

Ex. 4.1

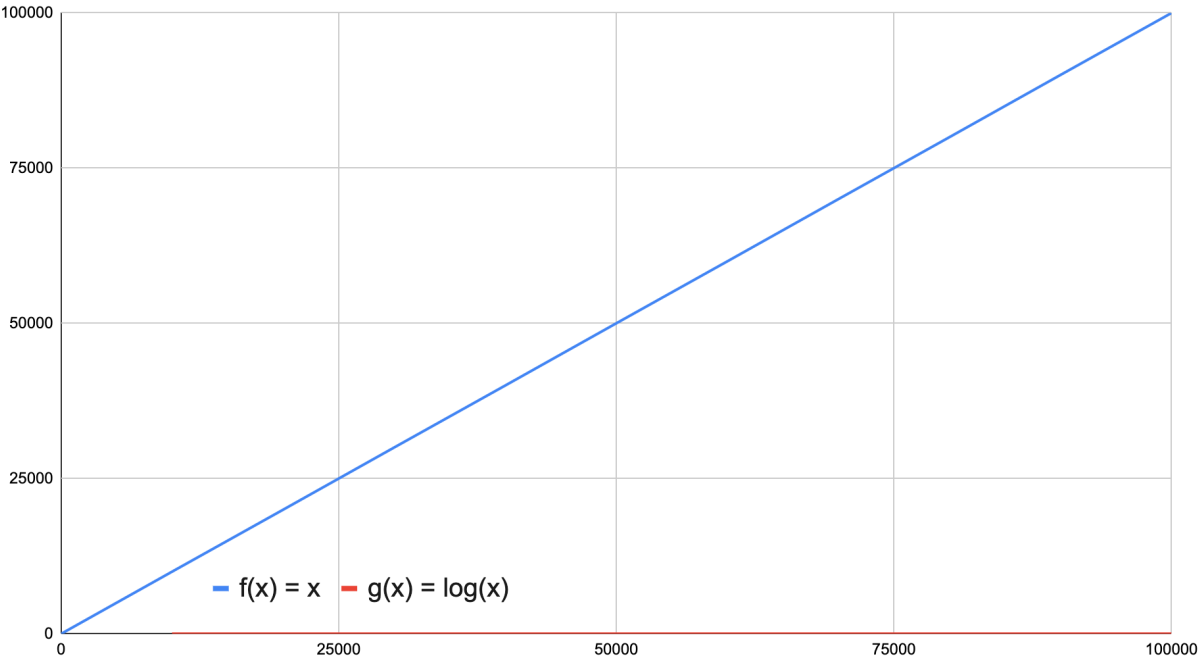
N	Output
100	Found=true Time taken: 0 Found=true Time taken: 0
1,000	Found=false Time taken: 0 Found=false Time taken: 0
100,000	Found=true Time taken: 1 Found=true Time taken: 0
100,000,000	Found=true Time taken: 11 Found=true Time taken: 0

Ex. 4.4

Curves $f(x) = x$ and $g(x) = \log(x)$ in the range 0 - 100,000

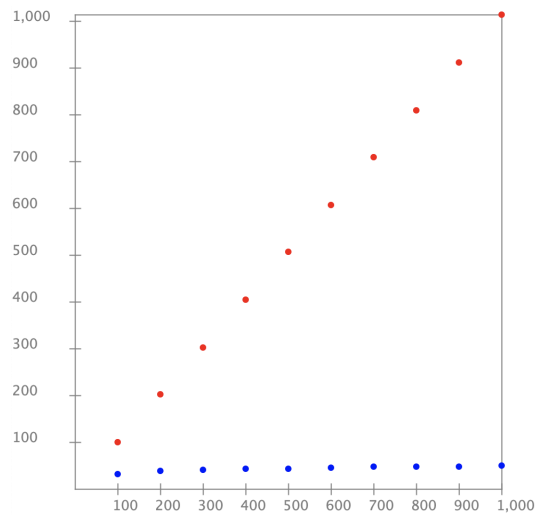
0	0	
10000	10000	13.28771238
20000	20000	14.28771238
30000	30000	14.87267488
40000	40000	15.28771238
50000	50000	15.60964047
60000	60000	15.87267488
70000	70000	16.0950673
80000	80000	16.28771238
90000	90000	16.45763738

100000	100000	16.60964047
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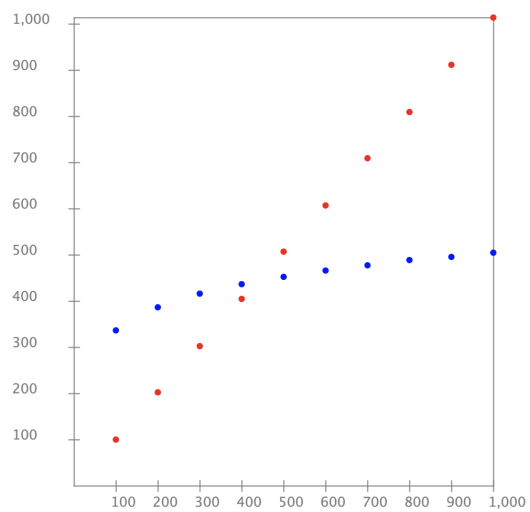


Ex. 4.6

With $q = 5$



With $q = 50$



Ex. 4.7

Five executions with $N = 10,000$

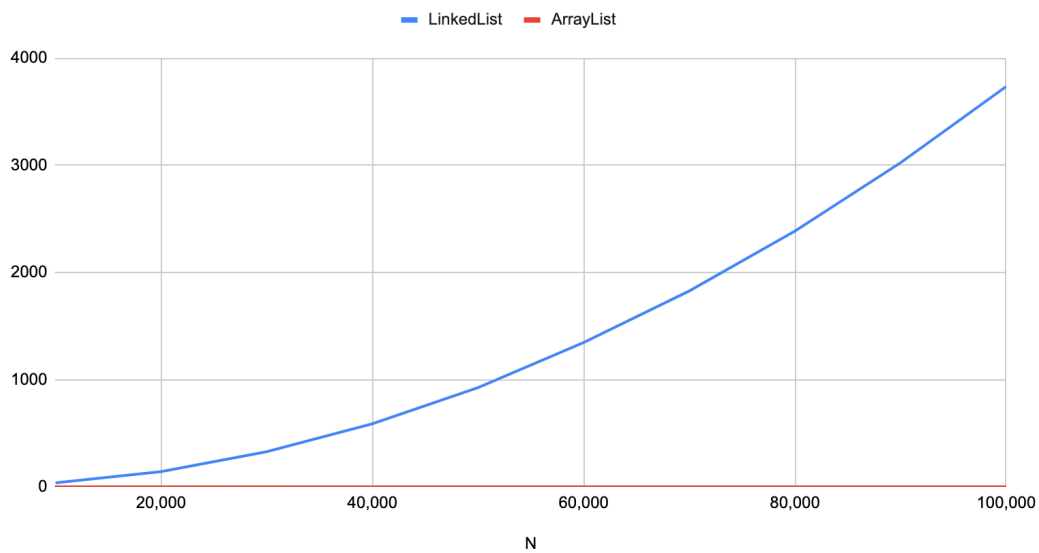
Test	LinkedList	ArrayList
1	40	1
2	40	0
3	42	0
4	43	0
5	40	0
Average	41	0.2

Ex. 4.8

Averages from five executions with N from 10,000 to 100,000

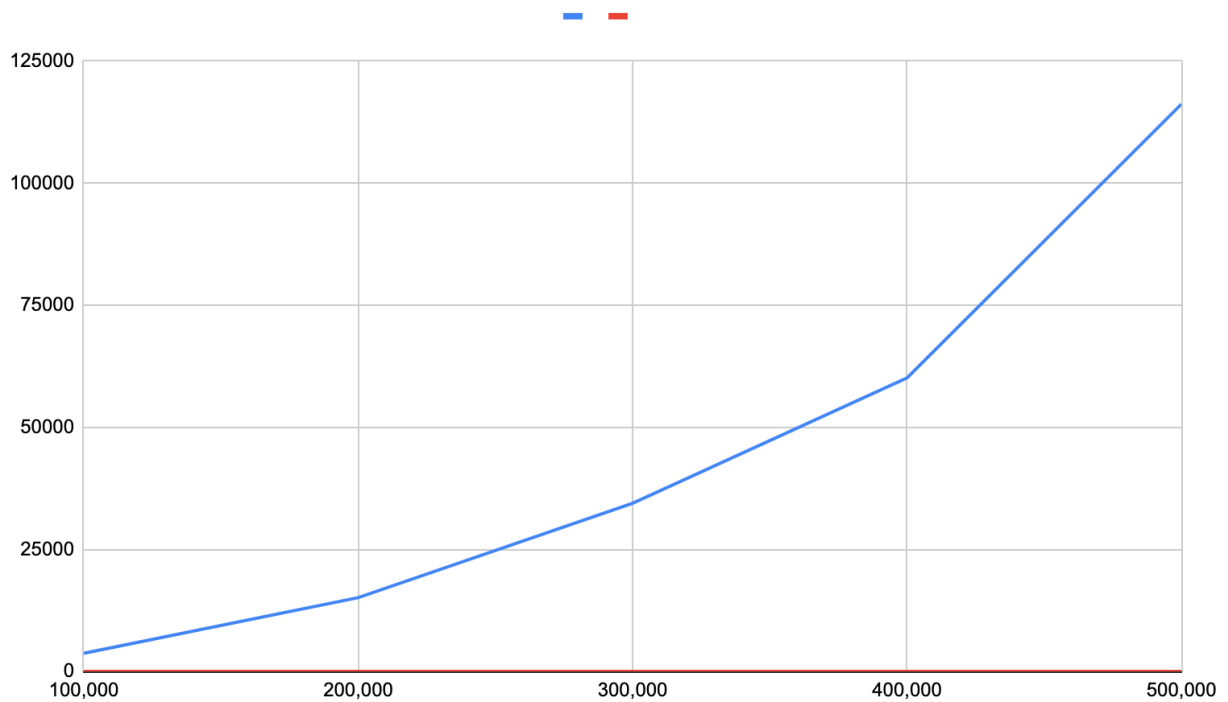
N	LinkedList	ArrayList
10,000	36.8	0.6
20,000	142.2	0.8
30,000	328	0.6
40,000	588.2	0.8
50,000	926.2	1
60,000	1346.4	0.6
70,000	1828	0.4
80,000	2385.8	0.8
90,000	3023.4	0.6
100,000	3736.2	0.6

LinkedList and ArrayList



Averages from five executions with N from 100,000 to 500,000

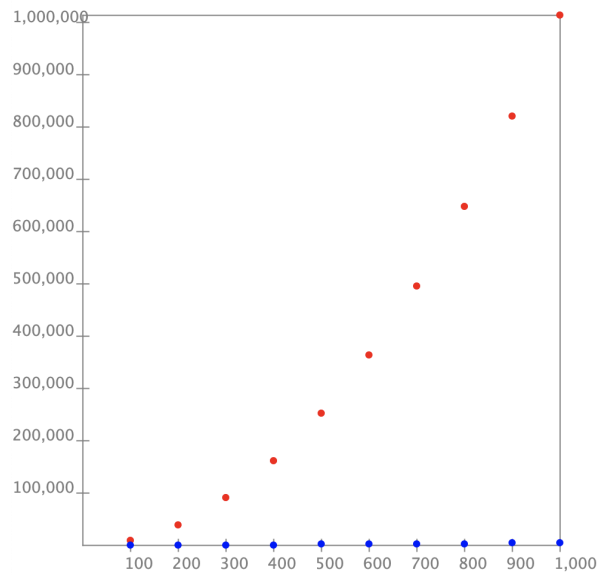
N	LinkedList	ArrayList
100,000	3736.2	0.6
200,000	15,113.20	1.2
300,000	34,467.20	1.6
400,000	60,114.40	1.6
500,000	116,221.80	1.6



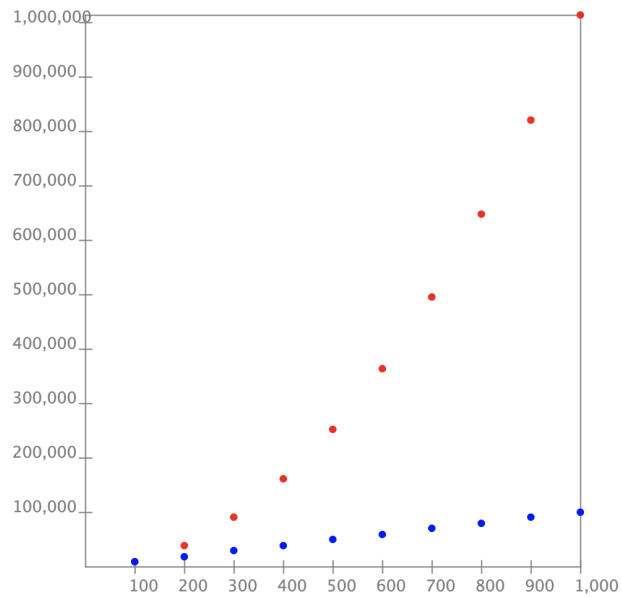
The time taken by the LinkedList grows at a greater than linear rate: When N doubles, the time grows four- to five-fold.

Ex. 4.9

$q=5$



$q=100$



Ex. 4.11

The curve of Algorithm B will rise above that of Algorithm A at $n=13$:

$n=12$ $f(n)=7104$ $g(n)=6912$
 $n=13$ $f(n)=8736$ $g(n)=8788$
 $n=14$ $f(n)=10612$ $g(n)=10976$

Ex. 4.12

Starting with an array of size 1, **10 doublings** will be required to insert 1,024 elements.

In general, the number of doublings for n items equals log to the base of 2 of n : **$\log_2 n$**

Ex. 4.14

With the optimization, in which the inner loop starts from the current item of the outer loop ($j=i$), the exact number of comparisons in terms of n will be

$$n * n/2 = n^2/2$$

In Big-Oh notation, this is equivalent to

$$O(n^2)$$

Ex. 4.15

The total amount of work is

$$n^2/2 - n/2 + n - 1$$

or

$$O(n^2)$$

Ex. 4.19

1. The length of the tour that is already in place: **50.22213534438458**
2. The length of the tour that I think is the shortest: **30.33029403737908**
3. Why the for-loop in main() goes up to n-2:

```
for (int i=0; i<n-1; i++) {  
    DrawTool.drawLine (x[tour[i]], y[tour[i]], x[tour[i+1]], y[tour[i+1]]);  
    tourLength += distance (x[tour[i]], y[tour[i]], x[tour[i+1]], y[tour[i+1]]);  
}
```

The for-loop works with i and with i+1. So, if i reached the value of the last index, i+1 would throw an exception LOOP OUT OF BOUND. Therefore, i should not be increased beyond n-2. The distance from the last point to point at [0] is added outside the for-loop.

Ex. 4.20

1. Number of tours evaluated: **40,322**
2. Best route and its length: **0 5 1 6 7 2 3 4; Length=29.492571697547458**
3. What does the number of tours evaluated have to do with n! (n factorial)?

$8! = 40,320$. This is the number of evaluations performed inside **recursiveFindTour()**. The remaining two evaluations are called from **findTour()**, one before the call to recursiveFindTour() and one after. Conclusion: This algorithm involves a factorial growth of n.