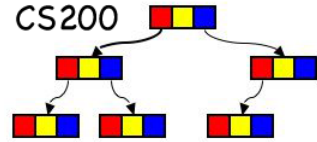


CS200: Recurrence Relations and the Master Theorem

Rosen Ch. 8.1 - 8.3

Recurrence Relations: An Overview

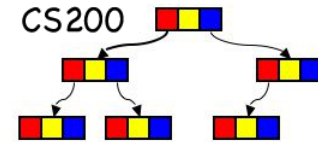


- What is a recurrence?
 - **A recursively defined sequence ...**

Example

- Arithmetic progression: $a, a+d, a+2d, \dots, a+nd$
 - $a_0 = a$
 - $a_n = a_{n-1} + d$

Formal Definition



A recurrence relation for the sequence $\{a_n\}$ is an equation that expresses a_n in terms of one or more of the previous terms of the sequence, namely, a_0, a_1, \dots, a_{n-1} , for all integers n with $n \geq n_0$ where n_0 is a nonnegative integer.

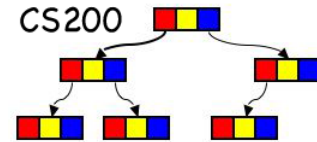
- A Sequence is called a solution of a Recurrence relation + Initial conditions (“*base case*”), if its terms satisfy the recurrence relation

- Example: $a_n = a_{n-1} + 2, a_1 = 1$

$a_1?, a_2? a_3?$
solution?

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

Compound Interest



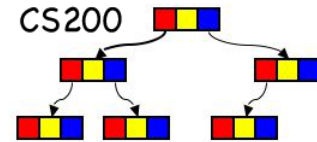
- You deposit \$10,000 in a savings account that yields 10% yearly interest. How much money will you have after 1,2, ... years? (b is balance, r is rate)

$$b_n = b_{n-1} + rb_{n-1} = (1 + r)^n b_0$$

$$b_0 = 10,000$$

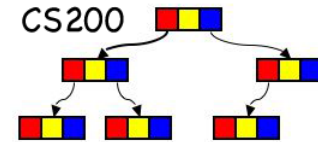
$$r = 0.1$$

Modeling with Recurrence



- Suppose that the number of bacteria in a colony triples every hour
 - Set up a recurrence relation for the number of bacteria after n hours have elapsed.
 - 100 bacteria are used to begin a new colony.

Recursively defined functions and recurrence relations



- **A recursive function**

$f(0) = a$ (base case)

$f(n) = f(n-1) + d$ for $n > 0$ (recursive step)

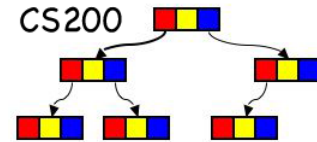
- The above recursively defined function generates the sequence

$$a_0 = a$$

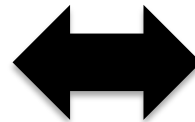
$$a_n = a_{n-1} + d$$

- A recurrence relation **produces a sequence**, an application of a recursive function produces a **value from the sequence**

How to Approach Recursive Relations



Recursive Functions



Sequence of Values

$$f(0) = 0 \text{ (base case)}$$

$$f(n) = f(n-1) + 2 \text{ for } n > 0$$

(recursive part)

$$f(0) = 0$$

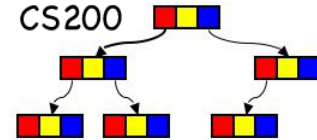
$$f(1) = f(0) + 2 = 2$$

$$f(2) = f(1) + 2 = 4$$

$$f(3) = f(2) + 2 = 6$$

Closed Form?(solution,
explicit formula)

Find a recursive function



- Give a recursive definition of $f(n)=a^n$, where a is a nonzero real number and n is a nonnegative integer.

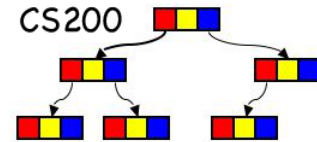
$$f(0) = 1,$$
$$f(n) = a * f(n-1)$$

- Give a recursive definition of factorial $f(n) = n!$

$$f(0) = 1$$
$$f(n) = n * f(n-1)$$

- Rosen Chapter 5 example 3-2 pp. 346

Solving recurrence relations



Solve $a_0 = 2$; $a_n = 3a_{n-1}$, $n > 0$

(1) What is the recursive function?

(2) What is the sequence of values?

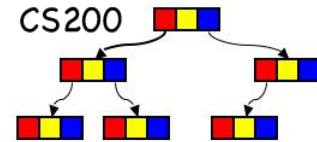
a

Hint: Solve by repeated substitution, recognize a pattern, check your outcome

■ $a_0 = 2$; $a_1 = 3(2) = 6$; $a_2 = 3(a_1) = 3(3(2))$; $a_3 = \dots$

Connection to Complexity...

Divide-and-Conquer



Basic idea:

Take large problem and **divide** it into **smaller problems** until problem is trivial, then **combine** parts to make solution.

Recurrence relation for the number of steps required:

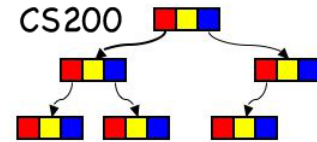
$$f(n) = a f(n / b) + g(n)$$

n/b : the size of the sub-problems solved

a : number of sub-problems

$g(n)$: steps necessary to **split** sub-problems and **combine** solutions to sub-problems

Example: Binary Search

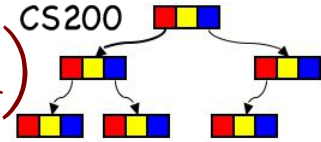


```
public int binSearch (int myArray[], int first,
                    int last, int value) {
    // returns the index of value or -1 if not in the array
    int index;
    if (first > last) { index = -1; }
    else {
        int mid = (first + last)/2;
        if (value == myArray[mid]) { index = mid; }
        else if (value < myArray[mid]) {
            index = binSearch(myArray, first, mid-1, value);
        }
        else {
            index = binSearch(myArray, mid+1, last, value);
        }
    }
    return index;
}
```

What are a, b, and g(n)?

$$f(n) = a \cdot f(n/b) + g(n)$$

Estimating big-O (Master Theorem)



Let f be an increasing function that satisfies

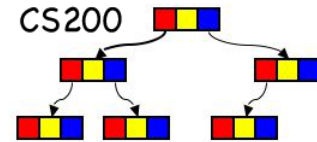
$$f(n) = a \cdot f(n/b) + c \cdot n^d$$

whenever $n = b^k$, where k is a positive integer, $a \geq 1$, b is an integer > 1 , and c and d are real numbers with c positive and d nonnegative. Then

$$f(n) = \left\{ \begin{array}{ll} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{array} \right\}$$

*Section 8.3 in Rosen
Proved using induction*

Binary Search using the Master Theorem



For binary search

$$\begin{aligned} f(n) &= a f(n / b) + c . n^d \\ &= 1 f(n / 2) + c \end{aligned}$$

$$f(n) = \left\{ \begin{array}{ll} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{array} \right.$$

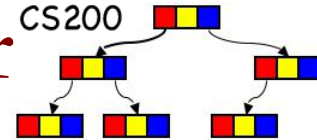
Therefore, **$d = 0$** (to make n^d a constant), **$b = 2$** , **$a = 1$** .

$$b^d = 2^0 = 1$$

It satisfies the second condition of the Master theorem.

$$\text{So, } f(n) = O(n^d \log_2 n) = O(n^0 \log_2 n) = \mathbf{O(\log_2 n)}$$

Complexity of MergeSort with Master Theorem

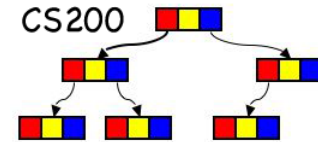


```
public void mergesort(Comparable[] theArray, int first, int last) {  
    // Sorts the items in an array into ascending order.  
    // Precondition: theArray[first..last] is an array.  
    // Postcondition: theArray[first..last] is a sorted permutation  
    if (first < last) {  
        int mid = (first + last) / 2; // midpoint of the array  
        mergesort(theArray, first, mid);  
        mergesort(theArray, mid + 1, last);  
        merge(theArray, first, mid, last);  
    } // if first >= last, there is nothing to do  
}
```

- $M(n)$ is the number of operations performed by mergeSort on an array of size n
- $M(0)=M(1) = 1$ $M(n) = 2M(n/2) + c.n$ **WHY + n ?**

the cost of merging two arrays of size $n/2$ into one of size n

Complexity of MergeSort



Master theorem

$M(n) = 2M(n/2) + c.n$
for the mergesort algorithm

$$\begin{aligned} f(n) &= a f(n/b) + c.n^d \\ &= 2 f(n/2) + c.n^1 \end{aligned}$$

$$f(n) = \left\{ \begin{array}{ll} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{array} \right.$$

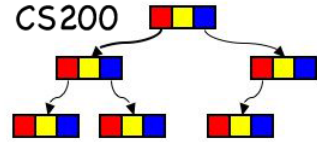
Notice that c does not play a role (big O)

$d = 1, b = 2, a = 2$. Therefore $b^d = 2^1 = 2$

It satisfies the second condition of the Master theorem.

$$\begin{aligned} \text{So, } f(n) &= O(n^d \log_2 n) \\ &= O(n^1 \log_2 n) \\ &= \mathbf{O(n \log_2 n)} \end{aligned}$$

Best Case QuickSort Recurrence



$$f(n) = a \cdot f(n/b) + cn^d$$

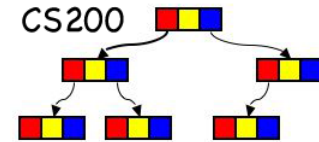
Best case: assume perfect division in equal sized partitions

- a=
- b=
- c=
- d=
- O(?)

$$f(n) = \left\{ \begin{array}{ll} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{array} \right\}$$

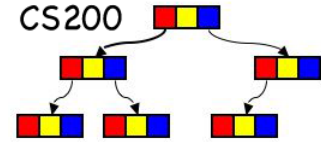
Worst Case: $n + (n-1) + \dots + 3 + 2 + 1 = O(n^2)$

CS320 Excursion: Tractability



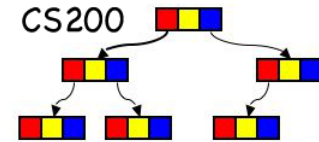
- A problem that is solvable using an algorithm with **polynomial worst-case complexity** is called **tractable**.
- If the polynomial has a high degree or if the coefficients are extremely large, the algorithm may take an extremely long time to solve the problem.

Intractable vs Unsolvable problems



- If the problem *cannot be solved* using an algorithm with worst-case polynomial time complexity, such problems are called **intractable**. **Have you seen such problems?**
- If it can be shown that no algorithm exists for solving them, such problems are called **unsolvable**.

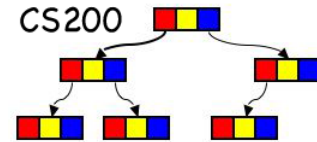
Hanoi



```
// pegs are numbers, via is computed
// number of moves are counted
// empty base case
public void hanoi(int n, int from, int to){
    if (n>0) {
        int via = 6 - from - to;
        hanoi(n-1,from, via);
        System.out.println("move disk " + n +
            " from " + from + " to " + to);
        count++;
        hanoi(n-1,via,to);
    }
}
```

**Recurrence for
number of moves?
Solution?
How did we prove
this earlier?**

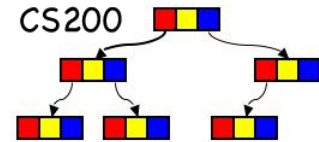
Permutations



```
public void permute(int from) {  
    if (from == P.length-1) { // suffix size one, nothing to permute  
        System.out.println(Arrays.toString(P));  
    }  
    else { // put every item in first place and recur  
        for (int i=from; i<P.length;i++) {  
            swapP(from,i); // put i in first position of suffix  
            permute(from+1); // permute the rest  
            swapP(from,i); // PUT IT BACK  
        }  
    }  
}
```

complexity? number of permutations? recurrence relation?

Interesting Intractable Problems

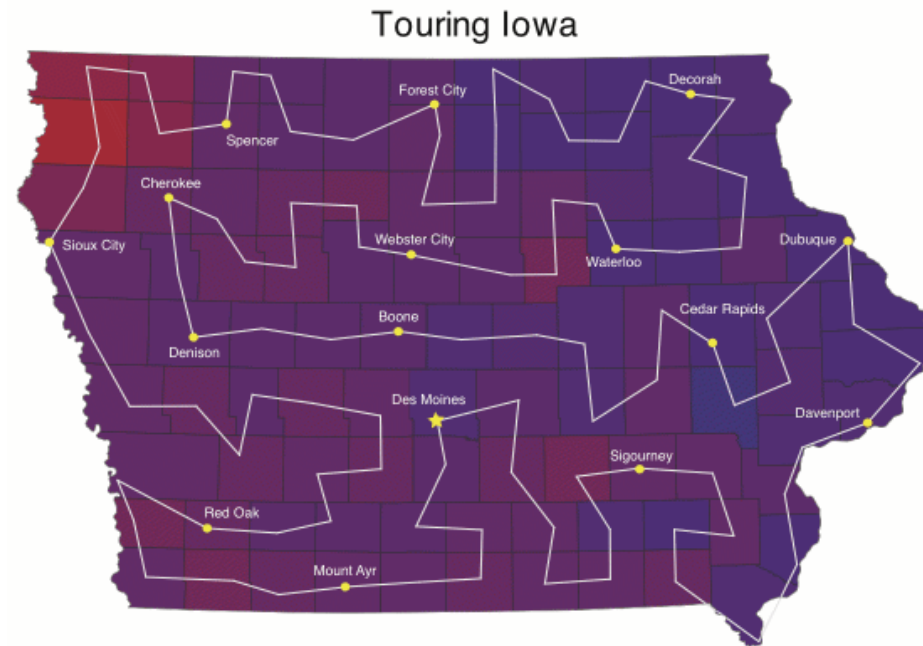


- Boolean Satisfiability 2^n

$$(A \vee \sim B \vee C) \wedge (\sim A \vee C \vee \sim D) \wedge (B \vee \sim C \vee D)$$

- TSP $n!$

- only solution:
trial and error



(c) W.J. Cook, A.L. Kornhauser, and R.J. Vanderbei

how many options for these problems?