# 5. Induction and Recursion

# **5.1 Mathematical Induction**

Consider the sum of the first *n* positive odd numbers:

1=1, 1+3=4, 1+3+5=9, 1+3+5+7=16, 1+3+5+7+9=25

Is it  $n^2$ ?

Induction is a powerful tool to prove assertions of this type.

#### **Mathematical Induction:**

Prove the theorem: P(n) is true  $\forall n \in Z^+$ 

#### **Proof by induction**:

Basis step
 Inductive step

P(1) is shown to be true

 $P(n) \rightarrow P(n+1)$  is shown to be true  $\forall n \in Z^+$ 

#### **Mathematical Induction:**

Prove the theorem: P(n) is true  $\forall n \in Z^+$ 

#### **Proof by induction:**

1. Basis stepP(1) is shown to be true2. Inductive step $P(n) \rightarrow P(n+1)$  is shown to be true  $\forall n \in Z^+$ 

Then if we apply the following rule of inference,

 $[\mathbf{P}(1) \land \forall n \ (\mathbf{P}(n) \to \mathbf{P}(n+1))] \to \forall n \ \mathbf{P}(n)$ 

to conclude that P(n) is true  $\forall n \in Z^+$ 

Basis step:

P(1):  $1 = 1^2$  (True)

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Suppose, for a fixed arbitrary *n*, P(*n*) is T, i.e.,  $1 + 3 + ... + (2n-1) = n^2$ Then show P(*n*+1) is also T.  $1 + 3 + ... + (2n-1) + (2n+1) = (n+1)^2$ ?

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Inductive step:

 $? \mathbf{P}(n) \to \mathbf{P}(n+1) \qquad \forall n \in Z^+$ 

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<u>Remark</u>: Here P(n) is called *inductive hypothesis* for a fixed arbitrary *n*.

*e.g.* Prove  $n < 2^n$   $\forall n \in Z^+$ 

Basis step:

P(1):  $1 < 2^1 = 2$  (T)

# Inductive step:

Show  $P(n) \rightarrow P(n+1) \quad \forall n \in Z^+$ .  $n < 2^n \rightarrow n+1 < 2^{(n+1)}$ ? *e.g.* Prove  $n < 2^n$   $\forall n \in Z^+$ 

Basis step:

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#### Inductive step:

Show  $P(n) \rightarrow P(n+1) \quad \forall n \in Z^+$ .  $n < 2^n \rightarrow n+1 < 2^{(n+1)}$ ?  $n < 2^n \Rightarrow n+1 < 2^n + 1 \le 2^n + 2^n = 2^{n+1}$   $\therefore P(n+1)$  is T  $\therefore n < 2^n \quad \forall n \in Z^+$  e.g. Inequality for Harmonic Numbers:

Harmonic number:  $H_k = 1 + 1/2 + 1/3 + \dots + 1/k, \qquad k = 1, 2, 3...$ 

Show that  $H_{2^n} \ge 1 + n/2$   $\forall n, n \text{ is a nonnegative integer.}$ 

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Basis step:

P(0) is true, since  $H_{2^0} = H_1 = 1 \ge 1 + 0/2 = 1$ .

Inductive step:

Assume P(n) is true  $\Rightarrow$   $H_2^n \ge 1 + n/2$ 

$$H_2^{n+1} = H_2^n + 1/(2^n + 1) + \dots + 1/(2^{n+1})$$
  
≥  $(1 + n/2) + 1/(2^n + 1) + \dots + 1/(2^{n+1})$   
≥  $(1 + n/2) + 2^n (1/2^{n+1}) = 1 + (n+1) / 2$   
∴ P(n+1) is true.  
Hence by induction  $H_2^n \ge 1 + n/2 \quad \forall n$ 

*e.g.* Prove that  $2^n < n!$  for n = 4, 5, 6, ....

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## Basis step:

P(4) is true, since  $2^4 = 16 < 4! = 24$ 

Inductive step:

Assume P(n) is true: 
$$2^n < n!$$
  

$$\Rightarrow 2^{n+1} < 2n!$$

$$< (n+1) n!$$

$$= (n+1)!$$
 $\therefore$  P(n+1) is true.

Hence by induction  $2^n < n!$  for n = 4, 5, 6, ...

# **5.2 Strong Induction**

- 1. **Basis step:** Show that P(1) is true.
- 2. Inductive step:

Show that  $[P(1) \land P(2) \land ... \land P(n)] \rightarrow P(n+1)$  is true  $\forall n \in Z^+$ .

2. **Inductive step:** Show that  $[P(1) \land P(2) \land ... \land P(n)] \rightarrow P(n+1)$  is true  $\forall n \in Z^+$ .

*e.g.* Show that if n > 1 integer, then *n* is either prime or can be written as a product of primes.

Let P(n) be "*n* is either prime or can be written as the product of primes".

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<u>Basis step</u>: P(2) is true since 2 is prime itself.

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Let P(n) be "*n* is either prime or can be written as the product of primes".

<u>Basis step</u>: P(2) is true since 2 is prime itself.

<u>Inductive step</u>: Assume P(k) is true for all  $2 \le k \le n$ . Show P(n+1) is also true.

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If (n+1) is prime then P(n+1) is already true.

If (n+1) is composite then  $n+1 = a \cdot b$  s.t.  $1 < a \le b < n+1$ .

Since we know that P(a) and P(b) are true by inductive hypothesis, *a* and *b* are either prime or can be written as product of primes.

 $\Rightarrow a \cdot b$  can also be written as product of primes.

 $\Rightarrow$  P(*n*+1) is also true.

 $\therefore$  P(n) is T  $\forall n > 1$  by strong induction

This completes the proof of the Fundamental Theorem of Arithmetic (see previous lectures, Ch. 4).

# **5.3 Recursive Definitions**

A function can often be defined also recursively:

Specify the value of the function at the beginning, e.g., at zero.
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 $\Rightarrow a_{n+1} = 2a_n$   $a_0 = 1, n = 0, 1, 2, \dots$ 

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 $\Rightarrow a_{n+1} = 2a_n$   $a_0 = 1, n = 0, 1, 2, ....$ 

e.g. F(n) = n! can be defined recursively: F(0) = 1F(n+1) = F(n)(n+1) *e.g.* Fibonacci numbers:

$$f_0 = 0, \quad f_1 = 1$$
  
 $f_{n+1} = f_n + f_{n-1} \quad n = 1, 2, \dots$ 

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$$\lim_{n \to \infty} \frac{f_{n+1}}{f_n} = \frac{1 + \sqrt{5}}{2} \approx \text{ Golden Ratio}$$



Golden ratio, found in nature and used in art and architecture

Let 
$$P(n): f_n > \alpha^{n-2}$$
  
Basis step:  $n = 3 \Rightarrow \alpha < 2 = f_3$   $\therefore$  P(3) is true  
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<u>Inductive step</u>: Assume  $f_k > \alpha^{k-2}$   $\forall k, 3 \le k \le n$  where  $n \ge 4$  $f_{n+1} > \alpha^{n-1}$ ?

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 $f_{n+1} > \alpha^{n-1}$  ?

Since  $\alpha$  is a solution of  $x^2 - x - 1 = 0$   $\therefore \alpha^2 = \alpha + 1$  $\Rightarrow \alpha^{n-1} = \alpha^2 \cdot \alpha^{n-3} = (\alpha + 1) \alpha^{n-3} = \alpha^{n-2} + \alpha^{n-3}$ 

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By inductive hypothesis,  $f_n > \alpha^{n-2}$  and  $f_{n-1} > \alpha^{n-3}$ 

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By inductive hypothesis,  $f_n > \alpha^{n-2}$  and  $f_{n-1} > \alpha^{n-3}$  $\Rightarrow f_{n+1} = f_n + f_{n-1} > \alpha^{n-2} + \alpha^{n-3} = \alpha^{n-1}$   $\therefore$   $f_n > \alpha^{n-2} \quad \forall n \ge 3$  by strong induction. *e.g.* Show that  $f_n < (5/3)^n \forall n \ge 0$ . Exercise (use strong induction).

<u>Theorem</u>: Lamé's Theorem Let  $a, b \in Z^+$  s.t  $a \le b$ . The number *n* of division steps used by **Euclidean algorithm** to find  $gcd(a,b) \le 5$  times the number of decimal digits in *a*, that is,  $n \le 5(\lfloor \log_{10} a \rfloor + 1)$ .

Hence the complexity of Euclidean algorithm is  $O(\log a)$ .

<u>Recall the code:</u> (Euclidean Algorithm,  $a \le b$ )

```
int gcd(int a, int b)
{
    int x,y,r;
    x = b;
    y = a;
    while (y != 0) {
        r = x % y;
        x = y;
        y = r;
    }
    return x;
}
```

Example:	
$155 = 125 \cdot 1 + 30$ $125 = 30 \cdot 4 + 5$ $30 = 5 \cdot 6$	
$\therefore$ gcd (155, 125) = 5	

<u>Proof</u> : Let $b = r_0$ and $a = r_1$	
$r_0 = r_1 q_1 + r_2$	$0 \le r_2 < r_1$
$r_1 = r_2 q_2 + r_3$	$0 \le r_3 < r_2$
$r_{n-2} = r_{n-1}q_{n-1} + r_n$	$0 \le r_n < r_{n-1}$
$r_{n-1}=r_nq_n$	

*n* division steps to find  $r_n$ , and  $q_i \ge 1$  and  $q_n \ge 2$ ,

Example:	
$155 = 125 \cdot 1 + 30$ $125 = 30 \cdot 4 + 5$ $30 = 5 \cdot 6$ $\therefore \text{ gcd } (155, 125) = 5$	

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*n* division steps to find  $r_n$ , and  $q_i \ge 1$  and  $q_n \ge 2$ ,  $r_n \ge 1 = f_2$  ( $f_n: n$ th Fibonacci number)  $r_{n-1} \ge 2r_n \ge 2f_2 = f_3$  Example:  $155 = 125 \cdot 1 + 30$   $125 = 30 \cdot 4 + 5$   $30 = 5 \cdot 6$  $\therefore$  gcd (155, 125) = 5 

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*n* division steps to find  $r_n$ , and  $q_i \ge 1$  and  $q_n \ge 2$ ,  $r_n \ge 1 = f_2$   $r_{n-1} \ge 2r_n \ge 2f_2 = f_3$  $r_{n-2} \ge r_{n-1} + r_n \ge f_3 + f_2 = f_4$  Example:  $155 = 125 \cdot 1 + 30$   $125 = 30 \cdot 4 + 5$   $30 = 5 \cdot 6$  $\therefore$  gcd (155, 125) = 5 

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*n* division steps to find  $r_n$ , and  $q_i \ge 1$  and  $q_n \ge 2$ ,  $r_n \ge 1 = f_2$   $r_{n-1} \ge 2r_n \ge 2f_2 = f_3$   $r_{n-2} \ge r_{n-1} + r_n \ge f_3 + f_2 = f_4$   $r_2 \ge r_3 + r_4 \ge f_{n-1} + f_{n-2} = f_n$  $a = r_1 \ge r_2 + r_3 \ge f_n + f_{n-1} = f_{n+1}$   $\therefore a \ge f_{n+1}$  Example:  $155 = 125 \cdot 1 + 30$   $125 = 30 \cdot 4 + 5$   $30 = 5 \cdot 6$   $\therefore \text{ gcd } (155, 125) = 5$ 

We also know that  $f_{n+1} \ge \alpha^{n-1}$  for n > 2, where  $\alpha = (1 + \sqrt{5}) / 2$ .

 $\Rightarrow a \ge \alpha^{n-1}$ 

Recall that *n* is the number of division steps required by the Euclidean algorithm.

We also know that  $f_{n+1} \ge \alpha^{n-1}$  for n > 2, where  $\alpha = (1 + \sqrt{5}) / 2$ .

$$\Rightarrow a \ge \alpha^{n-1}$$
  
$$\Rightarrow \log_{10} a \ge (n-1)\log_{10} \alpha > (n-1) / 5$$
  
$$\therefore n-1 < 5\log_{10} a \Rightarrow n < 5\log_{10} a + 1$$

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$$\therefore n-1 < 5\log_{10}a \Rightarrow n < 5\log_{10}a + 1$$

$$\Rightarrow n < 5(\lfloor \log_{10}a \rfloor + 1) + 1 \quad \text{since} \lfloor \log_{10}a \rfloor + 1 > \log_{10}a$$

$$\Rightarrow n \le 5(\lfloor \log_{10}a \rfloor + 1)$$

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Note also that (# of decimal digits in *a*) =  $\lfloor \log_{10}a \rfloor + 1$  $\therefore n \le 5 \cdot (\# \text{ of decimal digits in } a)$  (which is *Lamé's Theorem*)

# **5.4 Recursive Algorithms**

#### Definition:

An algorithm is called *recursive* if it solves a problem by reducing it to an instance of the same problem with smaller input.

$O \rightarrow OOD(1) < 1$	
<i>e.g.</i> Compute GCD( $a,b$ ), $a \le b$	Example:
<pre>int GCD(int a, int b){     int gcdAB;     if (a == 0) gcdAB = b;     else gcdAB = GCD(b % a, a);     return gcdAB; }</pre>	$155 = 125 \cdot 1 + 30$ $125 = 30 \cdot 4 + 5$ $30 = 5 \cdot 6$ $\therefore \text{ gcd } (125, 155) = \text{gcd } (30, 125)$ = gcd  (5, 30) = gcd  (0, 5) = 5

```
e.g. Compute a<sup>n</sup>, a∈R and n∈N
a<sup>0</sup> = 1, a<sup>n+1</sup> = aa<sup>n</sup>
int power(int a, int n) {
    int p;
    if (n == 0) p = 1;
    else p = a*power(a, n-1);
    return p;
}
```

# e.g. Product of two integers

```
int product(int m, int n) {
    int prod;
    if (n==1) prod = m;
    else prod = m + product(m, n-1);
    return prod;
}
```

```
e.g. Linear search x in \{a_0, a_1, ..., a_{n-1}\}
```

```
int search(int x, int a[], int n, int i){
    int location;
    if (a[i] == x)
        location = i;
    else if (i == n-1)
        location = -1;
    else
        location = search(x,a,n,i+1);
    return location;
}
```

```
Initially i = 0.
```

e.g. Binary search pseudocode

```
function binary search(x,a,i,j)

m = \lfloor (i + j)/2 \rfloor;

if (x == a[m]) loc = m;

else if (x < a[m] and i < m)

loc = binary search(x,a, i, m-1); (search in the first half)

else if (x > a[m] and j> m)

loc = binary search(x,a,m+1, j); (search in the second half)

else

loc = -1;

return loc;
```

Initially i = 0 and j = n - 1.

# **Recursion vs. Iteration:**

e.g. Computing factorial

## **Recursive:**

function factorial(n: positive integer)
if (n == 1)fact = 1;
else fact = n\*factorial (n-1);
return fact;

#### **Iterative:**

function factorial(n: positive integer)
fact = 1;
for (i = 1; i ≤ n; i++)
 fact = fact \* i;
return fact;

ALL recursive algorithms have an iterative equivalent, and vice versa.

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return fact;

Let T(n) be the time complexity of the recursive solution (in terms of comparison and arithmetic operations).

T(n) = T(n-1) + 3 for  $n \ge 2$  with T(1) = 1.

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Then T(n) = T(n-2) + 3 + 3
= T(n-3) + 3 + 3 + 3
.
.
= T(1) + 3 + 3 + ... + 3
= 3(n-1)+1 = 3n-2
Hence T(n) is O(n).
```

# **Recursion vs. Iteration:**

e.g. Computing factorial

#### **Recursive:**

function factorial(n: positive integer) if (n == 1)fact = 1; else fact = n\*factorial (n-1); return fact;

# **Iterative:**

function factorial(n: positive integer)
fact = 1;
for (i = 1; i ≤ n; i++)
 fact = fact \* i;
return fact;

Let T(n) be the time complexity of the recursive solution (in terms of comparison and arithmetic operations).

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Hence T(n) is O(n).
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Iterative solutions are usually faster than recursive solutions, though in this example both algorithms are of O(n) complexity.

# *e.g.* Computing Fibonacci numbers **Recursive:**

```
function Fibonacci(n: nonnegative integer)
if (n==0) fibonacci = 0;
else if (n==1) Fibonacci = 1;
else fibonacci = Fibonacci(n-1) + Fibonacci(n-2);
return fibonacci;
```

# **Iterative:**

```
function Fibonacci(n: nonnegative integer)
if (n==0) y = 0;
else{
    x=0;
```

```
y=1;
for (i=1; i < n; i++){
    z = x+y;
    x = y;
    y = z;
  }
}
return y;
```

# **Complexity analysis for recursive algorithms:**

Recursive solution for computing Fibonacci numbers seems more clever and compact; however the iterative solution is much more efficient!!!

Let T(n) be the time complexity of the recursive solution (in terms of comparison and arithmetic operations).

Then T(n) = T(n-1) + T(n-2) + 5 for  $n \ge 2$  with T(0) = 1 and T(1) = 2.

(The term 5 counts for 2 comparisons plus 1 addition plus 2 subtractions).

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(Or you can use big-Omega notation: T(n) is  $\Omega(\alpha^n)$ .)

Note that complexity of the iterative solution is  $\Theta(n)$  (big-Theta).

Moral of the story: Don't compute anything more than once!

We can use <u>mathematical induction</u> also to show that a given recursive algorithm produces the correct (i.e., the desired) output. Consider the recursive factorial example:

```
function factorial(n: positive integer)
if (n == 1)fact = 1; Is it correct?
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We have to show that the recursive code works correctly for all positive *n*, i.e., P(n) is true  $\forall n$ , where P(n): "The output of the algorithm is *n*!" <u>Basis step</u>: P(1): The output is 1 = 1! (T)

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```
function Fibonacci(n: nonnegative integer)
if (n==0) fibonacci=0;
else if (n==1) fibonacci=1;
else fibonacci=Fibonacci(n-1)+Fibonacci(n-2);
return fibonacci;
```

```
int GCD(int a, int n) {
    int gcdAB;
    if (a == 0) gcdAB = n;
    else gcdAB = GCD(n % a, a);
    return gcdAB;
}
```

Let P(n): The function GCD(a,n) correctly computes gcd(a, n) when  $a \le n$ ,  $\forall n > 0$ . <u>Basis step</u>: P(1) is true because the only possible cases are a = 0 and a = 1. Looking into the code, we see that the algorithm returns 0 in the first case and returns n = 1 in the latter, as expected.

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Inductive step: Assume P(k) for all  $1 \le k \le n$  and check whether GCD(a, n+1) returns gcd(a, n+1) for any  $a \le n$ .

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<u>Inductive step</u>: Assume P(k) for all  $1 \le k \le n$  and check whether GCD(a, n+1) returns gcd(a, n+1) for any  $a \le n$ . We see that in this case the function calls itself with input (c, a) where c = (n+1)% a, and this call generates gcd((n+1)% a, a) correctly because P(a) is true by inductive hypothesis since  $a \le n$ . Since gcd(a, n+1) = gcd((n+1)% a, a) from what we've learned in number theory, GCD(a, n+1) returns gcd(a, n+1). (Note also that a = n+1 case is trivial to check). So by strong induction we conclude that P(n) is true for all n, that is, GCD(a, n) returns gcd(a, n) whenever  $a \le n$ ,  $\forall n > 0$ .