

Artificial Intelligence 1

Lab 2

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CS4/LC8_team[public_static_void_main()]

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Exercise 1: N-queens problem

1.1 - Hill Climbing

1. We ran the program multiple times. The results are for 8 and 50 queens are in the appendix at the end of the document. We notice that, the larger the grid, the fewer queens are under attack. However, the program almost never solves the problem, but often gets close.
2. The algorithm loops over all queens, and puts them each in the most favourable position. This way, it might look like the solution is going to be optimal in the end, but the pseudo-optimal placement of the first few queens might hinder the yet-to-be-placed queens. Therefore, if this situation occurs, the found ‘solution’ might not actually be the optimal one, thus the program fails.
3. We might improve the algorithm by implementing a so called Random-Restart-Hill-Climbing algorithm. The idea behind this is as follows. If we don't find the solution, we try again, from another random point. The problem with this approach regarding this problem, is that, there are a lot of random restarts possible. Therefore, we must set a cap to the amount of restarts we allow, say, 25.

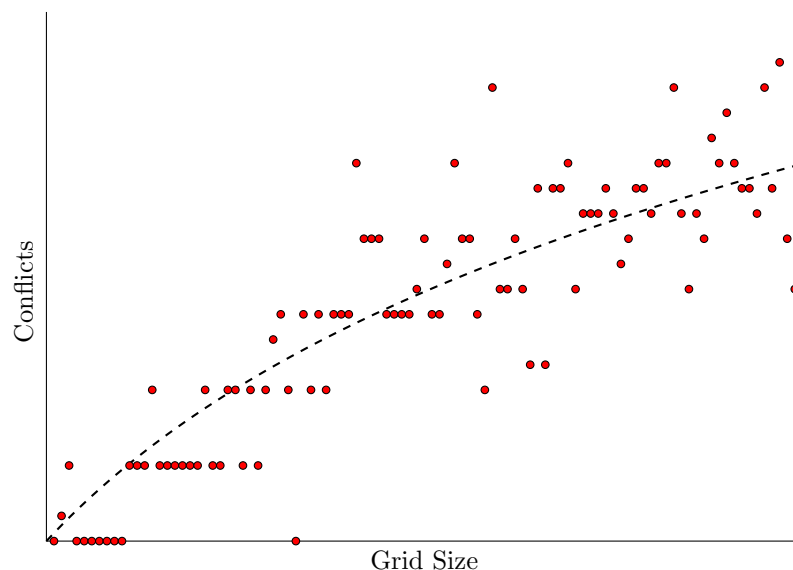
We ran a test, and immediately got a solution:

Final State

```
.....q..
..q.....
....q...
.....q
q.....
...q....
.q.....
.....q.
```

This is plenty of confirmation that this is a viable solution.

4. We use a bash script to run the program 3 times for every possible value for N . We can then graph the average of the results for every grid size (along the horizontal axis).



1.2 - Simulated Annealing

1. For the most part, we copied the algorithm from the textbook. However, instead of using the loop for $t = 1$ to infinity, we used a nested loop, the outer one going from $tries = 0$ to $tries = 1000$, and the inner one going from $i = 0$ to $i = nqueens$. In this inner loop, the queen of every row is evaluated. If the random next position, called *newPos* has less conflicts than the current position, the queen is moved to this *newPos*. If it has more (or the same amount of) conflicts than the current position, we generate a random number r and if the probability $p = e^{(\text{delta}E/T)}$ is bigger than this random float, we move the queen to the *newPos* anyway, even though it has more conflicts. We do this the queen of every row, and try it 500 times, which is the outer loop. Instead of using the delta E from the psuedo-code, we use $\text{delta}E = \text{current.conflicts} - \text{next.conflicts}$, which is exactly the same, just using one function less. One thing we added to improve the program was the same idea we used in question 3 from Hill Climbing. Basically, if t is smaller than a very small value like 0.001 and a solution has not been found yet, we restart the program, however on a new random board. This is implemented by a new outer loop which goes from 0 to, in this case, 50. In this outer loop, the temperature is set to the start temperature again and the queens are initiated by `InitiateQueens(1)`

(which sets the queens on the board randomly).

2. We chose this as the function timeToTemperature:

```
int timeToTemperature(float t){
    t = t*0.95;
    return t;
}
```

Which seemed to work fine in most cases. A smaller factor times t (in our code, we use start temperature 50, for a small problem like $n = 5$) could make it so that promising solutions are reset because t would get quite small quickly. The function creates some randomness, but it solves quite some problems. Without this function, certain problems would get stuck in a certain state with no chance of getting out of there, because there is no other state with less conflicts, however sometimes you need to take a step back to take two forward. In conclusion, this is our code for simulated annealing (which is in the nqueensa.c file):

Listing 1: Simulated Annealing

```
1  int timeToTemperature(float t){
2      t = t*0.95;
3      return t;
4  }
5
6
7  void simulatedAnnealing(){
8      float t = 50.0;
9      for(int restart = 0; restart < 50; restart++){
10         for(int tries = 0; tries < 500; tries++){
11             for(int i = 0; i < nqueens; i++){
12                 int pos = columnOfQueen(i);
13                 int newPos = pos;
14                 // loop so that the random number is
15                 // not the same
16                 while (newPos == pos) {
17                     newPos = random() % nqueens;
18                 }
19                 // current = current amount of
20                 // conflicts
21                 int current = countConflicts();
22                 // next = amount of conflicts if
23                 // moved to random newPos
24                 moveQueen(i,newPos);
25                 int next = countConflicts();
26                 int deltaE = current - next;
27                 // if the amount of conflicts at the
28                 // new is bigger than
```

```

25         // the old amount of conflicts, move
           the queen back to
26         // the 'current' configuration
27         if (deltaE < 0){
28             moveQueen(newPos,i);
29         }
30         // otherwise, the probability game
           starts, as
31         // the next state has more conflicts
           than the
32         // current one
33         else{
34             float r = ((float)rand() /
                           RAND_MAX);
35             float p = pow(euler,
                           (deltaE/t));
36             if (p>r){
37                 // do nothing, because
                   the queen is
                   already in
                   // newPos
38             }
39             else{
40                 moveQueen(newPos,i);
41             }
42         }
43     }
44 }
45 // if there are no conflicts, the solution is
   found
46 if(!countConflicts()){
47     break;
48 }
49 t = timeToTemperature(t);
50 }
51 if(!countConflicts()){
52     break;
53 }
54 // retry with different start
   initiateQueens(1);
55 t = 50.0;
56 }
57 if(!TESTING_MODE) printf("\nFinal State");
58 printState();
59 }
60 }

```

3. The algorithm will find a solution often for a small nqueens problem, for example when $n = 5$ or 6 . After this, it will struggle and find a solution every 5/6/7 tries or even more tries (or never) when the amount of queens

is big. For a large problem size, we choose a high starting temperature. Because it will take longer for the program to find a solution, the function `timeToTemperature()` is used more often, and so it will decrease more than you would maybe like to. For a small problem size a low starting temperature is better, as long as it is still good enough to make random decisions. This means that there may be less randomness, and there is at least a chance that you get out of a 'stuck' state.

4. The larger the amount of queens, the bigger the chance that there is some small mistake in one of the queen's placement, which influences the entire board. While the random 'hops' can solve these problems, this still only happens rarely, especially later on. And because the board is now at least 10x10, the chance that the 'perfect' random row and column is chosen is quite small.

We tried a few things to make the program work better for larger problems. An idea to improve the program was to add an inner loop, which goes from $j = 0$ to $j = n_{queens}$. In this loop, we search for every column which has the least amount of conflicts, instead of using the randomized way that simulated annealing is based on. Another idea was to create an outer loop which breaks if it is solved, however we decided not to implement this. It could take a very long time before a solution is found.

While it contributed, it did not work 'completely'. A problem with more than 10 queens will either take a long time or not find a solution at all. The full program is in Appendix B.

1.3 - Genetic Algorithm

Considerations

There can be much variation in genetic algorithms. For instance, one could change the generation size, the maximum amount of generations or even whole algorithms involved. By trying more combinations, you could discover that for that specific problem, one modification might benefit results.

In our case, we noticed that generation size above a certain threshold does not seem to influence the result anymore. We noticed the same when we modified the maximum amount of generations we allowed. We also implemented two different mutation functions, which can be toggled using a macro.

Design

We tried to make the algorithm as versatile as possible, so we can easily run tests. Therefore, we included macros that control the before named thresholds. Also, by adding command line arguments, we can easily run tests by changing little things.

For the algorithm itself, we used two mutation functions. One where every gene of a chromosome has a chance to be mutated, and one where there is a chance of mutating a single random gene. The last one makes mutation rarer, and

therefore less influential. This is beneficial when a generation of chromosomes is already close to convergence.

We also applied a little math to the randomness of the crossover function. We do generate each child as a crossover product of 2 parents, then apply mutation. However, when choosing the parents, we do not use the fitness function as a probability function. We apply a little math to give the ‘good’ chromosomes a much better chance to be selected than the ‘bad’ ones.

We did the following. We divide each chromosome’s fitness (we will refer to this as f) by a value depending on its fitness (y). We want y to be greater (say the size of the board, y_0) for lower values of f , and 1 for the maximum fitness (f_m). We can define a linear function

$$y = \frac{1 - y_0}{f_m} f + y_0$$

Since we know the maximum amount of the fitness is the same as the maximum amount of conflicts (see: `evaluateState()`), we can say that $f_m = \frac{y_0(y_0-1)}{2}$, and substitute

$$y = \frac{1 - y_0}{\frac{y_0(y_0-1)}{2}} f + y_0 = -\frac{2}{y_0} f + y_0$$

We can then find p (probability), and simplify to speed up the calculation by the computer

$$p = \frac{f}{y} = \frac{f}{-\frac{2}{y_0} f + y_0} = \frac{f y_0}{-2f + y_0^2}$$

Evaluation

For board sizes up to 10, the genetic algorithm solves the problem consistently. When going higher, it finds solutions which are often more than 90% correct. However, since it does not converge early because it found the solution, the time to terminate becomes exponentially larger, but remains manageable up to a board size of about 50.

To conclude, we can say that this algorithm is best to find a solution that is close to correct. Meaning to say, that it does not always find the global optima, but tries to seek out and compare the best local optima. We conclude that the Genetic Algorithm works best for the N-queens problem.

Exercise 2: Game of Nim

1. These graphs represent moves and evaluations: the nodes contain the state (value) and the evaluation and the edges are the moves. The nodes on the left represent the current player (the player that has to make a decision), where +1 is *MAX* and -1 *MIN*. A value of 1 is a terminal state and when a player is in such a state, the other player wins. *MAX* goes first. So for $n = 3$, *MAX* can either take 1 or 2 matches. Taking 2 matches results in a value of 1 and since it is then *MIN*’s turn, *MAX* wins and the evaluation

is $+1$. When MAX takes 1 match, MIN has only one option, that is take 1 match, and then MAX is in a terminal state, so MIN wins and the evaluation is -1 . The maximum of -1 and $+1$ is $+1$, so, assuming both players play optimally, MAX wins.

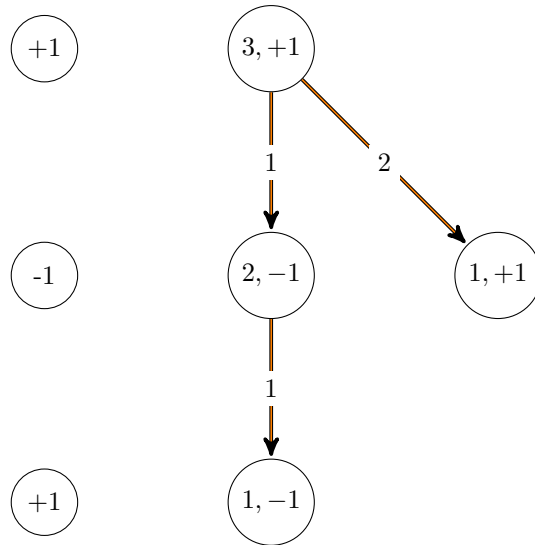


Figure 1: $n=3$

For $n = 4$, the three moves MAX can make result in values of 3, 2 and 1. Now MIN has to make a choice. 1 is terminal state, so in that case MAX wins and evaluate $+1$. In Figure 1 we see that if the value is 2 and it is MIN 's turn, the evaluation is -1 . We can also see that if the value is 3 and it is MAX 's turn, the evaluation is $+1$, so if it is MIN 's turn, all the signs are inverted and so the evaluation is -1 . The maximum of -1 , -1 and $+1$ is $+1$, so MAX wins. The result is in Figure 2.

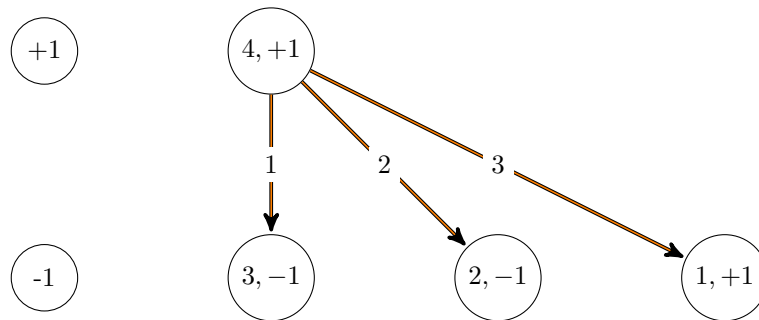


Figure 2: $n=4$

Using the same reasoning at $n = 4$, we can derive the graph for $n = 5$ (see Figure 3). The maximum of -1, -1 and -1 is -1, so *MIN* wins.

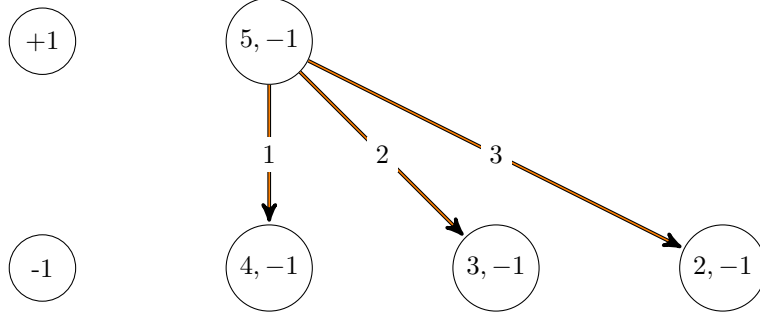


Figure 3: $n=5$

And again, using the same reasoning at $n = 4$ and $n = 5$, we can derive the graph for $n = 6$ (see Figure 4). The maximum of +1, -1 and -1 is +1, so *MAX* wins.

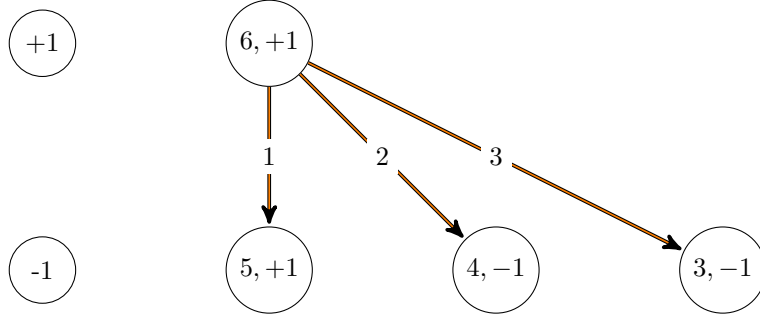


Figure 4: $n=6$

2. We can turn the *minimax* algorithm into the *negamax* algorithm by this rule: $\min(a, b) = -\max(-a, -b)$. The *negamax* function returns pairs (move + evaluation) and uses $colour \in \{-1, 1\}$, where -1 is *MIN* and 1 *MAX*. In our code, $pair[0]$ contains the move and $pair[1]$ the evaluation. This is a recursive function and our base case must return a pair: the evaluation is the opposite of the current colour, i.e. if 1 (*MAX*) is in the terminal state, -1 (*MIN*) wins and vice versa; the move isn't important in the base case, so we simply leave it as is (the best move and evaluation will be returned at the end of the function). Then, for every possible move, we check for the maximum or minimum evaluation. In this function, *minMax* is either a maximum or a minimum, depending on the colour. We don't want to assign the minimum/maximum of $pair[0]$ and *minMax* immediately to *minMax*, because we also want to determine the best move.

That's why we use an if-statement: $\max(a, b) = a > b ? a : b$. In the case that it's MAX' turn, we don't want to change this statement. If it's MIN's turn, we want to use: $\min(a, b) = -\max(-a, -b)$; and $-\max(-a - b) = -a > -b ? a : b$. *minMax* has a starting value of either *-INFINITY* or *+INFINITY*, depending on the colour: if it's MAX' turn, *minMax* is *-INFINITY* and *+INFINITY* at MIN's turn. We can achieve this by assigning $-\text{colour} * \text{INFINITY}$ to *minMax*. The result of the function is in Listing 2. (see appendix C for full program)

Listing 2: NegaMax function

```

1  int* negaMax (int state, int colour, int pair[2]) {
2      int move, bestmove, minMax = -colour*INFINITY;
3      /* terminal state ? */
4      if (state == 1) {
5          pair[1] = -colour; /* Max wins if min is in a
6                             terminal state */
7          return pair;
8      }
9      /* non-terminal state */
10     for (move = 1; move <= 3; move++) {
11         if (state - move > 0) { /* legal move */
12             pair = negaMax(state-move, -colour, pair);
13             if (colour*pair[1] > colour*minMax) {
14                 minMax = pair[1];
15                 bestmove = move;
16             }
17         }
18     }
19     pair[0] = bestmove;
20     pair[1] = minMax;
21     return pair;
22 }
23
24 int minimaxDecisionNegaMax(int state, int turn) {
25     int pair[2];
26     if (turn == MAX) {
27         negaMax(state, 1, pair);
28         return pair[0];
29     }
30     /* turn == MIN */
31     negaMax(state, -1, pair);
32     return pair[0];
33 }

```

3. Results of running the program with $n = 10, 20, 30, 40, 50$:

n=10: quick	n=20: quick	n=30: a bit slower
10: Max takes 1	20: Max takes 3	30: Max takes 1

9: Min takes 1	17: Min takes 1	29: Min takes 1
8: Max takes 3	16: Max takes 3	28: Max takes 3
5: Min takes 1	13: Min takes 1	25: Min takes 1
4: Max takes 3	12: Max takes 3	24: Max takes 3
1: Min loses	9: Min takes 1	21: Min takes 1
	8: Max takes 3	20: Max takes 3
	5: Min takes 1	17: Min takes 1
	4: Max takes 3	16: Max takes 3
	1: Min loses	13: Min takes 1
		12: Max takes 3
		9: Min takes 1
		8: Max takes 3
		5: Min takes 1
		4: Max takes 3
		1: Min loses

$n=40$: The program is way too slow and no result is given (within a reasonable amount of time). $n=50$: There's no point in trying $n = 50$, since it will take an eternity to finish

Other things we observed: Min loses every game and Max takes a lot of 3s and Min a lot of 1s.

We now use a transposition table, which is an integer array of size 101 (index 0 not used), with another dimension of size 3, so that for every state we can store three values: best move, evaluation and colour. Now, in our *negaMax* function, the first thing we do is check if we have already evaluated this state. Initially all values of the transposition table are 0, so if the evaluation (which is either +1 or -1) is 0, that means that we haven't evaluated this state before. However, if it isn't 0, we already know what to do and return the move and evaluation. The move to make is the same for MIN and MAX, but the evaluation is different: in the transposition table, we also store for which player this evaluation is; if the current player is the same as the player in the transposition table, we return the evaluation as is, if it is the other player, however, we return the negation of the evaluation (if the evaluation at state a is +1 for MAX, it will be -1 for MIN). After every possible move of this state has been checked, we can store the best move, the evaluation and the colour in the transposition table.

If we now run the program with $n = 50$ the program gives a result right away, so it definitely helps. The result of the function is in Listing 3. (see appendix D for full program)

Listing 3: NegaMax with transposition table

```

1 int* negaMax (int state, int colour, int pair[2], int
    transpositionTable[][3]) {

```

```

2      if (transpositionTable[state][1] != 0) {
3          pair[0] = transpositionTable[state][0];
4          if (transpositionTable[state][2] == colour)
5              pair[1] = transpositionTable[state][1];
6          else
7              pair[1] = -transpositionTable[state][1];
8          return pair;
9      }
10     int move, bestmove, minMax = -colour*INFINITY;
11     /* terminal state ? */
12     if (state == 1) {
13         pair[1] = -colour; /* Max wins if min is in a
14                             terminal state */
15         return pair;
16     }
17     /* non-terminal state */
18     for (move = 1; move <= 3; move++) {
19         if (state - move > 0) { /* legal move */
20             pair = negaMax(state-move, -colour, pair,
21                             transpositionTable);
22             if (colour*pair[1] > colour*minMax) {
23                 minMax = pair[1];
24                 bestmove = move;
25             }
26         }
27     }
28     transpositionTable[state][0] = pair[0] = bestmove;
29     transpositionTable[state][1] = pair[1] = minMax;
30     transpositionTable[state][2] = colour;
31     return pair;
32 }

```

Appendix

A: Exercise 1.1 output

Initial state:

Q

Final State

.q.....
 ...q.....
q....
q.....
Q....
 Q.....
Q....
 ...Q.....

Initial state:

A series of horizontal dotted lines for handwriting practice, each containing a lowercase letter 'q' at various positions.

Final State

(The page contains faint, illegible markings that appear to be bleed-through from the reverse side.)

B: full nqueens program (with hill climbing, simulated annealing and genetic algorithm)

```
1  /* nqueens.c: (c) Arnold Meijster (a.meijster@rug.nl) */
```

```

2
3 #include <stdio.h>
4 #include <stdlib.h>
5 #include <math.h>
6 #include <time.h>
7
8 #define MAXQ 100
9
10 #define MAXGEN 5000
11 #define GENSIZE 50
12 #define MUTATE_MODE 1
13
14 #define TESTING_MODE 0
15
16 #define FALSE 0
17 #define TRUE 1
18
19 #define euler 2.7182818
20
21 #define ABS(a) ((a) < 0 ? -(a) : (a))
22
23 int nqueens; /* number of queens: global variable */
24 int queens[MAXQ]; /* queen at (r,c) is represented by queens[r] == c */
25
26 void initializeRandomGenerator() {
27     /* this routine initializes the random generator. You are not
28      * supposed to understand this code. You can simply use it.
29      */
30     time_t t;
31     srand((unsigned) time(&t));
32 }
33
34 /* Generate an initial position.
35  * If flag == 0, then for each row, a queen is placed in the first
36  * column.
37  * If flag == 1, then for each row, a queen is placed in a random column.
38  */
39 void initiateQueens(int flag) {
40     int q;
41     for (q = 0; q < nqueens; q++) {
42         queens[q] = (flag == 0 ? 0 : random() % nqueens);
43     }
44 }
45
46 /* returns TRUE if position (row0,column0) is in
47  * conflict with (row1,column1), otherwise FALSE.
48  */
49 int inConflict(int row0, int column0, int row1, int column1) {
50     if (row0 == row1) return TRUE; /* on same row, */
51     if (column0 == column1) return TRUE; /* column, */

```

```

51     if (ABS(row0-row1) == ABS(column0-column1)) return TRUE; /* diagonal */
52     return FALSE; /* no conflict */
53 }
54
55 /* returns TRUE if position (row,col) is in
56 * conflict with any other queen on the board, otherwise FALSE.
57 */
58 int inConflictWithAnotherQueen(int row, int col) {
59     int queen;
60     for (queen=0; queen < nqueens; queen++) {
61         if (inConflict(row, col, queen, queens[queen])) {
62             if ((row != queen) || (col != queens[queen])) return TRUE;
63         }
64     }
65     return FALSE;
66 }
67
68 /* print configuration on screen */
69 void printState() {
70     if(TESTING_MODE) return;
71     int row, column;
72     printf("\n");
73     for(row = 0; row < nqueens; row++) {
74         for(column = 0; column < nqueens; column++) {
75             if (queens[row] != column) {
76                 printf(".");
77             } else {
78                 if (inConflictWithAnotherQueen(row, column)) {
79                     printf("Q");
80                 } else {
81                     printf("q");
82                 }
83             }
84         }
85         printf("\n");
86     }
87 }
88
89 /* move queen on row q to specified column, i.e. to (q,column) */
90 void moveQueen(int queen, int column) {
91     if ((queen < 0) || (queen >= nqueens)) {
92         fprintf(stderr, "Error in moveQueen: queen=%d "
93             "(should be 0<=queen<%d)...Abort.\n", queen, nqueens);
94         exit(-1);
95     }
96     if ((column < 0) || (column >= nqueens)) {
97         fprintf(stderr, "Error in moveQueen: column=%d "
98             "(should be 0<=column<%d)...Abort.\n", column, nqueens);
99         exit(-1);
100    }

```

```

101     queens[queen] = column;
102 }
103
104 /* returns TRUE if queen can be moved to position
105  * (queen,column). Note that this routine checks only that
106  * the values of queen and column are valid! It does not test
107  * conflicts!
108  */
109 int canMoveTo(int queen, int column) {
110     if ((queen < 0) || (queen >= nqueens)) {
111         fprintf(stderr, "Error in canMoveTo: queen=%d "
112             "(should be 0<=queen<%d)...Abort.\n", queen, nqueens);
113         exit(-1);
114     }
115     if (column < 0 || column >= nqueens) return FALSE;
116     if (queens[queen] == column) return FALSE; /* queen already there */
117     return TRUE;
118 }
119
120 /* returns the column number of the specified queen */
121 int columnOfQueen(int queen) {
122     if ((queen < 0) || (queen >= nqueens)) {
123         fprintf(stderr, "Error in columnOfQueen: queen=%d "
124             "(should be 0<=queen<%d)...Abort.\n", queen, nqueens);
125         exit(-1);
126     }
127     return queens[queen];
128 }
129
130 /* returns the number of pairs of queens that are in conflict */
131 int countConflicts() {
132     int cnt = 0;
133     int queen, other;
134     for (queen=0; queen < nqueens; queen++) {
135         for (other=queen+1; other < nqueens; other++) {
136             if (inConflict(queen, queens[queen], other, queens[other])) {
137                 cnt++;
138             }
139         }
140     }
141     return cnt;
142 }
143
144 /* evaluation function. The maximal number of queens in conflict
145  * can be 1 + 2 + 3 + 4 + .. + (nqueens-1)=(nqueens-1)*nqueens/2.
146  * Since we want to do ascending local searches, the evaluation
147  * function returns (nqueens-1)*nqueens/2 - countConflicts().
148  */
149 int evaluateState() {
150     return (nqueens-1)*nqueens/2 - countConflicts();

```

```

151 }
152
153 /*****
154
155  /* A very silly random search 'algorithm' */
156 #define MAXITER 1000
157 void randomSearch() {
158     int queen, iter = 0;
159     int optimum = (nqueens-1)*nqueens/2;
160
161     while (evaluateState() != optimum) {
162         printf("iteration %d: evaluation=%d\n", iter++, evaluateState());
163         if (iter == MAXITER) break; /* give up */
164         /* generate a (new) random state: for each queen do ...*/
165         for (queen=0; queen < nqueens; queen++) {
166             int pos, newpos;
167             /* position (=column) of queen */
168             pos = columnOfQueen(queen);
169             /* change in random new location */
170             newpos = pos;
171             while (newpos == pos) {
172                 newpos = random() % nqueens;
173             }
174             moveQueen(queen, newpos);
175         }
176     }
177     if (iter < MAXITER) {
178         printf ("Solved puzzle. ");
179     }
180     printf ("Final state is");
181     printState();
182 }
183
184 /*****
185
186 int sharesColumnWithPreviousQueen(int row, int column){
187     for(int i = 0; i < row; ++i){
188         if(queens[i] == column) return TRUE;
189     }
190     return FALSE;
191 }
192
193 void hillClimbing() {
194     int statelist[MAXQ];
195     for(int i = 0; i < nqueens; ++i) {
196         /* store state ranks in array statelist */
197         for(int j = 0; j < nqueens; ++j) {
198             if(!sharesColumnWithPreviousQueen(i,j)) {
199                 queens[i] = j;
200                 statelist[j] = evaluateState();

```



```

201         } else {
202             statelist[j] = -1;
203         }
204     }
205
206     /* find amount of maximal ranks */
207     int maxVal = -1, maxCnt = 0;
208     for(int j = 0; j < nqueens; ++j) {
209         if(statelist[j] == maxVal) ++maxCnt;
210         if(statelist[j] > maxVal) {
211             maxVal = statelist[j];
212             maxCnt = 1;
213         }
214     }
215
216     /* determine a random position from the options
217     * and move the queen there */
218     if(maxCnt == 0) continue;
219     int choice = random() % maxCnt;
220     int cnt = 0;
221     for(int j = 0; j < nqueens; ++j) {
222         if(statelist[j] == maxVal){
223             if(cnt == choice){
224                 queens[i] = j;
225                 break;
226             }
227             ++cnt;
228         }
229     }
230 }
231 if(!TESTING_MODE) printf("\nFinal State");
232 printState();
233 }
234
235 void randomRestartHillClimbing() {
236     int maxRestarts = 25;
237     int best[MAXQ];
238     int bestRank = 0;
239     for(int i = 0; i < maxRestarts; ++i) {
240         int statelist[MAXQ];
241         initiateQueens(1);
242         for(int j = 0; j < nqueens; ++j) {
243             /* store state ranks in array statelist */
244             for(int k = 0; k < nqueens; ++k) {
245                 if(!sharesColumnWithPreviousQueen(i,k)) {
246                     queens[i] = k;
247                     statelist[j] = evaluateState();
248                 } else {
249                     statelist[j] = -1;
250                 }

```

```

251     }
252
253     /* find amount of maximal ranks */
254     int maxVal = -1, maxCnt = 0;
255     for(int j = 0; j < nqueens; ++j) {
256         if(statelist[j] == maxVal) ++maxCnt;
257         if(statelist[j] > maxVal) {
258             maxVal = statelist[j];
259             maxCnt = 1;
260         }
261     }
262
263     /* determine a random position from the options
264     * and move the queen there */
265     if(maxCnt == 0) continue;
266     int choice = random() % maxCnt;
267     int cnt = 0;
268     for(int j = 0; j < nqueens; ++j) {
269         if(statelist[j] == maxVal){
270             if(cnt == choice){
271                 queens[i] = j;
272                 break;
273             }
274             ++cnt;
275         }
276     }
277
278     int rank;
279     if((rank = evaluateState()) > bestRank) {
280         for(int i = 0; i < nqueens; ++i) {
281             best[i] = queens[i];
282         }
283         bestRank = rank;
284     }
285 }
286 for(int i = 0; i < nqueens; ++i) {
287     queens[i] = best[i];
288 }
289 if(!TESTING_MODE) printf("\nFinal State");
290 printState();
291 }
292
293 /*****
294
295 int timeToTemperature(float t){
296     t = t*0.95;
297     return t;
298 }
299
300 void simulatedAnnealing(){

```

```

301     for(int restart = 0; restart < 50; restart++){
302         for(int tries = 0; tries < 1000; tries++){
303             for(int i = 0; i < nqueens; i++){
304                 int current = countConflicts();
305                 int min = current;
306                 int minRow = i;
307                 for (int j = 0; j < nqueens; j++){
308                     moveQueen(i,j);
309                     int currentRowConflict =
310                         countConflicts();
311                     if (currentRowConflict<min){
312                         min = currentRowConflict;
313                         minRow = j;
314                     }
315                 }
316                 moveQueen(i, minRow);
317                 // if there are no conflicts, the solution is found
318                 if(!countConflicts()){
319                     break;
320                 }
321             }
322             if(!countConflicts()){
323                 break;
324             }
325             initiateQueens(1);
326         }
327         if(!TESTING_MODE) printf("\nFinal State");
328     printState();
329 }
330
331 /*****
332
333 void printChromosome(int* chrom){
334     for(int i = 0; i < nqueens; ++i) {
335         queens[i] = chrom[i];
336     }
337     printState();
338 }
339
340 void freeGeneration(int** chromosomes) {
341     for(int i = 0; i < nqueens; ++i) {
342         free(chromosomes[i]);
343     }
344     free(chromosomes);
345 }
346
347 void initializeGeneration(int** generation) {
348     for(int i = 0; i < GENSIZE; ++i) {
349         for(int j = 0; j < nqueens; ++j) {

```

```

350         generation[i][j] = random() % nqueens;
351     }
352 }
353 }
354
355 int calcFitness(int* chrom) {
356     for(int i = 0; i < nqueens; ++i) {
357         queens[i] = chrom[i];
358     }
359     return evaluateState();
360 }
361
362 int getFittestIndex(int** chromosomes) {
363     int idx = 0, fitness = calcFitness(chromosomes[0]);
364     for(int i = 1; i < GENSIZE; ++i) {
365         int curFitness = calcFitness(chromosomes[i]);
366         if(curFitness > fitness) {
367             idx = i;
368             fitness = curFitness;
369         }
370     }
371     return idx;
372 }
373
374 /* Mutates one gene in 1/4 new chromosomes OR
375  * Mutates a gene with a probability of 10%
376  */
377 void mutate(int* chrom){
378     if(!MUTATE_MODE) {
379         if(!(random() % 4)) {
380             chrom[random() % nqueens] = random() % nqueens;
381         }
382     } else {
383         for(int i = 0; i < nqueens; ++i) {
384             if(!(random() % 10)) {
385                 chrom[i] = random() % nqueens;
386             }
387         }
388     }
389 }
390
391 int* crossover(int** a, int** b){
392     int cutPoint = random() % nqueens;
393     int* new = malloc(nqueens * sizeof(int));
394     for(int i = 0; i < nqueens; ++i) {
395         if(i < cutPoint) {
396             new[i] = (*a)[i];
397         } else {
398             new[i] = (*b)[i];
399         }

```

```

400     }
401     return new;
402 }
403
404 /* returns a new generation of chromosomes, and frees the given one */
405 int** reproduce(int** chromosomes, int generationIdx){
406     int** newGeneration = malloc(GENSIZE * sizeof(int*));
407     for(int i = 0; i < GENSIZE; ++i) {
408         newGeneration[i] = malloc(nqueens * sizeof(int));
409     }
410
411     /* instead of using the fitness as a probability function,
412      * we will use function of the fitness and some other variables
413      * to give better chromosomes a greater advantage over the
414      * worse chromosomes. This is to decrease the time of convergence */
415     int* probabilities = malloc(GENSIZE * sizeof(int));
416     int totalProb = 0;
417     for(int i = 0; i < GENSIZE; ++i) {
418         /* probability function yields values between [nqueens, 1]
419          * for lowest and highest fitnesses respectively */
420         int f = calcFitness(chromosomes[i]);
421         probabilities[i] = (f * nqueens) / (-2*f + nqueens*nqueens);
422         totalProb += probabilities[i];
423     }
424
425     for(int i = 0; i < GENSIZE; ++i) {
426         int cr1 = random() % totalProb;
427         int cr2 = random() % totalProb;
428         int a = 0, b = 0;
429         for(int j = 0; j < GENSIZE; ++j) {
430             if(cr1 < probabilities[j] && cr1 >= 0) a = j;
431             if(cr2 < probabilities[j] && cr2 >= 0) b = j;
432             cr1 -= probabilities[j];
433             cr2 -= probabilities[j];
434         }
435         // preventing inbreeding
436         if(a == b) {
437             --i; continue;
438         }
439         newGeneration[i] = crossover(&chromosomes[a], &chromosomes[b]);
440         mutate(chromosomes[i]);
441     }
442
443     free(probabilities);
444     freeGeneration(chromosomes);
445     return newGeneration;
446 }
447
448 void geneticPermutation(){
449     int** chromosomes = malloc(GENSIZE * sizeof(int*));

```

```

450     for(int i = 0; i < GENSIZE; ++i) {
451         chromosomes[i] = malloc(nqueens * sizeof(int));
452     }
453     initializeGeneration(chromosomes);
454
455     int maxFitness = (nqueens-1)*nqueens/2;
456     for(int i = 0; i < MAXGEN; ++i) {
457         chromosomes = reproduce(chromosomes, i);
458         int max = getFittestIndex(chromosomes);
459         if(calcFitness(chromosomes[max]) == maxFitness) {
460             printf("\nFound solution");
461             printChromosome(chromosomes[max]);
462             return;
463         }
464     }
465     int fittest = getFittestIndex(chromosomes);
466     printf("\nTerminated with %d/%d (%.2f%%)",
467           calcFitness(chromosomes[fittest]), maxFitness,
468           100*calcFitness(chromosomes[fittest]) / (float)maxFitness);
469     printChromosome(chromosomes[fittest]);
470
471     freeGeneration(chromosomes);
472 }
473
474 int main(int argc, char *argv[]) {
475     int algorithm;
476
477     do {
478         if(!TESTING_MODE) printf ("Number of queens (1<=nqueens<=%d): ",
479                                   MAXQ);
480         if(argc == 3)
481             nqueens = atoi(argv[1]);
482         else
483             scanf ("%d", &nqueens);
484     } while ((nqueens < 1) || (nqueens > MAXQ));
485
486     do {
487         if(!TESTING_MODE) {
488             printf ("Algorithm: (1) Random search (2) Hill climbing ");
489             printf ("(3) Simulated Annealing (4) Genetic Permutation: ");
490         }
491         if(argc == 3)
492             algorithm = atoi(argv[2]);
493         else
494             scanf ("%d", &algorithm);
495     } while ((algorithm < 1) || (algorithm > 4));
496
497     initializeRandomGenerator();

```

```

497     initiateQueens(1);
498
499     if(!TESTING_MODE) printf("\nInitial state:");
500     printState();
501
502     switch (algorithm) {
503     case 1: randomSearch();           break;
504     case 2: randomRestartHillClimbing(); break;
505     case 3: simulatedAnnealing();     break;
506     case 4: geneticPermutation();     break;
507     }
508
509     if(TESTING_MODE) {
510         FILE *out = fopen("1.out", "a");
511         fprintf(out, "%.1f,%d", (double)nqueens/10, countConflicts());
512         fclose(out);
513     }
514     return 0;
515 }

```

C: Nim with negaMax function [nimNegaMax.c]

```

1  #include <stdio.h>
2  #include <stdlib.h>
3
4  #define MAX 0
5  #define MIN 1
6
7  #define INFINITY 9999999
8
9  int* negaMax (int state, int colour, int pair[2]) {
10     int move, bestmove, minMax = -colour*INFINITY;
11     /* terminal state ? */
12     if (state == 1) {
13         pair[1] = -colour; /* Max wins if min is in a terminal
14                             state */
15         return pair;
16     }
17     /* non-terminal state */
18     for (move = 1; move <= 3; move++) {
19         if (state - move > 0) { /* legal move */
20             pair = negaMax(state-move, -colour, pair);
21             if (colour*pair[1] > colour*minMax) {
22                 minMax = pair[1];
23                 bestmove = move;
24             }
25         }
26     }
27     pair[0] = bestmove;

```

```

27         pair[1] = minMax;
28         return pair;
29     }
30
31     int minimaxDecisionNegaMax(int state, int turn) {
32         int pair[2];
33         if (turn == MAX) {
34             negaMax(state, 1, pair);
35             return pair[0];
36         }
37         /* turn == MIN */
38         negaMax(state, -1, pair);
39         return pair[0];
40     }
41
42     void playNim(int state) {
43         int turn = 0;
44         while (state != 1) {
45             int action = minimaxDecisionNegaMax(state, turn);
46             printf("%d: %s takes %d\n", state,
47                 (turn==MAX ? "Max" : "Min"), action);
48             state = state - action;
49             turn = 1 - turn;
50         }
51         printf("1: %s loses\n", (turn==MAX ? "Max" : "Min"));
52     }
53
54     int main(int argc, char *argv[]) {
55         if ((argc != 2) || (atoi(argv[1]) < 3)) {
56             fprintf(stderr, "Usage: %s <number of sticks>, where ", argv[0]);
57             fprintf(stderr, "<number of sticks> must be at least 3!\n");
58             return -1;
59         }
60
61         playNim(atoi(argv[1]));
62
63         return 0;
64     }

```

D: Nim with transposition table [nimTrans.c]

```

1  #include <stdio.h>
2  #include <stdlib.h>
3
4  #define MAX 0
5  #define MIN 1
6
7  #define INFINITY 9999999
8

```



```

9  int* negaMax (int state, int colour, int pair[2], int
    transpositionTable[][3]) {
10     if (transpositionTable[state][1] != 0) {
11         pair[0] = transpositionTable[state][0];
12         if (transpositionTable[state][2] == colour)
13             pair[1] = transpositionTable[state][1];
14         else
15             pair[1] = -transpositionTable[state][1];
16         return pair;
17     }
18     int move, bestmove, minMax = -colour*INFINITY;
19     /* terminal state ? */
20     if (state == 1) {
21         pair[1] = -colour; /* Max wins if min is in a terminal
22                             state */
23         return pair;
24     }
25     /* non-terminal state */
26     for (move = 1; move <= 3; move++) {
27         if (state - move > 0) { /* legal move */
28             pair = negaMax(state-move, -colour, pair,
29                             transpositionTable);
30             if (colour*pair[1] > colour*minMax) {
31                 minMax = pair[1];
32                 bestmove = move;
33             }
34         }
35     }
36     transpositionTable[state][0] = pair[0] = bestmove;
37     transpositionTable[state][1] = pair[1] = minMax;
38     transpositionTable[state][2] = colour;
39     return pair;
40 }
41
42 int minimaxDecisionNegaMax(int state, int turn, int
    transpositionTable[][3]) {
43     int pair[2];
44     if (turn == MAX) {
45         negaMax(state, 1, pair, transpositionTable);
46         return pair[0];
47     }
48     /* turn == MIN */
49     negaMax(state, -1, pair, transpositionTable);
50     return pair[0];
51 }
52
53 void playNim(int state, int transpositionTable[][3]) {
54     int turn = 0;
55     while (state != 1) {
56         int action = minimaxDecisionNegaMax(state, turn, transpositionTable);

```

```

55     printf("%d: %s takes %d\n", state,
56            (turn==MAX ? "Max" : "Min"), action);
57     state = state - action;
58     turn = 1 - turn;
59 }
60 printf("1: %s loses\n", (turn==MAX ? "Max" : "Min"));
61 }
62
63 int main(int argc, char *argv[]) {
64     if ((argc != 2) || (atoi(argv[1]) < 3)) {
65         fprintf(stderr, "Usage: %s <number of sticks>, where ", argv[0]);
66         fprintf(stderr, "<number of sticks> must be at least 3!\n");
67         return -1;
68     }
69
70     int transpositionTable[101][3] = {{0}};
71
72     playNim(atoi(argv[1]), transpositionTable);
73
74     return 0;
75 }

```