

Group Assignment 1

Taylor Dinkins

Due: Friday, July 12 at 11:59PM

You are encouraged to work in groups of up to three students. Only one member of each group should submit the group's work to Canvas, including the **project report as a pdf** and the code that implements the algorithms. The report should have **all member's names included**. You may use any language you choose to implement your algorithms. **No questions about this assignment will be answered after the due date above.**

For this project, you will design, implement and analyze (both experimentally and mathematically) *three* algorithms for the maximum subarray problem:

Given array of small integers $a[1, \dots, n]$ (that contains at least one positive integer), compute

$$\max_{i \leq j} \sum_{k=i}^j a[k]$$

For example, $\text{MAXSUBARRAY}([31, -41, \mathbf{59, 26}, -\mathbf{53, 58, 97}, -93, -23, 84]) = 187$

Description of the algorithms

Your three algorithms are to be based on these ideas:

Algorithm 1: Enumeration Loop over each pair of indices $i \leq j$ and compute the sum, $\sum_{k=i}^j a[k]$. Keep the best sum you have found so far.

Algorithm 2: Better Enumeration Notice that in the previous algorithm, the same sum is computed many times. In particular, notice that $\sum_{k=i}^j a[k]$ can be computed from $\sum_{k=i}^{j-1} a[k]$ in $O(1)$ time, rather than starting from scratch. Write a new version of the first algorithm that takes advantage of this observation.

Algorithm 3: Dynamic Programming Your dynamic programming algorithm should be based on the following idea:

- The maximum subarray either uses the last element in the input array, or it doesn't.

Describe the solution to the maximum subarray problem recursively and mathematically based on the above idea.

See pages 8-13 of <https://web.engr.oregonstate.edu/~glencora/wiki/uploads/max-subarray.pdf> for more details on these algorithms.

Testing for correctness You can test the correctness of your implementations using the test sets provided here: <http://www.eecs.orst.edu/~glencora/cs325/mstest.txt> The file has one test case per line (10 cases each with 100 entries). A line corresponding to the example above would be:

[31, -41, 59, 26, -53, 58, 97, -93, -23, 84], 187

with the input array followed by the sum of the maximum subarray.

Project report

Your typeset report **must** include:

Pseudocode Give **pseudocode** for each of the algorithms. Please try to follow the pseudocode suggestions in class (ie, not programming language specific). Your pseudocode should make clear how many times your algorithm will

- add *two* numbers together
- take the max of *two* numbers

Note that the pseudocode in the above-linked pdf does not do this.

Run-time analysis For each algorithm, express the number of + (addition) and max (between two numbers) operations that it makes for an input array of n numbers *as a sum* (e.g. $\sum_{i=1}^n \sum_{j=i}^{n-1} 2i + 3$). (Hint: think **for loop** ranges). Give asymptotic bounds for each. (That is, you should give three sums and three asymptotic bounds, one sum and asymptotic bound for each algorithm.)

Experimental run-time analysis For the experimental analysis you will plot running times as a function of input size. Every programming language should provide access to a clock (not necessarily in seconds). For example, in Python we have the **time** library. Run each of your algorithms on input arrays of size 100, 200, 300, ..., 900 and 1000, 2000, 3000, ..., 9000 (that is, you should have 18 data points for each algorithm). The first enumerative algorithm may be frustratingly slow, so you may compute running times for sizes 100, 200, 300, ..., 900.

To do this, generate random instances using a random number generator as provided by your programming language. **Remember to include both positive and negative numbers!** For each data point, you should take the average of a small number (say, 10) runs to smooth out any noise. For example, for the first data point, you will do something like:

```
for i = 1:10
    A = random array with 100 entries
    start clock
    maxsubarray(A)
    pause clock
return elapsed time divided by 10
```

Note that you should not include the time to generate the instance. Recall that this helps to account for the distribution over the inputs - that is, this will better represent the algorithm run-time rather than that of just receiving easy/hard input cases. This is just averaging out the time over 10 runs.

Plot the running times as a function of input size for each algorithm in a *single plot*. Label your plot (axes, title, etc.). Please put input size on the **x** axis, and run-time on the **y** axis. Additionally, please submit **3** additional plots, that include the run-time for each algorithm (as above), plotted with the theoretical run-time of the algorithm that you calculated in the **Run-time analysis** section. Discuss any discrepancies between the experimental and theoretical running times.

Deliverables

- Code for 3 implementations
- Project report described, with 1 plot comparing your different implementation run-times, and 3 plots (1 for each implementation) comparing the run-time of the specific implementation to its theoretical run-time.