Recursion vs Induction

CS3330: Algorithms
The University of Iowa
Recursion

- Recursion means defining something, such as a function, in terms of itself
  - For example, let \( f(x) = x! \)
  - We can define \( f(x) \) as
    \[
    f(x) = \text{if } x < 2 \text{ then } 1 \text{ else } x \times f(x-1)
    \]
- Recursion is a powerful problem-solving technique that often produces very clean solutions to even complex problems.
- Recursive solutions can be easier to understand and to describe than iterative solutions.
Recursion example

- Sequences are functions from natural numbers to reals:
  \[ f(i) = a_i \]

  \[ a_0, a_1, a_2, a_3, \ldots, a_n. \]

**Example:** Find \( f(1), f(2), f(3), \) and \( f(4), \) where \( f(0) = 1, \) and

\[ f(n+1) = f(n)^2 + f(n) + 1 \]

\[
\begin{align*}
  f(1) &= f(0)^2 + f(0) + 1 = 1^2 + 1 + 1 = 3 \\
  f(2) &= f(1)^2 + f(1) + 1 = 3^2 + 3 + 1 = 13 \\
  f(3) &= f(2)^2 + f(2) + 1 = 13^2 + 13 + 1 = 183 \\
  f(4) &= f(3)^2 + f(3) + 1 = 183^2 + 183 + 1 = 33673
\end{align*}
\]
Iterative vs. Recursive

- **Iterative**
  
  \[
  \text{factorial}(n) = \begin{cases} 
  1 & \text{if } n = 0 \\
  n \times (n-1) \times (n-2) \times \ldots \times 2 \times 1 & \text{if } n > 0 
  \end{cases}
  \]

- **Recursive**
  
  \[
  \text{factorial}(n) = \begin{cases} 
  1 & \text{if } n = 0 \\
  n \times \text{factorial}(n-1) & \text{if } n > 0 
  \end{cases}
  \]
Iterative Algorithm

```java
factorial(n) {
    i = 1
    factN = 1
    while (i <= n) {
        factN = factN * i
        i = i + 1
    }
    return factN
}
```

- The iterative solution is very straightforward. We simply loop through all the integers between 1 and n and multiply them together.
- In general, iterative solution is computed from small to big.
Recursive Algorithm

factorial(n) {
    if (n = 0)
        return 1
    else
        return n*factorial(n-1)
    end if
}

Note how much simpler the code for the recursive version of the algorithm is as compared with the iterative version.

We have eliminated the loop and implemented the algorithm with one ‘if’ statement.
Recursion Breakdown

1. Base cases:
   - Always have at least one case that can be solved without using recursion.

2. Recursive cases:
   - Any recursive call must make progress toward a base case.
Recursion Breakdown

- To see how the recursion works, let’s break down the factorial function to solve factorial(3).

\[
\text{Factorial}(3) = 3 \times \text{Factorial}(2)
\]

\[
\text{Factorial}(2) = 2 \times \text{Factorial}(1)
\]

\[
\text{Factorial}(1) = 1 \times \text{Factorial}(0)
\]

\[
\text{Factorial}(0) = 1
\]

\[
\text{Factorial}(3) = 3 \times 2 = 6
\]

\[
\text{Factorial}(2) = 2 \times 1 = 2
\]

\[
\text{Factorial}(1) = 1 \times 1 = 1
\]
Recursion Breakdown

- After looking at both iterative and recursive methods, it appears that the recursive method is much longer and more difficult to compute.
- If that’s the case, then why would we ever use recursion?
- It turns out that recursive techniques, although more complicated to solve by hand, are very simple and elegant to implement in a computer.
How Recursion Works

- A recursive solution solves a problem by solving a smaller instance of the same problem.
- It solves this new problem by solving an even smaller instance of the same problem.
- Eventually, the new problem will be so small that its solution will be either obvious or known.
- This solution will lead to the solution of the original problem.
To truly understand how recursion works we need to first explore how any function call works.

When a program calls a subroutine (function) the current function must suspend its processing.

The called function then takes over control of the program.

When the function is finished, it needs to return to the function that called it.

The calling function then ‘wakes up’ and continues processing.
How Recursion Works

- To do this we use a stack.
- Before a function is called, all relevant data is stored in a **stackframe**.
- This stackframe is then pushed onto the system stack.
- After the called function is finished, it simply pops the stackframe off the stack to return to the original state.
- By using a stack, we allow functions call other functions (including themselves) which can call other functions, etc.
Limitations of Recursion

- The main disadvantage of programming recursively is that, while it makes it easier to write simple and elegant programs, it also makes it easier to write inefficient ones.
- When we use recursion to solve problems, we are interested exclusively with correctness, and not at all with efficiency. Consequently, our simple, elegant recursive algorithms may be inherently inefficient.
Limitations of Recursion

- By using recursion, you can often write simple, short implementations of your solution.
- However, just because an algorithm can be implemented in a recursive manner doesn’t mean that it should be implemented in a recursive manner.
Fibonacci sequence

- \( \text{fibonacci}(0) = 0 \)
- \( \text{fibonacci}(1) = 1 \)
- \( \text{fibonacci}(n) = \text{fibonacci}(n-1) + \text{fibonacci}(n-2) \) for \( n > 1 \)

- This definition is a little different than the previous recursive definitions because it has two base cases, not just one; in fact, you can have as many as you like.
- In the recursive case, there are two recursive calls, not just one. There can be as many as you like.
Fibonacci sequence

- Definition of the Fibonacci sequence

  - Non-recursive:

  \[
  F(n) = \frac{(1 + \sqrt{5})^n - (1 - \sqrt{5})^n}{\sqrt{5} \cdot 2^n}
  \]

  - Recursive:

  \[
  F(n) = F(n-1) + F(n-2)
  \]

  or:

  \[
  F(n+1) = F(n) + F(n-1)
  \]

- Must always specify base case(s)!

  - \( F(0) = 0, \ F(1) = 1 \)
  - Note that some will use \( F(1) = 1, \ F(2) = 1 \)
Fibonacci sequence in Java

```java
long Fibonacci(int n) {
    if ( n == 0 ) return 0;
    else if ( n == 1) return 1;
    else return Fibonacci(n-2) + Fibonacci(n-1);
}

long Fibonacci2(int n) {
    return (long) ((Math.pow((1.0+Math.sqrt(5.0)),n)-
                      Math.pow((1.0-Math.sqrt(5.0)),n)) /
                    (Math.sqrt(5) * Math.pow(2,n)));
}
```
Exponential Time Algorithms

- Consider the recursive Fibonacci generator
- How many recursive calls does it make?
  - \( F(1) \): 1
  - \( F(2) \): 1
  - \( F(3) \): 3
  - \( F(4) \): 5
  - \( F(5) \): 9
  - \( F(10) \): 109
  - \( F(20) \): 13,529
  - \( F(30) \): 1,664,079
  - \( F(40) \): 204,668,309
  - \( F(50) \): 25,172,538,049
  - \( F(100) \): \( 708,449,696,358,523,830,149 \approx 7 \times 10^{20} \)
  - At 1 billion recursive calls per second (generous), this would take over 22,000 years
How many additions used by Fibonacci(n)?

```c
long Fibonacci(int n) {
    if ( n == 0 ) return 0;
    else if ( n == 1) return 1;
    else return Fibonacci(n-2) + Fibonacci(n-1);
}
```

let a(n) be the number of additions used by Fibonacci(n).

\[
a(0) = 0 \\
a(1) = 0 \\
a(n) = a(n-2) + a(n-1) + 1.
\]

**Theorem:** \( a(n) \geq \text{Fibonacci}(n) \) for \( n \geq 2 \).
There is very little overhead in calling this function, as it has only one word of local memory, for the parameter n. However, when we try to compute factorial(20), there will end up being 21 words of memory allocated - one for each invocation of the function. The input takes $O(\log n)$ space, but the code takes $O(n)$, exponential in terms of log $n$. 
Limitations of Recursion

- Multiple recursive calls may involve extensive overhead because they use calls.
- When a call is made, it takes time to build a stackframe and push it onto the system stack.
- Conversely, when a return is executed, the stackframe must be popped from the stack and the local variables reset to their previous values – this also takes time.
The overheads that may be associated with a function call are:

- **Space:** Every invocation of a function call may require space for parameters and local variables, and for an indication of where to return when the function is finished. Typically this space (allocation record) is allocated on the stack and is released automatically when the function returns. Thus, a recursive algorithm may need space proportional to the number of nested calls to the same function.

- **Time:** The operations involved in calling a function - allocating, and later releasing, local memory, copying values into the local memory for the parameters, branching to/returning from the function - all contribute to the time overhead.
Recursion is based upon calling the same function over and over, whereas iteration simply `jumps back' to the beginning of the loop. A function call is often more expensive than a jump.

In general, there is no reason to incur the overhead of recursion when its use does not gain anything.

Recursion is truly valuable when a problem has no simple iterative solution.
Hanoi Tower - *Instructions*

1. Transfer all the disks from pole A to pole B.

2. You may move only ONE disk at a time.

3. A large disk may not rest on top of a smaller one at anytime.
Try this one!

Shortest number of moves??
And this one

Shortest number of moves??
Now try this one!

Shortest number of moves??
How to solve Tower of Hanoi of n disks?

- If \( n = 1 \), “move disk 1 from A to B”, done.
- If \( n > 1 \),
  1. Solve the Tower of Hanoi of \( n-1 \) disks, from A to C;
  2. “move disk \( n \) from A to B”
  3. Solve the Tower of Hanoi of \( n-1 \) disks, from C to B.

```c
static Hanoi ( int n, char A, char B, char C ) {
    if (n==1) printf(“move disk 1 from ” + A + “ to “ + B);
    else {
        Hanoi(n-1, A, C, B);
        printf(“move disk “ + n + “ from “ + A + “ to “ + B);
        Hanoi(n-1, C, B, A);
    }
}
```

Counting the moves:
Let \( f(n) \) be the number of moves for \( n \) disks.

\[
f(1) = 1; \quad f(n) = 2f(n-1) + 1.\]
Let $f(n)$ be the least number of moves for $n$ disks.

$$
\begin{align*}
    f(1) &= 1; \\
    f(n) &= 2f(n-1) + 1.
\end{align*}
$$

<table>
<thead>
<tr>
<th>Number of Disks</th>
<th>Number of Moves</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>$f(1) = 1$</td>
</tr>
<tr>
<td>2</td>
<td>$f(2) = 2*1 + 1 = 3$</td>
</tr>
<tr>
<td>3</td>
<td>$f(3) = 2*3 + 1 = 7$</td>
</tr>
<tr>
<td>4</td>
<td>$f(4) = 2*7 + 1 = 15$</td>
</tr>
<tr>
<td>5</td>
<td>$f(5) = 2*15 + 1 = 31$</td>
</tr>
<tr>
<td>6</td>
<td>$f(6) = 2*31 + 1 = 63$</td>
</tr>
</tbody>
</table>
Let $f(n)$ be the number of moves for $n$ disks.

\[ f(1) = 1; \]
\[ f(n) = 2f(n-1) + 1. \]

**Prove:** $f(n) = 2^n - 1$.

**By induction.**

- **Base step:** $n = 1$.
  - **Left:** $f(1) = 1$;
  - **Right:** $2^1 - 1 = 1$
- **Induction hypothesis:** $f(n-1) = 2^{n-1} - 1$.
- **Inductive step:**
  - **Left:** $f(n) = 2f(n-1) + 1 = 2(2^{n-1} - 1) + 1 = 2^n - 1$.
  - **Right:** $2^n - 1$. 
Fascinating fact

So the formula for finding the number of steps it takes to transfer \( n \) disks from post A to post C is:

\[
2^n - 1
\]

- If \( n = 64 \), the number of moves of single disks is \( 2^ {64} - 1 \), or \( 18,446,744,073,709,551,615 \) moves! If one worked day and night, making one move every second it would take slightly more than 580 billion years to accomplish the job! - far, far longer than some scientists estimate the solar system will last.
Main Benefits of Recursive Algorithms

- Invariably recursive functions are clearer, simpler, shorter, and easier to understand than their non-recursive counterparts.
- The program directly reflects the abstract solution strategy (algorithm).
- From a practical software engineering point of view these are important benefits, greatly enhancing the cost of maintaining the software.
Consider the following program for computing the Fibonacci function.

```c
int s1, s2;
int fib (int n) {
    if (n == 0) return 0;
    else if (n == 1) return 1;
    else {
        s1 = fib(n-1);
        s2 = fib(n-2);
        return s1 + s2;
    }
}
```

fib(2) = ?  
fib(3) = ?  
fib(4) = ?
The main thing to note here is that the variables that will hold the intermediate results, S1 and S2, have been declared as global variables.

This is a mistake. Although the function looks just fine, its correctness crucially depends on having local variables for storing all the intermediate results. As shown, it will not correctly compute the Fibonacci function for \( n=4 \) or larger. However, if we move the declaration of \( s1 \) and \( s2 \) inside the function, it works perfectly.

This sort of bug is very hard to find, and bugs like this are almost certain to arise whenever you use global variables to store intermediate results of a recursive function.
Most recursive algorithms can be translated, by a fairly mechanical procedure, into iterative algorithms. Sometimes this is very straightforward - for example, most compilers detect a special form of recursion, called tail recursion, and automatically translate into iteration without your knowing. Sometimes, the translation is more involved: for example, it might require introducing an explicit stack with which to "fake" the effect of recursive calls.
What is Tail Recursion?

- Recursive methods are either
  - Tail recursive
  - Nontail recursive

- Tail recursive method has the recursive call as the last operation in the method.

- Recursive methods that are not tail recursive are called non-tail recursive
Is Factorial Tail Recursive?

- Is the factorial method a tail recursive method?

```java
int fact(int x) {
    if (x == 0)
        return 1;
    else
        return x * fact(x - 1);
}
```

- When returning back from a recursive call, there is still one pending operation, multiplication.
- Therefore, factorial is a non-tail recursive method.
Another Example

- Is this method tail recursive?

```java
void tail(int i) {
    if (i>0) {
        system.out.print(i+"")
        tail(i-1)
    }
}
```

It is tail recursive!
Third Example

- Is the following program tail recursive?

```java
void prog(int i) {
    if (i>0) {
        prog(i-1);
        System.out.print(i+" ");
    }
}
```

- No, because there is an earlier recursive call, other than the last one.

- In tail recursion, the recursive call should be the last operation, and there should be no earlier recursive calls whether direct or indirect.

- The last recursive call can still be treated as “tail recursion”.
Advantage of Tail Recursive Method

- Tail Recursive methods are easy to convert to iterative.

```java
void tail(int i){
    if (i>0) {
        System.out.println(i+" ");
        tail(i-1)
    }
}
```

- Smart compilers can detect tail recursion and convert it to iterative to optimize code

- Used to implement for loops in languages that do not support loop structures explicitly (e.g. prolog)
Converting Non-tail to Tail Recursive

A non-tail recursive method can be converted to a tail-recursive method by means of an "auxiliary" parameter used to form the result.

The technique is usually used in conjunction with an "auxiliary" function. This is simply to keep the syntax clean and to hide the fact that auxiliary parameters are needed.

```c
int fact_aux(int n, int result) {
    if (n == 0) return result;
    return fact_aux(n - 1, n * result);
}

int fact(n) {
    return fact_aux(n, 1);
}
```
A tail-recursive Fibonacci method can be implemented by using two auxiliary parameters for accumulating results.

```c
int fib_aux ( int i , int nextResult , int result )
{
  if ( i == 0 )
    return result;
  return fib_aux( i - 1 , nextResult + result , nextResult);
}
```

To calculate fib(n), call fib_aux(n,1,0)
Converting Tail Recursive to Iterative

F(x) {
    if (P(x)) return G(x);
    S(x);
    return F(H(x));
}

- If P(x) is true, the value of F(x) is the value of some other function G(x). Otherwise, the value of F(x) is the value of the function F on some other value, H(x)

F(x) {
    while (!P(x)) {
        S(x);
        x = H(x);
    }
    return G(x);
}
Converting Tail Recursive to Iterative

the function F is fact_aux

\( x \leftrightarrow (n, \text{result}), \) tuple of the two parameters

\( P(n, \text{result}) \leftrightarrow (n == 1) \)

\( G(n, \text{result}) \leftrightarrow \text{result} \)

\( H(n, \text{result}) \leftrightarrow (n - 1, n \times \text{result}) \)

\( S(n, \text{result}) \leftrightarrow \text{nothing} \)

```c
int fact_aux(int n, int result) {
    if (n == 1) return result;
    return fact_aux(n - 1, n * result);
}
```

```c
F(x) {
    if (P(x)) return G(x);
    S(x);
    return F(H(x));
}
```

```c
int fact_iter(int n, int result) {
    while (n != 1) {
        (n, result) = (n - 1, n * result);
    }
    return result;
}
```

```c
F(x) {
    while (!P(x)) {
        S(x);
        x = H(x);
    }
    return G(x);
}
```
Converting Tail Recursive to Iterative

The function $F$ is $fib\_aux$

$x \leftrightarrow (n, nRes, res)$, tuple of the three parameters

$P(n, nRes, res) \leftrightarrow (n == 0)$

$G(n, nRes, res) \leftrightarrow res$

$H(n, nRes, res) \leftrightarrow (n - 1, nRes + res, nRes)$

$S(n, result) \leftrightarrow$ nothing

```c
int fib_aux (int n, int nRes, int res){
    if (n == 0)return res;
    return fib_aux(n - 1, nRes + res, nRes);
}
```

```
fib_iter(int n, int nRes, int res) {
    while (n != 0) {
        (n, nRes, res) = (n - 1, nRes + res, nRes);
    }
    return res;
}
```
Recursive Insertion Sort

```c
void isort_rec(int n, int A[]) {
    if (n <= 1) return;
    isort_rec(n - 1, A);
    insert(n-1, A, A[n-1]);
    // insert will insert A[n-1] into sorted array A
}
```

Suppose insert(i, A[], x) takes O(i) time. What is the time complexity of isort_rec(n, A)?

Let f(n) be its time complexity.

\[ f(1) = 1 \]
\[ f(n) = f(n - 1) + cn \]

\[ c \sum_{i=2}^{n} i = \frac{n(n-1)}{2} \]

So \( f(n) = O(n^2) \)
Recursive Selection Sort

```c
void ssort_rec(int n, int A[]) {
    if (n <= 1) return;
    int i = arg_max(n, A);  // find index of max in A
    swap(A, i, n-1);        // swap elements at i and n-1.
    ssort_rec(n-1, A);
}
```

ssort_rec is tail-recursion!!!
Suppose arg_max(i, A[]) takes $O(i)$ time.
What is the time complexity of ssort_rec(n, A)?
Recursion

- Recursion is more than just a programming technique. It has two other uses in computer science and software engineering, namely:
- As a way of describing, defining, or specifying things.
- As a way of designing solutions to problems (divide and conquer).
Defining sets via recursion

- Same as mathematical induction:
  - Base case (or basis step)
  - Recursive step

- Example: the set of positive integers
  - Basis step: $1 \in S$
  - Recursive step: if $x \in S$, then $x+1 \in S$
Defining sets via recursion

- Give recursive definitions for:
  a) The set of odd positive integers
     - $1 \in S$
     - If $x \in S$, then $x+2 \in S$
  b) The set of positive integer powers of 3
     - $3 \in S$
     - If $x \in S$, then $3^x \in S$
  c) The set of polynomials with integer coefficients
     - $0 \in S$
     - If $p(x) \in S$, then $p(x) + cx^n \in S$
       - $c \in \mathbb{Z}$, $n \in \mathbb{Z}$ and $n \geq 0$
Terminology
- $\lambda$ is the empty string: ""
- $\Sigma$ is the set of all letters: \{ a, b, c, ..., z \}
  - The set of letters can change depending on the problem

We can define a set of strings $\Sigma^*$ as follows
- Base step: $\lambda \in \Sigma^*$
- If $w \in \Sigma^*$ and $x \in \Sigma$, then $wx \in \Sigma^*$
- Thus, $\Sigma^*$ is the set of all the possible strings that can be generated with the alphabet
Defining strings via recursion

- Let \( \Sigma = \{ 0, 1 \} \)
- Thus, \( \Sigma^* \) is the set of all binary strings
  - Or all possible computer files
String length via recursion

- How to define string length recursively?
  - Basis step: \( \text{len}(\lambda) = 0 \)
  - Recursive step: \( \text{len}(wx) = \text{len}(w) + 1 \) if \( w \in \Sigma^* \) and \( x \in \Sigma \)

- Example: \( \text{len}(\text{“aaa”}) \)
  - \( \text{len}(\text{“aaa”}) = \text{len}(\text{“aa”}) + 1 \)
  - \( \text{len}(\text{“aa”}) = \text{len}(\text{“a”}) + 1 \)
  - \( \text{len}(\text{“a”}) = \text{len}(\text{“”}) + 1 \)
  - \( \text{len}(\text{“”}) = 0 \)
  - Result: 3
Strings via recursion example

- Given a string $x = a_1a_2...a_n$, $x^R$ stands for its reversal: $x^R = a_na_{n-1}...a_1$. Eg. $x = abc$, $x^R = cba$.
- A string $x$ is a palindrome if $x = x^R$. Eg. $x = aba$.
- Give a recursive definition for the set of string that are palindromes
  - We will define set $P$, which is the set of all palindromes
  - Basis step: $\lambda \in P$
  - Second basis step: $x \in P$ when $x \in \Sigma$
  - Recursive step: $xpx \in P$ if $x \in \Sigma$ and $p \in P$
How many binary strings of length \( n \) that do not contain the pattern 11?

- \( n = 0 \): 1 string (\( \lambda \), the empty string)
- \( n = 1 \): 2 strings (0 and 1)
- \( n = 2 \): 3 strings (00, 01, 10)
- \( n = 3 \): 5 strings (000, 001, 010, 100, 101)
- \( n = 4 \): 8 strings (0000, 0001, 0010, 0100, 1000, 0101, 1001, 1010)

- Any pattern?
- A Fibonacci sequence!
How many binary strings of length $n$ that do not contain the pattern 11?

- $n = 2$: 3 strings (00, 01, 10)
- $n = 3$: 5 strings (000, 001, 010, 100, 101)
- $n = 4$: 8 strings (0000, 0001, 0010, 0100, 1000, 0101, 1001, 1010)

- The strings of $n=4$ can be divided into two classes:
  - $X = \{ 0000, 0001, 0010, 0100, 0101 \}$ and
  - $Y = \{ 1000, 1001, 1010 \}$
- $X$ can be obtained from $n = 3$: adding a leading 0
- $Y$ can be obtained from $n = 2$: adding leading 10.
How many binary strings of length $n$ that do not contain the pattern 11?

- Let $S_n$ be the set of binary strings of length $n$ that do not contain the pattern 11.
- For any string $x$ in $S_{n-1}$, $y = 0x$ is a string of $S_n$.
- For any string $x$ in $S_{n-2}$, $z = 10x$ is a string of $S_n$.
- Any string of $S_n$ is either $y$ or $z$ above.
- Hence $S_n = 0S_{n-1} \cup 10S_{n-2}$, or $|S_n| = |S_{n-1}| + |S_{n-2}|$
- From $|S_0| = 1$, $|S_1| = 2$, we can compute any $|S_n|$. 
Recursive Functions on Natural Numbers

The set of natural numbers
- Basis step: $0 \in \mathbb{N}$
- Recursive step: if $x \in \mathbb{N}$, then $x + 1 \in \mathbb{N}$

Define plus on $\mathbb{N}$:
- Basis step: $\text{plus}(0, y) = y$ for all $y \in \mathbb{N}$
- Recursive step: $\text{plus}(x+1, y) = \text{plus}(x, y) + 1$ for all $x, y \in \mathbb{N}$

Prove that $\text{plus}(x, y) = x + y$.

Induction on $x$:
- Basis step $x = 0$: $\text{plus}(0, y) = 0 + y$ for all $y \in \mathbb{N}$
- Induction hypothesis: $\text{plus}(x, y) = x + y$
- Recursive step $x = u + 1$: $\text{plus}(u + 1, y) = (u + 1) + y$ for all $u, y \in \mathbb{N}$
  - Left = $\text{plus}(u + 1, y) = \text{plus}(u, y) + 1 = u + y + 1$
  - Right = $(u + 1) + y = u + y + 1$

This induction proof is also called “structural induction”.

\(58\)
Recursive Functions on Natural Numbers

- Define plus on N:
  - Basis step: plus(0, y) = y for all y ∈ N
  - Recursive step: plus(x+1, y) = plus(x, y)+1 for all x, y ∈ N

- Define mult on N:
  - Basis step: mult(0, y) = 0 for all y ∈ N
  - Recursive step: mult(x+1, y) = plus(mult(x, y), y) for all x, y ∈ N

- Prove that mult(x, y) = x*y.

- Induction on x:
  - Basis step x=0: mult(0, y) = 0*y for all y ∈ N
  - Induction hypothesis: mult(x, y) = x*y
  - Recursive step x = u+1: mult(u+1, y) = (u+1)*y for all u, y ∈ N
    - Left = mult(u+1, y) = plus(mult(u, y), y) = u*y+y
    - Right = (u+1)*y = u*y+y
Recursion Functions on Strings

- Given a string $x = a_1a_2...a_n$, $x^R$ stands for its reversal: $x^R = a_n a_{n-1}...a_1$. Eg. $x = abc$, $x^R = cba$.

- How to define $x^R$ recursively?
  - Base step: $\lambda \in \Sigma^*$
  - Recursive step: If $w \in \Sigma^*$ and $x \in \Sigma$, then $wx \in \Sigma^*$

- Basis step: $\lambda^R = \lambda$

- Recursive step: $(wx)^R = x(w)^R$ if $x \in \Sigma$ and $w \in \Sigma^*$

- Theorem: $(aw)^R = (w)^Ra$ for all $a \in \Sigma$, $w \in \Sigma^*$.
How to prove \((aw)^R = (w)^Ra\) ?

- **Base step:** \(\lambda \in \Sigma^*\)
- **Recursive step:** If \(w \in \Sigma^*\) and \(x \in \Sigma\), then \(wx \in \Sigma^*\)
  - **Basis step:** \(\lambda^R = \lambda\)
  - **Recursive step:** \((wx)^R = x(w)^R\) if \(x \in \Sigma\) and \(w \in \Sigma^*\)

**Structural Induction Proof:**
- **Basis case** \(w = \lambda\): \((a\lambda)^R = (\lambda)^Ra\) (easy)
- **Induction hypothesis:** \((aw)^R = (w)^Ra\)
- **Inductive case:** \((a(wx))^R = (wx)^R a\) if \(x \in \Sigma\) and \(w \in \Sigma^*\)
  - **Left** = \((a(wx))^R = ((aw)x)^R = x(aw)^R = x(w)^Ra\).
  - **Right** = \((wx)^Ra = (x(w)^R)a = x(w)^Ra\).
How to prove \(((w)^R)^R = w\)?

- **Base step:** \(\lambda \in \Sigma^*\)
- **Recursive step:** If \(w \in \Sigma^*\) and \(x \in \Sigma\), then \(wx \in \Sigma^*\)
- **Basis step:** \(\lambda^R = \lambda\)
- **Recursive step:** \((wx)^R = x(w)^R\) if \(x \in \Sigma\) and \(w \in \Sigma^*\)

**Structural Induction Proof:**

- **Basis case:** \(((\lambda)^R)^R = \lambda\) (easy)
- **Induction hypothesis:** \(((w)^R)^R = w\)
- **Inductive case:** \(((wx)^R)^R = wx\) if \(x \in \Sigma\) and \(w \in \Sigma^*\)
- **Lemma:** \((aw)^R = (w)^R a\)
- **Left:** \(((wx)^R)^R = (x(w)^R)^R = ((w)^R)^R x = wx.\)
- **Right:** \(wx.\)
How to prove \( \text{len}(w^R) = \text{len}(w) \)?

- **Basis step**: \( \lambda^R = \lambda \)
- **Recursive step**: \( (wx)^R = x(w)^R \) if \( x \in \Sigma \) and \( w \in \Sigma^* \)
- **Basis step**: \( \text{len}(\lambda) = 0 \)
- **Recursive step**: \( \text{len}(wx) = \text{len}(w) + 1 \) if \( w \in \Sigma^* \) and \( x \in \Sigma \)

**Structural Induction Proof:**

- **Basis case**: \( \text{len}((\lambda)^R) = \text{len}(\lambda) \) (easy)
- **Induction hypothesis**: \( \text{len}((w)^R) = \text{len}(w) \)
- **Inductive case**: \( \text{len}((wx)^R) = \text{len}(wx) \) if \( x \in \Sigma \) and \( w \in \Sigma^* \)
- **Lemma**: \( \text{len}(aw) = 1 + \text{len}(w) \) if \( a \in \Sigma \) and \( w \in \Sigma^* \)
- **Left**: \( \text{len}((wx)^R) = \text{len}(x(w)^R) = 1 + \text{len}((w)^R) = 1 + \text{len}(w) \).
- **Right**: \( \text{len}(wx) = \text{len}(w) + 1 \).
The append function on strings

- **Basis step:** $\text{len}(\lambda) = 0$
- **Recursive step:** $\text{len}(wx) = \text{len}(w) + 1$ if $w \in \Sigma^*$ and $x \in \Sigma$

- The function $\text{app}(x, y)$ “appends” $x$ and $y$ together.
  - $\text{app}(\text{“abc”}, \text{“xyz”}) = \text{“abcxyz”}$.

- **Basis step:** $\text{app}(v, \lambda) = v$ if $v \in \Sigma^*$
- **Recursive step:** $\text{app}(v, wx) = \text{app}(v, w)x$ if $v, w \in \Sigma^*$ and $x \in \Sigma$.

- How to prove $\text{len}(\text{app}(v, w)) = \text{len}(v) + \text{len}(w)$?
How to prove $\text{len}(\text{app}(v, w)) = \text{len}(v) + \text{len}(w)$?

- **Basis step**: $\text{len}(\lambda) = 0$
- **Recursive step**: $\text{len}(wx) = \text{len}(w) + 1$ if $w \in \Sigma^*$ and $x \in \Sigma$
- **Basis step**: $\text{app}(v, \lambda) = v$ if $v \in \Sigma^*$
- **Recursive step**: $\text{app}(v, wx) = \text{app}(v, w)x$ if $v, w \in \Sigma^*$ and $x \in \Sigma$.

- **Structural Induction Proof**:
  - **Basis case**: $\text{len}(\text{app}(v, \lambda)) = \text{len}(v) + \text{len}(\lambda)$ (easy)
  - **Induction hypothesis**: $\text{len}(\text{app}(v, w)) = \text{len}(v) + \text{len}(w)$
  - **Inductive case**: $\text{len}(\text{app}(v, wx)) = \text{len}(v) + \text{len}(wx)$
  - **Lemma**: $\text{len}(aw) = 1 + \text{len}(w)$ if $a \in \Sigma$ and $w \in \Sigma^*$
  - **Left** = $\text{len}(\text{app}(v, wx)) = \text{len}(\text{app}(v, w)x) = \text{len}(\text{app}(v, w)) + 1 = \text{len}(v) + \text{len}(w) + 1$
  - **Right** = $\text{len}(v) + \text{len}(wx) = \text{len}(v) + \text{len}(w) + 1$. 
How to enumerate all subsets?

- A set of n elements has $2^n$ subsets
- An n-bit integer has $2^n$ values, each value corresponds to one subset.
- Can we use the values of an n-bit integer for printing out each subset?
How to enumerate all subsets?

- A set of n elements has $2^n$ subsets
- A n-bit integer has $2^n$ values, each value corresponds to one subset.

```java
class EnumerateDemo {
    public static void main(String[] args) {
        enumerateSubsets(10);
    }

    public static void enumerateSubsets (int n) {
        // Pre: n < 32
        for (int x = 0; x < (1 << n); x++) {
            System.out.println("{");
            for (int j = 1; j <= n; j++) if (x & (1 << (j-1)) != 0)
                System.out.println(j + ",");
            System.out.println("}\n");
        }
    }
}
```