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Reference Books

1. Artificial Intelligence

by Stuart Russell

Peter Norvig

Computer Science (Domain)

→ Artificial Intelligence (sub-domain)

→ Machine Learning (sub-domain)

→ Data Mining

→ Big Data

Artificial Intelligence

(not natural/regular)

- Man made thinking like a human;

- can be device or model of man for AI +

■ except AI, all → pre-programmed based

■ Turing test: is a method of inquiry in AI for determining whether or not a computer is capable of thinking like a human being.

(Alan Turing)

* AI → It is a branch of computer science by which we can create intelligent machine which can behave like a human, think like human and able to make decision.

- thinking humanly
- acting humanly
- thinking rationally
- acting rationally

- How do human and animals think and act?
- How does language relate to thought?

* In AI, you do not need to pre-program machine to do some work.

* In AI, you have to create a machine with programmed algorithm which can work on own intelligent.

■ Why AI?

→ With the help of AI, we can create such something/solution/software or device which can solve real-world problems very easily and accurately such as - health issues, marketing, traffic issues.

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- * With the help of AI, you can create your own personal assistant such as - google assistant, siri, etc.
- * You can build such robot that can work in an environment where survival of human can be at risk.
- * AI opens a new path to new technologies, new devices, and new opportunities.

Computer program -

Machine language - programming -

Human language - programming -

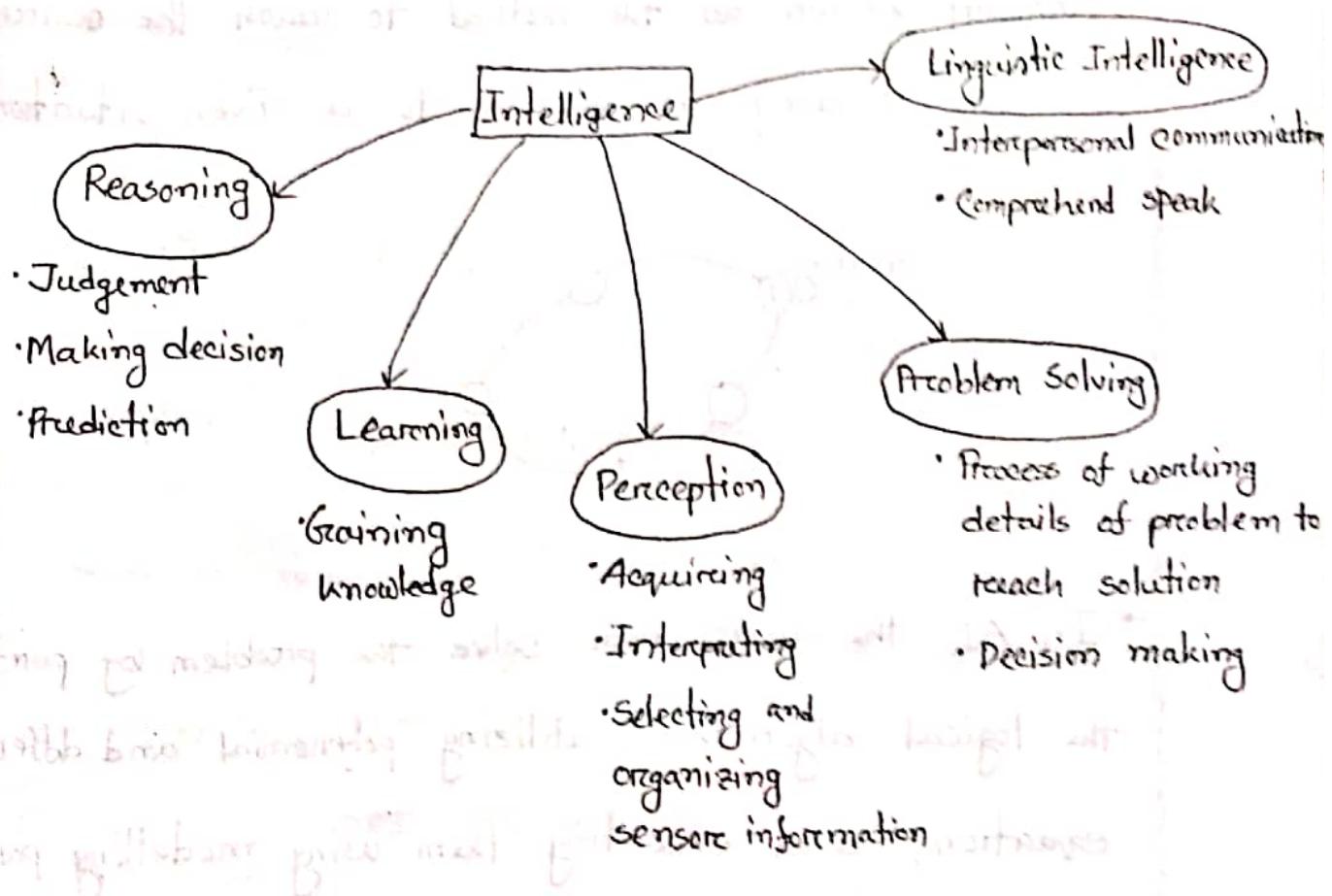
Machine language -

Human language -

Goal of AI

1. Replicate human intelligence
2. Solve knowledge intensive task
3. An intelligent connection of perception and action
4. Build a machine which can perform tasks that require human intelligence, such as -
 - providing a theorem
 - playing games (chess)
 - performing surgical operation
 - driving a car in traffic
5. Creating some system, which can exhibit intelligent behaviors, learn new things by itself, demonstrate, explain and advise to its users.

■ Intelligence is composed of:



Basic →

i) Definition

ii) Why AI

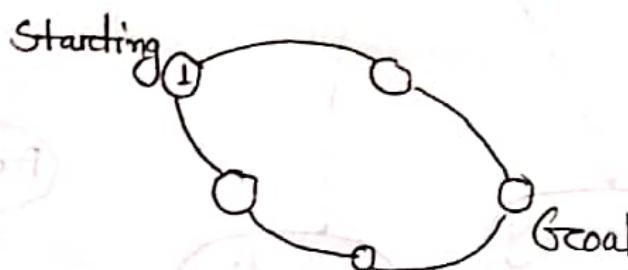
iii) Goal of AI

iv) Intelligence is Composed of

Part II

Problem Solving:

Commonly known as the method to reach the desired goal or finding a solution to a given situation.



* In AI, the users can solve the problem by performing the logical algorithms, utilizing polynomial and differential equation, and executing them using modelling paradigms.

* To do this, one need to define problem statement and generate the solution.

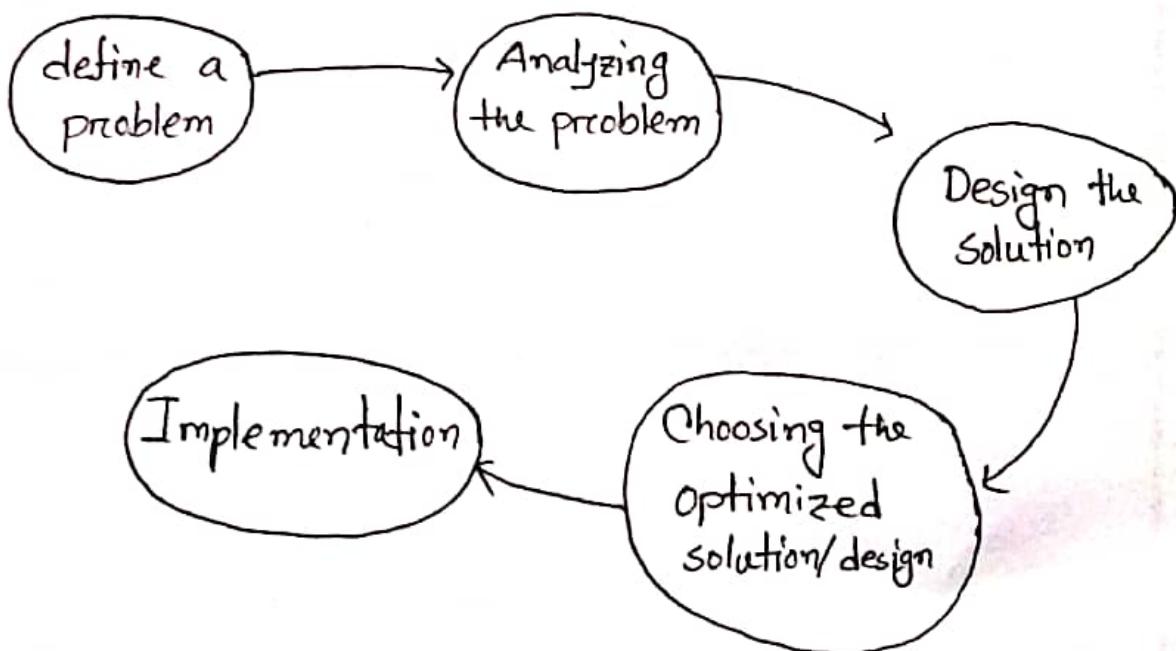
* Some of the most popularly used problems that are resolved by AI:

- Some kinds of games including chess
- N-Queen problems (puzzle)
- Travelling-Salesman problem
- Tower of Hanoi problem

* Problem solution by searching

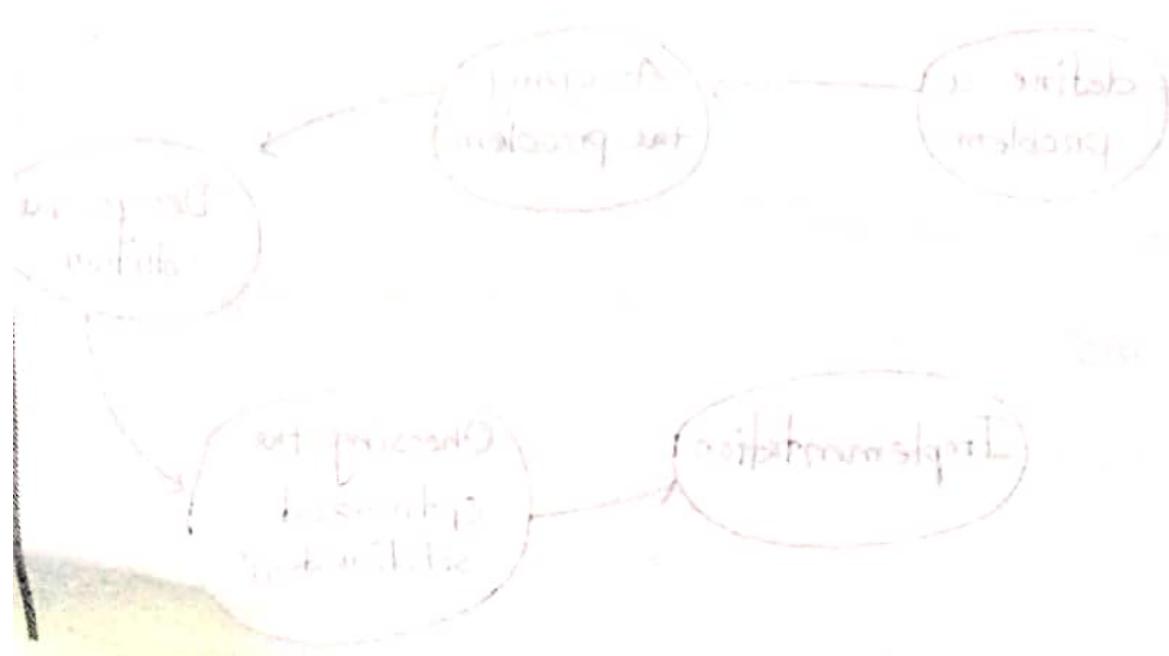
- finding information one need
- Most commonly used technique of problem solving in AI

* The process of solving problem



Properties of Searching Algorithm:

1. Completeness: It is said to be complete if it is guaranteed to find the solution if at least any solution for any random input.
2. Optimality: If a solution is found to be best solution (lowest-path-cost) among all.
3. Time complexity: It is a measure of time for an algorithm to complete its task.
4. Space complexity: How much memory is needed to perform the task.



Chapter 9

Types of Searching Algorithm:

- Uninformed Search (Blind Search)
- Informed Search (Heuristic Search)

Brute-Force Search - traditional

Blind Search algorithm → Brute Force Algorithm

Uninformed Search: Some important information are not given (no additional info; no domain knowledge)

- BFS (Breadth-First-Search)

- DFS (Depth-First-Search)

- Uniform Cost Search

updated - Depth Limited Search

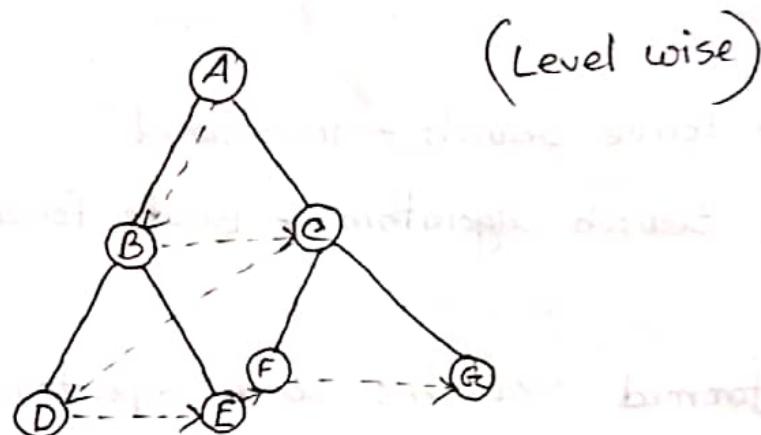
updated - Iterative Deepening Search

Informed Search: Directed search; heuristic → extra knowledge

- Greedy Breadth-First Search

keep list of stored frontier to stop searching

BFS: root node expanded first then successors respectively; Graph based; FIFO queue, sequence from left.



Traverse: $A \rightarrow B \rightarrow C \rightarrow D \rightarrow E \rightarrow F \rightarrow G$

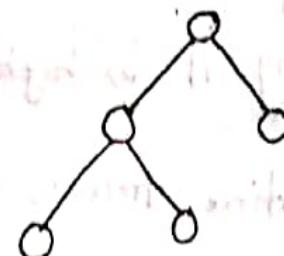
- Advantages of BFS:
- One of the simplest search strategies
- Complete. If there is a solution, BFS is guaranteed to find it
- If there are multiple solutions, then a minimal solution will be found
- The algorithm is optimal if all operators have the same cost. Otherwise, BFS finds a solution with the shortest path.
- Finds the path of minimal length to the goal

* Disadvantages of BFS:

- Consumes large memory space. Its time complexity is more.
- It has long pathways, when all paths to a destination are on approximately the same depth.

* Properties:

- Complete \rightarrow Yes (If b is finite)
- Optimal \rightarrow BFS, all actions have the same cost.
(cost = non-decreasing function)
- Time Complexity \rightarrow



d = depth
 b = branching factor

$$\text{Big } O \rightarrow O(b^d)$$

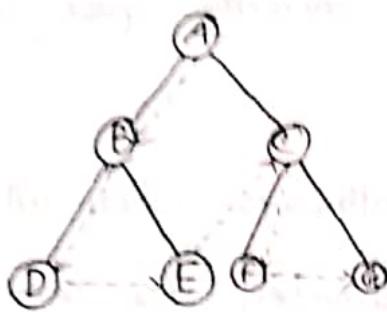
- Space Complexity \rightarrow

$$\text{Same} \rightarrow O(b^d)$$

BFS / DFS

- Searching Strategy
- Property
- Comparison

* DFS: Traverses a graph in depthward motion; LIFO



Traverse: A → B → D → E → C → F → G

* Properties

- Completeness → DFS is complete if the search tree is finite
- Optimality → DFS is not optimal, meaning the no. of steps in reaching the solution, or the cost spent in reaching it, is high.
- Time complexity → If the entire tree is traversed, $O(v)$
In case of a graph, $O(v+E)$
(V = no. of vertex; E = no. of edges)
- Space complexity → $O(h)$; h = maximum height of the tree.

Q) Problem Definition and Formulation

Agent - Perceiving information through sensors from environment.

Goal-based agent - A goal-based agent takes it a step further by using a goal in the future to help make decisions about how best to reach that outcome.

* A problem can be defined formally by 5 components:

1. Initial State: This stage state requires an initial state for the problem which starts the AI agent towards a specified goal. In this state new methods also initiate problem domain solving by a specific class.

2. Action: This stage of problem formulation works with function with a specific class taken from the initial state and all possible actions done in this stage.

3. Transition: This stage of problem formulation integrates the actual action done by the previous action stage and collects the final stage to forward it to their next stage.

4. Goal Test: This stage determines that the specified goal achieved by the integrated transition model or not, whenever the goal achieves the stop the action and forward into the next stage to determine the cost to achieve the goal.

5. Path Costing: This component of problem solving numerical assigned what will be the cost to achieve the goal. It requires all hardware, software and human working cost.

Optimal solution: is one whose measure of quality is close to the best that could theoretically be obtained.

Frontier → a set of leaf nodes.

Loopy path → infinite, unsolvable.

Explored set → keeps track of the nodes that have already been expanded.

Uniform Cost Search

UCS is an uninformed search algorithm that uses the lowest cumulative cost to find a path from the source to destination.

Nodes are expanded, starting from the root, according to the minimum cumulative cost. Then it is implemented using a Priority Queue.

✓ how many step - doesn't care

✓ cost should be less

* Properties

1. Completeness: completeness is guaranteed if only if the cost of every step is some positive number.
2. Optimal: UCS is optimal. This is because, at every step the path with the least cost is chosen, and paths never gets shorter as nodes are added, ensuring that

the search expands nodes in the order of their optimal path cost.

3. Time complexity:

Total Solution Cost = c^*

Each step cost at least = c

Steps = $1 + c^*/c$

Time complexity = $O(b^{[1+c^*/c]})$

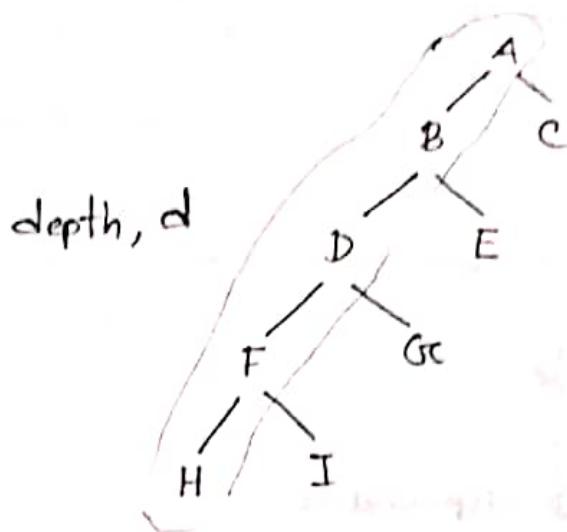
4. Space Complexity:

$O(b^{[1+c^*/c]})$

** Limitations of DFS and how can you address them?

• Inheriting characteristics of search problems where critical issue is life time of tree which is measured in terms of depth of evaluation tree. In terms of tree found after n steps of problem solving makes no sense as n is finite.

* DFS Limitations.



The failure of DFS in finite state spaces can be alleviated by supplying DFS with a predefined depth-limit l , i.e. nodes at depth l have no successors.

Depth Limited Search

- $l < d \rightarrow$ incomplete

- $l > d \rightarrow$ non optimal

- Time Complexity: $O(b^l)$

- Space complexity: $O(bl)$

unbounded length

shallowest depth

Iterative Deepening Search

- repeated iteration ; towards depth;
- gradually increases the limit

* Properties:

1. Completeness : It is complete
2. Optimality : It is optimal if step cost = 1
3. Time complexity : $O(b^d)$
4. Space complexity : $O(bd)$

Bidirectional Search

Two simultaneous searches -

- 1) forward from the initial state
- 2) backward from the goal

Time complexity $\rightarrow O(b^{d/2})$

Space complexity $\rightarrow O(b^{d/2})$

* Comparing uninformed search strategies

Criterion	BFS	UCS	DFS	DLS	IDS	BS
Completeness	Yes	Yes	No	No	Yes	Yes
Time Comp.	$O(b^d)$	$O(b^{d+\lceil c'/\epsilon \rceil})$	$O(b^m)$	$O(b^d)$	$O(b^d)$	$O(b^d)$
Space Comp.	$O(b^d)$	$O(b^{d+\lceil c'/\epsilon \rceil})$	$O(bm)$	$O(bd)$	$O(bd)$	$O(b^{d+1})$
Optimality	Yes	Yes	No	No	Yes	Yes

◻ Informed (Heuristic) Search Strategies

- uses problem specific knowledge
- general approach best-first search

Searching strategy: defined by picking the order of node expansion.

Heuristic function $h(n)$: estimated cost of the cheapest path from node n to a goal node.

$g(n)$

General Best first Search \rightarrow incomplete, not optimal.

A* Search.

heuristic function: is a shortcut to solve a problem when there is no exact solution for it and/or the time to find the solution is too long.

A* Search

minimizing the total estimated cost solution

The Cost:

$$f(n) = g(n) + h(n) \rightarrow n \text{ node to goal node}$$

↓

→ path cost from start node

→ estimated cost of the
cheapest solution through n .

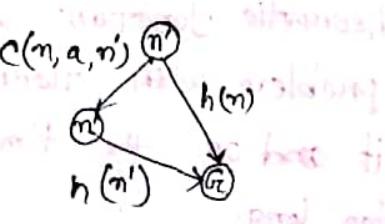
- form of BFS

- It evaluates nodes by combining $g(n)$, the cost to reach the node and $h(n)$, the cost to get from the node to the goal.

- complete and optimal

Admissible heuristic → never overestimates the cost to reach the goal

Consistent heuristic → $h(n) \leq c(n, a, n') + h(n')$



Properties:

1. Completeness: Yes, it is complete

2. Optimality: It is optimal

3. Time complexity: $O(b^d)$

4. Space complexity:

Memory Bounded Heuristic Search

Iterative Deepening A* algorithm

Recursive Best first search

A* Implementation

* Heuristic function $h(n)$ and how to make it

* Heuristic may be:

- 1. $h(n) >$ actual cost (optimal solution is not possible)
 - 2. $h(n) =$ actual cost (best case scenario)
 - 3. $h(n) <$ actual cost (admissible, consistent heuristic)
- Optimality of A* Search Algorithm

** Admissible एवं consistent रूप, किन्तु consistent एवं admissible वाले अलग होते।

Local Search Algorithms

- operate using a single current node
- generally moves only to neighbors of that node

A local search algorithm starts from a candidate solution and then iteratively moves to a neighbor solution.

* Advantages

- use very little memory
- can find reasonable solutions in large or finite state spaces

neighborhood \rightarrow set of all potential solutions that differ from the current solution

* Local Search for Optimization by the minimal possible extent.

$$\arg \min_x f_i(x)$$

$$\text{s.t. } f_i(x) \leq 0, i = \{1, \dots, k\}$$

$$h_j(x)$$

local search is a heuristic method for solving computationally hard optimization problems.

** Beyond Classical Search

Systematic solution / method \rightarrow classical search

* Hill-Climbing Search (8-queens)

- local domain (so knowledge) TTT
 - It is a greedy approach (best move at each step तक तक)
 - no back-tracking
 - also called greedy local search
 - if a state is better than current state then it is new current state

Global maxima \rightarrow target

Local maxima \rightarrow higher peak among the neighboring states

* Problems in hill climbing -

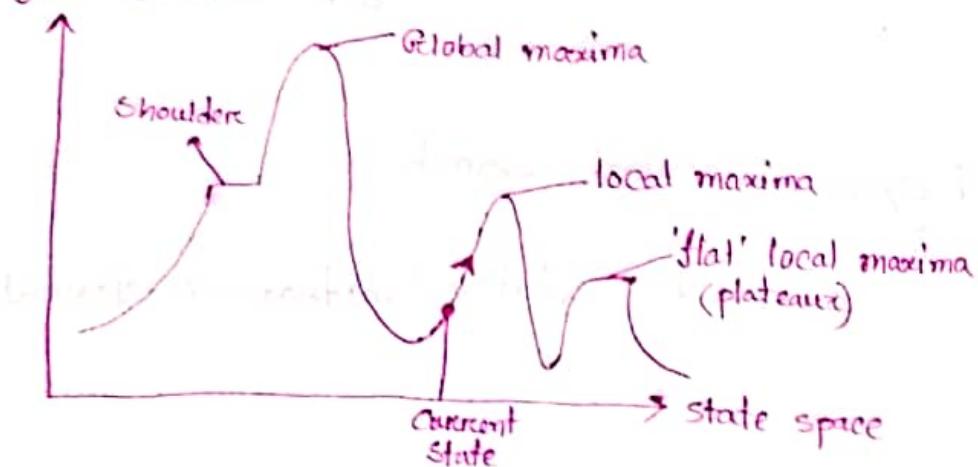
Shoulder - Plateaux

- i) Local maxima
 - ii) Plateau / flat maxima
 - iii) Ridge - direction change

} best move करें
उत्तिम जगह करें

Hill climbing is a heuristic search used for mathematical optimization problems in the field of AI. Given a large set of inputs and a good heuristic function, it tries to find a sufficiently good solution to the problem.

Objective function



Simulated Annealing Search

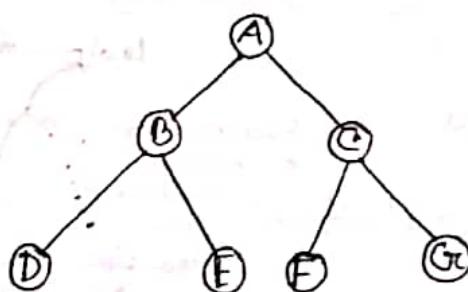
* Challenges of hill-climbing

- never makes "downhill" moves toward states with lower value is guaranteed to be incomplete, because it can get stuck on a local maximum.

- plateau
- ridge

* Random walk - moving to a successor chosen uniformly at random from the set of successors.

- A pure random walk is complete but extremely inefficient.



A B D B E B A C F C G

If it is complete but inefficient as there are **Repeating nodes**

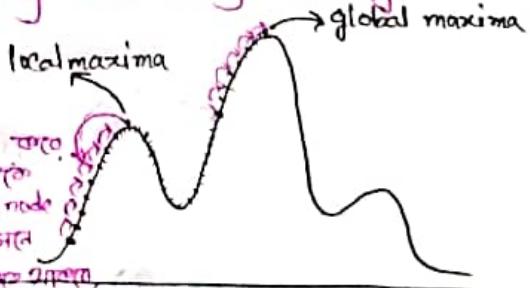
Combining hill-climbing and random walk yields both efficiency and completeness. Simulated annealing is such an algorithm. (allows downward steps)

Annealing - high temperature to gradually low

- instead of picking the best move, it picks random move
- if the move improves the situation, it is accepted.
- otherwise accepts move with probability < 1

improve move
badness move

Annealing is a process where in metallurgy where metals are slowly cooled to make them reach a state of low energy where they are very strong.



$T \rightarrow$ temperature

$\Delta E \rightarrow$ worsen evaluation

gradient ascent

gradient descent \rightarrow minimizing cost

** How simulated annealing works?

** Properties.

Advantages
checks all
neighbors
easy to code
gives good
solution
statistically
guarantees
optimal solution

Disadvantages
slow process
can't tell
whether an
optimal solution
is found.

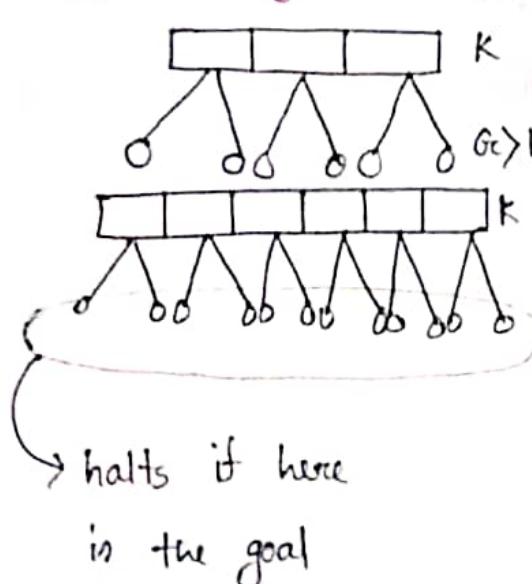
Annealing - Lowering temperature gradually

Quenching - Lowering temperature rapidly

* Local Search with Multiple Present States

Local Beam Search (space complexity related)

- keeps track of K states rather than just one.
- beam width $(B)/n$ is given. (B/n best value memory storage)
- begins with K randomly generated states
- space complexity constant
- At each step, all the successors of K are generated
- $B=1$ ~~is like hill climbing search~~
- if anyone is a goal, the algorithm halts. if $B=2$
- not complete; no guarantee to find optimal solution.



Space का नियम;
no. of nodes का
अंतर पर method का
बहुत ज्यादा रखना चाहिए
Space/time का नियम;
sp

Chapter - 6Constraints Satisfaction Problem

Constraints → restriction, condition

- have alternate ways, but bound to do the task in a certain way.

CSP - to solve a wide variety of problems efficiently.

* Problems have these things-

set of values, variables, values, constraints

** A variable problem is solved when each variable has a value that satisfies all the constraints on that variable is called CSP.

* CSP deals with general problems instead of specific problems; can solve complex problems.

* The main idea is to eliminate the large portion of the search space all at once by identifying the variable combination that violate the constraints.

* Defining CSP

A CSP consists of three components, X , D and C :

X is set of variables $\{x_1, \dots, x_n\}$

D is set of domains $\{D_1, \dots, D_n\}$, one for each variable

C is set of constraints that specify allowable combinations of values.

$$D_i \rightarrow \{v_1, \dots, v_k\}$$

$$C_i \rightarrow \{\text{scope}, \text{rel}\}$$

scope: a tuple of variables that participate in the constraints

rel: a relation that defines the values that those variables can take on.

Assignment →

I) Complete assignment: every variable is assigned

II) Consistent assignment: doesn't violate any constraints

III) Partial assignment: assign values to some of the variables.

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* CSP Example : Map coloring

- region शुल्काके colors कर्वा रखें; adjacent regions or neighbouring regions same colors कर्वा नाहीं

** Why formulate a problem as a CSP?

- Answer:
- i) yields natural representation of problems
 - ii) faster than state-space searchers

* Job-Shop scheduling

Mid Syllabus → Informed search & Uninformed search.

// Characteristics

// Complexity comparison

- * Variations on the CSP formalism
- * Discrete domain
- * Finite domain
- * Infinite domain
- * Continuous domain
- * Linear constraints - each variable appears only in linear form
- * Non-linear constraints - nonlinear variable
 - * Unary constraints - restricts value of a single variable
 - * Binary constraints - relates two variables



Map coloring problem (a प्रश्नापत्रिका)
दो रेजियन समान रंगे हो पायेते
SA \neq NSW

Binary CSP \rightarrow with only binary constraints

- * Global constraints - A constraint involving an arbitrary no. of variables.

↓
Alldiff \rightarrow all the variables must have different values

Example : Sudoku (All row - columns are different)

Cryptarithmic problems

Chapter 7 (Book 2)

Uncertainty is defined as - the lack of exact information or knowledge that hinders us to find correct conclusion.

Probability → Uncertainty; came from random variable.

Random variable → value comes from random experiment

→ Uncertain input (missing, noisy data)

Why Prob Uncertainty?

- Uncertain knowledge (multiple causes leads to multiple effects; incomplete knowledge of causality in domain)
- Uncertain output (abduction-induction are uncertain)

Prior probability → पूर्वानुमान या प्रारंभिक संभिति
 $P(A), P(B), BP(H), P(T) \dots$

Random experiment → Coin toss

Ω = Certain event

\emptyset = impossible event

(Omega) Ω = Random sample space $\rightarrow \{T, H\}$

Exact measurement of probability

IId → Identically Independent

- की अवली probability आणि, एकी experiment की दो dependent आहे (prior probability)

- दोन्ही random variable IId (विजेत्या असावित आहे)

Joint Probability → अंतर्गत संभिति

Example: 2 fr die roll याची अंतर्गतीली pair होती आहे

$$P(5, 6) = \frac{2}{6} = \frac{1}{3}$$

✓ Set Notations \rightarrow $\cap \cup \subset \in \{ \}$

$$\star \text{ Definition 7.2} \quad A \cup B = A \vee B$$

$$P(A) = \frac{|A|}{|\Omega|}$$

Summation of probability = 1

* Theorem 7.1

Conditional Probability

Example 7.4

* Definition 7.3

$$\rightarrow \begin{cases} P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A, B)}{P(B)} & '1' \rightarrow \text{Given sign; } \text{and} \\ & \text{sign be taken for} \\ & \text{given statement.} \\ P(B|A) = \frac{P(B \cap A)}{P(A)} & \left[\begin{array}{l} \text{Probability of } B \text{ when } A \text{ is} \\ \text{true.} \end{array} \right] \end{cases}$$

* Independent Event

$P(A)$ & $P(B)$ यदि Independent रहे तो

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

Example 7.5

* Chain Rule

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

$$\Rightarrow P(A \cap B) = P(A|B) \cdot P(B) \quad \text{--- ①}$$

for n variables

$$P(x_1, x_2, \dots, x_n) = P(x_n | x_1, x_2, \dots, x_{n-1}) \cdot P(x_1, x_2, \dots, x_{n-1})$$

$$= P(x_n \wedge (x_1, x_2, \dots, x_{n-1})) \cdot P(x_{n-1} | x_1, x_2, \dots, x_{n-2})$$

\prod → multiplication
denote अंतर्गत

$$= \prod_{i=1}^n P(x_i | x_1, x_2, \dots, x_{i-1})$$

* Marginalization

* Baye's Theorem

Conditional probability आणि Baye's Rule आवडते.

Baye's Rule आणि Baye's Bayesian Classifier आवडते.

↳ classification Algorithm

$$P(A \cap B) = P(A|B) \cdot P(B) \quad \text{--- ①}$$

$$P(B \cap A) = P(B|A) \cdot P(A) \quad \text{--- ②}$$

$$P(A \wedge B) = P(B \wedge A)$$

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

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

→ Baye's Rule.

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

↑ Posterior
(Prob. of B when A is true)
Likelihood
(Prob. of evidence)

↑ Marginal probability (Prob. of evidence)
Prior Prob. (Prob. of hypothesis)

Chapter 13 (Russell)Quantifying Uncertainty

$0 \leq P(\omega) \leq 1$ for ω $\sum_{\omega \in \Omega} P(\omega) = 1$

* Probability with a Proposition

$$\phi; P(\phi) = \sum_{\omega \in \phi} P(\omega)$$

$$P(\text{total} = 11) = P((5,6)) + P((6,5))$$

$$= \frac{1}{36} + \frac{1}{36}$$

$$= \frac{1}{18}$$

* Conditional Probability

$$P(\text{Cavity} | \text{toothache}) = 0.6$$

↳ given that

$$P(\text{cavity}) = ?$$

$$P(\text{covid-19} | \text{fever}) = 0.7$$

↳ given

$$P(\text{covid-19}) = ?$$

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$$P(a|b) = \frac{P(a \cap b)}{P(b)}$$

→ $P(a \cap b) = P(a|b) \cdot P(b)$ [Product Rule]

Example

$$T = \{\text{cavity, toothache, catch}\}$$

$$\text{Die}_1 = \{1, 2, 3, 4, 5, 6\}$$

$$\text{Age} = \{\text{juvenile, teen, adult}\}$$

$$\text{Weather} = \{\text{sunny, rainy, snow, cloudy}\}$$

$$P(\text{cavity} | \neg \text{toothache} \wedge \text{teen}) = 0.1$$

$$P(w = \text{sunny}) = 0.6$$

* Joint Probability (calculates the probability of two events occurring together and at the same point in time)

Marginalization or Marginal Probability

Summing out

$$P(Y) = \sum_{z \in Z} P(Y, z)$$

$Z \rightarrow$ another sample space

$$P(Y) = \sum_z P(Y|z) \cdot P(z) \quad [\text{conditioning rule}]$$

Marginalization (summing out) - task of marginal probability
 ↳ tells us just to add up some probabilities to get to the desired probabilistic quantity.

$$P(a|b) = \frac{P(b|a) \cdot P(a)}{P(b)} \xrightarrow{\text{marginalization}}$$

Observable value

Unobservable value

$$P(Y) = \sum_{z \in Z} P(Y, z)$$

$$= \sum_{z \in Z} P(Y|z) \cdot P(z) \quad (\text{rules of conditioning})$$

X, Y → joint probability

(., ' , \wedge \rightarrow similar sign)

X ^ Y → joint probability

Normalization → (The goal of normalization is to transform features to be on a similar scale. This improves the performance & training stability of model)

$$P(x|e) = \propto P(x, e) = \propto \sum_y P(x, e, y)$$

$$\alpha = \frac{1}{P(e)}$$

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Independence Probability

$$P(a|b) = P(a)$$

$$P(b|a) = P(b)$$

$$P(a \wedge b) = P(a) \cdot P(b)$$

$$\left. \begin{array}{l} P(a \wedge b) = P(a|b) P(b) \\ P(b \wedge a) = P(b|a) P(a) \end{array} \right\} \text{Product rule}$$

from product rules,

$$P(b|a) = \frac{P(a|b) \cdot P(b)}{P(a)} \rightarrow \text{Bayes rule}$$

Unconditional facts \rightarrow

Example

Bayesian classifier

13.16

13.17

Unit 1

$\prod_{i=1}^n P(\text{effect} | \text{cause}) P(\text{cause}) \rightarrow \text{Naive Bayesian Classifier}$

$$v_{NB} = P(\text{cause}) \prod_{i=1}^n P(\text{effect} | \text{cause})$$

Example of Naive Bayesian Classifier

$$\frac{\prod_{i=1}^n P(\text{effect}_i | \text{cause}) \cdot P(\text{cause})}{\text{Conditional Probability} \quad \text{Prior Probability}}$$

Prior probability of M, $P(M) = \frac{2}{5}$

" " " F, $P(F) = \frac{3}{5}$

$$P(M) + P(F) = \frac{2}{5} + \frac{3}{5} \\ = \frac{5}{5} = 1$$

	X	Y	
Name	H	W	
M	N	W	M
-	-	-	F
M			M
F			F
F			F

* Learning (Machine Learning) - 3 types

1. Supervised learning - output, level info, supervisor info
2. Unsupervised learning

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* Example 1

* Example 2

$$v_{NB} = \underset{v_j \in \{\text{yes, no}\}}{\arg \max} P(v_j) \prod_i P(a_i | v_j)$$

↑ argument

 a_i = attribute v_j = Y/N

** Normalization এর value যোগ করলে ১ হবে

*** অন্য example ফিল্টি implement করত হবে
for 3 instances

Lab mid → 23/10/2022

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*Example 1

Day	Outlook	Temperature	Humidity	Wind	Play Tennis
D1	Sunny	Hot	High	Weak	No
D2	Sunny	Hot	High	Strong	No
D3	Overcast	Hot	High	Weak	Yes
D4	Rain	Mild	High	Weak	Yes
D5	Rain	Cool	Normal	Weak	Yes
D6	Rain	Cool	Normal	Strong	No
D7	Overcast	Cool	Normal	Strong	Yes
D8	Sunny	Mild	High	Weak	No
D9	Sunny	Cool	Normal	Weak	Yes
D10	Rain	Mild	Normal	Weak	Yes
D11	Sunny	Mild	Normal	Strong	Yes
D12	Overcast	Mild	High	Strong	Yes
D13	Overcast	Hot	Normal	Weak	Yes
D14	Rain	Mild	High	Strong	No

Instance :

Outlook = Sunny

Temperature = Cool

Humidity = High

Wind = Strong

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SAT SUN MON TUE WED THU FR

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$$P(\text{Play Tennis} = \text{yes}) = \frac{9}{14} = 0.64$$

$$P(\text{Play Tennis} = \text{no}) = \frac{5}{14} = 0.36$$

Outlook	Y	N
Sunny	9/9	3/5
Overscast	4/9	0
Rain	3/9	2/5

Humidity	Y	N
High	3/9	4/5
Normal	6/9	1/5

Temperature	Y	N
Hot	9/9	2/5
Mild	9/9	2/5
Cool	3/9	2/5

Windy	Y	N
Strong	3/9	3/5
Weak	6/9	2/5

$$v_{NB} = \underset{v_j \in \{\text{yes, no}\}}{\operatorname{argmax}} P(v_j) \prod_i P(a_i | v_j)$$

$$= \underset{v_j \in \{\text{yes, no}\}}{\operatorname{argmax}} P(v_j)$$

$$P(\text{Outlook} = \text{Sunny} | v_j) P(\text{Temperature} = \text{Cool} | v_j)$$

$$P(\text{Humidity} = \text{High} | v_j) P(\text{Wind} = \text{Strong} | v_j)$$

$$v_{NB} = P(\text{yes}) P(\text{sunny} | \text{yes}) P(\text{cool} | \text{yes}) P(\text{high} | \text{yes}) P(\text{strong} | \text{yes}) \\ = 0.0053$$

$$v_{NB} = P(\text{no}) P(\text{sunny} | \text{no}) P(\text{cool} | \text{no}) P(\text{high} | \text{no}) P(\text{strong} | \text{no}) \\ = 0.0206$$

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$$V_{\text{yes}}(\text{yes}) = \frac{V_{\text{yes}}(\text{yes})}{V_{\text{yes}}(\text{yes}) + V_{\text{yes}}(\text{no})}$$

$$= 0.205$$

$$V_{\text{yes}}(\text{no}) = \frac{V_{\text{yes}}(\text{no})}{V_{\text{yes}}(\text{yes}) + V_{\text{yes}}(\text{no})}$$

$$= 0.795$$

~~Example-2~~

No.	Colour	Legs	Height	Smelly	Species
1	White	3	Short	Yes	M
2	Green	2	Tall	No	M
3	Green	3	Short	Yes	M
4	White	3	Short	Yes	M
5	Green	2	Short	No	H
6	White	2	Tall	No	H
7	White	2	Tall	No	H
8	White	2	Short	Yes	H

Instance: Colour = Green, Legs = 2, Height = Tall, Smelly = No

$$P(M) = \frac{4}{8} = 0.5$$

$$P(H) = \frac{4}{8} = 0.5$$

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SAT SUN MON TUE WED THU FRI

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Color	M	H
White	2/4	3/4
Green	2/4	1/4

Legs	M	H
2	1/4	4/4
3	3/4	0/4

Height	M	H
Tall	2/4	2/4
Short	1/4	2/4

Smelly	M	H
Yes	3/4	1/4
No	1/4	3/4

$$\begin{aligned}
 P(M \mid \text{Instance}) &= P(M) \cdot P(\text{Color} = \text{Green} \mid M) \cdot P(\text{Legs} = 2 \mid M) \cdot P(\text{Height} = \text{Tall} \mid M) \\
 &\quad P(\text{Smelly} = \text{No} \mid M) \\
 &= 0.5 * \frac{2}{4} * \frac{1}{4} * \frac{3}{4} * \frac{1}{4} \\
 &= 0.0117
 \end{aligned}$$

$$\begin{aligned}
 P(H \mid \text{Instance}) &= P(H) \cdot P(\text{Color} = \text{Green} \mid H) \cdot P(\text{Legs} = 2 \mid H) \cdot P(\text{Height} = \text{Tall} \mid H) \\
 &\quad P(\text{Smelly} = \text{No} \mid H)
 \end{aligned}$$

$$P(H \mid \text{Instance}) > P(M \mid \text{Instance})$$

- ∴ The instance belongs to species H.

Chapter 7 Logical Agents

Logical agents - in which we design agents that can form representations of a complex world, use a process of inference to derive new representations about the world, and use these new representations to deduce what to do.

Reasoning - The logical process of drawing conclusions, making predictions or constructing approaches towards a particular thought with the help of existing knowledge.

Logic - a general class as a general of representations to support knowledge-based agent.

- Propositional Logic
- First-order Logic

* Knowledge-based agent - are those agents who have the capability of maintaining an internal state of knowledge, reason over that knowledge, update their knowledge after observations and take action.

Knowledge-based agents are composed of -

- I) Knowledge-base
- II) Inference system

Knowledge-base - a set of sentences (technical terms) expressed in a language called knowledge representation language.

Axiom - when the sentence is taken as given without being derived from other sentences.

Inference - deriving new sentences from old.

Algorithm of generic knowledge based agent

function KB-AGENT (percept) returns an action
persistent: KB, a knowledge-base
to a counter, initially 0, indicating time

TELL (KB MAKE-PERCEP-SENTENCE)(percept, t))

action \leftarrow ASK (KB MAKE-ACTION-QUERY (t))

TELL (KB MAKE-ACTION-SENTENCE, (action, t))

$t \leftarrow t+1$

return action

Knowledge-based agent - 3 levels:

- 1) Knowledge level
- 2) Logical level
- 3) Implementation level

Declarative approach - TELL sentences one by one until the agent knows how to operate in its environment.

Procedural approach - encodes desired behaviors directly as a program code.

* The Wumpus World

The Wumpus world is a simple world example to illustrate the worth of a knowledge-based agent and to represent knowledge representation.

Stench		Breeze	(PIT)
Wumpus	Stench	PIT	Breeze
stench		Breeze	
Agent	Breeze	PIT	Breeze

Components of Wumpus:

1. Rooms adjacent to wumpus are smelly, it have stench
2. The rooms adjacent to PITS have breeze
3. There will be glitter in the room iff it has gold.
4. Wumpus can be killed by the agent and it would emit a horrible scream which can be heard anywhere in the cave

* PEAS Description of the Wumpus world:

Performance measure -

- +1000 points if agent comes out with gold
- -1000 points for being eaten by the wumpus or get killed by falling into the pit.
- -1 for each action, -10 for using arrow
- The game ends if the agent dies or came out of the cave.

United

Environment -

- A 4*4 grid of rooms
- The agent initially in room square [1,1], facing right.
- Location of gold, and wumpus, are random except [1,1]
- Each square except [1,1] can be pit with probability 0.2.

Actuators -

- Left turn
- Right turn
- Move forward
- Grab
- Release
- shoot

Sensors -

- Agent will receive stench in room adjacent to wumpus.
- Agent will perceive breeze in room adjacent to pit.
- Agent will perceive glitter in room where the gold is present.
- Agent will perceive bump if he walks into the wall
- When the wumpus is shot, it emits a horrible scream which can be heard anywhere in the cave.

* The Wumpus world Properties

- Partially observable
- Deterministic
- Sequential
- Static
- Discrete
- One agent

* Exploring the Wumpus World.

* Some Propositional Rules for Wumpus World:

- for a 4×4 square board, there will be $7 \times 4 \times 4 = 122$ propositional variables.

$$R1) \neg S_{11} \rightarrow \neg W_{11} \wedge \neg W_{12} \wedge \neg W_{13}$$

$$R2) \neg S_{11} \rightarrow \neg W_{11} \wedge \neg W_{21} \wedge \neg W_{22} \wedge \neg W_{31}$$

$$R3) \neg S_{12} \rightarrow \neg W_{11} \wedge \neg W_{12} \wedge \neg W_{21} \wedge \neg W_{13}$$

$$R4) S_{12} \rightarrow W_{13} \vee W_{12} \vee W_{22} \vee W_{31}$$

* Representation of Knowledge base for Wumpus World

* Prove that Wumpus is in room [1,3]

** Derive a propositional logic for wumpus world

* Propositional Logic - is a declarative statement which is either true or false but can't be both.

Conjunction \wedge

Disjunction \vee

Implication \rightarrow

Biconditional \leftrightarrow

Negation \neg

* A BNF grammar of sentences in propositional logic

Sentence -

i) Atomic \rightarrow True / false

ii) Compound \rightarrow sentence "connective" sentence

Connective \rightarrow AND, OR, NOT, implies, equivalent

Tautology \rightarrow universal truth/valid sentence

Contradiction \rightarrow inconsistent sentence

* Truth Table: five logical connectives

P	Q	$\neg P$	$P \wedge Q$	$P \vee Q$	$P \rightarrow Q$	$P \leftrightarrow Q$
F	F	T	F	F	T	T
F	T	T	F	T	T	F
T	F	F	F	T	F	F
T	T	F	T	T	T	T

* Proving things

Proof - sequences of sentences, where each sentence is either a premise or a sentence derived from earlier sentences in the proof by one or of the rules of inference.

* Horn Sentences

$$(P \rightarrow Q) = (\neg P \vee Q) \quad [\text{Truth table fact Proof}]$$

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- * Propositional Logic is a weak language
- * Studies propositional rules for wumpus world

7 marks: 2 questions

* Propositional logic

- quantification express ক্ষেত্রে পারে না

* First-Order logic/Predicate logic

- covers quantification & predicates

- extended part of propositional logic

Chapter - 8

First-Order Logic

* Representation Revisited

** Programming Language Lackings

* The Language of Thought

Communication goes অন্তর্ভুক্ত অবস্থায় হয়

Ambiguity - problem for a representation language

Whorf hypothesis =

* Combining the best of formal and natural languages

- Objects - people, houses, something
- Relations - unary relations, or properties
- Functions - father of,

↳ 3 for FOL natural language

Quantification → some, all

Figure 8.1

* Syntax & Semantics of FOL

Domain

Domain elements

Tuples

Page 291

* First Order Logic

Objects - things with individual identities

Properties - that distinguishes objects from other objects

Relations - that holds among sets of objects

Functions - subjects of relations

User Provides

- Constant symbols
- Functional symbols
- Predicate symbols

FOL Provides

- variable symbols
- Connectives
- Quantifiers

** Sentences are built from terms and atoms

* Quantifiers

- Universal Quantification
- Existential Quantification

Quantifier scope

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- Connections between All and Exists
 - Quantified Inference Rules

Final Examination Lab

Chatbot test with AI Using Python

* Models for FOL

Domain = set of objects

non-empty set of objects w.r.t. which an object refers to

Figure 8.2

* Symbols and Interpretations

Figure 8.3
 $\neg, =, \wedge, \vee, \Rightarrow, \Leftrightarrow$ } → Precedence

* Terms

* Atomic Sentences

* Complex sentences

* Quantifiers

- Universal (\forall)- Existential (\exists)

- Nested

* Equality

Page 300

★-Using FOL

- System को TELL/ASK करते हैं

Assertions and Queries in FOL

- Sentence को TELL को जारी करने के बाद knowledge base में add करते हैं
उसका assertion होता है
- ASK को जारी करते हैं → Query; always positive answer आता है

The Kinship Domain

- parenthood, brotherhood, marriage - etc

★ The Wumpus World

★ Knowledge Engineering in FOL

Knowledge engineer - someone who investigates a popular domain,
learns what concepts are important in

* Knowledge Engineering Steps:

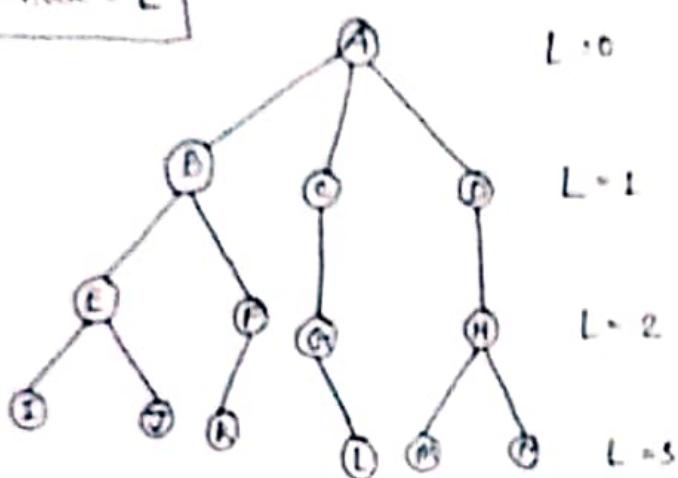
1. Identifying the task
2. Assemble the relevant knowledge
Knowledge acquisition
3. Decide a vocabulary
4. Encode general knowledge of the domain
5. Encode the specific problem
6. Post queries to the inference procedure
7. Debug the knowledge base

Exercises

* Iterative Deepening Search

IDS is a state space search strategy in which a depth-limited version of DFS is run repeatedly with increasing depth limits until the goal is found.

Goal node - L



1st iteration: A

2nd iteration: ABCD

3rd iteration: AB~~E~~FG~~C~~DH

4th iteration: AB~~I~~JFK~~G~~L~~D~~HMN

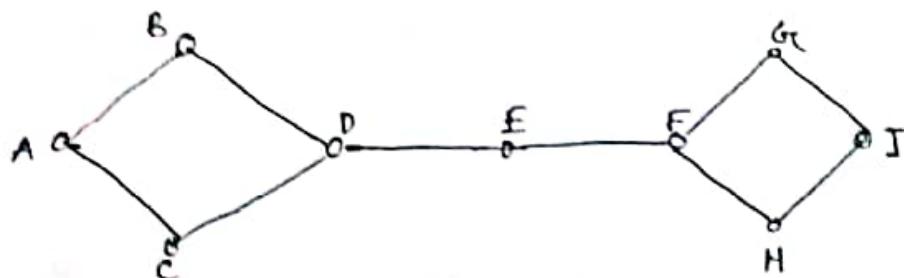
Here we found our goal node L after 4th iteration.

$$TC \rightarrow O(b^d)$$

$$SC \rightarrow O(bd)$$

* Bidirectional Search

Bidirectional search is a graph search algorithm that finds a shortest path from an initial vertex to a goal vertex/node. It runs two simultaneous searches - one forward from the initial state, and one backward from the goal, stopping when the two meets.

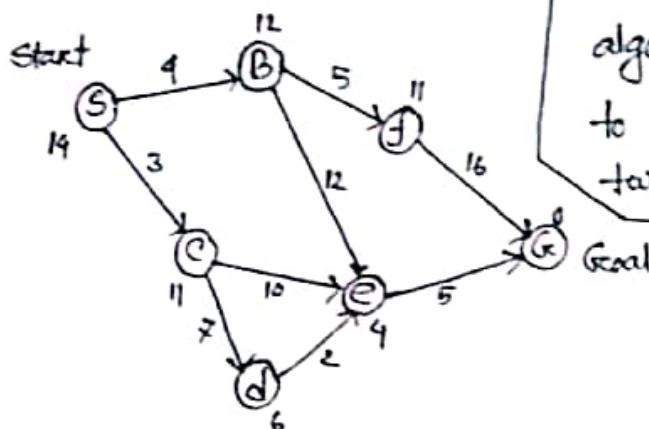


We want to find if there is a path from A to I. Hence we can execute two searches, one from the starting node A and others from goal node I. When both forward & backward search meet at node FE, we know that we've found a way from node A to I. Hence search can be terminated now.

* Why $\Rightarrow O(B^d + b^d)$ is less than $O(B^d)$

- * When \Rightarrow 1) branching factors same for A and I
2) when BFS is used

* A* Search



A* Search algorithm is a simple and efficient search algorithm that can be used to find the optimal path between two nodes in a graph. It is used for the shortest path finding.

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

actual cost from start node to n

Estimated cost from n to goal node

$$f(S) = 0 + 14 = 14$$

$$S \rightarrow B = 4 + 12 = 16$$

$$S \rightarrow C = 3 + 11 = 14$$

$$SC \rightarrow BE = (3 + 10) + 4 = 17$$

$$SC \rightarrow D = (3 + 7) + 6 = 16$$

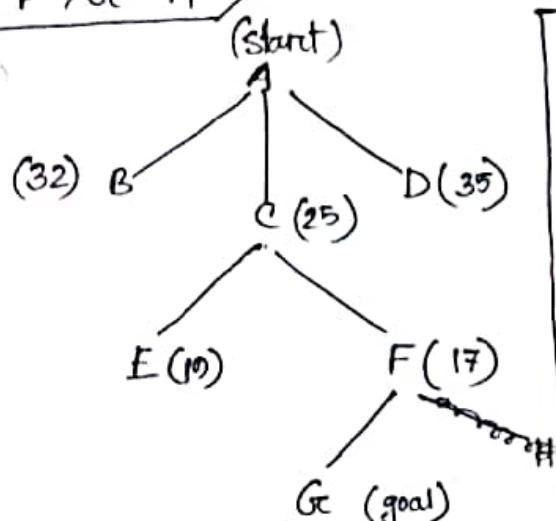
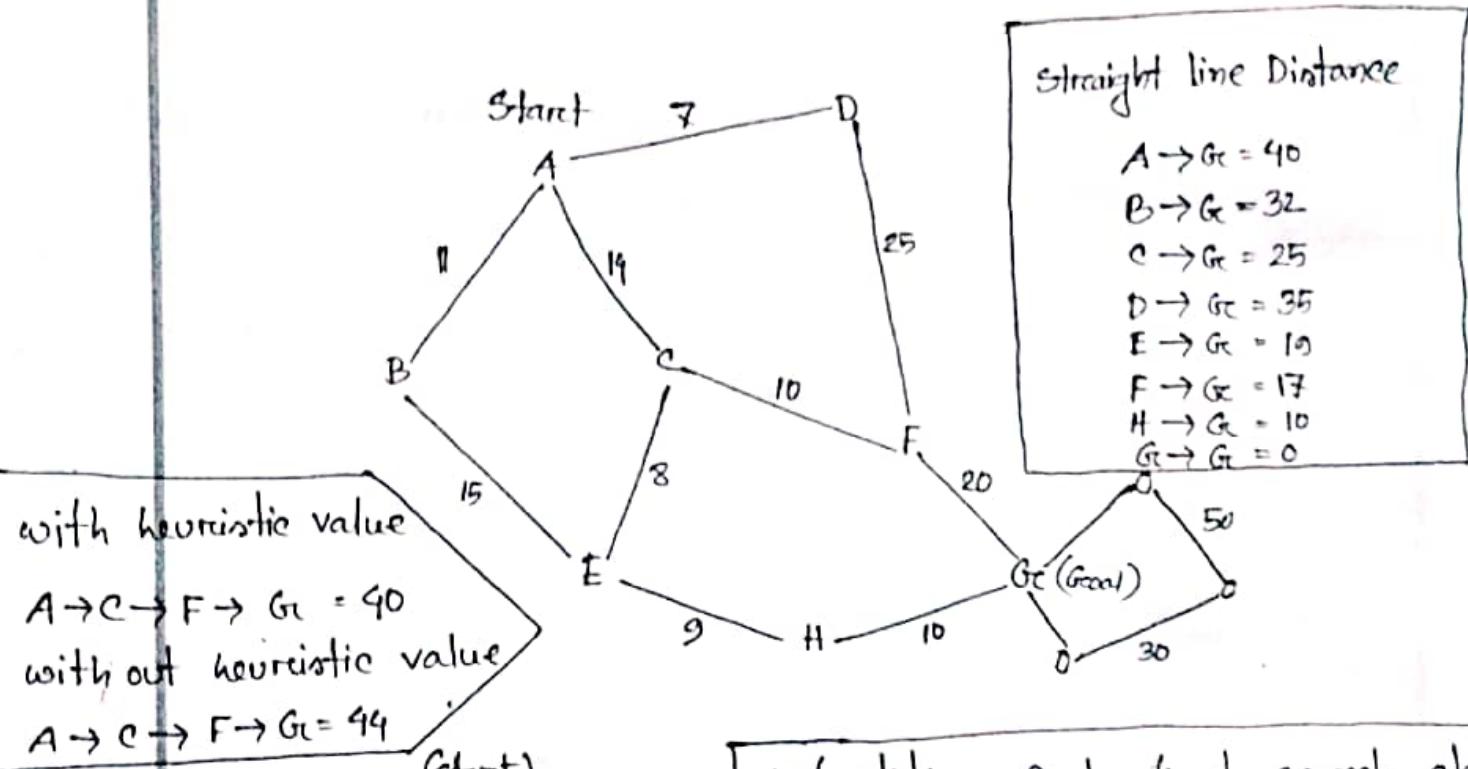
$$SCD \rightarrow E = (3 + 7 + 2) + 4 = 16$$

$$SCDE \rightarrow G = (3 + 7 + 2 + 5) + 0 = 17$$

$$T.C = O(b^d)$$

$$S.C = O(b^d)$$

* Best First Search (Informed, Heuristic)



Based on heuristic value

$A \rightarrow C \rightarrow F \rightarrow G_c$

Definition: Best first search algorithm is a traversal technique that decides which node is to be visit next by checking which node is the most promising one and then check it.

United

If gives good solution. Optimal solution is not guaranteed.

$T \cdot C \rightarrow O(b^m) / O(b^d)$

In worst case it performs same but in normal/average case it performs better than uninformed search.

* How to minimize the total estimated solution cost using the best-first search, A* search algorithm?

→ By making A* admissible, we can minimize the total estimated solution cost.

For A* search algorithm, the equation we use is

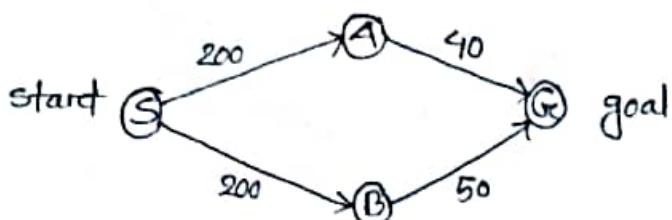
$$f(n) = g(n) + h(n)$$

where $g(n)$ = actual cost from start node to n .

$h(n)$ = estimated cost from n to goal node.

We know, when -

when $h(n) \geqslant$ actual cost, then it is overestimation and when $h(n) \leqslant$ actual cost, then it is underestimation.



hence,

$$g(A) = 200$$

$$g(B) = 200$$

actual cost, $A \rightarrow G = 40$

actual cost, $B \rightarrow G = 50$

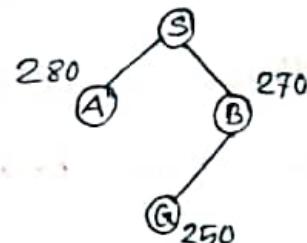
Case I : Overestimation

Let, $\begin{cases} h(A) = 80 \\ h(B) = 70 \end{cases} \} > \text{actual cost}$

$$\text{Now, } f(A) = 200 + 80 = 280$$

$$f(B) = 200 + 70 = 270$$

$$f(G) = g(G) + h(G) = 250 + 0 = 250$$



As A is greater so the search is stopped.

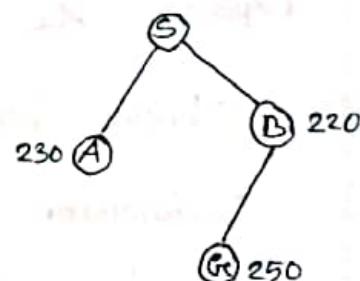
Case II : Underestimation

Let, $\begin{cases} h(A) = 30 \\ h(B) = 20 \end{cases} < \text{actual cost}$

$$f(A) = 200 + 30 = 230$$

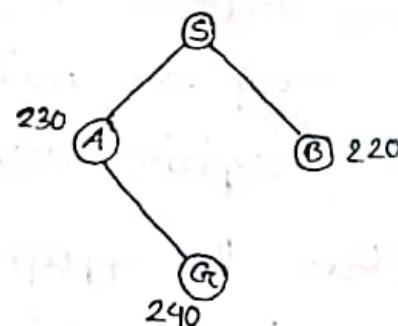
$$f(B) = 200 + 20 = 220$$

$$f(G) = 250$$



here A* doesn't stop but as A is less than G. So it explores A then G.

$$f(G) = g(G) + h(G) = 240 + 0 = 240$$



here 240 is less than 250, so the estimated cost is minimized.

Underestimation gives optimal solution

* What are the learning outcomes of Artificial Intelligence?

→ There are several learning outcomes of AI. Some of them are given below-

- Distinguishes between a conventional system and intelligent system
 - describes data, information, knowledge
 - explains algorithmic versus heuristic methods
 - defines structured and unstructured.
- Explains AI concepts and its applications
 - knows how to represent knowledge
 - describe searching methods
 - explains approximate reasoning
- Represents knowledge using various different techniques.
 - explains first order logic and predicate calculus
 - explains semantic network
 - explains rule-based system
 - explains case-based system
- Uses the appropriate searching techniques to reach desired goals
 - knows state space search
 - distinguishes between data driven and goal driven search
 - uses BFS, DFS; - uses some heuristic search method

* What is Uninformed Search?

→ Uninformed search is a class of general-purpose search algorithms which operates in brute force-way. Uninformed search algorithms do not have additional information about state or search space other than how to traverse the tree, so it is also called a blind search.

* What is Informed Search?

→ Informed search algorithms have information on the goal state which helps in more efficient searching. This information is obtained by a function that estimates how close a state is to the goal state.

* What is the heuristic function of informed search strategy?

→ Heuristic is a function which is used in informed search. It is a way to inform the search about the direction of the goal. It provides an informed way to guess which neighbors of a node will lead to a goal. It finds the most promising path. It takes the current state of the agent as its input and produces the estimation of how close the agent is from the goal.

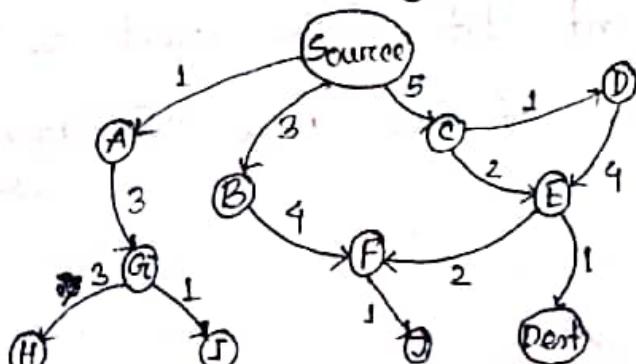
Mid Question (2017-18)

1. What is blind search technique?

→ Blind search technique applies a way in which search tree is searched without any information about the search space like initial state operators and test for the goal, so it is also called uninformed search. It examines each node of the tree until it achieves the goal node.

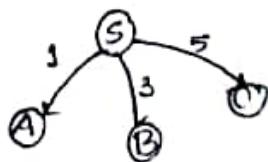
2. Describe Uniform-Cost Search method.

→ Uniform-cost search is an uninformed search algorithm that uses the lowest cumulative cost to find a path from the source to the destination. In this method, the nodes are expanded, starting from the root, according to the minimum cumulative cost. The uniform-cost search is then implemented using a priority queue.

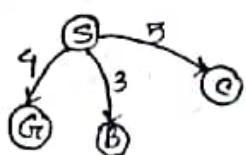


Time complexity needed to run ucs is:
 $O(b^{[1+c^*/e]})$ [steps = $1 + c^*/e$]
 b = branching factor,
 c^* = total solution cost
 e = each step cost

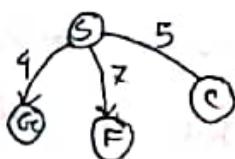
First we have the source node in the priority queue and we add their children with their cumulative distance as priority



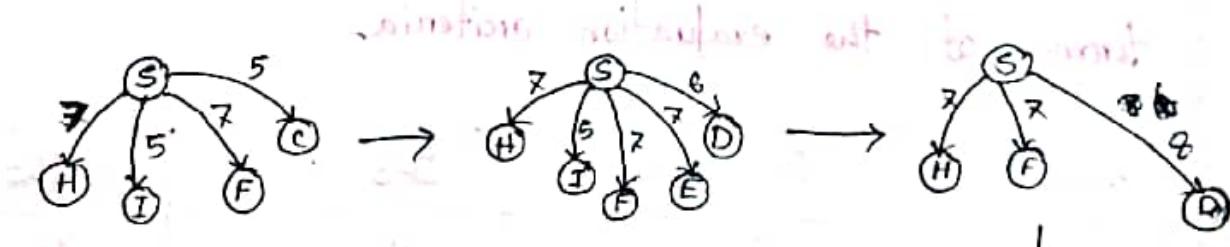
here 'A' has the minimum distance i.e maximum priority, so it is extracted from the destination and its children are added to the PQ.



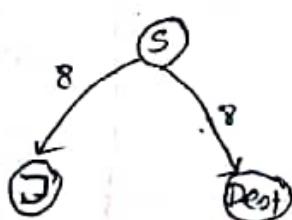
Similarly removing B and adding its children to the PQ



Following this procedure we get,



The minimum distance between the source & destination has been found i.e 8.



3. How to evaluate an algorithm's performance?

→ An algorithm's performance can be evaluated based on the four properties -

1. Completeness : An algorithm is said to be complete if it is guaranteed to find the solution if at least for any random solution.

2. Optimality : If a solution is found to be best among all the solutions for an algorithm

3. Time Complexity : It is a measure of time for an algorithm to complete its task.

4. Space Complexity : It defines how much memory is needed to complete the task for an algorithm.

4. Compare BFS, DFS & Iterative deepening search strategies in terms of the evaluation criteria.



Criterior	BFS	DFS	IDS
Completeness	Yes	No	Yes
Optimality	Yes	No	Yes
Time Complexity	$O(b^d)$	$O(b^n)$	$O(b^d)$
Space Complexity	$O(b^d)$	$O(bm)$	$O(bd)$

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DATE / /

Masud Sirc

Mahbub Sirc

1. Chapter 3

1. Chapter 7

2. Chapter 9

2. Chapter 8

3. Chapter 6

✓ 4. Chapter 13

Uncheck