

Lab4

September 15, 2021

```
[13]: from math import exp
import random
from copy import deepcopy
import time
```

```
[14]: N_QUEENS = 8
temp = 2000
```

make_board(n) function creates a n rows and places a queen in one of the columns of each of the rows. It stores the row, column(position) of every queen in chess_board dictionary and returns that.

```
[15]: def make_board(n):
    '''Create a chess board with a queen on a row'''
    chess_board = {}
    temp = list(range(n))
    # shuffle to make sure it is random
    random.shuffle(temp)
    column = 0

    while len(temp) > 0:
        row = random.choice(temp)
        chess_board[column] = row
        temp.remove(row)
        column += 1
    del temp
    return chess_board
```

```
[16]: def calculate_threat(n):
    '''Combination formula. It is choosing two queens in n queens'''
    if n < 2:
        return 0
    if n == 2:
        return 1
    return (n - 1) * n / 2
```

cost(chess_board) calculates how many queens are not in same position

```
[17]: def cost(chess_board):
    '''Calculate how many pairs of threaten queen'''
    threat = 0
```

```

m_chessboard = {}
a_chessboard = {}

for column in chess_board:
    temp_m = column - chess_board[column]
    temp_a = column + chess_board[column]
    if temp_m not in m_chessboard:
        m_chessboard[temp_m] = 1
    else:
        m_chessboard[temp_m] += 1
    if temp_a not in a_chessboard:
        a_chessboard[temp_a] = 1
    else:
        a_chessboard[temp_a] += 1

for i in m_chessboard:
    threat += calculate_threat(m_chessboard[i])
del m_chessboard

for i in a_chessboard:
    threat += calculate_threat(a_chessboard[i])
del a_chessboard

return threat

```

Simulated annealing (SA) is a probabilistic technique for approximating the global optimum of a given function. Specifically, it is a metaheuristic to approximate global optimization in a large search space for an optimization problem.

At each step, the simulated annealing heuristic considers some neighboring state s^* of the current state s , and probabilistically decides between moving the system to state s^* or staying in-state s . These probabilities ultimately lead the system to move to states of lower energy. Typically this step is repeated until the system reaches a state that is good enough for the application, or until a given computation budget has been exhausted.

0.1 Pseudocode for general simulated annealing

The following pseudocode presents the simulated annealing heuristic as described above. It starts from a state s_0 and continues until a maximum of k_{\max} steps have been taken. In the process, the call `neighbour(s)` should generate a randomly chosen neighbour of a given state s ; the call `random(0, 1)` should pick and return a value in the range $[0, 1]$, uniformly at random. The annealing schedule is defined by the call `temperature(r)`, which should yield the temperature to use, given the fraction r of the time budget that has been expended so far.

```

Let  $s = s_0$ 
For  $k = 0$  through  $k_{\max}$  (exclusive):
     $T \leftarrow \text{temperature}(1 - (k+1)/k_{\max})$ 
    Pick a random neighbour,  $s_{\text{new}} \leftarrow \text{neighbour}(s)$ 
    If  $P(E(s), E(s_{\text{new}}), T) \geq \text{random}(0, 1)$ :

```

```
s <- snw
Output: the final state s
```

```
[18]: def simulated_annealing():
        '''Simulated Annealing'''
        solution_found = False
        answer = make_board(N_QUEENS)

        # To avoid recounting when can not find a better state
        cost_answer = cost(answer)

        t = temp
        sch = 0.99

        while t > 0:
            t *= sch
            successor = deepcopy(answer)
            while True:
                index_1 = random.randrange(0, N_QUEENS - 1)
                index_2 = random.randrange(0, N_QUEENS - 1)
                if index_1 != index_2:
                    break
            successor[index_1], successor[index_2] = successor[index_2], \
                successor[index_1] # swap two chosen queens
            delta = cost(successor) - cost_answer
            if delta < 0 or random.uniform(0, 1) < exp(-delta / t):
                answer = deepcopy(successor)
                cost_answer = cost(answer)
            if cost_answer == 0:
                solution_found = True
                print_chess_board(answer)
                break
        if solution_found is False:
            print("Failed")
```

```
[19]: def print_chess_board(board):
        '''Print the chess board'''
        brd = [['_' for j in range(0,N_QUEENS)] for i in range(0,N_QUEENS)]
        for column, row in board.items():
            brd[row][column] = 'Q'
            print("{} => {}".format(column, row))
        for row in brd:
            print(row)
```

```
[20]: def main():
        start = time.time()
        simulated_annealing()
        print("Runtime in second:", time.time() - start)
```

```
[21]: main()
```

```
0 => 3
1 => 6
2 => 4
3 => 1
4 => 5
5 => 0
6 => 2
7 => 7
['_', '_', '_', '_', '_', 'Q', '_', '_']
['_', '_', '_', 'Q', '_', '_', '_', '_']
['_', '_', '_', '_', '_', '_', 'Q', '_']
['Q', '_', '_', '_', '_', '_', '_', '_']
['_', '_', 'Q', '_', '_', '_', '_', '_']
['_', '_', '_', '_', 'Q', '_', '_', '_']
['_', 'Q', '_', '_', '_', '_', '_', '_']
['_', '_', '_', '_', '_', '_', '_', 'Q']
Runtime in second: 0.01335453987121582
```

```
[ ]: from google.colab import drive
drive.mount('/content/drive')
```

```
[ ]:
```