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Section: K2



SCHOOL OF COMPUTING

**FACULTY OF ENGINEERING AND TECHNOLOGY SRM INSTITUTE
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Chengalpattu District

DEPARTMENT OF NETWORKING AND COMMUNICATIONS

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Exp 1: Implementation of toy problems

Aim: To design program to solve camel and banana toy problem

Pseudocode :

```
bananas = 3000 position = 0 carrying = 0 while position
< 3000: if carrying > 0 and carrying * 1000 + position
>= 3000: position += carrying * 1000 carrying = 0
    if carrying == 0 and bananas > 0: bananas -= 1
    if carrying == 0 and position < 3000:
        carrying = 1000
    if carrying > 0 and carrying * 1000 + position >= 3000: position
        += carrying * 1000
        carrying = 0
    position += 1
```

Optimisation Steps:

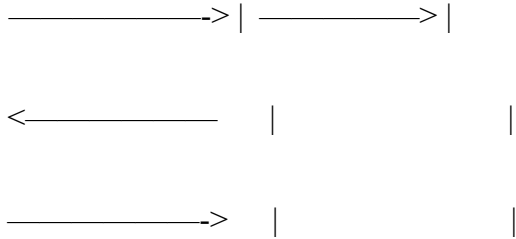
We have a total of 3000 bananas.
The destination is 1000KMs Only 1
mode of transport.

Camel can carry a maximum of 1000 bananas at a time. Camel eats
a banana every km it travels.

Source—————IP1—————IP2—————Destination

3000 x km 2000 y km 1000 z km

—————> | —————> | —————>
<————— | <————— |



$3000 - 5x = 2000$ so we get $x = 200$

$2000 - 3y = 1000$ so we get $y = 333.33$ but here the distance is also the number of bananas and it cannot be fraction so we take $y = 333$ and at IP2 we have the number of bananas equal 1001, so its $2000 - 3y = 1001$

So the remaining distance to the market is $1000 - x - y = z$ i.e $1000 - 200 - 333 \Rightarrow z = 467$.

Now, there are 1001 bananas at IP2. However the camel can carry only 1000 bananas at a time, so we need to leave one banana behind.

So from IP2 to the destination point the camel eats 467 bananas. The remaining bananas

are $1000 - 467 = 533$ **Output:**

```
Enter no. of bananas at starting of trip : 3000
Enter distance you want camel to cover : 1000
Enter max load capacity of your camel : 1000

Total banana delivered after 1000kms : 533

...Program finished with exit code 0
Press ENTER to exit console.█
```

Result: The Total Bananas Delivered were 533 i.e 17.7 % efficiency

Exp 2: Developing agent programs for real world problems

Aim: To Develop agent program for graph coloring problem

Pseudocode: function graph_coloring(Graph

G, int k):

 // G is the graph to be colored

 // k is the maximum number of colors to be used

 for each vertex v in G:

 v.color = null // Initialize all vertex colors to

 null for each vertex v in G: if v.color == null:

 // Use a set to keep track of the colors used by neighboring vertices

 used_colors = set()

 // Loop over neighboring vertices and add their colors to

 used_colors for each neighboring vertex u of v: if u.color is not null:

 used_colors.add(u.color)

 // Find the first color not in used_colors

 for i in range(k):

 if i not in used_colors:

 v.color = i break return G //

Return the colored graph

Optimisation Steps:

Lets take :

1---2

/\ \

3---4---5

1:null---2:null

| |

3:null---4:null---5:null

1:0---2:null

| |

3:null---4:null---5:null

v

used_colors = {0}

1:0---2:null

| |

3:null---4:null---5:null

v

used_colors = {}

4:1

1:0---2:null

| |

3:null---4:1---5:null

v

used_colors = {1}

1:0---2:null

| |

3:2---4:1---5:null

v

used_colors = {1, 2}

1:0---2:2

| |

3:2---4:1---5:null

v

used_colors = {1, 2, 0}

```

1:0---2:2
|      |
3:2---4:1---5:0 v
used_colors =
{0} 1:0---2:2
|      |
3:2---4:1---5:0

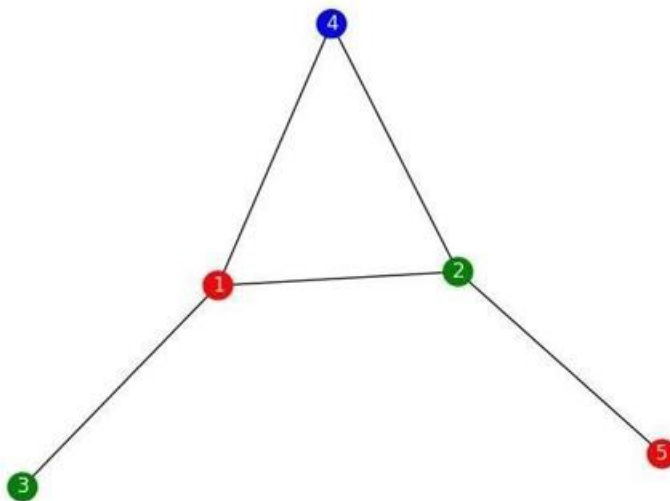
```

The chromatic number is 3 **Output:**

```

Enter the number of vertices: 5
Enter the number of edges: 5
Enter the vertices for edge 1 (separated by a space): 1 2
Enter the vertices for edge 2 (separated by a space): 1 3
Enter the vertices for edge 3 (separated by a space): 1 4
Enter the vertices for edge 4 (separated by a space): 2 4
Enter the vertices for edge 5 (separated by a space): 2 5

```



Result: The Chromatic number is 3

Exp 3: Implementation of constraint satisfaction problems

Aim: To design a constraint satisfaction problem such as cryptarithmic and sudoku solver

Pseudocode :

Cryptarithmic:

```
function solve_crypt_arithmetic(puzzle) : # Split the
    puzzle into the operands and the result operands =
    puzzle[:len(puzzle)-2].split("+") result =
    puzzle[len(puzzle)-1]

    # Get the unique letters from the puzzle letters =
    set(puzzle.replace("+", "").replace("=", ""))

    # Generate all possible combinations of digits for the unique letters
    digits = [i for i in range(10)] digit_combinations =
    itertools.permutations(digits, len(letters))

    # Check each combination of digits to see if it solves the
    puzzle for combination in digit_combinations: digit_map =
    dict(zip(letters, combination))

        # Check if the combination results in a valid solution if
        sum([int("".join([str(digit_map[char]) for char in op])) for op in operands]) ==
int("".join([str(digit_map[char]) for char in result])):
            return digit_map

    # If no solution is found, return None
```

return None **Sudoku solver:**

```
function solve_sudoku(board) # Find the next
    empty cell on the board row, col =
    find_empty_cell(board)
    # If there are no more empty cells, the board is solved
    if row is None: return True
```

```

# Try filling the cell with each number from 1 to 9 for
num in range(1, 10):
    # Check if the number is a valid choice for the
    cell if is_valid_choice(board, row, col, num): #
    Fill the cell with the number board[row][col] =
    num

    # Recursively try to solve the board with the new number added
    if solve_sudoku(board): return True

    # If the new number does not lead to a solution, backtrack and try the next number
    board[row][col] = 0

# If none of the numbers lead to a solution, return False
return False

```

Optimisation Steps:

H E R E

+ S H E

C O M E S

H

+ _

C O

$H + 1 = 9 + 1 = 0$ (1 carry to next step)

9 E R E

+ S 9 E

1 0 M E S

Now, they have already given S value as 8

Now, $E + E = 8$, thus, $E = 4$

Let's find value R

$R + 9 = E$ after substituting E value R
 $+ 9 = 4$ Thus, $R = 5$ and (1 carry to
next step) Lets, find the value of M

$E + S + 1$ (Carry) = M after substituting the value $4 + 8 + 1 =$
3 (1 carry to next step) The final values are –
 $H = 9, E = 4, R = 5, S = 8, C = 1, O = 0, M = 3$

Output :

```
+-----+-----+-----+
| 7 | 4 | 8 | 6 | 3 | 5 | 2 | 9 | 1 |
| 1 | 9 | 3 | 4 | 7 | 2 | 6 | 5 | 8 |
| 6 | 5 | 2 | 8 | 1 | 9 | 4 | 3 | 7 |
+-----+-----+-----+
| 2 | 6 | 5 | 9 | 4 | 8 | 7 | 1 | 3 |
| 8 | 7 | 9 | 1 | 2 | 3 | 5 | 4 | 6 |
| 3 | 1 | 4 | 7 | 5 | 6 | 8 | 2 | 9 |
+-----+-----+-----+
| 9 | 2 | 7 | 3 | 6 | 4 | 1 | 8 | 5 |
| 5 | 3 | 6 | 2 | 8 | 1 | 9 | 7 | 4 |
| 4 | 8 | 1 | 5 | 9 | 7 | 3 | 6 | 2 |
+-----+-----+-----+
```

```
Enter the number of elements in the array: 2
Enter element 1: HERE
Enter element 2: SHE
Enter the result: COMES
```

```
-----
HERE  5939
SHE   859
COMES 06798
-----
```

```
-----
HERE  9454
SHE   894
COMES 10348
-----
```

```
-----
HERE  2858
SHE   628
COMES 03486
-----
```

```
-----
HERE  3959
SHE   839
COMES 04798
-----
```

```
-----
HERE  2969
SHE   829
COMES 03798
-----
```

```
-----
HERE  5494
SHE   854
COMES 06348
-----
```

```
...Program finished with exit code 0
Press ENTER to exit console.
```

Result: Multiple solutions were found for constraint satisfaction programs.

Exp 4: Implementation and Analysis of DFS and BFS for an application

Aim: To design and implement Breadth first search and Depth first search

Pseudocode :

BFS: create a queue Q
mark v as
visited and put v into Q while Q is non-empty
remove the head u of Q mark and enqueue all (unvisited)
neighbours of u

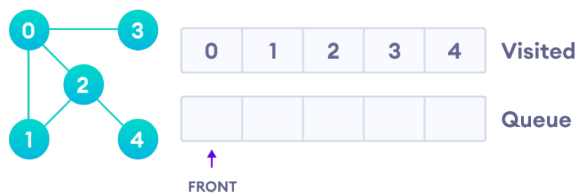
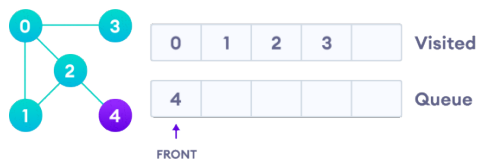
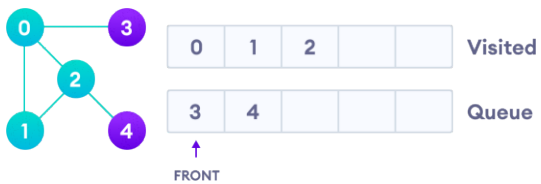
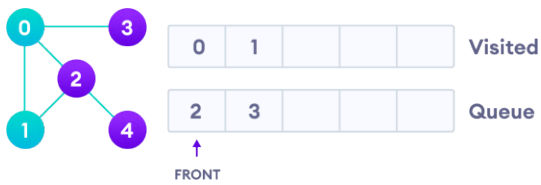
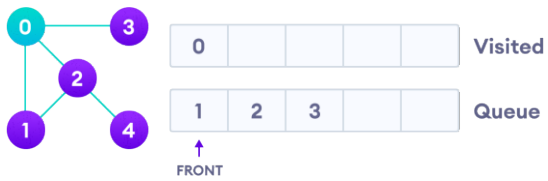
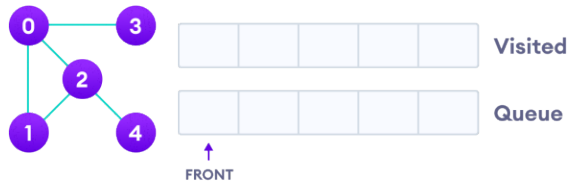
DFS:

DFS(G, u)
 u.visited = true for
 each $v \in G.Adj[u]$ if
 v.visited == false
 DFS(G,v)

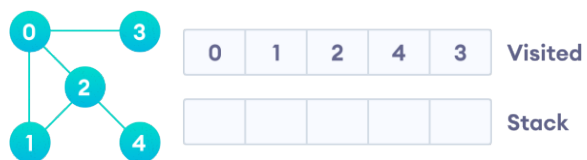
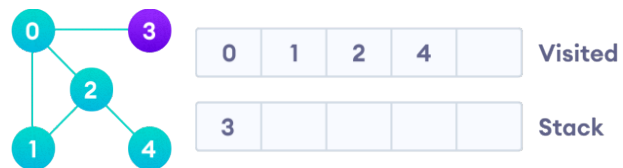
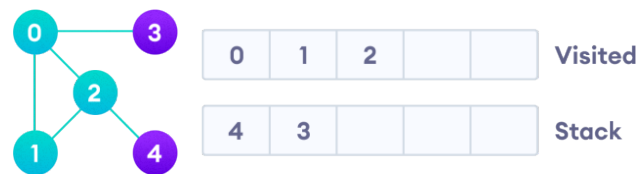
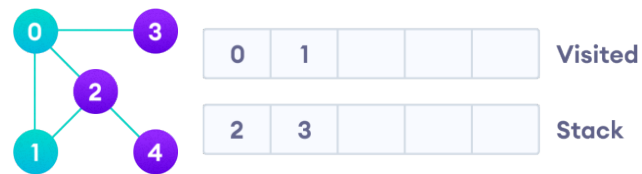
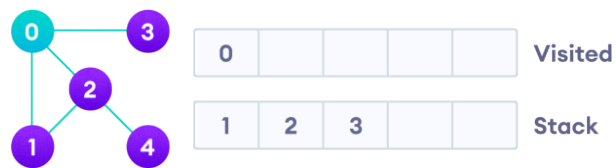
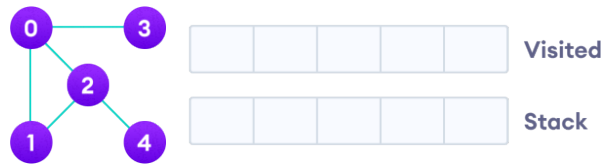
init() {
 For each $u \in G$
 u.visited = false
 For each $u \in G$
 DFS(G, u)
}

Optimisation Steps:

BFS:



DFS:



Output:

```
Enter the tree as a list of edges, with each edge represented by a pair of nodes separated by a space (e.g. '1 2'). Enter 'done' when finished:  
0 1  
0 3  
0 2  
1 2  
2 4  
done  
Enter the target node value: 4  
DFS is faster than BFS by 0.14 ms
```

Result: DFS is faster than BFS by 0.14ms

Exp 5: Developing Best first search and A* Algorithm for real world problems

Aim: To design best first search and A* program

Pseudocode :

Best first search: function best_first_search(initial_state, goal_test, successors, heuristic) returns solution or failure
frontier <- Priority Queue containing initial_state
explored <- empty set

```
while not frontier.empty() do state
  <- frontier.pop() if goal_test(state) then return solution
  explored.add(state) for successor in successors(state) do if
    successor not in explored and successor not in frontier then
      frontier.push(successor, heuristic(successor))
      else if successor in frontier and heuristic(successor) < frontier.get_priority(successor)
then frontier.replace(successor, heuristic(successor))
return failure
```

A*:

Input: A graph $G(V,E)$ with source node $start$ and goal node end .

Output: Least cost path from $start$ to end .

Steps:

Initialise

$open_list = \{ start \}$	<i>/* List of nodes to be traversed */</i>
$closed_list = \{ \}$	<i>/* List of already traversed nodes */</i>
$g(start) = 0$	<i>/* Cost from source node to a node */</i>
$h(start) = heuristic_function(start, end)$	<i>/* Estimated cost from node to goal node */</i>
$f(start) = g(start) + h(start)$	<i>/* Total cost from source to goal node */</i>

while $open_list$ is not empty

$m =$ Node on top of $open_list$, with least f

if $m == end$

return

remove m from $open_list$

add m to $closed_list$

for each n in $child(m)$

if n in $closed_list$

continue

$cost = g(m) + distance(m, n)$

if n in $open_list$ and $cost < g(n)$

remove n from $open_list$ as new path is better

if n in $closed_list$ and $cost < g(n)$

remove n from $closed_list$

if n not in $open_list$ and n not in $closed_list$

add n to $open_list$

$g(n) = cost$

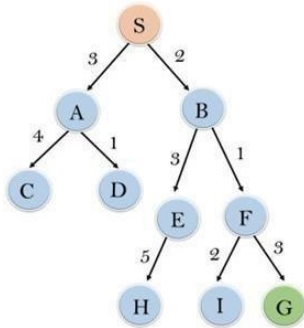
$h(n) = heuristic_function(n, end)$

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

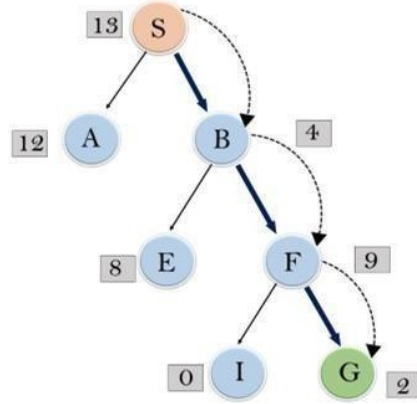
return failure

Optimisation Steps:

Best first search



node	H (n)
A	12
B	4
C	7
D	3
E	8
F	2
H	4
I	9
S	13
G	0



Initialization: Open [A, B], Closed [S]

Iteration 1: Open [A], Closed [S, B]

Iteration 2: Open [E, F, A], Closed [S, B] :

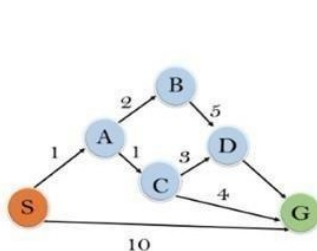
Open [E, A], Closed [S, B, F]

Iteration 3: Open [I, G, E, A], Closed [S, B, F]

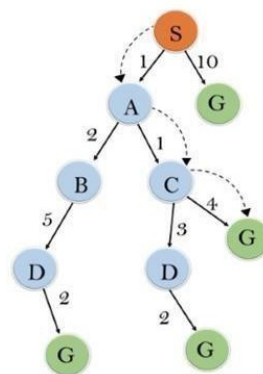
: Open [I, E, A], Closed [S, B, F, G]

Hence the final solution path will be: S----> B----->F-----> G

The worst case time complexity of Greedy best first search is $O(bm)$. **A***



State	h(n)
S	5
A	3
B	4
C	2
D	6
G	0



Initialization: {(S, 5)}

Iteration1: {(S--> A, 4), (S-->G, 10)}

Iteration2: {(S--> A-->C, 4), (S--> A-->B, 7), (S-->G, 10)}

Iteration3: {(S--> A-->C--->G, 6), (S--> A-->C--->D, 11), (S--> A-->B, 7), (S-->G, 10)} Iteration

4 will give the final result, as S--->A--->C--->G it provides the optimal path with cost 6

Output:

```
Enter the number of nodes: 5
Enter node 1: A
Enter neighbors of A (comma-separated): B,C
Enter node 2: B
Enter neighbors of B (comma-separated): A,C,D
Enter node 3: C
Enter neighbors of C (comma-separated): A,B,D,E
Enter node 4: D
Enter neighbors of D (comma-separated): B,C,E
Enter node 5: E
Enter neighbors of E (comma-separated): C,D
Enter node 1: A
Enter heuristic value for A: 8
Enter node 2: B
Enter heuristic value for B: 6
Enter node 3: C
Enter heuristic value for C: 4
Enter node 4: D
Enter heuristic value for D: 4
Enter node 5: E
Enter heuristic value for E: 2
Enter the start node: A
Enter the goal node: E
```

```
Goal node E found!  
Path: A -> C -> E
```

```
Enter the number of nodes: 5  
Enter node 1: A  
Enter neighbors of A (comma-separated): B,D  
Enter node 2: B  
Enter neighbors of B (comma-separated): A,C  
Enter node 3: C  
Enter neighbors of C (comma-separated): B,D,E  
Enter node 4: D  
Enter neighbors of D (comma-separated): A,C,E  
Enter node 5: E  
Enter neighbors of E (comma-separated): C,D  
Enter node 1: A  
Enter heuristic value for A: 8  
Enter node 2: B  
Enter heuristic value for B: 4  
Enter node 3: C  
Enter heuristic value for C: 3  
Enter node 4: D  
Enter heuristic value for D: 3  
Enter node 5: E  
Enter heuristic value for E: 0  
Enter the start node: A  
Enter the goal node: E  
Goal node E found!  
Path: A -> D -> E  
  
...Program finished with exit code 0  
Press ENTER to exit console.
```

Result: The worst case time complexity of Greedy best first search is $O(bm)$. The time complexity of A* search algorithm is $O(b^d)$, where b is the branching factor.

Exp 6: Implementation of minimax algorithm for an application

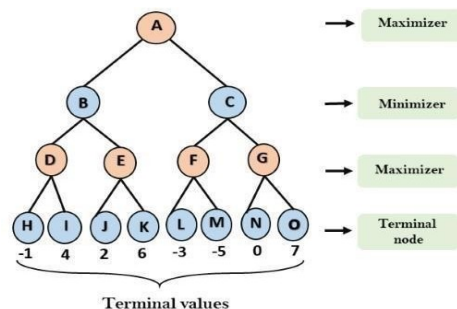
Aim: To implement minimax algorithm

Pseudocode:

```
function minimax(node, depth, maximizingPlayer)
    if depth == 0 or node is a terminal node return the
        heuristic value of node if maximizingPlayer
    bestValue = -infinity
    for each child of node
        v = minimax(child, depth - 1, FALSE)
        bestValue = max(bestValue, v)
    return bestValue

else // minimizing player
    bestValue = +infinity
    for each child of node
        v = minimax(child, depth - 1, TRUE)
        bestValue = min(bestValue, v)
    return bestValue
```

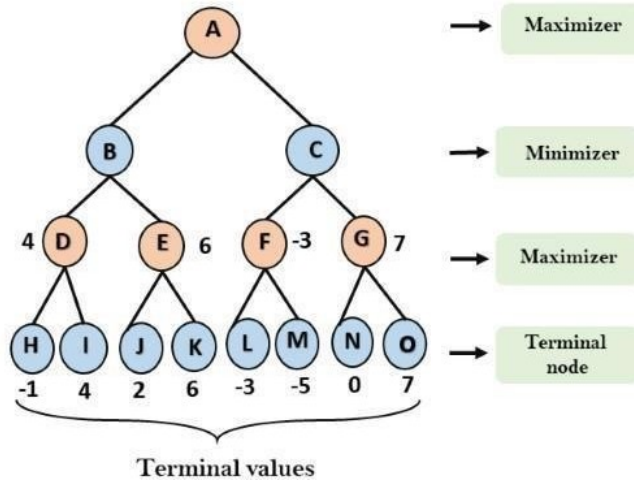
Optimisation Steps :



For node D $\max(-1, -\infty) \Rightarrow \max(-1, 4) = 4$

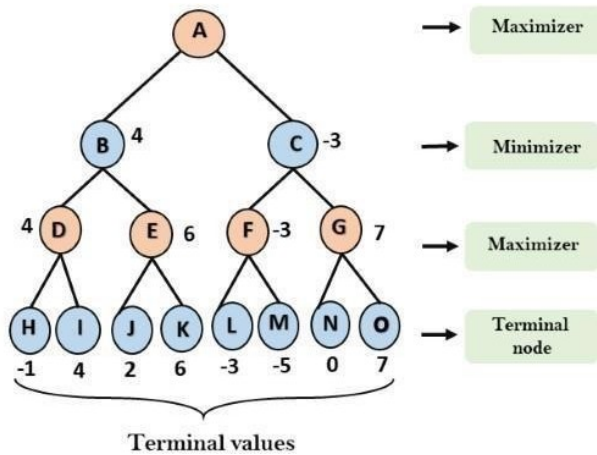
For Node E $\max(2, -\infty) \Rightarrow \max(2, 6) = 6$

For Node F $\max(-3, -\infty) \Rightarrow \max(-3, -5) = -3$ For node G $\max(0, -\infty) = \max(0, 7) = 7$

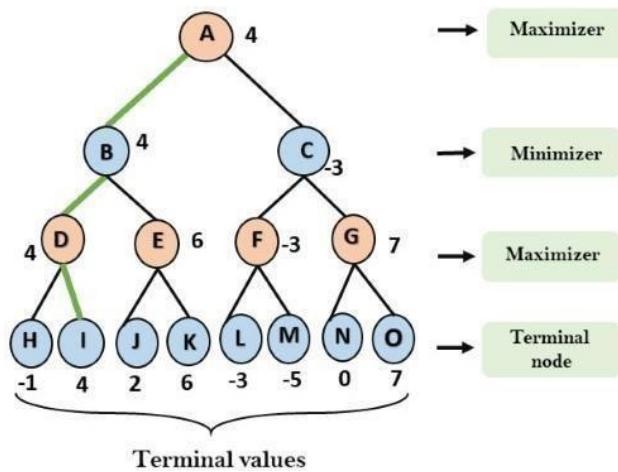


For node B $= \min(4, 6) = 4$

For node C $= \min(-3, 7) = -3$



For node A $\max(4, -3) = 4$



Output:

```

main.py
106
107
108
109
110
111
112
113
114
115
116
117
118
119
120
121
122

def analyzeboard(board):
    if (x==0):
        ConstBoard(board);
        print("Draw!!!")
    if (x==1):
        ConstBoard(board);
        print("X Wins!!! Y Loose !!!")
    if (x==2):
        ConstBoard(board);
        print("X Loose!!! O Wins !!!!")

    #-----#
    main()
    #-----#

def User2Turn(board):
    pass

def ConstBoard(board):
    for i in range(3):
        for j in range(3):
            print(board[i][j], end=" ")
        print()

x=0
y=0
board=[['X','X','O'],
        ['O','O','X'],
        ['X','X','O']]

main()

Enter X's position from [1..9]: 3
Current State Of Board :

X O X
O O X
X X O

Draw!!!
  
```

Result: The minimax algorithm was designed and tic tac toe game was designed

Exp 7: Implementation of unification and resolution for real world problems

Aim: To implement unification and resolution of in real life problem

Pseudocode :

Unification:

If Ψ_1 or Ψ_2 is a variable or constant, then:

 If Ψ_1 or Ψ_2 are identical, then return NIL.

 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.

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

If Ψ_1 and Ψ_2 have a different number of arguments, then return FAILURE.

Set Substitution set(SUBST) to NIL.

For $i=1$ to the number of elements in Ψ_1 .

 Call Unify function with the i th element of Ψ_1 and i th element of Ψ_2 , and put the result into S.

 If S = failure then returns Failure

 If $S \neq \text{NIL}$ then do,

 Apply S to the remainder of both L1 and L2.

 SUBST= APPEND(S, SUBST). Return

SUBST.

Resolution:

function resolution(KB, α): clauses

 = $\text{KB} \wedge \neg\alpha$ new

 = $\{ \}$ while True: for

c_i in clauses: for c_j in

clauses: if c_i is not

c_j :

resolvents = resolve(c_i, c_j) if

False in resolvents:

return True

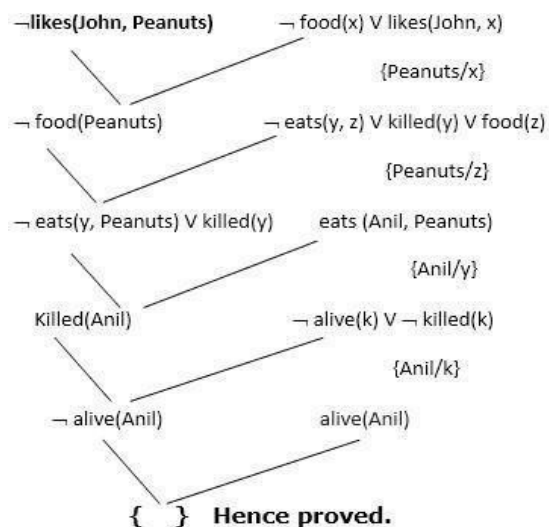
new = new \cup resolvents if

new is subset of clauses: return

False clauses = clauses \cup new

Optimisation Steps:

- 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. $\forall x: \text{eats}(\text{Anil}, x) \rightarrow \text{eats}(\text{Harry}, x)$
- f. $\forall x: \neg \text{killed}(x) \rightarrow \text{alive}(x)$ } **added predicates.**
- g. $\forall x: \text{alive}(x) \rightarrow \neg \text{killed}(x)$ }
- h. $\text{likes}(\text{John}, \text{Peanuts})$



1. Find the MGU of $\{p(f(a), g(Y)) \text{ and } p(X, X)\}$

Sol: $S0 \Rightarrow$ Here, $\Psi1 = p(f(a), g(Y))$, and $\Psi2 = p(X, X)$

$SUBST \theta = \{f(a) / X\}$

$S1 \Rightarrow \Psi1 = p(f(a), g(Y))$, and $\Psi2 = p(f(a), f(a))$

$SUBST \theta = \{f(a) / g(y)\}$, Unification failed.

Unification is not possible for these expressions.

2. Find the MGU of $\{p(b, X, f(g(Z))) \text{ and } p(Z, f(Y), f(Y))\}$

$S0 \Rightarrow \{p(b, X, f(g(Z))) ; p(Z, f(Y), f(Y))\}$

$SUBST \theta = \{b/Z\}$

$S1 \Rightarrow \{p(b, X, f(g(b))) ; p(b, f(Y), f(Y))\}$

$SUBST \theta = \{f(Y) / X\}$

$S2 \Rightarrow \{p(b, f(Y), f(g(b))) ; p(b, f(Y), f(Y))\}$ $SUBST$

$\theta = \{g(b) / Y\}$

$S2 \Rightarrow \{p(b, f(g(b)), f(g(b))) ; p(b, f(g(b)), f(g(b)))\}$ Unified Successfully. And
Unifier = $\{b/Z, f(Y) / X, g(b) / Y\}$.

3. Find the MGU of $\{p(X, X), \text{ and } p(Z, f(Z))\}$

Here, $\Psi1 = \{p(X, X)$, and $\Psi2 = p(Z, f(Z))$

$S0 \Rightarrow \{p(X, X), p(Z, f(Z))\}$

$SUBST \theta = \{X/Z\}$

$S1 \Rightarrow \{p(Z, Z), p(Z, f(Z))\}$

$SUBST \theta = \{f(Z) / Z\}$, Unification Failed.

Therefore, unification is not possible for these expressions.

5. Find the MGU of $Q(a, g(x, a), f(y)), Q(a, g(f(b), a), x)$ Here, $\Psi1 = Q(a, g(x, a), f(y))$, and $\Psi2 = Q(a, g(f(b), a), x)$

$S0 \Rightarrow \{Q(a, g(x, a), f(y)) ; Q(a, g(f(b), a), x)\}$

$SUBST \theta = \{f(b)/x\}$

$S1 \Rightarrow \{Q(a, g(f(b), a), f(y)) ; Q(a, g(f(b), a), f(b))\}$

$SUBST \theta = \{b/y\}$

SUBST $\theta = \{f(Y)/X\}$

$S2 \Rightarrow \{p(b, f(Y), f(g(b))); p(b, f(Y), f(Y))\}$

$S1 \Rightarrow \{Q(a, g(f(b), a), f(b)); Q(a, g(f(b), a), f(b))\}$, Successfully Unified.

Unifier: $[a/a, f(b)/x, b/y]$.

6. UNIFY(knows(Richard, x), knows(Richard, John))

Here, $\Psi1 = \text{knows(Richard, x)}$, and $\Psi2 = \text{knows(Richard, John)}$

$S0 \Rightarrow \{ \text{knows(Richard, x)}; \text{knows(Richard, John)} \}$

S SUBST $\theta = \{John/x\}$

$S1 \Rightarrow \{ \text{knows(Richard, John)}; \text{knows(Richard, John)} \}$, Successfully Unified.

Unifier: $\{John/x\}$.

Output:

```
top/MISC/Codes/aiexp7.py
John does not like Italian food
```

Result: Implementation was successful and output was obtained

Exp 8: Implementation of knowledge representation schemes

Aim: To construct knowledge base and apply inference in FOL , Generate query based family relationship

Pseudocode:

```
person = [] parent = [] male = [] female = []
mother(x, y): return True if x is the mother of y and False otherwise.
father(x, y): return True if x is the father of y and False otherwise.
grandparent(x, y): return True if x is a grandparent of y and False otherwise.
sibling(x, y): return True if x and y are siblings and False otherwise.
aunt(x, y): return True if x is the aunt of y and False otherwise.
uncle(x, y): return True if x is the uncle of y and False otherwise.
cousin(x, y): return True if x and y are cousins and False otherwise.
```

Optimisation Steps:

```
Enter the number of people: 5
Enter the name of person 1: Alice
Enter the gender of Alice (M/F): F
Enter the name of person 2: Bob
Enter the gender of Bob (M/F): M
Enter the name of person 3: Carol
Enter the gender of Carol (M/F): F
Enter the name of person 4: David
Enter the gender of David (M/F): M
Enter the name of person 5: Eve
Enter the gender of Eve (M/F): F
person = ['Alice',
```

```
'Bob', 'Carol', 'David', 'Eve'] male = ['Bob',
'David'] female = ['Alice',
'Carol', 'Eve']
Enter the name of parent 1: Alice
Enter the name of child 1: Bob
Enter the name of parent 2: Alice
Enter the name of child 2: Carol
Enter the name of parent 3: Bob
Enter the name of child 3: David
Enter the name of parent 4: Carol Enter the name of child 4: Eve parent
= [('Alice', 'Bob'), ('Alice', 'Carol'), ('Bob', 'David'), ('Carol', 'Eve')]
```

Output:

```
Enter the name of person 1: Alice
Enter the gender of Alice (M/F): F
Enter the name of person 2: Bob
Enter the gender of Bob (M/F): M
Enter the name of person 3: Carol
Enter the gender of Carol (M/F): F
Enter the name of person 4: David
Enter the gender of David (M/F): M
Enter the name of person 5: Eve
Enter the gender of Eve (M/F): F
```

```
Enter the name of parent 1: Alice
Enter the name of child 1: Bob
Enter the name of parent 2: Alice
Enter the name of child 2: Carol
Enter the name of parent 3: Bob
Enter the name of child 3: David
Enter the name of parent 4: Carol
Enter the name of child 4: Eve
```

```
Alice is the mother of Carol
Alice is the parent of Carol
Carol is the child of Bob
Eve is the parent of Frank
Dave is the parent of Frank
Dave is the father of Frank
```

Result: The result was obtained.

Exp 9: Implementation of uncertain methods for an application

Aim: To implement fuzzy logic in Monty Hall Problem

Pseudocode:

```
import random
```

```
A = "A"
```

```
B = "B"
```

```
C = "C"
```

```
doors = ["A", "B", "C"]
```

```
prize = random.choice(doors)
```

```
selection = input("Select door 'A', 'B', or 'C': ")
```

```
print("This problem relies on conditional probabilities.") print("It is suggested that you switch
```

```
doors, you will have a higher probability of winning if you
```

```
do.")
```

```

if selection == prize:    remaining =

list(set(doors) - set(selection))

    open_door = random.choice(list(set(doors) - set(prize) - set(random.choice(remaining))))

    alternate = list(set(doors) - set(open_door) - set(selection))[0] else:
    open_door = random.choice(list(set(doors) - set(selection) - set(prize)))

    alternate = list(set(doors) - set(open_door) - set(selection))[0]

print("The door I will now open is: %r" % open_door)


second_chance = input("Would you like to select the third door? Type 'Yes' or 'No': ")

if second_chance == "Yes":    print("The door you will

switch to is: %r" % alternate)

    if alternate == prize:

        print("Congrats, you win! The prize was behind your original selection, %r" % selection)

    else:        print("Sorry, the prize was behind the original door %r"
% prize)

if second_chance != "Yes":    print("You decided to keep your

```



```
initial door, %r" % selection)
```

```
if selection != prize:
```

```
    print("Sorry, the prize was behind the original door, %r" % prize)
```

```
else:
```

```
    print("Congrats, you win! The prize was behind your original selection, %r" % selection)
```

```
print("This is a check:")
```

```
print("Prize: %r" % prize)
```

```
print("Selection: %r " % selection)
```

```
print("Alternate: %r " % alternate) print("Door
```

```
opened: %r " % open_door)
```

Output:

```
main.py
32 def prize():
33     print("Congrats, you win! The prize was behind your original selection, %r" % selection)
34 else:
35     print("Sorry, the prize was behind the original door %r" % prize)
36
37 if second_chance != "Yes":
38     print("You decided to keep your initial door, %r" % selection)
39     if selection != prize:
40         print("Sorry, the prize was behind the original door, %r" % prize)
41     else:
42         print("Congrats, you win! The prize was behind your original selection, %r" % selection)
43
44 print("This is a check:")
45 print("Prize: %r" % prize)
46 print("Selection: %r " % selection)
47 print("Alternate: %r " % alternate)
48 print("Door opened: %r " % open_door)
```

Select door 'A', 'B', or 'C': C
This problem relies on conditional probabilities.
It is suggested that you switch doors, you will have a higher probability of winning if you do.
The door I will now open is: 'A'
Would you like to select the third door? Type 'Yes' or 'No': NO
You decided to keep your initial door, 'C'
Congrats, you win! The prize was behind your original selection, 'C'
This is a check:
Prize: 'C'
Selection: 'C'
Alternate: 'B'
Door opened: 'A'

Result: The chlorine level is 8.0

Exp 10: Implementation of block world problem

Aim: To implement block world problem

Pseudocode:

```
world_state = initialize_world_state() actions
= ['MOVE', 'STACK', 'UNSTACK']
def is_valid_action(action, state):
    # check if the action is valid based on the current state of the
world Return true def perform_action(action, state):
    # perform the given action on the world state
Return (block,source,target,state) goal_state =
define_goal_state()
    # perform search to find a sequence of actions that transforms the initial state into the goal state
search_alg(world_state, goal_state, actions, is_valid_action, perform_action) //heap run_time =
str(end-start)*1000
```

Optimisation Steps :

Initial state:

Blocks: A, B, C, D

On table: A

On blocks: B on A, C on B, D on C Goal

state:

Blocks: A, B, C, D

On table: C, B

On blocks: A on C, B on D

Pick up block D and place it on the table. Pick up block C and place it on the table.

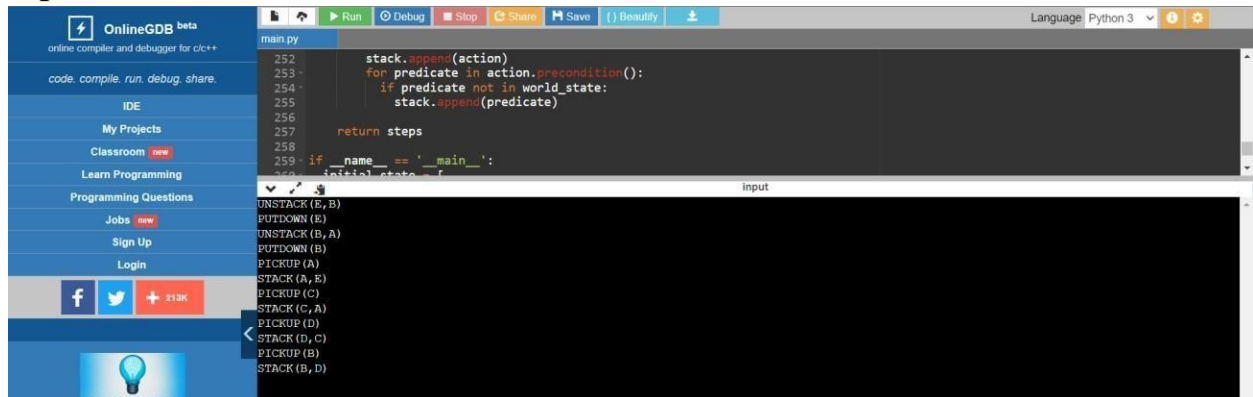
Pick up block B and place it on top of block D. Pick

up block C and place it on top of block A. Pick up

block A and place it on top of block C.

Pick up block B and place it on top of block D.

Output:



The screenshot shows the OnlineGDB IDE interface. The top bar includes buttons for Run, Debug, Stop, Share, Save, and Beautify, along with a language selector set to Python 3. The left sidebar contains navigation links for IDE, My Projects, Classroom, Learn Programming, Programming Questions, Jobs, Sign Up, and Login. The main editor displays a Python script with the following code:

```
252 stack.append(action)
253 for predicate in action.precondition():
254     if predicate not in world_state:
255         stack.append(predicate)
256
257 return steps
258
259 if __name__ == '__main__':
260     initial_state = f
```

The output window at the bottom shows the following sequence of actions:

```
UNSTACK (E, B)
PUTDOWN (E)
UNSTACK (B, A)
PUTDOWN (B)
PICKUP (A)
STACK (A, E)
PICKUP (C)
STACK (C, A)
PICKUP (D)
STACK (D, C)
PICKUP (B)
STACK (B, D)
```

Result: The block world problem was implemented with running time of 5.08618 ms