

1. The Big-Oh runtime for the brute force n-queens algorithm is $O(n!)$
2.
 - a. The brute force algorithm takes 1.374 seconds when $n=8$
 - b. The backtracking algorithm only takes .0003 seconds when $n=8$
 - c. This makes the backtracking solution approximately 458 times faster for $n=8$
3. Backtracking is not always faster than brute force. When n is small, they take the same amount of time. However, as n grows the backtracking method is significantly faster. Here are the running times:

Bruteforce (2)

Running time: 0.0 ms

Backtracking (2)

Running time: 0.0 ms

Bruteforce (3)

Running time: 0.0 ms

Backtracking (3)

Running time: 0.0 ms

Bruteforce (4)

Running time: 0.0 ms

Backtracking (4)

Running time: 0.0 ms

Bruteforce (5)

Running time: 0.0 ms

Backtracking (5)

Running time: 0.0 ms

Bruteforce (6)

Running time: 13.999700546264648 ms

Backtracking (6)

Running time: 0.9868144989013672 ms

Bruteforce (7)

Running time: 39.99066352844238 ms

Backtracking (7)

Running time: 0.0 ms

Bruteforce (8)

Running time: 1383.9564323425293 ms

Backtracking (8)

Running time: 2.997875213623047 ms

4. I will show that a square at position (i, j) is diagonal to a square at (x, y) if and only if $i + j == x + y$ or $i - j == x - y$. I will use the following definition of a diagonal square. Two squares (i, j) and (x, y) are diagonal if one of the following cases is true:

$$i - m = x \text{ and } j - m = y$$

$$i - m = x \text{ and } j + m = y$$

$$i + m = x \text{ and } j - m = y$$

$$i + m = x \text{ and } j + m = y$$

I will show that each definition resolves to either $i + j == x + y$ or $i - j == x - y$

- a. For the first definition I will subtract the second formula from the first
 - a. $(i - m = x) - (j - m = y)$
 - b. $i - m - j + m = x - y$
 - c. $i - j = x - y$
 - b. For the second definition I will add both formulas together
 - a. $(i - m = x) + (j + m = y)$
 - b. $i - m + j + m = x + y$
 - c. $i + j = x + y$
 - c. For the third equation I will add both formulas together
 - a. $(i + m = x) + (j - m = y)$
 - b. $i + m + j - m = x + y$
 - c. $i + j = x + y$
 - d. For the fourth and final definition I will subtract the second formula from the first
 - a. $(i + m = x) - (j + m = y)$
 - b. $i + m - j - m = x - y$
 - c. $i - j = x - y$
5. The Big-Oh run time for the fast_fib or iterative Fibonacci algorithm is $O(n)$.
- 6.
- a. The fastest of the three Fibonacci algorithms implemented is by far the matrix algorithm when referring to Big-Oh run times.
 - b.
 - i. Matrix Fibonacci $n = 4$ Running time: 0.0 ms
 - ii. Fast Fibonacci $n = 4$ Running time: 0.0 ms
 - iii. Recursive Fibonacci $n = 4$ Running time: 0.0 ms
 - iv. Matrix Fibonacci $n = 32$ Running time: 0.0 ms
 - v. Fast Fibonacci $n = 32$ Running time: 0.0 ms
 - vi. Recursive Fibonacci $n = 32$ Running time: 416.5472984313965 ms
 - vii. Matrix Fibonacci $n = 65536$ Running time: 7.00068473815918 ms
 - viii. Fast Fibonacci $n = 65536$ Running time: 51.44095420837402 ms

- ix. Recursive Fibonacci $n = 65536$ Running time: N/A
- c. All algorithms ran at the same speed when $n = 4$
- d. At $n = 32$ the Matrix and Fast Fibonacci algorithms were the fastest
- e. At $n = 65536$ the Matrix Fibonacci algorithm was the fastest
- f. The fastest Big-Oh runtime was the Matrix Fibonacci algorithm with a runtime of $O(\log n)$. Big-Oh represents the worst scenario for each algorithm relative to the size of the input. The second fastest was the Fast Fibonacci algorithm with a runtime of $O(n)$. The slowest algorithm was the Recursive Fibonacci algorithm with a runtime of $O(2^n)$ which is on the slower side of Big-Oh runtimes.