

Analysis of Algorithms

- Why to analyze algorithms?
- Analyzing Control Structure
 - Sequencing
 - Let $P1$ and $P2$ be two fragments of an algorithm
 - Sequencing Rule
 - Maximum Rule
 - Can we always consider $P1$ and $P2$ independent?

Analysis of Algorithms

➤ Analyzing Control Structure

➤ For Loop

for $i \leftarrow 1$ to m do $P(i)$

➤ If we neglect the time needed for loop control and assume that $P(i)$ does not actually depend on i and each time it is performed at a cost of t .

Analysis of Algorithms

➤ Analyzing Control Structure

➤ For Loop

for $i \leftarrow 1$ to m do $P(i)$

➤ If we do not neglect the time needed for loop control and assume that $P(i)$ does not actually depend on i .

**$i \leftarrow 1$
while $i \leq m$ do
 $P(i)$
 $i \leftarrow i + 1$**

In most situations, it is reasonable to count at unit cost the test $i \leq m$, the instructions $i \leftarrow 1$ and $i \leftarrow i + 1$, and the sequencing operations (**go to**) implicit in the **while** loop. Let c be an upper bound on the time required by each of these operations. The time ℓ taken by the loop is thus bounded above by

$$\begin{aligned}
 \ell &\leq c && \text{for } i \leftarrow 1 \\
 &+ (m+1)c && \text{for the tests } i \leq m \\
 &+ mt && \text{for the executions of } P(i) \\
 &+ mc && \text{for the executions of } i \leftarrow i + 1 \\
 &+ mc && \text{for the sequencing operations} \\
 &\leq (t+3c)m + 2c.
 \end{aligned}$$

Analysis of Algorithms

➤ Analyzing Control Structure

➤ For Loop

for $i \leftarrow 1$ to m do $P(i)$

➤ If we neglect the time needed for loop control and assume that $P(i)$ depends on i and cost of executing $P(i)$ each time is $t(i)$.

$$\sum_{i=1}^m t(i).$$

Analysis of Algorithms

➤ Analyzing Control Structure

➤ For Loop

➤ Iterative Fibonacci

```
function Fibiter(n)  
    i ← 1; j ← 0  
    for k ← 1 to n do j ← i + j  
                        i ← j - i  
    return j
```

0, 1, 1, 2, 3, 5, 8,
13, 21, 34, 55, 89,
144, 233, 377, 610,
987, 1597,

➤ If we count all arithmetic operations at unit cost and not taking loop control into account

➤ If we does not count additions involved in the computation of Fibonacci sequence at unit cost

$$\begin{cases} f_0 = 0; f_1 = 1 & \text{and} \\ f_n = f_{n-1} + f_{n-2} & \text{for } n \geq 2. \end{cases}$$

$$f_n = \frac{1}{\sqrt{5}}[\phi^n - (-\phi)^{-n}],$$

where $\phi = (1 + \sqrt{5})/2$ is the golden ratio. Since $0 < \phi^{-1} < 1$, the term $(-\phi)^{-n}$ can be neglected when n is large. Hence the value of f_n is roughly $\phi^n/\sqrt{5}$, which is exponential in n .

Analysis of Algorithms

- Analyzing Control Structure
 - For Loop
 - Iterative Fibonacci

$$\sum_{k=1}^n ck = c \sum_{k=1}^n k = c \frac{n(n+1)}{2} \in O(n^2).$$

Analysis of Algorithms

- Analyzing Control Structure
 - Recursive Calls

```
function Fibrec(n)  
  if n < 2 then return n  
  else return Fibrec(n - 1) + Fibrec(n - 2)
```

$$T(n) = \begin{cases} a & \text{if } n = 0 \text{ or } n = 1 \\ T(n-1) + T(n-2) + h(n) & \text{otherwise} \end{cases}$$

- Addition at unit cost or not at the unit cost.
- Running time

Analysis of Algorithms

➤ Analyzing Control Structure

➤ While Loops

- Why are they harder to analyze than for loops?
- Identifying and using a function of variables whose value decreases each time around
- Example of Binary_Search

```
function Binary_Search( $T[1..n]$ ,  $x$ )  
  {This algorithm assumes that  $x$  appears in  $T$ }  
   $i \leftarrow 1$ ;  $j \leftarrow n$   
  while  $i < j$  do  
    {  $T[i] \leq x \leq T[j]$  }  
     $k \leftarrow (i + j) \div 2$   
    case  $x < T[k]$ :  $j \leftarrow k - 1$   
            $x = T[k]$ :  $i, j \leftarrow k$  {return  $k$ }  
            $x > T[k]$ :  $i \leftarrow k + 1$   
  return  $i$ 
```


Analysis of Algorithms

➤ Analyzing Control Structure

➤ While Loops

➤ In binary search the number of elements of array T under consideration continuously decreases. Let this number be d and $d = j - i + 1$

➤ Let d and \hat{d} be the value of $j - i + 1$ before and after the iteration under consideration.

We use i , j , \hat{i} and \hat{j} similarly. If $x < T[k]$, the instruction

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➤ Analyzing Control Structure

➤ While Loops

If $x < T[k]$, the instruction “ $j \leftarrow k - 1$ ” is executed and thus $\hat{i} = i$ and $\hat{j} = [(i + j) \div 2] - 1$. Therefore,

$$\hat{d} = \hat{j} - \hat{i} + 1 = (i + j) \div 2 - i \leq (i + j)/2 - i < (j - i + 1)/2 = d/2.$$

Similarly, if $x > T[k]$, the instruction “ $i \leftarrow k + 1$ ” is executed and thus

$$\hat{i} = [(i + j) \div 2] + 1 \text{ and } \hat{j} = j.$$

Therefore,

$$\hat{d} = \hat{j} - \hat{i} + 1 = j - (i + j) \div 2 \leq j - (i + j - 1)/2 = (j - i + 1)/2 = d/2.$$

Finally, if $x = T[k]$, then i and j are set to the same value and thus $\hat{d} = 1$; but d was at least 2 since otherwise the loop would not have been reentered. We conclude that $\hat{d} \leq d/2$ whichever case happens, which means that the value of d is at least halved each time round the loop. Since we stop when $d \leq 1$, the process must eventually stop, but how much time does it take?

Analysis of Algorithms

- Analyzing Control Structure
 - While Loops

To determine an upper bound on the running time of binary search, let d_ℓ denote the value of $j - i + 1$ at the end of the ℓ -th trip round the loop for $\ell \geq 1$ and let $d_0 = n$. Since $d_{\ell-1}$ is the value of $j - i + 1$ *before* starting the ℓ -th iteration, we have proved that $d_\ell \leq d_{\ell-1}/2$ for all $\ell \geq 1$. It follows immediately by mathematical induction that $d_\ell \leq n/2^\ell$. But the loop terminates when $d \leq 1$, which happens at the latest when $\ell = \lceil \lg n \rceil$. We conclude that the loop is entered at most $\lceil \lg n \rceil$ times. Since each trip round the loop takes constant time, binary search takes a time in $O(\log n)$.

Analysis of Algorithms

➤ Analyzing Control Structure

➤ While Loops

➤ Alternative Approach : Treating while loops like recursive algorithms

➤ Let $t(d)$ be the maximum time needed to terminate the while loop when there are at most d elements under consideration

➤ Let b be the constant time required to go round the loop once

➤ Let c be the time to determine that the loop is finished when $d=1$ and return, then

$$t(d) \leq \begin{cases} c & \text{if } d = 1 \\ b + t(d - 1) & \text{otherwise} \end{cases}$$

Analysis of Algorithms

- Using a barometer
 - What is a barometer instruction?
 - Advantages
 - What should we do in case of nested for loops?
 - Sorting Example: $T[1..n]$, $U[1..s]$

```
procedure pigeonhole( $T[1..n]$ )  
  {Sorts integers between 1 and 10000}  
  array  $U[1..10000]$   
  for  $k \leftarrow 1$  to 10000 do  $U[k] \leftarrow 0$   
  for  $i \leftarrow 1$  to  $n$  do  
     $k \leftarrow T[i]$   
     $U[k] \leftarrow U[k] + 1$ 
```

```
 $i \leftarrow 0$   
for  $k \leftarrow 1$  to  $s$  do  
  while  $U[k] \neq 0$  do  
     $i \leftarrow i + 1$   
     $T[i] \leftarrow k$   
     $U[k] \leftarrow U[k] - 1$ 
```

Analysis of Algorithms

➤ Using a barometer

➤ What should we do in case of nested for loops?

➤ Sorting Example: $T[1...n]$, $U[1...s]$

➤ If we select any of the instructions in the inner loop as barometer then

➤ Total number of times they are executed is therefore

$$\sum_{k=1}^s U[k]$$

➤ Is it true?

➤ Let a be the time needed for the test $U[k] \neq 0$ and let b be the time taken by one execution of the instructions in the inner loop including implicit sequencing operation.

➤ To execute the inner loop completely for given k

$$t_k = (1 + U[k])a + U[k]b$$

➤ The complete process takes

$c + \sum_{k=1}^s (d + t_k)$, where c and d are new constants to take account of the time needed to initialize and control the outer loop, respectively. When simplified, this expression yields $c + (a + d)s + (a + b)n$. We conclude that the process takes a time in $\Theta(n + s)$

Analysis of Algorithms

- Using a barometer
 - What should we do in case of nested for loops?
 - Sorting Example: $T[1...n]$, $U[1...s]$
 - Shall we use barometer to analyze this pigeon-hole sorting?
 - $U[k] \neq 0$