

# Course - I, Week - 1

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## \* System of sentences:

System 1

Dog is Black

Bird is Red

complete

System 2

Dog is Black

~~Bird is Black~~

Redundant

System 3

Dog is Black

~~Dog is white~~

contradictory

Non-singular  
rank 2

singular  
rank 1

singular  
rank 0

## \* Systems of equations

$$\begin{array}{l} a + b = 10 \\ a + 2b = 12 \end{array}$$

Unique solution

$$a = 8$$

$$b = 2$$

complete

System 2

$$a + b = 10$$

$$2a + 2b = 20$$

infinite sol

$$a = 8, 7, 6, \dots$$

$$b = 2, 3, 4, \dots$$

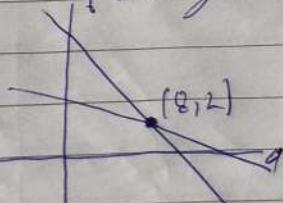
System 3

$$a + b = 10$$

$$2a + 2b = 24$$

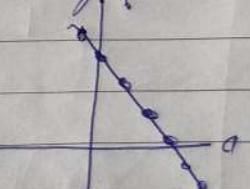
No solution

Non-singular



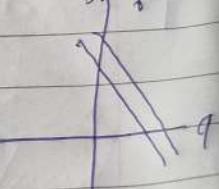
Rank 2

singular



Rank 1

singular



Rank 0

## System of equations as lines

(1)

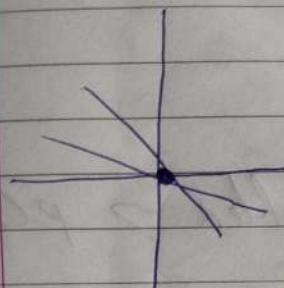
$$\begin{aligned} a+b &= 0 \\ a+2b &= 0 \end{aligned}$$

(2)

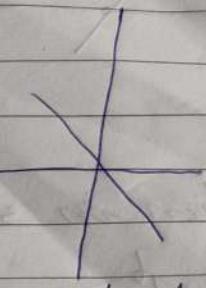
$$\begin{aligned} a+b &= 0 \\ 2a+2b &= 0 \end{aligned}$$

(3)

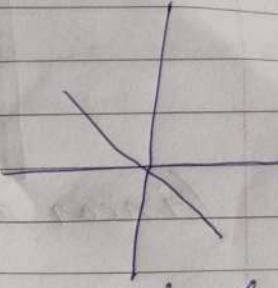
$$\begin{aligned} a+b &= 0 \\ 2a+2b &= 0 \end{aligned}$$



complete



Redundant



Redundant

## System of equations as matrices

(1)

$$\begin{aligned} a+b &= 0 \\ a+2b &= 0 \end{aligned}$$

non-singular  
System

$$\begin{matrix} 1 & 1 \\ 1 & 2 \end{matrix}$$

non-singular  
matrix

(2)

$$\begin{aligned} a+b &= 0 \\ 2a+2b &= 0 \end{aligned}$$

singular  
System

$$\begin{matrix} 1 & 1 \\ 2 & 2 \end{matrix}$$

singular  
matrix

Dependent - Singular  
Independent - Non-singular

↓  
Second row  
is dependent  
on first row

Second row  $\Rightarrow$  first row + 2

↓  
Linearly dependent

## Determinant :

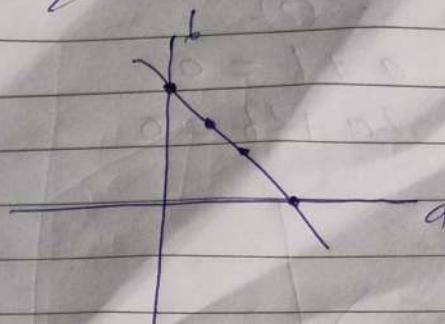
16

non-singular

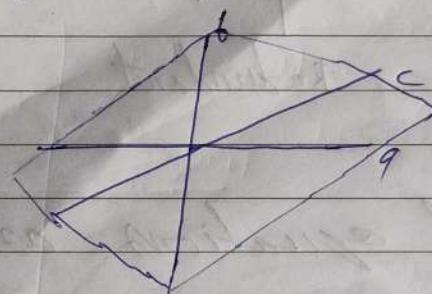
0

Singular

Linear equation in 2 variables  $\rightarrow$  Line



Linear equation in 3 variables  $\rightarrow$  plane



course - 1, week 2

System of equations to matrices

original system

$$5a + b = 17$$

$$4a - 3b = 6$$

intermediate system

$$a + 0.2b = 3.4$$

$$b = 2$$

solved system

$$a + 0.2b = 3$$

$$0a + 1b = 2$$

original matrix

$$\begin{matrix} 5 & 1 \\ 4 & -3 \end{matrix} \rightarrow$$

upper diagonal matrix

$$\begin{matrix} 1 & 0.2 \\ 0 & 1 \end{matrix}$$

Diagonal matrix

$$\begin{matrix} 1 & 0 \\ 0 & 1 \end{matrix}$$

Row echelon form

Reduced row echelon form.

Rank : In row echelon form  $\Rightarrow$   
 Number of ones in the diagonal

Reduced row echelon form

$$\begin{matrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{matrix}$$

If it is in row echelon form

$\Rightarrow$  each pivot is 1

$\Rightarrow$  Any number above a pivot is 0

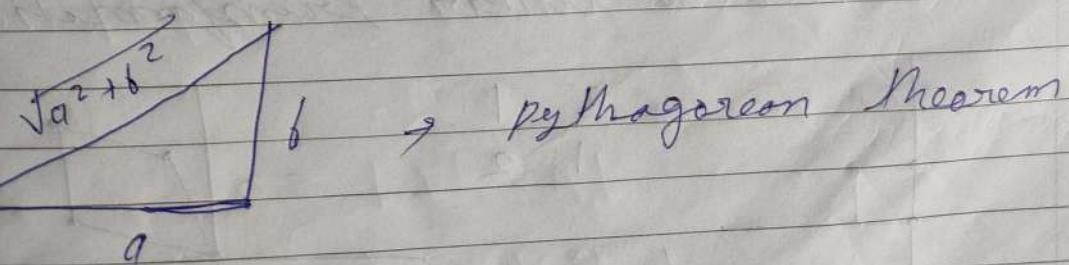
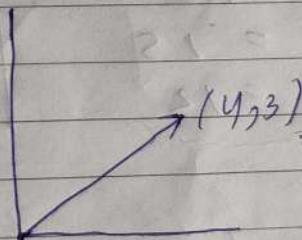


Rank  
5

$\Rightarrow$  Rank of the matrix is the number of pivots

course - 1, Week 3

Vectors

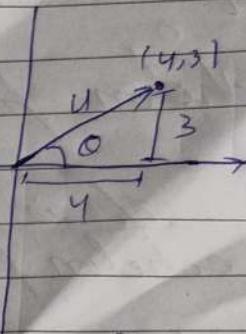


$\Rightarrow$  Pythagorean Theorem

$$\| (a, b) \|_1 = |a| + |b|$$

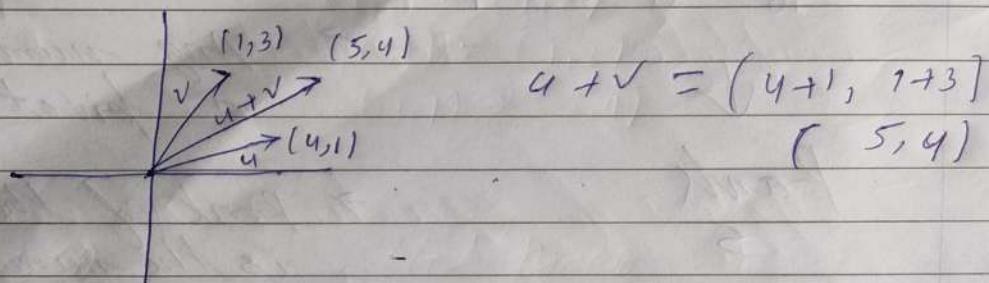
$$\| (a, b) \|_2 = \sqrt{a^2 + b^2}$$

## Direction of a vector



$$\tan(\theta) = \frac{3}{4}$$

## Sum of vectors



$$u + v = (4+1, 1+3) \\ (5, 4)$$

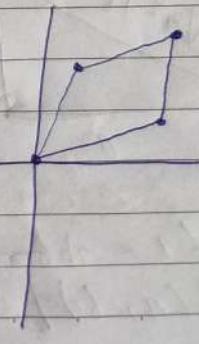
## Equation as dot product

system of equations $a+b+c = 10$ $a+2b+c = 15$ $a+b+2c = 12$	matrix product $\begin{vmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{vmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 10 \\ 15 \\ 12 \end{pmatrix}$
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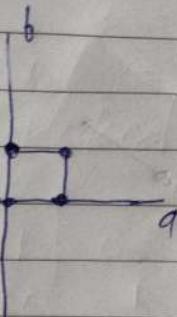
## Matrices as linear transformations

$$\begin{matrix} 3 & 1 \\ 1 & 2 \end{matrix}$$

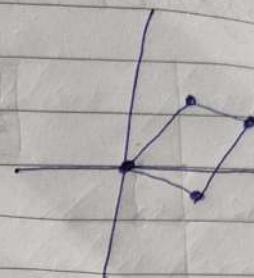
$(0,0) \rightarrow (0,0)$   
 $(1,0) \rightarrow (3,1)$   
 $(0,1) \rightarrow (1,2)$   
 $(1,1) \rightarrow (4,3)$



# Linear transformation as matrices



$$\begin{array}{|c|c|} \hline 2 & 2 \\ \hline \cdot & \cdot \\ \hline 2 & 2 \\ \hline \cdot & \cdot \\ \hline \end{array}$$



first

$$(0,0) \rightarrow (0,0)$$

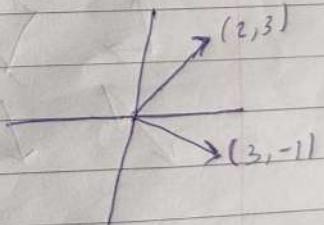
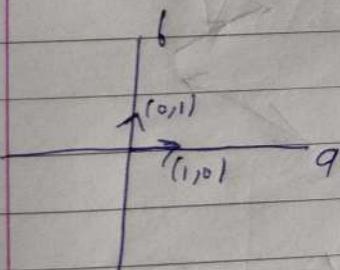
$$(1,0) \rightarrow (3, -1)$$

$$(0,1) \rightarrow (2, 3)$$

$$(1,1) \rightarrow (5, 2)$$

$$\left[ \begin{array}{cc|c} 2 & 2 & 0 \\ \cdot & \cdot & \cdot \\ 2 & 2 & 1 \\ \cdot & \cdot & \cdot \end{array} \right] = 2$$

$$\left[ \begin{array}{cc|c} 2 & 2 & 1 \\ \cdot & \cdot & 0 \\ 2 & 2 & 0 \\ \cdot & \cdot & \cdot \end{array} \right] = 3$$



$$\left[ \begin{array}{cc|c} 2 & 2 & 0 \\ \cdot & 3 & 1 \\ \cdot & \cdot & \cdot \end{array} \right] = 2$$

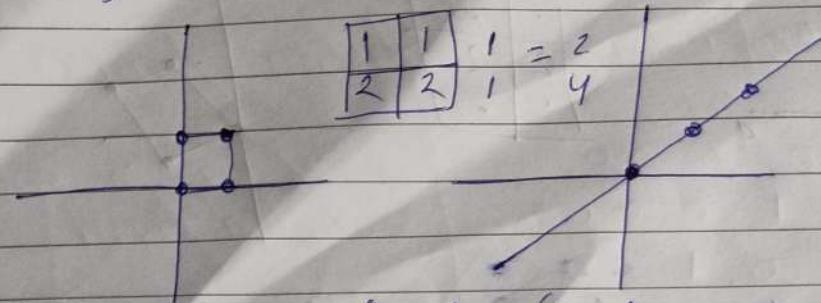
$$\left[ \begin{array}{cc|c} 2 & 2 & 1 \\ 2 & 3 & 0 \\ \cdot & \cdot & \cdot \end{array} \right] = 3$$

$$\left[ \begin{array}{cc|c} 3 & 2 & 0 \\ -1 & 3 & 1 \\ \frac{3}{-1} & 2 & 0 \end{array} \right] = 2$$

course - 1 week 4

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## singular transformation



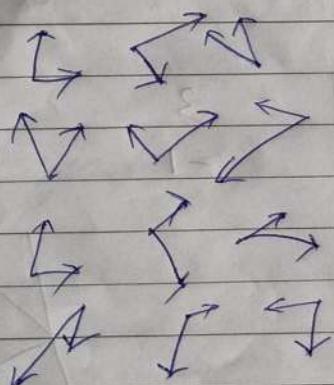
$$(0,0) \rightarrow (0,0)$$

$$(1,0) \rightarrow (1,2)$$

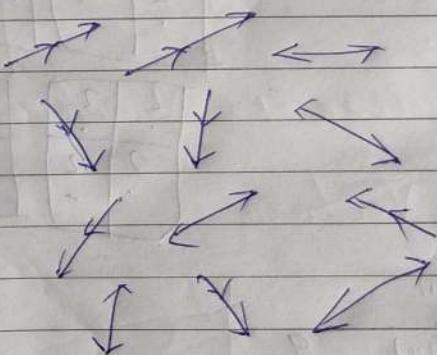
$$(0,1) \rightarrow (1,1)$$

$$(1,1) \rightarrow (2,4)$$

Bases



Not Bases



Inverse of a matrix  $A^{-1} = \frac{1}{|A|} \cdot \text{adj} A$

eigen vector :

$$|A - \lambda I| = 0$$

$$AV = \lambda V$$

slope = height  
width

$$f(n) = n^2$$

$$f'(n) = 2n$$

$$f'(n) = \frac{\Delta f}{\Delta n}$$

$$\frac{\Delta f}{\Delta n} = \frac{(n + \Delta n)^2 - n^2}{\Delta n}$$

$$= \frac{n^2 + 2n\Delta n + \Delta n^2 - n^2}{\Delta n}$$

$$= \frac{\Delta n (2n + \Delta n)}{\Delta n}$$

$$f'(n) = 2n + \Delta n$$

$$\Delta n \rightarrow 0$$

$$f'(n) \rightarrow 2n$$

$$f'(\sin n) = \cos n$$

$$f'(\cos n) = -\sin n$$

$$f'(e^n) = e^n$$

~~$$\lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n \rightarrow e^\bullet$$~~

JEP

corner | cusp  
M T W G S  
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YOU ARE  
jump discontinuity

$$f = g + h$$

$$f' = g' + h'$$

$$f = gh$$

$$f' = g'h + h'g$$

$$\frac{dy}{du} = \frac{dy}{du} \frac{du}{du}$$

$$f(u) = u^2 + y^2$$

$$\frac{d f(u)}{du} = 2u \Rightarrow 0, u=0$$

$$\frac{dy}{du} = 2y \Rightarrow 0, y=0$$

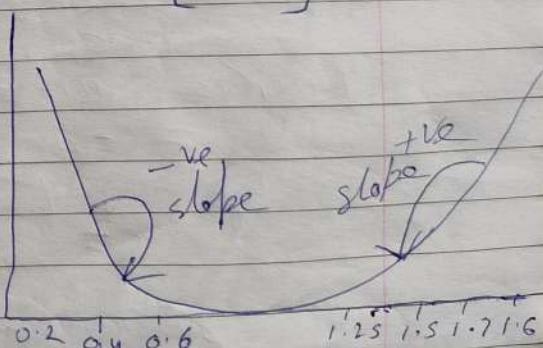
Gradient  $\begin{bmatrix} 2u \\ 2y \end{bmatrix}$  minimum

Gradient at (2, 3)  $\begin{bmatrix} 4 \\ 6 \end{bmatrix}$

new point  $\hat{u}$   
old point - slope

$$u_k = u_{k-1} - \frac{f(u_k)}{f'(u_k)}$$

one variable



2 variable

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define a &  
choose a starting point  $(n_0, g_0)$

update :

$$\begin{bmatrix} n_k \\ g_k \end{bmatrix} = \begin{bmatrix} n_{k-1} \\ g_{k-1} \end{bmatrix} - \alpha \nabla f(n_{k-1}, g_{k-1})$$

$$\sigma(z) = \frac{1}{1 + e^{-z}}$$

$$\sigma'(z) = d f' \left( \frac{1}{1 + e^{-z}} \right)$$

$$= (1 + e^{-z})^{-1}$$

$$= -1 (1 + e^{-z})^{-2}$$

$$= -(1 + e^{-z})^{-2} f'(e^{-z})$$

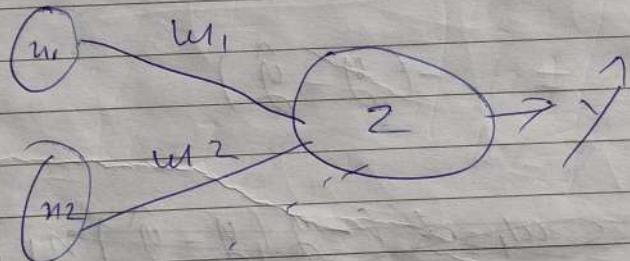
$$= (1 + e^{-z})^{-2} (e^{-z})$$

$$= (1 + e^{-z})^{-2} (e^{-z})$$

$$= \frac{e^{-z}}{(1 + e^{-z})^2}$$

$$= \frac{e^{-z} + 1 - 1}{(1 + e^{-z})^2}$$

$$\begin{aligned}
 &= \frac{1 + e^{-z}}{(1 + e^{-z})^2} - \frac{1}{(1 + e^{-z})^2} \\
 &= \frac{1}{1 + e^{-z}} - \left( \frac{1}{1 + e^{-z}} \right) \left( \frac{1}{1 + e^{-z}} \right) \\
 &= \frac{1}{1 + e^{-z}} \left( 1 - \frac{1}{1 + e^{-z}} \right) \\
 &= g(z) (1 - g(z))
 \end{aligned}$$



(b)

$$z = w_1 w_1 + w_2 w_2 + b$$

$$MSE = \frac{1}{2} (y - \hat{y})^2$$

$$\frac{dL}{d\hat{y}} f'(L) = -(y - \hat{y})$$

~~done~~

$$\frac{dL}{dw_1} = \frac{dL}{dy} \frac{dy}{dw_1}$$

$$\frac{dL}{db} = \frac{dL}{dy} \frac{dy}{db}$$

$$\frac{dL}{dw_2} = \frac{dL}{dy} \frac{dy}{dw_2}$$

$$\frac{dL}{dw_1} = -(y - \hat{y}) n_1$$

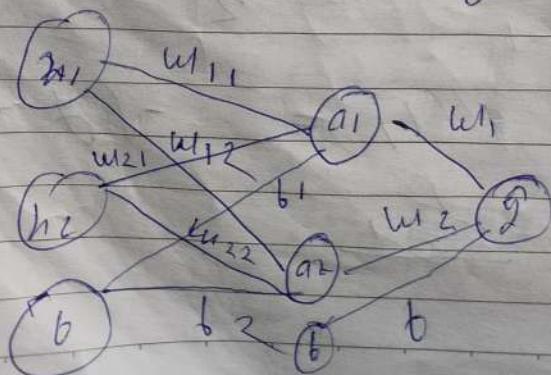
$$\frac{dL}{dw_2} = -(y - \hat{y}) n_2$$

$$\frac{dL}{db} = -(y - \hat{y})$$

~~$w_1 = w_1 - \alpha (-(\hat{y} - y) n_1)$~~

$$w_2 = w_2 - \alpha (-\hat{y})$$

$$\left. \begin{array}{l} w_1 = w_1 - \alpha (-(\hat{y} - y) n_1) \\ w_2 = w_2 - \alpha (-(\hat{y} - y) n_2) \\ b = b - \alpha (-\hat{y}) \end{array} \right\}$$



$$a_1 = \sigma(z_1)$$

$$z_1 = u_1 w_{11} + u_2 w_{21} + b_1$$

$$a_2 = \sigma(z_2)$$

$$z_2 = u_1 w_{12} + u_2 w_{22} + b_2$$

$$\hat{y} = \sigma(z)$$

$$z = a_1 u_1 + a_2 u_2 + b$$

$$\frac{dL}{du_{11}} = \frac{dz_1}{du_{11}} \frac{da_1}{dz_1} \frac{dz}{da_1} \frac{dy}{dz} \frac{dL}{dy}$$

$$L(y, \hat{y}) = -y \log(\hat{y}) - (1-y) \log(1-\hat{y})$$

$$\frac{dL}{d\hat{y}} = \frac{-y}{\hat{y}} - \frac{(1-y)}{(1-\hat{y})} (-1)$$

$$= \frac{-y}{\hat{y}} + \frac{(1-y)}{(1-\hat{y})}$$

$$= \frac{-y(1-\hat{y}) + (1-y)\hat{y}}{\hat{y}(1-\hat{y})}$$

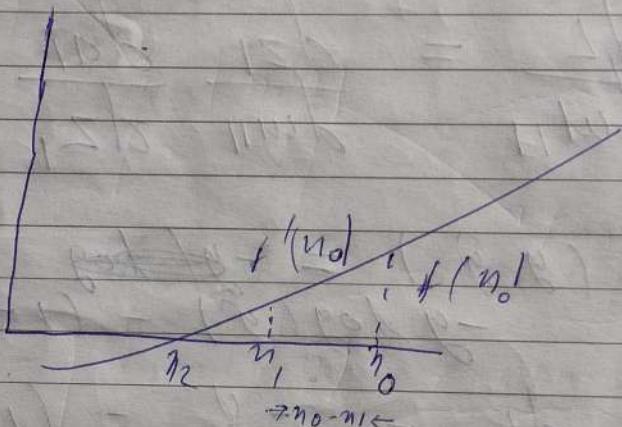
$$= \frac{-y + y\hat{y} + \hat{y} - y\hat{y}}{\hat{y}(1-\hat{y})}$$

$$= \frac{\hat{y} - y}{\hat{y}(1-\hat{y})}$$

$$\frac{dL}{du_{11}} = n_1 \cdot a_1 (1 - a_1) \cdot u_1 \cdot \frac{\partial}{\partial u} \left( \frac{y}{1+y} \right) \cdot \frac{\partial}{\partial y}$$

$$\boxed{f'(u_{11}) = n_1 u_1 a_1 (1 - a_1) (\hat{Q} - \hat{y})}$$

$$u_{11} = u_{11} - \alpha \frac{dL}{du_{11}}$$



$$f(n) = \frac{f(n)}{n_0 - n_1}$$

$$n_0 - n_1 = \frac{f(n)}{f'(n)}$$

$$n_1 = n_0 - \frac{f(n)}{f'(n)}$$

$$\text{Newton's method } n_{k+1} = n_k - \frac{f(n_k)}{f'(n_k)}$$

NM for optimization

- ↗ start with  $n_0$
- ↗ update

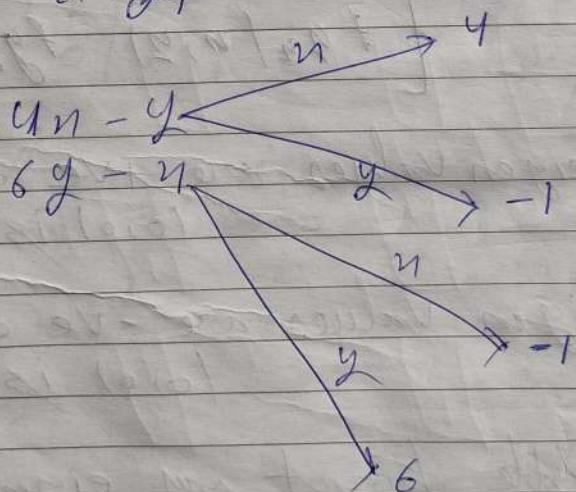
$$n_{k+1} = n_k - \frac{f'(n_k)}{f''(n_k)}$$

second

$$f(n, y) = 2n^2 + 3y^2 - ny$$

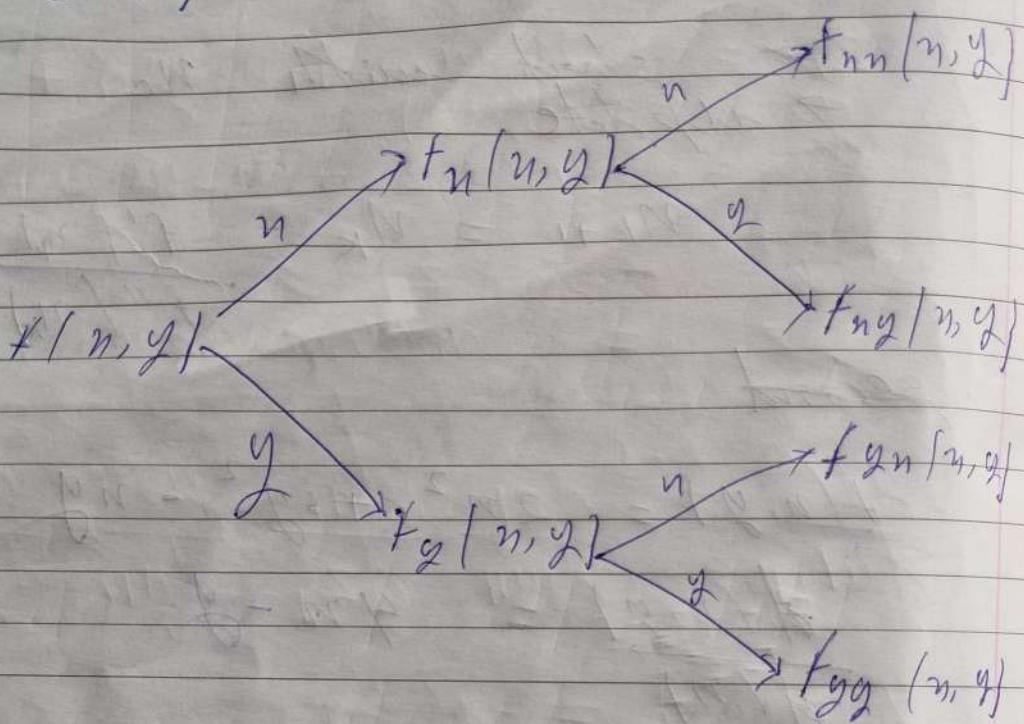
$$\frac{\partial f(n, y)}{\partial n} = 4n - y$$

$$\frac{\partial f(n, y)}{\partial y} = 6y - n$$



Hessian matrix  $H = \begin{bmatrix} 4 & -1 \\ -1 & 6 \end{bmatrix}$

~~$f(n,y)$~~



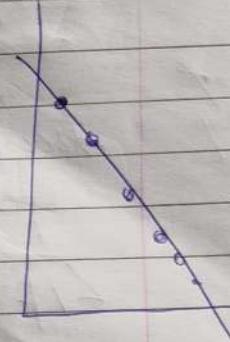
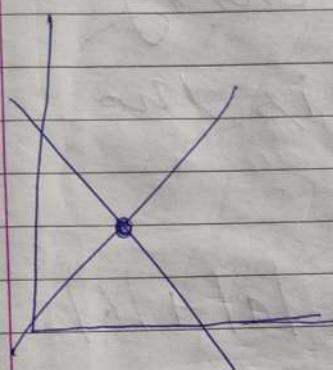
$$H(n, y) = \begin{bmatrix} f_{nn}(n, y) & f_{ny}(n, y) \\ f_{y_n}(n, y) & f_{yy}(n, y) \end{bmatrix}$$

- ↗ If eigen values are +ve or > 0  
(0,0) is a minimum
- ↗ If eigen values are -ve or < 0  
(0,0) is a maximum
- ↗ Saddle point any eigen value is -ve

Netwon

$$\begin{bmatrix} n_{k+1} \\ y_{k+1} \end{bmatrix} = \begin{bmatrix} n_k \\ y_k \end{bmatrix} - H^{-1}(n_k, y_k) \nabla f(n_k, y_k)$$

complete contradictory  
 ↓  
 non-singular      singular  
 ↓  
 unique solution    no solution      infinite solution



Row echelon form      Reduced row echelon

$$\begin{array}{|c|c|} \hline 1 & 0.6 \\ \hline 0 & 1 \\ \hline \end{array}$$

$$\begin{array}{|c|c|} \hline 1 & 1 \\ \hline 0 & 1 \\ \hline \end{array}$$

Rank of matrix + the number of 'ones' in diagonal is rank echelon.

↓  
 Rank 2

$$\begin{array}{|c|c|} \hline 0 & 0 \\ \hline 0 & 1 \\ \hline \end{array}$$

Rank 1

$$\begin{array}{|c|c|} \hline 0 & 0 \\ \hline 0 & 0 \\ \hline \end{array}$$

Rank 0

Row echelon form

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$$\begin{array}{|c|c|c|} \hline ② & * & x \\ \hline 0 & ① & x \\ \hline 0 & 0 & ③ \\ \hline \end{array}$$

rank  $\rightarrow 3$

Each row has 1 pivot

left most non-zero entry

\* Rank of the matrix is the number of pivots.

$$L_1\text{-norm} = \|(a, b)\|_1 = |a| + |b|$$

$$L_2\text{-norm} = \|(a, b)\|_2 = \sqrt{a^2 + b^2}$$

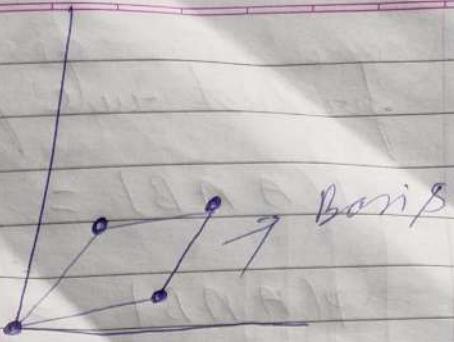
$\Rightarrow$  matrices as linear transformation

$$\begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

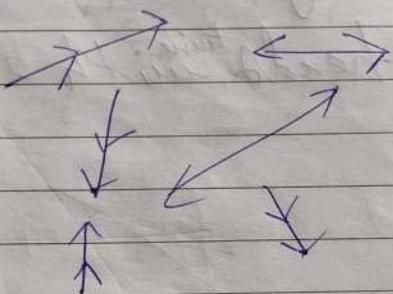
$$" \quad \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

$$" \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$" \quad \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$



Not Basis



Inverse of a matrix :

$$A^{-1} = \frac{1}{|A|} \cdot \text{adj } A$$

eigen values  $|A - \lambda I| = 0$

eigen vector  $AV = \lambda V$

course-3 states a prob

$$P'(n) = 1 - P(n)$$

Disjoint  $P(A \cup B) = P(A) + P(B)$

Joint  $P(A \cap B) = P(A) + P(B) - P(A \cup B)$

abq  
dbqcabq

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product rule for independent events

$$P(A \cap B) = P(A) \cdot P(B)$$

$$P(A \cap B) = \underset{\text{else}}{P(A)} \cdot P(B|A)$$

conditional probability

→ prob of getting heads twice with 2 coin

$$\frac{1}{4}$$

→ with condition the first one is head

$$\frac{1}{2}$$

Bayes Theorem -

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

→ prior → email → spam

posterior → Event - email contains lottery

$$P(\text{spam} | \text{lottery})$$

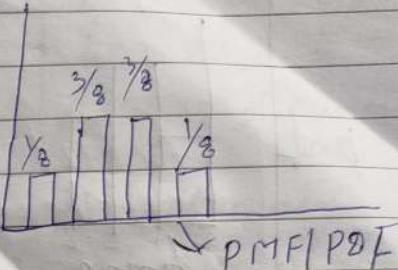
## discrete probability distribution :

number of heads 3 in one coin

$$P_n(y) \geq 0$$

$$\sum P_n(y) = 1$$

Binomial distribution :



$$I flip 5 coins, 2 of them land in head$$

$$n \in \{0, 1, 2, \dots, 5\}$$

$$\binom{n}{k} p^k (1-p)^{n-k}$$

Head Trail

Bernoulli distribution :

Number of heads flip a coin

$$P(X=1) = 0.5 \quad P(X=0) = 0.5$$

success

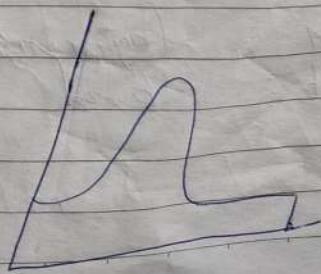
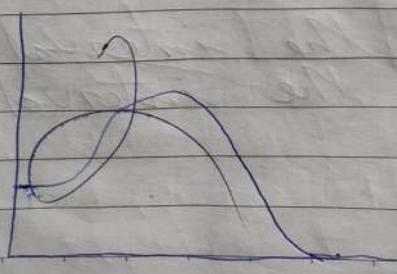
failure

discrete to continuous  $P$  <sup>interval</sup>:

fix

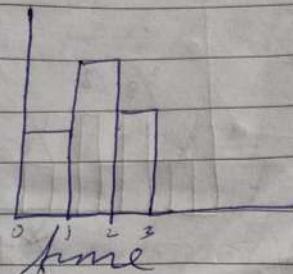
discrete : sum of heights equal 1

continuous : Area under the curve equals 1

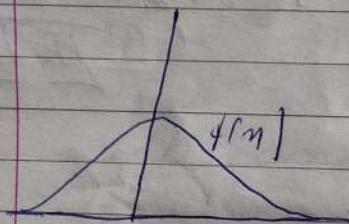


cumulative distributed probability

PDF

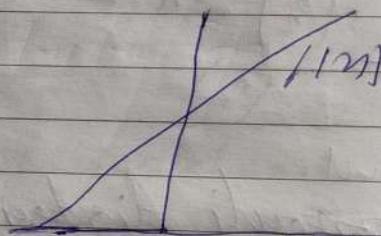


PDF



area = 1  
always +ve

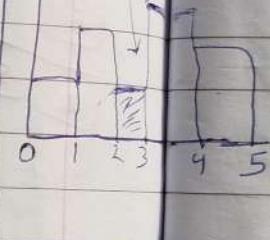
CDF



left end point is 0  
right end point is 1  
always +ve and  
increasing

f area

$p(2 \leq n^3)$



pleasable  
= 0 = a

6 hours

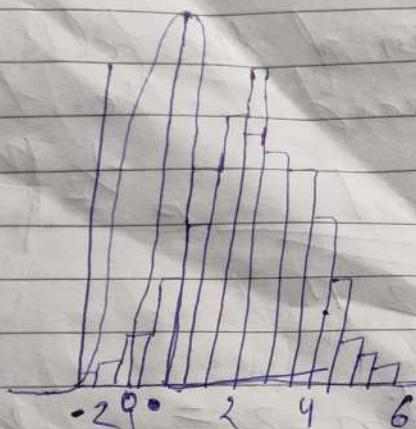
Uniform Distribution :

$$f_n(n) = \frac{1}{b-a} \quad a < n < b \\ n \in (a, b)$$

a : beginning of the interval  
b : end of the interval

## Normal Distribution :-

$$e^{-\frac{n^2}{2}}$$



uniform distri...  
any value b/w 0 & 15 minutes must

have the same  
frequency of occurrence

The pdf must be  
constant for all values  
in the interval (0, 15)

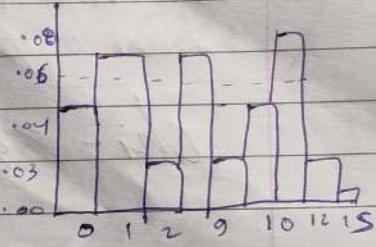
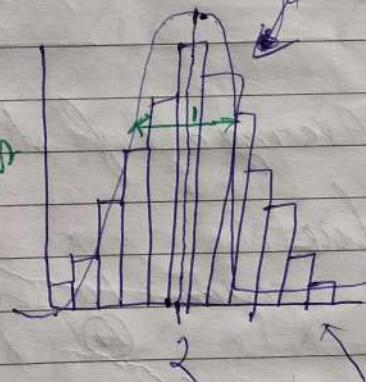
area



$$e^{-\frac{1}{2}(n-2)^2}$$

(exactly at 2 mins)  
 $\approx 0 = \text{area}$

the curve is

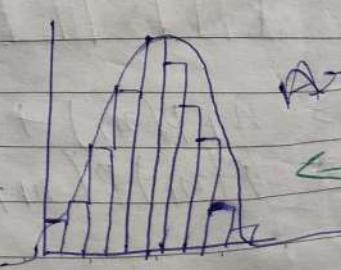
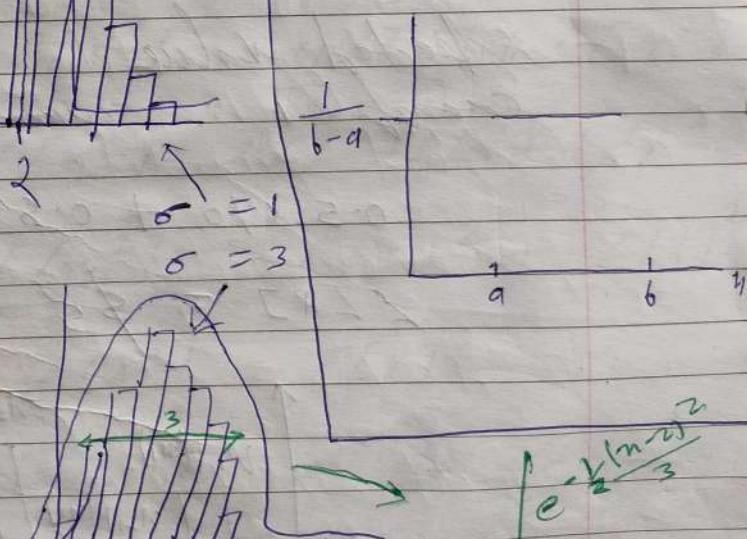


which constant  
 $15 \times h = 1, h = 0.06$

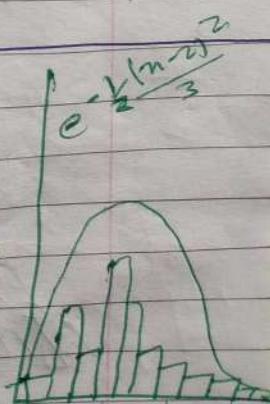
you think

$$\text{Area } \frac{1}{3\sqrt{2\pi}}$$

$$\frac{1}{3\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{n-2}{3}\right)^2}$$



Area 1



parameters :

$\mu$ : center of the bell

$\sigma$ : spread of the bell

$$f_x(n) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{n-\mu}{\sigma}\right)^2}$$

$$\text{standardization } z = \frac{n - \mu}{\sigma}$$

chi-square distribution

$$W = z^2$$

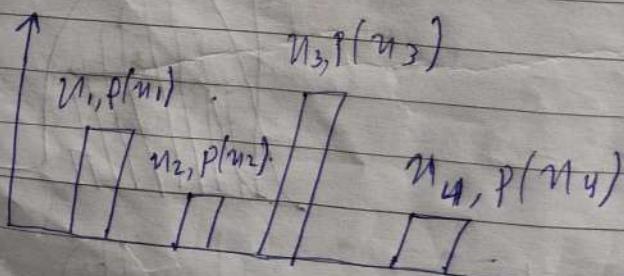
Median : Average of the 2 middle ones

Expected value

example 1 : min. 10 dollars or nothing

$$0.5 \times 10 + 0.5 \times 0 = \$5$$

$$E[X] = \$5$$



$$E[X] = n_1 p(n_1) + n_2 p(n_2) + n_3 p(n_3) + n_4 p(n_4)$$

$$E[E(Y)] = f(n_1) p(n_1) + f(n_2) p(n_2) + f(n_3) p(n_3) + f(n_4) p(n_4)$$

## Sum of Expectations

(H)  
T

$\min \$1 + \min \$1 + \$2 + \$3 + \$4 + \$5 + \$6$   
 $\min \text{nothing}$

$$E[X_{\text{coin}}] = \$0.5, E[X_{\text{dice}}] = \$3.5$$

$$\begin{aligned} E[X] &= E[X_{\text{coin}}] + E[X_{\text{dice}}] \\ &= 0.5 + 3.5 = \$4 \end{aligned}$$

$$\boxed{E[X_1 + X_2] = E[X_1] + E[X_2]}$$

## Variance

$$\text{Var}(X) = E[(X - \mu)^2]$$

$$\text{Var}(X) = E[(X - \mu)^2] = E[X^2 - 2\mu X + \mu^2]$$

$$= E[X^2] - 2\mu E[X] + E[\mu^2]$$

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

$$\text{std}(X) = \sqrt{\text{Var}(X)}$$

standardize a distribution

$$\frac{X - \mu}{\sigma}$$

Skewness

$$E\left[\left(\frac{n-u}{\sigma}\right)^3\right]$$

$\rightarrow$ +vely skewed	$> 0$	$\leftrightarrow$ not skewed	$= 0$	-vely skewed $< 0$
-------------------------------	-------	---------------------------------	-------	--------------------

kurtosis

$$E\left[\left(\frac{n-u}{\sigma}\right)^4\right]$$

Q2 Median  $\rightarrow$  second quartile  $\rightarrow$  50% quantile

Q1 first quartile  $\rightarrow$  25%.

Q3 third quartile  $\rightarrow$  75%.

Interquartile range (IQR)

$$IQR = Q_3 - Q_1$$

\* joint distribution:

$$P_{X,Y}(n, g) = P(X=n, Y=g)$$

Total	Age X	Height Y
child	$P(9, 30) = \frac{1}{10}$	30    40    50    60 0       0       0       0
	X	7       8       9 0       0       0 1/10    0       0/10    0
		prob      0       0       0

$X$  and  $Y$  are independent

$$P_{XY}(n, y) = P(n) \cdot P(y)$$

\* Marginal distribution

Distribution of one variable while ignoring the other.

\* conditional distribution

if age = 9, what is the distribution across the height variable

$$P_{Y|X=9}(y) = P(Y=y | X=9)$$

\* divide by row sum

$$P_{Y|X=n}(y) = \frac{P_{XY}(n, y)}{P_X(n)}$$

joint  
 PDF of  $X, Y$   
 ↑  
 marginal  
 distribution  
 of  $X$

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

equal probabilities

unequal probabilities :

$$\sum P_{xy} (u_i, y_i) / (u_i - \bar{u}_x) (y_i - \bar{y}_y)$$

covariance Matrix

$$\begin{bmatrix} \text{Var}(x) & \text{cov}(x, y) \\ \text{cov}(x, y) & \text{Var}(y) \end{bmatrix}$$

correlation coefficient  $\rightarrow$

$$\frac{\text{cov}(x, y)}{\sqrt{\text{Var}(x)} \cdot \sqrt{\text{Var}(y)}}$$

Multivariate Gaussian Distribution

if  $W, H$  were independent

$$f_{HW}(h, w) = f_H(h) f_W(w)$$

$$= \frac{1}{\sqrt{2\pi}\sigma_H} e^{-\frac{1}{2} \frac{(h-\bar{u}_H)^2}{\sigma^2_H}} \cdot \frac{1}{\sqrt{2\pi}\sigma_W} e^{-\frac{1}{2} \frac{(w-\bar{u}_W)^2}{\sigma^2_W}}$$

$$= \frac{1}{2\pi\sigma_H\sigma_W} e^{-\frac{1}{2} \left( \frac{(h-\bar{u}_H)^2}{\sigma^2_H} + \frac{(w-\bar{u}_W)^2}{\sigma^2_W} \right)}$$

Course - 3

population  $\rightarrow$  the entire group

Sample  $\rightarrow$  subset of the population

population proportion

$$P = \frac{\text{number of items with a given characteristic}}{\text{population } (n)}$$

Sample Variance

$$\text{Var}(n) = \frac{E(n - \bar{n})^2}{n-1}$$

Law of large numbers :-

As the sample size increases the average of the sample will tend to get closer to the average of the entire population.

$n$ : number of samples

$X_i$ : some estimate  $X$  for a sample size  $i$

as  $n \rightarrow \infty$

$$\frac{1}{n} \sum_{i=1}^n X_i \rightarrow E[X] = \mu_n$$

under certain condition :

- 7 Sample is randomly drawn
- 7 sample size must be sufficiently large
- 7 independent observation

Central Limit Theorem :

$$\begin{aligned} \mu &= np = np(H) \\ \sigma^2 &= np(1-p) = np(H)p(T) \end{aligned}$$

If As  $n$  increases, the probability distribution becomes closer to a Gaussian distribution

$\bar{X}_n = \mu$  ← population mean  
mean of the sample means

$$\text{Variance of the sample means} \quad \leftarrow \frac{\sigma^2}{n} = \frac{\sigma^2}{n} \leftarrow \text{population variance}$$

Maximum likelihood :

$$(H)(H)(H)(H)(H)(T)(T)(H)(H)(T)(T)$$

Coin 1

$$P(H) = 0.7$$

$$P(T) = 0.3$$

Coin 2

$$P(H) = 0.5$$

$$P(T) = 0.5$$

Coin 3

$$P(H) = 0.3$$

$$P(T) = 0.7$$

Coin 1 has the maximum likelihood to get more heads. ~~more heads~~

$$p = P(H)$$

$$\text{Likelihood } L(p; 8H) = p^8 (1-p)^2$$

$$\log \text{likelihood} = \log [p^8 (1-p)^2]$$

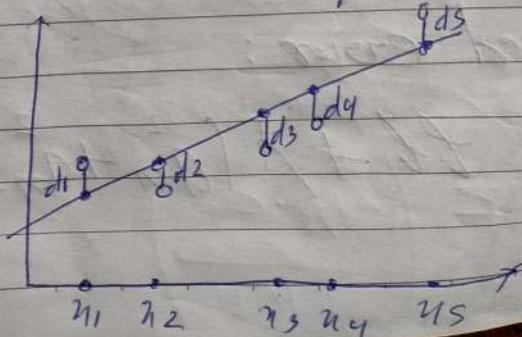
$$= 8 \log p + 2 \log (1-p)$$

$$\frac{d}{dp} (8 \log p + 2 \log (1-p)) = \frac{8}{p} + \frac{2}{1-p} (-1) \\ = 0$$

$$\hat{p} = \frac{8}{10}$$

\* The best distribution is the one where the mean of the distribution is the mean of the sample.

↓  
some for standard deviation



Likelihood :

$$\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}d_1^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}d_2^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}d_3^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}d_4^2}$$

Maximize :

$$e^{-\frac{1}{2}d_1^2} \cdot e^{-\frac{1}{2}d_2^2} \cdot e^{-\frac{1}{2}d_3^2} \cdot e^{-\frac{1}{2}d_4^2} \cdot e^{-\frac{1}{2}ds^2}$$

$$e^{-\frac{1}{2}(d_1^2 + d_2^2 + d_3^2 + d_4^2 + ds^2)}$$

minimize :

$$d_1^2 + d_2^2 + d_3^2 + d_4^2 + ds^2$$

Least squares error

Regularization term :

Model:  $y = a_n n^n + a_{n-1} n^{n-1} + \dots + a_1 n + b$

Log-likelihood:  $\ell l$

$L_2$  regularization error:  $a_n^2 + a_{n-1}^2 + \dots + a_1^2$

" parameter  $\lambda$

Regularized error:

$$\ell l + \lambda (a_n^2 + a_{n-1}^2 + \dots + a_1^2)$$

course - 4

## confidence interval :

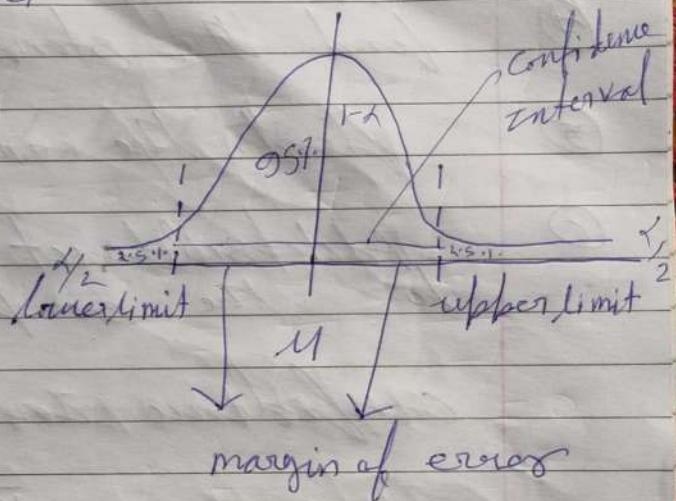
lower limit <  $\bar{x}$  < upper limit

confidence level

$$1 - \alpha, 1 - 0.05 \\ 0.95 \%$$

$$\alpha = 0.05$$

Significance level



mean of the sample  $\bar{x}$

standard deviation of the mean  $\frac{\sigma}{\sqrt{n}}$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

As  $n$  increases, the confidence interval shrinks.

$$\text{margin of error} = Z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

↓                      ↓  
 critical value   standard error

# confidence interval - calculation

STEPS :-

- ① find the sample mean
- ② define a desired confidence level ( $1-\alpha$ )
- ③ get the critical value ( $Z_{\alpha/2}$ )
- ④ find the standard error ( $\sigma/\sqrt{n}$ )
- ⑤ find the margin of error
- ⑥ add / subtract the margin of error to the sample mean

confidence interval =

$$\bar{x} \pm Z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

Assumptions:

- \* simple random sample
- \* sample size  $> 30$  or population is approx. normal
- \* calculating sample size.

$$n = \left( \frac{Z_{\alpha/2} \cdot \sigma}{E} \right)^2$$

## Confidence interval - F Distribution

$$\sigma^2$$

$$\bar{x} \pm Z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

$$\frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$$

$$\frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \text{ student's } t$$

unknown  $\sigma$

$$\bar{x} \pm t_{\alpha/2} \cdot \frac{s}{\sqrt{n}}$$

degrees of freedom :  $n - 1$

\* Confidence Interval for proportions

confidence interval =  $\hat{p} \pm \text{margin of error}$

$$\text{margin of error} = Z_{\alpha/2} \cdot \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

$$P = \frac{n}{N}$$

# Hypothesis testing

Motivation:

Null hypothesis  $H_0$ : email is ham

alternative hypothesis  $H_1$ : email is spam

- Not labeling the email spam, doesn't mean the email is ham  
(~~lack~~ lack of evidence)
- plenty of evidence against  $H_0$   
↓  
reject  $H_0$  (and accept  $H_1$ )

Type I error: False positive

Type II error: False negative

\* Significance Level

What is the greatest probability of type I error you are willing to tolerate?

(a ham email is spam 5% time if  $\alpha = 0.05$ )

Both options are awful decision makers

You want a small  $\alpha$ , but not zero

every time an email is marked ham you make a type error

Emails are always considered ham

no type I error

maximum prob of test

M T W T F S S  
error  
Date: 28/07/2018

YOUVA sunitha come

catch

↓ Type I error ↑ Type II error

for a given sample,  $\bar{x}$  will determine if you reject  $H_0$  or not.

$$\bar{n} = 68.442$$

↓ observed statistic

The mean height for 18 y/o in the CL is the  $\bar{x}$  was 66.7 is.

$$\text{Test statistic } \bar{x} = \frac{1}{10} \sum_{i=1}^{10} x_i$$

3 questions

3 sets of hypothesis

Right-Tailed Test  $H_0: \mu = 66.7$  vs  $H_1: \mu > 66.7$

Left-Tailed Test  $H_0: \mu = 66.7$  vs  $H_1: \mu < 66.7$

Two-Tailed Test  $H_0: \mu = 66.7$  vs  $H_1: \mu \neq 66.7$   
null hypothesis

Alternative hypothesis type I error  
 $\bar{x} > 66.7$

when mean didn't change

If  $\bar{x} > 66.7$

reject  $H_0$

Type II error  
don't reject  $\bar{x} = 66.7$  when  $\mu > 66.7$

known  
↓

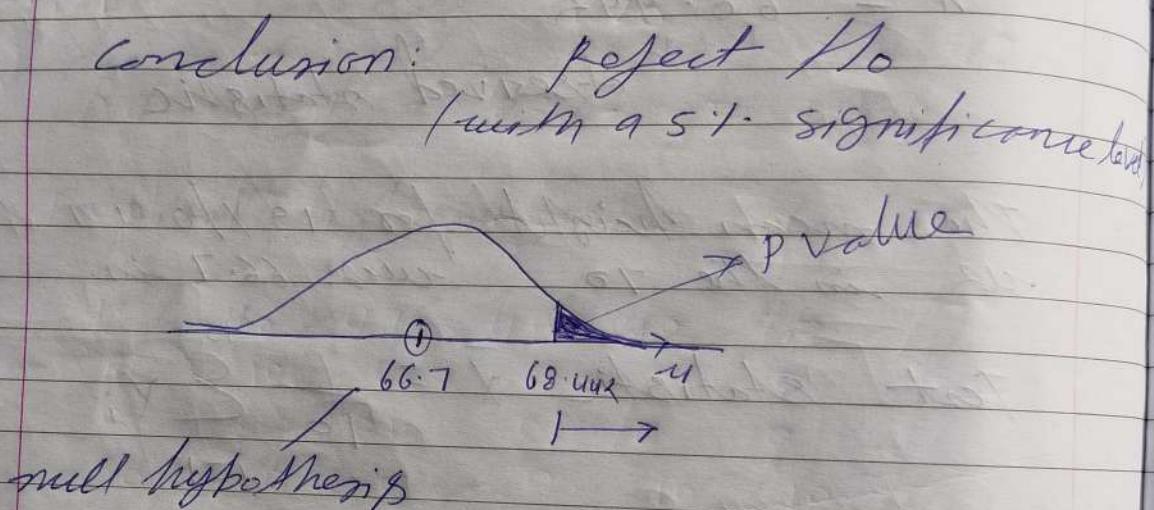
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## Right-Tailed Test for Gaussian Data

$$\sigma = 3, n = 10, \bar{v} = 68.442$$

Goal: Type I error prob  $\alpha = 0.01$

$$P(\bar{X} > 68.442 | \mu = 66.7) = 0.0407$$



Decision rule:

If p-value  $< \alpha$  reject  $H_0$

If p-value  $> \alpha$  don't reject  $H_0$

Z-statistics

$$Z = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

If  $H_0$  is true

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

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Decision

Reality

$H_0$  True  
(Not spam)

$H_0$  False  
(spam)

Reject  $H_0$   
(Decide spam)

Type I error

correct

Don't reject  $H_0$   
(Decide not spam)

correct

Type II error

steps for performing Hypothesis testing

① State your hypotheses

Null hypothesis: the baseline  $\rightarrow H_0: \bar{U} = 66.7$   
Alternative  $H_1$ : the statement u want to prove  
 $H_1: \bar{U} > 66.7$

② Design your test

Decide the test statistic to work with  $\bar{X}$

Decide the significance level  $\rightarrow \alpha = 0.05$

③ Compute the observed statistic  
(based on your sample)  $\rightarrow \bar{U} = 68.442$

④ Reach a conclusion:

If the p-value is less than the significance level reject  $H_0$

$$\rightarrow P(\bar{X} > 68.442 | \bar{U} = 66.7) > 0.05$$

unknown  $\sigma^2$

t-distribution and T-statistic

Degrees of freedom (V) =

s =

Sample size - 1

$$\sqrt{\frac{1}{10-1} \sum_{i=1}^{10} (x_i - \bar{x})^2} \quad T = \frac{\bar{x} - u}{\frac{s}{\sqrt{10}}} \sim t_9$$

V = 10-1

Remaining notes :-

Marginal distribution :

height 1

	45	46	47	48	49	50	
7	$\frac{3}{10}$	$\frac{1}{10}$	0	0	0	0	
8	0	0	$\frac{3}{10}$	0	0	0	$\frac{3}{10}$
Age 9	0	0	0	0	$\frac{3}{10}$	$\frac{1}{10}$	$\frac{2}{10}$
10	0	0	0	0	0	$\frac{1}{10}$	$\frac{4}{10}$
	$\frac{1}{10}$	$\frac{2}{10}$	$\frac{2}{10}$	0	$\frac{3}{10}$	$\frac{2}{10}$	$\frac{1}{10}$

Marginal distribution of age

Age	7	$\frac{3}{10}$
(X)	8	$\frac{2}{10}$
	9	$\frac{4}{10}$
	10	$\frac{1}{10}$

Marginal distribution of height

Height (y)	45	46	47	48	49
	$\frac{1}{10}$	$\frac{2}{10}$	$\frac{2}{10}$	0	$\frac{3}{10}$

$$P(y=49 | x=9) = \frac{3}{4}$$

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conditional PDF

$$P(A, B) = P(A) \cdot P(B|A)$$

$$\begin{aligned} P(Y=49|X=9) &= \frac{P(X=9, Y=49)}{P(X=9)} \\ &= \frac{\frac{3}{10}}{\frac{9}{10}} = \frac{3}{9} \end{aligned}$$

Covariance of a joint matrix with 3 variables

$$\begin{matrix} & x & y & z \\ x & \text{Var}(x) & \text{cov}(x,y) & \text{cov}(x,z) \\ y & \text{cov}(x,y) & \text{Var}(y) & \text{cov}(y,z) \\ z & \text{cov}(x,z) & \text{cov}(y,z) & \text{Var}(z) \end{matrix}$$

$$\begin{aligned} \text{Likelihood} &= \prod_{i=1}^n P_{U_i}(u_i) = \prod_{i=1}^n p^{u_i} (1-p)^{1-u_i} \\ &= p^{\sum_{i=1}^n u_i} (1-p)^{n - \sum_{i=1}^n u_i} \end{aligned}$$

$$\text{Log-likelihood} = \log(\text{Likelihood}) = \dots$$

$$\left( \sum_{i=1}^n u_i \right) \log(p) + \left( n - \sum_{i=1}^n u_i \right) \log(1-p)$$

## Independent Two sample t-test

Height of 18 y/o in the US

$$n_u = 10 \quad \bar{y} = 68.442 \\ S_x = 3.113$$

Height of 18 y/o in Argentina

$$n_u = 9 \quad \bar{y} = 65.949 \\ S_x = 3.106$$

$$H_0: \mu_{us} = \mu_{ar} \quad v/s \quad H_1: \mu_{us} > \mu_{ar}$$

$$\text{v/s } H_1: \mu_{us} < \mu_{ar}$$

$$\text{v/s } H_1: \mu_{us} \neq \mu_{ar}$$

$$x \quad (\bar{x} - \bar{y}) - (\mu_{us} - \mu_{ar}) \sim N(0, 1) \\ \sqrt{\frac{\sigma^2_{us}}{10} + \frac{\sigma^2_{ar}}{9}}$$

we don't know  $\sigma_{us}$ ,  $\sigma_{ar}$ , replace  
 $\sigma_{us}$ ,  $\sigma_{ar}$  with  $S_x^2$ ,  $S_y^2$ .

right tailed test

$$H_0: \mu_{us} - \mu_{ar} = 0 \quad v/s \quad H_1: \mu_{us} - \mu_{ar} > 0$$

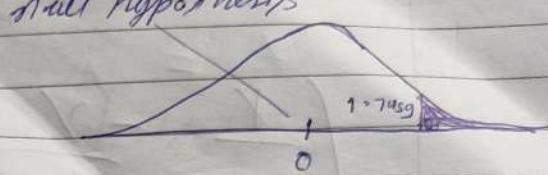
$$\alpha = 0.05$$

$$t = \frac{(\bar{x} - \bar{y}) - 0}{\sqrt{\frac{S_x^2}{10} + \frac{S_y^2}{9}}} = 1.7459$$

p-value  $0.0495 < 0.05$

reject  $H_0$

null hypothesis



paired t-test

$$\frac{(x_1 - y_1) + (x_2 - y_2) + \dots + (x_{10} - y_{10})}{10}$$

$$\bar{d} = \frac{d_1 + d_2 + d_3 + d_4 + d_5 + d_6 + d_7 + d_8 + d_9 + d_{10}}{10}$$

$$\frac{\bar{d} - M_d}{\sigma_d / \sqrt{10}} \sim N(0, 1)$$

$\sigma_d$  is ~~unk~~ unknown

$$\sigma_d \Rightarrow S_d \Rightarrow \sqrt{\frac{\sum_{i=1}^{10} (d_i - \bar{d})^2}{10-1}} \Rightarrow$$

$$T = \frac{\bar{d} - M_d}{S_d / \sqrt{10}} \sim t_{10-1}$$

A/B Testing : Conversion Rate

If  $H_0$  is true  $\Rightarrow P_A = P_B = P$

$$\left( \frac{X}{n_A} - \frac{Y}{n_B} \right) - (P - P) \sim N(0, 1)$$

$$\sqrt{\frac{P(1-P)}{n_A}} + \frac{P(1-P)}{n_B}$$



$$P(1-P) \left( \frac{1}{n_A} + \frac{1}{n_B} \right)$$

$$= \frac{P(1-P)(n_A + n_B)}{n_A n_B} + 1$$

$$\left( \frac{X}{n_A} - \frac{Y}{n_B} \right) - 0$$

$$\sqrt{\frac{(n_A + n_B)P(1-P)}{n_A n_B}} \xrightarrow{n_A, n_B \rightarrow \infty} N(0, 1)$$

You don't know  $P$

Replace it by estimation  $\hat{P} = \frac{X+Y}{n_A + n_B}$

### Test statistic

$$\frac{\left| \frac{n}{n_A} - \frac{y}{n_B} \right| - 0}{\sqrt{\frac{(n+y)(1-\frac{x+y}{n_A+n_B})}{n_A n_B}}} \sim N(0,1)$$

$H_0: p_A - p_B = 0$  v/s  $H_1: p_A - p_B < 0$   
 ( $p_A$  has high conversion rate)

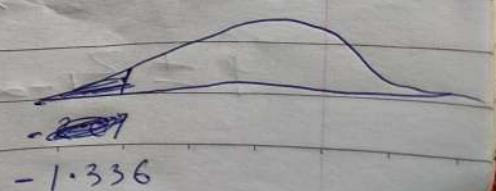
$$n_A = 80 \quad n_B = 20 \quad \alpha = 0.05 \\ n = 20 \quad y = 8$$

$$Z = \frac{\left| \frac{20}{80} - \frac{8}{20} \right| - 0}{\sqrt{\frac{(20+8)(1-20+8)}{80+20}}} \sim N(0,1)$$

$$Z = -1.336$$

$$P\text{-value} = 0.091$$

Don't reject  $H_0$



### Assignment - 3

t-test for 2 independent sample

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - (M_1 - M_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{\left( \frac{s_1^2}{n_1} + \frac{s_2^2}{n_2} \right)^2}{\frac{\left( \frac{s_1^2}{n_1} \right)^2}{n_1 - 1} + \frac{\left( \frac{s_2^2}{n_2} \right)^2}{n_2 - 1}}$$

z test for 2 independent sample

$$z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p}) \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$\hat{p} = \frac{n_1 + n_2}{n_1 + n_2}$$

~~$\hat{p} = \frac{1}{2}$~~

$$\hat{p} \pm z_{1-\alpha/2} \sqrt{\hat{p}(1-\hat{p})}$$