

4005-800 ALGORITHMS

HOMEWORK 2

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PROBLEM 1.

Solution.

The time complexity of the recurrence F_n can be defined as follows.

$$\begin{aligned}T_F(0) &= 1 \\T_F(1) &= 1 \\T_F(n) &= T_F(n-1) + T_F(n-2)\end{aligned}$$

The solution to $T_F(n)$ can be solved using the method of homogeneous equations, which yields the result that $T_F(n) = \Theta(\phi^n)$, where $\phi = \frac{1+\sqrt{5}}{2}$ (the golden ratio).

PROBLEM 2.

Solution.

The time complexity of the *fibIt* routine can be found by solving the recurrence relation that defines *fibIt*. More specifically, we can such a recurrence relation for *fibIt* by analyzing the number of additions performed, which corresponds to the following equation.

$$\begin{aligned}T_f(0) &= 0 \\T_f(1) &= 0 \\T_f(n) &= T_f(n-1) + 1\end{aligned}$$

This is because there is only one addition made in each recursive call from $f(n; a, b)$ to $f(n-1; b, a+b)$, and there are no additions made in the two cases where $n=0$ and $n=1$.

In order to solve this recurrence relation we can expand out the expression and attempt to identify the pattern. This process is shown below.

$$\begin{aligned}
T_f(n) &= T_f(n-1) + 1 \\
&= (T_f(n-2) + 1) + 1 = T_f(n-2) + 2 \\
&= (T_f(n-3) + 1) + 2 = T_f(n-3) + 3 \\
&= \dots \\
&= (T_f(n-k) + 1) + k = T_f(n-k) + k
\end{aligned}$$

Based on this pattern, we can reach the first base case of this recurrence relation ($T_f(1)$) when $(n-k) = 1$, meaning that $k = (n-1)$. Thus, we have the following.

$$\begin{aligned}
T_f(n) &= T_f(n - (n-1)) + (n-1) \\
&= T_f(1) + (n-1) \\
&= 0 + (n-1) \\
&= n-1
\end{aligned}$$

Based on this observation we can see that $T_f(n) \in \Theta(n)$, or simply $T_f(n) = \Theta(n)$.

PROBLEM 3.

Solution.

Base ($n = 0$)

When $n = 0$, we have the following equality.

$$\begin{aligned}
L_0(a, b) &= (f(0; a, b), f(1; a, b)) \\
&= (a, b)
\end{aligned}$$

Induction ($n > 0$)

First, we assume that $L^n(a, b) = (f(n; a, b), f(n+1; a, b))$. Now we show that $L^{n+1}(a, b) = (f(n+1; a, b), f(n+2; a, b))$.

$$\begin{aligned}
L^{n+1}(a, b) &= L(L^n(a, b)) \text{ (by multiplication powers)} \\
&= L(f(n; a, b), f(n+1; a, b)) \text{ (by induction)} \\
&= (f(n+1; a, b), f(n; a, b) + f(n+1; a, b)) \text{ (by definition of } L) \\
&= (f(n+1; a, b), f(n+2; a, b)) \text{ (by Theorem 1)}
\end{aligned}$$

Thus, $L^{n+1}(a, b) = (f(n+1; a, b), f(n+2; a, b))$, as desired. Therefore, we know that $f(n; a, b) = (L^n(a, b))_1$.

PROBLEM 4.

Solution. TODO: talk about representation, mention multiplication algorithm for logn time, give fibpow listing, list time complexity

TODO: matrix

TODO: repeated squaring

TODO: code here

TODO: time complexity

PROBLEM 5-a. Write down the definition of pseudo-polynomial time.

Solution.

Definition 1. A pseudo-polynomial-time algorithm is one that runs in time polynomial in the dimension of the problem and the magnitudes of the data involved (provided these are given as integers), rather than the base-two logarithms of their magnitudes .

PROBLEM 5-b. Is *fib* a pseudo-polynomial time algorithm? Explain.

Solution. No - exponential in value of input (not polynomial).

No, *fib* has a time complexity of $\Theta(\phi_n)$, where n is the input magnitude, which means that it is exponential in the magnitude of the input, which further implies that it is not polynomial in the magnitude of the input. Therefore, by definition, *fib* cannot be a pseudo-polynomial time algorithm.

PROBLEM 5-c. Is *fibIt* a pseudo-polynomial time algorithm? Explain.

Solution. Yes - polynomial in value of input, which is exponential in number of bits in the input (base 2 logarithm)

Yes, *fibIt* has a time complexity of $\Theta(n)$, which can also be defined as $\Theta(2^{\lg n})$, where n is the input magnitude, which means that it also has polynomial time complexity in the magnitude of n but exponential time in the bit size of n . Therefore, by definition, *fibIt* must be a pseudo-polynomial time algorithm.

PROBLEM 5-d. Is *fibPow* a pseudo-polynomial time algorithm? Explain.

Solution. Don't believe so, it isn't exponential in the magnitude of the input (see below) Is $2^{*(\lg \lg n)}$ exponential? If so, then yes, if not, then no.

No, *fibPow* has a time complexity of $\Theta(\lg n)$, which can also be defined as $\Theta(2^{\lg \lg n})$, where n is the input magnitude, which means that it is sub-polynomial in the magnitude of n and sub-exponential in the bit size of n . Therefore, by definition, *fibPow* must be a pseudo-polynomial time algorithm.