

Group Assignment 1

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January 15, 2015

Problem

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

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

Algorithm 1. *Enumeration*: Loop over each pair of indices $i \leq j$ and compute the sum $\sum_{k=i}^j a[k]$

Psuedocode

```
MaxSubArray(a[1, ..., n])
    max_sum = 0
    sum = 0
    for i = 1, ..., n
        for j = i, ..., n
            for k = i, ..., j
                sum = sum + a[k]
            if sum > max_sum
                max_sum = sum
    return max_sum
```

Run-time analysis

Number of operations

$$\sum_{i=1}^n \sum_{j=i}^n \sum_{k=i}^j a[k]$$

Asymptotic bounds

$$(\mathcal{O}(n^2) \times (\mathcal{O}(n)\text{time to compute each sum})) = \mathcal{O}(n^3)$$

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 $\mathcal{O}(1)$ time, rather than starting from scratch. Write a new version of the first algorithm that takes advantage of this observation.

Pseudocode

```
MaxSubArray(a[1, ..., n])
    max_sum = 0
    for i = 1, ..., n
        sum = 0
        for j = i, ..., n
            sum = sum + a[j]
            if sum > max_sum
                max_sum = sum
    return max_sum
```

Run-time analysis

Number of operations

$$\sum_{i=1}^n \sum_{j=i}^n a[j]$$

Asymptotic bounds

$$((\mathcal{O}(n)\text{i-iterations}) \times (\mathcal{O}(n)\text{j-iterations}) \times (\mathcal{O}(n)\text{time to update sum})) = \mathcal{O}(n^2)$$

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.

Recursive Formula

$$MaxSubArray(a[1, \dots, n]) = \begin{cases} MaxSubArray(a[1, \dots, \frac{n}{2}]) \\ MaxSubArray(a[\frac{n}{2}, \dots, n]) \\ MaxSuffix(a[1, \dots, \frac{n}{2}]) + MaxPrefix(a[\frac{n}{2}, \dots, n]) \end{cases}$$

MaxSuffix(a[low, ..., mid])

```
max = 0
sum = 0
for i = mid, ..., low
    sum = sum + a[i]
    if sum > max
        max = sum
return max
```

MaxPrefix(a[mid, ..., high])

```
max = 0
sum = 0
for i = mid, ..., high
    sum = sum + a[i]
    if sum > max
        max = sum
return max
```

Pseudocode

MaxSubArray(a[1, ..., n])

```
max = 0
sum = 0
for i = 1, ..., n
    sum = sum + a[i]
    if sum < 0
        sum = 0
    if sum > max
        max = sum
return max
```

Run-time analysis

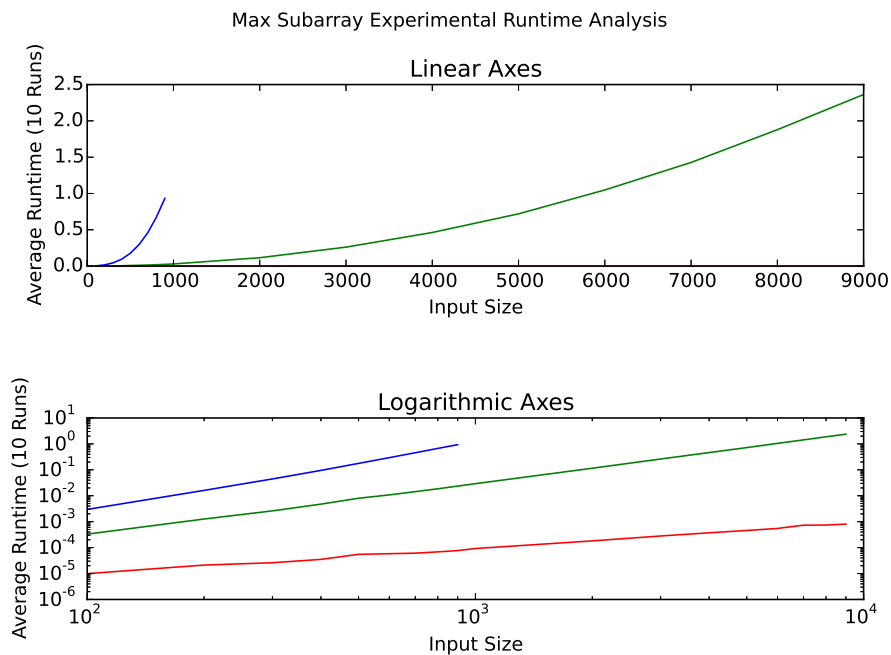
Number of operations

$$\sum_{i=1}^n a[i]$$

Asymptotic bounds

$$(\mathcal{O}(n)\text{i-iterations}) \times (\mathcal{O}(c)\text{to compute sum}) = \mathcal{O}(n)$$

1 Graphs



Slope comparison

During our experiments we obtained the following slopes for our algorithms:

- Algorithm 1: 2.6283
- Algorithm 2: 1.9803
- Algorithm 3: 0.9920

This would give us the following runtime complexity for each algorithm:

- Algorithm 1: $\theta(n^{2.6283})$
- Algorithm 2: $\theta(n^{1.9803})$
- Algorithm 3: $\theta(n^{0.9920})$

These values are extremely close to what we would have expected. The only algorithm which had a value that was not completely accurate was algorithm 1. This can be partially attributed to the fact that algorithm 1 was the only algorithm which was not tested with sets greater than 900 elements.