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1) Towers of Hanoi (Recursion)
def hanoi(n, source, target, auxiliary, moves):
  """Recursive solution for Towers of Hanoi."""
  if n == 0:
    return
  hanoi(n-1, source, auxiliary, target, moves)
  moves.append((n, source, target))
  hanoi(n-1, auxiliary, target, source, moves)
if __name__ == "__main__":
  n = 4 # number of disks
  moves = []
  hanoi(n, "A", "C", "B", moves)
  print(f"Total moves for {n} disks: {len(moves)}")
  for i, (disk, frm, to) in enumerate(moves, 1):
    print(f"{i}: Move disk {disk} from {frm} -> {to}")
2) Tic-Tac-Toe (Player vs Player)
def print_board(b):
  for row in b:
    print(" | ".join(row))
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print("-"*9)
def check_win(b, player):
  for i in range(3):
    if all(b[i][j]==player for j in range(3)): return True
    if all(b[j][i]==player for j in range(3)): return True
  if b[0][0]==b[1][1]==b[2][2]==player: return True
  if b[0][2]==b[1][1]==b[2][0]==player: return True
  return False
def board_full(b):
  return all(b[i][j] != " " for i in range(3) for j in range(3))
def play():
 board = [[" "]*3 for _ in range(3)]
  current = "X"
  while True:
    print_board(board)
    move = input(f"Player {current}, enter row,col (1-3): ").split()
    if len(move) != 2 or not all(s.isdigit() for s in move):
      print("Invalid input.")
      continue
    r, c = int(move[0])-1, int(move[1])-1
    if not (0 \le r \le 3 \text{ and } 0 \le c \le 3):
      print("Out of range.")
      continue
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if board[r][c] != " ":
     print("Cell occupied.")
      continue
    board[r][c] = current
   if check_win(board, current):
     print_board(board)
     print(f"Player {current} wins!")
     break
   if board_full(board):
     print_board(board)
     print("It's a draw.")
     break
    current = "O" if current == "X" else "X"
if __name__ == "__main__":
 play()
3) N-Queens Problem (Backtracking)
def solve_n_queens(n):
 solutions = []
  cols = set()
  diag1 = set()
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diag2 = set()
  board = [-1]*n
  def backtrack(r):
    if r == n:
      sol = []
      for i in range(n):
        row = ['.']*n
        row[board[i]] = 'Q'
        sol.append("".join(row))
      solutions.append(sol)
      return
    for c in range(n):
      if c in cols or (r-c) in diag1 or (r+c) in diag2:
        continue
      cols.add(c); diag1.add(r-c); diag2.add(r+c); board[r] = c
      backtrack(r+1)
      cols.remove(c); diag1.remove(r-c); diag2.remove(r+c); board[r] = -1
  backtrack(0)
  return solutions
if __name__ == "__main__":
  N = 8
  sols = solve_n_queens(N)
  print(f"Total solutions for N={N}: {len(sols)}")
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if sols:
   print("\n".join(sols[0]))
4) Random Maze Generator
import random
import numpy as np
import matplotlib.pyplot as plt
def make_maze(width, height):
  visual = np.zeros((2*height+1, 2*width+1), dtype=int)
 for i in range(height):
   for j in range(width):
     visual[2*i+1, 2*j+1] = 1
 visited = [[False]*width for _ in range(height)]
 stack = [(0,0)]
 visited[0][0] = True
  while stack:
   r,c = stack[-1]
   neighbors = []
   for dr, dc in [(0,1),(0,-1),(1,0),(-1,0)]:
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nr, nc = r+dr, c+dc
     if 0 <= nr < height and 0 <= nc < width and not visited[nr][nc]:
       neighbors.append((nr, nc, dr, dc))
   if neighbors:
     nr, nc, dr, dc = random.choice(neighbors)
     visual[2*r+1+dr, 2*c+1+dc] = 1
     visited[nr][nc] = True
     stack.append((nr, nc))
    else:
     stack.pop()
  return visual
def show_maze(visual):
  plt.figure(figsize=(6,6))
  plt.imshow(visual == 0, cmap="binary")
  plt.axis('off')
  plt.show()
if __name__ == "__main__":
 W, H = 20, 20
 vis = make_maze(W, H)
  show_maze(vis)
```

2. Model-Based Reflex Agents: Maintain an internal model of the environment to handle partially observable situations.
3. Goal-Based Agents: Choose actions to achieve specific goals, often using search or planning algorithms.
4. Utility-Based Agents: Select actions based on a utility function that measures desirability among multiple possible outcomes.
5. Learning Agents: Improve their performance over time by learning from experience.
Applications of intelligent agents include chatbots, recommendation systems, autonomous vehicles, and industrial robots. With increasing complexity, modern agents integrate planning, probabilistic reasoning, and machine learning to operate in dynamic and uncertain environments. Ethical considerations, such as fairness and safety, are essential when deploying these systems in real-world settings.