



CSc 28

Discrete Structures

Chapter 4

Data Structures & Algorithms

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Syllabus

- **Algorithm**
- **Searching**
- **Interesting Experience**
- **Number Theory**
- **Prime Numbers**
- **Representation of Integers**
- **Congruence**
- **Graphs**

Algorithm

Definition: Algorithm

- Informally: **Algorithm** is a discrete sequence of precise instructions to compute a solution to a problem in finite time
- Alternatively from Wikipedia [1]:
 - “The word *algorithm* itself is derived from the 9th century mathematician *Muḥammad ibn Mūsā al-Khwārizmī*, Latinized *Algoritmi*.
 - A partial formalization of what would become the **modern concept of algorithm** began with attempts to solve the **Entscheidungsproblem**, posed by Hilbert in 1928.
 - Later formalizations were framed as attempts to define **Effective Calculability** or **Effective Method**.”

Algorithm

Purpose & Properties of algorithm:

- Algorithm performs operations on data structures
- **Input** from a specified data set, e.g. file (data source)
- **Output** onto specified data set or file (solution)
- **Unambiguous** definition of every compute step
- Correct mapping defined for every possible input, even erroneous input data
- **Finiteness** of number of calculation steps
- Effectiveness of each calculation step

Algorithm



Named after Persian Mathematician **al-Khwarizmi** 780-850

Algorithm



Dijkstra's general programmer attributes:

- Critical **thinker**, uses good judgment
- **Creative** ideas, about algorithms, computing
- **Open** to others' ideas
- **Humble** 😊, see **E. Dijkstra** [1]

Practicing and mastering a craft:

- Programmer analyzes, improves programs
- Designs programs, i.e. SW solutions of computable problems
- Practices, reviews, corrects, refines programming process

Experience and practice:

- Knows how to design, implement, document, test, validate, improve, maintain, judge complex software

Mastery and skills:

- Advances the **state of the art** of SW development to a higher level
- Designs, implements cooperatively; true SW jobs generally are not solvable by 1 engineer; except **Linus Torvalds** implementing Linux 😊

System Programmer

Necessary **professional** (Pro) requirements for a successful **system programmer**:

- The **Pro** masters and uses some stable, well-defined, high-level programming language **L**
- Language **L** allows visibility of, and access to, low-level target machine, or to some OS resources
- Possible languages: Java, C, C++, assembler, . . .
- Pro knows target machine thoroughly
- Understands which operating system services can be expressed in **L** to manage system resources
- Pro can combine system calls to low-level machine from **L**, with overall goal of effective machine use

Good Judgment comes from Experience, and Experience comes from Bad Judgment

provided that one is smart!

Only a CS *genius* learns exclusively from the mistakes of others

The *smart Sac State CS student* learns from her own mistakes

A *dumb* person –not studying CS ☺ of course– repeats errors

Software Development Practices

SW with true value should be:

- **Functional**
- **Correct**
- **Reliable**
- **Efficient**
- **Easily readable: C. A. R. Hoare [4]**
- **Re-Usable**
- **Extendible, AKA open-ended**
- **Maintainable**
- **Complete; exceptions are documented clearly!**
- **Does such SW exist? Then why don't we see & touch it all the time? Ever touch an OS that locks up, freezes, slows down?**



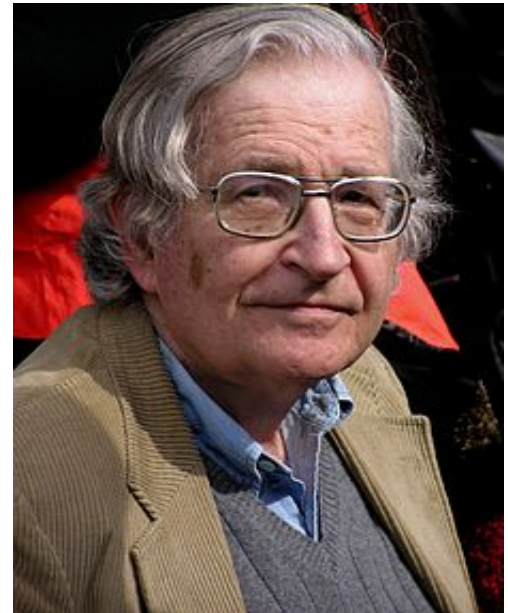
Language Complexity

Chomsky complexity measure for grammars, or for languages defined by such grammars, classified according to **Noam Chomsky** hierarchy:

- Type 3 Regular Expressions
- Type 2 Context Free Languages
- Type 1 Context Sensitive Languages
- Type 0 Recursively Enumerable (AKA **free**)

Listed in **simple (3)** to **complex (0)** order

See later detailed lecture on language complexity, and **program complexity**



Chomsky Hierarchy

Ref: <http://www.cs.nuim.ie/~jpower/Courses/Previous/parsing/node21.html>

Language Class	Grammar	Automaton
3	Regular	NFA or DFA
2	Context-Free	Push-Down Automaton
1	Context-Sensitive	Linear-Bounded Automaton
0	Unrestricted (or <i>Free</i>)	Turing Machine

Interesting Experiences

1. I re-typed ~50 lines of Pascal-like system code to render buggy SW fully functional; mystery never solved why the re-typed, “**identical**” program worked!
2. Designed and coded some days and nights w. almost no sleep and food, and crafted a few 1,000 lines of highly functional C++ code for **Ada Case Statement** in Ada compiler; worked! Until years later I observed an error in that Case Statement implementation!
3. Colleague left 1 page of pseudo-code for **Symbolic Differentiation** on copy machine; was intuitive, comprehensible, beautiful code; I stole with pride
4. Positive feedback about an almost *religious credo*: **do not trust your own SW**; but insert **checks** instead! Catch your own assumptions! Better: make few assumptions

What Can Be Programmed?

- Definition of algorithm; see [2]
- Guideline for what can be programmed: **Church Thesis** [6]: “any algorithmically computable function can be programmed and executed on a regular computer.” I.e. can be expressed as a C++ or Java program
- What is “Computable?”
 - see **Alonzo Church’s** Lambda Calculus
 - or **Alan Turing’s** “Turing Machine”
- Yet some problems remain **very hard** to solve programmatically, though being **computable**

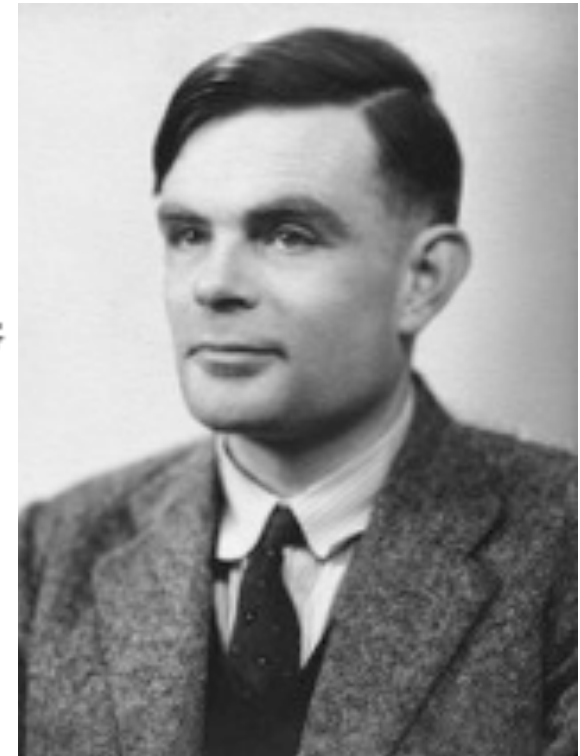
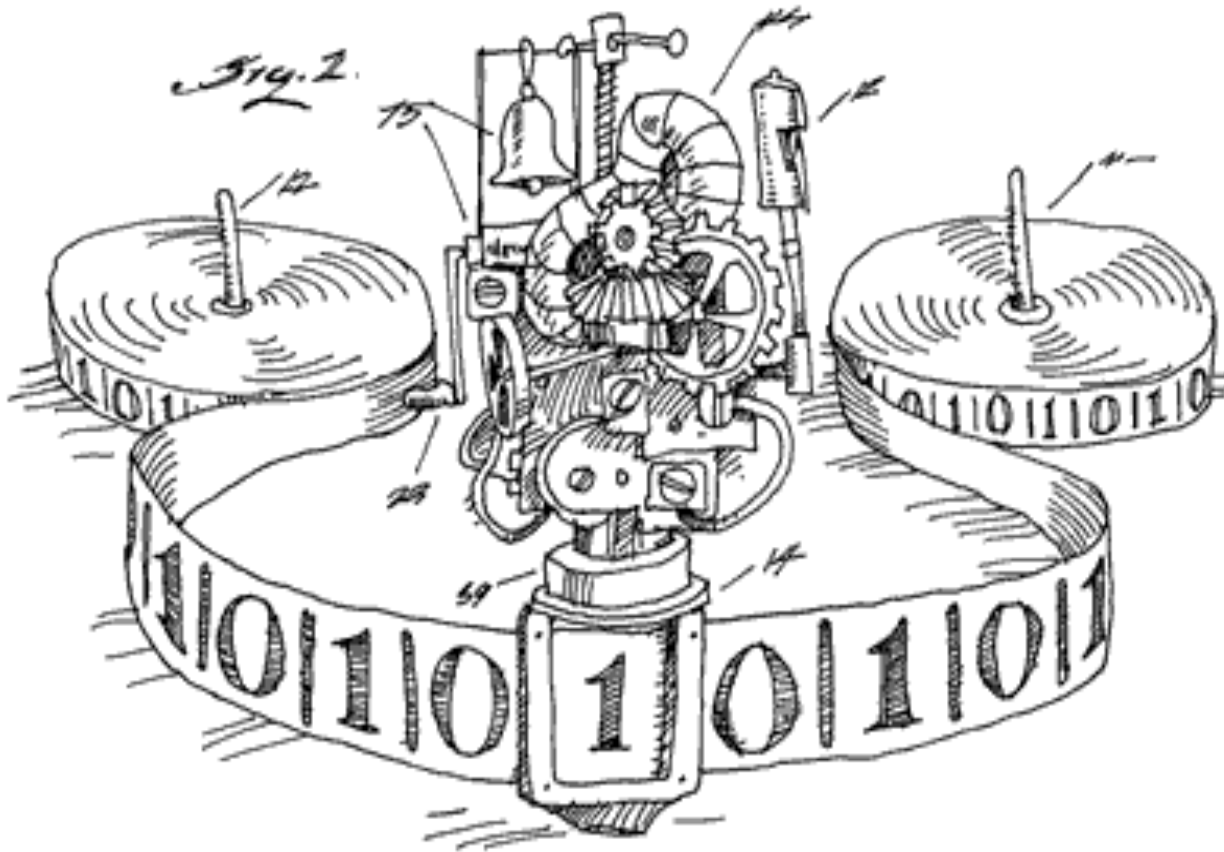
What Can Be Programmed?



American Mathematician **Alonzo Church** 1903 - 1995

What Can Be Programmed?

Maybe this playful **Turing Machine**, with bells & whistles, zeros & ones, can be programmed?



What Cannot be Programmed?

- **How to become a Perfect System Programmer?**
 - There is no methodical, algorithmic, guaranteed way ☹
 - But there *are* ways ☺ Don't give up! Includes learning, practicing, thinking
- **Design SW to win lottery?**
 - Is easy, provided unbounded time and all permutations allowed; you'd win, but then you'd *lose* money due to large number of ticket bought ☺ for all combinations
 - So you can't win!
- **Natural language translator?**
 - Yes, there exist automated translation tools! And your teacher uses them! Like *Babel Fish*, *Google translate*, etc.
 - If it were possible, how then would common ambiguities in human interaction be avoided? Forever?

What Cannot be Programmed?

- **Decryption of any encrypted code?**
 - May require too much time to render result interesting
 - Hence some nations impose limitations on encryption complexity (e.g. 128 bits) in the US for outside-national communication
- **Medical diagnosis?**
 - Yet great steps achieved to **automate some diagnostic steps**
 - Mainly to save time; final diagnosis generally left to MD
- **Judicial judgment?**

Algorithm

Markov (1954)

- Algorithm is ... an exact prescription, defining a **computational process**, leading from various **initial data** to the desired **result**

Donald Knuth (1968)

- A precisely defined sequence of **finite** steps to compute a result from given **inputs and initial values** –paraphrased by HGM 😊

Stone (1972)

- “An algorithm is a set of **rules that precisely defines** a sequence of operations such that each rule is effective and definite and that the **sequence terminates** in (very) finite time.”

Algorithm

Minsky (1967)

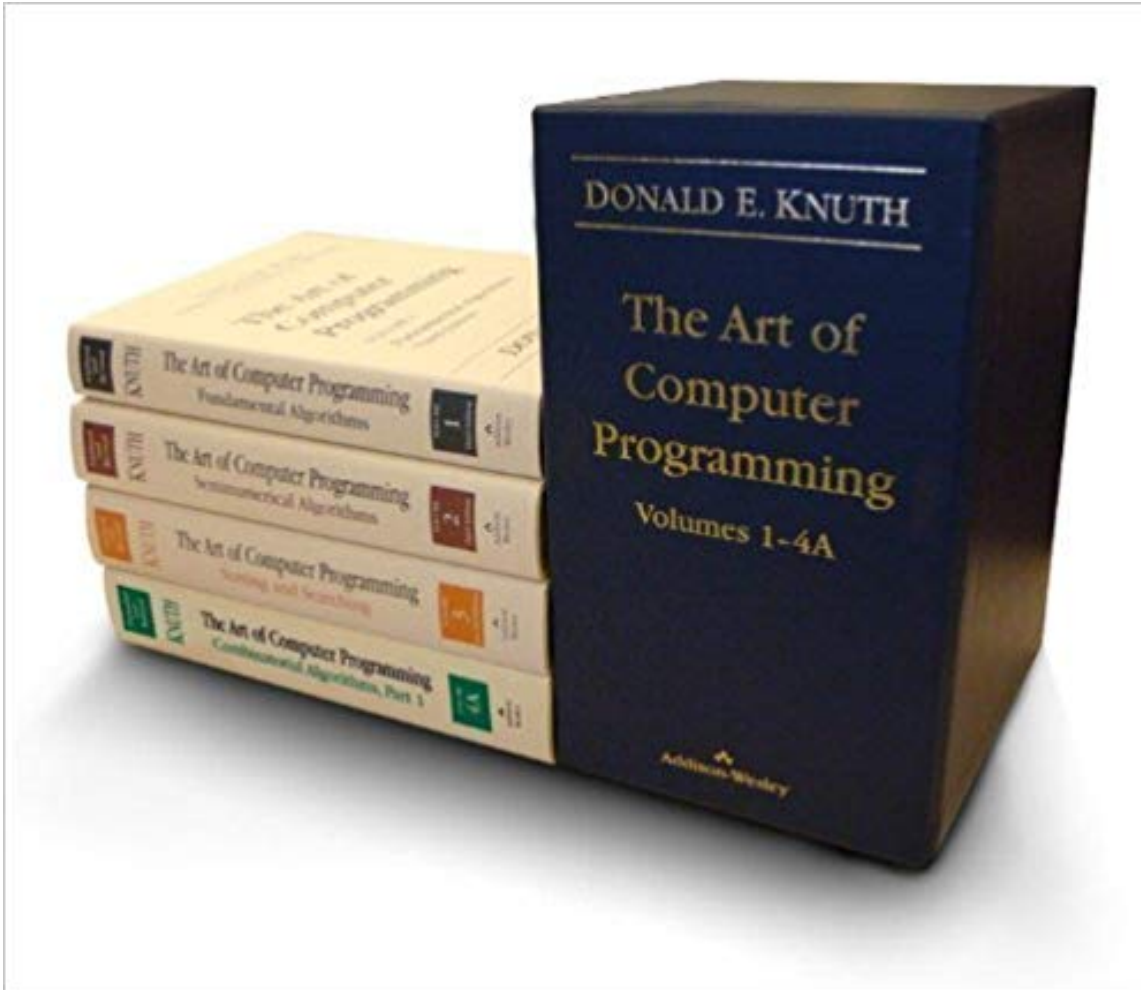
- Algorithm is a synonym for “effective procedure”

Berlinsky (2000)

- Algorithm is a **finite procedure**, written in a fixed symbolic vocabulary, governed by **precise** instructions, moving in **discrete steps**, 1, 2, 3, . . . , whose execution requires no insight, cleverness, intuition, intelligence, or perspicuity, and that sooner or later **comes to an end**

The Art of Computer Programming

Donald Knuth's **Opus Magnum** today 2021



Defensive Programming

Automation

- Some aspects of programming can be **automated**
- Many web interfaces to users/customers are automated
- How many times did you have to retype correct web-page information because 1 item further down was misspelled?
- Learning: simple things can be automated, but even for those, best to use good programming principles, consistency, clear common sense
- See Richard Sites, main designer of Alpha processor [5]:
“I’d rather write programs that write programs than write programs!”
- Automate what can be; and “manually” program the rest
- Programming the remaining portion, the hard problems, is a challenge

Programming



Defensive Programming

Future for Sac State CS students:

- Programmers will highly likely have very good work opportunities for a long time to come, in an exciting, fun profession!
- Like in all professions: The good ones will be in high demand, and craft new technologies through their work
- Others write web interfaces that clear out ☹ after a user makes 1 typo!

Don't be complacent; consider future enhancements!

- Today your SW works perfectly fine for some given spec.
- Tomorrow's spec. will change, and you need to adapt your SW to the modified requirements: Craft **maintainable SW**
- Tomorrow new errors pop up 😊

Defensive Programming

Don't trust your interface; *verify* even if checks seem unnecessary!

- Defined interface dictates your SW inputs, and specifies the output your SW is to generate
- **Verify the accuracy** of input, even your own; maybe **especially!**
- Generate messages where applicable and beneficial
- May not work for embedded SW, space mission 200 Mio miles from home planet; but appropriate default action should be meaningful

Don't trust your SW; check and *verify* all input instead!

- Other SW may automatically generate your program's input
- And even you!! make errors: so check, verify, report, mistrust
- Make on-line, live "reports" of suspicious logic, use **#ifdef** while developing SW product

Defensive Programming

Telerik Academy

What is Defensive Programming?

- ◆ Similar to defensive driving – you are never sure what other drivers will do



- ◆ Expect incorrect input and handle it correctly
- ◆ Think not only about the usual execution flow, but consider also unusual situations

Activate Windows

Go to Settings to activate Windows.

Algorithm Sample: Search

Goal of Search

- Have a data structure and an interesting datum:
- Searching: The process of inspecting a **data structure** for a specified **datum**
- Goal of search may be to determine, whether the **datum is included** in the searched structure
 - Upon finding the datum, the search may end successfully
 - Frequently the goal of a search is to know the specific **location** or **index**, where the datum was found
 - If not found, the complete data structure had to be searched
 - Yet in case of sorted data structures this may not be as costly as it sounds! I.e. the cost may be way $\ll O(n)$
- Another goal of a search may be to determine the number of occurrences of a datum in the data structure being inspected

Algorithm Sample:

Search

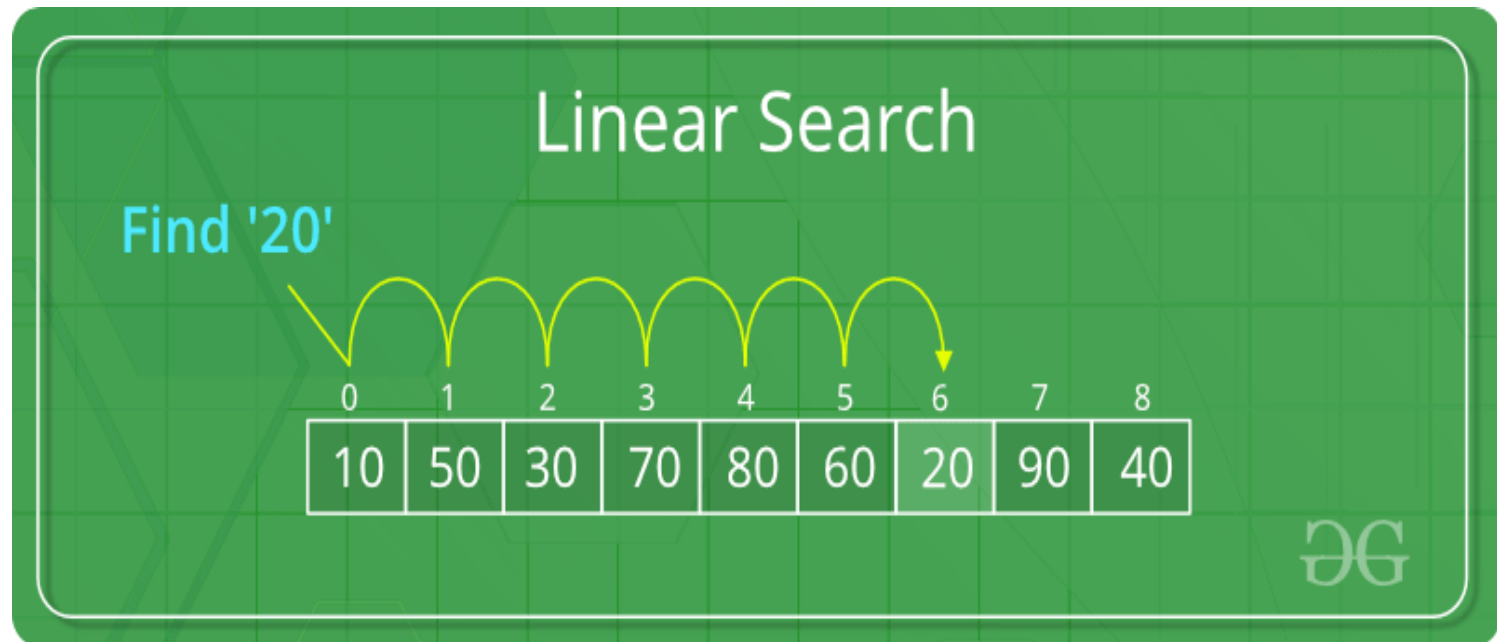
Cost of Search

- In linear search, the cost to find a datum is n , with n being the size of the data structure
- Or more concisely: Must look at up to all n elements in **linear search** to locate an item
- In **Big O notation** we say, the cost function for 1 lookup in a linear search is $O(n)$ (Big O, see later!)
- It doesn't matter that sometimes the first try may be successful, the overall cost function is still $O(n)$
- For the binary search **in a sorted list**, the lookup cost is $O(\log_2(n))$, since the worst case, to find an item, is to inspect $\log_2(n)$ elements
- For a balanced binary tree, cost is really $O(\log_2(n))$, but for an unbalanced tree it can be as bad as $O(n)$
- Redundant to say \log_2 as in Big O notation, the ratio of \log_2 by \log_{10} is constant, i.e. “noise” in Big O

Linear Search

- **Linear search** makes no assumption on the order, in which data are stored in a **data structure**, the data generally are unsorted!
- Data structure is often an **array**; but could be a linked list or any equivalent data structure
- If it is known that datum **x** is stored uniquely (occurs at most once) then search may terminate at first successful find
- To prove that a **datum** is not included, the complete data structure must be inspected in **linear search**
- Hence the cost function for a linear search is $O(n)$, with **n** being size of data structure
- If datum may be included more than once, then even after successful find the complete data structure must be inspected; result is **count of** occurrences

Linear Search



Linear Search

```
#include <stdio.h>
#include <iostream.h>
#define MAX 20
// slot at index 0 is NOT used to store data
// instead, index 0 used to indicate: "not found"
int unsorted[ MAX ] =
{
    0, 9, -99, 99, 999, -999, 7, -7, 77, 88,
    1, 2, 9, 8, 3, 4, 7, 16, 77, 22
};
```

Linear Search

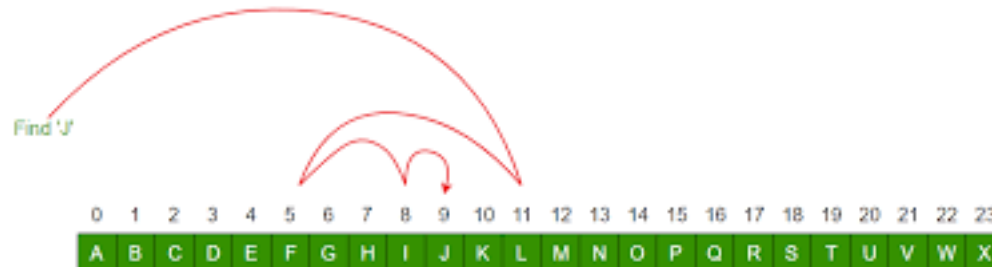
```
// assumes global array unsorted[] with MAX elements
// Note element unsorted[ 0 ] is NOT used:
// as index 0 does signify: Not Found!
// return 0 if element n is NOT found in unsorted[]
// searching n, return index of first occurrence
int linear( int n )
{ // linear
    // start search at index 1; index 0 not used
    for( int i = 1; i < MAX; i++ ) {
        if( unsorted[ i ] == n ) {
            // found it
            return i;
        } //end if
    } //end for
    return 0;    // 0 used for "not found!"
} //end linear
```

Binary Search, Play Game

- Practice the **Guessing Game** in class:
- Students think of an integer number **between 1 and 1,000**; keep it secret
- I can guess your secret number in at most 10 tries; exactly that secret number!
- No cheating ☺ and no changing of mind!
- Each time **not yet found**, all you need to say is: “no” and whether the actual number is $>$ or $<$
- Why does that work?
- Binary search cost is $O(\log_2(n))$

Binary Search

- Binary search assumes that data structure be **sorted**, generally in some **array[]** or equivalent data structure
- If data are unique –occurring at most once– then we say the **sorted** array holds data in **ascending** or **descending** order
- Alternatively we say: **non-ascending** and **non-descending**; sounds a bit awkward
- The method is:



Binary Search: Method

- Use: **left** and **right** indices; **array[]** sorted in **ascending** order, wanted element is: **n**
- Initially **left = 0** in C++, **right** is index of last array element: **right = size - 1**
- **Loop**: Repeatedly do the following “guess” and verify:
- Make a wild guess, that **n** be in the middle, named **mid**, between **left** and **right**, i.e. **mid = (left + right) / 2**
- If **array[mid] == n** the algorithm terminates: **Found!**

Binary Search: Method

- Else adjust indices: **left**: upward, or **right**: downward
- If wanted element is greater than the found one, search continues, with **left = mid + 1**; i.e. search in upper portion
- Else continue with **right = mid - 1**; i.e. search in lower portion
- When **left** is beyond **right**, in that case **n** is not found and the algorithm stops
- Else continue at **loop**

Binary Search 1

```
#include <stdio.h>
#include <iostream.h>
#define MAX 20
// also slot at index 0 not used in this binary search!
// simply a convention used here!
int sorted[ MAX ] =
{
    0, -88, -77, -12, -4, -2, -1, 0, 1, 4,
    5, 14, 17, 66, 77, 99, 100, 1001, 2015, 9999
};
```

Binary Search 1

```
// binary search for element n in sorted[] array
// sorted[] size MAX is known globally ☹
// element sorted[ 0 ] is NOT used
// stored in ascending order! (non-descending!)
// if n not found, return 0; else return index of n
int binary( int n ) // search for n
{ // binary
    int left  = 1;      // don't use index 0
    int right = MAX-1;  // left, right approach
    int mid   = ( left + right ) / 2;
    . . .
```


Binary Search 1

```
int binary( int n )      // search for n, return 0 if not found
{ // binary              // use of globals not always best
    int left  = 1;        // don't use index 0
    int right = MAX-1;    // left, right, middle approach
    int mid   = ( left + right ) / 2;
    while( ( sorted[ mid ] != n ) && ( left < right ) ) {
        // n not yet found, and not looked at all elements
        if( sorted[ mid ] > n ) {
            right = mid - 1;    // search in "lower" half
        }else if( sorted[ mid ] < n ) {
            left = mid + 1;     // search in "upper" half
        }else{
            break;              // found n
        } //end if
        mid = ( left + right ) / 2;
    } //end while
    return sorted[ mid ] == n ? mid : 0; // conditional expr.
} //end binary
```

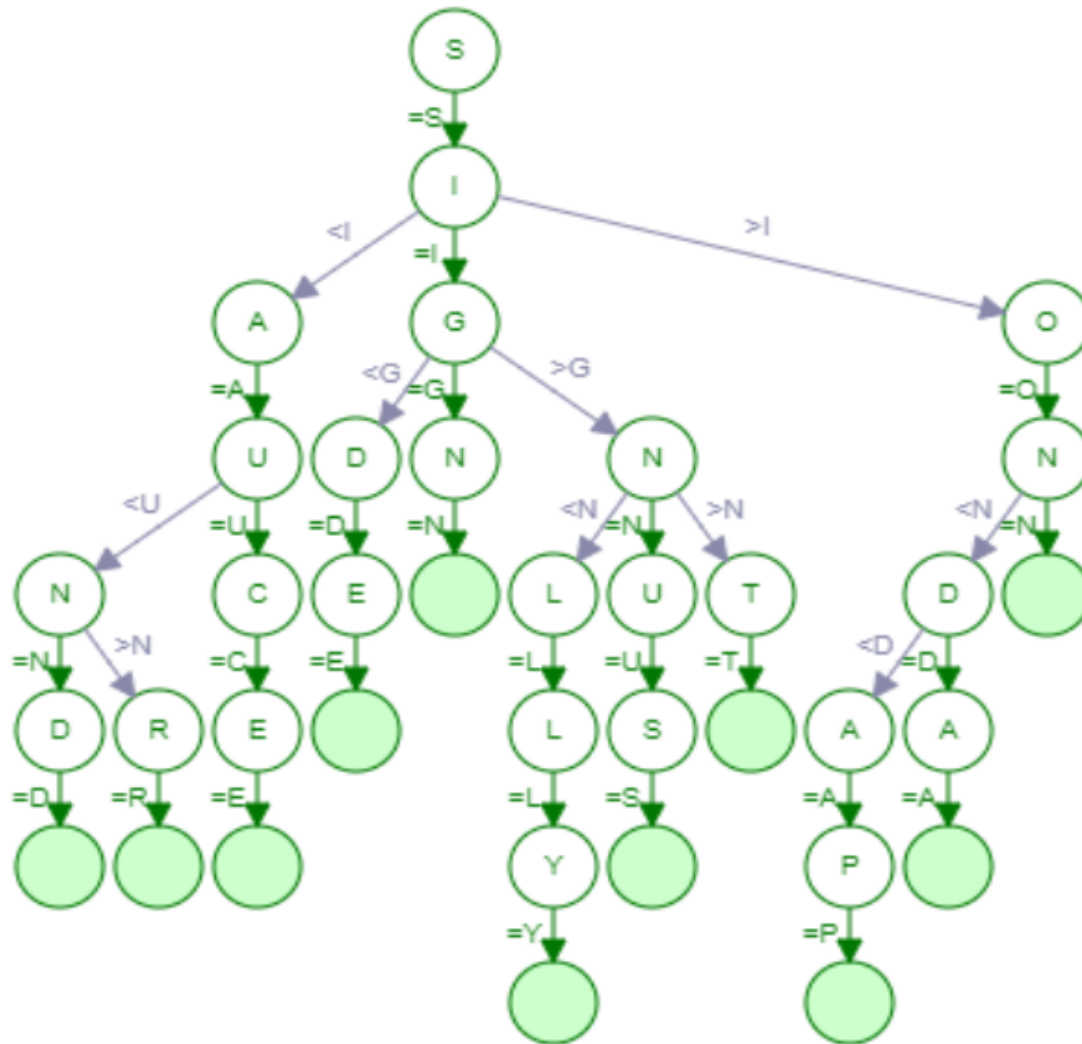
Better Binary Search 2

```
int sorted[ MAX ] = { -12, -6, -2, -1, 0, 1, 2, 16, 44, 120 };
// here element at index 0 is used
int binary_search( int first, int last, int key )
{ // binary_search
    int mid;
    if ( first > last ) {
        return NIL;          // not in range 0..MAX-1
    } //end if
    mid = ( first + last ) / 2;
    if ( key == sorted[ mid ] ) {
        return mid;
    }else if( key < sorted[ mid ] ) {
        return binary_search( first, mid-1, key );
    }else{
        // key < sorted[ mid ], so: first = mid+1
        return binary_search( mid+1, last, key );
    } // end if
} //end binary_search
```

Binary Search in Trees

- Searching for a datum **n** in a sorted binary tree is similar:
 - Initially node pointer **p** is set to the root
- If wanted element **n** is at **p->data**, the algorithm ends successfully, returning **p**
- Else, if **n > p->data**, search the right subtree: rooted at slot: **p = p->right**
- Else search the left subtree: **p = p->left**
- If **p** is ever null, **n** is not found; **n** is not in the tree!
- **Careful**, if tree is completely **unbalanced**, tree traversal can be as slow as linear search: **O(n)**
- Ternary Trees, next page, used in auto-completion techniques

Ternary Trees

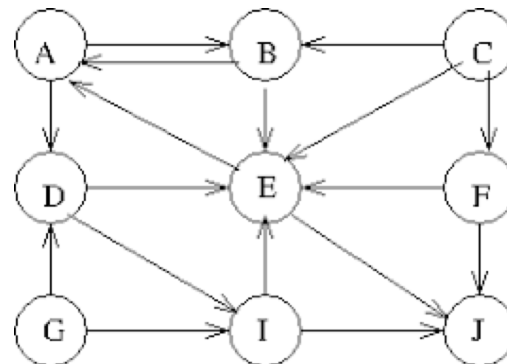


Search in Graph G

- A graph **G** is a data structure of **nodes** and connecting **edges**
- Edges may be undirected or directed
- A simplification is: to **view undirected edges a bi-directional**
- Any node in G may have **any number incident edges**
- Unlike tree: In a tree, nodes have exactly 1 incident edge; always!
- In contrast, graph G may have any # of incident edges
- A graph G may even be **unconnected!** In which case there are no incident edges to some nodes
- Use “trick” to traverse! See work-around later!

Search in Graph G

- From “G has any number incident edges” follows:
- . . . that G may be **unconnected**, when there are no incident edges to some nodes
- Hence a **binary search** for a general G is not feasible
- Unless an ancillary data structure is provided, even the **linear search** would be infeasible



Search in Graph G

- **Work-around: add a field to each node of G; but this field is NOT logically part of G proper!**
- **Instead, this field named *finger*, allows an algorithm to weave through all nodes of G, in linear (or other) fashion, *even if unconnected***
- **Don't forget to add a Boolean *visited* field to *avoid repeat visits!***

Bubble Sort

Sorting

- **Sorting to be handled in detail in later section**
- **Here outline of so called: “Bubble Sort”**
- **Repeatedly search through –the remainder of– an array, to identify the largest (or smallest) elements, and exchange with the current element**
- **Have 2 nested loops**
- **Repeated extract the one (smallest, or largest) and place into right array position**
- **Cost if high: doubly nested loop**

Bubble Sort Array

```
// bubble sort array a[] of ints; global an exception ☺  
// outer loop start index 0; ends 1 before list-end  
// inner loop starts at index+1; ends at last index
```

```
void sort()  
{ // sort  
    for( int left = 0; left < MAX-1; left++ ) {  
        for( int right = left+1; right < MAX; right++ ) {  
            // is right element greater? if so: manual swap  
            if( a[ right ] > a[ left ] ) { // descending  
                int temp      = a[ right ];  
                a[ right ]    = a[ left ];  
                a[ left ]     = temp;  
            } //end if  
        } //end for  
    } //end for  
} //end sort
```

Print Sorted Array

```
int main( void )
{ // main
    // print global array a[] unsorted
    print( "before sorting:" );

    sort();

    // print global array a[] sorted
    print( "after sorting:" );
} //end main
```

Growth of Functions:

Big O Notation

Big O Notation

- **Big O notation:** A mathematical notation expressing the limiting behavior of a function when its argument grows towards a particular boundary value, usually toward infinity
- Is a member of a family of notations invented by **Paul Bachmann, Edmund Landau**, et al.
- Practiced widely in Math and Computer Science
- Collectively called **Bachmann–Landau** notation, AKA **asymptotic** notation
- Used in CS to **characterize algorithms** according their runtime or space requirements, as a function of the input size or time for lookup

Big O Notation

- Use of letter capital **O**, as growth rate of a function is referred to as the: **Order of the function**
- Characterizing a function via **big O notation** provides Upper Bounds on the growth of such a function
- In typical use **Big O Notation** is asymptotical: It refers to very large **x** of $O(x)$
- Hence the contribution of terms that grow **most quickly** eventually renders other terms irrelevant
- Following simplification rules can be applied:
 1. If $f(x)$ is a **sum of several terms**, and one of these has a largest growth rate, it can be kept; all others can be omitted in Big O
 2. If $f(x)$ is a **product of several factors**, any other constant factors (that are not $f(x)$) can also be omitted

Growth of Functions

- Computational growth of functions is usually described using that **Big O notation**
- **Definition:** Let **f** and **g** be functions $f(x)$ and $g(x)$ from integer numbers x to the real numbers
- We say that **$f(x)$ is $O(g(x))$** if there are constants **C** and **k** such that:
(Operator symbols $|$ used for *absolute value*)

$$|f(x)| \leq C \cdot |g(x)|$$

whenever $x > k$

Growth of Functions

- When we analyze the growth of **complexity functions**, typically named $f(x)$ and $g(x)$, we view $f(x)$ and $g(x)$ always as positive
- Therefore, we can simplify the Big O requirement to

$$f(x) \leq C \cdot g(x) \text{ whenever } x > k$$

- If we want to show that $f(x)$ is $O(g(x))$, we only need to find **some** pair (C, k)

Growth of Functions

- Idea behind **Big O** notation: Establish **upper bound** for the growth of a function $f(x)$ for **large** x
- This boundary is specified by a function $g(x)$ that is usually much **simpler** than $f(x)$
- We accept (i.e. ignore) constant **C** in the requirement

$$f(x) \leq C \cdot g(x) \text{ whenever } x > k$$

- because **C does not grow**, or even change **with** x
- We are interested in large x , so it is OK if
 $f(x) > C \cdot g(x)$ for $x \leq k$

Growth of Functions

Example:

- Show that $f(x) = x^2 + 2x + 1$ is really just: $O(x^2)$
- For $x > 1$ we have:

$$\begin{aligned}x^2 + 2x + 1 &\leq x^2 + 2x^2 + x^2 \\ \Rightarrow x^2 + 2x + 1 &\leq 4x^2\end{aligned}$$

- Therefore, for $C = 4$ and $k = 1$:
- $f(x) \leq C \cdot x^2$ whenever $x > k$
- $\Rightarrow f(x)$ is $O(x^2)$

Growth of Functions

- Question: If $f(x)$ is $O(x^2)$, would its cost function also be bounded by (i.e. no greater than) $O(x^3)$?
- Or even higher
- Yes, $f(x)$ of order $O(x^2)$ is bounded by $O(x^3)$
- After all, x^3 grows faster than x^2 , so x^3 grows also faster than $f(x)$ which is limited by x^2
- We lose precision computing a cost function
- Should always aim to identify the simplest (cheapest) function $g(x)$ for which $f(x)$ is $O(g(x))$

Growth of Functions

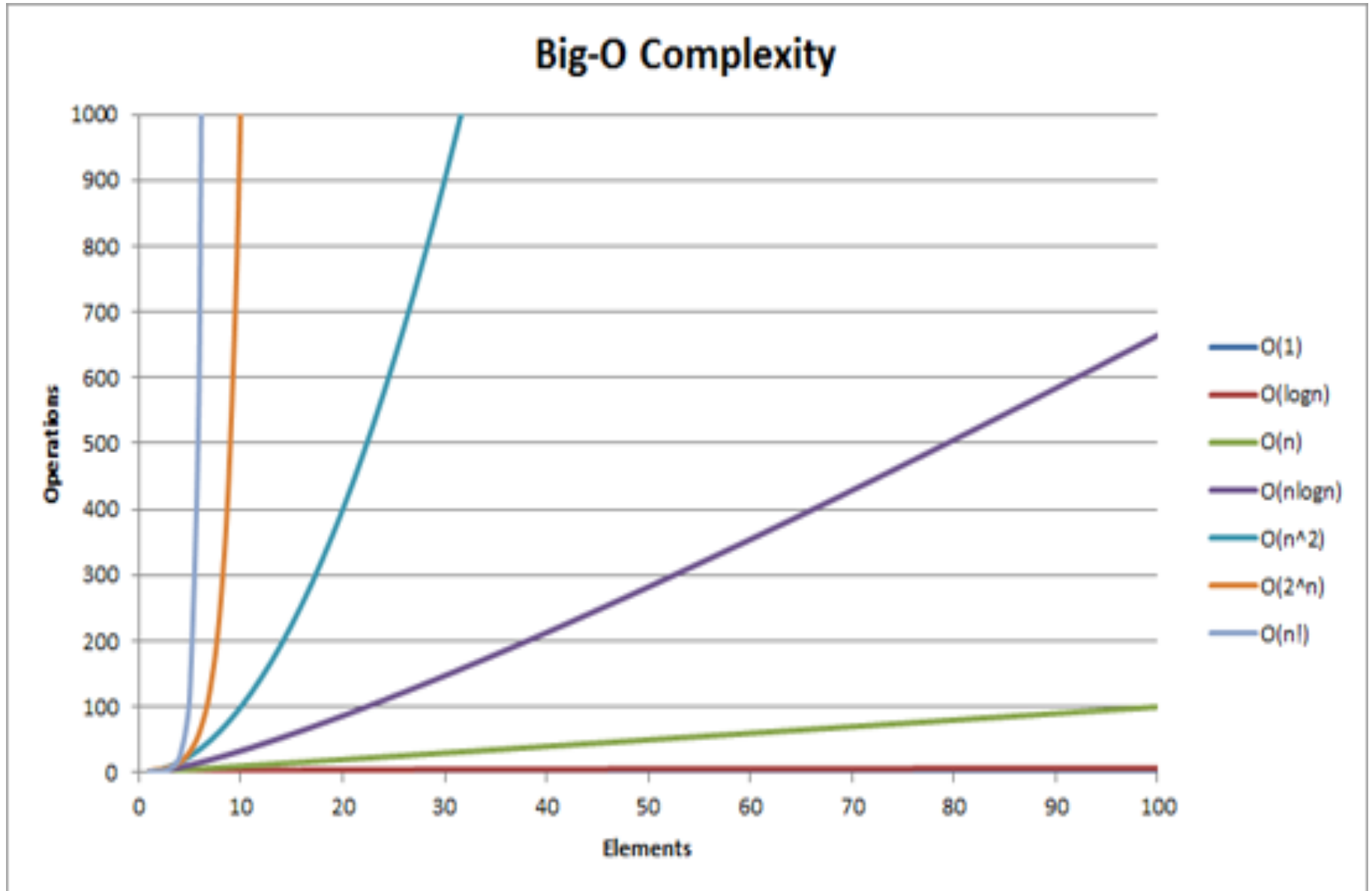
- Sample unordered generic functions $g(n)$ in Big O notation:

$n, \log n, 1, 2^n, n^2, n!, n, n^3, n \log n$

- Listed top-down from smallest to largest in terms of cost:

1
log n
n
n log n
 n^2
 n^3
 2^n
 $n!$
 n^n

Growth of Functions



Growth of Functions

- A problem that can be **solved with polynomial** worst-case complexity is called: **tractable**
- Problems of higher complexity are called: **intractable**
- Problems that no algorithm can solve are called: **unsolvable**

Useful Rules for **Big O**

- For any **polynomial** $f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_0$, where a_0, a_1, \dots, a_n are real numbers:
- Such an **$f(x)$** is $O(x^n)$

If cost function for **$f_1(x)$** is $O(g_1(x))$ and for $f_2(x)$ is $O(g_2(x))$:
 $(f_1 + f_2)(x)$ is **$O(\max(g_1(x), g_2(x)))$**

If $f_1(x)$ is $O(g(x))$ and $f_2(x)$ is $O(g(x))$, then
 $(f_1 + f_2)(x)$ is **$O(g(x))$**

Complexity Examples

What does the following algorithm compute?

```
procedure who_knows( a[n] : int )  
  m := 0  
  for i := 1 to n-1  
    for j := i + 1 to n  
      if  $|a[i] - a[j]| > m$  then m :=  $|a[i] - a[j]|$   
    end for  
  end for  
  . . .  
  ?
```


Complexity Examples

What does the following algorithm compute?

```
procedure who_knows( a[n] : int )  
  m := 0  
  for i := 1 to n-1  
    for j := i + 1 to n  
      if |a[i]-a[j]| > m then m := |a[i]-a[j]|  
    end for  
  end for  
  --m is max abs(diff) between any pair
```

Number of comparisons is $\sim (n-1)n/2 = 0.5n^2 - 0.5n$

Time complexity is $O(n^2)$

Factor 0.5 of n^2 falls by wayside; Summand $- 0.5n$ as well!

Complexity Examples

Another algorithm solving a different search problem:

```
procedure max_diff( a[n] : int)
min := a1
max := a1
for i := 2 to n
    if a[i] < min then
        min := a[i]
    elsif a[i] > max then -- elsif a la Ada
        max := a[i]
    end if
m := max - min
. . .
```

Comparisons: $2(n - 2)$; so time complexity is $O(n)$

Number Theory

Introduction to Number Theory

- Number theory **treats integers & their properties**
- We will start with the basic principles of
 - divisibility
 - greatest common divisors **gcd**
 - least common multiples, and
 - modular arithmetic

And analyze some relevant algorithms
Aside from **Fermat's Conjecture**

(No longer **Fermat's Conjecture**!

Proven in 2016:

No three positive integers a , b , and c satisfy
equation $a^n + b^n = c^n$ for any integer $n > 2$)



Pierre de Fermat

Division

- If a and b are integers with $a \neq 0$, we say that a divides b if there is an integer c so that $b = ac$
- When a divides b we say that a is a factor of b , or equivalently that b is a multiple of a
- Notation $a \mid b$ means: “integer a divides integer b ” evenly
- We write $a \nmid b$ when integer a does not divide b

Divisibility Theorems

For integers a , b , and c it is true that:

- if $a \mid b$ and $a \mid c$, then $a \mid (b + c)$
- Example:
 - $3 \mid 6$ and $3 \mid 9$, so $3 \mid 15$
- if $a \mid b$, then $a \mid b * c$ for all integers c
- Example:
 - $5 \mid 10$, so $5 \mid 20$, $5 \mid 30$, $5 \mid 40$, ...
- if $a \mid b$ and $b \mid c$, then $a \mid c$
- Example:
 - $4 \mid 8$ and $8 \mid 24$, so $4 \mid 24$

Primes

- A positive integer p greater than 1 is called **prime** if the only positive factors of p are 1 and p
- Note: Here 1 itself is not considered a prime
- A positive integer that is greater than 1 and not prime is called a **composite**
- A fundamental theorem of arithmetic:
 - Every positive integer can be written uniquely as the product of primes; AKA prime factorization
 - Whose prime factors are conventionally written in **increasing** order

Primes

Examples of Prime Factorization:

$$15 = 3 * 5$$

$$48 = 2 * 2 * 2 * 2 * 3$$

$$17 = 17 \text{ AKA } 1 * 17$$

$$100 = 2 * 2 * 5 * 5 = 2^2 * 5^2$$

$$512 = 2 * 2 * 2 * 2 * 2 * 2 * 2 * 2 * 2 = 2^9$$

$$515 = 3 * 103$$

$$28 = 2 * 2 * 7$$

Division Algorithm

- Let **a** be an integer and **d** be another positive integer
- Then there are unique integers **q** and **r**, with $0 \leq r < d$, that:

$$a = d * q + r$$

- In the above equation:
 - **a** is called the dividend
 - **d** is called the divisor
 - **q** is called the quotient, and
 - **r** is called the remainder

Division Algorithm

Example: When we divide 17 by 5, we have

$$17 = 5 * 3 + 2$$

- 17 is the dividend
- 5 is the divisor
- 3 is called the quotient, and
- 2 is called the remainder

Division Algorithm

Another example:

What happens when we divide -11 by 3 ?

Note that a *remainder can never be negative*

$$-11 = 3 * (-4) + 1$$

- -11 is the dividend
- 3 is the divisor
- -4 is called the quotient, and
- 1 is called the remainder

Greatest Common Divisor

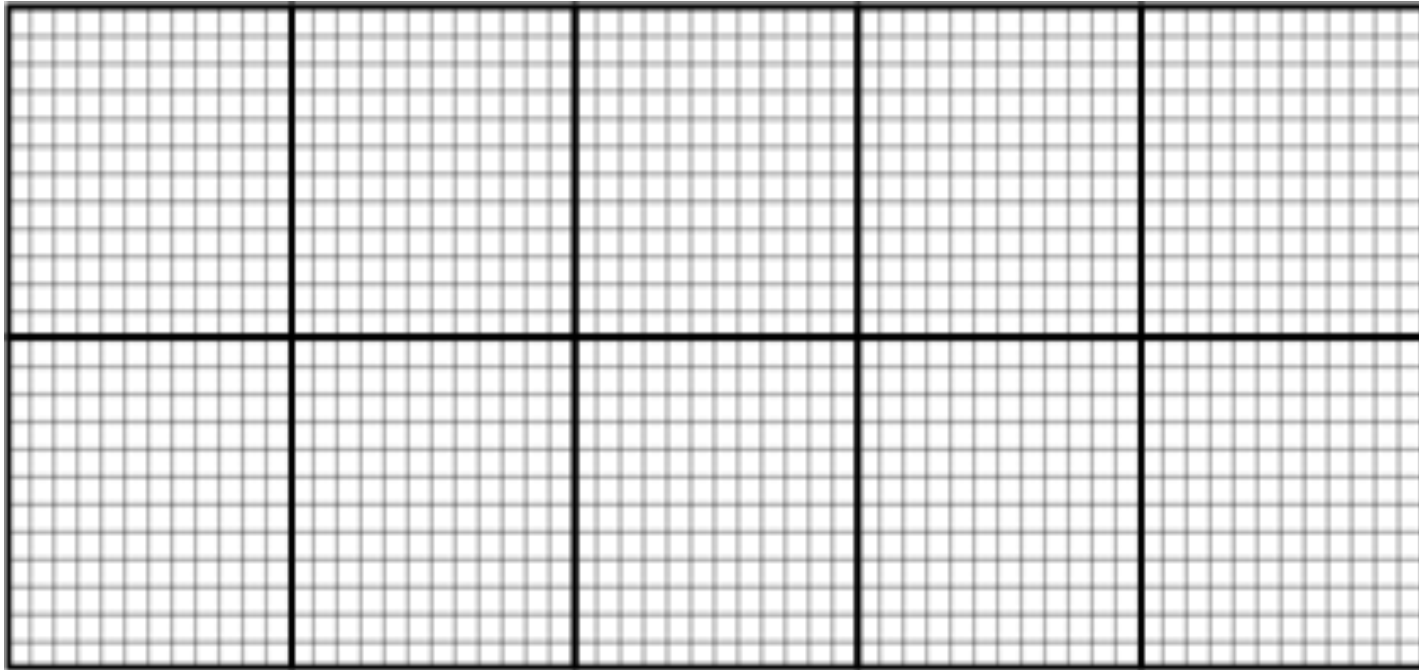
- Let a and b be integers, not both zero: The largest integer d such that $d \mid a$ and $d \mid b$ is called the greatest common divisor of a and b
- Greatest common divisor of a and b : is denoted by $\gcd(a, b)$
- Example 1: What is the $\gcd(48, 72)$?
- The positive common divisors of 48 and 72 are 1, 2, 3, 4, 6, 8, 12, 16, and 24, so $\gcd(48, 72) = 24$
- Example 2: What is $\gcd(19, 72)$?
- The only positive common divisor of 19 and 72 is 1, so $\gcd(19, 72) = 1$

Greatest Common Divisor

- There exists a nice geometric view of the GCD
- For example, a 24-by-60 rectangular area can be divided into a grid of 1-by-1 squares, 2-by-2 squares, 3-by-3 squares, 4-by-4 squares, 6-by-6 squares or 12-by-12 squares
- Thus we see that 12 is the greatest common divisor of 24 and 60!
- A 24-by-60 rectangular area can be divided into a grid of 12-by-12 squares, with two squares along one edge ($24/12 = 2$) and five squares along the other ($60/12 = 5$)
- See below:

Greatest Common Divisor

- Below you see the geometric partition of a $24 * 60$ rectangle partitioned into $60/12 = 5$; 12 is GCD
- 1,440 mini-squares; found at Wiki: [9]



Relatively Prime Integers

Definition: Two integers **a** and **b** are **relatively prime** if $\gcd(a, b) = 1$. Needed for example to define suitable size of hash table in fast compilers

Examples:

- Are 15 and 28 relatively prime?
- **Yes**, $\gcd(15, 28) = 1$
- Are 55 and 28 relatively prime?
- **Yes**, $\gcd(55, 28) = 1$
- Are 35 and 28 relatively prime?
- **No**, $\gcd(35, 28) = 7$

Relatively Prime Integers

- **Definition:** The integers a_1, a_2, \dots, a_n are pairwise relatively prime if $\gcd(a_i, a_j) = 1$ whenever $1 \leq i < j \leq n$
- **Example 1:** Are 15, 17, and 27 pairwise relatively prime?
- **No, because**
 $\gcd(15, 27) = 3$
- **Example 2:** Are 15, 17, and 28 pairwise relatively prime?
- **Yes, because**
 $\gcd(15, 17) = 1$
 $\gcd(15, 28) = 1$
 $\gcd(17, 28) = 1$

Challenge

- **Name the largest prime number!**
- **Or else prove, there exists no largest prime!**

Modular Arithmetic

Let a be an integer and m be a positive integer. Then we denote by: $a \bmod m$ the integer remainder, when a is divided by m

Examples:

$$9 \bmod 4 = 1$$

$$9 \bmod 3 = 0$$

$$9 \bmod 10 = 9$$

$$-13 \bmod 4 = 3$$

-- quotient being 0

-- Note quotient is -4

Congruence

- Let a and b be integers and m be a positive integer. We say that a is congruent to b modulo m if

m divides $(a - b)$

- We use the notation $a \equiv b \pmod{m}$ to indicate that a is congruent to b modulo m
- In other words:
 $a \equiv b \pmod{m}$ if and only if $(a \bmod m) = (b \bmod m)$

Congruence



Congruence

Examples:

- Is $46 \equiv 68 \pmod{11}$?
- Yes, because $11 \mid (46 - 68)$
- Is $46 \equiv 68 \pmod{22}$?
- Yes, because $22 \mid (46 - 68)$
- For which integers z is it true that $z \equiv 12 \pmod{10}$?
- It is true for any $z \in \{..., -28, -18, -8, 2, 12, 22, 32, ... \}$

Theorem: Let m be a positive integer. Then integers a and b are **congruent modulo m** if and only if there is an integer k such that $a = b + k * m$

Congruence

Theorem: Let m be a positive integer

If $a \equiv b \pmod{m}$ and $c \equiv d \pmod{m}$, then:

$a + c \equiv b + d \pmod{m}$ and $a * c \equiv b * d \pmod{m}$

- **Proof:** We know that $a \equiv b \pmod{m}$ and $c \equiv d \pmod{m}$ implies that there are integers s and t with:
 $b = a + s * m$ and $d = c + t * m$
- Therefore,
- $b + d = (a + s * m) + (c + t * m) = (a + c) + m*(s + t)$
and
- $b * d = (a + s * m) * (c + t * m) = a * c + m * (a * t + c * s + s * t * m)$
- Hence, $a + c \equiv b + d \pmod{m}$ and $a * c \equiv b * d \pmod{m}$

Congruence

Theorem: Let m be a positive integer. $a \equiv b \pmod{m}$ iff $a \bmod m = b \bmod m$

Proof:

Let $a = mq_1 + r_1$, and $b = mq_2 + r_2$.

Only if part: $a \bmod m = b \bmod m \rightarrow r_1 = r_2$, therefore

$$a - b = m(q_1 - q_2), \text{ and } a \equiv b \pmod{m}.$$

If part: $a \equiv b \pmod{m}$ implies

$$a - b = mq$$

$$mq_1 + r_1 - (mq_2 + r_2) = mq$$

$$r_1 - r_2 = m(q - q_1 + q_2).$$

Since $0 \leq r_1, r_2 < m$, $0 \leq |r_1 - r_2| < m$. The only multiple in that range is 0

Therefore $r_1 = r_2$, and $(a \bmod m) = (b \bmod m)$

Euclid of Alexandria



323 BC – 283 BC

Euclidean Algorithm

The Euclidean Algorithm finds the greatest common divisor of two integers **a** and **b**

- For example, if we want to find $\text{gcd}(287, 91)$, we divide 287 by 91, and find:

$$287 = 91 * 3 + 14$$

- We know that for integers a, b and c:
if $a \mid b$ and $a \mid c$, then $a \mid (b + c)$
- Therefore, any divisor (including their gcd) of 287 and 91 must also be a divisor of $287 - 91 \cdot 3 = 14$
- Consequently, $\text{gcd}(287, 91) = \text{gcd}(14, 91)$

Euclidean Algorithm

In the next step, we divide 91 by 14:

- $91 = 14 * 6 + 7$
- This means that $\gcd(14, 91) = \gcd(14, 7)$
- So we divide 14 by 7:
$$14 = 7 * 2 + 0$$
- We find that $7 \mid 14$, and thus $\gcd(14, 7) = 7$
- Therefore, $\gcd(287, 91) = 7$

Euclidean Algorithm

In Algol-like pseudo code, the algorithm can be implemented as follows, comments in { }:

```
procedure gcd( a, b: positive integers )  
  x := a  
  y := b  
  while y ≠ 0 do  
    r := x mod y  
    x := y  
    y := r  
  end  
  {x is gcd(a, b ) }  
end
```

Integer Representation Review

Representation of Integers

- Let **b** be a positive integer greater than 1: **base = b**
- Then if **n** is a positive integer, it can be expressed uniquely in the form:

$$n = a_k b^k + a_{k-1} b^{k-1} + \dots + a_1 b^1 + a_0 b^0$$

- where **k** is a nonnegative integer,
- a_0, a_1, \dots, a_k are nonnegative integers less than **b**

Example 859_{10} for base $b = 10$:

$$859 = 8 * 10^2 + 5 * 10^1 + 9 * 10^0$$

Representation of Integers

Example for $b = 2$ (binary expansion):

$$10110_2 = 1 \cdot 2^4 + 0 \cdot 2^3 + 1 \cdot 2^2 + 1 \cdot 2^1 + 0 \cdot 2^0 = 22_{10}$$

Example for base $b=16$ (hexadecimal expansion), using letters **A** to **F** to indicate hex digit values **10** to **15**:

$$3A0F_{16} = \underset{\text{A}}{3} \cdot 16^3 + \underset{\text{F}}{10} \cdot 16^2 + 0 \cdot 16^1 + 15 \cdot 16^0 = 14863_{10}$$

Representation of Integers

- How to construct base b expansion of an integer n ?
- First, divide n by b to obtain a quotient q_0 and remainder a_0 , that is:
$$n = bq_0 + a_0, \text{ where } 0 \leq a_0 < b$$
- The remainder a_0 is the rightmost digit in the base b expansion of n
- Next, divide q_0 by b to obtain:
$$q_0 = bq_1 + a_1, \text{ where } 0 \leq a_1 < b$$
- a_1 is the second digit from right in base b expansion of n . Continue these steps til you obtain a 0 quotient

Representation of Integers

Example: Base 8 expansion of 12345_{10}

First, divide 12345 by 8:

$$12345 = 8 \cdot 1543 + 1$$

$$1543 = 8 \cdot 192 + 7$$

$$192 = 8 \cdot 24 + 0$$

$$24 = 8 \cdot 3 + 0$$

$$3 = 8 \cdot 0 + 3$$

The result is: $12345_{10} = 30071_8$

Adding Integers

How do we (humans) add two integers?

Example:

$$\begin{array}{r} 7583 \\ + 4932 \\ \hline \text{carry: } 11100 \\ \hline 12515 \end{array}$$

Binary Numbers

10001100 1000
10001011 010
10100101 0
10011101 010
1000100101 111 01 001
1000 1001110 1110111 1010
1101 01110101 10010000 0111
000011 01000111 000111
1111001 00111001 110
101000 00010110 00
11010 1000110
010
010110
110
0000000
01100111 00
00110111 001
10111101 1101100
01011111 11010011
01011011 11000101
0111011 0100111
00100 0000101
11000 1001
0001 0011
1111 101
0000 01
1110
1011
010
01
10001100 1000
10001011 010
10100101 0
10011101 001
01
101010 0100011011101111 10010011 100
10110 00111010 10000100 10011001 11010100 10001101 0110011
10001 00 10101010 11101100 11100001 10100100 01001000 01100
11 00 00 01011111 00111001 01100111 11111100 0111 11
00010111 01000011 00011100 11100010 10011100 01100010
10 10110010 10110101 10011001 11001011 01001001 1110001
10100 00011010 1001010 0000010 00100011 00011111 0000100 01
111 10011 10000000 0 100110 10111001 01010010 01100 0
0100 100 0111 00000101 00100011 11010010 100 1100
00100 01001000 00110111 00100001 01000101 01010001 0010
11111100 01111101 0111101 1111001 11001010 11101000 1001
10011100 01100010 0101111 11010 0001100 10001101 011
01001001 11100010 0101111 1100 0101 000011
00011111 00001000 01011011 0
1010010 01100011 00000100 00
010010 10000111 00011000 1
100101 11110001
011110 111011
10101 100100
00111 000111
11001 110010 0
10110 0011 01
1100 1010 0
100 010
111 0
1100010 01011011 11000101 10001011 01000011 01000
011111 00001000 01011011 01001111 10100101 01111001 00
010010 01100011 00000100 00001011 10011100 0010100
01101 01100111 00000101 00100011 11010010 10000111 00011000 10011101 01111011 01011010

Binary Numbers

- **Binary Digit –AKA Bit--** is the smallest unit of computation on digital computers
- **Bit has two states:**
 - **0** represents 0 Volt [V], or ground; used for logical **False**, or for numeric **0**
 - **1** represents positive Voltage [+V]; used for logical **True**, or for numeric **1**
- **Computer word** consists of multiple bits, typically 32, 60, today mostly **64 bits**
- Often words are composed of **bytes**, units of 8 bits that are **addressable as one unit**: byte-addressable
- **Binary number representation works like decimal system**

$$\begin{array}{ccccccc} b_{n-1} & b_{n-2} & \dots & b_1 & b_0 & = & b_{n-1} \times 2^{n-1} + b_{n-2} \times 2^{n-2} + \dots + b_1 \times 2^1 + b_0 \times 2^0 \\ \uparrow & & & \uparrow & & & \\ \text{MSB} & & & \text{LSB} & & & \end{array}$$

Binary Numbers Using TCR

- Possible representations of binary numbers: *sign-magnitude (sm)*, *one's complement (ocr)*, and *two's complement representation (tcr)*
- Advantage of *tcr*: machine needs no subtract unit, the sole adder is sufficient
- When subtraction is needed, just add a negative number
- To create a negative number: invert the positive one; see full method below!
- In TCR, there is no need for signed- *and* unsigned arithmetic; unsigned is sufficient
- C++ allows signed & unsigned integers. In fact, arithmetic units with *tcr* effectively ignore the sign bit
- *Tcr* just needs an **adder**, **inverter**, and **check** for carry

Binary Numbers Using TCR

- Binary numbers in *tcr* use a **sign bit** and a **fixed number of bits** for the **magnitude**
- For example, old PCs have 32-bit integers, 1 bit for the sign, 31 bits for the magnitude
- PCs today use 64-bit integers, also use 1 for the sign
- When processed on *tcr* architecture, the most significant bit (MSB) is the **sign bit**, the other 63 bits hold the actual signed value of interest (AKA **magnitude**); usually in **two's complement**
- By convention, the sign bit **0** stands for **positive** and **1** for **negative** numbers

Binary Numbers Using TCR

- **Inverting positive numbers:** To create a negative number from a positive in *tcr*, start with the binary representation for the positive one, invert all bits, *and add 1*
- Note that overflow cannot happen by inversion alone: the negative value of positive numbers can always be created
- But **there is one more negative number** in *tcr* than positive numbers!
- There exists only **one single 0**, i.e. **no negative 0** in *tcr*, as we find it in *one's complement* and *sign-magnitude* representations

Binary Numbers Using TCR

- **Inverting negative numbers:** To invert a negative number, complement all bits of the original negative number and **add a 1** in *tcr*, do not subtract it!
- However, there will be one negative value, whose positive inverse cannot be represented; it will cause **overflow** instead!
- That value is the smallest, negative number. For example, an **8-bit signed *tcr* integer** can hold integers in the range from -128 .. 127. See the **asymmetry**? See the one negative value that cannot be inverted?
- On a 32-bit architecture, the range is from -2,147,483,648 to +2,147,483,647; similar for 64-bits
- Note the **asymmetry** of the numeric range!

Hexadecimal Numbers

- Hexadecimal (hex) numbers are simply numbers with a **base 16**; not 10, not 2, just 16, no magic 😊
- They have 16 different digits, 0..9 and, purely by convention, the digits **a . . f**
- Symbol 'a' or 'A' stands for some hex digit with value **10₁₀**, while the symbol 'f' stands for the value **15₁₀**
- Programming tools are not picky; they permit 6 extra digits to be lower- as well as uppercase letters 😊
- Here are a few hex numbers and their equivalent decimal values:

Hexadecimal Numbers

Decimal	Hexadecimal
0	0
1	1
9	9
10	a
11	b
15	f
16	10
33	21
127	7f
128	80
129	81
255	ff
256	100

Decimal	Hexadecimal
257	101
258	102
300	12c
16,383	3fff
16,384	4000
16,385	4001
32,767	7fff
32,768	8000
32,769	8001
65,535	ffff
65,536	1,0000
65,637	1,0001
4,294,967,295	ffff,ffff

Adding Hexadecimal Numbers

Adding Hexadecimal Numbers, $af_{16} + 65_{16}$, $10a42_{16} + 5be_{16}$

			a	f
			6	5
		1	1	4

1	0	a	4	2
		5	b	e
1	1	0	0	0

Building Graphs

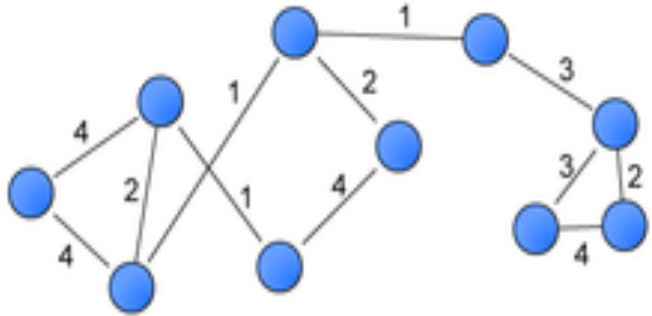
Formal Definition of Graph

- **Graph**: A graph G is a data structure $G = \{ V, E \}$ consisting of a set E of edges and a set of V vertices, AKA **nodes**. Any node $v_i \in V$ may be connected to any other node v_j . Such a connection is called an **edge**. Edges may be **directed**, even **bi-directed**
- Different from a tree, a node in G may have any number of predecessors –or incident edges: **THE main difference** between graph and tree!
- **Empty Graph**: For expediency we ignore the possibility of a graph G being empty; in an empty graph the data structure that points the graph is simply **NIL**
- **Connected Graph**: If all $n > 0$ nodes v_n in G are connected somehow, the graph G is called connected, regardless of edge directions

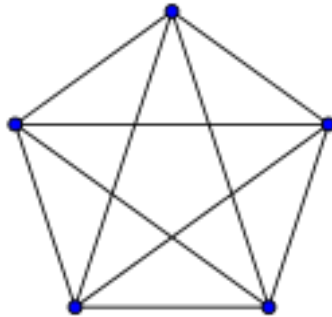
Formal Definition of Graph

- **Strongly Connected Component (SCC)**: A subset $SG \subseteq G$ is strongly connected, if for every node v_i in SG , $i > 0$, such a v_i can reach all v_j nodes in SG
- **Directed Acyclic Graph (DAG)**: A DAG is a graph with directed edges but **no cycles**. A node may still have multiple predecessors and/or successors
- When programming graphs, it is **convenient** to add fields to the node type for auxiliary functions; e.g. it is possible to process all nodes in a linear fashion by adding a link field, often called a “**finger**” or “**link**”
- Possible use: traversing all nodes in G , though **G may be disconnected!**

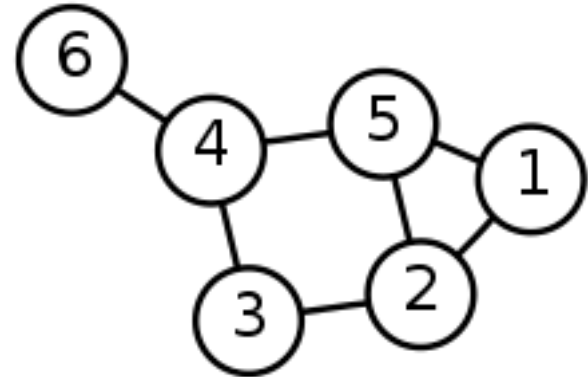
Sample Graphs



Weighted graph
with 10 vertices
and 12 **weighted**
edges



Complete graph
with 5 vertices
and all possible
10 edges

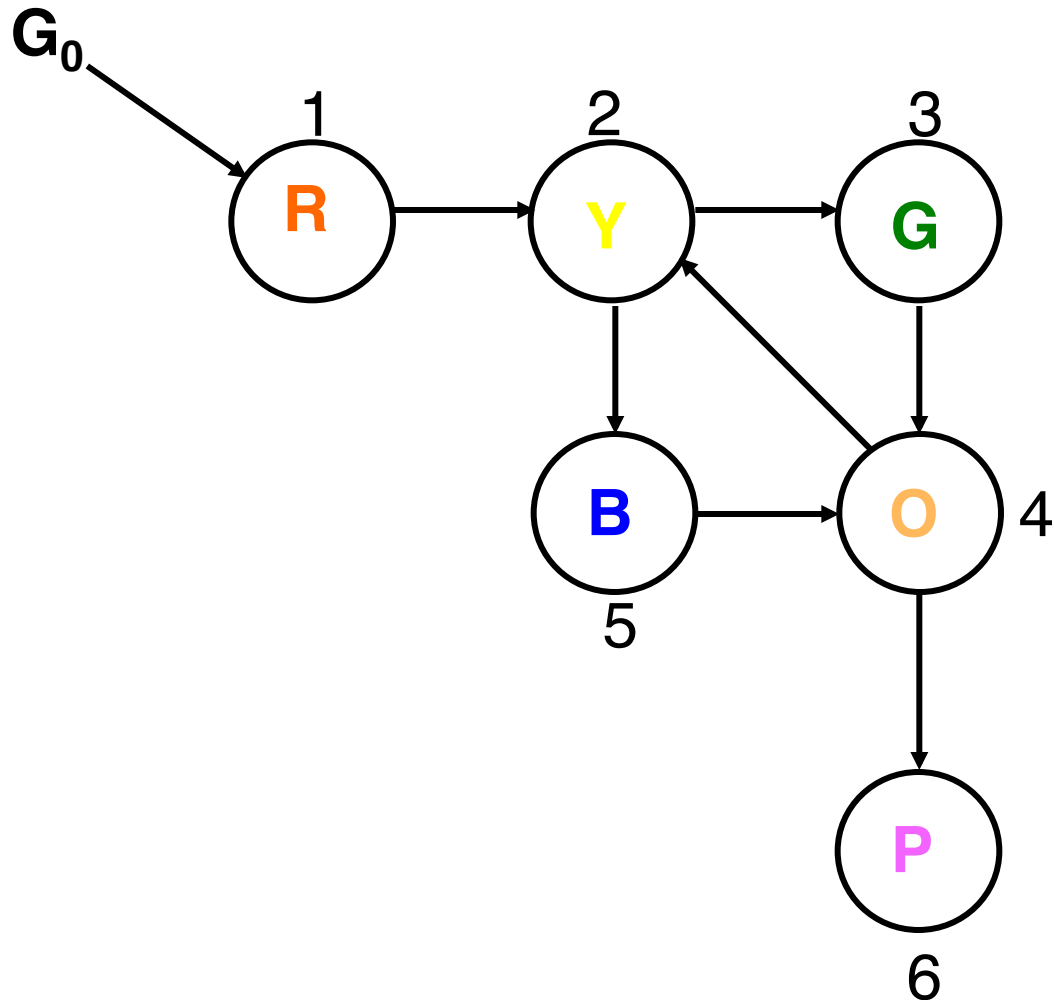


Sample graph
with 6 **named**
vertices and 7
edges

Graph Data Structure

- A graph $G(v, e)$ consists of nodes v and edges e
 - Implemented via some `node_type` data structure
- G often identified and accessible via one select node, called *entry node*, or simply entry, AKA *head*
- Or identified by a `pointer to node_type`; if NIL, the graph is empty
- **Caveat: G is not necessarily connected**
 - If parts of G are unconnected, how can they be retrieved in case of a necessary, complete graph traversal?
- **Several methods of forcing complete traversal:**
 - Either create a super-node, not part of G proper, in a way that each unconnected region is pointed at, thus reachable
 - Or have a linked-list (LL) meandering through each node of G , without this `link field` being part of G proper; e.g. “finger”

Sample Graph G_0



Graph Data Structure, Cont'd

- Sample Graph G_0 above has 6 nodes
- The ID of a node, AKA *name* of each node is shown next to the nodes, e.g. 1, 2, 3, 4, 5, 6
- The graph's node type data structure includes such name information as part of struct *node_type*
- In addition, each node in G_0 has *attributes*, such as **R**, **G**, **Y** etc. in the sample above; such attributes can be arbitrarily complex consisting of many fields
- There may be more attributes belonging to each node, depending on what the graph is used for
- Any of these attributes must also be **declared** in the *node_type* data structure
- Successors, if any, of each node must be encoded in the node; there is no upper limit on the number!
- G_0 has **3 SCCs**; 2 of those trivial, i.e. not interesting!

Graph Data Structure

- There is no upper bound on number of successor nodes; suitable way to define successors: via linked list of **tuples**, AKA **link** nodes
- Thus a possible data type for successor nodes is: **pointer to a link node**
- Link nodes can be allocated off the heap, as needed; they are not of type **node_type**, but of type **link_type**
- And each link is a tuple, i.e. consists of just 2 fields:
 - One field pointing to the next link; the type is **pointer to link_type**, in some languages expressed as ***link_type**
 - The other field pointing to the successor node; the type is **pointer to node_type**
- For convenience, the last link inserted is added at the head of the list, simplifying searches for the list end

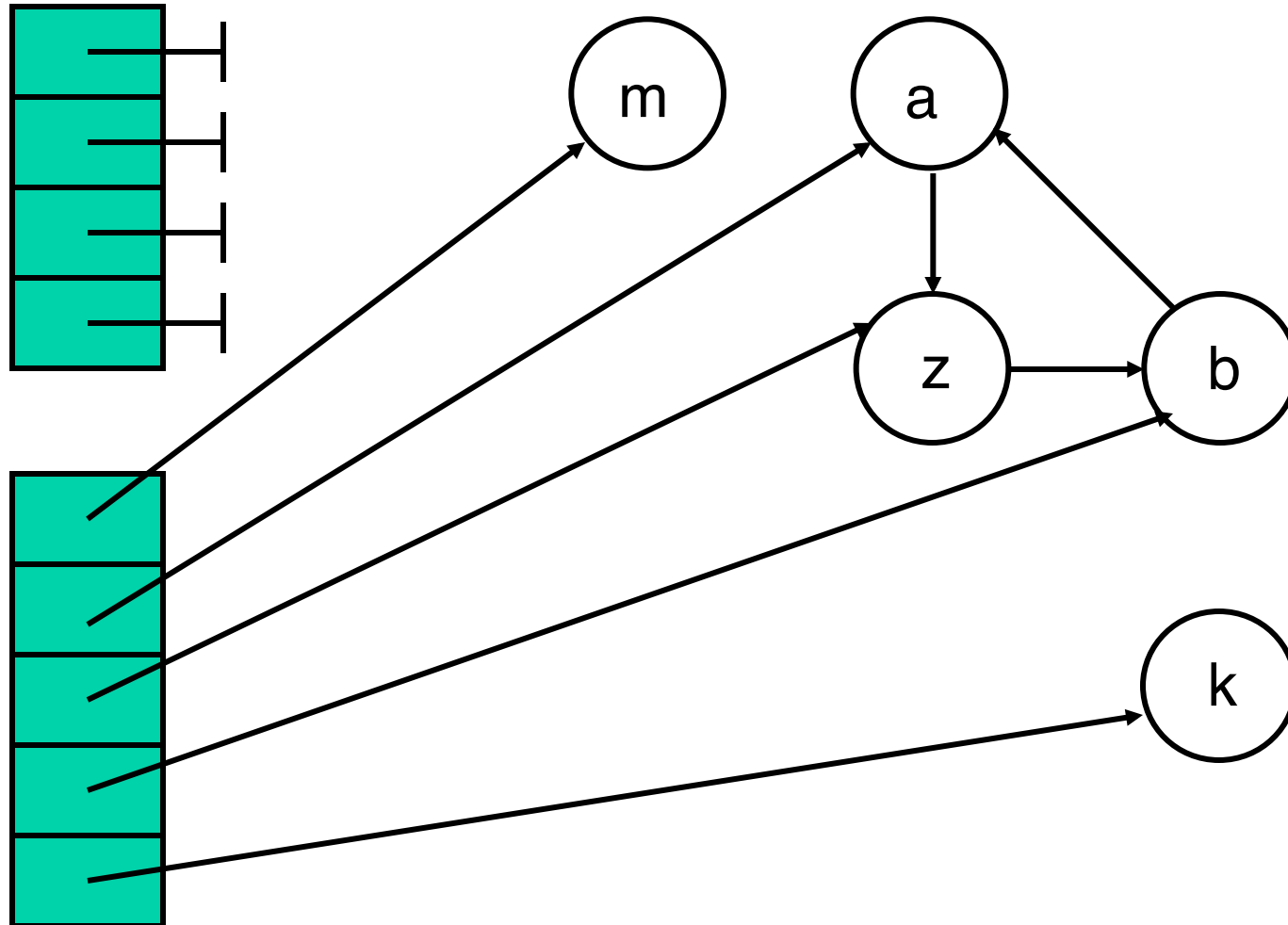
Graph Traversal

- Graph G with i nodes may be unconnected, and yet each unconnected region is part of G
- An algorithm may **require visiting each** node $n_i \in G$
- Requires additional data structure to guarantee a possible visit to node n_i
- Moreover, any full traversal of G must start with some node of G ; which one?

1. Graph Traversal: Fixed Array

- Specify some static array $a[k]$ of k node pointers, all initially **null**, not necessarily all used, in which each used element $a[i]$, $i < k$ points to node n_i
- Say, in increasing order of indices, starting at index 0
- As soon as $a[i]$ yields a null pointer, no more nodes are identified in $a[]$
- Thus G is known to have i nodes; or $k = i$
- **Advantage:**
 - Simple to implement
- **Disadvantage:**
 - **Not recommended!**
 - Wasted space almost always, since $a[]$ needs to be large
 - Too limited, the method fails if $i \geq k$

1. Graph Traversal Fixed Array



2. Graph Traversal Linked List

- Or use separate **linked list** of links w. pointer to nodes
- Each link element has 2 fields
 - Field **next** points to next link element, if any; ends with NIL
 - Field **finger** points to corresponding node $n_i \in G$
- Link element is allocated dynamically off the heap only when another node n_i in G is added
- Advantage: simple to implement
 - Space consumption **directly proportional** to the number of nodes in G ; thus no wasted static space
 - Runs out of space only when all memory space is exhausted
- Disadvantage:
 - Is data structure separate from actual graph G , ends up with 2 data structures to be synchronized
 - **Also not recommended**

3. Graph Traversal Added Field

- Extend the data structure of the graph G proper!
- Keep all fields needed for the graph, but add a **finger** field to each node, i.e. field of type **pointer to node**
- Traversing G requires an outside data structure, of type **pointer to node**, initially nil, to point to one select **start node** of the nodes $n_i \in G$
- Thus **finger** fields form a linked list of graph nodes
- **Advantage:**
 - Simple to implement
 - Space consumption proportional to number of nodes in G
 - Runs out of space only when all memory space is exhausted
- **Disadvantage:**
 - One more field, but the cost is contained!
 - You never get something for nothing, unless previously ... ☺

Graph Data Structure, Link

```
// node may have any number of successors, incl. 0
// all must be retrievable
// Hence each node in G has a link pointer,
// pointing to LL of all successor nodes
// Last node connected will be inserted at head

typedef struct link_tp * link_ptr_tp; // forward ref!
typedef struct node_tp * node_ptr_tp; // forward ref!

typedef struct link_tp
{
    link_ptr_tp next_link; // point to next link
    node_ptr_tp next_node; // point to successor node
} str_link_tp;

#define LINK_SIZE sizeof( str_link_tp )
```


Graph Data Structure, Node

```
// "name" is some unique ID given during creation
// Could use another way of "naming" nodes
// "link" is head of LList of successor nodes, while
// "finger" is direct linear link through all nodes
// "visited" only true after visit; initially FALSE

typedef struct node_tp
{
    link_ptr_tp link;    // points to LL of successors
    node_ptr_tp finger;  // finger through all nodes
    int         name;    // name=ID given at creation
    bool        visited; // to check traversal
    others ...           // other fields: attributes
} str_node_tp;

#define NODE_SIZE sizeof( str_node_tp )
```

Building a Graph, one Node

```
// create a node in graph G, identified by "name"
// connect to the global "finger" at head of LList
// assumption: no such node "name" exists in graph
// assume: global "finger" pointer, initially NULL

node_ptr_tp make_node( int name )
{ // make_node
    node_ptr_tp node = (node_ptr_tp) malloc( NODE_SIZE );

    // check once non-Null, not repeatedly on user side!
    ASSERT( node, "space for node missing" );
    node->finger      = finger;    // set: global finger!!
    node->link        = NULL;      // no successors yet
    node->name        = name;      // IDs this node
    node->visited      = FALSE;    // initially
    finger            = node;      // now link to "this"
    return node;
} //end make_node
```

Building a Graph from Linked Pairs

```
// input is list of pairs, each element being a node name
// craft edge from first node name=a to second node name=b;
// If a node is new: create it; else use ptr = exists()
while( scanf( "%d%d", &a, &b ) ) {           // a, b are ints
    if( ! ( first = exists( a ) ) ) {        // 'a' new node?
        first = make_node( a );              // allocate 'a'
    } //end if
    if( ! ( second = exists( b ) ) ) {       // 'b' new node?
        second = make_node( b );             // allocate 'b'
    } //end if
    // both exist. Either created, or pre-existed: Connect!
    if( new_link( first, second ) ) {
        link = make_link( first->link, second );
        ASSERT( link, "no space for link node" );
        first->link = link;
    }else{
        // link was there already, no need to add again!
        printf( "<><> skip duplicate link %d->%d\n", a, b );
    } //end if
} //end while
```

Building a Graph from Linked Pairs

```
// check, whether link between 2 nodes already exists
// if not, return true: New! Else return false, NOT new!
bool new_link( node_ptr_tp first, node_ptr_tp second )
{ // new_link
    int target          = second->name;
    link_ptr_tp link = first->link;

    while( link ) {
        if( target == link->next_node->name ) {
            return FALSE; // it is an existing link, NOT new
        } //end if
        // check next node; if any
        link = link->next_link;
    } //end while
    // none of successors equal the second node's name
    return TRUE;          // is a new link
} //end new_link
```

Building a Graph

- Any graph node may have 0 or more successors
- Even if some nodes have 0 successor nodes, graph must be **traversable**
- Needs extra data structure to link all nodes together
- If a link exists already, no need to duplicate
- A separate presentation later discusses detail of graph analysis and graph construction

Summary

- Defined **algorithm** as a finite sequence of discrete steps to perform computations
- Discussed algorithm, computational complexity, prime numbers and Euclid's method and other to test for primality (sic!)
- Introduced data structure **graph**, and representation in programming languages

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