$E[X] = (-1+0+1)/3 = 0$$

$$E[Y] = (1+0+1)/3 = 2/3$$

$$E[XY] = (-1+0+1)/3 = 0$$

$E[XY] = E[X] E[Y]$, but X & Y are dependent.

proof

Proof. $\text{Cov}(X, Y) = \text{Var}(X) = \text{Var}(Y)$ ~~if~~ $X = Y$

$$\begin{aligned} \text{if } X = Y, \Rightarrow \text{Cov}(X, X) &= E[X \cdot X] - E[X] E[X] \\ &= E[X^2] - (E[X])^2 = \text{Var}(X) \end{aligned}$$

~~13/9/19~~ $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2 \text{Cov}(X, Y)$

∴

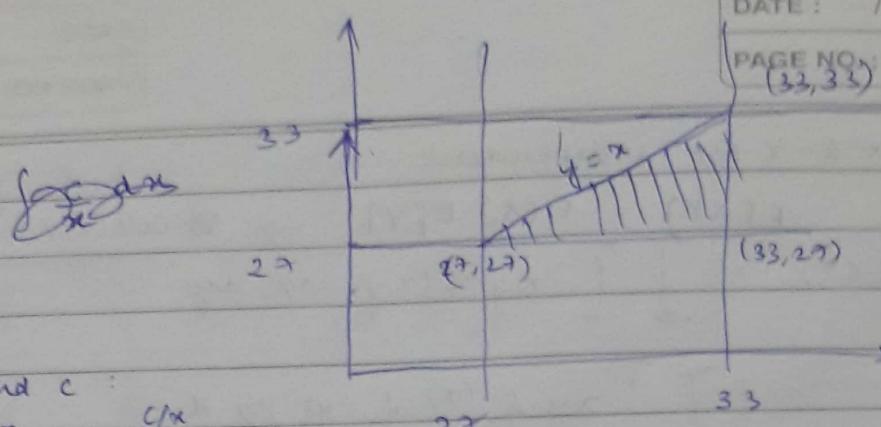
Exercise 19. The joint density is given by

$$f_{XY}(x, y) = \frac{c}{\pi} \quad 27 \leq y \leq x \leq 33$$

$$c = \frac{1}{(33-27) \times 33/27} = 1/72$$

(a) Find $E[X]$, $E[Y]$, $E[XY]$, $\text{Cov}(X, Y)$

(b) $E[X-Y]$



To find c :

$$\int_{-27}^{33} \int_{-\infty}^{\infty} f(x, y) dy dx = 1$$

$$= \int_{27}^{33} \frac{c}{x} dy dx \rightarrow \int_{27}^{33} \frac{c(x-27)}{x} dx$$

$$= c \int_{27}^{33} \left(1 - \frac{27}{x}\right) dx = 1 \rightarrow c \left[x - 27 \ln x \right]_{27}^{33} = 1$$

$$\therefore c = \frac{1}{6 - 27 \ln 33/27} = 1.72$$

(a) ~~E(X)~~

$$f_x(x) = \int_{27}^x \frac{c}{x} dx \Rightarrow \frac{(x-27)}{x} c$$

$$f_y(y) = \int_y^{33} \frac{c}{x} dx \Rightarrow \frac{(33-y)}{y} c (\ln 33 - \ln y)$$

$$E[X] = \int_{27}^{33} x f_x(x) dx = \int_{27}^{33} c(x-27) dx = \frac{x^2}{2} \Big|_{27}^{33} = 30.96$$

$$E[Y] = \int_{27}^{33} c(\ln 33 - \ln y) y dy = 28.99$$

$$E[XY] = \int_{27}^{33} \int_y^{33} xy \frac{c}{x} dx dy = 897.84$$

$$(b) E[X-Y] = Cov(XY) = E[XY] - E[X]E[Y] = 0.31$$

$$E[X-Y] = \int_{27}^{33} \int_{27}^y (x-y) \frac{c}{x} dy dx = 1.97$$

PROVE

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2 \text{Cov}(X, Y)$$

If, X & Y are independent:

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y), \text{Var}(X-Y) = \text{Var}(X) + \text{Var}(Y)$$

$$\begin{aligned}\text{Var}(X+Y) &= E[(X+Y)^2] - (E[X+Y])^2 \\ &= E[X^2 + Y^2 + 2XY] - (E[X] + E[Y])^2 \\ &= E[X^2] + E[Y^2] + 2E[XY] - (E[X])^2 - (E[Y])^2 - 2E[X]E[Y] \\ &= \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)\end{aligned}$$

$$\text{ex. } f_{XY}(x, y) = \frac{1}{2(e-1)} \left[\frac{1}{x} + \frac{1}{y} \right] \quad \begin{array}{l} 1 \leq x \leq e \\ 1 \leq y \leq e \end{array}$$

(a) Show that it is valid joint density

$$(b) \text{ Find } E[X], E[Y], E[XY] \quad \frac{3e-1}{4}, \frac{3e-1}{4}, \frac{e^2-1}{2}$$

(c) X & Y independent? No

$$\begin{aligned}(a) \quad &\int_1^e \int_1^y \frac{1}{2(e-1)} \left[\frac{1}{x} + \frac{1}{y} \right] dy dx \\ &= \int_1^e \frac{1}{2(e-1)} \left(\frac{y}{x} + \ln y \right) \Big|_1^e dx = \frac{1}{2(e-1)} \int_1^e \left(\frac{(e-1)}{x} + 1 \right) dx \\ &= \frac{1}{2(e-1)} \left((e-1) \ln x + x \right) \Big|_1^e = \frac{1}{2(e-1)} [e-1 + e-1] = 1\end{aligned}$$

(b) $E[X]$:

$$f_X(x) = \int_1^e \frac{1}{2(e-1)} \left[\frac{1}{x} + \frac{1}{y} \right] dy = \frac{1}{2(e-1)} \left(\frac{e-1}{x} + 1 \right)$$

$$\begin{aligned}E(X) &= \int_1^e x f_X(x) dx = \int_1^e \left(\frac{x}{2(e-1)} \left[\frac{e-1}{x} + 1 \right] \right) dx \\ &= \int_1^e \frac{1}{2} + \int_1^e \frac{x}{2(e-1)} dx = \frac{x^2}{2} \Big|_1^e\end{aligned}$$

$$= \left\{ \frac{e-1}{2} + \frac{1}{2(e-1)} \left[\frac{e^2-1}{2} \right] \right\} = \frac{e-1}{2} + \frac{e+1}{4} = \frac{3e-1}{4}$$

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Conditional Distribution

Conditional densities :

$$f_{X|Y} = \frac{P(X=x_0 | Y=y)}{P(Y=y)} = \frac{f_{XY}(x,y)}{f_Y(y)}, f_Y(y) > 0$$

satisfies

$$\text{1) } \int_{-\infty}^{\infty} f_{X|Y} dy \geq 0$$

$$\text{2) } \int_{-\infty}^{\infty} f_{X|Y} dx = 1$$

Conditional mean & variance :

$$E(Y|x) = \sum y f_{Y|x}(y) = \int y f_{Y|x}(y) dy$$

$$\text{var}(Y|x) = \sum y^2 f_{Y|x}(y) - \mu_{Y|x}^2$$

Ex. $f(x,y) = \begin{cases} x^2 + \frac{xy}{3}, & 0 \leq x \leq 1 \\ 0 & 0 \leq y \leq 2 \end{cases}$
e.w.

solⁿ Marginal density for x :

$$f_X(x) = \int_0^2 \left(x^2 + \frac{xy}{3} \right) dy = x^2 y + \frac{x}{6} y^2 \Big|_0^2 \\ = 2x^2 + \frac{2}{3} x \quad 0 \leq x \leq 1$$

$$f_Y(y) = \int_0^1 \left(x^2 + \frac{xy}{3} \right) dx = \frac{1}{3} + \frac{y}{6}$$

$$f_{X|Y} = \frac{f_{XY}(x,y)}{f_Y(y)} = \frac{x^2 + \frac{xy}{3}}{\frac{1}{3} + \frac{y}{6}} = \frac{6x^2 + 2xy}{2 + 4y}, \frac{\partial^2 f_{X|Y}}{\partial x \partial y}$$

$$f_{Y|x} = \frac{f_{XY}(x,y)}{f_X(x)}$$

Mean & Variance of linear Combⁿ of Random Variables

$$Y = c_1 X_1 + c_2 X_2 + \dots + c_p X_p \quad X_i \rightarrow r.v. \\ c_i \rightarrow \text{const}$$

$$E(Y) = c_1 E(X_1) + c_2 E(X_2) + \dots + c_p E(X_p)$$

$$V(Y) = c_1^2 V(X_1) + c_2^2 V(X_2) + \dots + c_p^2 V(X_p) + 2 \sum_{i \neq j} c_i c_j \text{cov}(X_i, X_j)$$

If X_1, X_2, \dots, X_p are independent

$$V(Y) = c_1^2 V(X_1) + c_2^2 V(X_2) + \dots + c_p^2 V(X_p)$$

$$\begin{aligned} \rightarrow V(Y) &= E[(Y - E(Y))^2] \\ &= E[(c_1 X_1 + c_2 X_2 + \dots + c_p X_p - (c_1 \mu_1 + c_2 \mu_2 + \dots + c_p \mu_p))^2] \\ &= E[(c_1(X_1 - \mu_1) + c_2(X_2 - \mu_2) + \dots + c_p(X_p - \mu_p))^2] \\ &= E[c_1^2(X_1 - \mu_1)^2 + c_2^2(X_2 - \mu_2)^2 + \dots + \dots] \end{aligned}$$

Ex. X & Y : independent r.v.

$$\mu_X = 7 \quad \mu_Y = -5$$

$$V(X) = 9 \quad V(Y) = 3$$

$$\begin{aligned} (i) \quad E[4X - 2Y + 6] &= 4E[X] - 2E[Y] + 6 = 4(7) - 2(-5) + 6 \\ &= 28 + 10 + 6 \\ &= 44 \end{aligned}$$

$$\begin{aligned} (ii) \quad V(4X - 2Y + 6) &= 16V(X) + 4V(Y) + 36 \\ &= 16(9) + 4(3) + \cancel{36} = 144 + 12 + 36 \\ &= 156 \end{aligned}$$

Ex. Let X_1 & X_2 be 2 independent r.v. having normal distrib' having mean 5 & 6, var 2 & 3. Find mean of $Y = X_1 + X_2$. Find distrib' of Y

Soln (Take help of moment generating func'')

$$M_{X_1}(t) = e^{5t + \frac{1}{2} \cdot 2t^2}$$

Given $X_1 \sim N(5, 2)$

$$M_{X_1}(t) = e^{5t + \frac{1}{2} \cdot 2t^2}$$

$$M_{X_2}(t) = e^{6t + \frac{1}{2} \cdot 3t^2}$$

$$\begin{aligned} M_Y(t) &= M_{(X_1+X_2)}(t) = M_{X_1}(t) \cdot M_{X_2}(t) && \{X_1 \& X_2: \text{independent}\} \\ &= (e^{5t+t^2}) \cdot (e^{6t+\frac{3}{2}t^2}) \\ &= e^{11t+\frac{5}{2}t^2} \rightarrow N(11, 5) \end{aligned}$$

$Y \rightarrow$ also a Normal distrib'

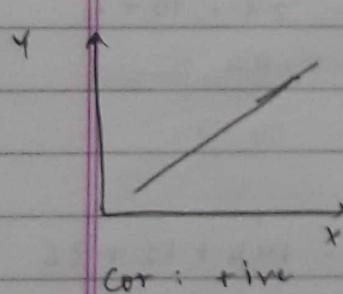
Correlation

$\rho \rightarrow$ Pearson Coefficient

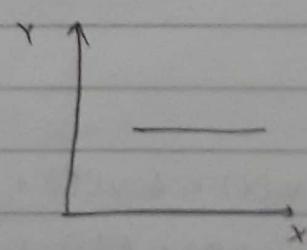
$\star \star \quad \boxed{\rho = \frac{\text{cov}(X, Y)}{\sqrt{(\text{var}(X))(\text{var}(Y))}}}$

\Rightarrow linear rel' b/w X & Y

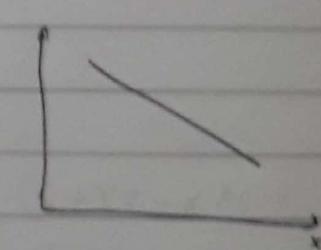
If distrib' is given, use this



Cor: +ive



Cor: 0



Cor: -ive

$\rho = -1 \Rightarrow$ Perfect -ive correlation

$\rho = 0 \Rightarrow$ Independent var (Not correlated) \rightarrow Non-linear rel' ship

$\rho = 1 \Rightarrow$ Perfect +ve correl'.

\rightarrow Not possible for large pop'n to calculate ρ , taken random samp' of size n .

$$\text{Var } X = \sum_{i=1}^n (x_i - \bar{x})^2 / n = s_{xx}/n, \text{ similarly for } Y$$

$$\text{Cov}(x, y) = E[(x - \mu_x)(y - \mu_y)]$$

$$\Rightarrow \text{cov}(x, y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n} = S_{xy}/n$$

$$\hat{r} = R = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}$$

Computational formula :

$$\star \hat{r} = \frac{n \sum xy - \sum x \sum y}{\left[\left(n \sum x^2 - (\sum x)^2 \right) \left(n \sum y^2 - (\sum y)^2 \right) \right]^{1/2}} \Rightarrow \text{use when data is given}$$

Q. Find sample corr. coeff.

$$1.) x : 1, 2, 3, 4, 5$$

$$Y = 2, 4, 6, 8, 10$$

$$\text{Soln: } \hat{r} = ?$$

x	y	x^2	y^2	xy
1	2	1	4	2
2	4	4	16	8
3	6	9	36	18
4	8	16	64	32
5	10	25	100	50
\sum	15	55	220	110
				75
				1100 - 400

$$\hat{r} = r = \frac{5(110) - (15)(30)}{\sqrt{[5 \times 55 - 225][5 \times 220 - 900]}}$$

$$= \frac{550 - 450}{\sqrt{50 \times 200}} = \frac{100}{\sqrt{10000}} = \frac{100}{100} = 1$$

2) $f_{XY}(x, y) = \frac{1}{6640} (4x+2y+1)$ $0 \leq x \leq 40$
 $0 \leq y \leq 2$

Find $E(X)$, $E(Y)$, $E(X^2)$, $E(Y^2)$, $\text{Var}(X)$, $\text{Var}(Y)$, $\text{Cov}(XY)$.

Sol? $E(XY) = \int_0^{40} \int_0^2 xy \frac{1}{6640} (4x+2y+1) dy dx = \frac{26.586}{6640}$

$$E(X) = \int_0^{40} \int_0^2 x f_{XY}(x) dx dy = 26.426 \quad E(X^2) = 700.361$$

$$E(Y^2) = 1.3494$$

$$E(Y) = 1.008$$

$$\text{Cov}(XY) = E(XY) - E(X)E(Y) = 26.586 - (26.426)(1.008)$$

$$= -0.051$$

$$\text{Var}(X) = 700.361 - (26.426)^2 = 2.027$$

$$\text{Var}(Y) = 1.3494 - (1.008)^2 = 0.333$$

$$\rho = \frac{\text{Cov}(XY)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} = -0.009 \Rightarrow \text{negatively correlated}$$

Chi-square distribution

Let $Z \sim N(0, 1)$

$\Rightarrow Z^2 \sim \chi^2 \rightarrow \text{chi-square}$

Let X_1, X_2, \dots, X_n be n independent variables with mean μ_i & σ_i^2 ($i=1, 2, \dots, n$) then,

$\sum_{i=1}^n \frac{(X_i - \mu_i)^2}{\sigma_i^2}$ is a χ^2 -variate with n d.f.

prob. Let X be χ^2 variable with n d.f. n

$$E(X) = n$$

$$\text{Var}(X) = 2n$$

→ This distribution is used for Goodness of fit

Observed & Expected frequency

Chi-square (χ^2) Test

of goodness of fit :

check if theoretical

value is same as observed

Let O_i ($i=1, 2, \dots, n$) be a set of observed frequencies & e_i ($i=1, 2, \dots, n$) corresponding set of expected (theoretical) frequencies. Then,

$$\chi^2 = \sum_{i=1}^n \frac{(O_i - e_i)^2}{e_i}$$

has a χ^2 dist. with $(n-1)$ d.f.

given, all $e_i \geq 5$. (Tukey rule)

to
degree of freedom

If $\chi^2_{\text{calc}} < \chi^2_{\text{Tab}}$ \Rightarrow we have made correct assumption. If no significant, we can accept it.
difference was found to vary

Ex. The demand of a spare part in a factory from day to day. In a sample study, foll. info. was obtained.

Days	M	T	W	Th	F	S.	Total
No. of parts demanded	1124	1125	1110	1120	1126	1115	6720
(O_i)							"

→ Observed frequency

Test the hypothesis, that the no. of parts doesn't depend on the day.

sol"

$P(\text{demand})$ is same for all day

① find total of all days = 6720 = n

H_0 : The demand of no. of parts doesn't depend on day

H_1 : H_0 is not true

$$p_i = \frac{1}{6}, i=1, 2, \dots, n$$

$$e_i = n \times p_i$$

No. of parts : 1120 1120 1120 1120 1120 1120
(e_i)

$$\chi^2 = \sum \frac{(o_i - e_i)^2}{e_i}$$

$$= \frac{(1124 - 1120)^2}{1120} + \frac{(1125 - 1120)^2}{1120} + \dots \\ = 0.179$$

No. of cells = 6

→ Degree of freedom here = 5

$$\chi^2_{\text{tab with } 5 \text{ d.o.f.}} = 11.07 \quad (\text{from table})$$

$0.179 < 11.07 \Rightarrow$ no significant difference
 $\Rightarrow H_0 \text{ is accepted.}$

→ If any $e_i < 5 \rightarrow$ can add it with any preceding or succeeding cell (add both e_i & o_i). Then, d.o.f. --;

student t-distribution

→ It is a sample distrib'

→ Can be used whenever z-distrib' is used, only cond' is $n \leq 30$

Let x_1, x_2, \dots, x_n be random sample of small size n with mean μ and variance σ^2 respectively. ($i=1, 2, \dots, n$) from a normal population with mean μ & variance σ^2 . Let \bar{x} be mean of sample & s^2 be variance of sample.

$$\text{i.e., } \bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$$

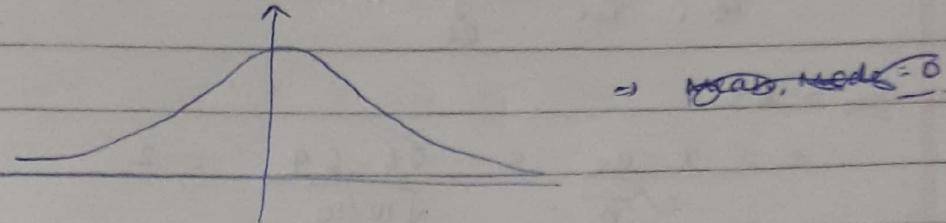
$$\text{and } s^2 = \frac{1}{(n-1)} \sum_{i=1}^n (x_i - \bar{x})^2$$

↓

In sample, only $n-1$ samples are linearly independent

then $t = \frac{\bar{x} - \mu}{S/\sqrt{n}}$ has a t-distribⁿ with $(n-1)$ d.f.

$$f(-t) = f(t) \text{ : symmetric}$$



$$\mu_{2k+1}^* = 0$$

$$\mu_{2k} = \frac{n k (2k-1)(2k-2) \dots 3 \cdot 1}{(n-2)(n-4) \dots (n-2k)}, \quad \frac{n}{2} > k$$

$$\mu_2 (\text{Var}) = \frac{n}{n-2}, \quad n > 2 \quad \boxed{\text{if mean} = 0} \quad (\mu_1 = 0)$$

④ Significance Test for sample Mean

Let x_1, x_2, \dots, x_n be a random sample of size n from a normal population with a specified popuⁿ mean μ .

To test $\mu = \mu_0$ or there is no significant difference between

sample mean \bar{x} and samⁿ popuⁿ mean μ_0 , cat compute

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \quad \text{where } \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad s^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2$$

- if $|t| < t_{\text{tab}}$ \Rightarrow no significant diff. between \bar{x} & μ_0

~~If~~ If $n > 30 \rightarrow$ then t will become Z .

Questⁿ: The heights of 10 men of a locality are found to be:

70, 67, 62, 68, 61, 68, 70, 64, 64, 66

Is it reasonable to believe that avg. height is > 64 inches?

Test at 5% level of significance

$$\begin{aligned} t &> t_{\alpha} \\ t &< t_{\alpha} \\ |t| &< t_{\alpha/2} \end{aligned}$$

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Soln. $\bar{x} = \frac{\sum x_i}{n} = \frac{660}{10} = 66$ $\alpha = 0.05$

$$s^2 = \frac{1}{10-1} \sum_{i=1}^{10} (x_i - \bar{x})^2 = 10$$

$$\mu_0 = 64$$

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} = \frac{66 - 64}{\sqrt{10}/10} = 2$$

$$t_{\text{tab with 9 d.f.}} = 2.26$$

$t_{\text{calc}} < t_{\text{tab}} \Rightarrow$ Avg. height is not greater than 64 inches.

F-distribution

Let X & Y be two independent Chi-square variates with ν_1 & ν_2 d.f. Then,

$$F = \frac{X/\nu_1}{Y/\nu_2} = F(\nu_1, \nu_2)$$

App's:

↳ entire analysis of variance of
↳ Agriculture, Medical

- Let x_1, x_2, \dots, x_n & y_1, y_2, \dots, y_n be 2 independent random samples drawn from same normal population with same ~~mean~~ ^{variance}.
- Let \bar{x} & \bar{y} be the sample mean & s_x^2 & s_y^2 be sample variance of 2 samples.

$$\Rightarrow \bar{x} = \frac{1}{n_1} \sum x_i \quad \bar{y} = \frac{1}{n_2} \sum y_i$$

$$s_x^2 = \frac{1}{n_1-1} \sum (x_i - \bar{x})^2 \quad s_y^2 = \frac{1}{n_2-1} \sum (y_i - \bar{y})^2$$

Then,
$$\boxed{F = \frac{s_x^2/n_1}{s_y^2/n_2}}$$