

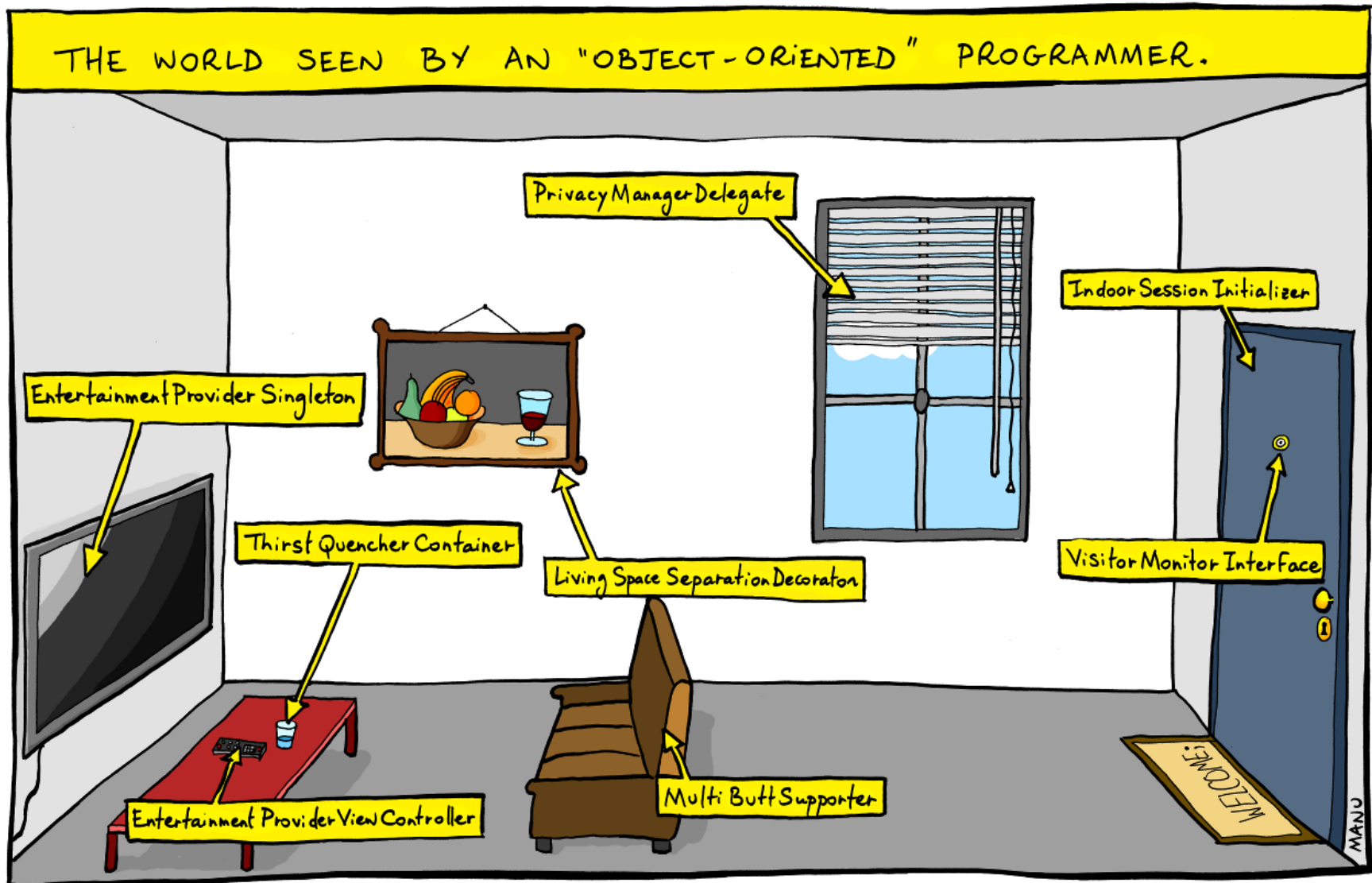
CSCI 104

Rafael Ferreira da Silva

rafsilva@isi.edu

Slides adapted from: Mark Redekopp and David Kempe

THE WORLD SEEN BY AN "OBJECT-ORIENTED" PROGRAMMER.



CLASSES (CONT.)

Composite Objects

- Fun Fact: Memory for an object comes alive before the code for the constructor starts at the first curly brace '{'

```
#include <string>
#include <vector>
using namespace std;

struct Student
{
    string name;
    int id;
    vector<double> scores;
    // say I want 10 test scores per student

    Student() /* mem allocated here */
    {
        // Can I do this to init. members?
        name("Tommy Trojan");
        id = 12313;
        scores(10);
    }
};

int main()
{
    Student s1;
}
```

string name
int id
scores

Composite Objects

- You cannot call constructors on data members once the constructor has started (i.e. passed the open curly '{')
 - So what can we do??? Use initialization lists!

```
#include <string>
#include <vector>
using namespace std;

struct Student
{
    string name;
    int id;
    vector<double> scores;
    // say I want 10 test scores per student

    Student() /* mem allocated here */
    {
        // Can I do this to init. members?
        name("Tommy Trojan");
        id = 12313;
        scores(10);
    }
};

int main()
{
    Student s1;
}
```

string name
int id
scores

This would be
"constructing"
name twice. It's
too late to do it in
the {...}

Constructor Initialization Lists

```
Student::Student()  
{  
    name = "Tommy Trojan";  
    id = 12313  
    scores.resize(10);  
}
```

If you write this...

```
Student::Student() :  
    name(), id(), scores()  
    // calls to default constructors  
{  
    name = "Tommy Trojan";  
    id = 12313  
    scores.resize(10);  
}
```

The compiler will still generate this.

- Though you do not see it, realize that the **default constructors** are implicitly called for each data member before entering the {...}
- You can then assign values but this is a **2-step** process

Constructor Initialization Lists

```
Student:: Student() /* mem allocated here */  
{  
    name("Tommy Trojan");  
    id = 12313;  
    scores(10);  
}
```

**You can't call member
constructors in the {...}**

```
Student::Student() :  
    name("Tommy"), id(12313), scores(10)  
{ }
```

**You would have to call the member
constructors in the initialization list context**

- Rather than writing many assignment statements we can use a special initialization list technique for C++ constructors
 - Constructor(param_list) : member1(param/val), ..., memberN(param/val)
{ ... }
- We are really calling the respective constructors for each data member

Constructor Initialization Lists

```
Student::Student()  
{  
    name = "Tommy Trojan";  
    id = 12313  
    scores.resize(10);  
}
```

You can still assign data
members in the {...}

```
Student::Student() :  
    name(), id(), scores()  
    // calls to default constructors  
{  
    name = "Tommy Trojan";  
    id = 12313  
    scores.resize(10);  
}
```

But any member not in the initialization list will
have its default constructor invoked before the
{...}

- You can still assign values in the constructor but realize that the default constructors will have been called already
- So generally if you know what value you want to assign a data member it's good practice to do it in the initialization list

Member Functions

- Have access to all member variables of class
- **Use “const” keyword if it won't change member data**
- Normal member access uses dot (.) operator
- Pointer member access uses arrow (->) operator

```
class Item
{ int val;
  public:
    void foo();
    void bar() const;
};

void Item::foo() // not Foo()
{ val = 5; }

void Item::bar() const
{ }

int main()
{
    Item x;
    x.foo();
    Item *y = &x;
    (*y).bar();
    y->bar(); // equivalent
    return 0;
}
```

Exercises

- `cpp/cs104/classes/const_members`
- `cpp/cs104/classes/const_members2`
- `cpp/cs104/classes/const_return`

C++ Classes: Other Notes

- Classes are generally split across two files
 - ClassName.h – Contains interface description
 - ClassName.cpp – Contains implementation details
- Make sure you remember to prevent multiple inclusion errors with your header file by using `#ifndef`, `#define`, and `#endif`

```
#ifndef CLASSNAME_H  
#define CLASSNAME_H  
class ClassName { ... };
```

```
#endif
```

```
#ifndef ITEM_H  
#define ITEM_H  
  
class Item  
{ int val;  
public:  
    void foo();  
    void bar() const;  
};  
  
#endif
```

item.h

```
#include "item.h"  
  
void Item::foo()  
{ val = 5; }  
  
void Item::bar() const  
{ }
```

item.cpp

CONDITIONAL COMPILATION

Multiple Inclusion

- Often separate files may #include's of the same header file
- This may cause compiling errors when a duplicate declaration is encountered
 - See example
- Would like a way to include only once and if another attempt to include is encountered, ignore it

```
class string{  
... };
```

string.h

```
#include "string.h"  
class Widget{  
public:  
    string s;  
};
```

widget.h

```
#include "string.h"  
#include "widget.h"  
int main()  
{ }
```

main.cpp

```
class string { // inc. from string.h  
};  
  
class string{ // inc. from widget.h  
};  
class Widget{  
... }  
int main()  
{ }
```

main.cpp after preprocessing

Conditional Compiler Directives

- Compiler directives start with '#'
 - #define XXX
 - Sets a flag named XXX in the compiler
 - #ifdef, #ifndef XXX ... #endif
 - Continue compiling code below until #endif, if XXX is (is not) defined
- Encapsulate header declarations inside a
 - #ifndef XX
 - #define XX
 - ...
 - #endif

```
#ifndef STRING_H
#define STRING_H
class string{
... };
#endif
```

String.h

```
#include "string.h"
class Widget{
public:
  string s;
};
```

Character.h

```
#include "string.h"
#include "string.h"
```

main.cpp

```
class string{ // inc. from string.h
};

class Widget{ // inc. from widget.h
...
}
```

main.cpp after preprocessing

Conditional Compilation

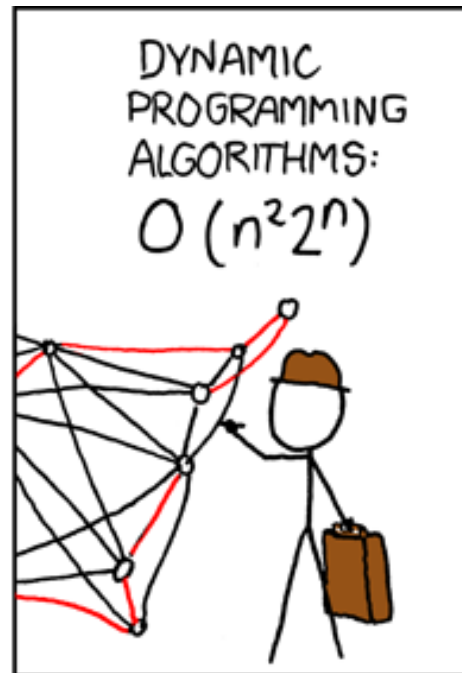
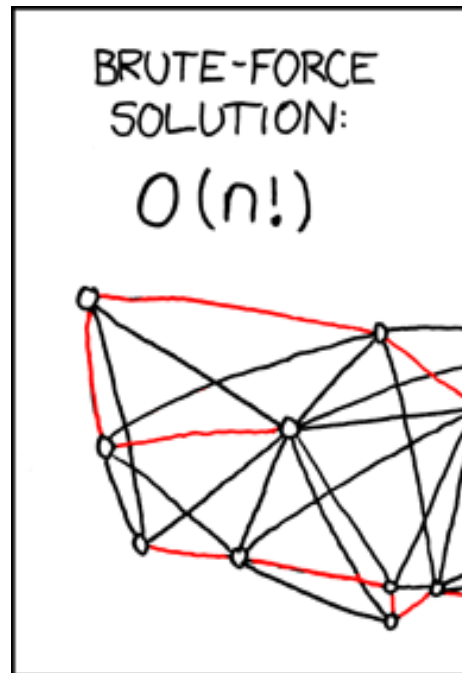
- Often used to compile additional DEBUG code
 - Place code that is only needed for debugging and that you would not want to execute in a release version
- Place code in an `#ifdef XX...#endif` bracket
- Compiler will only compile if a `#define XX` is found
- Can specify `#define` in:
 - source code
 - At compiler command line with `(-Dxx)` flag
 - `g++ -o stuff -DDEBUG stuff.cpp`

```
int main()
{
    int x, sum=0, data[10];
    ...
    for(int i=0; i < 10; i++){
        sum += data[i];
#ifdef DEBUG
        cout << "Current sum is ";
        cout << sum << endl;
#endif
    }

    cout << "Total sum is ";
    cout << sum << endl;
}
```

stuff.cpp

```
$ g++ -o stuff -DDEBUG stuff.cpp
```



XKCD #399

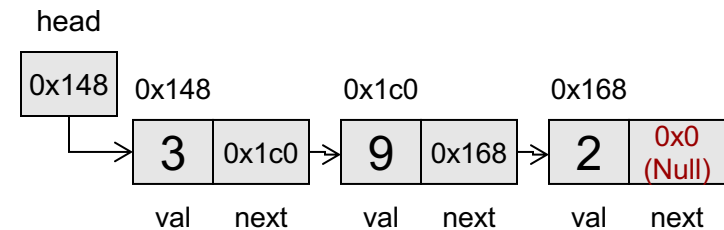
RUNTIME ANALYSIS

Runtime

- It is hard to compare the run time of an algorithm on actual hardware
 - Time may vary based on speed of the HW, etc.
 - The same program may take 1 sec. on your laptop but 0.5 second on a high performance server
- If we want to compare 2 algorithms that perform the same task we could try to count operations (regardless of how fast the operation can execute on given hardware)...
 - But what is an operation?
 - How many operations is: `i++` ?
 - `i++` actually requires grabbing the value of `i` from memory and bringing it to the processor, then adding 1, then putting it back in memory. Should that be 3 operations or 1?
 - Its painful to count 'exact' numbers of operations
- Big-O, Big-Ω, and Θ notation allows us to be more general (or "sloppy" as you may prefer)

Complexity Analysis

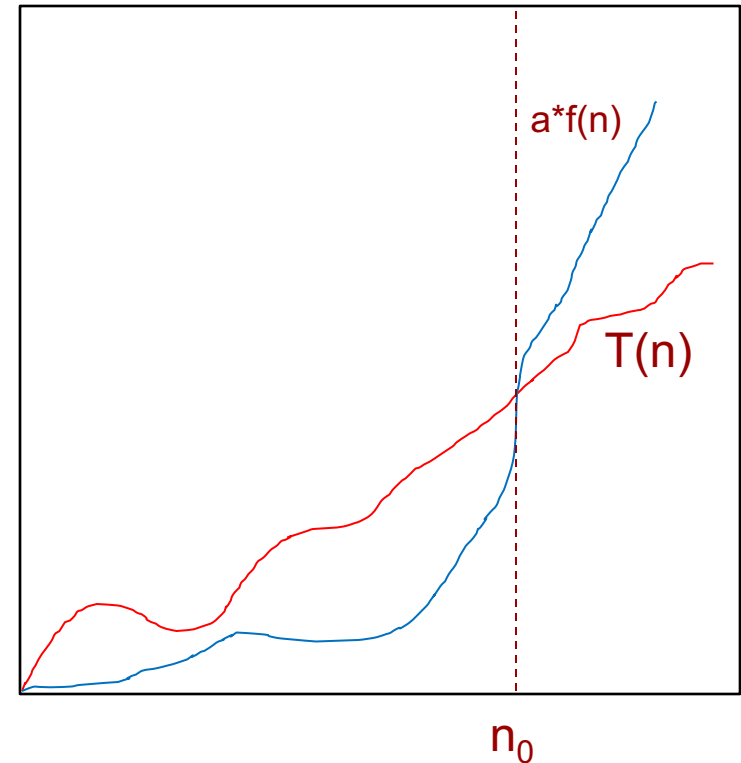
- To find upper or lower bounds on the complexity, we must consider the set of all possible inputs, I , of size, n
- Derive an expression, $T(n)$, in terms of the input size, n , for the number of operations/steps that are required to solve the problem of a given input, i
 - Some algorithms depend on i and n
 - Find(3) in the list shown vs. Find(2)
 - Others just depend on n
 - Push_back / Append
- Which inputs though?
 - Best, worst, or "typical/average" case?
- We will always apply it to the "worst case"
 - That's usually what people care about



Note: Running time is not just based on an algorithm, BUT algorithm + input data

Big-O, Big- Ω

- $T(n)$ is said to be $O(f(n))$ if...
 - $T(n) < a \cdot f(n)$ for $n > n_0$ (where a and n_0 are constants)
 - Essentially an upper-bound
 - We'll focus on big-O for the **worst case**
- $T(n)$ is said to be $\Omega(f(n))$ if...
 - $T(n) > a \cdot f(n)$ for $n > n_0$ (where a and n_0 are constants)
 - Essentially a lower-bound
- $T(n)$ is said to be $\Theta(f(n))$ if...
 - $T(n)$ is both $O(f(n))$ AND $\Omega(f(n))$



Worst Case and Big- Ω

- What's the lower bound on List::find(val)
 - Is it $\Omega(1)$ since we might find the given value on the first element?
 - Well it could be if we are finding a lower bound on the 'best case'
- Big- Ω does **NOT** have to be ***synonymous*** with 'best case'
 - Though many times it mistakenly is
- You can have:
 - Big-O for the best, average, worst cases
 - Big- Ω for the best, average, worst cases
 - Big- Θ for the best, average, worst cases

Worst Case and Big- Ω

- The key idea is an algorithm may perform differently for different input cases
 - Imagine an algorithm that processes an array of size n but depends on what data is in the array
- Big- O for the **worst-case** says **ALL** possible inputs are bound by $O(f(n))$
 - Every possible combination of data is at MOST bound by $O(f(n))$
- Big- Ω for the **worst-case** is attempting to establish a lower bound (at-least) for the worst case (the worst case is just one of the possible input scenarios)
 - If we look at the first data combination in the array and it takes n steps then we can say the algorithm is $\Omega(n)$.
 - Now we look at the next data combination in the array and the algorithm takes $n^{1.5}$. We can now say worst case is $\Omega(n^{1.5})$.
- To arrive at $\Omega(f(n))$ for the **worst-case** requires you simply to find **AN** input case (i.e. the worst case) that requires **at least** $f(n)$ steps
- Cost analogy...

Deriving $T(n)$

- Derive an expression, $T(n)$, in terms of the input size for the number of operations/steps that are required to solve a problem
- If is true $\Rightarrow 4$
- Else if is true $\Rightarrow 5$
- Worst case $\Rightarrow T(n) = 5$

```
#include <iostream>

using namespace std;

int main()
{
    int i = 0;           1
    x = 5;               1

    if(i < x){           1
        x--;            1
    }
    else if(i > x){      1
        x += 2;         1
    }
    return 0;
}
```

Deriving T(n)

- Since loops repeat you have to take the sum of the steps that get executed over all iterations
- $T(n) =$
- $= \sum_{i=0}^{n-1} 5 = 5 * n$
- Or you can setup a relationship like:
- $T(n) = T(n - 1) + 5$
- $= T(n - 2) + 5 + 5$
- $= \sum_{i=0}^{n-1} 5 = 5 * n$
- $= \sum_{i=0}^{n-1} O(1) = O(n)$

```
#include <iostream>
using namespace std;

int main()
{
    for(int i=0; i < N; i++){
        x = 5;
        if(i < x){
            x--;
        }
        else if(i > x){
            x += 2;
        }
    }
    return 0;
}
```

Common Summations

- $\sum_{i=1}^n i = \frac{n(n+1)}{2} = \theta(n^2)$
 - This is called the arithmetic series
- $\sum_{i=1}^n \theta(i^p) = \theta(n^{p+1})$
 - This is a general form of the arithmetic series
- $\sum_{i=1}^n c^i = \frac{c^{n+1}-1}{c-1} = \theta(c^n)$
 - This is called the geometric series
- $\sum_{i=1}^n \frac{1}{i} = \theta(\log n)$
 - This is called the harmonic series

Skills You Should Gain

- To solve these running time problems try to break the problem into 2 parts:
- FIRST, setup the expression (or recurrence relationship) for the number of operations
- SECOND, solve
 - Unwind the recurrence relationship
 - Develop a series summation
 - Solve the series summation

Loops

- Derive an expression, $T(n)$, in terms of the input size for the number of operations/steps that are required to solve a problem
- $T(n) =$
- $= \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \theta(1) = \sum_{i=0}^{n-1} \theta(n) = \Theta(n^2)$

```
#include <iostream>

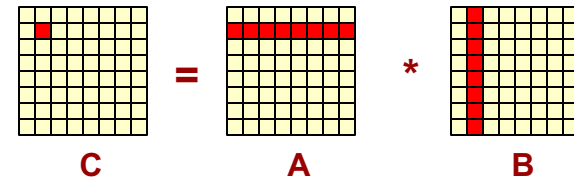
using namespace std;
const int n = 256;
unsigned char image[n][n]
int main()
{
    for(int i=0; i < n; i++){
        for(int j=0; j < n; j++){
            image[i][j] = 0;
        }
    }
    return 0;
}
```

Matrix Multiply

- Derive an expression, $T(n)$, in terms of the input size for the number of operations/steps that are required to solve a problem

- $T(n) =$

- $= \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \sum_{k=0}^{n-1} \theta(1) = \theta(n^3)$



Traditional Multiply

```
#include <iostream>
using namespace std;
const int n = 256;
int a[n][n], b[n][n], c[n][n];
int main()
{
    for(int i=0; i < n; i++){
        for(int j=0; j < n; j++){
            c[i][j] = 0;
            for(int k=0; k < n; k++){
                c[i][j] += a[i][k]*b[k][j];
            }
        }
    }
    return 0;
}
```

Sequential Loops

- Is this also n^3 ?
- No!
 - 3 for loops, but not nested
 - $O(n) + O(n) + O(n) = 3 * O(n) = O(n)$

```
#include <iostream>

using namespace std;
const int n = 256;
unsigned char image[n][n]
int main()
{
    for(int i=0; i < n; i++){
        image[0][i] = 5;
    }
    for(int j=0; j < n; j++){
        image[1][j] = 5;
    }
    for(int k=0; k < n; k++){
        image[2][k] = 5;
    }
    return 0;
}
```

Counting Steps

- It may seem like you can just look for nested loops and then raise n to that power
 - 2 nested for loops $\Rightarrow O(n^2)$
- But be careful!!
- You have to count steps
 - Look at the update statement
 - Outer loop increments by 1 each time so it will iterate N times
 - Inner loop updates by dividing x in half each iteration?
 - After 1st iteration $\Rightarrow x=n/2$
 - After 2nd iteration $\Rightarrow x=n/4$
 - After 3rd iteration $\Rightarrow x=n/8$
 - Say k^{th} iteration is last $\Rightarrow x = n/2^k = 1$
 - Solve for k
 - $k = \log_2(n)$ iterations
 - $O(n \cdot \log(n))$

```
#include <iostream>
using namespace std;
const int n = 256;

int main()
{
    for(int i=0; i < n; i++){
        int y=0;
        for(int x=n; x != 1; x=x/2){
            y++;
        }
        cout << y << endl;
    }
    return 0;
}
```

Analyze This

- Count the steps of this example?

```
for (int i = 0; i <= log2(n); i++)  
    for (int j=0; j < (int) pow(2,i); j++)  
        cout << j;
```

- $\sum_{i=0}^{\lg(n)} \sum_{j=0}^{2^i} 1$
- $= \sum_{i=0}^{\lg(n)} 2^i$
- Use the geometric sum eqn.
- $= \sum_{i=0}^{n-1} a^i = \frac{1-a^n}{1-a}$
- So our answer is...
- $\frac{1-2^{\lg(n)+1}}{1-2} = \frac{1-2*n}{-1} = O(n)$

Another Example

- Count steps here...
 - Think about how many times if statement will evaluate true
- $T(n) = \sum_{i=0}^{n-1} (\theta(1) + O(n))$
- $T(n) =$

```
for (int i = 0; i < n; i++)  
{  
    cout << "i: ";  
    int m = sqrt(n);  
    if( i % m == 0){  
        for (int j=0; j < n; j++)  
            cout << j << " ";  
    }  
    cout << endl;  
}
```

Another Example

- Count steps here...
 - Think about how many times if statement will evaluate true

```
for (int i = 0; i < n; i++)  
{  
    cout << "i: ";  
    int m = sqrt(n);  
    if( i % m == 0){  
        for (int j=0; j < n; j++)  
            cout << j << " ";  
    }  
    cout << endl;  
}
```

- $T(n) = \sum_{i=0}^{n-1} (\theta(1) + O(n))$
- $T(n) = \sum_{i=0}^{n-1} \theta(1) + \sum_{k=1}^{\sqrt{n}} \sum_{j=1}^n \theta(1)$
- $T(n) = \theta(n) + \sum_{k=1}^{\sqrt{n}} \theta(n)$
- $T(n) = \theta(n) + \theta(n \cdot \sqrt{n})$
- $T(n) = \theta(n^{3/2})$

What about Recursion

- Assume N items in the linked list
- $T(n) = 1 + T(n-1)$
- $= 1 + 1 + T(n-2)$
- $= 1 + 1 + 1 + T(n-3)$
- $= n = O(n)$

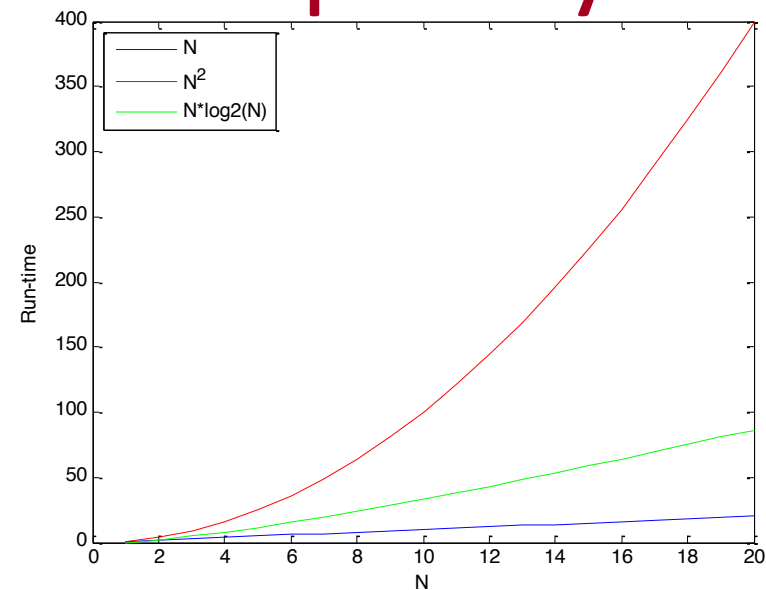
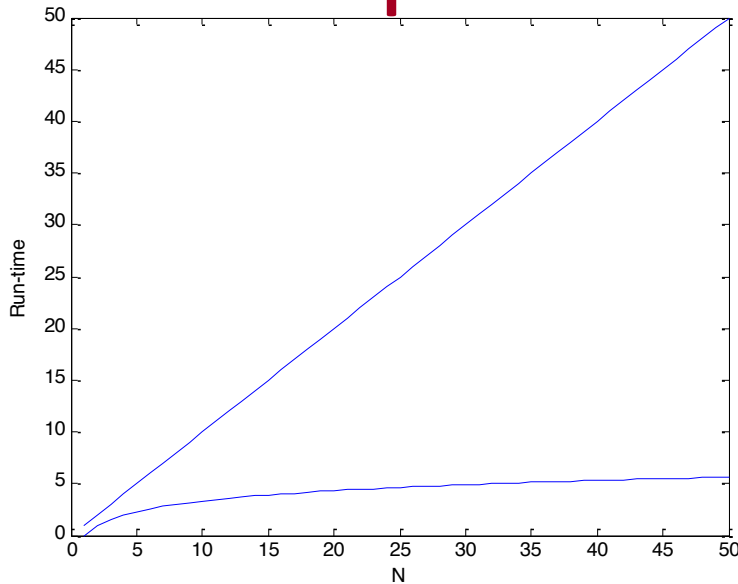
```
void print(Item* head)
{
    if(head==NULL) return;
    else {
        cout << head->val << endl;
        print(head->next);
    }
}
```

Binary Search

- Assume N items in the data array
- $T(n) =$
 - $O(1)$ if base case
 - $O(1) + T(n/2)$
- $= 1 + T(n/2)$
- $= 1 + 1 + T(n/4)$
- $= k + T(n/2^k)$
- Stop when $2^k = n$
 - Implies $\log_2(n)$ recursions
- $O(\log_2(n))$

```
int bsearch(int data[],
            int start, int end,
            int target)
{
    if(end >= start)
        return -1;
    int mid = (start+end)/2;
    if(target == data[mid])
        return mid;
    else if(target < data[mid])
        return bsearch(data, start, mid,
                        target);
    else
        return bsearch(data, mid, end,
                        target);
}
```

Importance of Complexity



N	$O(1)$	$O(\log_2 n)$	$O(n)$	$O(n \log_2 n)$	$O(n^2)$	$O(2^n)$
2	1	1	2	2	4	4
20	1	4.3	20	86.4	400	1,048,576
200	1	7.6	200	1,528.8	40,000	1.60694E+60
2000	1	11.0	2000	21,931.6	4,000,000	#NUM!