

19 attr

L
14

→ entropy closer to zero

iska info gain 0.94 ho jaye gg

$$G(S, A=\text{out}_0) = 0.94 - 0$$

→ gain ratio use Karlson to
Solve this biasedness problem



post rule post pruning

poora chap 3
ahi hogा

→ jtna shallower decision tree htna
bhtar

rule post pruning

→ stop cut off / prune off

after you have done your training,
over-fitting ke bad rules

Woh kinsaasay, aap unki pre conditions
ko nikaal den

A ~~AB~~ \cap X → Class 1

ak ko nikaal kar dekھئ
kya arha.

validation error banta toh you
can't cut off C, otherwise you
can remove

X \cap BAC

root ko bhi nikaal saktey

→ missing values

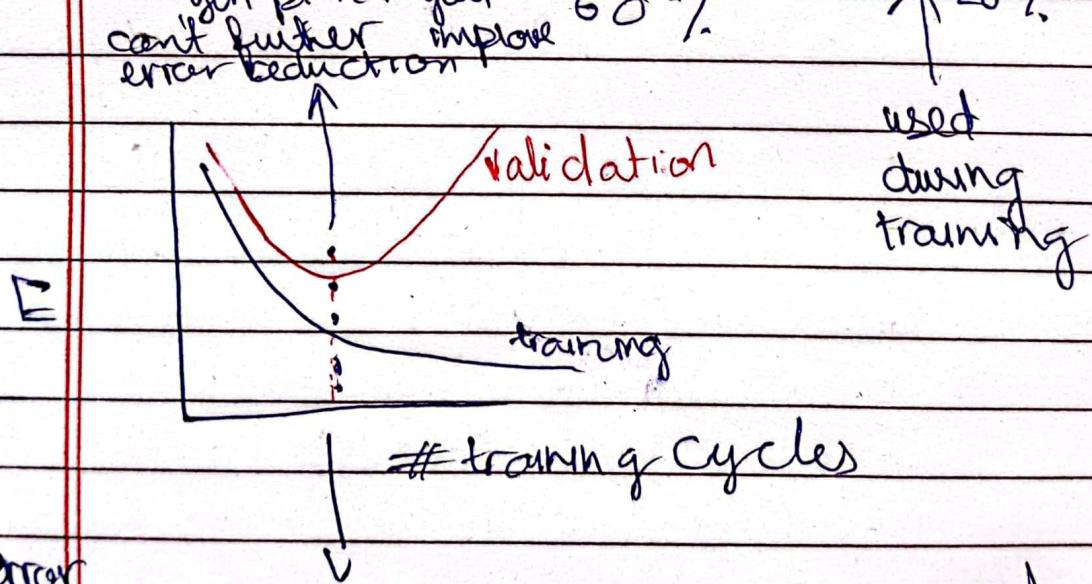
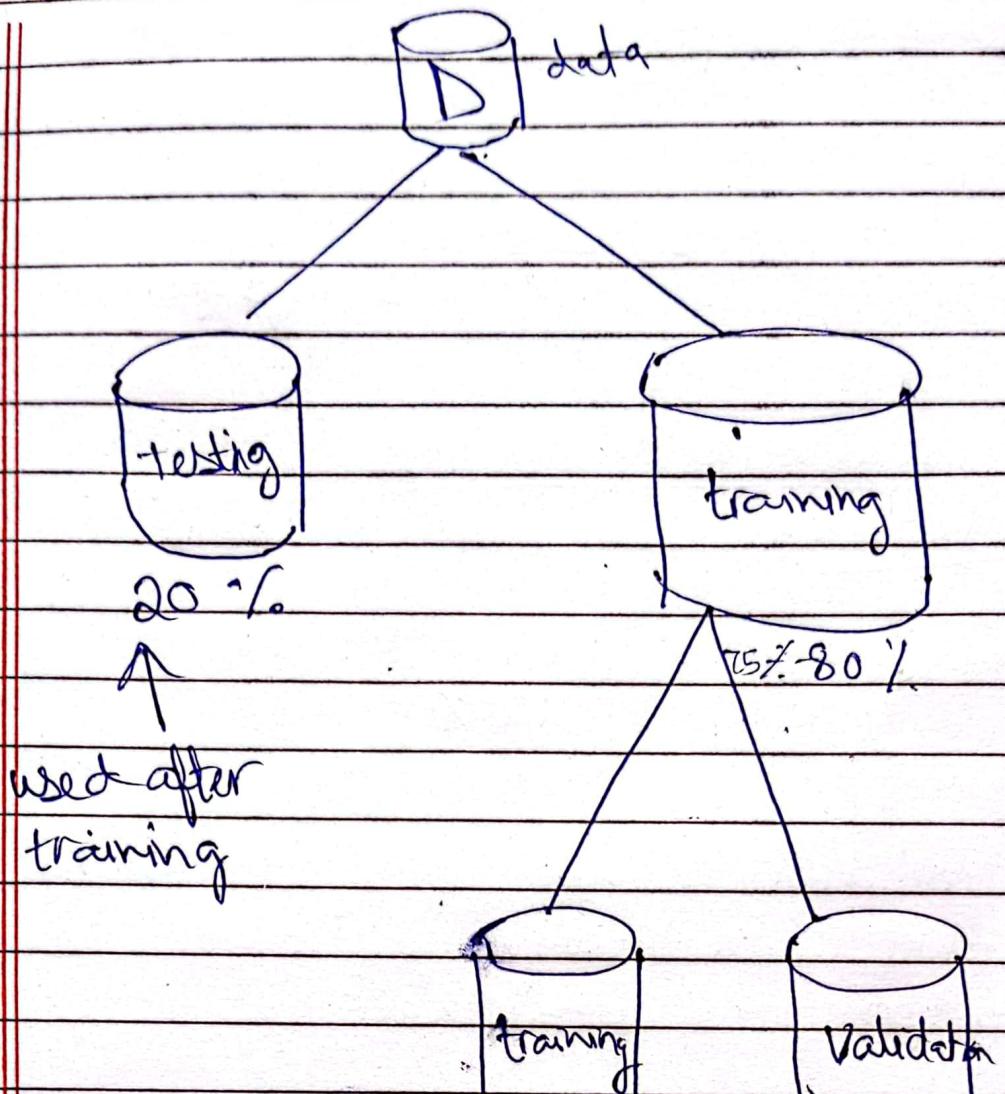
→ dt pure partitions karne chahta
hai

ak attr 10 possible val ak kli 4

↓
individual entropy
array less array ko
change

ak kli 2

purity toh wala karne ziyada



error detection cycle
 validation used as yardstick
 (Early stopping)
 beyond your it is over fitting

M.M. Tues. Wed. Thurs. Fri. Sat.

examples empty : no value against it
no, existence

ID3 → bases best on entropy
(info gain)

issues associated with virtually any ML model:

→ overfitting (fitting perfect)
100%
near 100%

→ test error is high, applying data set to predict learn karta but unseen data per pool work karta

noise in data, model learned noise & regularities of data

high variance scenario

how to avoid over-fitting?

→ ID3 would always overfit, unless there are some checks to stop it from learning to more than some certain depth

ya toh 7-8 samples hain,
usek further split na karen

→ Validation set banata, aik portion reserve for testing

agar sacay no kahi toh entropy = 0

Mon Tue Wed Thu Fri Sat

Date: 1/1/20

$$C_G(S, A=\text{outlook}) = 0.24$$

Pure ki entropy 0, pure isse kehtay
jiski sirf aik hi class aayeg.

recursion seh khatuk call karay
thoray attributes reh saath ga
reduced set keh liye kaam karey ga

agar root node ka poochay
toh sabka calculate karengay

Waala specific poochay toh aik
ka hi.

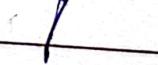
Informative attr

a bit less informative
attr



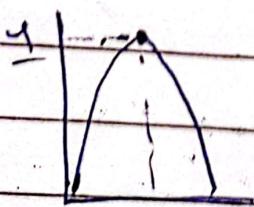
lesser lesser

a bit less
informative
attr



lesser

for boolean classification
toh entropy can be max 1
et mn 0



Max entropy 1
tab aati jab dono
classes ki prob same.

$$C_1(S, A=\text{outlook}) = 0.94 - \left(\frac{|S|}{|S_A=\text{sunny}|} \cdot E(S_A=\text{sunny}) + \frac{|S|}{|S_A=\text{overcast}|} \cdot E(S_A=\text{overcast}) + \frac{|S|}{|S_A=\text{rain}|} \cdot E(S_A=\text{rain}) \right)$$

$$E(S_A=\text{sunny}) = \frac{|S|}{|S_A=\text{sunny}|}$$

$$\cdot E(S_A=\text{overcast})$$

±

$$0.94 - \left(\frac{|S|}{|S_A=\text{sunny}|} \cdot E(S_A=\text{sunny}) + \frac{|S|}{|S_A=\text{overcast}|} \cdot E(S_A=\text{overcast}) + \frac{|S|}{|S_A=\text{rain}|} \cdot E(S_A=\text{rain}) \right)$$

$$E(S_A=\text{overcast}) = \frac{|S|}{|S_A=\text{overcast}|}$$

$$\cdot E(S_A=\text{overcast})$$

$$+ \frac{|S|}{|S_A=\text{rain}|} \cdot E(S_A=\text{rain})$$

$$|S| = 14$$

$$|S_A=\text{sunny}| = 5$$

$$|S_A=\text{overcast}| = 4$$

$$|S_A=\text{rain}| = 5$$

$$E(S_A=\text{sunny}) = -\left(\frac{2}{5} \log \frac{2}{5} + \frac{3}{5} \log \frac{3}{5}\right)$$

* Saaray yes of overcast teh entropy = 0

$E(S_A=\text{overcast}) = 0$ since only one class exists that's why entropy = 0

decision trees used for taking decisions

given your data, temperature doesn't matter

Criteria for selecting attribute as root:

$\rightarrow G(S, A)$ is maximum

$$G(S, A) = E(S) - \sum_{v=1}^m |S_v| \cdot E(S_v)$$

↓ ↓
entropy entropy

S barai set hai

$$E(S) = - \sum_{c=1}^2 P_c \log_2 P_c$$

$$= - (P_{\text{yes}} \log_2 P_{\text{yes}} + P_{\text{no}} \log_2 P_{\text{no}})$$

$$= 0.94$$

$c=1 \rightarrow$ means yes

$c=2 \rightarrow$ No

$$P_{\text{yes}} = \frac{9}{14}$$

$\log_2 \text{yes} \text{ tah } \log_{10} \frac{9}{14}]$

Karo

$$P_{\text{no}} = \frac{5}{14}$$

$\log_{10} \frac{5}{14}$

$\log_{10} \frac{5}{14}$

corr

$\log_{10} \frac{5}{14}$

So

★ So

AI

Kuch main sirf input provide karay,
Ex ML looks for patterns. categorical data
 \rightarrow discrete labels { e.g.: sunny, rain, 0C }

$$S = \{ \langle x^1, y^1 \rangle, \langle x^2, y^2 \rangle, \dots, \langle x^n, y^n \rangle \}$$

n tuples in data set

$x^i \rightarrow$ table with attr & value

$$x^i = \{ (a_j, v_j) \}$$

$$y^i = \{ 0, 1 \}, \{ \text{yes}, \text{No} \}, \{ +, - \}$$

$\{ 1, 2, \dots, k \}$ Binary classification
 multi-class classification

2 tasks of supervised learning

- classification
- regression

Numeric values

$y^{(1)} \in \mathbb{R}$ } regression

$y^{(2)}$

isi input
normally
numeric
attributed kint
categorical
khi ha sakay
but output
is always
numeric.

probability = 1 if all yes; message $\neq 0$.

adhi gets adhi no, 1 bit to encode yes,
other bit to encode no.

$$\frac{1}{2} \log_2 \frac{1}{2}$$

$$\log_2 2^1$$

$$\frac{1}{2} \times (-1) = -\frac{1}{2}$$

$$\begin{aligned}\log_2 \frac{1}{2} &= \log_2^1 - \log_2^2 \\ &= 0 - 1 \\ &= -1\end{aligned}$$

tak negative sign iss liege lagay
khn positive ho jaye

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

$$\text{Entropy}(S) = - \left[\left(\frac{9}{14} \log_2 \frac{9}{14} \right) + \left(\frac{5}{14} \log_2 \frac{5}{14} \right) \right]$$

$$\begin{aligned}&= -(-0.94) \\ &= 0.94\end{aligned}$$

id3 algorithm
iterative descenizer version 3

id3 algorithm : 3, 4, 12 \rightarrow 12.1 + decision tree

numericals, mcqs, short conceptual
questions

Gain(S, outlook) = 0.246

Entropy : degree of purity ko number assign karta hai

2 classes hain (binary) : yes or no

$$\text{Entropy}(S) = - P_{\oplus} \log_2 P_{\oplus} - P_{\ominus} \log_2 P_{\ominus}$$

probability of seeing a yes

$$P_{\oplus} = \frac{9}{14} \quad \begin{matrix} 9 \rightarrow \text{yes} \\ 5 \rightarrow \text{no} \end{matrix}$$

$$P_{\ominus} = \frac{5}{14} \quad \text{probability of seeing a no}$$

$$\log_2 P_{\oplus} = \log_2 \frac{9}{14}$$

log kyun kiya gaya
is saath no. kyun multiply
kiya
or why the minus sign

entropy in physics: measure of disorder
why logs used?

↳ How many bits do you need to encode the qty

Same classification → so you need message with 0, don't need any bit to encode this message



In a set of decision trees, it selects a simple decision tree.

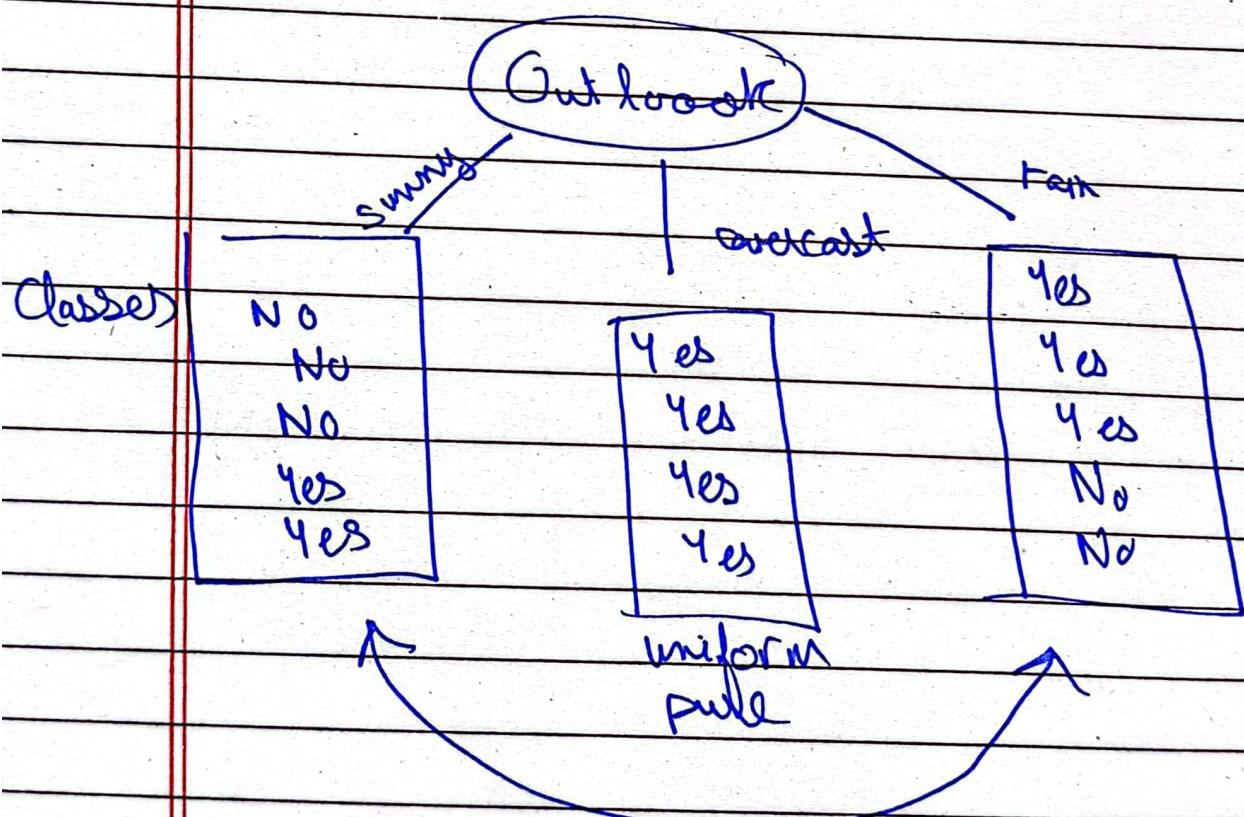
→ aik hi node wala sabse kam level wala decision tree ho sakte

(Yes) → simplest possible decision tree

→ shallowest, with least attributes

→ rarely possible

Sabse ziyada uniform classes ki distribution ha.



also root node select karen jisse partitions as pure as possible boundary

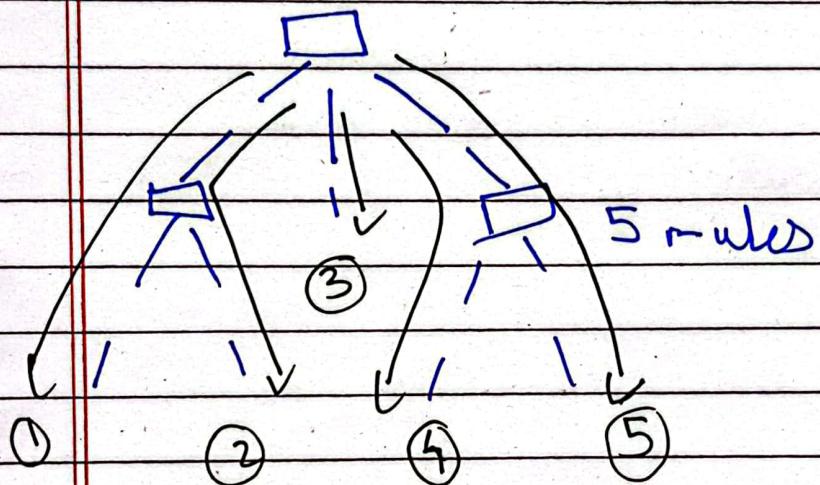
Decision Trees

→ categorical data
(no numeric attribute)

ch3 of another book

Whether seh tell koi tennis khelta.

decision tree → temperature ^{dt} men included
whi hai



how to decide konai attribute use kona konai whi, kisko root node banana.

agar temperature numerical data hai
toh you want to make into categorical
data interval bana saktey, there
are other systematic methods as well.

→ recursive algo (data men aik attribute
select karke data wo binaa
per split karay)

T/P O/P

x	f(x)	= x ²
-1	1	
0	0	
1	1	
2	4	
3	9	
:	:	

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$h(x)$
program

$g \rightarrow$ unknown function (generated your data)

element

(Hypothesis Space)

H (set H)

h is closest to g

We cannot access g function

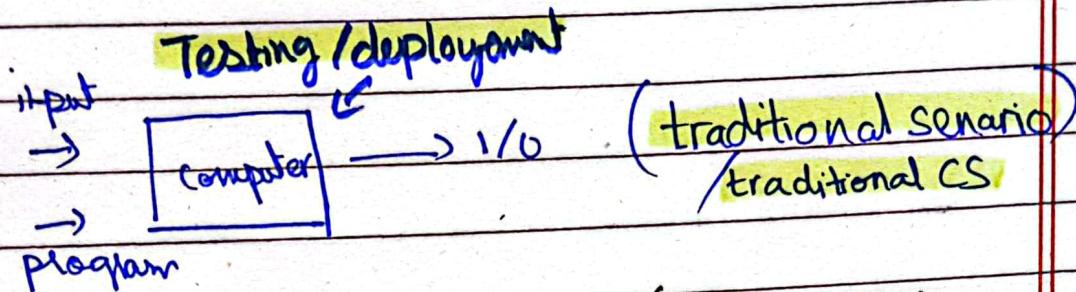
Consistently (pooray
data ka explain karday)

Set of all
possible

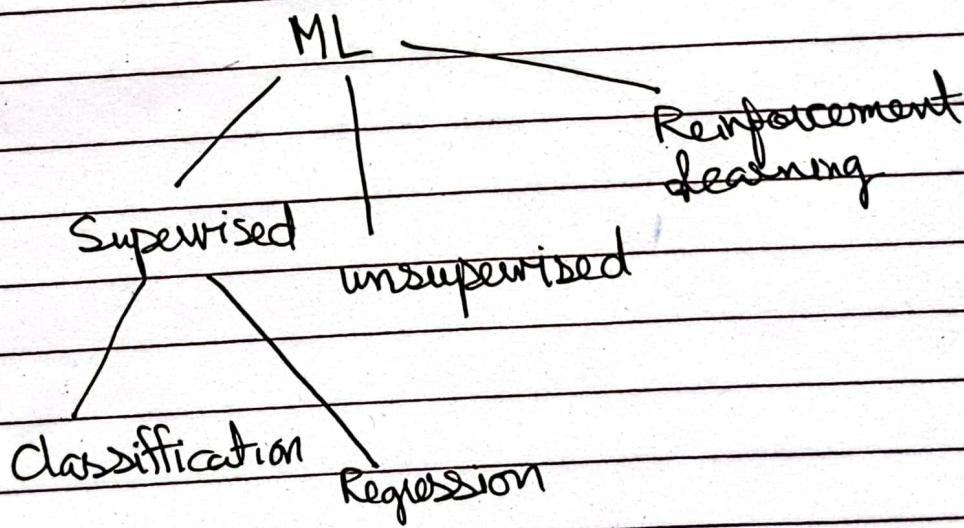
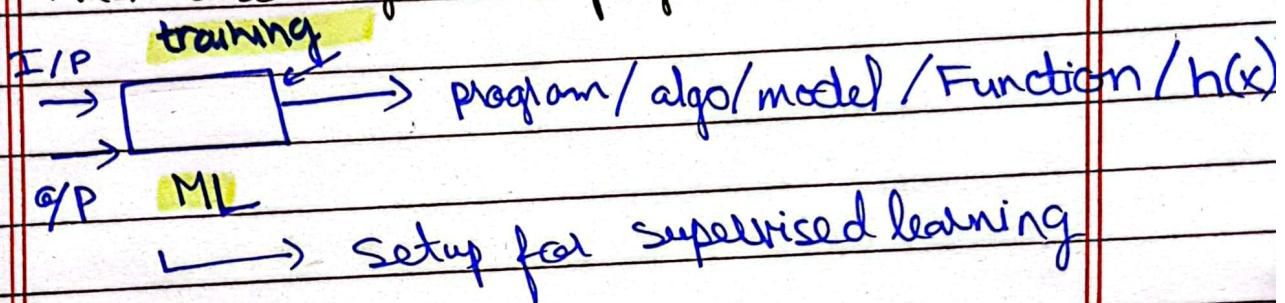
polynomials
in this case

AI

ML - Supervised learning mainly



Machine learning ki category (Supervised Learning):



in Supervised Learning (ML)

Learning: is performance improvement
in a sufficient task given sufficient
data.

Date: _____

$$f_i = \frac{1}{c_i}$$

fitness; jiski cost kam

fitness

selection

crossover → single parent
mutation | not allowed

invalid sol:

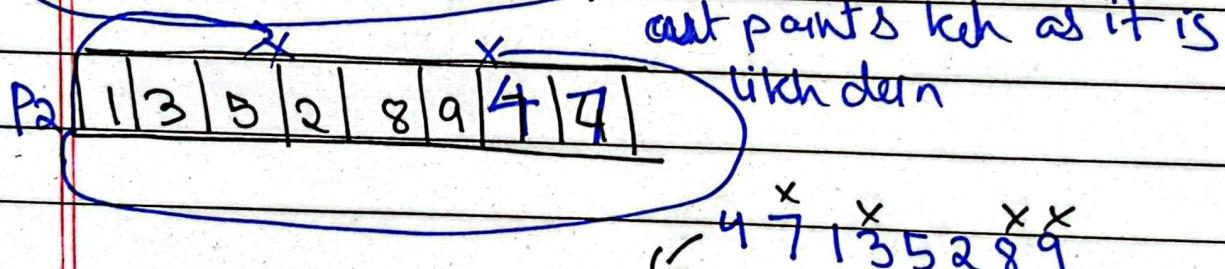
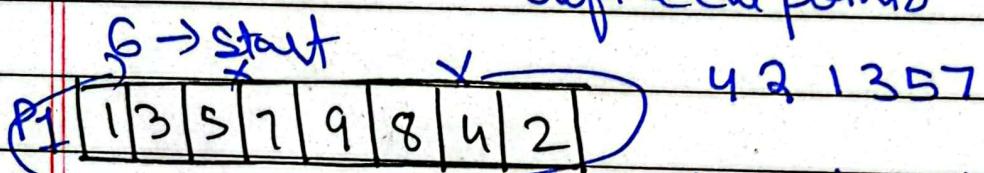
Sab ko visit

whi kar paye ga,
we cannot visit
one node twice

1985

order crossover

define cut points



C₁ | 4 | 1 | 3 | 7 | 9 | 8 | 5 | 2 |

C₂ | 4 | 1 | 3 | 2 | 8 | 9 | 5 | 7 |

mutation : 2 random cities ko pakar
kar swap karen

Date: _____

Totology

homeshi hi false hou

garanteet karta sol aa jaye gi, koh loop
chalti jaye & sol exists \rightarrow please confirm it

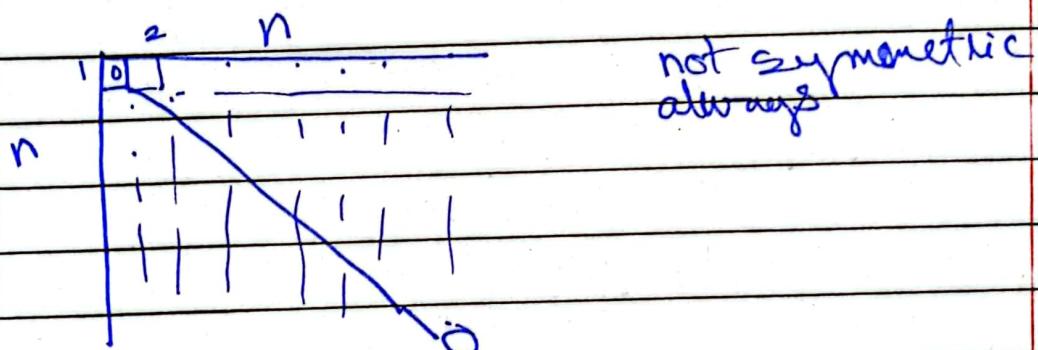
TSP problem

n cities to visit

host per stat karna & return to it

n-1 nodes \rightarrow one time visit

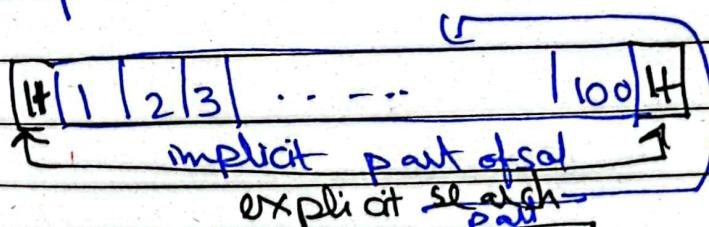
1 node \rightarrow 2 times visit



give best loop

$n!$ solutions \rightarrow agar brute force
n factorial

Representation kesa?



Nearest neighbor heuristic is good if not
genetic algo.

Selection:

roulette's wheel

25 dafa crossover kartaay 50 create ho jayen gaay child.

We set limit for crossover.

Crossover

- single point bhi ho sakte
- multipoint bhi ho sakte (define parts)
- uniform bhi ho sakte

can generate index randomly and
bachon ki bhi wahi pass wohhi index meri
hoy.

uniform; probability assign karein
0-1 keh between.

phir fill kartay bit by bit, uniform probability.

fair coin

intraligation selection

crossover

mutation bhi bari asan hai.

replacement

25% weak nikaal do, define
your replacement, full, partial

* Soch lein and implement it.

Date: _____

1) CNF Satisfiability Prob

Conjunctive normal form

boolean expression $A \neq B$

$\neg a \vee b$

$a=0$

$b=0; \neg a \vee b = \text{True}$

$0, 1 \rightarrow a, b$

boolean algebra seh nikli

not in cnf form

disjunction

$\neg a \wedge b = \text{True}$

$a=0$

$b=1$

conjunction

can be arbitrarily long :

$(\neg a \vee b) \wedge (a \vee d \vee \neg e) \wedge (\neg c \vee f) \dots$

clauses

CNF is conjunction of disjunction (clauses)

Exponential growth
3 var 2^3

hard problem

genetic algo for CNF problem

1	0	0	1	1	0
a	b	c	d	e	f

goal iska True karna poora eg.

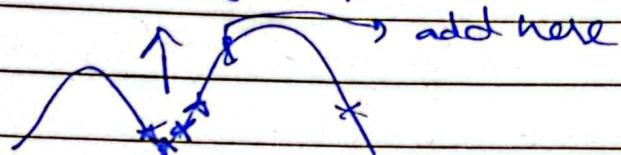
fitness: no. of true clauses
total no. of clauses

Date: _____

If there is only one hill, greedy search or hill climbing search is ^{the} best approach, would give optimal solution

Genetic algo works on multiple solutions simultaneously

ak hi jaga per sab clustered



diversity ka kya farda?

like sab fitness ratio almost same ac rahi hain
jab stagnant ho jaye toh stop

As the number of size increases, so
brute force is not feasible

there is a method to madness
Search meta heuristic

Hill climbing mein best ko b kai chalen.

Neural networks mein aur jaye gi hill
climbing ho jani

Q. 1 It is not necessary for knapsack problem to use genetic algo, you can just use hill climbing

tournament selection

Knapsack 0-1 → for genetic algorithm problem

It works

characteristics

it helps transfer fitness traits parent to offspring

Q. Why we do crossover in genetic algo? ↗

Q. What is the relationship of genetic algo & hill climbing search?

Q. What is the benefit of using mutation operator?
→ it creates diversity (variety)

→ genetic algo is a parallel version of hill climbing.

Hill climbing main aim possibility.

Faida kya agar 1 banda hui ya 100 banday
any benefit?

(take a step in the direction of the goal)

we have multiple hills.

Hills ka kya mukhabbat with an algo?

→ fitness vs ratio / quality
landscape



min find karay
global min

Ab yeh optimization prob. We have

found minimum using differentiation.

$$\frac{dy}{dx} = 0$$

Date: _____

CS2 10111000

CS3 10110100

$$\frac{200 + 150 + 250 + 120}{5 + 15 + 4 + 7} = 23.22$$

$$\frac{200 + 150 + 230 + 160}{5 + 15 + 4 + 8} = 23.75$$

Weight is exceeding in both cases

31, 32

So for our sake of purpose, we ignore
the weight was 30 & set it to 35 kg

Ab weakest population ko may replace
korden. In CS'2' & CS'3' soh

CS-2, CS-9 to replace korden

if invalid sol (weight exceed):

choose an existing sol or initialize
randomly a CS-5

go with the flow of algorithms

otherwise for selection:

tournament Selection

$r_2 \in [0,1] = 0.5$ CS-2

CS-3 & CS-2 per gene) crossover

CS2 0 1 2 3 | 4 5 6 7
 00 11 | 0 1 0 0

CS3 1 0 1 1 | 1 0 0 0

define type of crossover, then apply it

point

* single pt. crossover

question
pertinent

CS2' 0 0 1 1 1 0 0 0

bohot ziada

mixing hi

Kamii

→ preserve wall

ches khatam

no jayegi

CS3' 1 0 1 1 0 1 0 0

Mutation

→ Pick a random location (gene)

& flip it.

let the nature take its course

→ 2-3% of gene

16 genes have to pick random location
per change Kartay, first set
randomly CS set them bits within
the random search.

Iteration = generation

Termination conditions:-

i) desired fitness threshold has been achieved

(ii) Defined # of generation have expired

$$\frac{f_i}{\sum f_i}$$

Fitness ratio

$$f_1 = 22.14$$

$$22.14 / 87.63 = 0.25$$

$$f_2 = 20.74$$

$$= 0.23$$

$$f_3 = 23.72$$

$$= 0.28$$

$$f_4 = 21.53 \quad 0.56$$

$$= 0.19$$

$$\sum = 87.64$$

Roulette wheel Selection

also called probability
but better to call it fitness

Cumulative fitness

0.27	0.27
0.428	0.52
	0.8
	0.99

↓ CS3

CS-1 CS-2 CS-3 CS-4

0 0.27 0.52 0.8 0.99

selected

$$r_1 \in [0,1] = 0.74 \text{ C}$$

$W_{\max} \leftarrow 30 \text{ kg}$

Item	Value 1000PKR	Weight
1	200	5
2	100	10
3	150	15
4	250	4
5	120	7
6	160	8
7	50	12
8	40	6

Q-1 Knapsack Problem (Solution via Genetic Algorithm)

CS-1

0	1	2	3	4	5	6	7
1	0	0	1	1	0	1	0

CS-2 ~~000110100~~ 00110100

CS-3 10111000

CS-4 00110110

Fitness

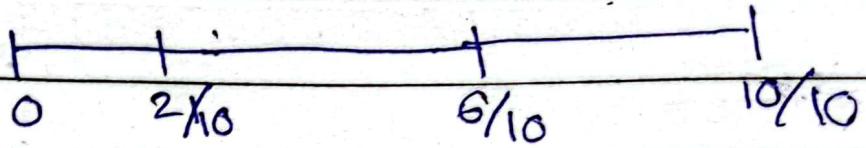
$$f_1 = \underbrace{200 + 250 + 120 + 50}_{5+4+7+12} = 22.14$$

$$f_2 = \frac{\cancel{200} + \cancel{250} + \cancel{150} + \cancel{160}}{\cancel{5} + \cancel{4} + \cancel{7} + \cancel{6}} = 20.74$$

$$f_3 = 23.22$$

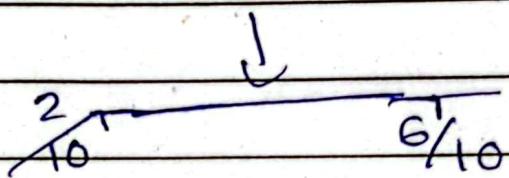
$$f_4 = 21.53$$

Date: _____



roulette wheel logic

O seh I keh damyaan random no.
generate kallen gay



random no ka kya probability hai
→ uniform

gaussion whikallen gay

selected ka tag laga den gay.
wskay apne crossover ki hoga
and fab tag O ho jaay ga.

When to stop?

1. Job desired fitness achieve ho jaye
2. Predefined no. of generations expired
3. Stagnation

Can deteriorate best
We say evolution mein
improve hi hata

$t+1$

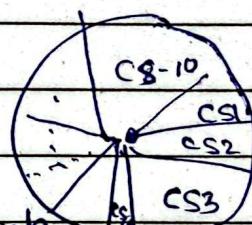
ake aur baar chalay ga for another time
(second iteration.)

Selection:

1. Fittest %

2. Roulette wheel selection
gaming wheel

lets assume fair roulette wheel hai



unequal slice
hai.

fittest slice ka
portion ziyada tak

	Fitness	Prob
CS-1	2	2/10
CS-2	4	4/10
CS-10	3	3/10

It is Fair, as it
favours fitter ones more
but there is an element
of uncertainty

$$\sum f_i = 10$$

Fitness ratio (Prob)
2/10
4/10
3/10

Cumulative fitness ratio

2/10

6/10

10/10

Date: _____

Crossover point can be selected at middle.

crossover
after creating children:
→ ab mutation (agnani (random
change))

bits flip randomly
in rhapsody

→ mutation is a random change

8 genes mutation
1 gene 5-6%
2 genes 10% or more

⇒ 100 genes mean 2 ya 3 genes generate
no solution nahi.

Diversity, Stagnation
variety

→ replace P(t) with some offsprings

10 ki population

2 children create

aj population 12.

but same rakhne ke liye

you have to remove some

add based on fitness.

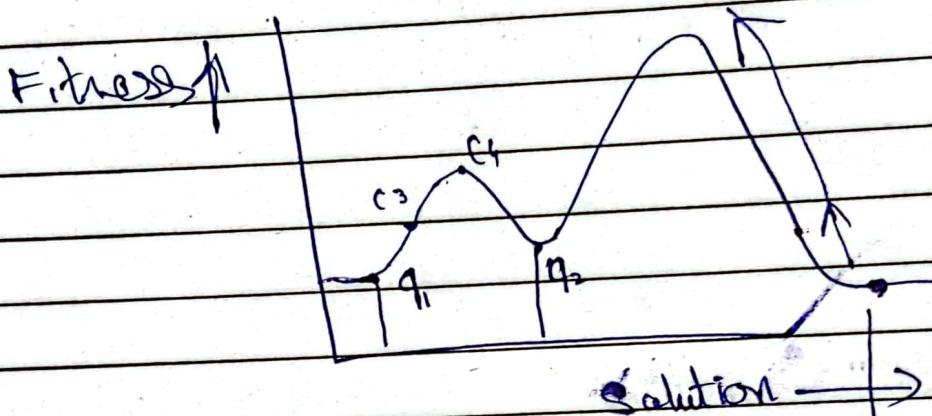
partially replace

total replacement 3 crossover karna padega
all old parents die

Date: _____

- Select a pool of fit solutions (parent)
top 4, top 2

agar saay elite hon toh we can be stuck after awhile, and improve nhi hoga ziyada



q_1, q_2 seh C_3 banay a

C_3 seh C_4

ak random

ak ruch topped

skr igne
kya

toh agar

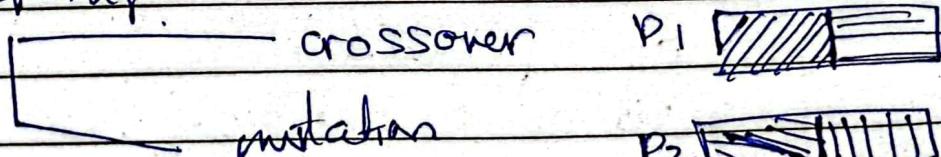
exploit karty

toh kya

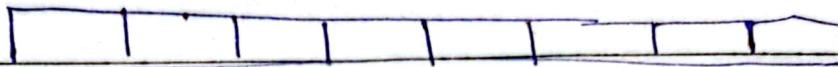
pata global

max per

- apply genetics ops to create offspring



one of the ways to create, mix children
children G1 [|||||]
G2 [|||||]
G3 [|||||]



18

Item	Value	Weight	$\sum w_i \leq w_{max}$
i_1	v_1	w_1	
i_2	v_2	w_2	
:	:	:	
i_8	v_8	w_8	

TSP is a description of problem not only for actual travelling salesperson

Fitness (CS-1)

$$= \frac{v_1 + v_4}{w_1 + w_4}$$

CS-1 [11010] 11

$$w_1 + w_4 \leq w_{max}$$

Can generate illegal solution as well.

CS - 1

:

CS-10

- evaluate
every
time

(1) human legal change, toh illegal
to stop & generate more till requirement
fulfill

Date:

$$S = 2(b) - 1$$

$$S = b^{(d+1)/2} + b^{(d-1)/2} - 1 \quad \text{add}$$

even

Genetic Algorithm ($P(t)$)

representation
of solution
↓

$$P(t) = \{x_1^t; x_2^t, \dots, x_n^t\}$$

genetic algo ($P(t)$) ↓

$$\rightarrow t=0$$

→ initialize $P(t)$ [initialization framework]

(generations per iteration) while termination criteria isn't satisfied }
↓

- evaluate every x_i^t (find its fitness)

- select a pool of fit solutions (parents)

- apply genetic operators to create children (offspring)

- Replace $P(t)$ with some offspring

$$\} - t = t+1$$

↓

dk if we've reached solution

Knapsack mem sol is array of 1s, 0s of len =
amt of object.

knapsack problem

bool vals

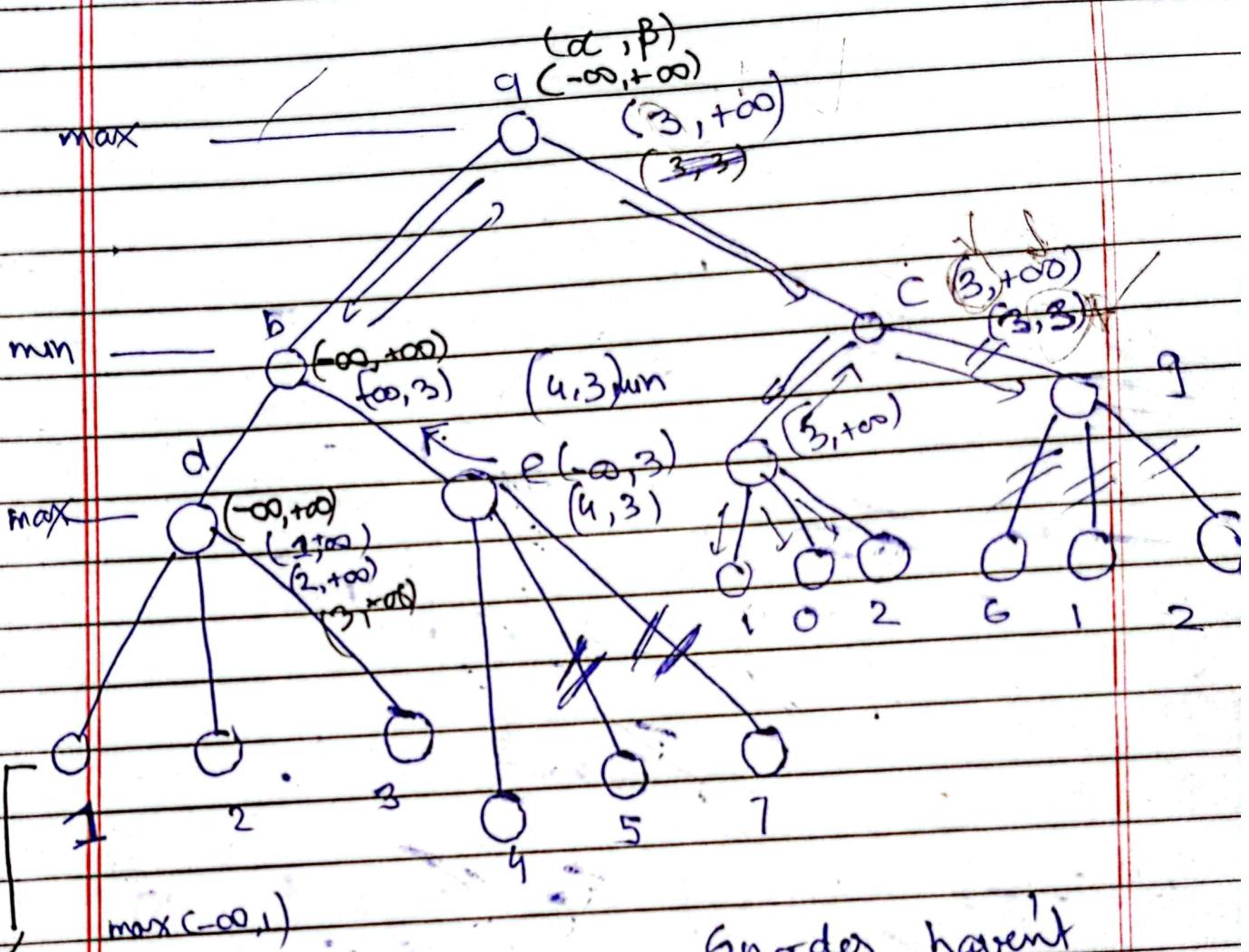
initialized
nai

$$x_i^t = [0 | 1 | 1 | . | 1 | 1]_{i \in n}$$

Date: 24/10/24

AI

↙ generated on the fly



max $(-\infty, 1)$

ply depth

no max level

jahan tak aap

apna game

tree generate kar

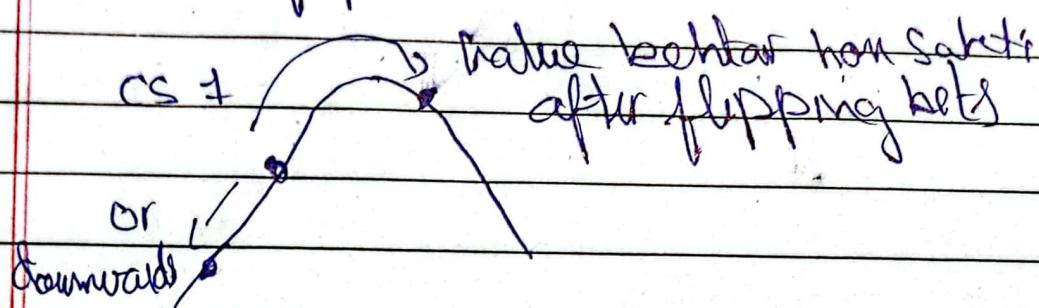
sakday

6 nodes haven't
been explored

$$d \leq \beta$$

Date: _____

3.b) Mutation
→ flip bits



can discard some

gives good result with such simplicity

TSPs via 4 generations
(iterations)

until you meet the termination criteria.

Solve alphabeta pruning, Monotonicity
bit away q.

Date: _____

2) Fitness criteria

$$\frac{V_1}{W_1} \text{ value of sol 1}$$
$$\frac{V_2}{W_2} \text{ weight of sol 1}$$

$$W_1 \leq W_2$$

jiski value ziyada woh ziyada acha
ratio achi tak woh acha

$$\frac{V_1}{W_1} \frac{V_2}{W_2} \overline{110} \text{ illegal val x}$$

3) Improving Solution Quality

- isman randomness kaifi, random search
- but there is a method to randomness
- better than random
- make changes in existing solution

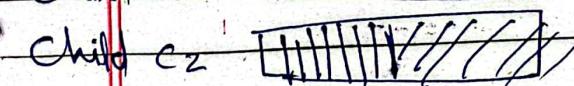
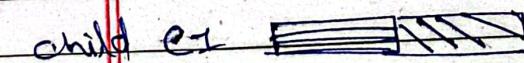
Operators:

3(a) crossover

→ mix solution keh pieces



take 1 part from parent 1
+ other part from parent 2



→ adaptability
→ inheritance mein usko intelligence mil gayi

→ inheritance is biased
→ fitter parents, fitter kids

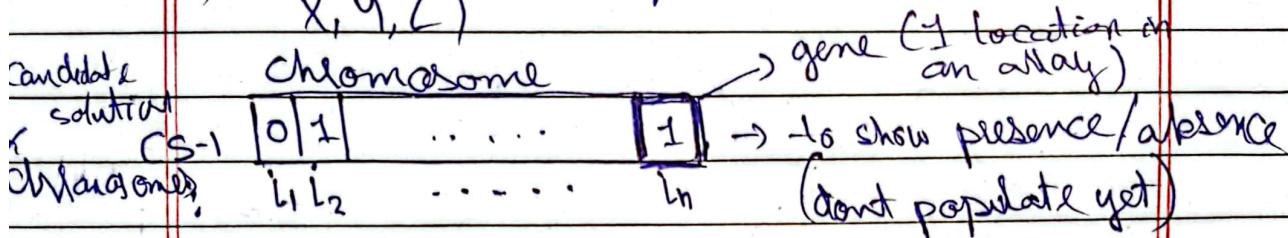
Date: _____

Genetic Algorithm

- inspired by evolution of living things.
- not just biologically inspired (like neural network)

1) Representation of problem solution
(What is the format of solution?)

X, Y, Z



CSN → we look at the army of solutions.

timetable
ki sol kai?
table,
visual representation

The solution isn't proposed, a set of
solution is proposed.

veghair sochay canhay initializing karin

CS-1 CS-2
Instead of following
one path, we follow
many paths!

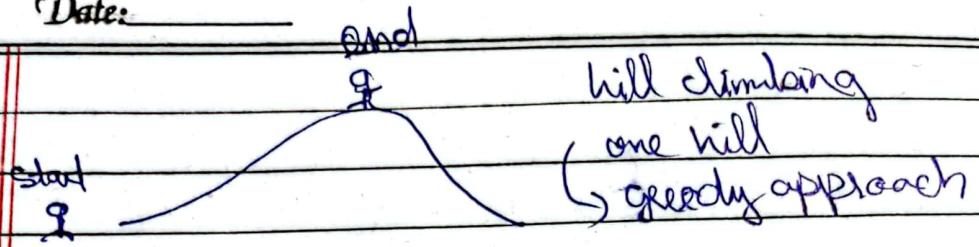
initialized

so many solutions,
parallel computing

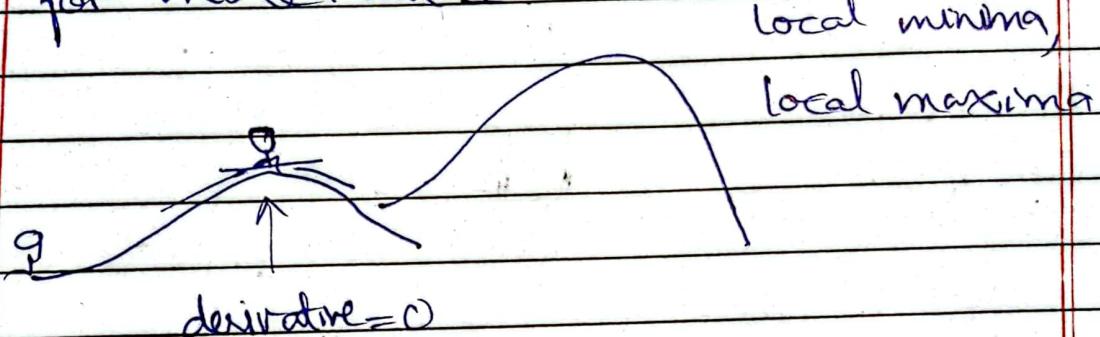
survival of the fittest

represent n solutions

Date: _____



You can always climb a hill (single) using greedy approach but not optimal for more than 1



Kya optimal timetable hai? we can try our best, it is NP hard problem
↳ factorial complexity

TSP: had city ka airfa visit kar kai apna starting node pe terminate karega, and best waal hogya jo min cost.
→ can't guarantee optimal sol.
→ $(n-1)!$ time complexity
→ circuit board par bhi apply hota.

minimum spanning tree (like TSP)
but doesn't end on start node

Date: 22/10/24

quiz 2

Thursday
24th Oct

monotony
self oriented
till alpha
beta learning

Genetic Algorithm

read 12.1 article book

→ not for normal problems, it is for
NP hard problems

Knapsack problem

optimization problem

Knapsack hai articles iss tarah rakhna keh
capacity mein rakhay huy rakhne ka bag mil
itney wajah ka rakhna ka ussay ziyada whi.
Bag nahi khakab karna.

→ 0-1 Knapsack problem

yes or no, lein gawey ya nahi. Apni
Space ko optimize karna

→ Fractional Knapsack problem

- immovable things (no share)

-> You can add things in fractions as
well, not necessarily complete (Powered Form)

→ can be solved using greedy
approach (optimal sol) for knapsack

→ powder, dust

→ optimal sol by greedy approach

→ hard

0-1 Knapsack can be solved using
greedy approach but not optimal,
but fractional knapsack is
optimal by greedy approach

Date: _____

kitni nodes by mini maximax &
alpha beta pruning?

26 15

 $\frac{15}{26}$

best case : $S_E = 2b^{\frac{d}{2}} - 1$ (even
ply depth)

no. of nodes
examined / checked / generated

$$S_O = b^{\frac{(d+1)}{2}} + b^{\frac{(d-1)}{2}} - 1 \quad (\text{odd ply depth})$$

$$\mathcal{O}(b^{\frac{d}{2}}) = \mathcal{O}(\sqrt{b})$$

$$S_E = 2b^{\frac{d}{2}} - 1$$

$$S_O = b^{\frac{(d+1)}{2}} + b^{\frac{(d-1)}{2}} - 1$$

best case = $b^{\frac{d}{2}}$

worst case $\frac{1-b^{d+1}}{1-b}$

Date: _____

$$\max - A(d, B) = (-\infty, +\infty)$$

$\max(-\infty, 8) = 8$
 $(8, +\infty)$

$$\min - B(d, C) = (-\infty, +\infty)$$

$(-\infty, 8)$
 $\min(8, 9) = 8$

$$C(8, +\infty)$$

$$\max - E(d, L) = (-\infty, +\infty)$$

$\max(-\infty, 4) = 4$ or $(4, +\infty)$ like
 $(8, +\infty)$

$$L \quad 4$$

$$M \quad 8$$

No label of play depth

normally $\rightarrow (\alpha < \beta)$

pruning tab hogi jab ($\alpha \geq \beta$)

$$E(-\infty, 8)$$

$(9, 8)$

X

$$N \quad 9$$

X not generated
O (just for reference)

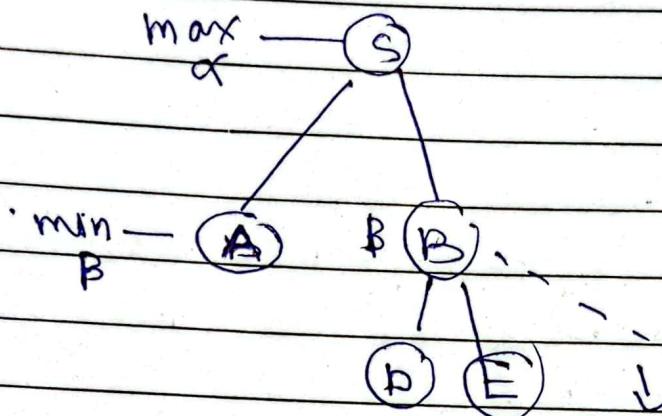
$$O(bd) \rightarrow O(Tbd)$$

abhi bhi linear hi hai

Q. given tree, alpha
beta pruning
implement karta
d, B ki final value
beta in paper mein

Mini Max ko alpha beta pruning
seh kaise implement karen

(S) root node keh saaybachay
generate hota, since wo leaf node
nhi thaey.



$$\alpha(i) \geq \beta(j)$$

pruning condition

wasn't generated

since it was
so already,
min less than
A's min

min-nodes: $\beta(j) \leq \alpha(i)$ **d-cut off**

Max-nodes: $\alpha(j) \geq \beta(i)$ **\beta-cut off**

alpha cutoff on min nodes

beta cutoff on max nodes

associate alpha, beta for each node

α ↗ inc monotonically
(non-decreasing)

β ↗ dec monotonically
(non-increasing)

Date: _____

Singular extension heuristic

→ aik individual node ko further down tak check kartaayi:

* phlephth kesaay kartayi & tracking kesaay
Karma

Mini Max is exhaustive search

agar bfs way mein chalataayi tah

mem: $O(b^d)$

time: $O(b^d)$

agar dfs way:

mem: $O(b^d)$

time: $O(b^d)$

Node S keh perspective seh dekhna:

B ko further explore karne useless,
since wahan adversary ka node, wo hamesha
apko kam peh hi le kar aaye ga.

A → rain is 100

B → min 20 → don't explore
further.

Deep blue: Besh futher dekhta tha.

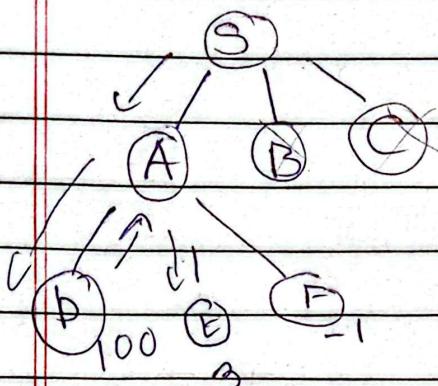
5 depth alpha beta pruning
max min

Date: 17/10/24

AI

agar sirf aik bacha daalein stack mein & explore to depth, then bad meen sibling generate karne.

* Know the diff b/w
this func logic &
DFS ka normal logic



$$A = \min(100, 3, -1) = -1$$

$$B = \min(9, 1) = 5$$

$$C = \min(2, 0) = 2$$

→ B is best, maximum advantage

plywood

Trade-off of More ply depth:

→ more memory usage

→ Ziyada achi analysis

→ time ziyada leta

horizon effect

→ can't see beyond the ply depth

Quiescent states: jismein aik dum
Seh pieces kam ho jaen aur ziyada

Date: _____

$$f \ g(n) = 0$$

$$f(n) = h(n)$$

best first search

jon silk distances ki basis par

chaltay hain algorithms ko

greedy algorithms kehtay hain-

irrevocable: hill climbing algo

greedy algo cannot guarantee
optimal path (they don't back
track and are irrevocable).

(Syllabus of quiz 1)

mini max algo

To find ' G_2 ' first

$$f(n) > f(G_{1,2}) = c_2 \quad (\text{ii})$$

$$c_2 < f(n) \leq c_1 \quad (\text{from (i) \& (ii)})$$

$\rightarrow c_1 > c_2$ (by transitivity)

But $c_2 > c_1$

\therefore our assumption is contradictory.

\therefore Therefore A^* is always optimal.

$$f(n) = g(n) + h(n)$$

If $h(n) = 0$

$$f(n) = \boxed{g(n)}$$

if not constant
it is uniform
cost search
Dijkstra's algo

↓
Branch \& bound
If constant, it is

is BFS

Date: _____

rule of induction :

deduction is based on logic,
so it can't be wrong.

flawed reasoning

- we can infer, it is all about
ML, and it is all about capturing
intelligence

best-first search is a framework
heuristic O Kadar to bfs
bam jaye ga..

? jiski cost
3iyada

Optimality of A* (formal proof)

proof by contradiction :

Assume that A* finds a
suboptimal goal before it hits the
optimal one.

$$f(G_1) = c_1; f(G_2) = c_2 \quad \Rightarrow \quad c_2 > c_1$$

suboptimal

$$f(n) \leq c_1 \quad (\text{by admissibility})$$

①

$$h_1(n) \subseteq h_2(n) \subseteq h_3(n) \subseteq h_4 \subseteq h^+(n)$$

Date:

M	T	W	T	F	S
---	---	---	---	---	---

$h_2(n)$ $h_3(n)$ $h_4(n)$

euclidean

$h(n) = 0$

$h^+(n)$

overestimate

underestimate

↑
no. of nodes explored
increases.

→ koi bhi solution explore

Kar saktay, but kare nodes

Ziyada bhtar toh seh se

$h(q)$

Informedness :

it is more informed

$h_1(n) \subseteq h_2(n)$

bigger

is better in

the underestimate

memory

takes less space

Now:



koi aur heuristic karna saktay

jai manhattan distance seh

kisi ziyada informed hoy?



Date:

Admissible

oracle } bata raha

admissibility:

$\left\{ \begin{array}{l} \text{genuine lab min. cost} \\ \rightarrow \text{actual path ki} \\ \text{actual cost} \end{array} \right.$

$$h(n) \leq h^*(n)$$

heuristic always underestimate

Now,

\rightarrow it is always an underestimate
 of the ^(actual) cost to the goal.

overestimate kyun hi kartay?

Kapdon ki sale

underestimate hoy

toh ziyada
 Sastay karta
 ke saktey

overestimate hoy

toh ojri mehnga
 hi milegi

how will we know our heuristic
 underestimates?

problem relaxation problem

constraint ko follow kartay huay

heuristic underestimate hi hogi
 heuristic design

problem ki restriction ko follow
 ya reby. karta won't be admissible
 heuristic

Date: _____

 $\underline{5}, \underline{5}, 5, 6$ ↑
top

agar $h(C_n)$ zero toh hum goal state
dekh pocheinch gaye.

 $f(n)$

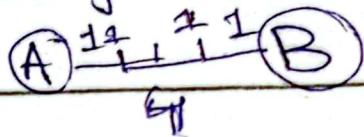
total evaluation function

problem relaxation problem

Date: _____

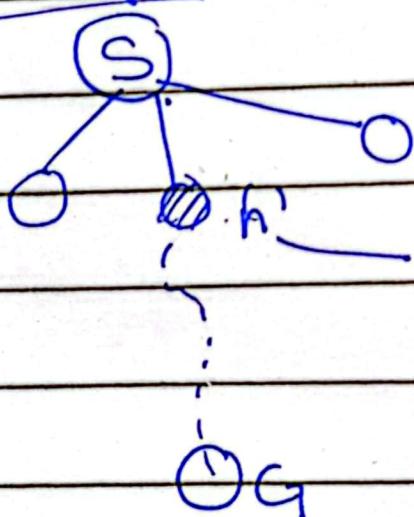
bfs, dfs \rightarrow for unweighted graphs

but wrong around?



weighted graphs \rightarrow in assignment
cost in 8 puzzle moves

Ignore the weights

A* algo :

$$g(n) = 2$$

$$h(n) = 4 \rightarrow \text{eg (estimate)}$$

$$f(n) = 6$$

$$g(n) + h(n)$$

\hookrightarrow also always equal to

toh A* algorithm degenerate ho!
 $\overset{\text{zero}}{\text{Kai bfs bari jaye ga}}$

Kai bfs bari jaye ga

Lecture #5

mn
max
algo

A* algorithm (Best first search + admissible heuristic)

→ most optimal path

→ complete algorithm

→ competitor to Dijkstra's algo

→ uses admissible heuristics

$$f(n) = g(n) + h(n)$$

l , true / actual cost → estimated cost from node n to the goal

from start node to ' n '

Algorithm

→ best first search + not admissible search

= A* algorithm

→ selective search

- A star is optimally more efficient than Dijkstra's algorithm

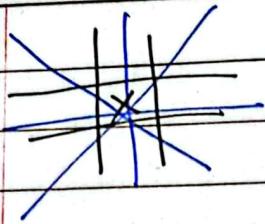
- $f(n) \rightarrow$ gives total evaluation

gives estimate of distance b/w Start state & goal state

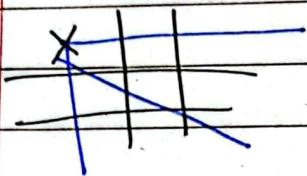
Start state : $g(n)=0$

goal state : $f(n)=0$

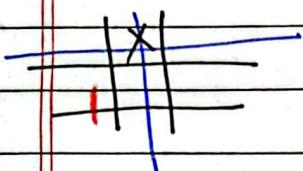
Date: _____



④ possible successes



② possible successes



② possible successes

The no. of winning lines you have available

VS

The no. of winning lines your rival was available

agai deficit mein toh select the less deficit ones

Yeh heuristic negative values bhi de sak�ay.

quiz next lec: (3-4), slides

next lab → heuristic function use along with priority queue.

→ blind search implement karna hoga next lab me.

Assignment

imp → output of pseudocode, code, descriptive, main function
Sub a'sakti quiz, mein → A* → be an underestimate
normally backtrack

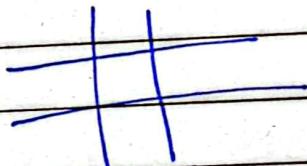
Date: _____

obstacles ziyada hi hon.
trail will be zigzag.
→ Optimal path kyun hogा? kyun
keh wao hamesha heuristic
underestimate dega, robot will try
since uska heuristic hamesha hi
underestimate hogा.

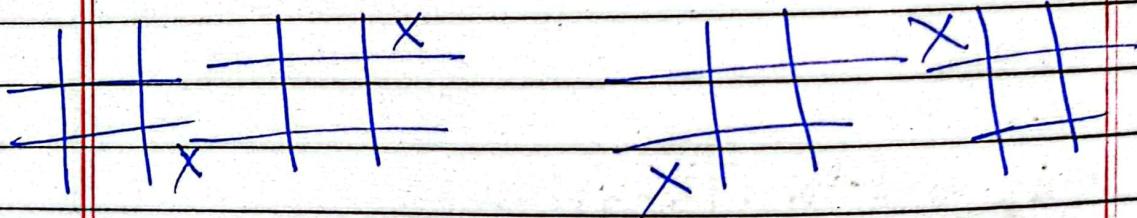
Zero-Sum games

→ profit at the expense of someone
else's loss.

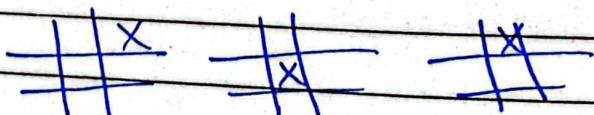
tic tac toe



states 9! ban sakti at max,
but ismein 2 log hain, alternatively
dono chalen gaay, teh it is not
the same as 8 puzzle problem



yeh chaalon baabbar hain,
since we only need to complete a
line of 3.



3 possible states

Date: _____

Best-First Search

→ priority queue used here.

priority kiski?

→ from heuristics, the number comes

- Kaise implement karte hain priority queue?

tree, heap,
min max heap.

- linked list seh bhi ka saletay, make priority, sort on the basis of number associated

- top per hoga woh path laya ga

ab!

Comparison with Uninformed search :

→ won't exhaust all possibilities

→ won't be exponential in memory

→ time in the worst case, at least is exponential unfortunately

→ optimal search?

→ optimistic path deti,

heuristic apka underestimate hona chahiye

Robot in a maze

heuristic kaise banayein gaay?

→ distance (SLD) maintained rakhay

→ jis state mein kam distance hua uss state mein isse rakhne

Date:

(1)	2	8	3		1	2	3
(2)	1	6	4		8		4
	7	5			7	6	5

out of place tiles = 4

somewhat 4 moves
but distance per
move

variable → nested loops in x direction
in y direction

underestimate
of the actual
moves

heuristic function

blank is not a tile

heuristic

* $h(n) = \text{sum of distances out of place} = 5$

manhattan distance

$x_1 - x_2$ coordinates

$y_1 - y_2$ coordinates

Euclidean heuristics distance

$$d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

+ve real number

euclidean will be lower than
manhattan distance

underestimate

$e_n \leq a_n$

estimated
number of
moves

actual moves

Date: 3/10/24 AI Lecture #5

(till today syllabus) quiz in next lec
descriptive quiz & assignment
MCQs

Ch 4 → heuristics search

Rule of thumb rules jo overtime seekhay

→ doctor keh paas jataay woh ha test perform nahi kartaay, mechanic bhi saari cheezain possibilities nahi dekhti.

rule of thumb → ml hi hai, machine khud seh pick up kerti.

Once you provide it with heuristic & provide a method to use it, it will do selective search.

$h: b \rightarrow \mathbb{R}$
board real
ki to number
state

per map karti

* be able to write a heuristic function.

Heuristic → is a rule of thumb.

→ it is not supposed to be right always, it can make mistakes.

Date: _____

$$N = b^0 + b^1 + b^2 + \dots + b^d$$

$$N = (d+1)b^0 + db^1 + (d-1)b^2 + \dots + 2b^{d-1} + 1b^0$$

IDFS

itni baar generate hua hai

$$N_{BFS} = b_0 + b^1 + \dots + b^d$$

→ Sif constant ka falg hai tab
itna koi falg nahi hai

$$N_{IDFS} \rightarrow O(b^d) \rightarrow \text{time complexity}$$

$$N_{BFS} \rightarrow O(b^d)$$

time \propto no of nodes

IDDFS → DFS hi hai but level save
nhi karey,

have to store depth info

global variable hogा jou

persist karey ga across
iterations

heuristic search

hill climbing

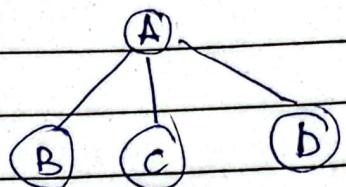
gradient descent

gradient ascent

* Red physical system hypothesis

analytically can't prove

→ empirically prove ho rahi



generate & test goal hai kuch hi

DFS
BFS

incomplete search
complete search

Space complexity

linear vs
exponential

DFID / IDS

depth-first with iterative deepening
iterative deepening search

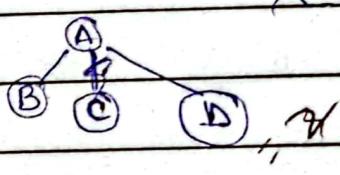
IDS

Space
 $O(b^d)$

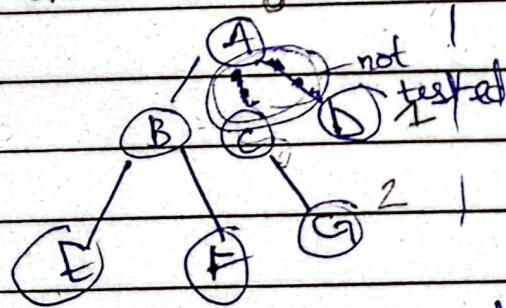
Time

$O(b^d)$

IDS
 $d=0$ A : $d=1$
 \cancel{X} discard



$d=2$ depth 0 1



$1+2+3$, n^b

2

hal tution man chal level
by level hi rakhay hain

dekhda kyun bana rakhay? wasteful

manzil aa raha hai but we want to
avoid storing/occupying extra memory

every problem is not solvable (halting problem)

DFS is very time efficient (linear behaviour)
memory

BFS se path chondray keh liye predecessor
list seh kartay rahan.

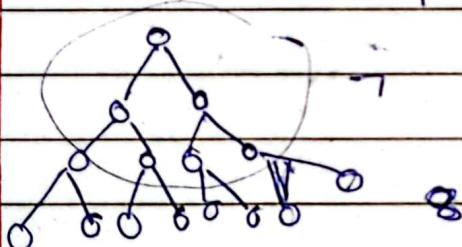
khatay nodes hain:

$$\hookrightarrow N_t = N_0 + N_1 + \dots + N_d \quad \text{no. of nodes at depth } \dots$$

$$N_t = b^0 + b^1 + b^2 + \dots + b^d$$

$$N_t = \frac{1 - b^{d+1}}{1 - b} \quad \begin{matrix} \text{total no. of nodes in a} \\ \text{tree till depth } 'd'. \end{matrix}$$

$$\frac{1 - 2^{2+1}}{1 - 2} = \frac{-7}{-1} = 7$$



chess \rightarrow branching factor b/w no search

ak level add karday toh pick by coarray
levels keh stem seh ziyaada nodes hain.

No. of nodes grow exponentially.

BFS \rightarrow hamisha optimal output degi.

You have not left any stone unturned
BFS is a complete search, agar ha ga
goal toh BFS chondray lega

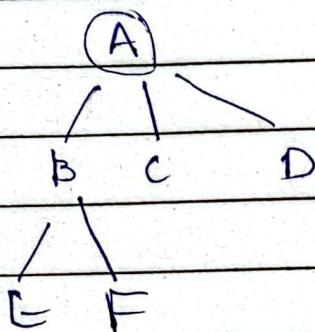
DFS, agar levels ziyaada toh guarantee nahi
hai keh goal mile ga ya nahi, kam levels
mem no issue

Date: 10/24

AI Lecture #4

Informed

backtracking search is essentially DFS search



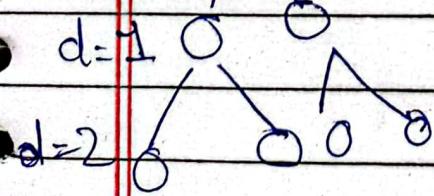
open	closed
A B C D	[A]
A B C D E F	BA]

branching factor
 $\frac{3+4+5}{3} = \frac{12}{3}$

avg branching factor

	Space	Time
BFS	(optimal complexity) $O(b^d)$	$O(b^d)$
DFS	(Memory efficient) $O(bd)$	$O(b^d)$

$d=0$ depth level



Date: _____

Historic person hai aapka ancestor,
which direction will you check?

From you to historic person or historic
person to you?

→ First one as bf sirf 2 hai, while
in the other case it could be too much,
unkay descendants bohot hotay hain
toh agar bf bohot ziyaada ho sakte.

2 types of search:

- 1) data-driven search (forward chaining)
- 2) goal-driven search (backward chaining)

→ disease diagnose, saalay test kawao, dekha kya hua wa

→ starts off with the goal, kis taregay seh goal password.

Doctors use goal-driven reasoning.

Kya malaria toh nahi? Symptoms kya hain?

→ less time consuming

Branching factor: aik node keh kitney an average children hain.

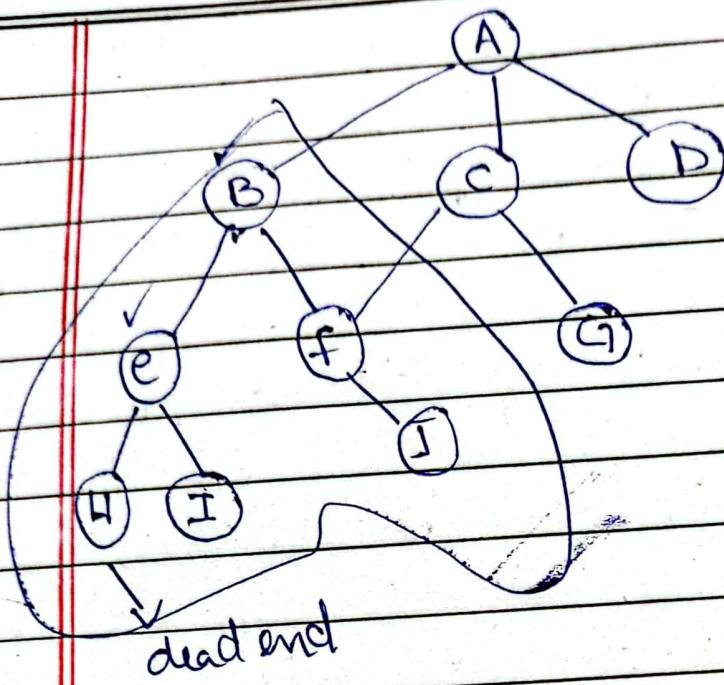
→ Baaz problems mein k states like 8-puzzle problem.

$$\text{avg bf} = \frac{4+3}{2}$$

goal driven

Kya aap start state seh goal state per jaana chahen gaay ya goal State seh start state jaana chahen gaay? Done hi done in case of 8 puzzle problem.

Date: _____



C mein jaakay aap F create Karen
gaay but won't add it.

$A \rightarrow C \rightarrow G$

NSL, SL \rightarrow stack

Hamen graph diya jata hai, toh kya
yeh graph hamen bhi pehla bana
banaya milta?

\rightarrow Nhi hamne graph nahi banaya,
explicitly nodes create nhi kiye, implicitly
define kar rahan graph.

\rightarrow Actual creation nahi kartay graph
keh, possibilities ka tree bana rahay
and checks lagatay.

Inversion as a state space graph

Date: _____

In algorithms where you can backtrack it is called an revocable algorithm.

BFS → complete algo

DFS → incomplete algo

Backtrack, DFS → behan bhai

Systematic → not random

1

Backtracking

3 lists that keep track of algo:

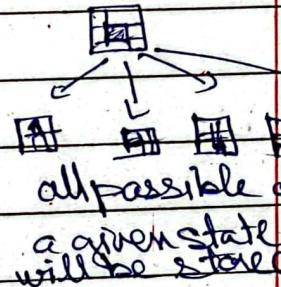
1) SL (state list)

2) NSL (new state list)

3) DE (dead ends)

node under examination 4) CS (current state) pointer
we check yeh goal hai, pointer
hai yeh

start state mein multiple nodes kisi hou
saletay.



Date: _____

Blind Search Algorithm

→ Systematic search karni, problem-solving karni without explicit problem solving knowledge.

→ Brute force hi hogi

→ Uninformed search

heuristic: explicit problem solving knowledge.

4 algos:

1 & 2) Backtracking & DFS

→ mein path bhi store kar raha hoga

hotay but dfs men whikata.

3) BFS

→ orthogonal

4) DF with iterative deepening

Search strategies:

(i) Complexity $O(C)$

search algorithm maximum kitna time ya space le sakte.

(ii) Completeness

A gal goal exist karta, complete algorithm guarantees keh goal achieve karay ga.

(iii) Optimality

Algorithm that returns best solution, best is what you define, like in terms of cost, moves according to your criteria.

(iv) Irrevocability

An algorithm ^{in which} you can't undo or backtrack.

e.g.: Hill climbing search
is irrevocable

(weight search)
edge

Date: 26/9/24

AI Lecture #3

2 things required for intelligent problem solving:

1) representation (Knowledge representation) KR
2) Search

you need a formal structure, chahiye structured
unstructured db, but you need it in some
form.

→ Sequence of actions we perform to reach
our required goal.

In 2 cheezon ko interplay algorithms
kehلاتا ہے۔

Random search → koi random permutation aa jaye
but woh goal state tak search random

Systematic search → namayeh karen gaay

DFS (page by page search)

heuristic in DFS in nature

Kabhi kabhi instead of random search
bhi karta

Weak AI follower \rightarrow sr

Itni bhi AI tarazu karay, insaan
hamesha upar hai.

Physical symbol system hypothesis:

Physical symbol system ^{eg computer}
has necessary & means for sufficient
means for general intelligent action.

Modern AI

8 puzzle

2⁸

3x3 puzzle hai 2^9 tak kitni
 $9!$ = combinations possible.
there are possible $9!$ permutations.

OOP object class \rightarrow puzzle state ki

graph mein nodes & edges hain
search tech andai implement karna

When you search a graph:

you actually develop or generate
on the go, a search tree

control repetitions through
parity checks.

Phir ismein intelligence kaisey?
exhaustive hai yeh, toh
aap heuristic search karein gray,

Date: _____

physical system hai yahan per?

→ computer is a physical symbol system.

→ algebra, logic

→ input hain, output hain

→ Insaan aik ~~symbol~~ ^{Physical} system hai.

but kya ^{Physical} symbol system hai?
mean?

Insaan aik physical system but kya
woh aik physical symbol system
hai?

neurologist have shown keh
kay memory store hoti, konsay regions
kya store karta, language area.

Brain cannot feel pain, surgery brain
ki jaagay huay hoti hai

→ Insaan physical symbol system hai.

emotions kya?

There is more to humans then physical
symbol system that is a sense of
spirituality.

Date:

Sentience

Weak AI

John Seral

Mind, brain, program

robottal 1980
tuung test



chinese room,

doesn't know chinese, but
resources hain, wah kousa
Samjh lek koota.

He thinks kabhi bhi

Strong AI nhi ho sakte

chatgpt bhi pattern matching
koota

System

Physical Symbol Hypothesis

It's just a hypothesis, a statement that
can be proven true or false

for solutions

Intelligence = KR + search

1.
using
symbols

Symbol → nishan / representation

woi language use karein

Date: _____

Kya tuning test reliable hai?

intelligence cannot be defined
specifically

self-awareness kya hai?

[chatgpt keh context mein se�xious?]

[Sentience] - cognitive state jismein aapkay
pass self-awareness hai.

STRONG AT

Machine = Human

Sentience

mainning self-awareness hai, aapko
environment, aapka bolay mein jaati.

WEAK AT

AI = helpful tool for humans

dirty bombs
smart bombs

Date: 24/9/24 AI Lecture # 2

AI - big body of knowledge

Ch 3 → seh start

AGI

node 2 Partay class

AT Ko computational point of view seh path rakhay
main focus: understanding
no definition in exams like AT

Turing Test kya tha?

Interrogator
I ♀

H ♀ C ♂

text ke question

Utkhma, aur text per reply.

Pehchaana kahan computer konsa hai.

Alan turing:

Machine ko pehchaan nahi saken
galey toh yeh intelligence hai

Date: _____

clustering reinforcement learning
practical applications.

Is chatgpt really intelligent?

↪ how can we check

Turing test

judge kah human kah machine, key answer
test

Dr. Hugh Lebow

ILMs aamay kah baad that turing test was
done.

There are many other turing test.

movie : A beautiful mind

ch2, 3 seh baat shuru hoti course mein

DFS, BFS karkay aao

19/8/24

Advanced Stats
Lecture #1

Assignments:

• Research paper relevant etc

Date: _____

chair

Konsi fursi, how do you class

-ify?

how you do that? you don't know.

how machines does that? you should know.

yeh billi kesa yahi?

Alan Turing

John von Neumann

Arthur Samuel's checkers program 1958? C

Perception couldn't do XOR (at that time?)

back propagation algorithm

MLP

ask neuron

ask perception or memory

Will shift to machine learning in this course

In this course → conventional machine learning

→ conventional AI

feature engineering

Deep networks

Q: Kisi peh bandai, kya
A: yeh kisi hai?
no

AlexNet no need to specify bands
convolution network

watch: video of

AI chatbots can converse with each others

automated reasoning

evolutionary approaches
to learning

heuristic knowledge

not deep learning in this
course.

Date: 19/8/24

AI

Lecture #1

Computer Core Course

↳ CS-203

$$(3+1=4)$$

Sentience

15 marks → quiz

5 marks → sem proj

2+3

↓
Assignment

Class participation

Prerequisites: OOP

Objects not numbers but a combination

Characteristics of Intelligence:

→ problem-solving

→ learning to make better decisions

called AI (John McCarthy)

1956

list

lost

in
stup
parenthesis

Inventive: 1942 ↓ 2024

artificial
method known
as

ability to learn

AGI

e
n
a
r
t
a
l

Artificial Super Intelligence

Super Intelligence

Micron (Expert system)