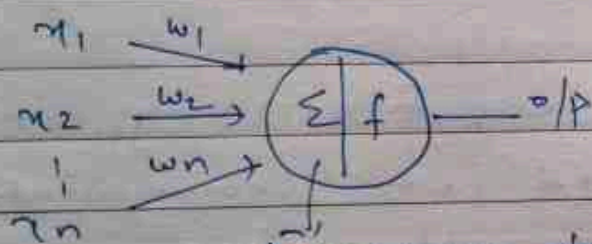


McCulloch - Pitts Neuron.

- First mathematical model of biological neuron in 1943 by Warren McCulloch & Walter Pitts
- Also called Linear-threshold Gate model
- Basic building block of neural n/w
- Directed weight graph is used for connected neuron.
- Two possible state of neuron
 - Active (1)
 - Silent (0)



Aggregates the weighted i/p into.

Single numeric value

$$\sum (x_1 w_1 + x_2 w_2 + x_3 w_3 + \dots + x_n w_n) \xrightarrow{\text{denoted}} X$$

f → produced the o/p using threshold (T) value.

$$o/p = 1 \text{ if } X > T$$

$$0 \text{ if } X < T$$

Bias / Threshold → It is the minimum value of weighted active i/p of neuron to fire.

If effective/weighted i/p $> T$ then o/p is 1 else o/p is 0.

i.e.

$$o/p : f(x)$$

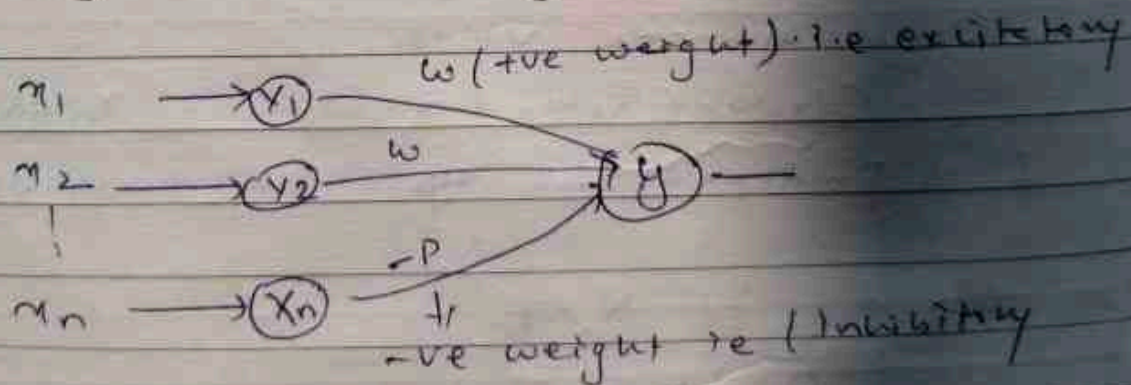
$$\hookrightarrow \sum w_i x_i \geq T$$

$$\text{function } f(x) = \begin{cases} 1 & \text{if } x > 0 \\ 0 & \text{else} \end{cases}$$

what this model do

There is weighted i/p & o/p function compared it with weighted i/p. If it is greater than threshold then o/p is 1 else the o/p is zero.

Binary - neuron may fire (1) may not fire (0)
weight associated with communication link may be excitatory (positive) or inhibitory (-ve)



$$y_{in} = x_1 w_1 + x_2 w_2 + x_n w_n$$

$$y = f(y_{in}) = \begin{cases} 1 & \text{if } y_{in} \geq \theta \text{ threshold } (0) \\ 0 & \text{if } y_{in} < \theta \end{cases}$$

To find threshold value
for inhibitory, threshold will be the activation
function should satisfy the following function

$$\theta > n\omega - p$$

↓
Inhibitory

for no particular training program.

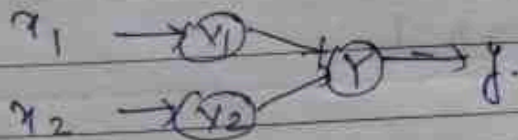
Ex of using Mc-culloch Pitts ne

Q. Implement AND NOT function using Mc-Culloch-Pitts neuron (Binary data)

x_1	x_2	y
0	0	0
0	1	0
1	0	1
1	1	0

o/p = 1 if 1st i/p is 1 & 2nd i/p is 0

Soln



Case-1

$$\omega_1 = \omega_2 = 1$$

$$\theta \geq n\omega - p$$

$$> 2(1) - 0$$

$$\theta \geq 2$$

So $y_{in} = x_1 w_1 + x_2 w_2$

for i/p

$w_1 = w_2 = 1$

$(0, 0) \quad y_{in} = 0 \times 1 + 0 \times 1 = 0$

$(0, 1) \quad y_{in} = 0 \times 1 + 1 \times 1 = 1$

$(1, 0) \quad y_{in} = 1 \times 1 + 0 \times 1 = 1$

$(1, 1) \quad y_{in} = 1 \times 1 + 1 \times 1 = 2$

This is not the i/p for AND-NOR function

Case-2 one weight is excitatory & other is inhibitory.
i.e. $w_1 = 1 \quad w_2 = -1$

for i/p

$(0, 0) \quad y_{in} = 0 \times 1 + 0 \times -1 = 0$

$(0, 1) \quad y_{in} = 0 \times 1 + 1 \times -1 = -1$

$(1, 0) \quad y_{in} = 1 \times 1 + 0 \times -1 = 1$

$(1, 1) \quad y_{in} = 1 \times 1 + 1 \times -1 = 0$

It is possible to fire the neuron with i/p (1, 0) only
for $w_1 = 1 \quad w_2 = -1$

$\theta \geq n w - p$

$\geq 2(1) - 1$

$\theta \geq 1$

The o/p of Neuron γ can be written as

o/p $y = f(y_{in}) = \begin{cases} 1 & \text{if } y_{in} \geq 1 \\ 0 & \text{if } y_{in} < 1 \end{cases}$



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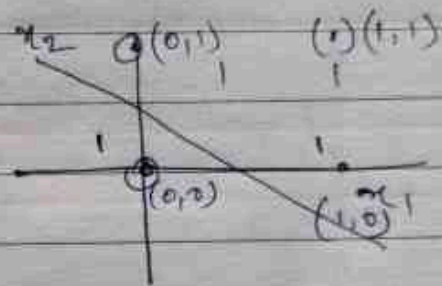
Linear Separability

- A decision line is drawn to separate positive and negative response
- When two class that can be separated by decision boundary they are said to be Linear Separable problem.

Eg.

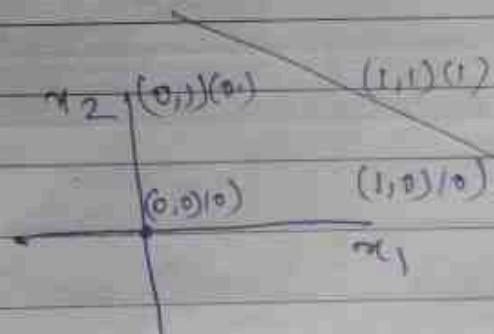
OR gate

x_1	x_2	OR
0	0	0
0	1	1
1	0	1
1	1	1



AND gate

x_1	x_2	o/p
0	0	0
0	1	0
1	0	0
1	1	1



Hebb network

Algorithm

- 1) Initialize all weight & bias to zero
 $w_i = 0$ for $i = 1$ to n
- 2) for each i/p training vector & target vector o/p pair (s, t) perform step 3 to 6
- 3) Set activation for i/p unit i/p vector
 $x_i = s_i$ ($i = 1$ to n)
i/p unit x i/p vector.
- 4) Set Activation for o/p unit with o/p neuron
 $y = t$
- 5) Adjust the weight by applying hebb rule.
 $w_i(\text{new}) = w_i(\text{old}) + x_i y$ to $i = 1$ to n .
- 6) adjust bias.
 $b(\text{new}) = b(\text{old})$

Eg. Hebb n/w to implement AND using hebb rule

Training data for AND fun.

x_1	x_2	bias	Target y
1	1	1	1
1	-1	1	-1
-1	1	1	-1
-1	-1	1	-1

Teacher's Signature

Initially the weight & bias to zero

$$w_1 = w_2 = b = 0$$

First input $[x_1, x_2, b] = [1, 1, 0]$ and target = 1
Setting the initial weight as old weight
and applying the Hebb rule, we get

$$w_i(\text{new}) = w_i(\text{old}) + \Delta w_i$$

Change of weight

first find Δw_i

$$\Delta w_i = x_i y$$

i/p \Rightarrow target

$$\Delta w_1 = x_1 y = 1 \times 1 = 1$$

$$\Delta w_2 = x_2 y = 1 \times 1 = 1$$

$$\Delta b = y = 1$$

$$w_1(\text{new}) = w_1(\text{old}) + \Delta w_1 = 0 + 1 = 1$$

$$w_2(\text{new}) = w_2(\text{old}) + \Delta w_2 = 0 + 1 = 1$$

$$b(\text{new}) = b(\text{old}) + \Delta b = 0 + 1 = 1$$

Second i/p.

$$[x_1, x_2, b] = [1, -1, 1] \text{ and } y = -1$$

The weight change has

$$\Delta w_1 = x_1 y = 1 \times -1 = -1$$

$$\Delta w_2 = x_2 y = -1 \times -1 = 1$$

$$\Delta b = y = -1$$

$$w_1(\text{new}) = w_1(\text{old}) + \Delta w_1 = 1 - 1 = 0$$

$$w_2(\text{new}) = w_2(\text{old}) + \Delta w_2 = 1 + 1 = 2$$

$$b(\text{new}) = b(\text{old}) + \Delta b = 1 - 1 = 0$$

third i/p

$$[-1, 1, 1] \text{ and } y = -1$$

$$\Delta w_1 = x_1 y = -1 \times 1 = -1$$

$$\Delta w_2 = x_2 y = 1 \times -1 = -1$$

$$\Delta b = y = -1$$

$$w_1(\text{new}) = w_1(\text{old}) + \Delta w_1 = 0 - 1 = -1$$

$$w_2(\text{new}) = w_2(\text{old}) + \Delta w_2 = 2 - 1 = 1$$

$$b(\text{new}) = b(\text{old}) + \Delta b = 0 - 1 = -1$$

fourth i/p

$$[-1, -1, 1]$$

$$y = -1$$

$$\Delta w_1(\text{new}) = w_1(\text{old}) + \Delta w_1 = 0 + 1 = 1$$

$$w_2(\text{new}) = w_2(\text{old}) + \Delta w_2 = 1 + 1 = 2$$

$$b(\text{new}) = b(\text{old}) + \Delta b = -1 - 1 = -2$$

$$\Delta w_1 = x_1 y = -1 \times -1 = 1$$

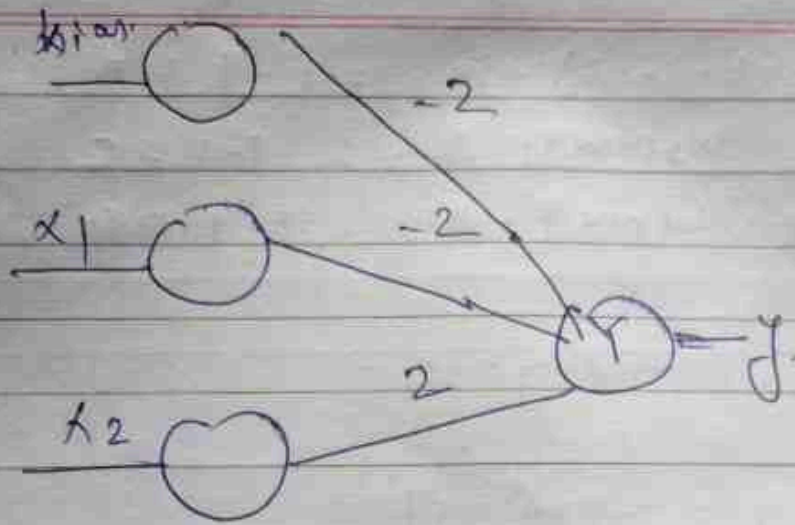
$$\Delta w_2 = x_2 y = -1 \times -1 = 1$$

$$\Delta b = y = -1$$

New weight calculated are:

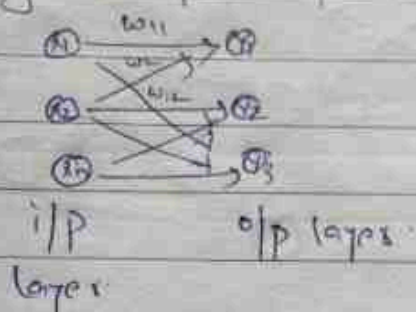
x_1	x_2	b	y	Δw_1	Δw_2	Δb	w_1	w_2	b
1	1	1	1	1	1	1	1	1	0
1	-1	1	-1	-1	1	-1	0	2	0
-1	1	1	-1	1	-1	-1	1	1	-1
-1	-1	1	-1	1	-1	-1	2	2	-2

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Models of artificial neural n/w

1) Single layer feed forward

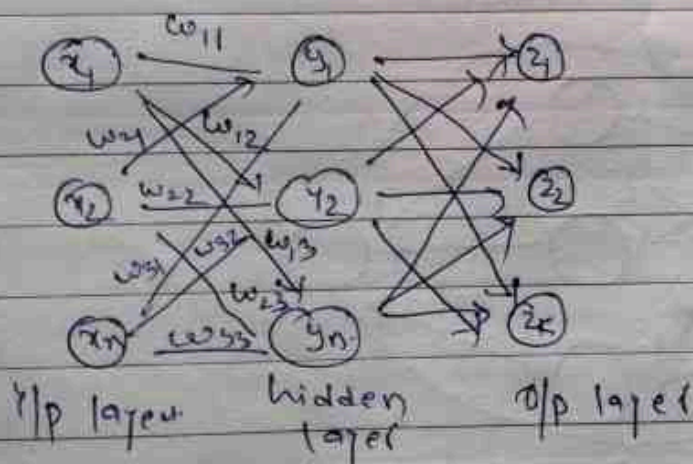


Two layers only
i/p layer & o/p layers

2) Multi layer feed forward n/w

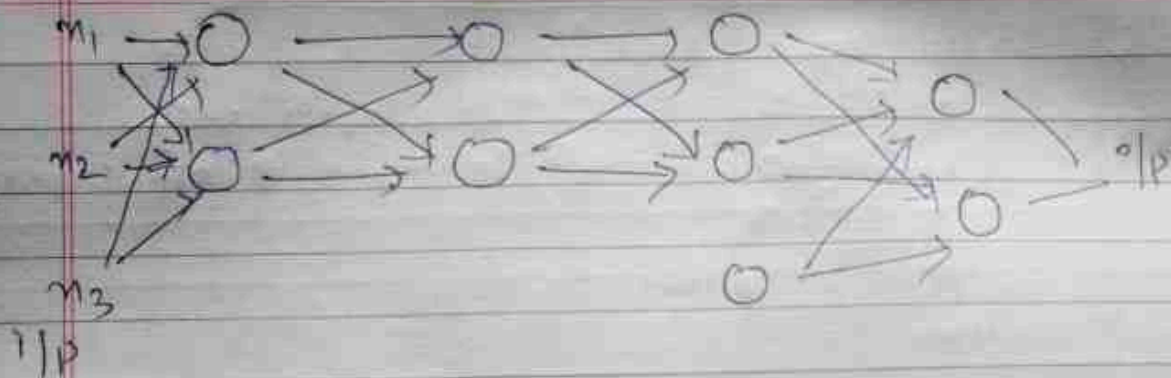
There is hidden layer.

→ It become Computationally more strong.



3) Multilayer Perceptron

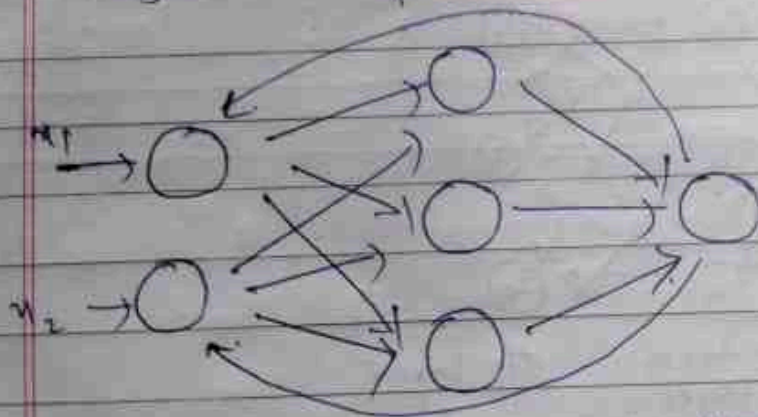
3 or more layers are used to classify non-linearly separable data.



It is fully connected n/w that uses non linear activation functions

④ feedback ANN

feedback is given to i/p layers to adjust the parameters



A feedback is available toward the first layer so that error according to error parameters can be adjusted
It is used to minimise the error.

Perception network



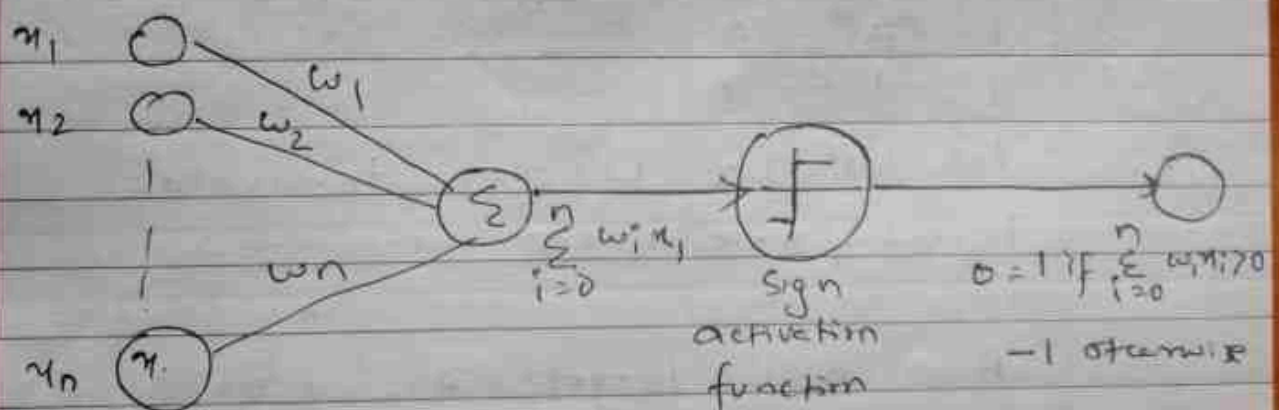
- Perception unit is used to build ANN system
- Perception takes a vector of real-valued i/p, calculates a linear combination of these i/p, then o/p 0 or 1 if the result is greater than some threshold and -1 otherwise Mathematically.

if given i/p and $x_1, x_2, x_3, \dots, x_n$
o/p and $0, 1, 0, 3, \dots, 0_n$

then computed by perception is

$$o(x_1, \dots, x_n) = \begin{cases} 1 & \text{if } w_0 + w_1x_1 + w_2x_2 + \dots + w_nx_n \geq 0 \\ -1 & \text{otherwise} \end{cases}$$

w_i is real valued constant or weight that determines the contribution of i/p x_i to the perception o/p



The actual o/p is compared with target o/p if both target & actual o/p are same, these are final wt. otherwise go back and

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modify the weights (w_1, w_2 & w_3). Again the same process continue i.e. first calculate the summation term, apply activation function and calculate the actual o/p. If the actual o/p is same as target these are final wt. otherwise go back and modify the wt.

How to learn the wt.

Ex - logical AND gate using perceptron net

Truth Table

A	B	$A \wedge B$
0	0	0
0	1	0
1	0	0
1	1	1

Let $w_1 = 1.2$ $w_2 = 0.6$ threshold = 1

learning rate $\eta = 0.5$

$A = 0$ $B = 0$ target = 0

$$w, x_1 = 0 \times 1.2 + 0 \times 0.6 = 0$$

not greater than threshold of 1 so o/p = 0.

The actual o/p calculate is zero and target is 0, because both are same. No need of wt. updation

2. $A=0$ $B=1$ target $= 0$

$$w, \pi_1 = 0 \times 1.2 + 0.6 \times 1 = 0.6$$

not greater than threshold of 1, So $o/p = 0$

target = actual o/p , So no need of wt. modification

3) $A=1$ $B=0$ target $= 0$

$$w, \pi_1 = 1 \times 1.2 + 0.6 \times 0 = 1.2$$

o/p is greater than threshold of 1 So $o/p = 1$
target \neq actual o/p , So wt need to be modified.

A/c to perceptron rule

$$w_i = w_i + \eta(t - o) \pi_i$$

$$w_1 = 1.2 + 0.5(0 - 1)1 = 0.7$$

$$w_2 = 0.6 + 0.5(0 - 1)0 = 0.6$$

new modified w

Check from beginning:

$A=0$ $B=0$ target $= 0$

$$w, \pi_1 = 0.7 \times 0 + 0.6 \times 0 = 0$$

no need of wt. updation

$A=0$ $B=1$ target $= 0$

$$w, \pi_1 = 0 \times 0.7 + 1 \times 0.6 = 0.6$$

no need of wt. updation

3) $A=1$, $B=0$ target $= 0$

$$w, \pi_1 = 1 \times 0.7 + 0 \times 0.6 = 0.7$$

no need wt. updation

4) $A=1$ $B=1$ target $= 1$

$$w, \pi_1 = 0.7 \times 1 + 0.6 \times 1 = 1.3$$

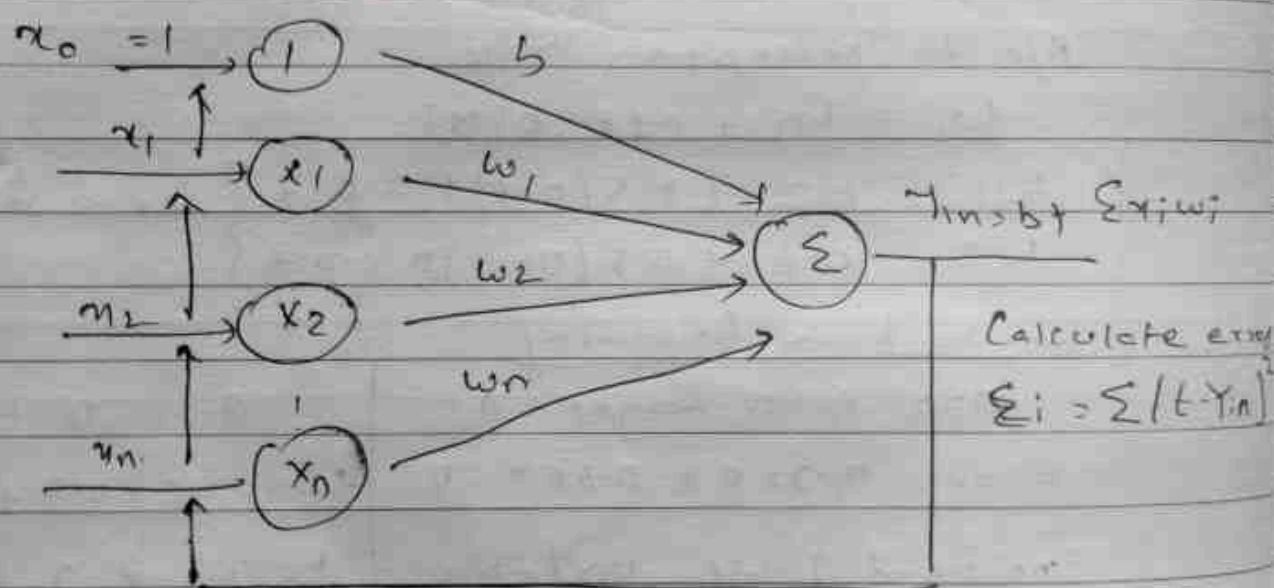
$w.1 >$ threshold (no wt. updation)

Adaptive Linear neuron (Adaline)

Date _____
Page _____

The units with linear activation function are called linear units.

- A n/w with a single linear unit is called adaptive linear neuron (Adaline)
- In adaptive linear neuron the i/p & o/p relationship is linear
- Adaline uses bipolar activation function for its i/p signal and its target o/p
 $m_0 = 1$ bias

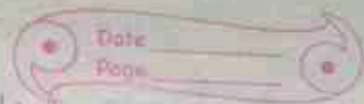


Weight updation

$$w_i(\text{new}) = w_i(\text{old}) + \eta (t - y_{in}) x_i$$

$$b(\text{new}) = b(\text{old}) + \eta (t - y_{in})$$

Unsupervised learning n/w



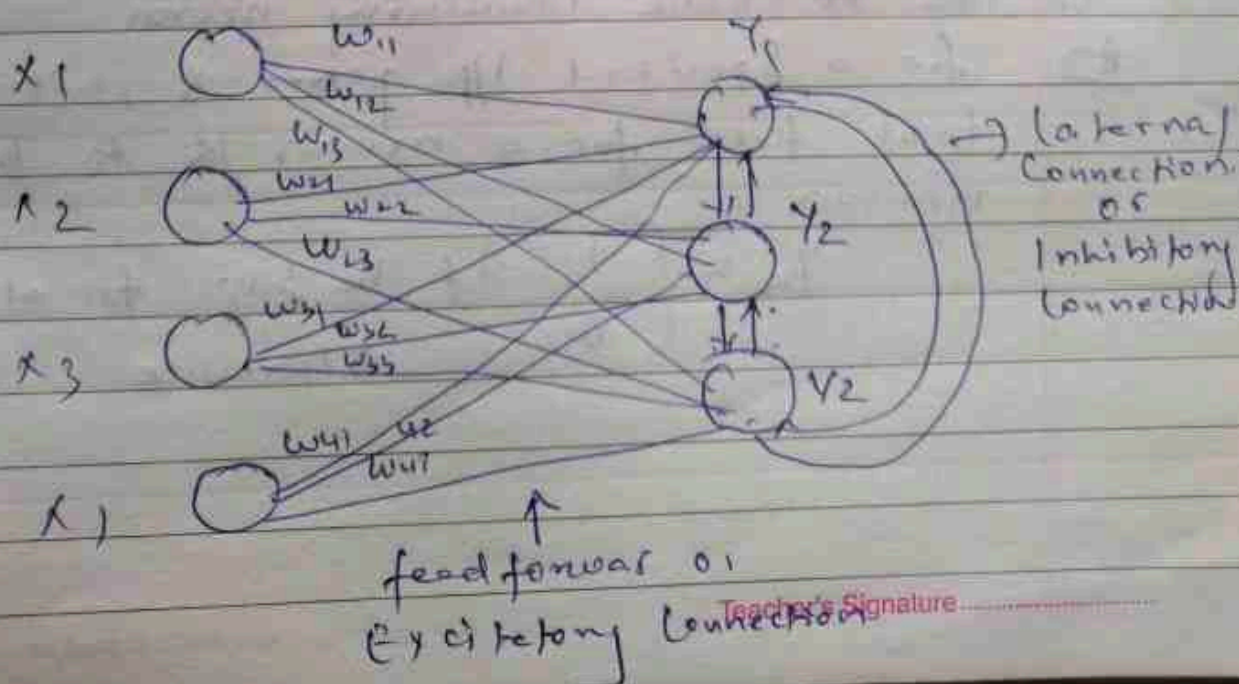
1) Competitive learning n/w

→ The o/p neurons of a neural n/w compete among themselves to become active (fire).
- In Hebbian learning, several o/p neurons may be active but in competitive only single o/p neuron is active at any one time.

→ This feature helps to classify a set of i/p pattern.

Three basic elements of competitive learning

- 1) a set of neuron that are all to serve. (except for synaptic weights)
- 2) Unit imposed on the strength of each neuron.
- 3) a mechanism that permits the neuron to compete → a winner takes all



Analogy:

All student read all text book, video, lecture
but only one will be the topper.

Among Y_1, Y_2, Y_3 only one o/p will win
topper always wants to be topper, so he
inhibit other o/p student by not saying secret.

Elements

1) all i/p neuron x_1, x_2, x_3 & x_4 are same
but their weights are different

2) Limit imposed on each neuron.

Constraint can be:

$$\left. \begin{aligned} \sum_j w_{kj} &= 1 \quad \text{for all } k \\ \sum_j w_{kj}^2 &= 1 \quad \text{for all } k \end{aligned} \right\} \text{these are constraints.}$$

3rd elements

Q How to choose winning neuron

Ans for a specified i/p pattern x , with induced
local field for a neuron k to be winning
neuron

$$y_k = \begin{cases} 1 & \text{if } v_k > v_j \text{ for all } j, j \neq k \\ 0 & \text{otherwise.} \end{cases}$$

membership - function

$$A = \{x, \mu(x)\}$$

Let

$$\mu(x) = \frac{1}{1+x^2}$$

↓

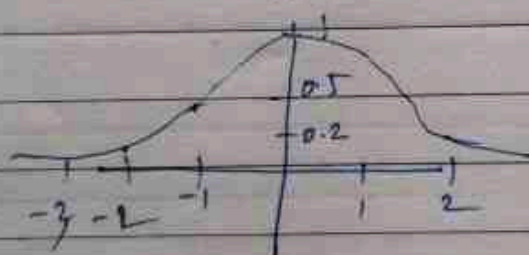
membership function.

eg.

$$A = \{ \text{Real no. close to zero} \}$$

$$= \{x, \mu(x)\} \quad x \in \mathbb{R}$$

$$\mu(x) = \frac{1}{1+x^2}$$



degree of membership

$$\mu(-1) = 0.5$$

$$\mu(1) = 0.5$$

$$\mu(-2) = 0.2$$

$$\mu(2) = 0.2$$

1) Support Set

2) height of fuzzy set

3) Normalised & non-normalised fuzzy set

$$A = \{(x_1, \mu(x_1)), (x_2, \mu(x_2)), (x_3, \mu(x_3)) + \dots\}$$

Another way to represent

$$A = \left\{ \frac{\mu(x_1)}{x_1} + \frac{\mu(x_2)}{x_2} + \frac{\mu(x_3)}{x_3} + \dots \right\}$$

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$$= \sum_{i=1}^n \frac{\mu(x_i)}{n_i} \quad \text{for Discrete data.}$$

$$\int \frac{\mu(x)}{x} \quad \text{for Continuous data}$$

Support^{Set} of fuzzy set:

$$\vec{A} = \{ (20, 0.0), (40, 0.0), (60, 0.8), (80, 0.9) \}$$

$$S(\vec{A}) = \{ x | \mu(x) > 0 \}$$

Accordingly Support set is all set for which the value are greater than 0
i.e. all element for which $\mu(x) > 0$

$$S(\vec{A}) = \{ x | \mu(x) > 0 \}$$

$$\{ 40, 60, 80 \}$$

Height of fuzzy set:-

$$\mu_{\max}$$

$$H(\vec{A}) = 0.9$$

Max. value of $\mu(x)$ is ht. of fuzzy set

Normalised set:

$$\text{If } \mu_{\max} = 1$$

i.e. If height of $\mu(x) = 1$ then normalised.

Q4

Standard operation on fuzzy set

- i) Union
- 2) Complement
- 3) Intersection
- iv) α cut set or α level set
- ii) Equivalent fuzzy set.
- iii) Sub-set of fuzzy set
- iv) Standard operation on fuzzy set

α -cut set or α level set

If \bar{A} then α -cut set of \bar{A} is $\alpha \bar{A}$

$$\bar{A}_\alpha = \{x \in A \mid \mu(x) \geq \alpha\}$$

eg

$$\bar{A} = \{(10, 0.5) \mid (20, 0.6) \mid (30, 0.7) \mid (40, 0.8)\}$$

then

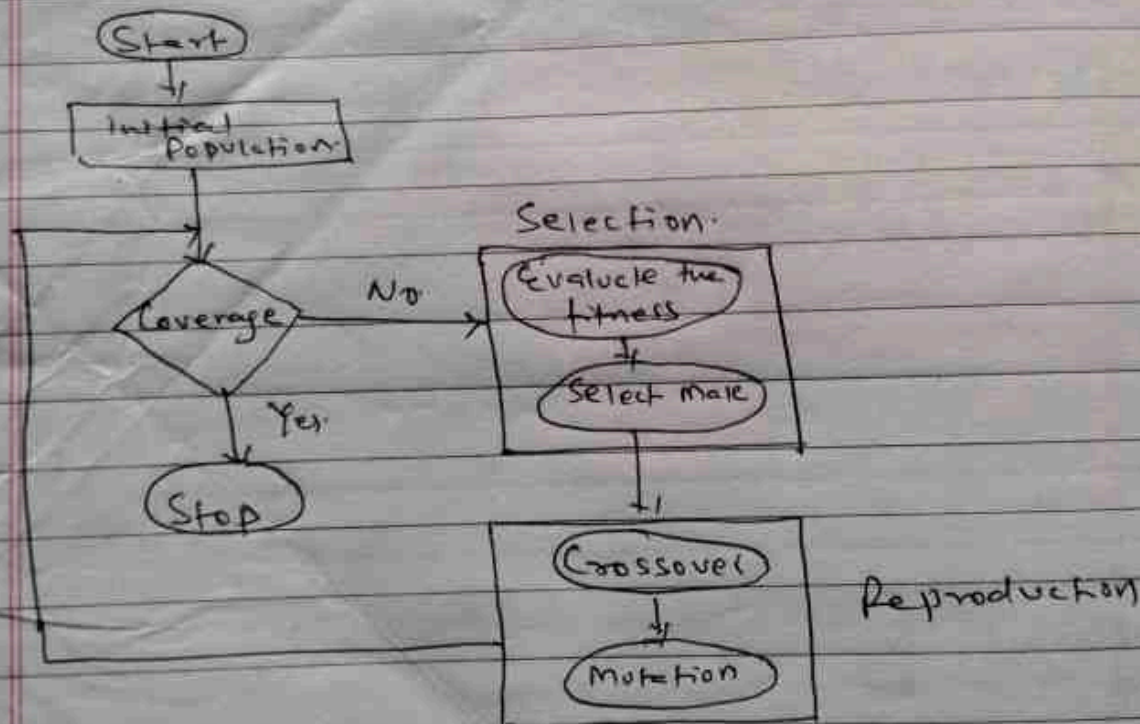
$$\bar{A}_{0.6} = \{20, 30, 40\}$$

we write value which 0.6 value greater than 0.6

equivalent fuzzy set

GA is adaptive-heuristic Search Algorithm
It is evolutionary Algorithms
Genetic & natural selection.
To Generate high-Quality Solution for optimization
Problem

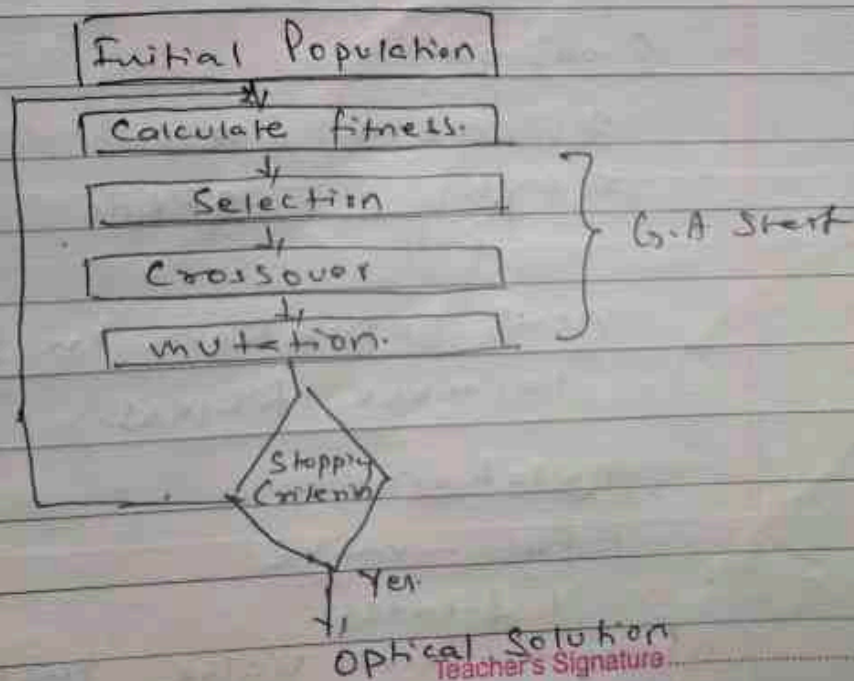
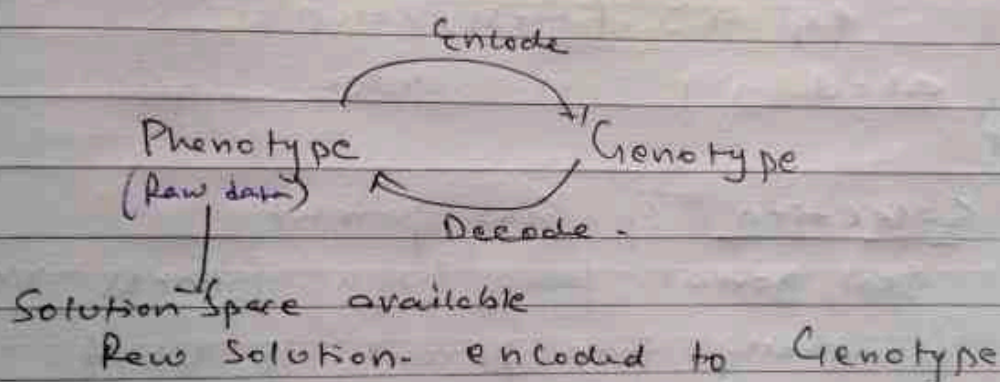
GA



Genetic Algorithm

John Holland gave idea of G.A

- It is abstraction of a biological evolution
- Solve Complex Problem (like NP)
- Focus on optimization
- Application of possible solution for a given problem
- From a group of individual, the best will survive



Initial population → Sort already available
Don't focus on one or two point.

Eg. Exit Poll → Consider all population

eg young, old, baby

Calculate fitness

eg Make we have to form from
word take. So according to fitness
how many alphabet take is fit to
form make

eg → 3 fitness value

abcd → 1 1 1 value

Selection of best parent.

Best parent have high fitness value.

Generally two parent is used

Cross-over

a b c d e

e f g h i

swap

a b o h i

e f c d e

Swap policy.

new population is generated to
increase fitness.

Mutation

e f g h — make

1 fitness

Generate such value that have fitness
value better than previous.

one fitness value highest is achieved.
then stopping criterion is reduced.

eg

9 km e > both have highest fitness
1 km e value.

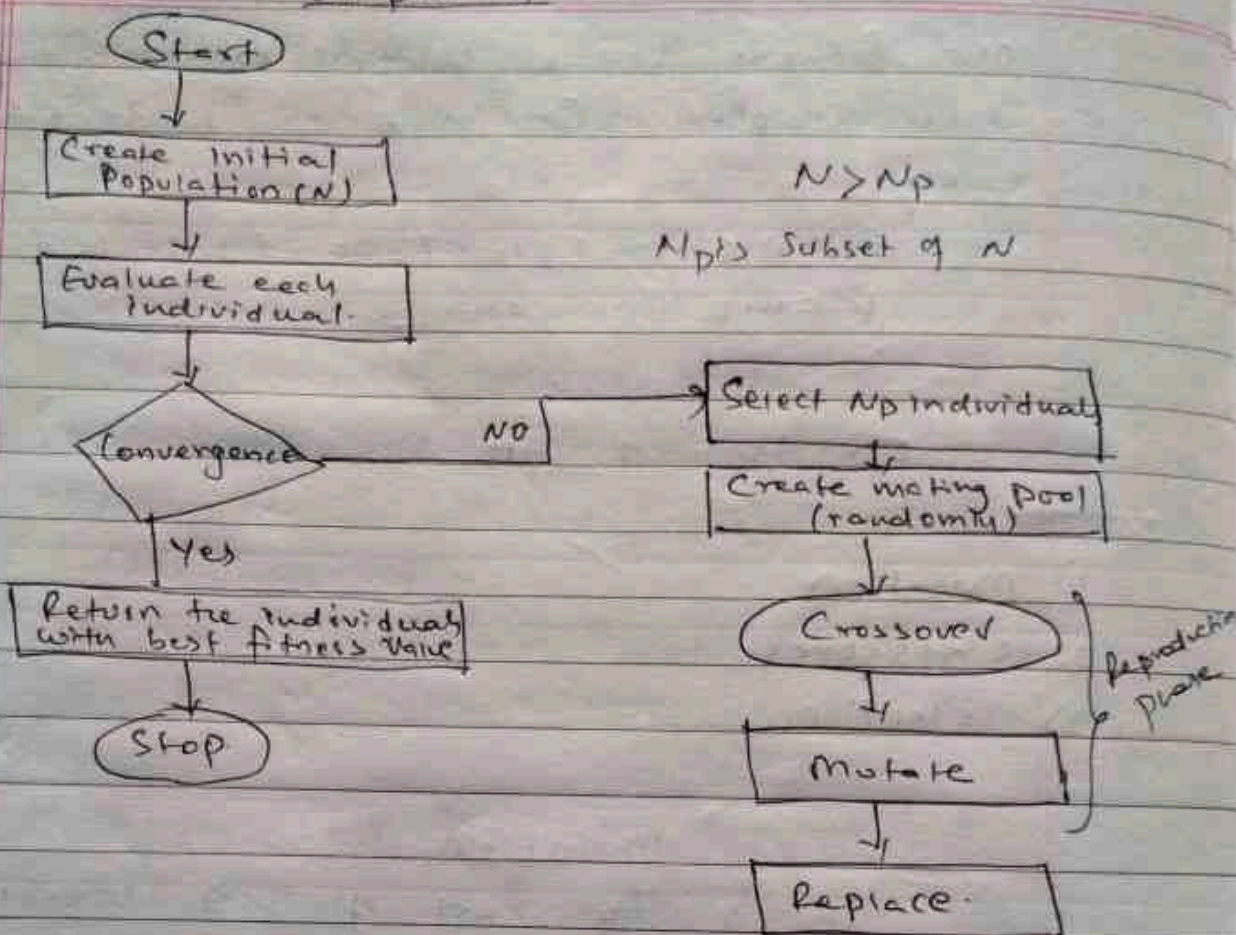
operators of GA

- 1) Mutation
- 2) Cross-over
- 3) Selection.

Mutation

It is the part of GA which is related to the 'exploration' of the search space. It has been observed that mutation is essential to the convergence of GA while crossover is not.

Simple GA



GA Vs Traditional Algorithms

GA	TA
1) Genetics & natural Selection to solve optimization Problem	Step by step procedure to solve a given problem
2) More advance	→ Not as advance
3) Use the field such as ML, AI	→ used in field such as programming, mathematics
4) Probabilistic Rules	→ Fully deterministic rules
5) Search on a population of points	→ Search on a single point

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Convergence Criterion / Termination Condition Convergence Test

1) Manual Checking

It is tedious, time consuming
Manual checked

2) Solution found that satisfy objective Criterion eg. if threshold value is satisfied.

It is satisfy the Criterion. It is not
not optimized value. It is applicable when
certain Criterion is achieved

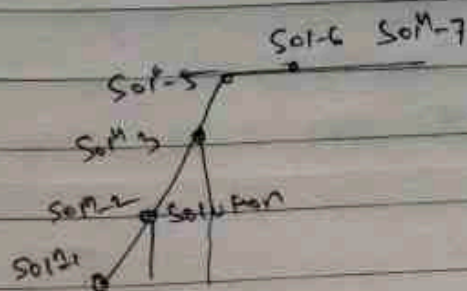
3) Fixed no. of generation

eg. 10 iteration will be perform. no. of
iteration is increase, we are reaching
toward global optimization, so we get go
set no iteration

4) Budget limit reached

base on time available or available
space Criterion is fixed. So if computation
time is one hour then optimization is fixed
for one hour.

5)



Solⁿ 2 is more fitter than Solⁿ 3

Solⁿ 3 is less fit than Solⁿ 2

after Solⁿ 5, Solⁿ 6, Solⁿ 7 is
equal. We have reached
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6) Combination of above said Solⁿ

better Solⁿ

Encoding:

Binary Encoding:

1st step Creating initial population. So Creating initial population Encoding is performed.

Binary Encoding

Individual is represented by Chromosome. It contain Genes. Material in form of Genes. So Chromosome contain Genes. Various Genes in one Chromosome.

Binary Encoding

Q. Ex.

$$f(x) = x^3 ; 0 \leq x \leq 255$$

- 8-bit

1 0 0 1 1 0 1 1 1 1 1 1 1 1 1 1

8bit

1st

255

digit

Chromosome

↓

→ It can be a Solution

It can also be a Solⁿ

Assigning Chromosome in form of String of bits is Encoding.

Population contain Collection of possible Combination of 8-bits

Ex-2

Binary Encoding

include one item or not called knapsack problem
 include - 1
 not include - 0

(A)	-7	= 10	= 5	} Profit	= 100 cost
(B)		Answer	= 10		= 150
(C)			= 15		= 200

M=25 max is 25 called constraint

Choose one

(A)	(AB)	(AC)	(BC)
100	110	101	011

A = 100 Choose.

↓
 item-1 include → item-2 included } item-3 included

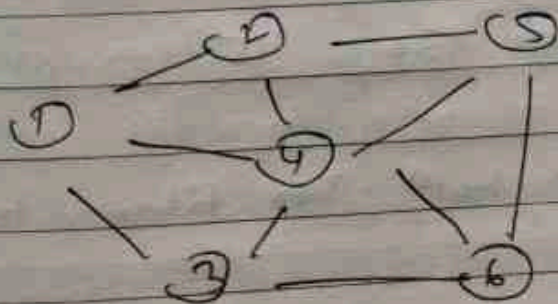
AB = 110

↓
 item-1 include → item-2 included

Fitness Evaluation

fitness function or objective function

TSP (Travelling Salesman Problem)



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Ex-1

Let Fitness value \propto

- Path-1 - 1 2 5 6 4 3 1 - (18)
 2 - 1 2 5 4 6 3 1 - (20)
 3 - 1 2 4 5 6 3 1 - (12)
 4 - 1 2 5 6 3 4 1 - (17)

Ex-2

Objective function \rightarrow Shortest path

Constraint \rightarrow one visit to every city is

Constraint

fitness function is one which has highest objective accomplishment

Path-3 is fittest if meet the objective

Ex-2 01 napsad

- A = 5 100
 B = 10 150
 C = 15 200

profit

M = 25 (constraint)

Profit
P/100

A = 1 0 0	=	W/5	
AB = 1 1 0	=	15	250
AC = 1 0 1	=	20	300
BC = 0 1 1	=	25	350
ABC = 1 1 1	=	30	450 \rightarrow not a soln

BC is the best soln. Its fitness test is highest

Crossover (Binary Code)

P ₁	0	1	1	0	1	0
P ₂	1	1	0	1	0	0

P₁ & P₂ are Parent 1 & parent-2

Single Point Crossover

0	1	1	0	1	1
1	1	0	1	0	0

K-Single point

P₁ 1st part P₂ 2nd part

P₂ 1st part P₁ 2nd part

O ₁	0	1	0	1	0	0
O ₂	1	1	1	0	1	0

→ Aug Single point Crossover

O₁ & O₂ are offspring
Reproduction phase

Two point Crossover

0	1	1	0	1	0
1	1	0	1	0	0

1st part
K₁-point
2nd part
K₂-point
3rd part

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1st & 3rd part are extreme part. Write it as it is

is

01	0	1	0	1	1	0
02	1	1	1	0	0	0

Swipe 2nd part

multipoint Crossover

0	1	1	0	1	0
1	1	0	1	0	0

\downarrow \downarrow \downarrow \downarrow
 swipe k_3 k_4 swipe k_2

Alternat part Swape with each other.

1	1	0	1	1	0
0	1	0	0	0	0

Uniform Crossover

0	0	1	0	1	1
1	0	0	1	0	0

Does every bit is to be Swiped. So tossing is performed. - on basis result swapping is decided. If it comes 1 then Swipe and if comes 0 then don't Swipe.

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Tossing result



To S

0	0	1	0	1	1
1	0	0	1	0	0

1	0	0	1	1	0
1	0	1	1	0	1
0	0	0	0	1	0

→ Toss Result

Cross-over mask

0	0	1	0	1	1
1	0	0	1	0	0

→ Parent Chromosome

C-m

0 1 0 1 1 0

Cross-over mask is randomly generated.

0 ₁	0	0	1	1	1	1
0 ₂	1	0	0	0	1	0

Rules

If bit is 0 the parent bit-1 is copied to

offspring

offspring

If C-m is 0 the parent bit-2 is copied to

offspring.

For offspring-2

If 0 the Parent bit-2 is copied & if 1 comes the Parent bit 1 is copied

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Mutation In GA

off spring	0	1	1	0	0	1
MO (up)	0	1	0	0	0	1

It is used for more diversity

flipping of bit as per mutation Probability μ_p
 up decide if it is to be ~~mut~~ flipped or not
 If $\mu_p = 0$ Let it be as it is - not flip
 $\mu_p = 1$ flipped

mutated offspring MO	0	0	1	0	0	1
----------------------------	---	---	---	---	---	---

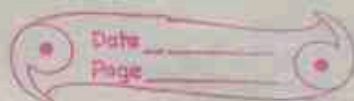
2nd method

off spring	0	1	1	0	0	1
MO	0	*	1	0	*	1

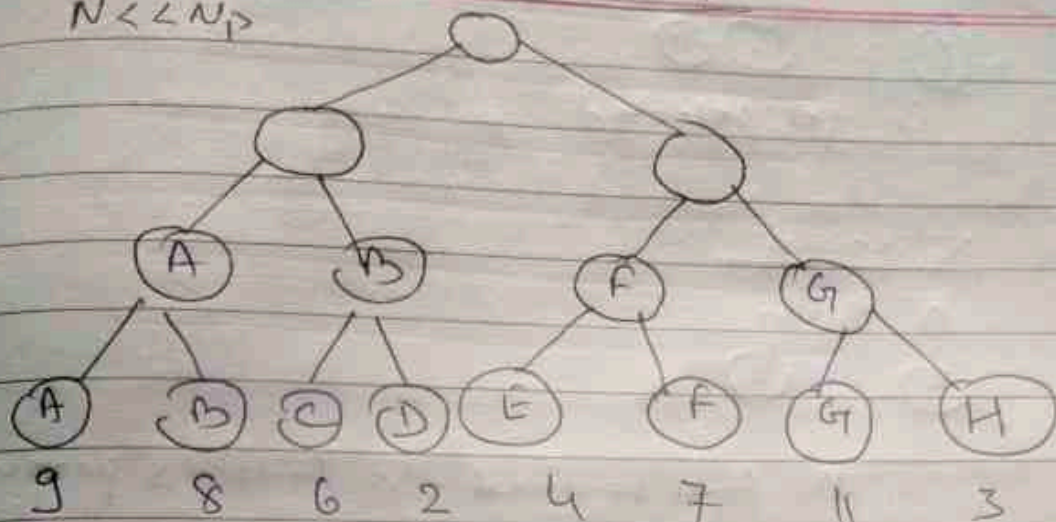
Randomly Select 2-bit from off Spring
 and then replace each other.. * indicate
 two random bit and are flipped to get
 MO

Tournament Conducted to find fittest parent.

Tournament Selection



$N \ll N_p$



To select N_p from N tournament is conducted.

How tournament is conducted

Randomly select any one without seeing the weightage

Ex Randomly select

from $\begin{pmatrix} A \\ 9 \end{pmatrix} \& \begin{pmatrix} B \\ 8 \end{pmatrix} \rightarrow A \text{ got selected.}$

$\begin{pmatrix} C \\ 6 \end{pmatrix} \rightarrow \text{get winner.}$

$\begin{pmatrix} A \\ 9 \end{pmatrix} \rightarrow \text{get selected}$

$\begin{pmatrix} G \\ 11 \end{pmatrix} \rightarrow \text{get selected.}$

} first round

Round-2

A - C compare
 $\frac{1}{9}$ 6

A/9 - get select

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(F)-7 (G)-11
G₁ got selected

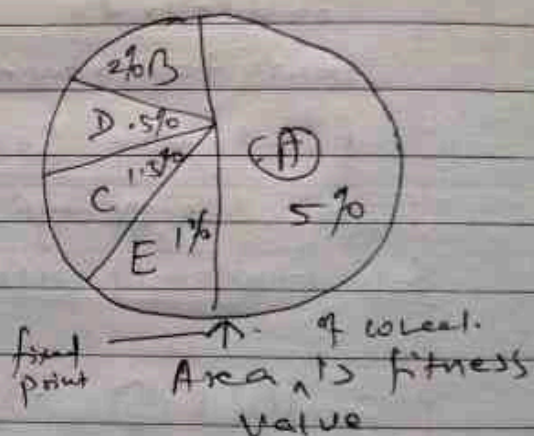
3rd round.

A & G₁
(5) (11)

G₁₁ is final winner. So G₁₁ is parent.

Roulette wheel Selection

C	F
(A)	5
(B)	2
(C)	0.5
(D)	1.5
(E)	1



fitness proportionate Selection. i.e

Selection \propto fitness

Population are A B C D E

fitness score 5 2 0.5 1.5 1 $\rightarrow 10$

50% is occupied by A

Rotate the wheel. When it stop pointer will

point to the area. That particular area is

Selected as parent.

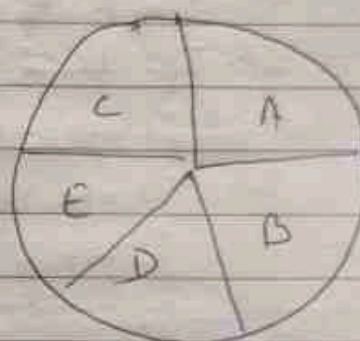
low Area has little chances to get selected

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fitness & Area occupied & Selection

Rank-Based Selection

	C	F	
(A)	5	5	33.33%
(B)	2	4	26.67%
(C)	0.5	1	6.67%
(D)	1.5	3	20%
(E)	1	2	13.33%
	10	15	



1st method

Travelling Salesman Problem

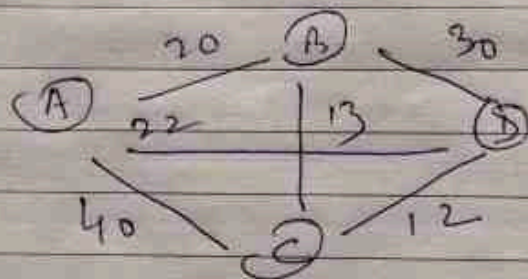
TSP and its state representation.

→ Salesman has list of Cities, each of which must be visited once.

→ In list there are direct roads b/w each pair of cities.

→ To find route, Salesman should follow shortest possible looped trip.

→ State is represented as pair of any two Cities and distance between them.



A, B, C, D are Cities.

Initial state space is A

$$L \left\{ A \xrightarrow{20} B, A \xrightarrow{40} C, A \xrightarrow{22} D \right\} \text{ One state}$$

To find shortest possible path.

$$A \xrightarrow{20} B \xrightarrow{13} C \xrightarrow{12} D \xrightarrow{22} A = 67$$

$$A \xrightarrow{20} B \xrightarrow{30} D \xrightarrow{12} C \xrightarrow{40} A \rightarrow 102$$

$$A \rightarrow C \rightarrow D \rightarrow B \rightarrow A \rightarrow 105$$

$$A \rightarrow D \rightarrow B \rightarrow C \rightarrow A \rightarrow 105$$

$$A \rightarrow D \rightarrow C \rightarrow B \rightarrow A = 67$$

$$A \rightarrow C \rightarrow B \rightarrow D \rightarrow A \rightarrow 105$$

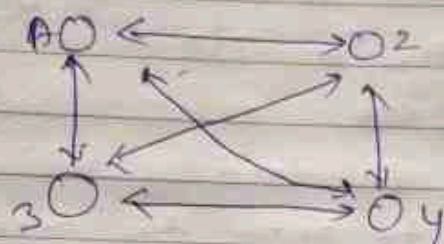
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Shortest path is

ABCD A & ADCBA

2nd method

TSP

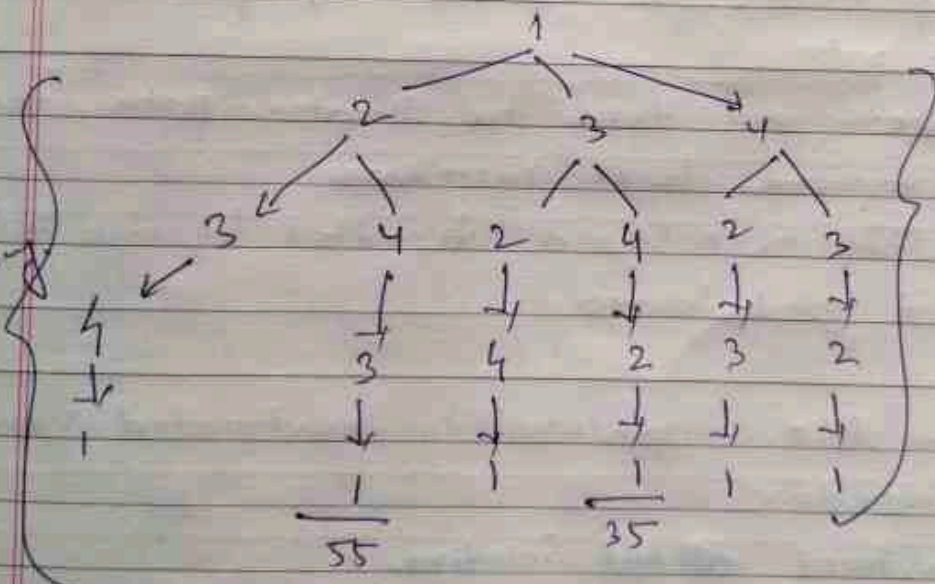


	1	2	3	4
1	0	10	15	20
2	5	0	25	10
3	15	30	0	5
4	15	10	20	0

General method



Dynamic programming



Brute force method

Check for all possibility and take the least state route

$(n-1)$ nodes are possible

Dynamic Programming for ants

Ant Colony Optimization

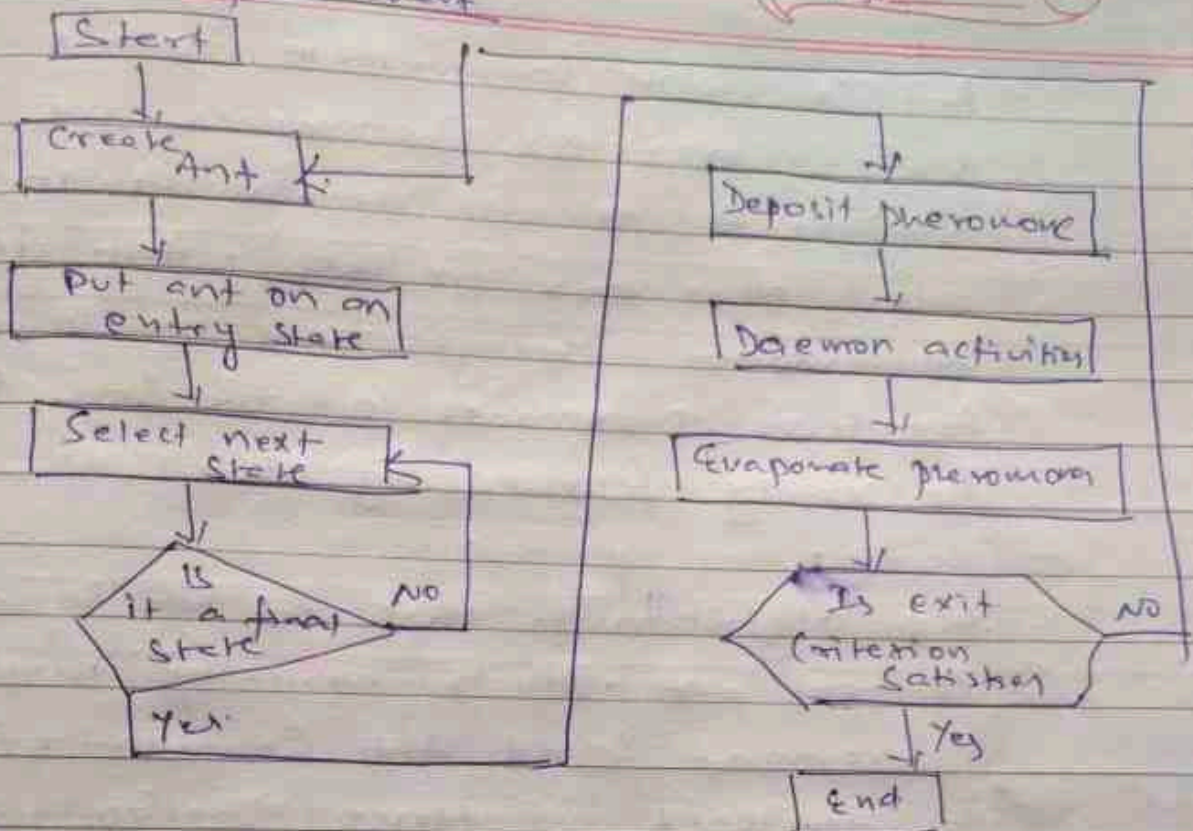
ACO is probabilistic technique for solving computation problem which can be reduced to find good path through the graph. i.e. finding shortest path in TSP

- * Ant Search for food.
- * The shorter the path the greater the pheromone left by an ant.
- * The probability of taking a route is directly proportional to the level of pheromone on that path.
- * As more ants take the shorter path, the pheromone level increases.
↓
It is a liquid deposited by ant when they found good source.

The ACO Algorithm is constructed using three Procedure

- 1) Construct Ant Solution
- 2) update Pheromones
- 3) Daemon Actions

flow chart



→ Applying Ant Colony optimization to TSP

Step-1 Initialize Important Parameters.

Population size = 04

Max. iterations = 10

Pheromone Evaporation rate = 0.05

Artificial pheromone = 0.0453

$\alpha = 1$

$\beta = 1$

$Q = 1$

→ Ants can easily communicate with each other using pheromones. Pheromones are chemical signals. Ants release pheromones in danger (to alert other ants for help). Ants detect pheromones through their mobile antennae. Ants leave pheromones on the soil, that can be easily followed by other ants.

==

Aco algorithm is inspired by social behaviour of real ants. It is basically inspired by pheromone based Ant Communication. Aco is developed by Marco Dorigo in 1992. This technique is used to find optimal path. Aco is used for Routing and load balancing problems, for eg TSP. Aco can be continuous optimization problem.

→ Ants communicate with each other indirectly using a chemical known as pheromone. With the help of pheromone signal, ants can easily find shortest path between nest/colony and food.

- Ant ~~are~~ explore area randomly
- They release chemical ~~are~~ called pheromones when ant find food source they choose the way have strong pheromones
- A group of Ants foraging in the environment for food randomly. Forger will mark trail on the way using pheromones while going back ~~to~~ to the colony. Forger will mark trail on the way using pheromones when food source is finished. No new way are mark by returning ant

Case-1 All ants are in nest. There is no pheromone mark on the ground

- ② Foraging start → 50% ant take shortest path & 50% take longest path.
- ③ Pheromone mark on the shorter path have strong pheromone signal. Probability of this path selected by other ant increases

Aco algorithm

- 1) Initialize Aco parameters
- 2) Ant Solution Construction
- 3) Position each ant in to starting node
- 4) Each ant will select next node by applying State transition rule
- 5) Repeat until ant build the best solution, then compute the fitness value.
- 6) update best solution
- 7) Apply offline pheromone update
- 8) Display best result

Particle Swarm optimization

350	1350	10793
719	6147	1500
520	61	4000
580	950	16293
	1226	+ 6000
	150	22293
	909	
	<u>320</u>	

Traveling Salesman Problem

Process of patent filing:

First Provisional date filing.

2-kind of application — Provisional date

→ non-provisional date

Provisional → we want to protect the priority date because we have just developed the concept and need some time to work on that concept & idea. It is just to protect the idea and priority date

2nd

Non-provisional or Complete Specification → Filed when invention is complete.

→ Within 12-month of filing provisional application go for non-provisional application

→ 18-month from provisional application patent publish

→ 48-month to file request for examination. They will file examined. They will come with few objection or clarification. They will submit first examination report, which will be given to you. I need to take action.

→ If not examination they will call for patent hearing. If still not satisfied, patent not granting.

→ for Speed

Trade-mark

Date _____
Page _____

Notes

→ Brand → 1958 (Trade & Merchandise ^{Mark} Act)

Brand name and brand mark.

Brand Register Trade mark.

→ It is any work, name or symbol (or their combination) that let us identify the goods made by any individual, company, organisation etc.

Conventional

work, label, logo, packaging

Shape of goods

Non-Conventional

→ Sound mark IPL music

→ Special taste.

→ Dynamic mark.

Smell

Patent Used for Invention

→ Scientific Inventions.

→ Industrial application. } Condition
non obvious.
new-invention

Act 1970 Sec-384 } not patented → Scientific principle, method of agriculture, horticulture, method of treatment, traditional knowledge, atomic energy Inventions
Original literary work } Valid for 60 years after death of author
Dramatic work
Musical work
Artistic work

60 counted from next calendar year.

→ Cinematograph
Sound.

Posthumous publication (Published after author death)
Anonymous and pseudonymous publication (Pen name)

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Type of IPR

- ① Copyright → 60 years valid. Counted after death of author.
Book, music, movie, software, writing, photograph, audio, video, sculpture, architectural work
- ② Trademark → artistic work, literary work.
- ③ Patent → 20 years right
- ④ Geographical Indicators
- ⑤ Design
- ⑥ Plant Variety
- ⑦ Semiconductor IC layout Design

1957 → Copyright Act (Right to not copy) (effective 1957)

Right include

- Right to copy (reproduce) a work
- Right to create derivative work based on it
- Right to distribute copy of the work to the public
- Right to publicly display or perform the work

What is protected

Literary work → Pamphlet, Brochure, Novel, Books, poem, Song lyrics, Computer program

Artistic work → Drawing, painting, Architectural drawing, Tech. drawing, Map, logos

Dramatic work → Dance, movies, Screenplay, Musical work, sound recording, Cinematographic

fi — → 8 → 2118.

Act Amended on 1983, 1984, 1992, 1994, 2012

118218
1983
1984
92
93

→ work of govt & work of international organisation

Type of IPR

- 60 years valid. Counted after death of author.
- ① Copyright - Book, music, movie, software, writing, photograph, audio, video, sculpture, architectural work
 - ② Trademark - artistic work, literary work.
 - ③ Patent → 20 years right
 - ④ Geographical Indicators
 - ⑤ Design
 - ⑥ Plant Variety
 - ⑦ Semiconductor IC layout Design

1957 → Copyright Act (Right to not copy) (effective 1970) for

Right include

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