# CP468 Final Project (N Queens) Group 7

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# **Discussion of design choices**

# N-queens representation

To represent a state in N-queens, it was our goal to represent all the necessary information in as little space possible. We fixed a given queen to each of the N columns. Therefore, a 1xN vector was used such that each element represents the row number in that given column. More specifically, if the 1xN vector was named puzz, the queen in the ith column would be placed in the puzz[i]th row. A 1 dimensional vector was chosen as opposed to an NxN binary representation of N-queens in order to be more space efficient.

To represent each possible conflict, three data structures were used:

- Row vector (1xN) such that the ith element represents the number of queens in row i
- Diag1 vector (1x2N-1) such that the ith element represents the number of queens in the ith diagonal in the set of diagonals beginning at the top right corner
- Diag2 vector (1x2N-1) such that the ith element represents the number of queens in the ith diagonal in the set of diagonals beginning at the top left corner

We chose this design because we felt that it was able to represent all the necessary information about a queen's positioning and conflicts in the least amount of space. These four data structures are used together through an object oriented programming design. We took an OOP approach to the problem so that we could easily call methods to reassign queens at any point in the program. Using an OOP design also helped in our organization of the code through the use of classes.

For an example of a given state, we have that the queen in column i, row puzz[i] is in row conflict with row[puzz[i]] number of queens, diagonal # 1 conflict with diag1[i+puzz[i]] number of queens, diagonal #2 conflict with diag1[i-puzz[i]+N] number of queens, and column conflict with 0 queens.

Since N-queens is framed as a constraint satisfaction problem (CSP), these three data structures essentially store constraint information about each queen. If a queen is not violating any row, diag1, and diag2 constraints then the queen will have a value of 1 at its corresponding position in each of these vectors.

An N-queens is solved when the entire diag1, diag2, and row vectors have no values above 1.

#### **Initial state creation**

The initial state was chosen by a greedy assignment process that assigns a queen to a row with the minimum number of conflicts. This was done so to reduce the number of steps that the min conflicts algorithm had to run.

#### Language choice

We initially wrote the program in python and were able to solve up to N = 100,000. However, we realized that for N=1M, the program would take several days to complete. As a result, we rewrote the program in C++ which greatly improved run time.

# **Optimization challenges**

A challenge we faced throughout the entire project was efficiency in time and space, especially time. We initially wrote the program in python, but we did not believe that it would solve N=1M in a reasonable amount of time, so we rewrote the program in C++ to solve this challenge. The second challenge was that it was taking too many steps (and thus too much time) to find a solution for large N. We knew that it should take an average of 50 steps to find a solution, independent of N size, so we were initially confused on why it was taking so many steps. After further research, we found that it takes an average of 50 steps after a good initial state had been found. Therefore, we wrote a method that would build a good initial state so that a smaller number of steps (closer to average of 50) was required to solve the problem.

# Results of test runs for n = 10, 100, 10000, 100000, 1000000 Machine specs used to generate results

#### **Hardware Overview:**

Model Name: MacBook Pro Model Identifier: MacBookPro17,1

Chip: Apple M1

Total Number of Cores: 8 (4 performance and 4 efficiency)

Memory: 8 GB System Firmware Version: 6723.81.1

Serial Number (system): FVFFG9NVQ05D

Hardware UUID: 930C556E-166F-5542-9866-0E06127FD45D

Provisioning UDID: 00008103-00182920118A001E

Activation Lock Status: Disabled

N = 10

N: 10

Initial state created in 0.000146 seconds

After initial state creation, solution found in 0.001758 seconds

Total duration: 0.001918 seconds

Number of steps: 44

N = 100

```
N: 100
Initial state created in 0.002099 seconds
After initial state creation, solution found in 0.004255 seconds
Total duration: 0.006371 seconds
Number of steps: 113
```

#### N = 10,000

```
Assigned 9000
Initial state created in 3.02511 seconds
After initial state creation, solution found in 0.017272 seconds
Total duration: 3.0424 seconds
Number of steps: 42
```

#### N = 100,000

```
Initial state created in 298.04 seconds
After initial state creation, solution found in 0.496 seconds
Total duration: 298.536 seconds
Number of steps: 142
```

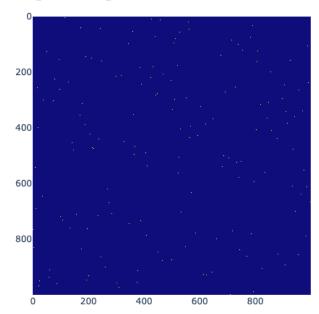
#### N = 1,000,000

```
Assigned 999000
Initial state created in 29742.5 seconds
After initial state creation, solution found in 5.0688 seconds
Total duration: 29747.6 seconds
Number of steps: 104
1
```

To see the actual output of the program, view the output\_N.txt files for a list of row numbers that the queen in column i is assigned to. When looking at the amount of time to create a good initial state for the min conflicts algorithm, we can see that the time increases relatively quickly with N. The time it takes to create a good initial state is low at 0.000153 seconds for N=10 and very high at 29742.5 seconds (8.2 hrs) for N = 1M. This increase in time makes sense since the search for an optimal state is  $O(N^2)$  since for every column (xN), we iterate through every row (xN) to find the one with the minimum number of conflicts, which itself takes constant time. However, we can see that once a good initial state has been reached, the actual time it takes to find a solution is relatively quick, even for a large N (5 seconds for 1M). This is because the time it takes to find a solution is  $O(\max_s teps^*N) = O(N)$ . This linear increase in time is shown in the images above. Amazingly, after a good initial state has been built, there is research that shows that only an average of 50 steps is needed and it becomes independent of problem size. This appears to be true for our program as well seeing that N=1M does not require particularly more steps than any other N. Since the jump from 100,000 to 1M was more than 8 hours, we assume that the program we wrote cannot solve for N larger than 1M because the initial state creation

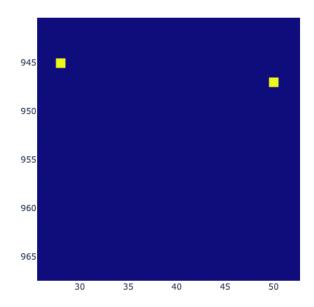
takes too long. However, if a good initial state was passed to the min conflicts algorithm, then it can find a solution in a few seconds, regardless of N size.

# **Graphical representation**



For a graphical representation of a solution found with our code, we decided to create a multichannel image data using a graphing python library named plotly. The image above is the graphical representation of a solution our code found for n = 1,000. The yellow squares represent a queen's placement on the board and a blue square represents an empty square. To create this image, we converted the program's output of a list of the row numbers for each queen in a column to a 2D nxn list such that 0 represents an empty space and 1 represents the presence of a queen at position i,j on the chess board.

This poster is also interactive. If you suspect that two points have been placed on the same row, column, or diagonal, you can zoom into those points to verify. For example, let's say that we suspect that the two points near 0,1000 are on the same row. After zooming in using the interactive feature, we can see that they are actually on different rows:



# How to compile code

# The steps to running the code for N-queens are as follows:

- 1. Specify the output file that you would like the results written to in the main. The default is 'output.txt'
- 2. Compile n\_queens.cpp to create an executable file named n\_queens.exe
- 3. Run the executable file n queens.exe
- 4. The program will wait until you specify an n as user input in console after 'N:'
- 5. The program output will be in 'output.txt' which will have a list of the row number that each queen is in for a given column

Note: if you do not have stdc++.h, copy the contents of this <u>code</u> and place inside a folder named 'bits' and place that folder in the path where C++ libraries are stores in your computer

#### The steps to running the graphical representation of a solution are as follows:

- 1. compile and run C++ program which will create a output\_n.txt' file with the a list of row positions for each column's queen needed to create the visual
- 2. upload the file 'CP468\_project\_poster.ipynb' into Google colab
- 3. upload the grid.txt file into your google drive
- 4. Run each code chunk in 'CP468\_project\_poster.ipynb'

# Code

```
#include <ctime>
using namespace std;
class NQueenCSP {
  NQueenCSP(int _N, int _MS): N(_N), max_steps(_MS) {
```

```
diag2[0 + N - 1]=1; // have to add N - 1 because i-j can be as small as -(N-1)
    row[0]=1;
void init cnts() {
```

```
for (int i = 0; i < 2*N-1; i++) {
           diag2.push_back(0);
           if (i < N) row.push_back(0);</pre>
           if (i < N) puzz.push back(0);</pre>
       rng.seed(0);
           int j = puzz[i];
ret.push_back(make_pair(i,j));
```

```
bool solved() {
      return conflicted().empty();
counted once
and 0 if i == 0
```

```
fh.open("grid viz.txt", ios::trunc);
            vector<int>::iterator itr = find(puzz.begin(), puzz.end(), col);
            int index = distance(puzz.begin(), itr);
    fh.close();
void upd vals(int i, int j, bool subtract) {
   diag1[i+j] += !subtract - subtract;
   row[j] += !subtract - subtract;
int select col to change() {
```

```
pair<int, int> ch = conf[rng() % (int)conf.size()];
upd vals(ch.first, ch.second, 1);
return ch.first;
int mini = N+1;
    if (n_conflicts(i,j) == mini) choices.push_back(j);
int new row = choices[rng() % (int)choices.size()];
upd_vals(i,new_row,0);
puzz[i]=new row;
```

```
number of clinflicts
       for(int x = 0; x < max steps; x++) {
           if (solved()) {
(std::clock() - is_creation_duration) / (double) CLOCKS PER SEC<< " seconds" << endl;</pre>
 seconds" << endl;
           int i = select_col_to_change();
int main() {
```

```
cout << nqcsp.min_conflicts() << endl;

// Log our solved configuration to text file 'output.txt' in json format for list
ofstream f;
f.open("output.txt", ios::trunc);
f << '[';
for(int i = 0; i < N-1; i++) {
    f << nqcsp.puzz[i] << ',';
}
f << nqcsp.puzz[N-1] << ']';
f.close();
// only uncomment for n <= 100, otherwise use graphical poster representation tool
in google colab
//nqcsp.print_grid();
}</pre>
```