

```

Knowledge_Base = {
    frozenset({'Mother', 'Leela', 'Oshin'}),
    frozenset({'Alive', 'Leela'}),
    frozenset({'not', 'Mother', 'x', 'y'}),
    frozenset({'Parent', 'x', 'y'}),
    frozenset({'not', 'Parent', 'w', 'z'}),
    frozenset({'not', 'Alive', 'w', 'z'}),
    frozenset({'Older', 'w', 'z'}),
}

```

```

query = ('Older', 'Leela', 'Older')
result = proof_by_resolution(Knowledge_Base, query)
if result:
    print("Leela is older than Oshin.\nProved by resolution.")
else:
    print("Cannot prove. Leela is not older than Oshin.")

```

Output Snapshot:

```

Leela is older than Oshin.
Proved by resolution.

```

Program 10:

Implement Alpha-Beta Pruning.

Algorithm:

Lab - 11

classmate

Write algorithm pseudocode. Alpha-Beta Pruning search algorithm.

Function alpha-beta (node, depth, alpha, beta, maxing, path):

if depth is 0 or node is a terminal node then
return node's value, path.

if maximizing-player then
maxeval = negative infinity
optimal-path = null
for each child of node do
child-value, child-path = alpha-beta (child, depth+1, alpha, beta, false, path + child's name)
if child-value > maxeval then
maxeval = child-value
optimal-path = child-path
alpha = maximum(alpha, maxeval)
if beta <= alpha then
break
return maxeval, optimal-path
else
minval = positive infinity
for each child of node do
child-value, child-path = alpha-beta (child, depth+1, alpha, beta, True, path + child's name)

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if childval < min_eval then
    min_eval = child-value
    optimal_path = child-path

beta = min(beta, min_eval)
if beta <= alpha then
    break
return min_eval, optimal-path

maximizing_player = True
initial_alpha = negative inf
initial_beta = positive inf
depth = 3

optimal_val, optimal_path = alpha_beta (root,
depth, initial_alpha, initial_beta, maximizing_player)

print (optimal_val)
print (optimal_path)

```

Amrith

Code:

```
class Node:
    def __init__(self, value=None, children=None):
        self.value = value
        self.children = children if children else []

def alpha_beta_pruning(node, depth, alpha, beta, maximizing_player):
    if not node.children or depth == 0:
        return node.value

    if maximizing_player:
        max_eval = float('-inf')
        for child in node.children:
            eval = alpha_beta_pruning(child, depth - 1, alpha, beta, False)
            max_eval = max(max_eval, eval)
            alpha = max(alpha, eval)
            if beta <= alpha:
                print(f'Pruned at MAX node with alpha={alpha}, beta={beta}')
                break
        node.value = max_eval
        return max_eval
    else:
        min_eval = float('inf')
        for child in node.children:
            eval = alpha_beta_pruning(child, depth - 1, alpha, beta, True)
            min_eval = min(min_eval, eval)
            beta = min(beta, eval)
            if beta <= alpha:
                print(f'Pruned at MIN node with alpha={alpha}, beta={beta}')
                break
        node.value = min_eval
        return min_eval

def print_tree(node, level=0):
    print(" " * level * 2 + f'Value of Node: {node.value}')
    for child in node.children:
        print_tree(child, level + 1)

if __name__ == "__main__":
    tree = Node(None, [
        Node(None, [
            Node(None, [Node(10), Node(9)]),
            Node(None, [Node(14), Node(18)])
        ]),
    ]),
```

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Node(None, [
    Node(None, [Node(5),Node(4)]),
    Node(None, [Node(50),Node(3)])
])

print("Game Tree Before Alpha-Beta Pruning:")
print_tree(tree)
final_value = alpha_beta_pruning(tree, depth=3, alpha=float('-inf'), beta=float('inf'),
maximizing_player=True)
print("\nGame Tree After Alpha-Beta Pruning:")
print_tree(tree)
print("\nFinal Value at MAX node:", final_value)

```

Output Snapshot:

```

Game Tree Before Alpha-Beta Pruning:
Value of Node: None
  Value of Node: None
    Value of Node: None
      Value of Node: 10
      Value of Node: 9
    Value of Node: None
      Value of Node: 14
      Value of Node: 18
  Value of Node: None
    Value of Node: None
      Value of Node: 5
      Value of Node: 4
    Value of Node: None
      Value of Node: 50
      Value of Node: 3
Pruned at MAX node with alpha=14, beta=10
Pruned at MIN node with alpha=10, beta=5

Game Tree After Alpha-Beta Pruning:
Value of Node: 10
  Value of Node: 10
    Value of Node: 10
    Value of Node: 9
  Value of Node: 14
    Value of Node: 14
    Value of Node: 18
  Value of Node: 5
    Value of Node: 5
    Value of Node: 4
  Value of Node: None
    Value of Node: 50
    Value of Node: 3

Final Value at MAX node: 10

```