

Lab 2 Continued, Problems 1- 6

- (1) Prove by induction that for all $n > 4$, $F_n > \left(\frac{4}{3}\right)^n$. Then use this result to explain the approximate asymptotic running time of the recursive algorithm for computing the Fibonacci numbers. Is the recursive Fibonacci algorithm fast or slow? Why?

We use the hint and first prove by induction that, for all $n > 1$,

$$\phi(n) : \quad \left(\frac{4}{3}\right)^{n-2} + \left(\frac{4}{3}\right)^{n-1} > \left(\frac{4}{3}\right)^n.$$

Since

$$1 + \left(\frac{4}{3}\right)^1 = \frac{7}{3} = \frac{21}{9} > \frac{16}{9} = \left(\frac{4}{3}\right)^2,$$

the base case $\phi(2)$ holds.

Assuming the result holds for all $j < n$ (that is, $\phi(j)$ holds for $j < n$), where $n > 2$, we have

$$\left(\frac{4}{3}\right)^n = \frac{4}{3} \cdot \left(\left(\frac{4}{3}\right)^{n-1}\right) < \frac{4}{3} \cdot \left(\left(\frac{4}{3}\right)^{n-3} + \left(\frac{4}{3}\right)^{n-2}\right) = \left(\frac{4}{3}\right)^{n-2} + \left(\frac{4}{3}\right)^{n-1},$$

as required.

We use this result to solve the problem. Let $\psi(n)$ be the statement " $F_n > \left(\frac{4}{3}\right)^n$." We show by induction that $\psi(n)$ holds for all $n > 4$. The base case follows immediately from the table provided in the exercise: $F_5 = 5 > 4.21 = \left(\frac{4}{3}\right)^5$. For the induction step, assume the result hold for all $j < n$. Then

$$F_n = F_{n-2} + F_{n-1} > \left(\frac{4}{3}\right)^{n-2} + \left(\frac{4}{3}\right)^{n-1} > \left(\frac{4}{3}\right)^n.$$

- (2) (a) False:

$$\lim_{n \rightarrow \infty} \frac{4^n}{2^n} = \lim_{n \rightarrow \infty} \left(\frac{4}{2}\right)^n = \lim_{n \rightarrow \infty} 2^n = \infty.$$

(b) True: It is enough to prove that $\lim_{n \rightarrow \infty} \frac{\log n}{\log_3 n}$ is finite but nonzero:

$$\lim_{n \rightarrow \infty} \frac{\log n}{\log_3 n} = \lim_{n \rightarrow \infty} \frac{\log n}{\frac{\log n}{\log 3}} = \log 3,$$

as required.

(c) True: It is enough to prove that $\lim_{n \rightarrow \infty} \frac{(n/2) \log n/2}{n \log n}$ is finite nonzero.

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{(n/2) \log n/2}{n \log n} &= \lim_{n \rightarrow \infty} \frac{\log n/2}{2 \log n} \\ &= \lim_{n \rightarrow \infty} \frac{\frac{1}{2} \frac{1}{n} \cdot \log e}{\frac{2 \log e}{n}} \\ &= \frac{1}{2}, \end{aligned}$$

as required. The second to last line was obtained from the line previous to it using L'Hopital's Rule.

- (3) Below, pseudo-code is given for the recursive factorial algorithm `recursiveFactorial`. Use the Guessing Method to determine the worst-case asymptotic running time of this algorithm. Then verify correctness of your formula.

```

Algorithm recursiveFactorial(n)
  Input: A non-negative integer n
  Output: n!
  if (n = 0 || n = 1) then
    return 1
  return n * recursiveFactorial(n-1)

```

Step 1: Formulate the recurrence relation. We aren't concerned with the running time when $n = 0$, so we consider positive n only.

$$T(n) = \begin{cases} 4 & \text{if } n = 1 \\ T(n-1) + 4 & \text{if } n > 1 \end{cases}$$

Step 2: Guess a closed form solution.

$$\begin{aligned} T(1) &= 4 \\ T(2) &= T(1) + 4 = 4 + 4 \\ T(3) &= T(2) + 4 = 4 + 4 + 4 \\ T(n) &= T(n-1) + 4 = 4n. \end{aligned}$$

Step 3: Prove the formula from Step 2 is a solution.

Let $f(n) = 4n$. We show $f(1) = 4$ and $f(n) = f(n-1) + 4$. The first part is obviously true. We also have

$$f(n) = 4n = 4(n-1) + 4 = f(n-1) + 4,$$

as required.

Step 4: Prove correctness.

- (a) *Verify valid recursion.* The recursion is valid because “ $n == 0 \parallel n == 1$ ” is the base case and repeated self-calls in the algorithm lead to the base case since each self-call reduces the input size by 1.
 - (b) *Verify base case outputs are correct.* This follows since $0! = 1$ and $1! = 1$.
 - (c) *Verify inductively that outputs are correct for all n .* Assume `recursiveFactorial(j)` outputs $j!$ for all $j < n$, where $n > 1$. Then `recursiveFactorial` on input n returns `recursiveFactorial(n-1) * n`, which, by inductive hypothesis, is $(n-1)! * n = n!$.
- (4) Devise an iterative algorithm for computing the Fibonacci numbers and compute its running time. Prove your algorithm is correct.

Below is an iterative Java method that computes F_n on input n . It executes a single loop that depends on n , so running time is $O(n)$. It also uses $O(n)$ space, using the `store` array.

```
int[] store;
//precondition: n is non-negative integer
public int fib(int n) {
    store = new int[n+1];
    store[0] = 0;
    store[1] = 1;

    //Loop Invariant: I(i): store[i] = F_i
    for(int i = 2; i <= n; i++) {
        store[i] = store[i-1] + store[i-2];
    }
    //postcondition: store[n] = F_n

    return store[n];
}
//postcondition: F_n is returned
```

We verify correctness. We first establish a loop invariant I and show that $I(k)$ holds at the end of the $i = k$ pass, for $2 \leq k \leq n$. Our loop invariant is:

$$I(i) : \text{store}[i] = F_i$$

For the Base Case, we establish $I(2)$ at the end of the $i = 2$ pass. But this is clear since $I[0] = F_0$ and $I[1] = F_1$ and $\text{store}[2] = \text{store}[0] + \text{store}[1]$.

For the Induction Step, assume $I(j)$ holds at the end of the $i = j$ pass for each $j < k$, where $2 \leq k \leq n$; we show $I(k)$ holds at the end of the $i = k$ pass. By the Induction Hypothesis, $\text{store}[k-1] = F_{k-1}$ and $\text{store}[k-2] = F_{k-2}$. By inspection of the algorithm, it is clear therefore that

$$\text{store}[k] = \text{store}[k-1] + \text{store}[k-2] = F_{k-1} + F_{k-2} = F_k.$$

This completes the induction and shows that the loop invariant holds for $2 \leq k \leq n$. Therefore, at the end of the $i = n$ pass, we have that $\text{store}[n]$ stores F_n , and this is the value returned by the algorithm. The algorithm is therefore correct.

- (5) We use the Master Formula to solve this recurrence relation:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + n & \text{otherwise} \end{cases}$$

Here, $a = 1, b = 2, c = 1, d = 1, k = 1$. Note $a < b^k$. The Master Formula tells us therefore that $T(n) \in \Theta(n)$.

- (6) See the Java file ZeroesAndOnes.java. Since BinarySearch is used to locate a 0 followed by a 1, this part of the algorithm takes $O(\log n)$, and the rest takes constant time.
We can also easily show that f/g reaches 0 as n tends to infinity. Hence, it is $o(n)$.

