Note: Print to submit the last three pages where the homework problems are listed.

Due: by the next class hour (4/1, 2 points)

## **Recurrence Relations**

**Recurrence relation** is an equation that recursively defines a sequence or multidimensional array of values, once one or more initial terms are given: each further term of the sequence or array is defined as a function of the preceding terms.

There are various methods for solving recurrence relations.

Here we will use a method called "telescoping". The method consists in the following:

- Re-write the recurrence relation for subsequent smaller values of n so that the left side of the re-written relation is equal to the first term of the right side of the previous relation. Rewrite the relation until a relation involving the initial condition is obtained.
- Add the left sides and the right sides of the relations. Cancel the equal terms on the left and the right side.
- Add the remaining terms on the right side. The sum will give the general formula of the sequence.

The standard method for solving recurrence relations, called "unfolding", makes repeated substitutions applying the recursive rule until the base case is reached.

### **Useful formulas:**

$$1 + 2 + 3 + ... + N = N(N+1)/2$$
  
 $1 + 2 + 4 + 8 + ... + 2^{n} = 2^{n+1} - 1$ 

# Example 0:

We may describe that the time complexity of the linear search is

$$T(0)=c_0$$
 // the time to process an array of 1 element is 1 
$$T(n)=T(n-1)+c$$
 // to process n items is to process (n-1) elements + 1 element

The cost of searching n elements is the cost of looking at 1 element, plus the cost of searching n-1 elements. We may express the constants  $c_0$  and c as 1, respectively, for our purpose. The expressions become:

$$T(0) = 1$$

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

### **Telescoping:**

$$T(n) = T(n-1) + 1$$
  
 $T(n-1) = T(n-2) + 1$   
 $T(n-2) = T(n-3) + 1$   
...  
 $T(1) = T(0) + 1$ 

Add the left and right sides:

$$T(n) + T(n-1) + T(n-2) + ... + T(1) = T(n-1) + T(n-2) + ... + T(0) + 1 + 1 + ... + 1$$
  
Cancel equal terms on both sides: 
$$T(n) = T(0) + 1 + 1 + ... + 1$$
  
How many n's are there? 
$$T(n) = T(0) + n$$
$$= 1 + n$$
  
Therefore,  $T(n) = O(n)$ .

#### **Unfolding:**

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

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

$$= T(n-3) + 3$$
...
$$= T(n-n) + n = T(0) + n$$

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

$$T(n) = O(n)$$

# Example 1:

Consider the sequence 1, 3, 6, 10, 15, 21, 28, 36, ... Each term is obtained from the previous by adding the number that shows the position of the current term to the previous term, e.g. 10 is in position 4 and the previous term is 6: 10 = 6 + 4. We want to obtain a formula for the general term of this sequence.

$$a_1 = 1$$
  
 $a_n = a_n-1 + n, n = 2, 3, 4, ...$ 

#### Telescoping:

$$a_n = a_{n-1} + n$$
 $a_{n-1} = a_{n-2} + (n-1)$ 
 $a_{n-2} = a_{n-3} + (n-2)$ 
....
 $a_4 = a_3 + 4$ 
 $a_3 = a_2 + 3$ 
 $a_2 = a_1 + 2$ 

Add the left and right sides:

$$a_n + a_{n-1} + a_{n-2} + a_{n-3} + ... + a_4 + a_3 + a_2 =$$
  
 $a_n - 1 + a_n - 2 + a_n - 3 + ... + a_4 + a_3 + a_2 + a_1 + 2 + 3 + 4 + ... + (n-1) + n$ 

Cancel equal terms on both sides:

$$a_n = a_1 + 2 + 3 + 4 + ... + (n - 1) + n$$

Replace with the value given in the initial condition:

$$a_n = 1 + 2 + 3 + 4 + ... + (n - 1) + n = n(n+1)/2$$

Thus we have obtained the general formula for each term in the sequence:  $a_n = n(n+1)/2$ 

# Example 2: findmax()

Find recursively the max element in an array.

Algorithm:

If the array has 1 element,

this is the maximum

Else,

- a. Find the maximum in the array of the first n-1 elements.
- b. Compare with the n-th element and return the greater one

```
int findmax(int [] array, int size) {
if (size == 1)
    return array[0];
else {
    int max = findmax(array, size-1)
    if (max >= array[size-1])
       return max;
    else
       return array[size-1];
    }
}
```

Here we reason in the following way:

- The time to find the maximum in an array of size N is equal to the time to find the maximum in an array of size N-1 plus the time to do the comparison.
- Since the time to do the comparison does not depend on the size of the array, we consider it to be a constant and we use 1.
- When the array has only one element the time is a constant (again we use 1)

Thus we come up with the following recurrence relation:

```
T(1) = 1 // the time to process an array of 1 element is 1 T(N) = T(N-1) + 1
```

#### **Telescoping:**

```
T(N) = T(N-1) + 1

T(N-1) = T(N-2) + 1

T(N-2) = T(N-3) + 1

....

T(3) = T(2) + 1

T(2) = T(1) + 1

T(1) = 1
```

Adding the left and the right side and canceling equal terms, we obtain:

```
T(N) = 1 + 1 + ... + 1
```

There are N lines in the telescoping, thus the sum of the 1s is

```
T(N) = N

T(N) = O(n)
```

Notice: Print to submit the last three page where the homework problems exist.

## Problem 1 - insertion sort

**Recurrence relation:** The time to sort an array of N elements is equal to the time to sort an array of N-1 elements plus N-1 comparisons. Initial condition: the time to sort an array of 1 element is constant:

$$T(1) = 1$$

$$T(N) = T(N-1) + N-1$$

Next we perform **telescoping**: re-writing the recurrence relation for N-1, N-2, ..., 2

$$T(N) = T(N-1) + N-1$$

.....

Next we sum up the left and the right sides of the equations above:

$$T(N) + T(N-1) + T(N-2) + T(N-3) + \dots + T(3) + T(2) =$$

$$T(N-1) + T(N-2) + T(N-3) + \dots + T(3) + T(2) + T(1) + \dots$$

Finally, we cross the equal terms on the opposite sides and simplify the remaining sum on the right side:

Therefore, the running time of insertion sort is:

$$T(N) =$$
\_\_\_\_\_\_ =  $O(N^2)$ 

#### Problem 2

$$T(1) = 1$$

$$T(N) = T(N-1) + 2$$
 // 2 is a constant like c

## Telescoping:

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

. . . . . .

Next we sum up the left and the right sides of the equations above:

Finally, we cross the equal terms on the opposite sides and simplify the remaining sum on the right side:

$$T(N) = T(1) +$$
\_\_\_\_\_

T(N) = 1 +
------------

Therefore, the running time of reversing a queue is:

$$\mathsf{T}(\mathsf{N}) = \underline{\hspace{1cm}} = \mathsf{O}(\mathsf{N})$$

# Problem 3 - Power()

```
long power(long x,long n) {
  if(n==0)
    return1;
  else
    return x * power(x,n-1);
}
```

T(n) = Time required to solve a problem of size n

Recurrence relations are used to determine the running time of recursive programs—recurrence relations themselves are recursive

T(0) = time to solve problem of size 0

Base Case

T(n) = time to solve problem of size n

- Recursive Case

$$T(0) = 1$$

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

// +1 is a constant

## Solution by telescoping:

If we knew T(n-1), we could solve T(n).

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

$$T(n-1) =$$
\_\_\_\_\_

. . . .

$$T(1) =$$

Next we sum up the left and the right sides of the equations above:

$$T(n) + T(n-1) + T(n-2) + T(n-3) + \dots$$
  $T(3) + T(2) + T(1) =$   $T(n-1) + T(n-2) + T(n-3) + \dots$   $T(3) + T(2) + T(1) + T(0) + \dots$ 

Finally, we cross the equal terms on the opposite sides and simplify the remaining sum on the right side:

$$O(n) = n$$

# Problem 4 - Power()

```
long power(long x,long n) {
  if (n == 0) return 1;
  if (n == 1) return x;
  if ((n % 2) == 0)
   return power(x * x, n/2);
  else
    return power(x * x, n/2) * x;
}
T(0) = 1
T(1) = 1
T(n) = T(n/2) + 1
                       // Assume n is power of 2, +1 is a constant
Solution by unfolding:
T(0) = 1
T(1) = 1
T(n) = T(n/2) + 1
                                    since T(n/2) = T(n/4) + 1
                                    since T(n/4) = T(n/8) + 1
    = T(n/16) + 4
                                     in terms of n, 2k, k
We want to get rid of T(n/2^k).
We solve directly when we reach T(1)
n/2^k = 1
  n = 2^k
\lg n = k
T(n) = _{--}
                                  in terms of n, 2k, k
```

Therefore, its time complexity is O(lg n)