Week 6 - Wednesday



Last time

- What did we talk about last time?
- Finished Exam 1 post mortem
- Data compression example
- Mergesort

Questions?

Assignment 3

Logical warmup

- On the planet Og, there are green people and red people
- Likewise, there are northerners and southerners
- Green northerners tell the truth
- Red northerners lie
- Green southerners lie
- Red southerners tell the truth
- Consider the following statements by two natives named Ork and Bork:
 - Ork: Bork is from the north
 - Bork: Ork is from the south
 - Ork: Bork is red
 - Bork: Ork is green
- What are the colors and origins of Ork and Bork?

Divide and Conquer

Recursive running time

- If we can, we want to turn the recursive version of T(n) into an explicit (non-recursive) Big Oh bound
- Before we do, note that we could similarly have written:

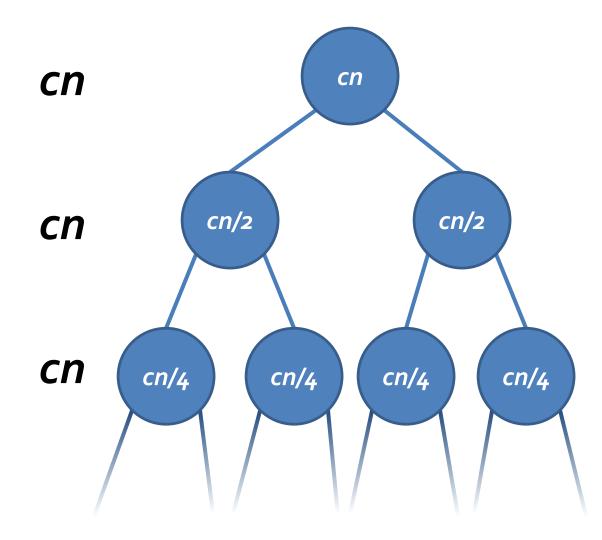
$$T(n) \le 2T\left(\frac{n}{2}\right) + O(n)$$

- Also, we can't guarantee that *n* is even
- A more accurate statement would be

$$T(n) \le T\left(\left[\frac{n}{2}\right]\right) + T\left(\left[\frac{n}{2}\right]\right) + cn$$

 Usually, we ignore that issue and assume that *n* is a power of 2, evenly divisible forever

Intuition about mergesort recursion



- Each time, the recursion cuts the work in half while doubling the number of problems
 - The total work at each level is thus always *cn*
- To go from *n* to 2, we have to cut the size in half (log₂ *n*) 1 times

Checking a solution

- We know that there's *cn* work at each level and approximately log₂
 n levels
- If we think that the running time O($n \log n$), we can guess that T(n)≤ $cn \log_2 n$ and substitute that in for T(n/2)

$$T(n) \le 2T\left(\frac{n}{2}\right) + cn$$

$$\le 2c\left(\frac{n}{2}\right)\log_2\left(\frac{n}{2}\right) + cn$$

$$= cn\left(\log_2 n - 1\right) + cn$$

$$= cn\log_2 n - cn + cn$$

$$= cn\log_2 n$$

Divide and conquer

- Divide and conquer algorithms are ones in which we divide a problem into parts and recursively solve each part
- Then, we do some work to combine the solutions to each part into a final solution
- Divide and conquer algorithms are often simple
- However, their running time can be challenging to compute because recursion is involved

Recursively defined sequences

- Defining a sequence recursively as with Mergesort is called a recurrence relation
- The **initial conditions** give the starting point
- Example:
 - Initial conditions
 - **T**(0) = 1
 - *T*(1) = 2
 - Recurrence relation
 - T(k) = T(k-1) + kT(k-2) + 1, for all integers $k \ge 2$
 - Find T(2), T(3), and T(4)

Writing recurrence relations in multiple ways

- Consider the following recurrence relation:
 - T(k) = 3T(k-1) 1, for all integers $k \ge 1$
- Now consider this one:
 - T(k+1) = 3T(k) 1, for all integers $k \ge 0$
- Both recurrence relations have the same meaning

Differences in initial conditions

- Even if the recurrence relations are equivalent, different initial conditions can cause a different sequence
- Example:
 - T(k) = 3 T(k-1), for all integers $k \ge 2$
 - *T*(1) = 2
 - S(k) = 3S(k-1), for all integers $k \ge 2$
 - **S**(1) = 1
 - Find T(1), T(2), and T(3)
 - Find S(1), S(2), and S(3)

Compound interest

- Interest is compounded based on some period of time
- We can define the value recursively
- Let *i* is the **annual percentage rate** (APR) of interest
- Let *m* be the number of times per year the interest is compounded
- Thus, the total value of the investment at the kth period is
 - $P(k) = P(k-1) + P(k-1)(i/m), k \ge 1$
 - **P**(o) = initial principle

Solving Recurrence Relations

Recursion

- ... is confusing
- We don't naturally think recursively (but perhaps you can raise your children to think that way?)
- With an interest rate of *i*, a principle of *P*(o), and *m* periods per year, the investment will yield *P*(o)(*i*/*m* + 1)^k after *k* periods

Finding explicit formulas by iteration

- We want to be able to turn recurrence relations into explicit formulas whenever possible
- Often, the simplest way is to find these formulas by **iteration**
- The technique of iteration relies on writing out many expansions of the recursive sequence and looking for patterns
- That's it

Iteration example

- Find a pattern for the following recurrence relation:
 - *T*(*k*) = *T*(*k*-1) + 2
 - **T**(0) = 1
- Start at the first term
- Write the next below
- Do not combine like terms!
- Leave everything in expanded form until patterns emerge

Arithmetic sequence

- In principle, we should use mathematical induction to prove that the explicit formula we guess actually holds
- The previous example (odd integers) shows a simple example of an arithmetic sequence
- These are recurrences of the form:
 - T(k) = T(k-1) + d, for integers $k \ge 1$
- Note that these recurrences are always equivalent to
 - T(n) = T(o) + dn, for all integers $n \ge o$

Geometric sequence

- Find a pattern for the following recurrence relation:
 - $T(k) = rT(k-1), k \ge 1$
 - **T**(0) = **a**
- Again, start at the first term
- Write the next below
- Do not combine like terms!
- Leave everything in expanded form until patterns emerge

Geometric sequence

- It appears that any geometric sequence with the following form
 - $T(k) = rT(k-1), k \ge 1$
- is equivalent to
 - $T(n) = T(o)r^n$, for all integers $n \ge o$
- This result applies directly to compound interest calculation

Employing outside formulas

- Intelligent pattern matching gets you a long way
- However, it is sometimes necessary to substitute in some known formula to simplify a series of terms
- Recall
 - Geometric series: $1 + r + r^2 + ... + r^n = (r^{n+1} 1)/(r 1)$
 - Arithmetic series: 1 + 2 + 3 + ... + n = n(n + 1)/2

How many edges are in a complete graph?

- In a complete graph, every node is connected to every other node
- If we want to make a complete graph with k nodes, we can take a complete graph with k 1 nodes, add a new node, and add k 1 edges (so that all the old nodes are connected to the new node)
- Recursively, this means that the number of edges in a complete graph is
 - $S(k) = S(k-1) + (k-1), k \ge 2$
 - S(1) = 0 (no edges in a graph with a single node)
- Use iteration to solve this recurrence relation

How long does binary search take?

- We can model the running time for binary search as a recurrence relation
 - $T(n) = T(n/2) + c, k \ge 2$
 - **T**(1) = **c**
- Use iteration to solve this recurrence relation
- Instead of plugging in values 1, 2, 3,..., try powers of two: 1, 2, 4, 8,...

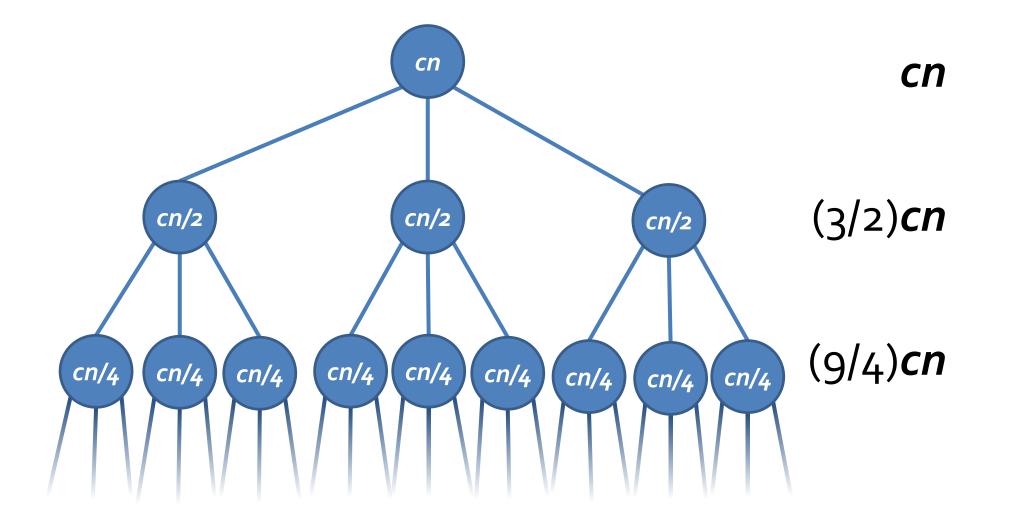
Further Recurrence Relations

Three-sentence Summary of Further Recurrence Relations

Further recurrence relations

- We have seen that recurrence relations of the form $T(n) \le 2T\left(\frac{n}{2}\right) + cn$ are bounded by O($n \log n$)
- What about $T(n) \le qT\left(\frac{n}{2}\right) + cn$ where **q** is bigger than 2 (more than two sub-problems)?
- There will still be $\log_2 n$ levels of recursion
- However, there will not be a consistent *cn* amount of work at each level

Consider q = 3



Converting to summation

• For
$$\boldsymbol{q} = 3$$
, it's $T(n) \le \sum_{j=0}^{\log_2 n-1} \left(\frac{3}{2}\right)^j cn$

In general, it's

$$T(n) \leq \sum_{j=0}^{\log_2 n-1} \left(\frac{q}{2}\right)^j cn = cn \sum_{j=0}^{\log_2 n-1} \left(\frac{q}{2}\right)^j$$

This is a geometric series, where $r = \frac{q}{2}$

$$T(n) \leq cn \left(\frac{r^{\log_2 n} - 1}{r - 1}\right) \leq cn \left(\frac{r^{\log_2 n}}{r - 1}\right)$$

Final bound

$$T(n) \leq cn\left(\frac{r^{\log_2 n} - 1}{r - 1}\right) \leq cn\left(\frac{r^{\log_2 n}}{r - 1}\right)$$

• Since r - 1 is a constant, we can pull it out
• $T(n) \leq \left(\frac{c}{r-1}\right)nr^{\log_2 n}$
• For $a > 1$ and $b > 1$, $a^{\log b} = b^{\log a}$, thus $r^{\log_2 n} = n^{\log_2 r} = n^{\log_2 (q/2)} = n^{(\log_2 q) - 1}$
• $T(n) \leq \left(\frac{c}{r-1}\right)n \cdot n^{(\log_2 q) - 1} \leq \left(\frac{c}{r-1}\right)n^{\log_2 q}$ which is $O(n^{\log_2 q})$

What about a single sub-problem?

- We will still have $\log_2 n 1$ levels
- However, we'll cut our work in half each time

$$T(n) \le T\left(\frac{n}{2}\right) + cn \le \sum_{j=0}^{\log_2 n-1} \left(\frac{1}{2}\right)^j cn = cn \sum_{j=0}^{\log_2 n-1} \frac{1}{2^j}$$

- Summing all the way to infinity, $1 + \frac{1}{2} + \frac{1}{4} + \cdots = 2$
- Thus, $T(n) \leq 2cn$ which is O(n)

What might that look like in code?

Here's a non-recursive version in Java

```
int counter = 0;
for( int i = 1; i <= n; i *= 2 )
    for( int j = 1; j <= i; j++ )
        counter++;</pre>
```

 We've just shown that this is O(n), in spite of the two for loops



Upcoming



Counting inversions

Reminders

- Assignment 3 is due on Friday
- Read section 5.3
- Extra credit opportunities (0.5% each):
 - Hristov teaching demo: 2/19 11:30-12:25 a.m. in Point 113
 - Hristov research talk: 2/19 4:30-5:30 p.m. in Point 139