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Lab Task -4 (N-Queen Problem)

N-Queen Problem Report

Introduction

The N-Queen problem is a classic combinatorial problem in computer science and artificial intelligence. It involves placing N queens on an $N \times N$ chessboard in such a way that no two queens threaten each other. This means that no two queens should share the same row, column, or diagonal. The problem is a generalization of the well-known 8-Queen problem.

Problem Definition

Given an $N \times N$ chessboard, the objective is to place N queens on the board so that:

1. No two queens share the same row.
2. No two queens share the same column.
3. No two queens share the same diagonal (both major and minor diagonals).

Applications of the N-Queen Problem

The N-Queen problem has several practical applications, including:

- **Constraint Satisfaction Problems (CSPs):** Used in artificial intelligence and operations research.
- **Parallel Processing:** Helps in optimizing parallel computing algorithms.
- **VLSI Design:** Used in designing circuits without conflicts.
- **Robotics:** Applied in autonomous decision-making processes.

Solution Approaches

Several methods exist to solve the N-Queen problem efficiently:

1. Backtracking Algorithm

Backtracking is a brute-force recursive approach that places queens one by one in different rows and backtracks when a conflict is encountered.

- It systematically explores possible solutions by placing a queen in a row and recursively solving the subproblem for the next row.
- If a solution is found, it is stored; otherwise, it backtracks and tries a different placement.
- Time Complexity: $O(N!)$.

2. Branch and Bound

This optimization of backtracking reduces the number of possibilities by using bounding functions to eliminate infeasible solutions early.

- It improves efficiency by pruning the search tree.
- Time Complexity: Better than $O(N!)$ but still exponential in nature.

3. Constraint Programming (CSP Approach)

- The problem is modeled as a constraint satisfaction problem.
- Uses constraint propagation techniques to reduce search space.
- Efficient for larger values of N .

4. Genetic Algorithms

- Uses evolutionary techniques to generate and improve solutions over generations.
- Suitable for approximate solutions rather than exact ones.
- Time Complexity varies depending on implementation.

5. Heuristic and Metaheuristic Approaches

- **Hill Climbing:** Places queens randomly and iteratively moves them to minimize conflicts.
- **Simulated Annealing:** Uses probabilistic techniques to escape local optima.

Complexity Analysis

The time complexity of solving the N-Queen problem varies based on the approach used:

- **Backtracking:** $O(N!)$ (worst-case exponential complexity).
- **Branch and Bound:** More optimized than backtracking but still exponential.
- **Heuristic Methods:** Can achieve near-optimal solutions in polynomial time for larger N .

Challenges in the N-Queen Problem

1. **Exponential Growth:** The number of possible board configurations grows exponentially with N .
2. **Large N values:** Finding an exact solution for very large values of N (e.g., $N > 1000$) is computationally expensive.

3. **Memory Constraints:** Recursive methods may lead to excessive memory usage for large N.

Conclusion

The N-Queen problem is a fundamental problem in computer science with significant applications in artificial intelligence and optimization. While brute-force methods provide solutions for small values of N, heuristic and metaheuristic methods are necessary for solving large instances efficiently. The problem continues to be a rich area of research in algorithm design and constraint solving.

Code:

```
1 class nqueen:
2     def __init__(self,n):
3         self.n = n
4         self.board = [-1]*n
5         self.solution = []
6
7     def safe(self,row,col):
8         for pre_row in range(row):
9             pre_col = self.board[pre_row]
10            if pre_col == col or abs(pre_col-col)==abs(pre_row-row):
11                return False
12            return True
13
14    def solve_queen(self,row=0):
15        if row==self.n:
16            self.solution.append(self.board[:])
17            return
18        for col in range(self.n):
19            if self.safe(row,col):
20                self.board[row] = col
21                self.solve_queen(row+1)
22                self.board[row] = -1
23
```

```

23
24     def print_sol(self):
25         for sol in self.solution:
26             for row in range(self.n):
27                 l = ["." for _ in range(self.n)]
28                 l[sol[row]] = 'Q'
29                 print(" ".join(l))
30             print('\n'+ '-'*(2*self.n-1))
31     def solve(self):
32         self.solve_queen()
33         print(f'total solution for {self.n}-Queen:{len(self.solution)}')
34         self.print_sol()
35 if __name__ == '__main__':
36     n = 8
37     game = nqueen(n)
38     game.solve()

```

total solution for 8-Queen:92

```

Q . . . . . . .
. . . . Q . . .
. . . . . . . Q
. . . . . Q . .
. . Q . . . . .
. . . . . . . Q
. Q . . . . . .
. . . Q . . . .

```

```

-----
Q . . . . . . .
. . . . . Q . .
. . . . . . . Q
. . Q . . . . .
. . . . . Q . .
. . . Q . . . .
. Q . . . . . .
. . . . Q . . .

```