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Artificial Intelligence.

SNO	Date	Title	sign
1	24/09/24	Tic-Tac-Toe	✓ 24/9/24
2	1/10/24	Vacuum Cleaner	✓ 1/10/24
3	8/10/24	8 puzzle - BFS, DFS	✓ 8/10/24
4	15/10/24	8 puzzle - A* implementation	✓ 15/10/24
5	22/10/24	8 puzzle - IDS N-Queens - Hill climbing	✓ 22/10/24
6	29/10/24	Simulated Annealing	✓ 29/10/24
7	12/11/24	Proposition Logic	✓ 12/11/24
8	19/11/24	Unification for FOL	✓ 19/11/24
9	26/11/24	Forward Chaining/Reasoning	
10)	26/11/24	FOL to Resolution	
11)	3/12/24	Alpha Beta pruning	

1) Tic-Tac-Toe :

Algorithm :

1) Initialize the game board :

- Create 3×3 matrix
- Each position in the board is initialized to '-'

2) Start the loop :

- The game runs for a maximum of 9 turns
- The game alternates between two players, "X" and "O"

3) Display the board :

- Print the current status of the board

4) Player Input :

- The current player selects the position
- The selected position is checked

5) Update the board :

- Place the current player mark in the selected position

6) Check for win :

- After every move, check if the current player has won
- Any row or Any column has same all the same mark or either of the diagonal has all the same mark

7) Check for draw :

- If all 9 positions are filled, then it is a draw.

8) If a player wins or a draw, then end the game.

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2) Implement vacuum world problem

function REFLEX-VACUUM-AGENT([LOCATION, STATUS]) returns an action

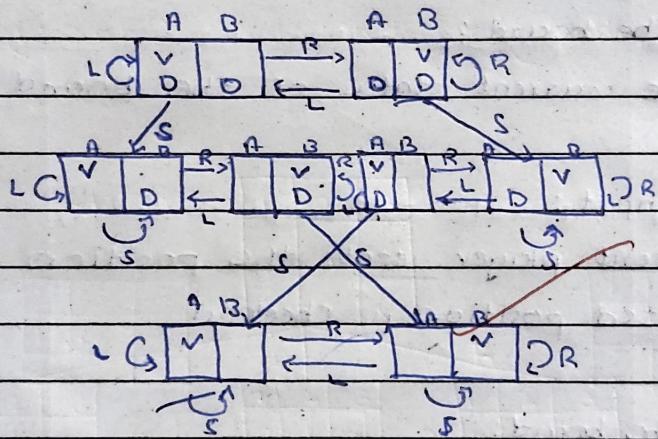
if status = Dirty then return Suck

else if location = A then return Right

else if location = B then return Left

state space diagram of vacuum world.

function simple



Problem formulation steps:

* States

* Initial state

* Actions

* Transition model

* Reach test

* Path cost

Algorithm:

- 1) Initialize room and cost
- 2) Check the room are clean or not
If room is dirty clean the room and move to the next location.
- 3) Next check whether previously cleaned is room is dirty if dirty clean.
- 4) If both the rooms are clean then return the total cost and off the vacuum cleaner.

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- 3) solve 8 puzzle problem using DFS & BFS
- 4) Implement depth limited search algorithm & Uniform cost search algorithm
- 5) Implement iterative deepening search

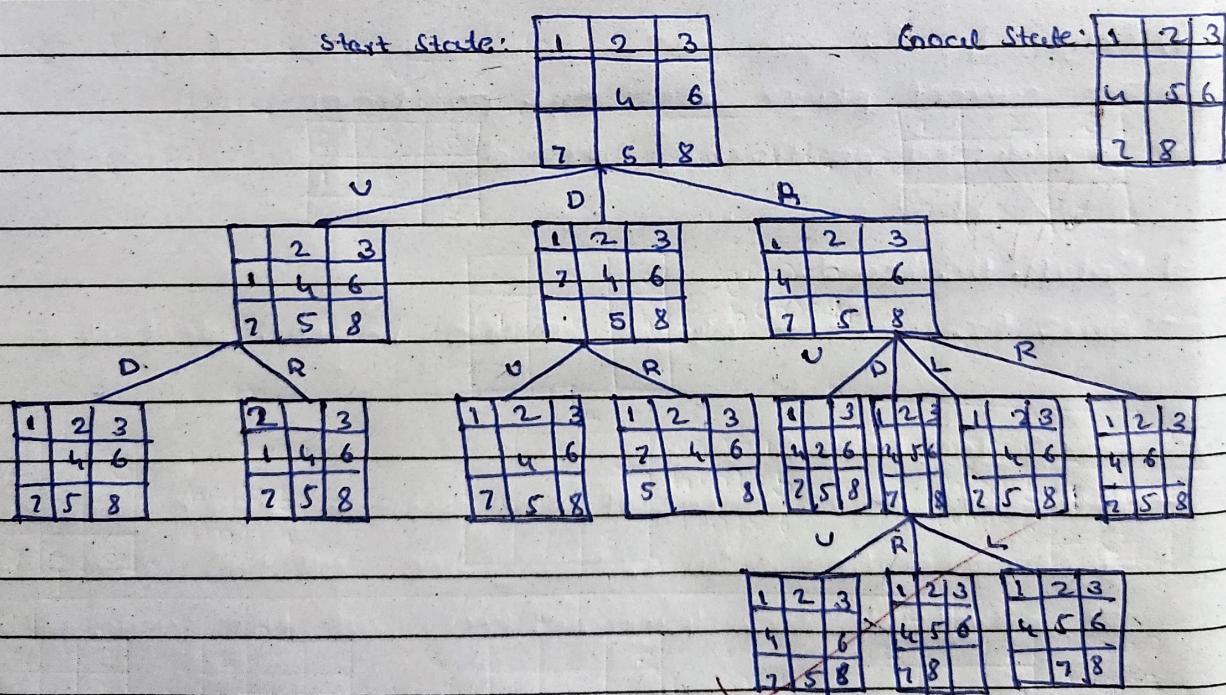
Breadth-First-Search:

Complete? Yes

Optimal? Yes, if path cost is nondecreasing function of depth

Time complexity: $O(b^d)$

Space complexity: $O(b^d)$, note that every node in the fringe is kept in the queue.



3) Solve 8 puzzle problem using DFS & BFS.

BFS:

- The initial state & goal state will be given.
- We need to achieve goal state of 8 puzzle using BFS algorithm.
- Initially the parent node will be the initial state, possible movements of blank space will be recorded i.e., up, down, right, left.
- These possible movements will be the child nodes of initial state.
- Define helper function findBlankTile(state) to locate the position of blank tile
- Define another function getpossibleMoves(state) that generates all valid moves for the current state:
 - Determine the position of the blank tile.
 - For each possible direction (up, down, left, right), record direction & move the tile.
- Run the BFS loop until we reach the goal state.

Top
left

Q) For 8 puzzle problem using A* implementation, to calculate f(n) using a) misplaced tiles b) Manhattan Distance
 a) g(n) - depth of node b) h(n). 15/10/24

A* Misplaced Tiles:

- 1) Define the initial & goal state
- 2) Define list & $f(n) = g(n) + h(n)$ where,
 $g(n)=0$, & $h(n) \rightarrow$ misplaced tiles
 where $g(n)$ is cost to reach each node & $h(n)$ is heuristic value.
- 3) While list is not empty, extract node with lowest frequency.
 - If current state is goal state return path & add the current state to closed set.
 - a) Generate new state by moving up, down, left, right, if new state is not in closed set, calculate $g(n)$ for new state & $h(n)$ & $f(n)$. Continue the process until goal is reached.

Give state space diagram for

A* Manhattan Distance Puzzles

initial state :

2	8	3
1	6	4
7		5

Goal state :

1	2	3
8		4
7	6	5

2	3	3
1		4
7	6	5

$g(n)=0$
 $h(n)=3$
 $f(n)=4$

2	8	3
1	6	4
7		5

$g(n)=0$
 $h(n)=4$
 $f(n)=4$

$g(n)=0$
 $h(n)=4$
 $f(n)=4$

L
 S
 R

2	8	3
1	6	4
7		5

$g(n)=1$
 $h(n)=5$
 $f(n)=6$

2	8	3
1	6	4
7	5	

$g(n)=1$
 $h(n)=5$
 $f(n)=6$

U			R	D			L	U			R	D			L	U			R	D			L
2	3	3	$g(n)=2$	2	8	3	$g(n)=2$																
1	8	4	$h(n)=3$	1	6	4	$h(n)=4$																
7	6	5	$f(n)=5$	7		5	$f(n)=6$																

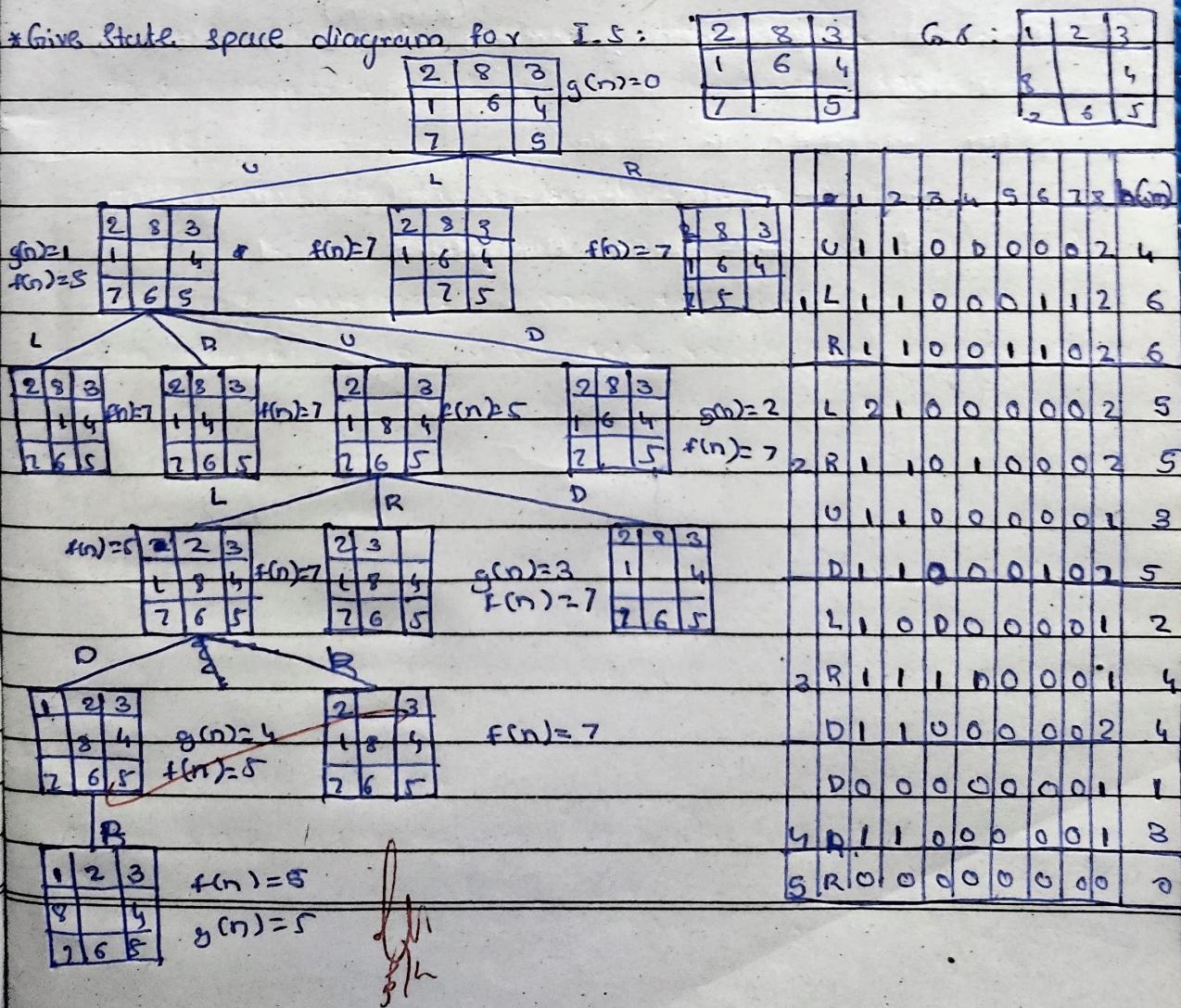
L			R	D			U	L			R	D			U	L			R	D			U
2	3	3	$g(n)=3$	2	8	3	$g(n)=3$																
1	8	4	$h(n)=4$	1	6	4	$h(n)=4$																
7	6	5	$f(n)=5$	7		5	$f(n)=6$																

D			R	U			L	D			R	U			L	D			R	U			L
2	4	3	$g(n)=4$	2	3	3	$g(n)=4$																
1	8	4	$h(n)=3$																				
7	6	5	$f(n)=5$	7		5	$f(n)=6$																

Manhattan Distance:

- 1) Define goal state & also create a dictionary.
- 2) Create a list containing $f(n)$, $g(n)$ & path with $g(n) \geq 0$.
- Define a closed set to track visited states.
- 3) Calculate manhattan distance,
 - for each tile from 1 to 8 in current state
 - find this current position & find target by goal state & calculated distance.
 - sum these distances to get $h(n)$.
- 4) Expand nodes, while list is not empty, extract node with low frequency & return path.

* Give state space diagram for 1-5:



Lab 5: Implement Iterative Deepening Search Algorithm.

- 1) For each child of the current node
- 2) If it is the target node return
- 3) If the curr max depth is reached return
- 4) Set the current node to this node & go back to 1
- 5) After having gone through all children go to the next child of the parent (the next sibling)
- 6) After having gone through all children of the start node increase the maximum depth & go back to 1
- 7) If we have reached all leaf (bottom) nodes, the goal back doesn't exist.

function Iterative-Deepening-search (^(problem) returns a solution or failure

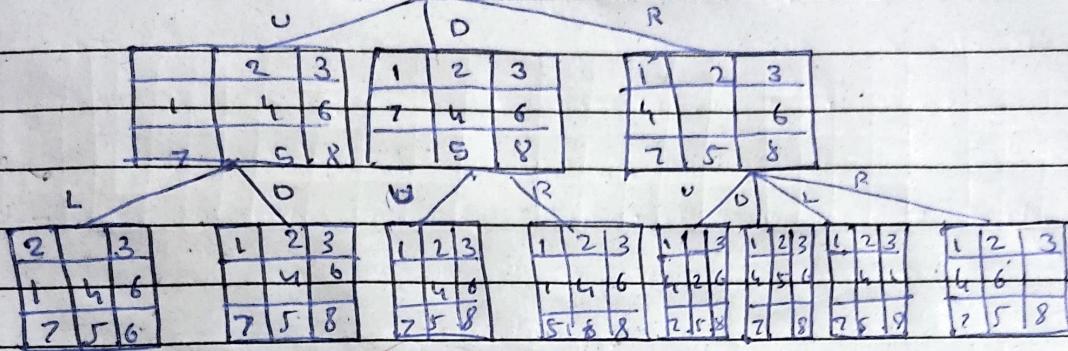
for depth = 0 to ∞ do

result ← Depth-limited-search (^(problem, depth))

if result ≠ cutoff then return result

1	2	3
4		6
7	5	8

1	2	3
4	5	6
7	8	



✓ 12
22/12 ~

e) Hill climbing search Algorithm to solve, N Queens problem

Hill climbing search Algorithm :

function HILL-CLIMBING (problem) returns a state that is a local maximum

 current \leftarrow MAKE-NODE (problem, INITIAL-STATE)

 loop do

 neighbours \leftarrow a highest-valued successor of current

 if neighbours.Value \leq current.Value then return
 current.STATE

 current \leftarrow neighbours.

Problem Formulation :

• State : 4 queens on the board one queen per column

- Variables : x_0, x_1, x_2, x_3 - where x_i is the new position of the queen in column i. Assume that there is one queen per column.

- Domain for each variable $x_i \in [0, 1, 2, 3] \forall i$

• Initial state : a random state

• Goal state : 4 queens on the board. No pair of queens are attacking each other.

• Neighbour relation :

 swap the row position of two queens

• cost function : The no of pairs of queens attacking each other, directly or indirectly.

	Q		
1		Q	
2			Q
3			
4			Q

$$n(n) = 12$$

1,2	3,4	1,3	1,4	2,3	2,4
a	a	a	a	a	a
a	a	a	a	a	a
a	a	a	a	a	a
4	4	4	4	4	4

(1,3)	(1,u)	(2,3)	(2,4)	(3,4)
2	2	2	2	4
a	a	a	a	a
a	a	a	a	a
a	a	a	a	a
2	2	2	2	4

(3,4)	(1,3)	(2,4)	(3,u)	(1,u)
1	2	1	1	1
a	a	a	a	a
a	a	a	a	a
a	a	a	a	a
1	2	1	1	1

12
22/10/2020

29/10/21

6)

a) Write a program to implement simulated Annealing Algorithm.

function SIMULATED-ANNEALING(problem, schedule) returns
a solution state

inputs: problem, a problem

schedule, a mapping from time to "temperature"

current \leftarrow MAKE-NODE(problem, INITIAL-STATE)

for $t=1$ to ∞ do

$T \leftarrow$ schedule(t)

if $T = 0$ then return current

next \leftarrow a randomly selected successor of current

$\Delta E \leftarrow$ next.VALUE - current.VALUE

if $\Delta E > 0$ then current \leftarrow next

else current \leftarrow next only with probability $e^{\Delta E/T}$

SA - Steps :

1. Start at a random point x

2. Choose a new point x_j , on a neighborhood $N(x)$.

3. Decide whether or not to move to the new point x_j , The decision will be made based on the probability function $P(x, x_j, T)$

$$P(x, x_j, T) = \begin{cases} 1, & F(x_j) \geq F(x) \\ e^{\frac{F(x_j) - F(x)}{T}}, & F(x_j) < F(x) \end{cases}$$

4. Reduce T

8-Queens Problem

Output:

The best position found is : [4 0 2 5 2 6 1 3]

The number of queens that are not attacking each other is = 8.0

TSP :

Result structure : (array [1, 0, 3, 5, 4, 2],
21.02934853206 ,None)

Best route found : [1 0 3 5 4 2]

Total distance of best route : 21.02934853206

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1024
29

* Wumpus World using Proposition Logic

$KB + \alpha \rightarrow \text{query}$

↳ Knowledge base

$P_{x,y}$ - pit in $[x,y]$

$W_{x,y}$ - wumpus in $[x,y]$

$B_{x,y}$ - Breeze in $[x,y]$

$S_{x,y}$ - stench in $[x,y]$

$R_1 : \neg P_{1,1}$

$R_2 : B_{1,1} \iff P_{1,2} \vee P_{2,1}$

$R_3 : B_{2,1} \iff (P_{1,1} \vee P_{2,2} \vee P_{3,1})$

function TT-ENTAILS? (KB, α) returns true or false

inputs: KB , the knowledge base, a sentence in PL

α , the query, a sentence in PL

symbols ← a list of the proposition symbols in KB and
return TT-CHECK-ALL ($KB, \alpha, \text{symbols}, \{ \}$)

function TT-CHECK-ALL ($KB, \alpha, \text{symbols}, \text{model}$) returns true or
false

if EMPTY? (symbols) then

if PL-TRUE? (KB, model) then return PL-TRUE? (α, model)

else return true // when KB is false, always return true

else do

$R \leftarrow \text{FIRST}(\text{symbols})$

$\text{rest} \leftarrow \text{REST}(\text{symbols})$

$\text{return } (\text{TT-CHECK-ALL } (\text{KB}, \alpha, \text{rest}, \text{model } U \setminus \{P=\text{true}\}))$
and

$\text{TT-CHECK-ALL } (\text{KB}, \alpha, \text{rest}, \text{model } U \setminus \{P=\text{false}\}))$

Propositional Inference : Enumeration Method

$$\alpha = A \vee B$$

$$\text{KB} = (A \vee C) \wedge (B \vee \neg C)$$

checking that $\text{KB} \models \alpha$

A	B	C	$A \vee C$	$B \vee \neg C$	KB	α
F	F	F	F	T	F	F
F	F	T	T	F	F	F
F	T	F	F	T	F	T
*	F	T	T	T	T	T
*	T	F	F	T	T	T
-	F	T	T	F	F	T
*	T	T	F	T	T	T
*	T	T	T	T	T	T

O/P seen
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8) Implement Unification in First Order Logic

Algorithm : Unity (Ψ_1, Ψ_2)

Step 1: If Ψ_1 or Ψ_2 is a variable or constant then :

- If Ψ_1 or Ψ_2 are identical, then return NZL.
- Else if Ψ_1 is a variable,
 - then if Ψ_1 occurs in Ψ_2 , then return FAILURE.
 - Else return $\{(\Psi_2/\Psi_1)\}$.
- Else if Ψ_2 is a variable
 - if Ψ_2 occurs in Ψ_1 , then return FAILURE.
 - Else return $\{(\Psi_1/\Psi_2)\}$.
- Else return FAILURE.

Step 2: If the initial Predicate symbol in Ψ_1 and Ψ_2 are not same, then return FAILURE.

Step 3: If Ψ_1 & Ψ_2 have a diff. no. of args, then return FAILURE

Step 4: Set Substitution set (SUBST) to NZL.

Step 5: For i=1 to the no of ele in Ψ_1 ,

- call unity function with the ith ele of Ψ_1 & ith ele of Ψ_2 & put the result into s
- If s = failure, then return Failure
- If s ≠ NZL then do,
 - Apply s to the remainder of both L1 & L2.
 - $SUBST = APPEND(s, SUBST)$

Step 6: Return SUBST.

$$p(x, f(y)) \rightarrow ①$$

$$p(a, f(g(x))) \rightarrow ②$$

① & ② are identical if x is replaced with a in ①

$$p(a, f(y)) \rightarrow ①$$

if y is replaced with $g(x)$

$$p(a, f(g(x))) \rightarrow ①$$

$$q(a, g(x, a), f(y)) \rightarrow ①$$

$$q(a, g(f(a), a), x) \rightarrow ②$$

$f(b)$ is replaced with x in ②

$$q(a, g(x, a), x)$$

x is replaced with $f(y)$ in ②

$$q(a, g(x, a), f(y)) \rightarrow ①$$

$$\begin{aligned} ① \quad \psi_1 &= p(f(a), g(y)) & \psi_2 &= p(x, x) \\ \rightarrow \text{SUBST } &(f(a)/x) \end{aligned}$$

$$\begin{aligned} \psi_1 &= p(f(a), g(y)) & \psi_2 &= p(f(a), f(a)) \\ \rightarrow \text{SUBST } &(f(a)/g(y)) \end{aligned}$$

Failed

$$\begin{aligned} ② \quad \psi_1 &= p(b, x, f(g(x))) & \psi_2 &= p(z, f(y), f(y)) \\ \rightarrow \text{SUBST } &(b, z) \end{aligned}$$

$$\begin{aligned} \psi_1 &= p(b, x, f(g(b))) & \psi_2 &= p(b, f(y), f(y)) \\ \rightarrow \text{SUBST } &(f(y), x) \end{aligned}$$

$$\begin{aligned} \psi_1 &= p(b, f(y), f(g(b))) & \psi_2 &= p(b, f(y), f(g(y))) \end{aligned}$$

LAB-8

a) Create a knowledge base consisting of FOL statements
to prove the given query using forward reasoning

function FOL-FC-ASK(KB, α) returns a substitution or fail
inputs: KB , the knowledge base, a set of first-order
definite clauses

α , the query, an atomic sentence

local variables: new , the new sentences inferred on
each iterations

repeat until new is empty

$\text{new} \leftarrow \{\}$

for each rule in KB do

$(P_1 \wedge \dots \wedge P_n \Rightarrow Q) \leftarrow \text{STANDARDIZE-VARIABLES}(rule)$

for each rule Θ such that $\text{SUBST}(\Theta \Rightarrow P_1 \wedge \dots \wedge P_n) = \text{SUBST}$

$(\Theta, P_1 \wedge \dots \wedge P_n)$

for some $(P'_1 \wedge \dots \wedge P'_n)$ in KB

$g' \leftarrow \text{SUBST}(\Theta, g)$

if g' does not unify with some sentence already
in KB or new then

add g' to new

$\alpha \leftarrow \text{UNIFY}(g', \alpha)$

if α is not fail then return α

add new to KB

return fail

Representation in FOL:

It is a crime for an American to sell weapons to hostile nations.

Let's say p , q & r are variables

$\text{American}(p) \wedge \text{weapon}(q) \wedge \text{sellr}(p, q, r) \wedge \text{Hostile}(r) \Rightarrow \text{Criminal}(p)$

→ Country A has some missiles

$\exists x \text{ owns}(A, x) \wedge \text{missile}(x)$

Existential instantiation, Introducing a new constant t_1 : $\text{owns}(A, t_1)$, $\text{missile}(t_1)$

→ All the missiles were sold to country A by Robert

$\forall x \text{ missile}(x) \wedge \text{owns}(A, x) \Rightarrow \text{sellr}(\text{Robert}, x, A)$

→ Missiles are weapons: $\text{missile}(x) \Rightarrow \text{weapon}(x)$

→ Robert is an American: $\text{American}(\text{Robert})$

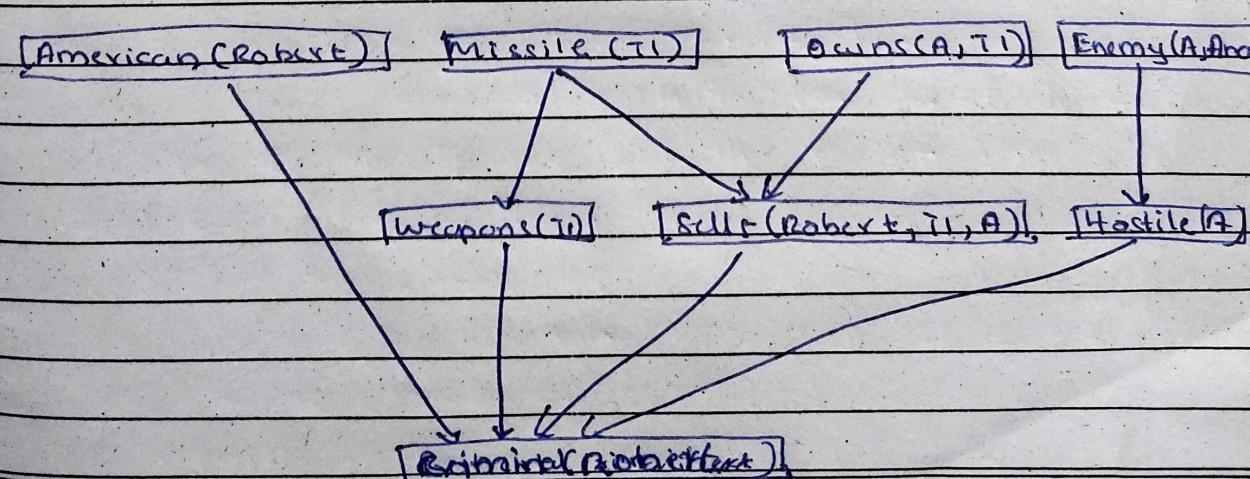
→ Enemy of America is known as hostile

$\forall x \text{ Enemy}(x, \text{America}) \Rightarrow \text{Hostile}(x)$

→ The country A, an enemy of America

$\text{Enemy}(A, \text{America})$

To Prove: Robert is a criminal: $\text{Criminal}(\text{Robert})$



$\text{American}(\text{Robert}) \wedge \text{weapon}(t_1) \wedge \text{sellr}(\text{Robert}, t_1, A) \wedge \text{Hostile}(A)$
 $\Rightarrow \text{Criminal}(\text{Robert})$

Q) Convert a given first order logic statement into Resolution:

Basic steps for proving a conclusion S given premises
Premise, --- Premise

1. Convert all sentences to CNF
2. Negate conclusion S & convert result to CNF
3. Add negated conclusion S to the premise clauses
4. Repeat until contradiction or no progress is made
 - a. select 2 clauses (call them parent clauses)
 - b. Resolve them together performing all required unifications
 - c. If resolvent is the empty clause, a contradiction has been found
 - d. If not, add resolvent to the premises

If we succeed in step 4, we have proved the conclusion.

Representation in FOL:

Given KB or premises:

- a. John likes all kind of food
 - b. Apple & vegetables are food
 - c. Anything anyone eats & not killed is food
 - d. Anil eat peanut & still alive
 - e. Harry eats everything that Anil eats
 - f. Anyone who is alive implies not killed
 - g. Anyone who is not killed implies alive
 - h. Prove by resolution that ? John likes peanuts
- a. $\forall x : \text{food}(x) \rightarrow \text{likes}(\text{John}, x)$
 - b. $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetables})$
 - c. $\forall x \forall y : \text{eats}(x, y) \wedge \neg \text{killed}(x) \rightarrow \text{food}(y)$
 - d. $\text{eats}(\text{Anil}, \text{peanuts}) \wedge \text{alive}(\text{Anil})$
 - e. $\forall x \forall y : \text{eats}(\text{Anil}, x) \rightarrow \text{eats}(\text{Harry}, x)$
 - f. $\forall x : \neg \text{killed}(x) \rightarrow \text{alive}(x)$
 - g. $\forall x : \text{alive}(x) \rightarrow \neg \text{killed}(x)$
 - h. likes (John, peanuts)

1) Eliminate Implication :

- a) $\forall x \neg \text{food}(x) \vee \text{likes}(\text{John}, x)$
- b) $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetables})$
- c) $\forall x \forall y \neg [\text{eats}(x, y) \wedge \neg \text{killed}(x)] \vee \text{food}(y) \rightarrow \forall x \forall y \neg \text{eats}(x, y) \vee \text{killed}(x) \wedge \text{food}(y)$
- d) $\text{eats}(\text{Anil}, \text{Peanut}) \wedge \text{alive}(\text{Anil})$
- e) $\forall x \neg \text{eats}(x, z) \vee \text{eats}(\text{Harry}, z)$
- f. $\forall x \neg [\text{killed}(x)] \vee \text{alive}(x)$
- g) $\forall z \neg \text{alive}(z) \vee \neg \text{killed}(z)$
- h) $\text{likes}(\text{John}, \text{Peanuts})$

2) Move Negation (\neg) inward & write

3) Rename variables, or standardize variables

- a. $\forall x \neg \text{food}(x) \vee \text{likes}(\text{John}, x)$
- b. $\text{food}(\text{Apple}) \wedge \text{food}(\text{vegetable})$
- c. $\forall y \forall z \neg \text{eats}(y, z) \vee \text{killed}(y) \vee \text{food}(z)$
- d. $\text{eats}(\text{Anil}, \text{Peanut}) \wedge \text{alive}(\text{Anil})$
- e. $\forall w \neg \text{eats}(w, z) \vee \text{eats}(\text{Harry}, w)$
- f. $\forall g \neg [\text{killed}(g)] \vee \text{alive}(g)$
- g. $\forall k \neg \text{alive}(k) \vee \neg \text{killed}(k)$
- h. $\text{likes}(\text{John}, \text{Peanuts})$

4) Drop Universal Quantification

- a. $\neg \text{food}(x) \vee \text{likes}(\text{John}, x)$
- b. $\text{food}(\text{Apples})$
- c. $\text{food}(\text{vegetables})$
- d. $\neg \text{eats}(y, z) \vee \text{killed}(y) \vee \text{food}(z)$
- e. $\text{eats}(\text{Anil}, \text{Peanut})$
- f. $\text{alive}(\text{Anil})$
- g. $\neg \text{eats}(w, z) \vee \text{eats}(\text{Harry}, w)$
- h. $\text{killed}(g) \vee \text{alive}(g)$
- i. $\neg \text{alive}(k) \vee \neg \text{killed}(k)$
- j. $\text{likes}(\text{John}, \text{Peanuts})$

Proof by Resolution:

$\neg \text{likes}(\text{John}, \text{Peanuts})$

$\neg \text{food}(x) \vee \text{likes}(\text{John}, x)$

{Peanuts/z}

$\neg \text{food}(\text{Peanuts})$

$\neg \text{eats}(y, z) \vee \text{killed}(y) \vee \text{food}(z)$

{Peanuts/z}

$\neg \text{eats}(y, \text{Peanuts}) \vee \text{killed}(y) \quad \text{eats}(\text{Anil}, \text{Peanuts})$

{Anil/y}

$\text{killed}(\text{Anil})$

$\neg \text{alive}(k) \vee \neg \text{killed}(k)$

{Anil/k}

$\neg \text{alive}(\text{Anil})$

$\text{alive}(\text{Anil})$

{3} Hence proved

Implement Alpha-Beta Pruning Algorithm

$\text{Alpha}(\alpha) - \text{Beta}(\beta)$ process to find the optimal path without looking at every node in the game tree.

max contains $\text{Alpha}(\alpha)$ and min contains $\text{Beta}(\beta)$ bound during the calculation

In both MIN and MAX node, we return $\alpha > \beta$, which compares with its parent node only

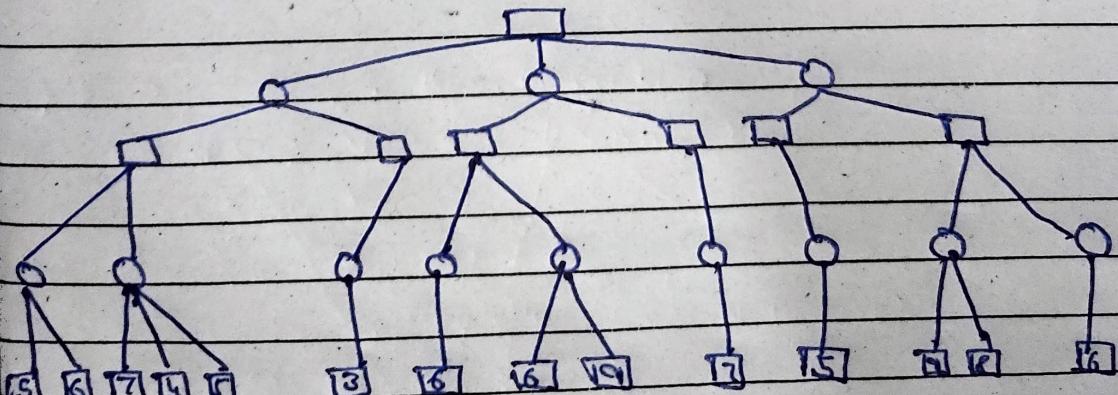
Both minimax & $\text{Alpha}(\alpha) - \text{Beta}(\beta)$ cut-off give same path.

$\text{Alpha}(\alpha) - \text{Beta}(\beta)$ gives optimal solution as it takes less time to get the value for the root node

MAX

MIN

Input:



Output : Max Min

