

Recurrence Relations



Computer Science Department
University of Central Florida

COP 3502 – Computer Science I



Announcements

- Cheating:
 - ProGurl2, SpudMonkey, vette9890, sauve81, nesyou5, rpaul, srabbis (and others)
 - Guess what:
 - We now have code “snippets” from the “developers”
 - And what did that result in:
 - We have TWO students that have been identified.
 - SNAP!
 - And we’re still working to get the others students/cheaters identified
 - Cheaters have NOT been informed they were caught
 - Why?



Announcements

■ Cheating:

- ProGurl2, SpudMonkey, Vette9890, sauve81, nesyou5, rpaul, srabbis (and others)

■ **FINAL CHANCE:**

- Turn yourself in and get a zero for the assignment
 - and NOTHING MORE
- Call the following scare tactics, I don't care:
 - If you do NOT turn yourself in, you take the chance that you are from those we already identified
 - Is it with a roll of the dice? A 50/50 chance?
 - Again, if you do NOT turn yourself in, and we identify you, then you will get a ZERO for the course and it goes on your permanent UCF record



Outline

- Recursion
 - Simple warm up example (Factorial n)
- Recurrence Relations
 - Factorial N
 - Power N



Recursion

- What is Recursion?
 - Powerful, problem-solving strategy
 - Solves large problems by **reducing** them to **smaller** problems of the **same form**

- Example: Compute Factorial of a Number
 - $4! = 4 * 3 * 2 * 1 = 24$
 - $n! = n * (n-1) * (n-2) * \dots * 2 * 1$
 - Also, $0! = 1$
 - (just accept it!)



Recursion

- Example: Compute Factorial of a Number
 - Recursive Solution
 - Note that each factorial is related to a factorial of the next smaller integer
 - $n! = n * (n-1)!$
 - $4! = 4 * (4-1)! = 4 * (3!)$
 - But we need something else
 - We need a stopping case, or this will just go on and on and on
 - NOT good!
 - We let $0! = 1$
 - So in “math terms”, we say
 - $n! = 1$ if $n = 0$
 - $n! = n * (n-1)!$ if $n > 0$



Recursion

- Example: Compute Factorial of a Number

- Recursive Solution --- in C code

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

And notice how this function is very clean and basically follows the mathematical definition of factorial.

- This is recursive. Why?

- It defines the factorial of n in terms of the factorial of $(n-1)$, thus reducing the problem



Recurrence Relations

- Today we go over Recurrence Relations
 - The Question: What is a recurrence relation?
 - an **equation** that defines a sequence recursively
 - each term of the sequence is defined as a function of the preceding term
 - What is the purpose?
 - In response, let us ask, what is the purpose using Summations in Big-O analysis?
 - Answer:
 - Summations are a tool to assist in measuring the running time of **iterative** algorithms



Recurrence Relations

- Today we go over Recurrence Relations
 - What is the purpose?
 - But can we use this same method of analysis, along with summations, to decipher the running time of recursive algorithms?
 - You cannot!
 - You cannot simply “eyeball” a recursive function for a minute or two, in the way you can an iterative function, and come up with a Big-O. Just doesn’t work.
 - So just like summations are a tool to help find the Big-O of iterative algorithms
 - **Recurrence Relations are a tool to help find the Big-O of recursive algorithms**



Recurrence Relations

■ Back to Factorial N...

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

■ The GOAL:

- We want to come up with an **equation** that properly expresses this `fact` function in a **recursive manner**.
- Then we will need to **solve** this newly found equation.
 - We do so by putting it into its “closed form”.
- Here’s the process...



Recurrence Relations

■ Back to Factorial N...

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

■ What is happening in this problem?

- At every step of the recursion,
 - meaning, each time the function is recursively called,
- What happens?
 - We see that the input size (n) reduces by 1
 - So if n was 100, it is reduced to 99 when the function is called recursively for the first time.



Recurrence Relations

■ Back to Factorial N...

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

- What is happening in this problem?
 - Also, at every step of the recursion,
 - TWO mathematical operations are performed
 - The '*' and the '-' in `return (n * fact(n-1));`
 - So now we want to write an equation expressing these two facts.



Recurrence Relations

■ Back to Factorial N...

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

■ What is happening in this problem?

■ We can say the following:

- The total number of operations needed to execute this `fact` function for any given input, `n`, can be expressed as
 - 1) the sum of the 2 operations (the '*' and the '-')
 - 2) plus the number of operations needed to execute the function for `n-1`



Recurrence Relations

■ Back to Factorial N...

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

■ In techno talk:

- $T(n)$ represents the # of operations of this function
- $T(n)$ can be expressed as a sum of:
 - $T(n-1)$
 - and the two arithmetic operations



Recurrence Relations

■ Back to Factorial N...

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

■ In techno talk:

- $T(n)$ can be expressed as a sum of:
 - $T(n-1)$
 - and the two arithmetic operations

$$T(n) = T(n-1) + 2$$

$$T(0) = 1 \quad \text{Meaning, we it takes constant time to simply return.}$$



Recurrence Relations

■ Back to Factorial N...

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

■ So what did we just do?

- We came up with an equation that properly expresses this fact function in a recursive manner.

$$T(n) = T(n-1) + 2$$

$$T(0) = 1$$

- This equation is our Recurrence Relation



Recurrence Relations

- Back to Factorial N...
 - From this recurrence relation, $T(n)$, we can come up with a Big-O
 - Great, so we solved it, so let's move on!
 - **Not so fast.**
 - As it is, the recurrence relation,
$$T(n) = T(n-1) + 2$$
$$T(1) = 1$$
 - doesn't tell us about the # of operations of $T(n)$
 - Does anyone know how many operations are in $T(n-1)$?
 - Is it 487 operations? Perhaps 515,243 operations?
 - We DON'T know!



Recurrence Relations

- Back to Factorial N...
 - The problem is only “**solved**” once we **remove all T(...)'s from the right side of the equation**
 - Again, here's the equation:
$$T(n) = T(n-1) + 2$$
 - So T(n-1) needs to go bye-bye
 - **Then the problem is in its “closed form” and is solved.**
 - So how do we make this happen?
 - **BUCKLE UP and HOLD ON.**



Recurrence Relations

- Back to Factorial N
 - We need to solve $T(n)$ in terms of n
 - For the recurrence relation,
 - $T(n) = T(n-1) + 2$
 - Do we know what $T(n-1)$ equals?
 - Does it equal 8,572 operations?
 - Who knows? We surely don't know!
 - So we want to REDUCE the right side
 - specifically, the $T(n-1)$
 - UNTIL we get to that which we do know!



Recurrence Relations

■ Back to Factorial N

- We need to solve $T(n)$ in terms of n

- Starting from this equation:

$$T(n) = T(n-1) + 2$$

- We reduce the right side until we get to $T(1)$.

- Why?

- CUZ we know $T(1)$.

- What is $T(1)$?

- It is $1!$...this was from our Recurrence Relation earlier.

- So then we can put 1 in the place of $T(1)$

- Effectively eliminating all $T(\dots)$ s from the right side of eqn!



Recurrence Relations

■ Back to Factorial N

- We need to solve $T(n)$ in terms of n

$$T(n) = T(n-1) + 2$$

- We reduce the right side until we get to $T(1)$.
- Here's the idea:

$T(n-1)$

$T(n-2)$

$T(n-3)$

...

$T(n\text{-something}) = T(1)$

if we assume
that $n = 100$,
we have...

$T(100-1)$

$T(100-2)$

$T(100-3)$

...

$T(100-99) = T(1)$



Recurrence Relations

■ Back to Factorial N

- We need to solve $T(n)$ in terms of n

$$T(n) = T(n-1) + 2$$

- We reduce the right side until we get to $T(1)$.

- So, we do this in steps

- 1) We **replace n with $n-1$** on both sides of the equation
- 2) We plug the result back in
- 3) And then we do it again
and again and again and again...
till a “light goes off” and we see something



Recurrence Relations

Or you're like this guy, whose lights never turned on.





Recurrence Relations

■ Back to Factorial N

- $T(n) = T(n-1) + 2$ ----- call this Eq. 1
 - Replace n with n-1

DON'T overcomplicate this step.

It is REALLY this SIMPLE.

Wherever you see an n in Eq. 1, simply replace with n-1.

So if you have $T(n-1)$ and you replace that n with an n-1, you will get $T((n-1)-1)$, which equates to $T(n-2)$.

Simple right?

Right.



Recurrence Relations

■ Back to Factorial N

- $T(n) = T(n-1) + 2$ ----- call this Eq. 1

- Replace n with n-1

- $T(n-1) = T(n-2) + 2$ ----- call this Eq. 2

- Now substitute the result of Eq. 2 into Eq. 1

- $T(n) = T(n-2) + 2 + 2$

Wait? How'd we get this?

$$T(n) = T(n-1) + 2 \quad \text{----- Eq. 1}$$

And from Eq. 2, we also have, $T(n-1) = T(n-2) + 2$

So we simply plug in the result (the right side) of the Eq. 2 into Eq. 1 where we see $T(n-1)$

$$T(n) = T(n-1) + 2$$

$$T(n) = (T(n-2) + 2) + 2 \quad \text{removing parantheses, we get}$$

$$T(n) = T(n-2) + 2 + 2$$



Recurrence Relations

■ Back to Factorial N

- $T(n) = T(n-1) + 2$ ----- call this Eq. 1

- Replace n with n-1

- $T(n-1) = T(n-2) + 2$ ----- call this Eq. 2

■ Now substitute the result of Eq. 2 into Eq. 1

- $T(n) = T(n-2) + 2 + 2$

- We can look at $2 + 2$ as $2 * 2$ you'll see why we do this shortly

- $T(n) = T(n-2) + 2 * 2$ ----- call this Eq. 3

■ So what did we do:

- We made ANOTHER equation for $T(n)$

- But this one is in terms of $T(n-2)$

- REDUCED from being in terms of $T(n-1)$



Recurrence Relations

■ Back to Factorial N

- So we now have this new equation for $T(n)$:
 - $T(n) = T(n-2) + 2*2$
- Are we done?
 - NO! Cuz we still have $T(\dots)$ s on the right
- And do we know how many operations are performed by $T(n-2)$?
 - Perhaps 5,219 operations? We don't know!
- So we now need to REDUCE this equation further
- We have $T(n)$ in terms of $T(n-2)$
- We want to get $T(n)$ in terms of $T(n-3)$



Recurrence Relations

■ Back to Factorial N

- So we now need to REDUCE this equation further
- We want to get $T(n)$ in terms of $T(n-3)$
- How are we going to do this?
 - We currently have $T(n) = T(n-2) + 2 \cdot 2$
 - We want to develop an equation with $T(n-2)$ on the **left**
 - and in terms of $T(n-3)$
- So, in Eq. 2, once again, replace n with $n-1$
 - $T(n-1) = T(n-2) + 2$ ----- Eq. 2
 - Replace n with $n-1$
 - $T(n-2) = T(n-3) + 2$ ----- call this Eq. 4
- Ah! So we now have our “ $T(n-2)$ ” equation



Recurrence Relations

■ Back to Factorial N

■ Now substitute the result of Eq. 4 into Eq. 3

- $T(n-2) = T(n-3) + 2$ ----- Eq. 4

- $T(n) = T(n-2) + 2 * 2$ ----- Eq. 3

- $T(n) = T(n-3) + 2 + 2 * 2$

- $2 + 2 * 2$ really is $2 * 3$...again, you'll see why we do this in a bit

- $T(n) = T(n-3) + 2 * 3$

■ Again, what did we accomplish?

- We made ANOTHER equation for $T(n)$

- But this one is in terms of $T(n-3)$

- REDUCED from being in terms of $T(n-2)$



Recurrence Relations

■ Back to Factorial N

- Thus far, we have three equations with $T(n)$ on the left side

- $T(n) = T(n-1) + 2*1$
 - Note that I added the $*1$ next to the 2
 - This doesn't change anything right?
 - $2*1$ is the same as just plain 'ole 2
 - You'll see why we did this in a second.
- $T(n) = T(n-2) + 2*2$
- $T(n) = T(n-3) + 2*3$



Recurrence Relations

- Back to Factorial N
 - Is there a pattern developing? Perhaps some “light” going off?
 - 1st step of recursion, we have: $T(n) = T(n-1) + 2^1$
 - 2nd step of recursion, we have: $T(n) = T(n-2) + 2^2$
 - 3rd step of recursion, we have: $T(n) = T(n-3) + 2^3$
 - If we followed the process one more time, we get
 - $T(n) = T(n-4) + 2^4$...for the 4th step of the recursion
 - So on the **kth step/stage of the recursion**, we get a **generalized recurrence relation**:
 - $T(n) = T(n-k) + 2^k$



Recurrence Relations

- Back to Factorial N
 - So on the **kth step/stage of the recursion**, we get a **generalized recurrence relation**:
 - $T(n) = T(n-k) + 2^k$
 - WHEW!
 - That was a lot!
 - But we're finally done! Right?.
 - WRONG!!! Why aren't we done yet?
 - CUZ we still have $T(\dots)$ s on the right side of the equation
 - So now we need to actually solve this generalized recurrence relation



Recurrence Relations

- Back to Factorial N
 - We need to solve this generalized rec. relation
 - $T(n) = T(n-k) + 2^k$
 - How?
 - Remember we said we wanted to reduce the right side of the equation to $T(1)$
 - Again, why?
 - Because we know what $T(1)$ equals...it equals 1!
 - So we have $T(n-k)$ and we want $T(1)$
 - Simple! Let $n - k = 1$
 - Solve for k leaving $k = n - 1$
 - Plug back into equation



Recurrence Relations

■ Back to Factorial N

- We need to solve this generalized rec. relation
 - $T(n) = T(n-k) + 2 \cdot k$
 - $k = n - 1$
 - Plug into above equation
 - $T(n) = T(n-(n-1)) + 2(n-1) = T(1) + 2(n-1)$
 - And we know that $T(1) = 1$
 - Therefore....
 - $T(n) = 2(n-1) + 1 = 2n - 1$
 - And we are done!
- Right side does not have any $T(\dots)$'s
- This rec. relation is now solved!
- This algorithm runs in LINEAR time.



Brief Interlude: Human Stupidity





Recurrence Relations

- Let's look at a function that calculates powers

```
int power (int x, int n) {           // calculates the value of x^n
    if (n == 0)
        return 1;
    if (n == 1)
        return x;
    if (n is even)
        return power(x*x, n/2);
    else // if n is odd
        return power(x*x, n/2)*x;
}
```

- What's going on in this problem?
 - At every step, the problem size is reduced by **half**
 - If n is even, 2 arithmetic operations are computed
 - If n is odd, 3 arithmetic operations are computed



Recurrence Relations

■ Power Function

- What's going on in this problem?
 - At every step, the problem size is reduced by **half**
 - If n is even, 2 arithmetic operations are computed
 - If n is odd, 3 arithmetic operations are computed
- When computing time complexity, **we assume the worst case**
 - We assume n is odd at each step
 - So 3 operations are assumed to be always needed
- Thus, $T(n)$ can be expressed as the sum of $T(n/2)$ and the 3 operations needed at each step

$$T(n) = T(n/2) + 3$$

$$T(1) = 1$$



Recurrence Relations

- Power Function

- So here's our recurrence relation:

$$T(n) = T(n/2) + 3$$

$$T(1) = 1$$

- We need to solve this by removing all $T(\dots)$'s from the right side.
 - $T(n/2)$ needs to hit the road
 - Then the problem is in its “closed form” and is solved.



Recurrence Relations

■ Power Function

- We need to solve $T(n)$ in terms of n
- Starting from this equation

$$T(n) = T(n/2) + 3$$

We reduce the right side until we get to $T(1)$.

- Why?
 - $T(1)$ is known to us (it equals 1)
- We do this in steps
 - We replace n with $n/2$ on both sides of the equation
 - We plug the result back in
 - And then we do it again...till a “light goes off” and we see something



Recurrence Relations

■ Power Function

- This time we'll do a slightly different order of things...just so you see two different ways
 - Start with the base recurrence relation
 - $T(n) = T(n/2) + 3$ ----- call this Eq. 1
 - Replace n with $n/2$, and go ahead and do this several times
 - $T(n/2) = T(n/4) + 3$ ----- call this Eq. 2
 - $T(n/4) = T(n/8) + 3$ ----- call this Eq. 3
 - $T(n/8) = T(n/16) + 3$ ----- call this Eq. 4
- Now we substitute each one of these back into Eq.1 and hopefully see a pattern



Recurrence Relations

■ Power Function

- Here's the four current equations we have:

- $T(n) = T(n/2) + 3$ ----- Eq. 1

- $T(n/2) = T(n/4) + 3$ ----- Eq. 2

- $T(n/4) = T(n/8) + 3$ ----- Eq. 3

- $T(n/8) = T(n/16) + 3$ ----- Eq. 4

- Now substitute the result of Eq. 2 into Eq. 1

- $T(n) = T(n/4) + 3 + 3$

- We can look at $3 + 3$ as $3 * 2$ you remember why...right.?

- $T(n) = T(n/4) + 3 * 2$ ----- call this Eq. 5



Recurrence Relations

■ Power Function

- Here's the four current equations we have:

- $T(n) = T(n/2) + 3$ ----- Eq. 1

- $T(n/2) = T(n/4) + 3$ ----- Eq. 2

- $T(n/4) = T(n/8) + 3$ ----- Eq. 3

- $T(n/8) = T(n/16) + 3$ ----- Eq. 4

- Now substitute the result of Eq. 3 into Eq. 5

- $T(n) = T(n/8) + 3 + 3*2$

- $T(n) = T(n/8) + 3*3$ ----- call this Eq. 6

- One more substitution of Eq. 4 into Eq. 6:

- $T(n) = T(n/16) + 3*4$ ----- call this Eq. 7



Recurrence Relations

■ Power Function

- Now show all the equations we developed with $T(n)$ on the left...is there a pattern developing?
 - $T(n) = T(n/2) + 3*1$ $= T(n/2^1) + 3*1$
 - $T(n) = T(n/4) + 3*2$ $= T(n/2^2) + 3*2$
 - $T(n) = T(n/8) + 3*3$ $= T(n/2^3) + 3*3$
 - $T(n) = T(n/16) + 3*4$ $= T(n/2^4) + 3*4$
- So on the k th step/stage of the recursion, we get a generalized recurrence relation:
 - $T(n) = T(n/2^k) + 3*k$
- We're not done yet right.
- Cuz we need to get rid of the $T(n/2^k)$



Recurrence Relations

■ Power Function

- We need to solve this generalized rec. relation
 - $T(n) = T(n/2^k) + 3 \cdot k$
- How?
 - Remember we said we wanted to reduce the right side of the equation to $T(1)$
 - Again, why?
 - Because we know what $T(1)$ equals...it equals 1!
 - So we have $T(n/2^k)$ and we want $T(1)$
 - Simple! Let $n = 2^k$
 - Solve for k
 - Take log base 2 of both sides
 - $k = \log n$

Plug back into equation



Recurrence Relations

■ Power Function

- We need to solve this generalized rec. relation
 - $T(n) = T(n/2^k) + 3 \cdot k$
 - So $n = 2^k$ and $k = \log n$
 - Plug into above equation
 - $T(n) = T(1) + 3(\log n)$
 - And we know that $T(1) = 1$
 - Therefore....
 - $T(n) = 1 + 3\log(n)$
 - And we are done! This algorithm runs in logarithmic time.
- Right side does not have any $T(\dots)$'s
- This rec. relation is now solved!



Recurrence Relations

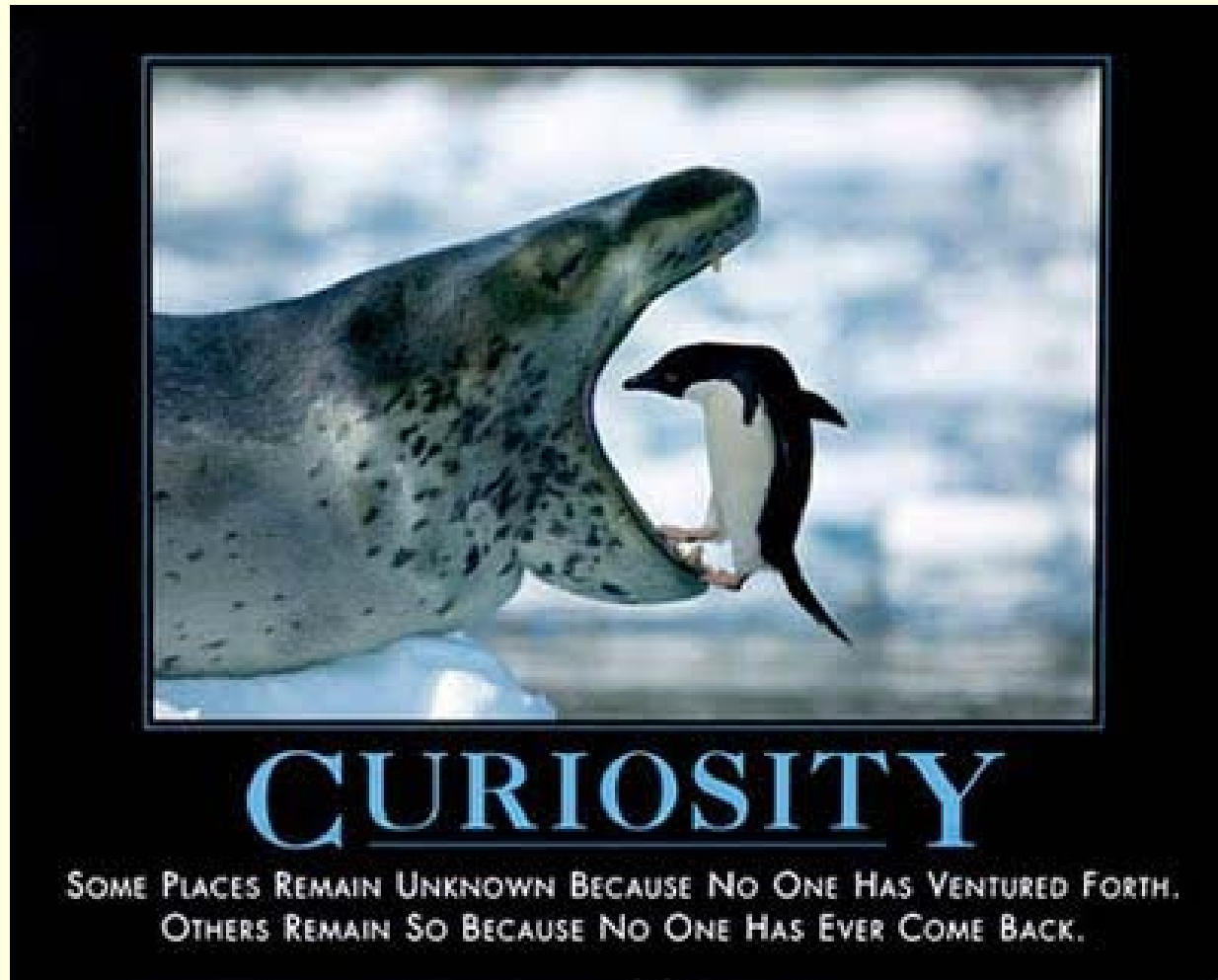
WASN'T

THAT

**(Let's admit it:
that sucked!)**



Daily Demotivator



Recurrence Relations



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COP 3502 – Computer Science I