

# **COMP9020**

Foundations of Computer Science Term 3, 2024

Lecture 18: Algorithmic Analysis

## Outline

Motivation

Standard Approach

Examples

Simplifying with Worst Case and Big-O

Recursive Examples

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## Algorithmic analysis: motivation

Want to compare algorithms – particularly ones that can solve *arbitrarily large* instances.

We would like to be able to talk about the resources (running time, memory, energy consumption) required by a program/algorithm as a function f(n) of some parameter n (e.g. the size) of its input.

#### **Example**

How long does a given sorting algorithm take to run on a list of n elements?

#### ssues

#### **Problems**

- The exact resources required for an algorithm are difficult to pin down. Heavily dependent on:
  - Environment the program is run in (hardware, software, choice of language, external factors, etc)
  - Choice of inputs used

#### Issues

#### **Problems**

- The exact resources required for an algorithm are difficult to pin down. Heavily dependent on:
  - Environment the program is run in (hardware, software, choice of language, external factors, etc)
  - Choice of inputs used
- Cost functions can be complex, e.g.

$$2n\log(n) + (n-100)\log(n)^2 + \frac{1}{2^n}\log(\log(n))$$

Need to identify the "important" aspects of the function.

# Order of growth

## **Example**

Consider two time-cost functions:

- $f_1(n) = \frac{1}{10}n^2$  milliseconds, and
- $f_2(n) = 10n \log n$  milliseconds

Input size	$f_1(n)$	$f_2(n)$
100	0.01s	2s
1000	1s	30s
10000	1m40s	6m40s
100000	2h47m	1h23m
1000000	11d14h	16h40h
10000000	3y3m	8d2h

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## Algorithmic analysis

Asymptotic analysis is about how costs **scale** as the input increases.

Standard (default) approach:

- Consider asymptotic growth of cost functions
- Consider worst-case (highest cost) inputs
- Consider running time cost: number of elementary operations

#### **Take Notice**

Other common analyses include:

- Average-case analysis
- Space (memory) cost

# Elementary operations

Informally: A single computational "step"; something that takes a constant number of computation cycles.

#### Examples:

- Arithmetic operations
- Comparison of two values
- Assignment of a value to a variable
- Accessing an element of an array
- Calling a function
- Returning a value
- Printing a single character

#### **Take Notice**

Count operations up to a constant factor, O(1), rather than an exact number.

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#### **Example**

Squaring a number (First version):

```
square(n): return n * n
```

```
Example
Squaring a number (First version):

square(n):
return \ n*n \ O(1)
```

## Running time vs Execution time

Previous example shows one difference between running time and execution time.

In general, running time only approximates execution time:

- Simplifying assumptions about elementary operations
- Hidden constants in big-O
- Big-O only looks at limiting performance as *n* gets large.

#### **Examples**

- Implementations of square(n) will take longer as n gets bigger
- A program that "solves chess" will run in O(1) time.

## **Example**

```
square(n):

r := 0

for i = 1 to n:

r := r + n

return r
```

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## **Example**

```
\begin{aligned} & \text{square}(n): \\ & r := 0 \\ & \text{for } i = 1 \text{ to } n: \quad \textit{O}(1) \\ & r := r + n \\ & \text{return } r \end{aligned}
```

#### **Example**

```
\begin{array}{l} \operatorname{square}(n): \\ r:=0 \\ \text{for } i=1 \text{ to } n: & \textit{O}(1) \\ r:=r+n & \textit{O}(1) \\ \text{return } r \end{array}
```

#### **Example**

```
\begin{array}{lll} & \operatorname{square}(n): \\ & r:=0 \\ & \operatorname{for}\ i=1\ \operatorname{to}\ n: & O(1) \\ & r:=r+n & O(1) \end{array} \mid n \ \operatorname{times} \\ & \operatorname{return}\ r \end{array}
```

#### **Example**

```
square(n):
r:=0
for i=1 to n: O(1)
r:=r+n
O(1)
n times
O(n)
```

#### **Example**

```
\begin{array}{lll} \operatorname{square}(n): & & & & & & & & & \\ r:=0 & & & & & & & & & & \\ \operatorname{for} i=1 \text{ to } n: & & & & & & & & & \\ r:=r+n & & & & & & & & & & & \\ return r & & & & & & & & & & & \\ \end{array} \begin{array}{c|cccc} O(1) & & & & & & & & & & \\ n \text{ times} & & & & & & & & \\ O(n) & & & & & & & & & \\ \end{array}
```

#### **Example**

Squaring a number (Second version):

```
\begin{array}{lll} \operatorname{square}(n): & & & & & & & & & & \\ r:=0 & & & & & & & & & & O(1) \\ & \operatorname{for}\ i=1\ \operatorname{to}\ n: & & & & & & & O(1) \\ & r:=r+n & & & & & & O(1) \\ & \operatorname{return}\ r & & & & & & O(1) \end{array} \hspace{0.2cm} \begin{array}{c} n\ \operatorname{times} & & & & O(n) \\ & O(1) & & & & & & O(1) \end{array}
```

Running time: O(1) + O(n) + O(1) = O(n)

## **Example**

Cubing a number (using second squaring program):

```
\begin{aligned} & \operatorname{cube}(n) : \\ & r := 0 \\ & \text{for } i = 1 \text{ to } n : \\ & r := r + \operatorname{square}(n) \\ & \text{return } r \end{aligned}
```

#### **Example**

Cubing a number (using second squaring program):

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Cubing a number (using second squaring program):

```
\begin{array}{l} \operatorname{cube}(n): \\ r:=0 \\ \text{for } i=1 \text{ to } n: \\ r:=r+\operatorname{square}(n) \\ \end{array} \begin{array}{c} O(1) \\ O(1)+O(n) \\ \end{array} \begin{array}{c} n \text{ times} \\ O(1) \\ O(1) \end{array}
```

#### **Example**

Cubing a number (using second squaring program):

Running time:  $O(1) + O(n^2) + O(1) = O(n^2)$ 

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Worst-case input assumption and big-O combine to *simplify* the analysis:

## **Example**

Sum of squares (Using second squaring program):

```
sumOfSquares(n):
r := 0
for i = 1 to n:
r := r + square(i)
return r
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

```
Example
Sum of squares (Using second squaring program):
      sumOfSquares(n):
                                                           O(1)
         r := 0
          for i = 1 to n:
                                     O(1)
            r := r + \operatorname{square}(i)
                                                           O(1)
         return r
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

## **Example** Sum of squares (Using second squaring program): sumOfSquares(n): O(1)r := 0for i = 1 to n: O(1)O(?) $r := r + \operatorname{square}(i)$ O(i)O(1)return r

Worst-case input assumption and big-O combine to *simplify* the analysis:

# Example Sum of squares (Using second squaring program): sumOfSquares(n): $r := 0 \qquad O(1)$ $for i = 1 \text{ to } n: \qquad O(1)$ $r := r + square(i) \qquad O(n)$ $return r \qquad O(1)$

Worst-case input assumption and big-O combine to *simplify* the analysis:

```
Example
Sum of squares (Using second squaring program):
                                                               sumOfSquares(n):
                                                                                                                       or i = 1 to n:
color i = 1 \text{ to } n:

                                                                                          r := 0
                                                                                                     for i = 1 to n:
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   O(1)
                                                                                          return r
   Running time: O(1) + O(n^2) + O(1) = O(n^2)
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

#### **Example**

Finding an element (x) in an array (L) of length n:

```
 \begin{aligned} & \operatorname{find}(x,L): \\ & \text{for } i = 0 \text{ to } n-1: \\ & \text{if } L[i] == x: \\ & \text{return } i \end{aligned}
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

```
Example

Finding an element (x) in an array (L) of length n:

find(x, L):
for i = 0 \text{ to } n - 1: \qquad O(1)
if L[i] == x: \qquad O(1)
return i \qquad O(1)
return -1 \qquad O(1)
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

## **Example**

Finding an element (x) in an array (L) of length n:

```
\begin{array}{ll} \operatorname{find}(x,L): \\ & \operatorname{for}\ i=0\ \operatorname{to}\ n-1: & O(1) \\ & \operatorname{if}\ L[i]==x: & O(1) \\ & \operatorname{return}\ i & O(1) \end{array} \begin{array}{c} \operatorname{?\ times} & O(?) \\ & O(1) \end{array}
```

Worst-case input assumption and big-O combine to *simplify* the analysis:

## **Example**

Finding an element (x) in an array (L) of length n:

Worst-case input assumption and big-O combine to *simplify* the analysis:

```
Example

Finding an element (x) in an array (L) of length n:

\begin{array}{c}
\text{find}(x,L):\\ \text{for } i=0 \text{ to } n-1: & O(1)\\ \text{if } L[i]==x: & O(1)\\ \text{return } i & O(1)\\ \end{array}

\begin{array}{c}
\text{return } I & O(n) \\ \text{return } I & O(n)
\end{array}

Running time: O(n)+O(1)=O(n)
```

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Worst-case input assumption and big-O combine to *simplify* the analysis:

#### **Take Notice**

Simplifications might lead to sub-optimal bounds, may have to do a better analysis to get best bounds:

- Finer-grained upper bound analysis
- Analyse specific cases to find a matching lower bound (big- $\Omega$ )

#### **Take Notice**

Big- $\Omega$  is a **lower bound** analysis of the worst-case; NOT a "best-case" analysis.

Analyse specific cases to find a matching lower bound (big- $\Omega$ )

## **Example**

```
Let L_n be an n-element array of 0's.
Finding an element (x) in an array (L) of length n:
```

```
\begin{aligned} & \text{find}(x,L): \\ & \text{for } i = 0 \text{ to } n-1: \\ & \text{if } L[i] == x: \\ & \text{return } i \\ & \text{return } -1 \end{aligned}
```

Analyse specific cases to find a matching lower bound (big- $\Omega$ )

## **Example**

```
Let L_n be an n-element array of 0's.
```

Finding an element (x) in an array (L) of length n:

```
\begin{array}{l} \operatorname{find}(x,L): \\ \operatorname{for}\ i = 0\ \operatorname{to}\ n-1: & \Omega(1) \\ \operatorname{if}\ L[i] == x: & \Omega(1) \\ \operatorname{return}\ i & \Omega(1) \\ \operatorname{return}\ -1 & \Omega(1) \end{array}
```

Analyse specific cases to find a matching lower bound (big- $\Omega$ )

## **Example**

```
Let L_n be an n-element array of 0's.
```

Finding an element (x) in an array (L) of length n:

```
\begin{array}{ll} \operatorname{find}(x,L): \\ & \text{for } i=0 \text{ to } n-1: & \Omega(1) \\ & \text{if } L[i] == x: & \Omega(1) \\ & \text{return } i & \Omega(1) \end{array} \middle| \begin{array}{l} \Omega(n) \text{ times} \\ \Omega(1) \\ \end{array} \\ & \text{return } -1 & \Omega(1) \end{array}
```

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Analyse specific cases to find a matching lower bound (big- $\Omega$ )

#### **Example**

Let  $L_n$  be an n-element array of 0's.

Finding an element (x) in an array (L) of length n:

Analyse specific cases to find a matching lower bound (big- $\Omega$ )

#### **Example**

Let  $L_n$  be an *n*-element array of 0's.

Finding an element (x) in an array (L) of length n:

```
\begin{array}{lll} \operatorname{find}(x,L): & & & & & & & & & & \\ \operatorname{for}\ i=0\ \operatorname{to}\ n-1: & & & & & & & & & & & \\ \operatorname{if}\ L[i]==x: & & & & & & & & & & & & & & \\ \operatorname{return}\ i & & & & & & & & & & & & & & \\ \operatorname{return}\ i & & & & & & & & & & & & & & & \\ \operatorname{return}\ -1 & & & & & & & & & & & & & & & \\ \end{array} \right] \Omega(n)\ \operatorname{times} \qquad \Omega(n)
```

Running time of find(1,  $L_n$ ):  $\Omega(n)$ 

Analyse specific cases to find a matching lower bound (big- $\Omega$ )

#### **Example**

Let  $L_n$  be an *n*-element array of 0's.

Finding an element (x) in an array (L) of length n:

```
\begin{array}{lll} \operatorname{find}(x,L): & & & & & & & & & & & \\ \operatorname{for}\ i=0\ \operatorname{to}\ n-1: & & & & & & & & & & & & \\ \operatorname{if}\ L[i]==x: & & & & & & & & & & & & & & & \\ \operatorname{return}\ i & & & & & & & & & & & & & & & \\ \operatorname{return}\ i & & & & & & & & & & & & & & & & \\ \operatorname{return}\ -1 & & & & & & & & & & & & & & & & \\ \end{array}
```

Running time of find(1,  $L_n$ ):  $\Omega(n)$ 

Therefore, running time of find(x, L):  $\Theta(n)$ 

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## **Example**

## Factorial:

```
fact(n):
if n == 0:
return 1
else:
return n * fact(n-1)
```

## **Example**

## Factorial:

```
\begin{array}{ll} \operatorname{fact}(n): \\ & \text{if } n == 0: \\ & \operatorname{return} \ 1 \\ & \text{else}: \\ & \operatorname{return} \ n * \operatorname{fact}(n-1) \end{array} \begin{array}{ll} O(1) \\ O(1) +? \end{array}
```

## **Example**

Factorial:

```
\begin{array}{ll} \operatorname{fact}(n): \\ & \text{if } n == 0: \\ & \operatorname{return} \ 1 \\ & \text{else}: \\ & \operatorname{return} \ n * \operatorname{fact}(n-1) \end{array} \begin{array}{ll} O(1) \\ O(1) +? \end{array}
```

Running time for fact(n): T(n)

## **Example**

Factorial:

```
\begin{split} & \text{fact}(n): \\ & \text{if } n == 0: \\ & \text{return 1} \\ & \text{else:} \\ & \text{return } n * \text{fact}(n-1) \\ \end{split} \begin{subarray}{c} O(1) \\ O(1) + T(n-1) \\ \end{subarray}
```

Running time for fact(n): T(n)

## **Example**

Factorial:

```
\begin{array}{l} \operatorname{fact}(n): \\ \text{if } n == 0: \\ \text{return 1} \\ \text{else:} \\ \text{return } n * \operatorname{fact}(n-1) \end{array} \qquad \begin{array}{l} O(1) \\ O(1) + T(n-1) \end{array}
```

Running time for fact(n): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(1)$ 

## **Example**

Factorial:

```
\begin{array}{l} \operatorname{fact}(n): \\ \text{if } n == 0: \\ \text{return 1} \\ \text{else:} \\ \text{return } n * \operatorname{fact}(n-1) \end{array} \qquad \begin{array}{l} O(1) \\ O(1) + T(n-1) \end{array}
```

Running time for fact(n): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(1)$   
 $\in O(n)$ 

#### **Example**

Factorial:

```
fact(n):

if n == 0:

return 1

else:

return n * fact(n-1)

O(1) + T(n-1)
```

Running time for fact(n): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(1)$   
 $\in O(n)$ 

Running time:  $T(n) \in O(n)$ 

## **Example**

Summing elements of a linked list (length n):

```
sum(L):
    if L.isEmpty():
        return 0
    else:
        return L.data + sum(L.next)
```

## **Example**

Summing elements of a linked list (length n):

```
\begin{array}{lll} & \text{sum}(\texttt{L}): \\ & \text{if } \texttt{L.isEmpty()}: & \textit{O(1)} \\ & \text{return 0} & \textit{O(1)} \\ & \text{else:} \\ & \text{return L.data} + \text{sum}(\texttt{L.next}) & \textit{O(1)} + \\ \end{array}
```

# **Example** Summing elements of a linked list (length n): sum(L): O(1) O(1)if L.isEmpty(): return 0 else: return L.data + sum(L.next)O(1) +Running time for sum(L): T(n)

## **Example**

Summing elements of a linked list (length n):

```
\begin{array}{lll} & \text{sum}(\texttt{L}): \\ & \text{if } \texttt{L.isEmpty()}: & \textit{O(1)} \\ & \text{return 0} & \textit{O(1)} \\ & \text{else:} \\ & \text{return L.data} + \text{sum}(\texttt{L.next}) & \textit{O(1)} + \textit{T(n-1)} \end{array}
```

Running time for sum(L): T(n)

## **Example**

Summing elements of a linked list (length n):

```
\begin{array}{lll} & \text{sum}(\texttt{L}): \\ & \text{if } \texttt{L.isEmpty}(\texttt{)}: & \textit{O}(\texttt{1}) \\ & \text{return 0} & \textit{O}(\texttt{1}) \\ & \text{else:} \\ & \text{return L.data} + \text{sum}(\texttt{L.next}) & \textit{O}(\texttt{1}) + \textit{T}(\textit{n}-\texttt{1}) \end{array}
```

Running time for sum(L): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(1)$ 

## **Example**

Summing elements of a linked list (length n):

```
\begin{array}{lll} & \text{sum}(\texttt{L}): \\ & \text{if } \texttt{L.isEmpty()}: & \textit{O(1)} \\ & \text{return 0} & \textit{O(1)} \\ & \text{else:} \\ & \text{return L.data} + \text{sum}(\texttt{L.next}) & \textit{O(1)} + \textit{T(n-1)} \end{array}
```

Running time for sum(L): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(1)$   
 $\in O(n)$ 

## **Example**

Insertion sort (L has n elements):

```
sort(L):
    if L.isEmpty():
        return L
    else:
        L2:= sort(L.next)
        insert L.data into L2
        return L2
```

## **Example**

Insertion sort (L has n elements):

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Running time for sort(L): T(n)

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Insertion sort (L has n elements):

Running time for sort(L): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(n) + O(1)$ 

### Example

Insertion sort (L has n elements):

Running time for sort(L): T(n), where:

$$T(0) \in O(1) + O(1) = O(1)$$
  
 $T(n) = T(n-1) + O(n) + O(1)$   
 $\in O(n^2)$ 

## **Example**

Euclidean algorithm for gcd(m, n) (N = m + n):

```
\gcd(m, n):

if m > n:

return \gcd(m - n, n)

else if n > m:

return \gcd(m, n - m)

else:

return m
```

## **Example**

Euclidean algorithm for gcd(m, n) (N = m + n):

```
\begin{split} \gcd(m,n): \\ &\text{if } m>n: \\ &\text{return } \gcd(m-n,n) \\ &\text{else if } n>m: \\ &\text{return } \gcd(m,n-m) \\ &\text{else: } &\text{return } m \end{split} \qquad \begin{array}{c} O(1) \\ O(1) \\ & O(
```

## **Example**

```
Euclidean algorithm for gcd(m, n) (N = m + n):
```

```
\gcd(m,n):

if m>n:

return \gcd(m-n,n)

else if n>m:

return \gcd(m,n-m)

else:

return m

O(1)
```

Running time for gcd(m, n): T(N)

#### Example

```
Euclidean algorithm for gcd(m, n) (N = m + n):
```

```
\begin{array}{lll} \gcd(m,n): & & O(1) \\ & \text{if } m>n: & O(1) \\ & \text{return } \gcd(m-n,n) & \leq T(N-1) \\ & \text{else if } n>m: & O(1) \\ & \text{return } \gcd(m,n-m) & \leq T(N-1) \\ & \text{else}: & \text{return } m & O(1) \end{array}
```

Running time for gcd(m, n): T(N)

#### **Example**

Euclidean algorithm for gcd(m, n) (N = m + n):

```
\gcd(m,n):

if m > n:

return \gcd(m-n,n)

else if n > m:

return \gcd(m,n-m)

else : return m

O(1)

\leq T(N-1)

\leq T(N-1)
```

Running time for gcd(m, n): T(N), where:

$$T(1) \in O(1)$$
  
 $T(N) \leq T(N-1) + O(1)$ 

#### **Example**

Euclidean algorithm for gcd(m, n) (N = m + n):

```
\gcd(m,n):

if m > n:

return \gcd(m-n,n)

else if n > m:

return \gcd(m,n-m)

else:

return \gcd(m,n-m)

else:

O(1)

O(1)

O(1)
```

Running time for gcd(m, n): T(N), where:

$$T(1) \in O(1)$$
 $T(N) \leq T(N-1) + O(1)$ 
 $\in O(N)$ 

### **Example**

Euclidean algorithm for gcd(m, n) (N = m + n):

Running time: O(N)

### **Take Notice**

*N* is not the input size. Input size is  $\log(m) + \log(n)$ 

### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

```
\begin{split} \gcd(m,n): \\ &\text{if } m>n>0: \\ &\text{return } \gcd(m\ \%\ n,n) \\ &\text{else if } n>m>0: \\ &\text{return } \gcd(m,n\ \%\ m) \\ &\text{else : } &\text{return } \max(m,n) \end{split}
```

### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

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Faster Euclidean algorithm for gcd(m, n) (N = m + n):

```
 \gcd(m,n): \\  \text{if } m>n>0: \\  \text{return } \gcd(m\ \%\ n,n) \\  \text{else if } n>m>0: \\  \text{return } \gcd(m,n\ \%\ m) \\  \text{else : } \text{return } \max(m,n)
```

Running time for gcd(m, n): T(N)

#### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

Running time for gcd(m, n): T(N)

#### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

Running time for gcd(m, n): T(N), where:

#### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

Running time for gcd(m, n): T(N), where:

$$T(1) \in O(1)$$
  
 $T(N) \leq T(N/1.5) + O(1)$ 

#### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

Running time for gcd(m, n): T(N), where:

$$T(1) \in O(1)$$

$$T(N) \leq T(N/1.5) + O(1)$$

$$\in O(\log N)$$

### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

What about lower bounds?

#### **Example**

Faster Euclidean algorithm for gcd(m, n) (N = m + n):

What about lower bounds?

- Can show algorithm takes k steps to compute  $gcd(F_k, F_{k-1})$  where  $F_k$  is the k-th Fibonacci number
- Can show  $1.5^k \le F_k \le 2^k$ , so  $k \in \Theta(\log F_k)$
- Therefore  $gcd(F_k, F_{k-1}) \in \Omega(\log(F_k + F_{k-1}))$

### Exercise

#### **Exercise**

RW: 4.3.22 The following algorithm raises a number a to a power n.

$$\exp(a, n)$$
:  
 $p = 1$   
 $i = n$   
while  $i > 0$ :  
 $p = p * a$   
 $i = i - 1$   
return  $p$ 

Determine the running time of this algorithm.

#### Exercise

#### **Exercise**

RW: 4.3.21 The following algorithm gives a fast method for raising a number a to a power n.

fast-exp(
$$a$$
,  $n$ ):
$$p = 1$$

$$q = a$$

$$i = n$$
while  $i > 0$ :
if  $i$  is odd:
$$p = p * q$$

$$q = q * q$$

$$i = \left\lfloor \frac{i}{2} \right\rfloor$$
return  $p$ 

Determine the running time of this algorithm.