



RECURSION

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- A **self referential** style of definition useful when it is difficult to directly define objects
- We can use recursion to define
 - Sequences
 - Functions
 - Algorithms
 - Data Structures
- A **recursive** or **inductive definition** requires two components
 - **Basis Step (or Base Case)**: which defines an initial element or defines the simplest form of a problem that can be directly solved
 - **Recursive Step**: provides a rule by which the current element uses a previous one, or a means by which a larger problem is subdivided into the smaller problem
- The functional form of recursion is a form of the **Divide and Conquer** algorithm design strategy

Outline



The lecture is structured as follows:

- Recursively Defined Functions
- Algorithms
 - Search
 - Sorting
 - String Matching
 - Greedy
- Data Recursion



§ Recursively Defined Functions

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Recursively Defined Functions



- Recursively defined functions are **well-defined**
 - for every positive integer, the value of the function at this integer is determined in an unambiguous way.

- Suppose f is defined recursively by:

$$\begin{aligned}f(0) &= 3 \\f(n+1) &= 2f(n) + 3\end{aligned}$$

- Find $f(1)$, $f(2)$, $f(3)$, $f(4)$:

$$\begin{aligned}f(1) &= 2f(0) + 3 = 2 \cdot 3 + 3 = 9 \\f(2) &= 2f(1) + 3 = 2 \cdot 9 + 3 = 21 \\f(3) &= 2f(2) + 3 = 2 \cdot 21 + 3 = 45 \\f(4) &= 2f(3) + 3 = 2 \cdot 45 + 3 = 93\end{aligned}$$

- Give a recursive definition of:

$$\sum_{k=0}^n a_k$$

Basis Case: $\sum_{k=0}^0 a_k = a_0$

Recursive Case: $\sum_{k=0}^{n+1} a_k = \left(\sum_{k=0}^{n+1} a_k \right) + a_{n+1}$

Factorial



- We can define the function $n!$ as: $n! = 1 \times 2 \times \dots \times n$
- However, this is far too imprecise for implementation
- We can define $n!$ recursively

Basis Step: $0! = 1$

Recursive Step: $(n + 1)! = (n + 1) \times n!$

Haskell Implementation:

```
factorial :: Int -> Int
factorial 0 = 1
factorial (n + 1) = (n + 1) * factorial n
```

Recursion Over Lists



- Recursion over lists
 - *Base Case*: [], the empty list
 - *Recursive Case*: the non-empty list i.e., (x:xs)

- General Form:

```
f :: [a] -> type of result
f []      = result of empty list
f (x:xs) = result defined using (f xs) and x
```

- Example: length

```
length :: [a] -> Int
length []      = 0
length (x:xs) = 1 + length xs
```

```
length [1,2,3]
= 1 + length [2,3]
= 1 + (1 + length [3])
= 1 + (1 + (1 + length []))
= 1 + (1 + (1 + 0))
= 3
```

- It is better to think of recursion as a systematic calculation using a high-level equational view rather than via a low-level machine view

Recursion Over Lists



- Another Simple Example: `sum`

```
sum :: Num a => [a] -> a
sum []      = 0
sum (x:xs)  = x + sum xs
```

```
sum [1,2,3]
= 1 + sum [2,3]
= 1 + (2 + sum [3])
= 1 + (2 + (3 + sum []))
= 1 + (2 + (3 + 0))
= 6
```

- Returning a List: `(++)`

```
(++) :: [a] -> [a] -> [a]
[] ++ ys      = ys
(x:xs) ++ ys  = x : (xs ++ ys)
```

```
[1,2,3] ++ [9,8,7,6]
= 1 : ([2,3] ++ [9,8,7,6])
= 1 : (2 : ([3] ++ [9,8,7,6]))
= 1 : (2 : (3 : ([] ++ [9,8,7,6])))
= 1 : (2 : (3 : [9,8,7,6]))
= 1 : (2 : [3,9,8,7,6])
= 1 : [2,3,9,8,7,6]
= [1,2,3,9,8,7,6]
```


- Recursing over 2 lists: zip

```
zip :: [a] -> [b] -> [(a, b)]
zip [] ys          = []
zip xs []          = []
zip (x:xs) (y:ys) = (x, y) : zip xs ys
```

```
zip [1,2,3,4] ['A','*', 'q']
= (1, 'A') : zip [2,3,4] ['*', 'q']
= (1, 'A') : ((2, '*') : zip [3,4] ['q'])
= (1, 'A') : ((2, '*') : ((3, 'q') : zip [4] []))
= (1, 'A') : ((2, '*') : ((3, 'q') : []))
= (1, 'A') : ((2, '*') : [(3, 'q')])
= (1, 'A') : [(2, '*'), (3, 'q')]
= [(1, 'A'), (2, '*'), (3, 'q')]
```

- Recursing a list of lists: concat

```
concat :: [[a]] -> [a]
concat []          = []
concat (xs:xss) = xs ++ concat xss
```

```
concat [[1], [2,3], [4,5,6]]
= [1] ++ concat [[2,3], [4,5,6]]
= [1] ++ ([2,3] ++ concat [[4,5,6]])
= [1] ++ ([2,3] ++ [4,5,6])
= [1] ++ [2,3,4,5,6]
= [1,2,3,4,5,6]
```

Higher-Order Recursive Functions



- Each of the prior recursive functions are quite similar
- It would be elegant if we had a function which express this general computational pattern
- Such a general function would need to be provided both
 - the functions inputs
 - the computation (a function) to perform
- Such functions are called **higher order functions** or a **combinator**
- We have several choices of **combinators**
 - `map` - takes a function and applies it to all items in a list \Rightarrow List
 - `zipWith` - takes a function and applies it to all items in two lists \Rightarrow List
 - `foldr` and `foldl` - takes a function, aggregation variable, and applies to the function to combine the list values into the var \Rightarrow singleton variable

Algorithms

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- There are many general classes of problems that arise in Discrete Mathematics and Computing
- The key to solving such problems is to
 1. Construct a model that translates the problem into a mathematical context
 2. Define a method that will solve the general problem using the model
- The second step is the purview of *algorithm* design
- **Algorithm:** a finite sequence of precise instructions for performing a computation or for solving a problem
 - Typically expressed in English or Pseudocode
- **Pseudocode:** an intermediate step between an English language description of an algorithm and an implementation of the algorithm in a programming language

Pseudocode Example



- Finding the maximum element in a finite sequence

procedure MAX(A)

$max := A_1$

for $i := 2$ **to** n **do**

if $max < A_i$ **then** $max := A_i$

return max

- To gain insight into how an algorithm works it is useful to construct a **trace** that shows the steps for a given specific input.

- Algorithms generally share several properties:
 - **Input:** An algorithm has input values from a specified set
 - **Output:** From each set of input values an algorithm produces output values from a specific set.
 - The *output* values are the solution to the problem
 - **Definiteness:** The steps of an algorithm must be defined precisely
 - **Correctness:** An algorithm should produce the correct output values for each set of input values
 - **Finiteness:** An algorithm should produce the desired output after a finite (but perhaps large) number of steps for any input in the set
 - **Effectiveness:** It must be possible to perform each step of an algorithm exactly and in a finite amount of time
 - **Generality:** The procedure should be applicable for all problems of the desired form, not just for a particular set of input values.

Search Algorithms

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- **Search Problem Definition:** Locate an element x in a list of distinct elements, a_1, a_2, \dots, a_n , or determine that it is not in the list
- The solution to this problem is the location of the term in the list that equals x and 0 if x is not in the list.
- This is one of the most commonly occurring problems in computer science, and occurs in many different contexts

Linear Search



- **Linear Search (sequential search):** searches an ordered list (a_1, a_2, \dots, a_n) for some value x , starting at a_1 and ending at a_n terminating when either the value is found (i.e., $x = a_i$) or the end of the list is reached.

Iterative Linear Search Alg:

procedure LINEARSEARCH(A, x)

$i := 1$

while $i \leq n$ **and** $x \neq A_i$ **do**

$i := i + 1$

if $i \leq n$ **then** $location := i$

else $location := 0$

return $location$



Linear Search



Recursive Linear Search Alg:

- $A \rightarrow$ array/list to search
- $i \rightarrow$ current index
- $j \rightarrow$ size of list
- $x \rightarrow$ value to find

procedure LINSEARCH(A, i, j, x)

if $A_i = x$ **then**

return i

else if $i = j$ **then**

return 0

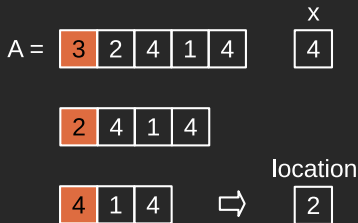
else

return LINSEARCH($A, i + 1, j, x$)

- Requires: $O(n)$ comparisons

Haskell Implementation:

```
linSearch :: Eq a => [a] -> Int -> a -> Int
linSearch [] _ _ = 0
linSearch (y:ys) i x =
  if x == y then i
  else linSearch ys (i + 1) x
```



Binary Search



- Can be used when the list is ordered in either ascending or descending order
- Successively searches smaller and smaller sections, until either the item is found or not
- Requires $O(\log n)$ comparisons

procedure BINSEARCH(A, x)

$i := 1$

$j := n$

while $i < j$ **do**

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

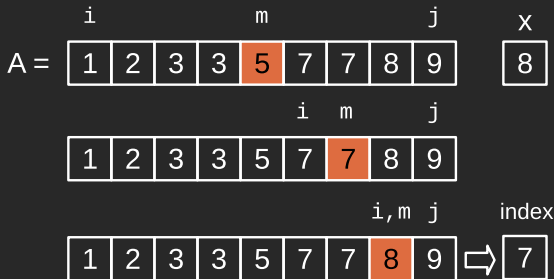
if $x > A_m$ **then** $i := m + 1$

else $j := m$

if $x = A_j$ **then** $location := i$

else $location := 0$

return $location$



Recursive Binary Search Alg:

- $A \rightarrow$ array/list to search
- $i \rightarrow$ current index
- $j \rightarrow$ size of list
- $x \rightarrow$ value to find

procedure BINSEARCH(A, i, j, x)

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

if $x = A_m$ **then return** m

else if $x < A_m$ **and** $i < m$ **then**

return BINSEARCH($A, i, m - 1, x$)

else if $x > A_m$ **and** $j > m$ **then**

return BINSEARCH($A, m + 1, j, x$)

elsereturn 0

- Requires $O(\log n)$ comparisons

Haskell Implementation

```
binSearch :: (Ord a) => [a] -> a -> Int -> Int
binSearch arr x lo hi
  | hi < lo = -1
  | pivot > x = binSearch arr x lo (mid - 1)
  | pivot < x = binSearch arr x (mid + 1) hi
  | otherwise = mid
where
  mid = lo + (hi - lo) `div` 2
  pivot = arr!!mid
```

Sorting Algorithms

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- **Sorting:** the problem of ordering a collection of element (i.e., a list or set)
 - This problem occurs in many contexts, including:
 - Telephone directory
 - Addresses in mailing lists
 - Directory of songs for download
 - Dictionaries
- A significant amount of computing resources is devoted to sorting \Rightarrow a large amount of effort has gone into developing efficient sorting algs
 - 100+ existing sorting algorithms
 - Recently Timsort and Library Sort were developed

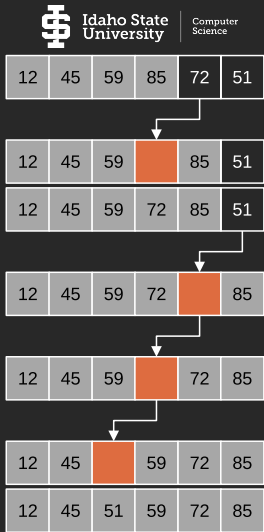
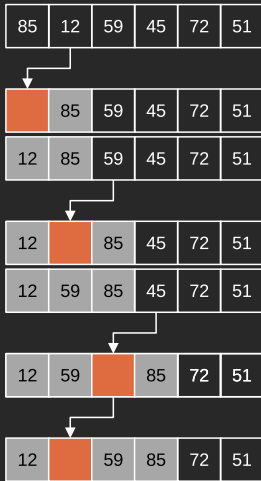
Insertion Sort

```
procedure SORT(A)
  for j := 2 to n do
    i := 1
    while Aj > Ai do
      i := i + 1
    m := Aj
    for k := 0 to j - i - 1 do
      Aj-k := Aj-k-1
    Ai := m
```

Haskell Implementation:

```
insert :: (Ord a) => a -> [a] -> [a]
insert x [] = [x]
insert x (y:ys)
  | x < y    = x:y:ys
  | otherwise = y : (insert x ys)

sort :: (Ord a) => [a] -> [a]
sort []     = []
sort (x:xs) = insert x (sort xs)
```



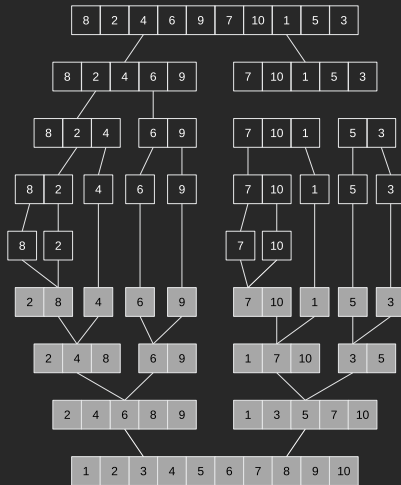
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Merge Sort



- Idea is to recursively split the list in half until each piece is size 1 or less
- Each sublist is then merged to form a sorted combined list
- **Lemma:** Two sorted lists with m and n elements can be merged into a sorted list in no more than $m + n - 1$ comparisons.
- **Theorem:** The number of comparisons needed to merge sort a list with n elements is $O(n \log n)$



Merge Sort



The Algorithm

procedure MSORT(L)

if $n > 1$ **then**

$m := \lfloor n/2 \rfloor$

$L_1 \leftarrow L_1, L_2, \dots, L_m$

$L_2 \leftarrow L_{m+1}, L_{m+2}, \dots, L_n$

$L := \text{MERGE}(\text{MSORT}(L_1), \text{MSORT}(L_2))$

procedure MERGE(L_1, L_2)

$L := []$

while L_1 and L_2 are both nonempty **do**

remove smaller of L_{11} and L_{21} , add to L

if one list is empty **then**

remove all elements of the other list and

append to L

return L

Haskell Implementation

```
merge :: (Ord a) => [a] -> [a] -> [a]
```

```
merge [] [] = []
```

```
merge [] ys = ys
```

```
merge xs [] = xs
```

```
merge allX@(x:xs) allY@(y:ys)
```

```
  | x > y      = y : merge allX ys
```

```
  | otherwise = x : merge xs allY
```

```
sort :: (Ord a) => [a] -> [a]
```

```
sort [] = []
```

```
sort [a] = [a]
```

```
sort [a,b]
```

```
  | a > b      = [b, a]
```

```
  | otherwise = [a, b]
```

```
sort list =
```

```
  let split = splitAt(length list `div` 2) list
```

```
      firstHalf = sort (fst split)
```

```
      secondHalf = sort (snd split)
```

```
  in merge firstHalf secondHalf
```

- A sorting approach based on the idea of *divide and conquer* where we take a list and we attempt to successively cut it in half to make the problem size smaller
- The goal is to gain more than by reducing by one while also ensuring the recursion will complete
- The algorithm in a nutshell works as follows:
 - **Base Case:** empty list \rightarrow empty list
 - **Recursive Case:** non-empty list
 - Select a *pivot* (typically the first or last item in the list)
 - We then select all items from the list $<$ pivot and quick sort those and add them before the pivot
 - We select all items from the list \geq pivot and quick sort them and place them after the pivot

QuickSort



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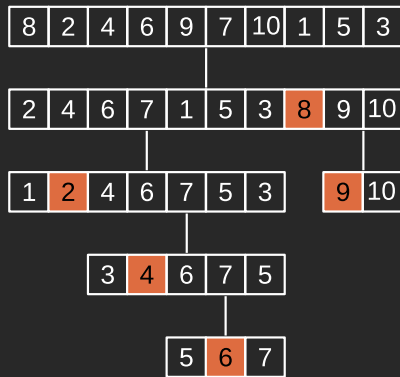
The Algorithm:

```
procedure SORT( $L, lo, hi$ )  
  if  $lo \geq hi$  or  $lo < 0$  then  
    return  
   $p := \text{PARTITION}(L, lo, hi)$   
  SORT( $L, lo, p - 1$ )  
  SORT( $L, p + 1, hi$ )
```

```
procedure PARTITION( $L, lo, hi$ )  
   $pivot := L_{lo}$   
   $i := lo$   
  for  $j := lo$  to  $hi - 1$  do  
    if  $L_j \leq pivot$  then  
       $i := i + 1$   
      swap  $L_i$  with  $L_j$   
  swap  $L_i$  with  $L_{lo}$   
  return  $i$ 
```

Haskell Implementation:

```
quickSort :: Ord a => [a] -> [a]  
quickSort [] = []  
quickSort (pivot:xs) =  
  quickSort [y | y <- xs, y < pivot]  
  ++ [pivot]  
  ++ quicksort [y | y <- xs, y >= pivot]
```



String Matching

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String Matching: finding where a particular string of characters P , called a *pattern*, occurs, if it does, within another string T , called the *text*

- this is another commonly occurring problem with a wide array of applications, including:
 - Text editing
 - Spam filters
 - Detecting network attacks
 - Search engines
 - Plagiarism detection
 - Bioinformatics
 - and many others



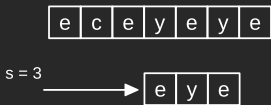
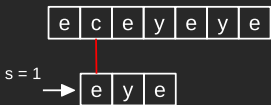
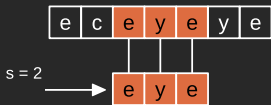
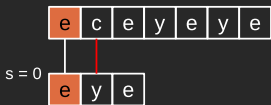
String Matching

Naive String Matcher

```
procedure MATCH( $n, m, T, P$ )  
  for  $s := 0$  to  $n - m$  do  
     $j := 1$   
    while  $j \leq m$  and  $T_{s+j} = P_j$  do  
       $j := j + 1$   
    if  $j > m$  then print "s is a valid shift"
```

Haskell Implementation

```
match :: [Char] -> [Char] -> Int -> [Int]  
match [] _ _ = []  
match _ [] _ = []  
match p allT@(t:ts) i =  
  let l = length p  
      n = take l allT  
  in if (n == p) then i : (match p ts (i + 1))  
      else match p ts (i + 1)
```



§ Greedy Algorithms

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- **Optimization Problems:** Problems where the goal is to find a solution to the given problem that either minimizes or maximizes the value of some parameter. Examples include:
 - Finding a route between two cities with the least total mileage
 - Encoding a message using the fewest bits possible
- **Greedy Algorithms:** Algorithm design strategy, wherein we select the “best” choice at each step rather than attempt to consider all sequences of steps that may lead to the optimal solution
 - Once we know a greed alg finds a feasible solution, then we must prove it is an optimal one
- Greedy algs are often the approach used to solve optimization problems

Greedy Algorithms



- Make n cents change with Quarters, dimes, nickels, and pennies using the least total number of coins.

procedure CHANGE($Coins, n$)

for $i := 1$ **to** r **do**

$D_i := 0$

while $n \leq Coins_i$ **do**

$D_i := D_i + 1$

$n := n - Coins_i$

return D

- The proof of optimality can be found in DMA on page 211

$Coins = [0.25, 0.10, 0.05, 0.01]$

| | | | | | Value |
|-------|---|---|---|---|-------|
| D = | 0 | 0 | 0 | 0 | .86 |
| i = 1 | 1 | 0 | 0 | 0 | .61 |
| i = 1 | 2 | 0 | 0 | 0 | .36 |
| i = 1 | 3 | 0 | 0 | 0 | .11 |
| i = 2 | 3 | 1 | 0 | 0 | .01 |
| i = 3 | 3 | 1 | 0 | 0 | .01 |
| i = 4 | 3 | 1 | 0 | 1 | .00 |
| | 3 | 1 | 0 | 1 | .00 |

Data Recursion

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- This data structure serves as an example of a recursive ADT

```
data Peano = Zero | Succ Peano deriving Show
```

- **Example:**

```
1 = Succ Zero  
2 = Succ (Succ (Succ Zero))
```

- Some operations:

```
decrement :: Peano -> Peano  
decrement zero = Zero  
decrement (Succ a) = a
```

```
add :: Peano -> Peano -> Peano  
add Zero b      = b  
add (Succ a) b = Succ (add a b)
```

```
sub :: Peano -> Peano -> Peano  
sub a Zero      = a  
sub Zero b      = Zero  
sub (Succ a) (Succ b) = sub a b
```

```
lt :: Peano -> Peano -> Bool  
lt a      Zero      = False  
lt Zero   (Succ b) = True  
lt (Succ a) (Succ b) = lt a b
```

- Recursive functions are useful in nearly all programming languages
 - They are especially important for data structures such as Trees and Graphs.
- **Data Recursion:** An important technique that uses recursion to define data structures
 - The idea is to define *circular* data structures
 - **Example:** An infinite list of 1's

```
f :: a -> [a]
f x = x : fx
ones = f 1
```

- Rather than a function, we could simply use a circular list

```
twos = 2 : twos
```





- Data recursion is possible in languages like Haskell due to the use of *lazy evaluation*
- **Lazy Evaluation:** is a technique where expressions are not evaluated until their value is actually needed
- However, most imperative languages (such as C or Java) do not support this and thus we cannot construct infinite data structures in this manner
 - Rather, they would cause an infinite loop
- Yet, we can create circular data structures in other ways

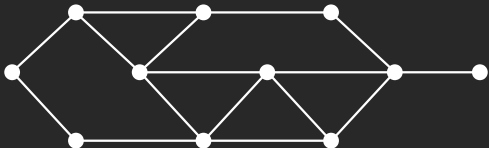
- Recursion can play a role when working with strings
- We can define a string over an alphabet Σ as a finite sequence of symbols from Σ
 - We can then define Σ^* as the set of strings over Σ
 - The recursive definition is:
 - Basis Step:** $\lambda \in \Sigma^*$ (where λ is the empty string)
 - Recursive Step:** if $w \in \Sigma^*$ and $x \in \Sigma^*$, then $wx \in \Sigma^*$
- **Example:** $\Sigma = \{0, 1\}$, $\Sigma^* = \{\lambda, 0, 1, 00, 01, 10, 11, \dots\}$
- Basic string operations can also be defined recursively, for example
 - Concatenation
 - Length

Recursively Defined Trees

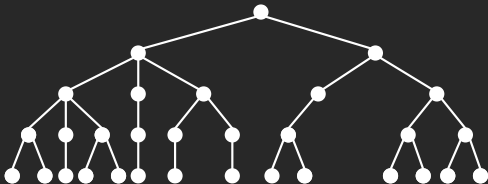


- **Graph:** A data structure composed of vertices and edges connecting pairs of vertices
 - Graphs can be constructed by defining each node with an equation in a `let` expression
 - Thus, each node can be referred to by any other node (including itself)

```
object = let a = 1 : b
          b = 2 : c
          c = [3] ++ a
        in a
```



- **Tree:** A special type of graph



Rooted Trees



- **Rooted Tree:** a tree consisting of vertices containing a distinguished vertex called the *root* and edges connecting these vertices.
 - We can define such a structure recursively

Basis Step: A single vertex r is a rooted tree

Recursive Step: Suppose that T_1, T_2, \dots, T_n are disjoint rooted trees with roots r_1, r_2, \dots, r_n . Then the graph formed by starting with a root r not in any T_i and adding an edge from r to each of the vertices r_1, \dots, r_n , is also a rooted tree.

Step 1



Step 2



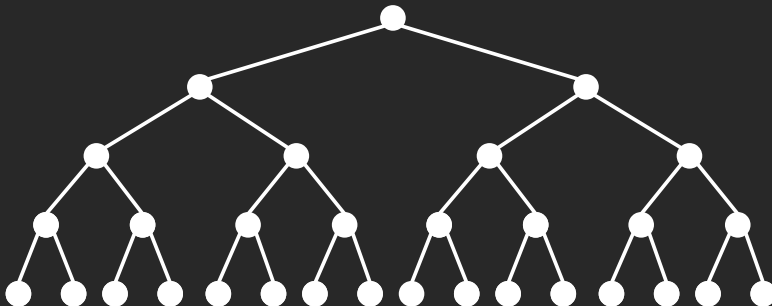
Step 3



Binary Trees



- **Binary Tree:** A rooted tree in which a vertex may have only two children, each of which is a subtree
 - **Full Binary Tree:** if a vertex has children, it must have both a left and right child
 - **Extended Binary Tree:** either the left or right subtree may be empty



Extended Binary Trees



- The set of *extended binary trees* is defined as:

Basis Step: the empty set is an extended binary tree

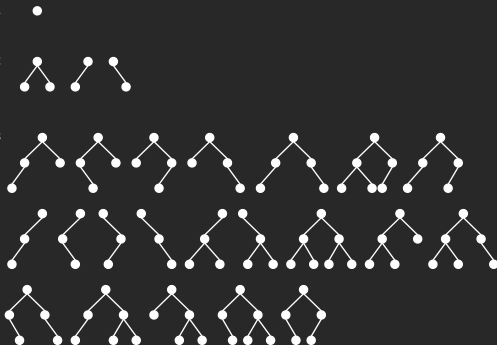
Recursive Step: If T_1 and T_2 are disjoint extended binary trees, there is an extended binary tree denoted $T_1 \cdot T_2$, consisting of a root r together with edges connecting r to the roots of T_1 (left) and T_2 (right) when T_1 and T_2 are nonempty.

Basis Step

Step 1

Step 2

Step 3



Full Binary Trees



- Recursively defined as:

Basis Step: There is a full binary tree consisting of only a single vertex r

Recursive Step: If T_1 and T_2 are disjoint fully binary trees, there is a full binary tree denoted $T_1 \cdot T_2$, consisting of a root r together with edges connecting r to the roots of T_1 and T_2

Basis Step



Step 1



Step 2





- One approach for defining sets is to simply enumerate all of its elements.
 - Unfortunately, this is impractical for all but the smallest sets
 - For larger sets, we could simply use an ellipsis “...” to indicate the definition continues.
 - However, this is an informal approach which is both imprecise and ambiguous
- What we need is an approach that can define these types of sets which is concise, precise, and unambiguous

The Idea Behind Induction



- Induction is sort of a form of mathematical programming which produces a proof when needed
 - i.e., we can assert that something is a member of a set defined by induction

- **Example:**

$$\begin{aligned} 0 &\in S \\ n \in S &\rightarrow n + 1 \in S \end{aligned}$$

- By *modus ponens* and the first assertion we can deduce $1 \in S$, by similar reasoning we can also deduce $2 \in S$
- Furthermore, we can continually build up this chain for **any** natural number

The Idea Behind Induction



- Such inductive definitions can show that a set contains a value v , but requires us to enumerate the values prior to v
- **Sequence:** a set with an ordering
 - The inductively enumerated values form a *sequence*
- Computationally we can use this idea to generate sets

```
imp1 :: Integer -> Integer
imp1 1 = 2
imp1 x = error "premise does not match"
```

```
imp2 :: integer -> Integer
imp2 2 = 3
imp x = error "premise does not match"
```

```
s :: [Integer]
s = [1, imp1 (s!!0), imp2 (s!!1)]
```

The Induction Rule



- Recall,

$$\begin{array}{ll} 0 \in S & \{\text{base case}\} \\ n \in S & \rightarrow \{\text{induction case}\} \end{array}$$

- The *induction case* generate the links of the chain which define the set, starting from the base case
 - By simply modifying our induction rule, we can create completely different sets

$$n \in S \rightarrow n + 2 \in S$$

This rule generates the set of even natural number, however if we change the base case to be $1 \in S$, this same induction case then generates the set of odd natural numbers.

The Induction Rule



- Our prior implementation was fairly restricted
- If we want to implement the following set:

$$0 \in S$$

$$x \in S \rightarrow x + 1 \in S$$

- We can do the following

```
increment :: Integer -> Integer
```

```
increment x = x + 1
```

```
s :: [Integer]
```

```
s = 0 : map increment s
```

- This style of programming is called **data recursion**
- *map* will proceed down s, creating each value it needs, then using it.

⌘ Defining Sets Inductively

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- Beyond the base and inductive cases, inductive set definition needs one more component: the *extremal clause*
- **Extremal Clause:** A statement which excludes anything from the set that are not introduced by the base case, or are instantiations of the induction case, it reads something like the following:
“Nothing is an element of the set unless it can be constructed by a finite number of uses of the first two clauses”
- Thus all inductive set definitions include 3 parts:
 - **Base Case:** a simple statement of some mathematical fact: i.e., $1 \in S$
 - **Induction Case:** an implication in a general form: $\forall x \in U, x \in S \rightarrow x + 1 \in S$
 - **Extremal Clause:** Nothing is in the set being defined unless it got there by a finite number of uses of the first two cases

- The set of natural numbers, \mathbb{N} , is defined as follows
 - **Base Case:** $0 \in \mathbb{N}$
 - **Induction case:** $x \in \mathbb{N} \rightarrow x + 1 \in \mathbb{N}$
 - **Extremal clause:** nothing is an element of the set \mathbb{N} unless it can be constructed with a finite number of uses of the base and induction cases.
- We can show that an arbitrary number above and including 0 are in \mathbb{N}
 1. $0 \in \mathbb{N}$ Base Case
 2. $0 \in \mathbb{N} \rightarrow 1 \in \mathbb{N}$ *instantiation rule, induction case*
 3. $1 \in \mathbb{N}$ 1, 2, Modus Ponens
 4. $1 \in \mathbb{N} \rightarrow 2 \in \mathbb{N}$ instantiation rule, induction case
 5. $2 \in \mathbb{N}$ 3, 4, Modus Ponens



- Let *BinDigit* be the set $\{0, 1\}$. The set *BinWords* of machine words in binary is defined as follows:
 - Base Case:** $x \in \text{BinDigit} \rightarrow x \in \text{BinWords}$
 - Induction Case:** if x is a binary digit and y is a binary word, then their concatenation xy is also a binary word

$$(x \in \text{BinDigit} \wedge y \in \text{BinWords}) \rightarrow xy \in \text{BinWords}$$

- Extremal Clause:** Nothing is an element of *BinWords* unless it can be constructed with a finite number of uses of the base and induction cases
- A set based on another set S in this way is given the name S^+
 - it is the set of all possible non-empty strings over S
 - S^* is similar to S^+ except S^* includes the empty string
 - BinWords* could have also been written as BinDigit^+

- We can define a function to create two new BinWords based on one that has been provided
 - i.e., given [1, 0] it will return [0, 1, 0] and [1, 1, 0]

```
newBinaryWords :: [Integer] -> [[Integer]]  
newBinaryWords ys = [0:ys, 1:ys]
```

- We then define the set of BinWords as:

```
mappend :: (a -> [b]) -> [a] -> [b]  
mappend f []      = []  
mappend f (x:xs) = f x ++ mappend f xs  
  
binWords = [0] : [1] : (mappend newBinaryWords binWords)
```

The Set of Integers



- Both of the prior sets are **well-founded**, meaning they are infinite in only one direction, and they have a *least* element
- **Countable Set:** a set which can be counted using the natural numbers
 - Are the integers countable?
 - Doesn't have a least element
 - Infinite in two directions
 - However we can count them using natural numbers as follows:
 - Start at 0
 - For every number $n \in \mathbb{N}$, we count both n and $-n$ in \mathbb{Z}
 - That is, we can consider the set of integers as an infinitely long tape folded in half at 0, and then count the overlapping numbers $(i, -i)$ for each $i \in \mathbb{N}$
- Yet, this does not specify \mathbb{Z}

The Set of Integers



- The set \mathbb{Z} of integer is defined as follows:

- Base Case:** $0 \in \mathbb{Z}$
- Induction Case:**
 $(x \in \mathbb{Z} \wedge x \geq 0) \rightarrow x + 1 \in \mathbb{Z} \wedge -(x + 1) \in \mathbb{Z}$
- Extremal Clause:** nothing is in \mathbb{Z} unless its presence is justified by a finite number of uses of the base and induction cases

Thus, we can define integers using Haskell, as follows

```
build :: a -> (a -> a) -> Set a
build a f = set
    where set = a : map f set

builds :: a -> (a -> [a]) -> Set a
builds a f = set
    where set = a : mappend f set

nextInteger :: Integer -> [Integer]
nextInteger x
    = if x > 0 \ / x == 0
      then [x + 1, -(x + 1)]
      else []

integer :: [Integer]
integers = builds 0 next Integers
```

For Next Time



- Review DMUC Chapter 3 and 9
- Review DMA Chapters 3.1 and 5.3 - 5.5
- Review this Lecture
- Read DMUC Chapter 4
- Read DMA Chapters 5.1 - 5.2





Are there any questions?