

3. Examples

Show Correctness, Recursion and Recurrences
[References to literature at the examples]

3.1 Ancient Egyptian Multiplication

Ancient Egyptian Multiplication– Example on how to show correctness of algorithms.

Ancient Egyptian Multiplication

3

Compute $11 \cdot 9$

$$11 \mid 9$$

$$9 \mid 11$$

³Also known as russian multiplication

Ancient Egyptian Multiplication

3

Compute $11 \cdot 9$

$$11 \mid 9$$

$$9 \mid 11$$

1. Double left, integer division by 2
on the right

³Also known as russian multiplication

Ancient Egyptian Multiplication

3

Compute $11 \cdot 9$

$$\begin{array}{r|l} 11 & 9 \\ 22 & 4 \end{array}$$

$$\begin{array}{r|l} 9 & 11 \\ 18 & 5 \end{array}$$

1. Double left, integer division by 2
on the right

³Also known as russian multiplication

Ancient Egyptian Multiplication

3

Compute $11 \cdot 9$

11	9
22	4
44	2

9	11
18	5
36	2

1. Double left, integer division by 2
on the right

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Ancient Egyptian Multiplication

3

Compute $11 \cdot 9$

11	9	9	11
22	4	18	5
44	2	36	2
88	1	72	1

1. Double left, integer division by 2 on the right
2. Even number on the right \Rightarrow eliminate row.

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1. Double left, integer division by 2 on the right
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Ancient Egyptian Multiplication

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Compute $11 \cdot 9$

11	9
22	4
44	2
88	1
99	-

9	11
18	5
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72	1
99	-

1. Double left, integer division by 2 on the right
2. Even number on the right \Rightarrow eliminate row.
3. Add remaining rows on the left.

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Advantages

- Short description, easy to grasp
- Efficient to implement on a computer: double = left shift, divide by 2 = right shift

left shift $9 = 01001_2 \rightarrow 10010_2 = 18$
right shift $9 = 01001_2 \rightarrow 00100_2 = 4$

Questions

- For which kind of inputs does the algorithm deliver a correct result (in finite time)?
- How do you prove its correctness?
- What is a good measure for Efficiency?

The Essentials

If $b > 1, a \in \mathbb{Z}$, then:

$$a \cdot b = \begin{cases} 2a \cdot \frac{b}{2} & \text{falls } b \text{ gerade,} \\ a + 2a \cdot \frac{b-1}{2} & \text{falls } b \text{ ungerade.} \end{cases}$$

Termination

$$a \cdot b = \begin{cases} a & \text{falls } b = 1, \\ 2a \cdot \frac{b}{2} & \text{falls } b \text{ gerade,} \\ a + 2a \cdot \frac{b-1}{2} & \text{falls } b \text{ ungerade.} \end{cases}$$

Recursively, Functional

$$f(a, b) = \begin{cases} a & \text{falls } b = 1, \\ f(2a, \frac{b}{2}) & \text{falls } b \text{ gerade,} \\ a + f(2a, \frac{b-1}{2}) & \text{falls } b \text{ ungerade.} \end{cases}$$

Implemented as a function

```
// pre: b>0
// post: return a*b
int f(int a, int b){
    if(b==1)
        return a;
    else if (b%2 == 0)
        return f(2*a, b/2);
    else
        return a + f(2*a, (b-1)/2);
}
```

Correctnes: Mathematical Proof

$$f(a, b) = \begin{cases} a & \text{if } b = 1, \\ f(2a, \frac{b}{2}) & \text{if } b \text{ even,} \\ a + f(2a \cdot \frac{b-1}{2}) & \text{if } b \text{ odd.} \end{cases}$$

Remaining to show: $f(a, b) = a \cdot b$ for $a \in \mathbb{Z}, b \in \mathbb{N}^+$.

Correctness: Mathematical Proof by Induction

Let $a \in \mathbb{Z}$, to show $f(a, b) = a \cdot b \quad \forall b \in \mathbb{N}^+$.

Base clause: $f(a, 1) = a = a \cdot 1$

Hypothesis: $f(a, b') = a \cdot b' \quad \forall 0 < b' \leq b$

Step: $f(a, b') = a \cdot b' \quad \forall 0 < b' \leq b \stackrel{!}{\Rightarrow} f(a, b + 1) = a \cdot (b + 1)$

$$f(a, b + 1) = \begin{cases} f(2a, \underbrace{\frac{b+1}{2}}_{0 < \cdot \leq b}) \stackrel{i.H.}{=} a \cdot (b + 1) & \text{if } b > 0 \text{ odd,} \\ a + f(2a, \underbrace{\frac{b}{2}}_{0 < \cdot < b}) \stackrel{i.H.}{=} a + a \cdot b & \text{if } b > 0 \text{ even.} \end{cases}$$



[Code Transformations: End Recursion]

The recursion can be written as *end recursion*

```
// pre: b>0  
// post: return a*b  
int f(int a, int b){  
    if(b==1)  
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    else  
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}
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```
// pre: b>0  
// post: return a*b  
int f(int a, int b){  
    if(b==1)  
        return a;  
    int z=0;  
    if (b%2 != 0){  
        --b;  
        z=a;  
    }  
    return z + f(2*a, b/2);  
}
```

[Code-Transformation: End-Recursion \Rightarrow Iteration]

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        z=a;
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}
```



```
int f(int a, int b) {
    int res = 0;
    while (b != 1) {
        int z = 0;
        if (b % 2 != 0){
            --b;
            z = a;
        }
        res += z;
        a *= 2; // neues a
        b /= 2; // neues b
    }
    res += a; // Basisfall b=1
    return res;
}
```

[Code-Transformation: Simplify]

```
int f(int a, int b) {  
    int res = 0;  
    while (b != 1) {  
        int z = 0;  
        if (b % 2 != 0){  
            --b; → Teil der Division  
            z = a; → Direkt in res  
        }  
        res += z;  
        a *= 2;  
        b /= 2;  
    }  
    res += a; → in den Loop  
    return res;  
}
```



```
// pre: b>0  
// post: return a*b  
int f(int a, int b) {  
    int res = 0;  
    while (b > 0) {  
        if (b % 2 != 0)  
            res += a;  
        a *= 2;  
        b /= 2;  
    }  
    return res;  
}
```

Correctness: Reasoning using Invariants!

```
// pre: b>0
// post: return a*b
int f(int a, int b) {                      Sei  $x := a \cdot b$ .
    int res = 0;
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            res += a;
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        }
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```

Sei $x := a \cdot b$.

here: $x = a \cdot b + res$

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here: $x = a \cdot b + res$ und $b = 0$

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 b even

here: $x = a \cdot b + res$

here: $x = a \cdot b + res$ und $b = 0$

Also $res = x$.

Conclusion

The expression $a \cdot b + res$ is an **invariant**

- Values of a, b, res change but the invariant remains basically unchanged: The invariant is only temporarily discarded by some statement but then re-established. If such short statement sequences are considered atomic, the value remains indeed invariant
- In particular the loop contains an invariant, called *loop invariant* and it operates there like the induction step in induction proofs.
- Invariants are obviously powerful tools for proofs!

3.2 Fast Integer Multiplication

[Ottman/Widmayer, Kap. 1.2.3]

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \\ 6 \quad 2 \\ \cdot \quad 3 \quad 7 \\ \hline 1 \quad 4 \end{array} \qquad \left| \begin{array}{l} d \\ 7 \\ \hline d \cdot b \end{array} \right.$$

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \\ 6 \quad 2 \\ \cdot \quad 3 \quad 7 \\ \hline 1 \quad 4 \\ 4 \quad 2 \end{array} \quad \left| \begin{array}{l} d \cdot b \\ d \cdot a \end{array} \right.$$

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \\ 6 \quad 2 \\ \cdot \quad 3 \quad 7 \\ \hline 1 \quad 4 \\ 4 \quad 2 \\ 6 \end{array} \quad \begin{array}{l} d \cdot b \\ d \cdot a \\ c \cdot b \end{array}$$

Example 2: Multiplication of large Numbers

Primary school:

a	b	c	d	
6	2	.	3	7
		1	4	$d \cdot b$
	4	2		$d \cdot a$
		6		$c \cdot b$
	1	8		$c \cdot a$

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \quad & c \quad d \\ 6 \quad 2 \quad . \quad 3 \quad 7 \\ \hline & & 1 & 4 & | d \cdot b \\ & & 4 & 2 & | d \cdot a \\ & & 6 & & | c \cdot b \\ & 1 & 8 & & | c \cdot a \\ \hline = & 2 & 2 & 9 & 4 \end{array}$$

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \quad & c \quad d \\ 6 \quad 2 & \cdot \quad 3 \quad 7 \\ \hline & 1 \quad 4 & d \cdot b \\ & 4 \quad 2 & d \cdot a \\ & 6 & c \cdot b \\ 1 \quad 8 & & c \cdot a \\ \hline = & 2 \quad 2 \quad 9 \quad 4 & \end{array}$$

$2 \cdot 2 = 4$ single-digit multiplications.

Example 2: Multiplication of large Numbers

Primary school:

$$\begin{array}{r} a \quad b \quad & c \quad d \\ 6 \quad 2 & \cdot \quad 3 \quad 7 \\ \hline & 1 \quad 4 & d \cdot b \\ & 4 \quad 2 & d \cdot a \\ & 6 & c \cdot b \\ 1 \quad 8 & & c \cdot a \\ \hline = & 2 \quad 2 \quad 9 \quad 4 & \end{array}$$

$2 \cdot 2 = 4$ single-digit multiplications. \Rightarrow Multiplication of two n -digit numbers: n^2 single-digit multiplications

Observation

$$ab \cdot cd = (10 \cdot a + b) \cdot (10 \cdot c + d)$$

Observation

$$\begin{aligned} ab \cdot cd &= (10 \cdot a + b) \cdot (10 \cdot c + d) \\ &= 100 \cdot \textcolor{red}{a} \cdot \textcolor{red}{c} + 10 \cdot \textcolor{red}{a} \cdot \textcolor{red}{c} \\ &\quad + 10 \cdot \textcolor{blue}{b} \cdot \textcolor{blue}{d} + \textcolor{blue}{b} \cdot \textcolor{blue}{d} \\ &\quad + 10 \cdot (a - b) \cdot (d - c) \end{aligned}$$

Improvement?

$$\begin{array}{r} a \quad b \quad | \quad c \quad d \\ 6 \quad 2 \quad . \quad 3 \quad 7 \\ \hline 1 \quad 4 \quad | \quad \color{red}{d \cdot b} \end{array}$$

Improvement?

$$\begin{array}{r} \begin{array}{cccc|c} a & b & c & d \\ 6 & 2 & \cdot & 3 & 7 \end{array} \\ \hline \begin{array}{cc|c} & & \\ 1 & 4 & d \cdot b \\ 1 & 4 & d \cdot b \end{array} \end{array}$$

Improvement?

$$\begin{array}{r} \begin{array}{cccc|c} a & b & c & d \\ 6 & 2 & \cdot & 3 & 7 \end{array} \\ \hline \begin{array}{ccc|c} & 1 & 4 & d \cdot b \\ 1 & 4 & & d \cdot b \\ 1 & 6 & & (a - b) \cdot (d - c) \end{array} \end{array}$$

Improvement?

a	b	c	d	
6	2	.	3	7
		1	4	$d \cdot b$
	1	4		$d \cdot b$
	1	6		$(a - b) \cdot (d - c)$
	1	8		$c \cdot a$

Improvement?

a	b	c	d	
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Improvement?

$$\begin{array}{r} \begin{array}{cccc|c} a & b & c & d \\ 6 & 2 & \cdot & 3 & 7 \end{array} \\ \hline \begin{array}{ccccc} 1 & 4 & d \cdot b \\ 1 & 4 & d \cdot b \\ 1 & 6 & (a - b) \cdot (d - c) \\ 1 & 8 & c \cdot a \\ 1 & 8 & c \cdot a \end{array} \\ \hline = \quad \begin{array}{ccccc} 2 & 2 & 9 & 4 \end{array} \end{array}$$

Improvement?

$$\begin{array}{r} a \quad b \quad c \quad d \\ 6 \quad 2 \quad . \quad 3 \quad 7 \\ \hline & 1 \quad 4 & d \cdot b \\ & 1 \quad 4 & d \cdot b \\ & 1 \quad 6 & (a - b) \cdot (d - c) \\ & 1 \quad 8 & c \cdot a \\ \hline & 1 \quad 8 & c \cdot a \\ = & 2 \quad 2 \quad 9 \quad 4 & \end{array}$$

→ 3 single-digit multiplications.

Large Numbers

$$6237 \cdot 5898 = \underbrace{62}_{a'} \underbrace{37}_{b'} \cdot \underbrace{58}_{c'} \underbrace{98}_{d'}$$

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Recursive / inductive application: compute $a' \cdot c'$, $a' \cdot d'$, $b' \cdot c'$ and $c' \cdot d'$ as shown above.

Large Numbers

$$6237 \cdot 5898 = \underbrace{62}_{a'} \underbrace{37}_{b'} \cdot \underbrace{58}_{c'} \underbrace{98}_{d'}$$

Recursive / inductive application: compute $a' \cdot c'$, $a' \cdot d'$, $b' \cdot c'$ and $c' \cdot d'$ as shown above.

→ 3 · 3 = 9 instead of 16 single-digit multiplications.

Generalization

Assumption: two numbers with n digits each, $n = 2^k$ for some k .

$$\begin{aligned}(10^{n/2}a + b) \cdot (10^{n/2}c + d) &= 10^n \cdot a \cdot c + 10^{n/2} \cdot a \cdot c \\&\quad + 10^{n/2} \cdot b \cdot d + b \cdot d \\&\quad + 10^{n/2} \cdot (a - b) \cdot (d - c)\end{aligned}$$

Recursive application of this formula: algorithm by Karatsuba and Ofman (1962).

Algorithm Karatsuba Ofman

Input: Two positive integers x and y with n decimal digits each: $(x_i)_{1 \leq i \leq n}$,
 $(y_i)_{1 \leq i \leq n}$

Output: Product $x \cdot y$

```
if  $n = 1$  then  
| return  $x_1 \cdot y_1$ 
```

```
else
```

```
| Let  $m := \lfloor \frac{n}{2} \rfloor$ 
```

```
| Divide  $a := (x_1, \dots, x_m)$ ,  $b := (x_{m+1}, \dots, x_n)$ ,  $c := (y_1, \dots, y_m)$ ,  
|  $d := (y_{m+1}, \dots, y_n)$ 
```

```
| Compute recursively  $A := a \cdot c$ ,  $B := b \cdot d$ ,  $C := (a - b) \cdot (d - c)$ 
```

```
| Compute  $R := 10^n \cdot A + 10^m \cdot A + 10^m \cdot B + B + 10^m \cdot C$ 
```

```
| return  $R$ 
```

Analysis

$M(n)$: Number of single-digit multiplications.

Recursive application of the algorithm from above \Rightarrow recursion equality:

$$M(2^k) = \begin{cases} 1 & \text{if } k = 0, \\ 3 \cdot M(2^{k-1}) & \text{if } k > 0. \end{cases} \quad (\text{R})$$

Iterative Substitution

Iterative substitution of the recursion formula in order to guess a solution of the recursion formula:

$$M(2^k) = 3 \cdot M(2^{k-1})$$

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$$M(2^k) = 3 \cdot M(2^{k-1}) = 3 \cdot 3 \cdot M(2^{k-2}) = 3^2 \cdot M(2^{k-2})$$

Iterative Substitution

Iterative substitution of the recursion formula in order to guess a solution of the recursion formula:

$$\begin{aligned}M(2^k) &= 3 \cdot M(2^{k-1}) = 3 \cdot 3 \cdot M(2^{k-2}) = 3^2 \cdot M(2^{k-2}) \\&= \dots \\&\stackrel{!}{=} 3^k \cdot M(2^0) = 3^k.\end{aligned}$$

Proof: induction

Hypothesis $H(k)$:

$$M(2^k) = F(k) := 3^k. \quad (\mathsf{H})$$

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Claim:

$H(k)$ holds for all $k \in \mathbb{N}_0$.

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$H(k)$ holds for all $k \in \mathbb{N}_0$.

Base clause $k = 0$:

$$M(2^0) \stackrel{R}{=} 1 = F(0). \quad \checkmark$$

Proof: induction

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Claim:

$H(k)$ holds for all $k \in \mathbb{N}_0$.

Base clause $k = 0$:

$$M(2^0) \stackrel{R}{=} 1 = F(0). \quad \checkmark$$

Induction step $H(k) \Rightarrow H(k + 1)$:

$$M(2^{k+1}) \stackrel{R}{=} 3 \cdot M(2^k) \stackrel{H(k)}{=} 3 \cdot F(k) = 3^{k+1} = F(k + 1). \quad \checkmark$$



Comparison

Traditionally n^2 single-digit multiplications.

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Karatsuba/Ofman:

$$M(n) = 3^{\log_2 n} = (2^{\log_2 3})^{\log_2 n} = 2^{\log_2 3 \log_2 n} = n^{\log_2 3} \approx n^{1.58}.$$

Comparison

Traditionally n^2 single-digit multiplications.

Karatsuba/Ofman:

$$M(n) = 3^{\log_2 n} = (2^{\log_2 3})^{\log_2 n} = 2^{\log_2 3 \log_2 n} = n^{\log_2 3} \approx n^{1.58}.$$

Example: number with 1000 digits: $1000^2/1000^{1.58} \approx 18$.

Best possible algorithm?

We only know the upper bound $n^{\log_2 3}$.

There are (for large n) practically relevant algorithms that are faster.

Example: Schönhage-Strassen algorithm (1971) based on fast Fouriertransformation with running time $\mathcal{O}(n \log n \cdot \log \log n)$. The best upper bound is not known.⁴

Lower bound: n . Each digit has to be considered at least once.

⁴In March 2019, David Harvey and Joris van der Hoeven have shown an $\mathcal{O}(n \log n)$ algorithm that is practically irrelevant yet. It is conjectured, but yet unproven that this is the best lower bound we can get.

3.3 Maximum Subarray Problem

Algorithm Design – Maximum Subarray Problem [Ottman/Widmayer, Kap. 1.3]
Divide and Conquer [Ottman/Widmayer, Kap. 1.2.2. S.9; Cormen et al, Kap. 4-4.1]

Algorithm Design

Inductive development of an algorithm: partition into subproblems, use solutions for the subproblems to find the overall solution.

Goal: development of the asymptotically most efficient (correct) algorithm.

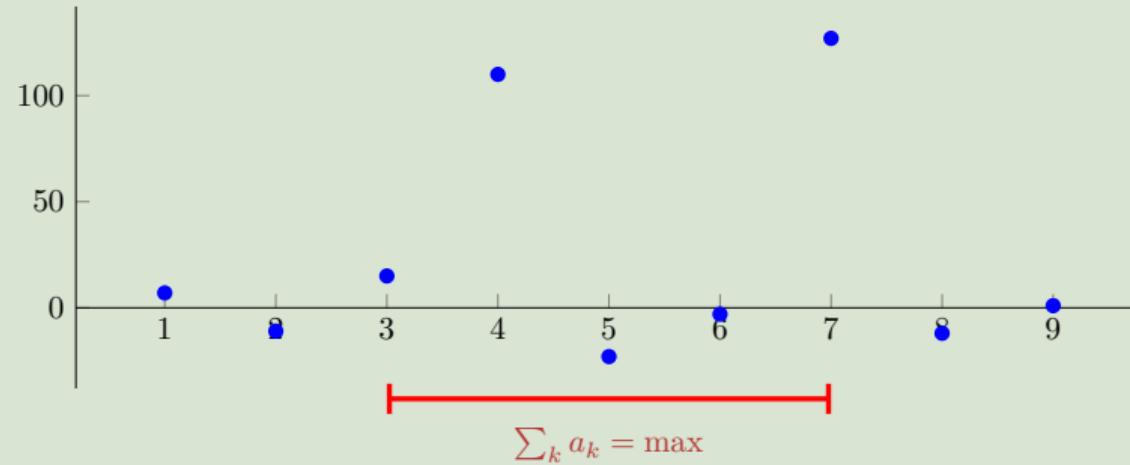
Efficiency towards run time costs (# fundamental operations) or /and memory consumption.

Maximum Subarray Problem

Given: an array of n real numbers (a_1, \dots, a_n) .

Wanted: interval $[i, j]$, $1 \leq i \leq j \leq n$ with maximal positive sum $\sum_{k=i}^j a_k$.

$$a = (7, -11, 15, 110, -23, -3, 127, -12, 1)$$



Naive Maximum Subarray Algorithm

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: I, J such that $\sum_{k=I}^J a_k$ maximal.

$M \leftarrow 0; I \leftarrow 1; J \leftarrow 0$

for $i \in \{1, \dots, n\}$ **do**

for $j \in \{i, \dots, n\}$ **do**

$m = \sum_{k=i}^j a_k$

if $m > M$ **then**

$M \leftarrow m; I \leftarrow i; J \leftarrow j$

return I, J

Analysis

Theorem 3

The naive algorithm for the Maximum Subarray problem executes $\Theta(n^3)$ additions.

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The naive algorithm for the Maximum Subarray problem executes $\Theta(n^3)$ additions.

Proof:

$$\begin{aligned} \sum_{i=1}^n \sum_{j=i}^n (j - i + 1) &= \sum_{i=1}^n \sum_{j=0}^{n-i} (j + 1) = \sum_{i=1}^n \sum_{j=1}^{n-i+1} j = \sum_{i=1}^n \frac{(n - i + 1)(n - i + 2)}{2} \\ &= \sum_{i=0}^n \frac{i \cdot (i + 1)}{2} = \frac{1}{2} \left(\sum_{i=1}^n i^2 + \sum_{i=1}^n i \right) \\ &= \frac{1}{2} \left(\frac{n(2n + 1)(n + 1)}{6} + \frac{n(n + 1)}{2} \right) = \frac{n^3 + 3n^2 + 2n}{6} = \Theta(n^3). \end{aligned}$$



Observation

$$\sum_{k=i}^j a_k = \underbrace{\left(\sum_{k=1}^j a_k \right)}_{S_j} - \underbrace{\left(\sum_{k=1}^{i-1} a_k \right)}_{S_{i-1}}$$

Observation

$$\sum_{k=i}^j a_k = \underbrace{\left(\sum_{k=1}^j a_k \right)}_{S_j} - \underbrace{\left(\sum_{k=1}^{i-1} a_k \right)}_{S_{i-1}}$$

Prefix sums

$$S_i := \sum_{k=1}^i a_k.$$

Maximum Subarray Algorithm with Prefix Sums

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: I, J such that $\sum_{k=J}^J a_k$ maximal.

```
 $S_0 \leftarrow 0$ 
for  $i \in \{1, \dots, n\}$  do // prefix sum
     $S_i \leftarrow S_{i-1} + a_i$ 
 $M \leftarrow 0; I \leftarrow 1; J \leftarrow 0$ 
for  $i \in \{1, \dots, n\}$  do
    for  $j \in \{i, \dots, n\}$  do
         $m = S_j - S_{i-1}$ 
        if  $m > M$  then
             $M \leftarrow m; I \leftarrow i; J \leftarrow j$ 
```

Analysis

Theorem 4

The prefix sum algorithm for the Maximum Subarray problem conducts $\Theta(n^2)$ additions and subtractions.

Analysis

Theorem 4

The prefix sum algorithm for the Maximum Subarray problem conducts $\Theta(n^2)$ additions and subtractions.

Proof:

$$\sum_{i=1}^n 1 + \sum_{i=1}^n \sum_{j=i}^n 1 = n + \sum_{i=1}^n (n - i + 1) = n + \sum_{i=1}^n i = \Theta(n^2)$$

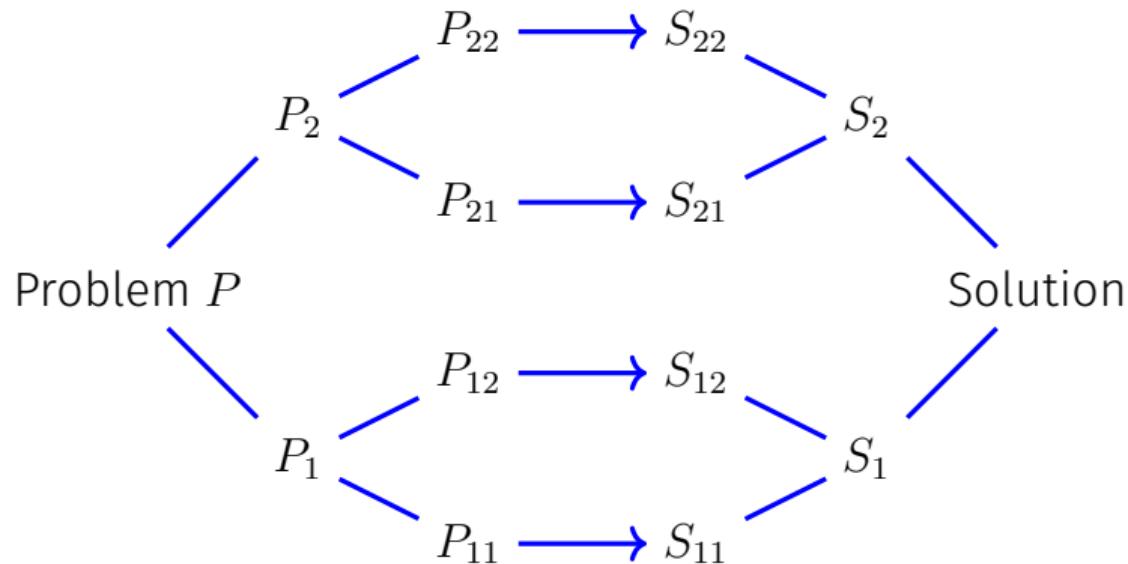


divide et impera

Divide and Conquer

Divide the problem into subproblems that contribute to the simplified computation of the overall problem.

divide et impera



Maximum Subarray – Divide

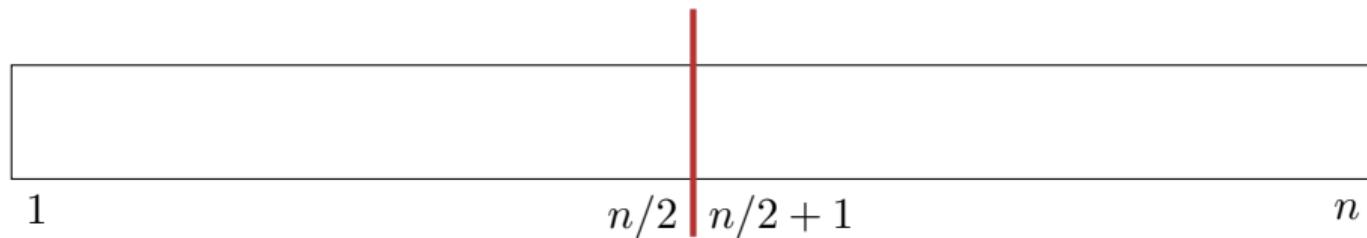
- Divide: Divide the problem into two (roughly) equally sized halves:
 $(a_1, \dots, a_n) = (a_1, \dots, a_{\lfloor n/2 \rfloor}, \quad a_{\lfloor n/2 \rfloor + 1}, \dots, a_n)$

Maximum Subarray – Divide

- Divide: Divide the problem into two (roughly) equally sized halves:
 $(a_1, \dots, a_n) = (a_1, \dots, a_{\lfloor n/2 \rfloor}, \quad a_{\lfloor n/2 \rfloor + 1}, \dots, a_n)$
- Simplifying assumption: $n = 2^k$ for some $k \in \mathbb{N}$.

Maximum Subarray – Conquer

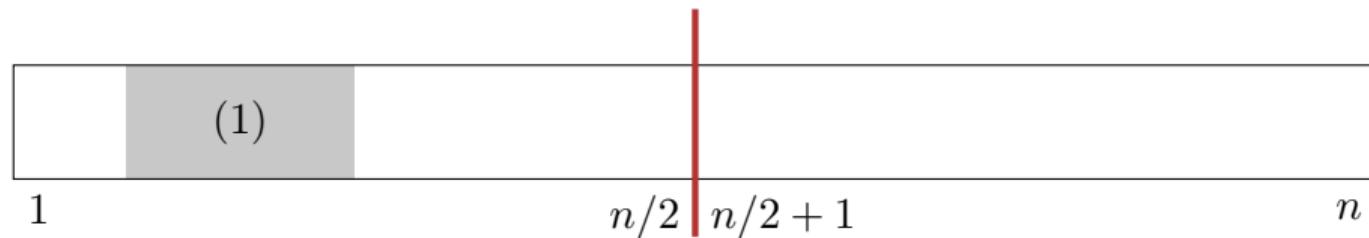
If i and j are indices of a solution \Rightarrow case by case analysis:



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

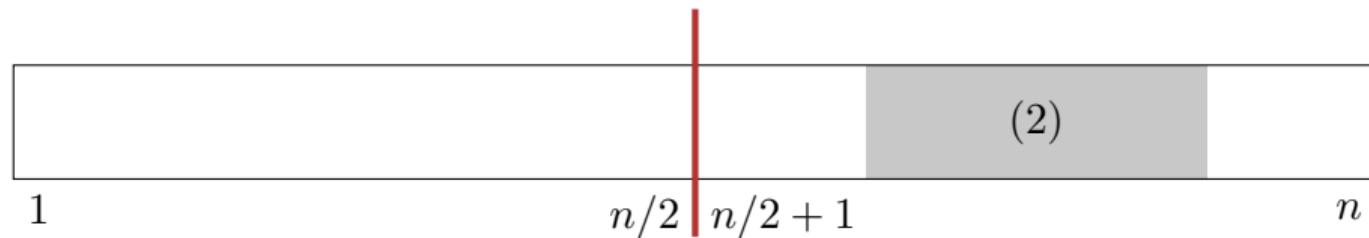
1. Solution in left half $1 \leq i \leq j \leq n/2$



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

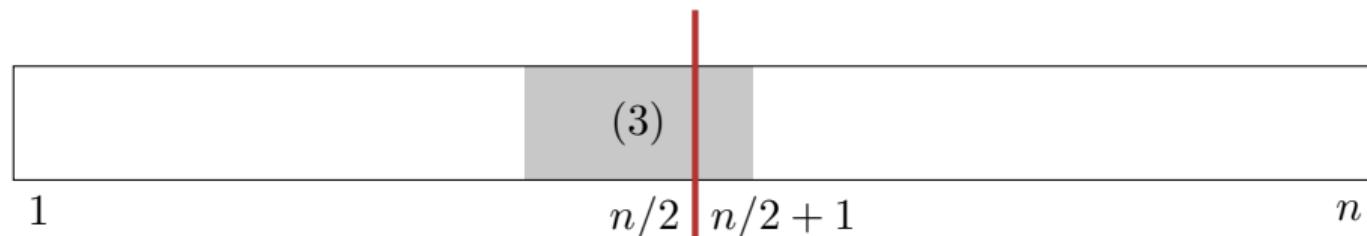
1. Solution in left half $1 \leq i \leq j \leq n/2$
2. Solution in right half $n/2 < i \leq j \leq n$



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

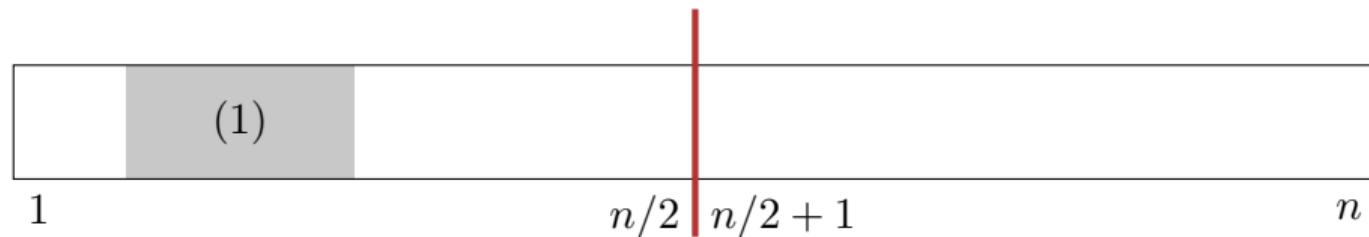
1. Solution in left half $1 \leq i \leq j \leq n/2$
2. Solution in right half $n/2 < i \leq j \leq n$
3. Solution in the middle $1 \leq i \leq n/2 < j \leq n$



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

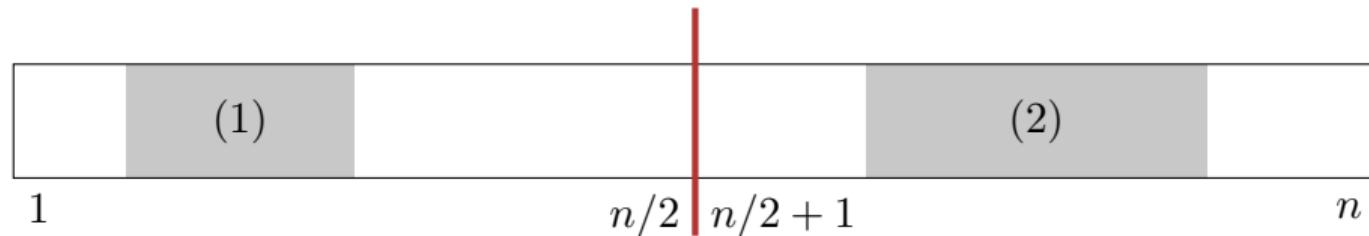
1. Solution in left half $1 \leq i \leq j \leq n/2 \Rightarrow$ Recursion (left half)
2. Solution in right half $n/2 < i \leq j \leq n$
3. Solution in the middle $1 \leq i \leq n/2 < j \leq n$



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

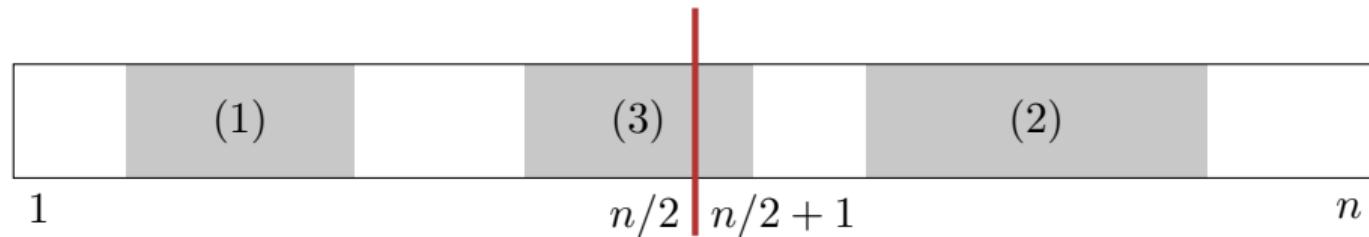
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3. Solution in the middle $1 \leq i \leq n/2 < j \leq n$



Maximum Subarray – Conquer

If i and j are indices of a solution \Rightarrow case by case analysis:

1. Solution in left half $1 \leq i \leq j \leq n/2 \Rightarrow$ Recursion (left half)
2. Solution in right half $n/2 < i \leq j \leq n \Rightarrow$ Recursion (right half)
3. Solution in the middle $1 \leq i \leq n/2 < j \leq n \Rightarrow$ Subsequent observation



Maximum Subarray – Observation

Assumption: solution in the middle $1 \leq i \leq n/2 < j \leq n$

$$S_{\max} = \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \sum_{k=i}^j a_k$$

Maximum Subarray – Observation

Assumption: solution in the middle $1 \leq i \leq n/2 < j \leq n$

$$S_{\max} = \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \sum_{k=i}^j a_k = \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \left(\sum_{k=i}^{n/2} a_k + \sum_{k=n/2+1}^j a_k \right)$$

Maximum Subarray – Observation

Assumption: solution in the middle $1 \leq i \leq n/2 < j \leq n$

$$\begin{aligned} S_{\max} &= \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \sum_{k=i}^j a_k = \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \left(\sum_{k=i}^{n/2} a_k + \sum_{k=n/2+1}^j a_k \right) \\ &= \max_{1 \leq i \leq n/2} \sum_{k=i}^{n/2} a_k + \max_{n/2 < j \leq n} \sum_{k=n/2+1}^j a_k \end{aligned}$$

Maximum Subarray – Observation

Assumption: solution in the middle $1 \leq i \leq n/2 < j \leq n$

$$\begin{aligned} S_{\max} &= \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \sum_{k=i}^j a_k = \max_{\substack{1 \leq i \leq n/2 \\ n/2 < j \leq n}} \left(\sum_{k=i}^{n/2} a_k + \sum_{k=n/2+1}^j a_k \right) \\ &= \max_{1 \leq i \leq n/2} \sum_{k=i}^{n/2} a_k + \max_{n/2 < j \leq n} \sum_{k=n/2+1}^j a_k \\ &= \max_{1 \leq i \leq n/2} \underbrace{S_{n/2} - S_{i-1}}_{\text{suffix sum}} + \max_{n/2 < j \leq n} \underbrace{S_j - S_{n/2}}_{\text{prefix sum}} \end{aligned}$$

Maximum Subarray Divide and Conquer Algorithm

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: Maximal $\sum_{k=i'}^{j'} a_k$.

if $n = 1$ **then**

return $\max\{a_1, 0\}$

else

 Divide $a = (a_1, \dots, a_n)$ in $A_1 = (a_1, \dots, a_{n/2})$ und $A_2 = (a_{n/2+1}, \dots, a_n)$

 Recursively compute best solution W_1 in A_1

 Recursively compute best solution W_2 in A_2

 Compute greatest suffix sum S in A_1

 Compute greatest prefix sum P in A_2

 Let $W_3 \leftarrow S + P$

return $\max\{W_1, W_2, W_3\}$

Analysis

Theorem 5

The divide and conquer algorithm for the maximum subarray sum problem conducts a number of $\Theta(n \log n)$ additions and comparisons.

Analysis

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: Maximal $\sum_{k=i'}^{j'} a_k$.

if $n = 1$ **then**

return $\max\{a_1, 0\}$

else

 Divide $a = (a_1, \dots, a_n)$ in $A_1 = (a_1, \dots, a_{n/2})$ und $A_2 = (a_{n/2+1}, \dots, a_n)$

 Recursively compute best solution W_1 in A_1

 Recursively compute best solution W_2 in A_2

 Compute greatest suffix sum S in A_1

 Compute greatest prefix sum P in A_2

 Let $W_3 \leftarrow S + P$

return $\max\{W_1, W_2, W_3\}$

Analysis

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: Maximal $\sum_{k=i'}^{j'} a_k$.

if $n = 1$ **then**

$\Theta(1)$ **return** $\max\{a_1, 0\}$

else

$\Theta(1)$ Divide $a = (a_1, \dots, a_n)$ in $A_1 = (a_1, \dots, a_{n/2})$ und $A_2 = (a_{n/2+1}, \dots, a_n)$

 Recursively compute best solution W_1 in A_1

 Recursively compute best solution W_2 in A_2

 Compute greatest suffix sum S in A_1

 Compute greatest prefix sum P in A_2

$\Theta(1)$ Let $W_3 \leftarrow S + P$

$\Theta(1)$ **return** $\max\{W_1, W_2, W_3\}$

Analysis

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: Maximal $\sum_{k=i'}^{j'} a_k$.

if $n = 1$ **then**

$\Theta(1)$ **return** $\max\{a_1, 0\}$

else

$\Theta(1)$ Divide $a = (a_1, \dots, a_n)$ in $A_1 = (a_1, \dots, a_{n/2})$ und $A_2 = (a_{n/2+1}, \dots, a_n)$

 Recursively compute best solution W_1 in A_1

 Recursively compute best solution W_2 in A_2

$\Theta(n)$ Compute greatest suffix sum S in A_1

$\Theta(n)$ Compute greatest prefix sum P in A_2

$\Theta(1)$ Let $W_3 \leftarrow S + P$

$\Theta(1)$ **return** $\max\{W_1, W_2, W_3\}$

Analysis

Input: A sequence of n numbers (a_1, a_2, \dots, a_n)

Output: Maximal $\sum_{k=i'}^{j'} a_k$.

if $n = 1$ **then**

$\Theta(1)$ **return** $\max\{a_1, 0\}$

else

$\Theta(1)$ Divide $a = (a_1, \dots, a_n)$ in $A_1 = (a_1, \dots, a_{n/2})$ und $A_2 = (a_{n/2+1}, \dots, a_n)$

$T(n/2)$ Recursively compute best solution W_1 in A_1

$T(n/2)$ Recursively compute best solution W_2 in A_2

$\Theta(n)$ Compute greatest suffix sum S in A_1

$\Theta(n)$ Compute greatest prefix sum P in A_2

$\Theta(1)$ Let $W_3 \leftarrow S + P$

$\Theta(1)$ **return** $\max\{W_1, W_2, W_3\}$

Analysis

Recursion equation

$$T(n) = \begin{cases} c & \text{if } n = 1 \\ 2T\left(\frac{n}{2}\right) + a \cdot n & \text{if } n > 1 \end{cases}$$

Analysis

Mit $n = 2^k$:

$$\overline{T}(k) := T(2^k) = \begin{cases} c & \text{if } k = 0 \\ 2\overline{T}(k-1) + a \cdot 2^k & \text{if } k > 0 \end{cases}$$

Solution:

$$\overline{T}(k) = 2^k \cdot c + \sum_{i=0}^{k-1} 2^i \cdot a \cdot 2^{k-i} = c \cdot 2^k + a \cdot k \cdot 2^k = \Theta(k \cdot 2^k)$$

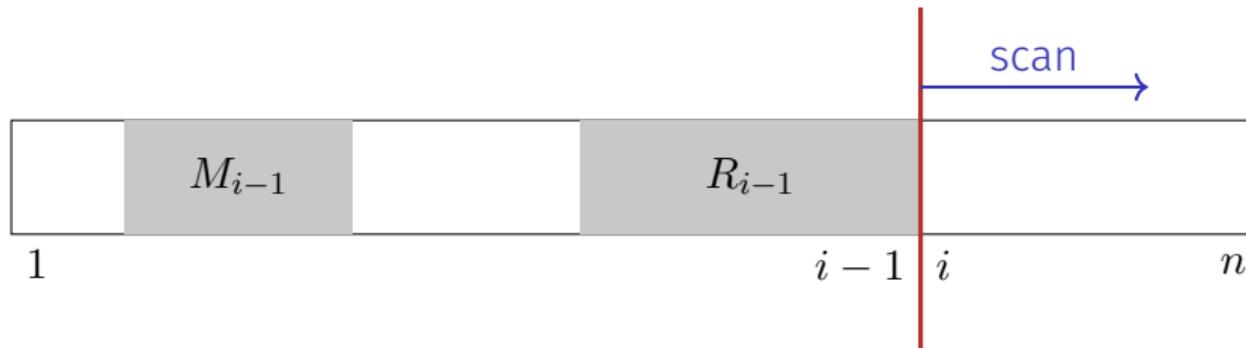
also

$$T(n) = \Theta(n \log n)$$



Maximum Subarray Sum Problem – Inductively

Assumption: maximal value M_{i-1} of the subarray sum is known for (a_1, \dots, a_{i-1}) ($1 < i \leq n$).



a_i : generates at most a better interval at the right bound (prefix sum).
 $R_{i-1} \Rightarrow R_i = \max\{R_{i-1} + a_i, 0\}$

Inductive Maximum Subarray Algorithm

Input: A sequence of n numbers (a_1, a_2, \dots, a_n) .

Output: $\max\{0, \max_{i,j} \sum_{k=i}^j a_k\}$.

$M \leftarrow 0$

$R \leftarrow 0$

for $i = 1 \dots n$ **do**

$R \leftarrow R + a_i$

if $R < 0$ **then**

$R \leftarrow 0$

if $R > M$ **then**

$M \leftarrow R$

return M ;

Analysis

Theorem 6

The inductive algorithm for the Maximum Subarray problem conducts a number of $\Theta(n)$ additions and comparisons.

Complexity of the problem?

Can we improve over $\Theta(n)$?

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Every correct algorithm for the Maximum Subarray Sum problem must consider each element in the algorithm.

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Every correct algorithm for the Maximum Subarray Sum problem must consider each element in the algorithm.

Assumption: the algorithm does not consider a_i .

1. The algorithm provides a solution including a_i . Repeat the algorithm with a_i so small that the solution must not have contained the point in the first place.

Complexity of the problem?

Can we improve over $\Theta(n)$?

Every correct algorithm for the Maximum Subarray Sum problem must consider each element in the algorithm.

Assumption: the algorithm does not consider a_i .

1. The algorithm provides a solution including a_i . Repeat the algorithm with a_i so small that the solution must not have contained the point in the first place.
2. The algorithm provides a solution not including a_i . Repeat the algorithm with a_i so large that the solution must have contained the point in the first place.

Complexity of the maximum Subarray Sum Problem

Theorem 7

The Maximum Subarray Sum Problem has Complexity $\Theta(n)$.

Proof: Inductive algorithm with asymptotic execution time $\mathcal{O}(n)$.

Every algorithm has execution time $\Omega(n)$.

Thus the complexity of the problem is $\Omega(n) \cap \mathcal{O}(n) = \Theta(n)$. ■

3.4 Appendix

Derivation and repetition of some mathematical formulas

Logarithms

$$\log_a y = x \Leftrightarrow a^x = y \quad (a > 0, y > 0)$$

$$\log_a(x \cdot y) = \log_a x + \log_a y$$

$$a^x \cdot a^y = a^{x+y}$$

$$\log_a \frac{x}{y} = \log_a x - \log_a y$$

$$\frac{a^x}{a^y} = a^{x-y}$$

$$\log_a x^y = y \log_a x$$

$$a^{x \cdot y} = (a^x)^y$$

$$\log_a n! = \sum_{i=1}^n \log i$$

$$\log_b x = \log_b a \cdot \log_a x$$

$$a^{\log_b x} = x^{\log_b a}$$

To see the last line, replace $x \rightarrow a^{\log_a x}$

Sums

$$\sum_{i=0}^n i = \frac{n \cdot (n + 1)}{2} \in \Theta(n^2)$$

Trick

$$\begin{aligned}\sum_{i=0}^n i &= \frac{1}{2} \left(\sum_{i=0}^n i + \sum_{i=0}^n n - i \right) = \frac{1}{2} \sum_{i=0}^n i + n - i \\ &= \frac{1}{2} \sum_{i=0}^n n = \frac{1}{2} (n + 1) \cdot n\end{aligned}$$

Sums

$$\sum_{i=0}^n i^2 = \frac{n \cdot (n+1) \cdot (2n+1)}{6}$$

Trick:

$$\sum_{i=1}^n i^3 - (i-1)^3 = \sum_{i=0}^n i^3 - \sum_{i=0}^{n-1} i^3 = n^3$$

$$\sum_{i=1}^n i^3 - (i-1)^3 = \sum_{i=1}^n i^3 - i^3 + 3i^2 - 3i + 1 = n - \frac{3}{2}n \cdot (n+1) + 3 \sum_{i=0}^n i^2$$

$$\Rightarrow \sum_{i=0}^n i^2 = \frac{1}{6}(2n^3 + 3n^2 + n) \in \Theta(n^3)$$

Can easily be generalized: $\sum_{i=1}^n i^k \in \Theta(n^{k+1})$.

Geometric Series

$$\sum_{i=0}^n \rho^i \stackrel{!}{=} \frac{1 - \rho^{n+1}}{1 - \rho}$$

$$\begin{aligned}\sum_{i=0}^n \rho^i \cdot (1 - \varrho) &= \sum_{i=0}^n \rho^i - \sum_{i=0}^n \rho^{i+1} = \sum_{i=0}^n \rho^i - \sum_{i=1}^{n+1} \rho^i \\ &= \rho^0 - \rho^{n+1} = 1 - \rho^{n+1}.\end{aligned}$$

For $0 \leq \rho < 1$:

$$\sum_{i=0}^{\infty} \rho^i = \frac{1}{1 - \rho}$$