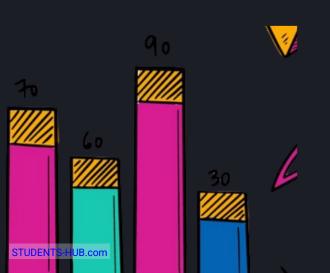
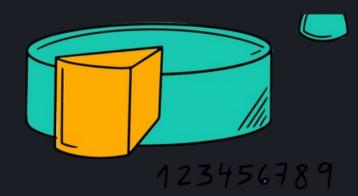




STATESTES.

By Rawan Alfares





Recorded lec

chA elementary statistics

1

we take a random Sample (X,, X2, X3,....Xn) of Size (n) from a Population. Some of Sample properties :-

1. Sample mean
$$\mathcal{L}_{x}^{0}$$
 is defined as 3-
 $\mathcal{L}_{x}^{0} = \perp \hat{\mathcal{L}}_{x}^{0}$

2. Sample variance
$$\theta_{X}^{2}$$
 when f_{X}^{2} is known $\theta_{X}^{2} = \frac{1}{n} \mathcal{E}(X_{i} - f_{X}^{2})$

when f_{X}^{2} is unknown $\theta_{X}^{2} = \frac{1}{n} \mathcal{E}(X_{i} - f_{X}^{2})$

when f_{X}^{2} is unknown $\theta_{X}^{2} = \frac{1}{n} \mathcal{E}(X_{i} - f_{X}^{2})$

$$\theta_{X}^{2} = n \mathcal{E}(X_{i}^{2} - f_{X}^{2})$$

$$\eta(0-1)$$

3. Sample Standard diviation
$$8-6\hat{y} = \int \hat{\theta_x^2}$$

$$C_{xy} \triangleq n \underbrace{\sum_{i=1}^{n} X_i y_i - \sum_{i=1}^{n} X_i \underbrace{\sum_{i=1}^{n} y_i}_{i=1}}_{n (n-1)}$$

Example 8-

X	75	80	90	85	95	99	70	Exi = 594
yi	80	78	92	83	90	95	75	$E_{i} = 593$

1)
$$M_{\hat{x}}^2 = \frac{594}{7} = 84.8$$
 $M_{\hat{y}}^2 = \frac{593}{7} = 84.714$

2)
$$6x^{2} = \frac{1}{n-1} \mathcal{E} (\chi_{\dot{c}} - \mu_{\chi}^{2})^{2} = \frac{690.857}{6} = 115.143$$

$$\theta'y = \frac{1}{n-1} \mathcal{E} (y_{i} - \mu y_{i})^{2} = \frac{351.446}{6} = 58.574$$

$$3) = \frac{459.761}{6} = 76.627$$

$$\frac{4)}{82.138} = \frac{76.627}{82.138} = 0.933$$

seconded lecture Regression Techniques

$$\epsilon = \left[\sum_{i=1}^{n} \left[\sum_{i=1}^{n} (\alpha_{i} x_{i} + \beta_{i}) \right]^{2} \rightarrow \text{least Square error.}$$

Cueve live error live for data sit in a staight line

- we derivative the equation two times, one to & the other to B. $\frac{\partial \mathcal{E}}{\partial x} = -\frac{\partial}{\partial x} \left[\left[Y_i - (\alpha X_i + \beta) \right] X_i = 0 \rightarrow n\beta + \alpha \left[\frac{\partial}{\partial x} X_i = \frac{\partial}{\partial x} Y_i \right]$

$$\frac{\partial \mathcal{E}}{\partial \beta} = -\partial \left[\frac{\partial}{\partial x} \left[Y_i - (\alpha X_i + \beta) \right] = 0 \qquad \Rightarrow \beta \left[\frac{\partial}{\partial x} X_i + \alpha \left[\frac{\partial}{\partial x} X_i \right] = \left[\frac{\partial}{\partial x} X_i \right]$$

using Creamer's Rule

$$\alpha = \frac{C \times y}{6 \hat{\kappa}_{y}^{2}} \qquad \beta = \hat{\mu}_{y} - \alpha \hat{\mu}_{x}^{2}$$

Fitting a polynomial by the method of least squares

ن کیرہ- احنا بنحل عشان نوجد φ و α .

examples												
·	y: speed (mls)	χi	Xi Yi	Xi -Mx	y; - 129							
1	5.7	١	5.7									
1.3	6.3	1.69	8.4									
a.3	7.4	5.29	17.02									
3	8.4	q	as. a									
3.5	11.9	12.25										
4	13.7	16	54.8									
EXi=15.1	રપા= 53	45.23	152.56									

find 8-

$$\lim_{X \to \infty} \frac{1}{n} = \lim_{x \to \infty} \frac{1}{x} = \lim_{x$$

2)
$$\mu_{\gamma}^2 = \frac{1}{6}(53) = 8.8333.$$

$$3) Cxy =$$

I find a linear equation $y = \alpha x + \beta$ that describes the speed of the vichle (y) vs. time (x) based on the above measurements:

to find & and B

$$\begin{bmatrix} x_i & \xi x_i^2 \end{bmatrix} \begin{bmatrix} \beta \\ \alpha \end{bmatrix} = \begin{bmatrix} \xi y_i \\ \xi x_i x_i \end{bmatrix}$$

$$\begin{bmatrix} 6 & 15.1 \\ 15.1 & 45.23 \end{bmatrix} \begin{bmatrix} B \\ \alpha \end{bmatrix} = \begin{bmatrix} 53.4 \\ 152.56 \end{bmatrix}$$

$$\alpha = 0.514$$
 $\beta = 0.573$

$$y = ae$$
 $\rightarrow lny = lnae$
 $lny = lna + lne$

$$y' = B' + \alpha' x$$
 $y' = B' + \alpha' x$
 $y' = B' +$

$$y' = \ln y$$
, $\beta' = \ln \alpha$, $\alpha' = b$

$$\begin{bmatrix} n & \xi X_{i} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \hat{\beta} \\ \alpha' \end{bmatrix} = \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi X_{i} \end{bmatrix} \begin{bmatrix} \xi y_{i} \text{ new} \\ \xi$$

$$\begin{bmatrix} 6 & 15.1 \\ 15.1 & 45.23 \end{bmatrix} \begin{bmatrix} B \\ A \end{bmatrix} = \begin{bmatrix} 12.804 \\ 34.258 \end{bmatrix}$$

$$\begin{bmatrix} 6.3 & 1.84055 \\ 2.00148 & 4.603404 \\ 8.4 & 2.128232 & 6.384695 \\ 11.9 & 2.476538 & 8.667884 \\ 13.7 & 2.617396 & 10.46958 \\ 53.4 & 12.80466 & 34.25875 \end{bmatrix}$$

$$\alpha = 0.6,8724$$
) as the work 10 11th $\alpha = 0.6$

$$\alpha = e = e = 4.137$$

$$B = \alpha' = 0.28$$

$$\frac{\partial E}{\partial b} = -2 \sum_{i=1}^{6} (y_i - bx_i) (x_i) = 0 \implies b = \frac{Ey_i x_i}{Ex_i^2} = \frac{152.56}{45.23}$$

example (5-a):-

$$y = ax^b$$
, find the error furthis Curve.

 $y = \alpha x + \beta$
 $\ln y = \ln (ax^b) = \ln a + b \ln x$
 $y' = \alpha + bx$
 $y = \beta + dx$
 $y = \ln y$, $x = \ln x$, $b = \alpha$, $\beta = \ln \alpha$

$$6 \, \alpha^{1} + 4.83 \, 2943 \, \beta = 12.80466$$

 $4.83 \, 2943 \, \alpha^{1} + 5.460749 \, \beta = 11-219043$

$$\alpha' = 1.669158992 = B$$

$$\beta = 0.5772271778 = \infty$$

$$\ln a = B \implies a = \frac{B}{e} = 5.30770$$

$$Y = 5.30770 X$$

If
$$y = 1 - e^{\frac{-x^b}{a}}$$

$$\frac{-\frac{x^{b}}{a}}{\sum_{a=1}^{b}-1}$$

$$\frac{x^{b}}{e^{a}} = \frac{1}{1-y} \implies \ln \frac{x^{b}}{e^{a}} = \ln \frac{1}{1-y}$$

$$\ln \frac{x^{b}}{a} = \ln \ln \frac{1}{1-y}$$

$$b \ln x - b \ln a = \ln \ln \frac{1}{1-y}$$

b
$$\ln x - b \ln a = \ln \ln \frac{1}{1-x}$$

Tinear Regression

$$\beta'x' + \alpha' = \gamma$$

$$\alpha'=b \ln \alpha$$
, $\beta'=b$, $\chi'=\ln x$, $\gamma=\ln \ln \frac{1}{1-\gamma}$

EXAMPLE (5-4):

If
$$y = \frac{L}{1 + e^{a+bx}}$$

$$y + ye = L$$

$$e = L-y \longrightarrow a+bx = ln(L-y)$$

$$y'=\ln\left(\frac{L-y}{y}\right)$$
 , $\beta'=\alpha$, $\alpha'=b$

Recorded lecture y = C1 X1 + C2 X2

if
$$x_1$$
 and x_2 are indep = un carollated $\Rightarrow P_{x_1x_2} = 0$

Theorem: independent Gaussian Radom variables

then y is also gaussian distribution.

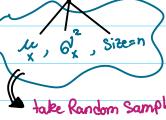
$$y = C_1 x_1 + C_2 x_2 + C_3 x_3$$
, find μ_y and θ_y^2

$$\frac{y_{2}-C_{1}y_{1}+C_{2}y_{2}+C_{3}y_{3}}{x_{1}+x_{2}}$$

$$\theta_{y}^{2} = C_{1}^{2} \theta_{x_{1}}^{2} + C_{2}^{2} \theta_{x_{2}}^{2} + C_{3}^{2} \theta_{x_{3}}^{2} + 2C_{1}C_{2} \theta_{x_{1}}^{2} \theta_{x_{2}}^{2} P_{x_{1}} P_{x_{2}}^{2}$$
 if x_{1} , x_{2} and x_{3} are indep $+ 2C_{2}C_{3} \theta_{x_{2}}^{2} \theta_{x_{3}}^{2} P_{x_{3}}^{2} P_{x_{3}}^{2}$ then θ_{y}^{2} is without this paret.

+
$$\partial C_1 C_3 G_{X_1} G_{X_3} F_{X_1} F_{X_3}$$
 means $f_{X_1 X_2}, f_{X_2 X_3}, f_{X_1 X_3}$ are Q

all of them are known



) Population

at Size n



examples
$$\mathcal{P}(\mu_{x}^{2} \leqslant 80) \rightarrow \text{we use "Central limit theorem" only for Sample mean$$

Assume that
$$y = \mu_{\hat{x}} = \perp \hat{\xi} x_{\hat{x}}$$

VOIT[Y] =
$$\theta_{\lambda}^{3} = \text{VOIT}[\lambda_{\lambda}^{2}] = (\frac{1}{2})^{2}\theta_{\lambda}^{2} + (\frac{1}{2})^{2}\theta_{\lambda 2}^{2} + \dots + (\frac{1}{2})^{2}\theta_{\lambda n}^{2}$$

$$= M \cdot 1 \cdot 6^{2} = \frac{6^{2}}{100} = Var[M]$$
 bis

EXAMPLE (5-6):

Let X_1 and X_2 be two Gaussian random variables such that: $\mu_1 = 0$, $\sigma_1^2 = 4$, $\mu_2 = 10$, $\sigma_2^2 = 9$, $\rho_{1,2} = 0.25$. Define $Y = 2X_1 + 3X_2$

- a. Find the mean and variance of Y
- b. Find $P(Y \le 35)$.

$$\theta_{y}^{2} = 4 \theta_{x_{1}}^{2} + 9 \theta_{x_{2}}^{2} + (2) C_{1} C_{2} \theta_{x_{1}} \theta_{x_{2}} P_{x_{1}} P_{x_{2}}$$

$$= (4)(4) + (9)(9) + (2)(2)(3)(2)(3)(0.25)$$

$$= 115$$

$$\phi\left(\frac{35-30}{\sqrt{115}}\right) = \phi(0.466) \rightarrow 0$$

this gonna both are gaussian be gaussian

EXAMPLE (5-7):

Let X_1 and X_2 be two independent Gaussian random variables such that: $\mu_1 = 0$, $\sigma_1^2 = 4$,

$$\mu_2 = 10$$
, $\sigma_2^2 = 9$. Define $Y = 2X_1 + 3X_2$

- c. Find the mean and variance of Y
- d. Find $P(Y \le 35)$.

a.
$$\mu_y = 2 \mu_{x_1} + 3 \mu_{x_2}$$

$$= (2)(0) + (3)(10) = 30$$

$$= (4)(4) + (9)(9) = 97$$

b. Since its gaussian

$$\Phi\left(\frac{35-30}{\sqrt{97}}\right) = \Phi\left(0.5077\right)$$

Soft-drink cans are filled by an automated filling machine. The mean fill volume is 330 ml and the standard deviation is 1.5 ml. Assume that the fill volumes of the cans are independent Gaussian random variables. What is the probability that the average volume of 10 cans selected at random from this process is less than 328 ml.

$$\mu_{\rm x} = 330 \, {\rm ml.}$$
 , $\theta_{\rm x} = 1.5 \, {\rm ml.}$, gaussian

$$\mu_{x}^{x} = \mu_{x}^{x} = 330$$

$$6^{1/2}_{1} = \frac{6^{1/2}_{1}}{6^{1/2}_{1}} = 0.885$$

$$6x = \frac{0}{2} = \frac{0.30}{10} = 0.30$$

$$\rho(M<328) = \phi(328-330) = \phi(-4.31) = 1-\phi(4.31)$$

Recorded lecture
$$y = x_1 + x_2 + x_3 + \dots + x_n$$

y Gaussian

$$A^2 = \prod_{n} [X_1 + X_2 + \ldots + X_n]$$



EXAMPLE (5-10):

An electronic company manufactures resistors that have a mean resistance of 100 Ω and a standard deviation of 10Ω . Find the <u>probability</u> that a random sample of n = 25 resistors will have an average resistance less than 95 Ω .

$$\mu = 100$$
 , STD = ω , Size = 25

$$\rho(\mu_{\chi}^{\circ} < 95) = \phi\left(\frac{95 - \mu_{\chi}^{\circ}}{9\chi^{\circ}}\right) = \phi\left(\frac{95 - \log}{2}\right) = \phi(-2.5)$$

$$= 1 - \phi(2.5)$$

$$0 \mu_{X}^{2} = \mu_{X} = 100$$

$$\begin{array}{cccc}
0 & \mu_{x}^{0} &= \mu_{x} &= 100 \\
0 & \theta_{x}^{0} &= \frac{\theta_{x}^{0}}{n} &= \frac{100}{25} &= 4
\end{array}$$

EXAMPLE (5-11):

The lifetime of a special type of battery is a random variable with mean 40 hours and standard deviation 20 hours. A battery is used until it fails, then it is immediately replaced by a new one. Assume we have 25 such batteries, the lifetime of which are independent, approximate the probability that at least 1100 hours of use can be obtained.

$$L + G_{\lambda}^{2} = u G_{\lambda}^{2} = (35)(400) = 10,000$$

$$\sqsubseteq$$
 $\epsilon_{y} = loo$.

$$P(\lambda \ge 109) = 1 - b(\lambda < 1100) = 1 - \phi(1100 - 1000)$$

EXAMPLE (5-12):

Suppose that the random variable X has a uniform distribution: over the interval $0 \le X \le 1$. A random sample of size 30 is drawn from this distribution.

a. Find the probability distribution of the sample mean $\hat{\mu}_{x}$

b. Find $P(\hat{\mu}_{x}) < 0.52$

withorm dis.
$$\mu_{x} = \frac{a+b}{2} = 0.5$$

$$\theta_{x}^{2} = \frac{(b-a)^{2}}{12} = 1/12$$

a)
$$\mu_{X}^{2} = \mu_{X}^{2} = 0.5$$

$$\theta_{X}^{2} = \theta_{X}^{2} = \frac{1}{2} = \frac{1}{30} = \frac{1}{30}$$

b)
$$\rho(\hat{\mu}_{X} < 0.52) = \rho(\hat{\mu}_{X} < 0.52 - 0.5) = \phi(0.379)$$

EXAMPLE (5-13):

Suppose that X is a discrete distribution which assumes the two values 1 and 0 with equal probability. A random sample of size 50 is drawn from this distribution.

a. Find the probability distribution of the sample mean $\hat{\mu}_{x}$

b. Find $P(\hat{\mu}_{x}) < 0.6$

$$pdf = f_X(x) = \begin{cases} 0.5 & x = 1 \\ 0.5 & x = 0 \end{cases}$$

a)
$$\mu_{X}^{0} = \mu_{X} = 0.5$$
 $\rightarrow 6^{\frac{2}{50}} = \frac{6^{\frac{2}{50}}}{50} = \frac{0.25}{50} = \frac{1}{200}$
b) $P(\mu_{X}^{0} < 0.6) = \Phi(0.6 - 0.5) = \Phi(1.414)$

ومن لمح فجر الأجر: هان عليه ظلام التكليف،