In this handout, we discuss a typical situation in the analysis of algorithms: the number of steps required by the algorithm satisfies some recurrent inequality; from this we want to infer an upper bound on the order of magnitude of the number of steps. We illustrate the method on a specific recurrent inequality which will occur in class in the analysis of a “branch-and-bound” algorithm.

Suppose we have an algorithm which takes at most $T(n)$ steps on all inputs of size $n$. Suppose in addition that $T(n)$ is known to satisfy the following recurrent inequality:

$$T(n) \leq T(n-1) + T(n-2) \quad (n \geq 2) \quad (1)$$

How can we use this information to give a good upper bound on $T(n)$?

**Theorem 1.** Suppose the function $g(n) > 0$ has the following property: there exists a threshold value $n_0$ such that for all $n \geq n_0$ the inequality

$$g(n) \geq g(n-1) + g(n-2) \quad (2)$$

holds. Then $T(n) = O(g(n))$.

**Warning.** Note that inequality (2) is the same as inequality (1) except that it goes in the opposite direction!

**Proof:** First, let us choose a positive constant $C$ such that

(a) $Cg(n_0) \geq T(n_0)$, and

(b) $Cg(n_0 + 1) \geq T(n_0 + 1).$

(The choice $C := \max\{T(n_0)/g(n_0), T(n_0 + 1)/g(n_0 + 1)\}$ will do.) Now we claim that for all $n \geq n_0$, the inequality

$$T(n) \leq Cg(n) \quad (3)$$

holds. This clearly justifies the $T(n) = O(g(n))$ claim. (Note that on the basis of the data given, we have no information about the magnitude of the constant $C$ implicit in the big-oh notation.)
The proof of inequality (3) is an easy application of mathematical induction. By inequalities (a) and (b) we know that inequality (3) holds for \( n = n_0 \) and \( n = n_0 + 1 \) (starting cases).

For the inductive step, let now \( n \geq n_0 + 2 \) and assume that \( T(k) \leq Cg(k) \) holds for all \( k \) in the interval \( n_0 \leq k \leq n-1 \). Under this inductive hypothesis we need to prove that \( T(n) \leq Cg(n) \).

Indeed,

\[
T(n) \leq T(n-1) + T(n-2) \leq Cg(n-1) + Cg(n-2) \leq Cg(n). \tag{4}
\]

(The first inequality is just inequality (1); the second inequality holds by the inductive hypothesis; and the third inequality comes from inequality (2).) This completes the inductive step, and thereby the inductive proof of Theorem 1. Q. E. D. [End of proof, from the Latin “Quod erat demonstrandum.”]

Questions.

1. Why did we need to verify two starting cases \( (n = n_0, n_0 + 1) \) (rather than just one starting case)?

2. Why did \( g(n) \) need to satisfy the reverse of the recurrent inequality satisfied by \( T(n) \)? (Where and how did the proof exploit this condition?)

3. Prove that \( T(n) \leq 2T(n-1) \) for \( n \geq 4 \).

4. Infer from item 3 that \( T(n) = O(2^n) \).

5. Prove that \( g(n) \geq 2g(n-2) \) for \( n \geq n_0 + 3 \).

6. Infer from item 5 that \( g(n) = \Omega(2^{n/2}) \).

The next step in evaluating the recurrence (1) is to find some function \( g(n) \) satisfying inequality (2). Item 4 above suggests the choice \( g(n) := 2^n \). Indeed, \( 2^n > 2^{n-1} + 2^{n-2} \) (why?). But this may not be the smallest such \( g(n) \). The upper bound in item 4 and the lower bound in item 6 together suggest that we should be looking for an exponential function, of the form \( g(n) = \alpha^n \) (geometric progression) for some fixed \( \alpha \), and try to find the smallest such \( \alpha \) in order to obtain the best possible upper bound. (From item 6 we know that any such \( \alpha \) must be \( \geq \sqrt{2} \).) The condition then is

\[
\alpha^n \geq \alpha^{n-1} + \alpha^{n-2}, \tag{5}
\]

[End of proof.]
or equivalently (dividing by $\alpha^{n-2}$),

$$\alpha^2 \geq \alpha + 1.$$  \hspace{1cm} (6)

Recall that we wish to minimize $\alpha$. The smallest value of $\alpha > 1$ that satisfies (6) actually gives equality:

$$\alpha^2 = \alpha + 1.$$  \hspace{1cm} (7)

This is a quadratic equation. Of the two solutions, one is negative and therefore has no meaning in our context; the other solution is

$$\alpha = \phi := \frac{1 + \sqrt{5}}{2} \approx 1.6180339$$  \hspace{1cm} (8)

(this is the golden ratio).

The following summarizes our conclusion:

**Theorem 2.** If the function $T(n) \geq 0$ satisfies inequality (1) then

$$T(n) = O(\phi^n)$$

where $\phi$ is the golden ratio (equation (8)).

*Comment.* Of course $T(n)$ could be much smaller; what we have is an upper bound. However, this is the best upper bound that can be inferred from the information given (inequality (1)), as demonstrated by the example of the Fibonacci sequence in the role of $T(n)$.

*Remark.* The Fibonacci sequence is the sequence defined by the recurrence $F_n = F_{n-1} + F_{n-2}$ with initial values $F_0 = 0$ and $F_1 = 1$. (So the first few terms of the sequence are 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89.) The asymptotic value of $F_n$ is $F_n \sim \phi^n/\sqrt{5}$.

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"Divide and Conquer" algorithms are the most frequent sources of recurrent inequalities. We illustrate the method on the recurrence arising from the Karatsuba–Ofman integer multiplication algorithm.

For simplicity we assume $n = 2^k$. Let $B(n)$ be the number of bit-operations required by the K-O algorithm to multiply two $n$-digit integers. We do not ignore additions/subtractions. The inequality then is:

$$B(n) \leq 3B(n/2) + O(n).$$  \hspace{1cm} (9)
We cannot handle the $O(n)$ term directly because it is not a well-defined quantity: all we know about it is that it is between $-Cn$ and $Cn$ for some positive constant $C$ (for all $n \geq 1$). With this (unknown, but constant) value of $C$ we obtain the inequality

$$B(n) \leq 3B(n/2) + Cn.$$  \hspace{1cm} (10)

Now all terms of the inequality are sufficiently well defined to allow us to do basic operations with them.

**Theorem 3.** If the function $B(n) \geq 0$ satisfies inequality \eqref{10} then $B(n) = O(n^\alpha)$ where $\alpha = \log 3 \approx 1.58$.

Again, we need to find a function $g(n)$ which satisfies the reverse of inequality \eqref{10} for $n \geq n_0$ and $g(n) > 0$ for all $n \geq 1$. Then, as before, we shall know that $B(n) = O(g(n))$.

We try to find $g(n)$ in the form $g(n) = An^\alpha - Dn$ for some constants $A > 0$ and $D$. The reason for this choice will be apparent from the calculations below; here we only note that we are entitled to look for $g(n)$ in any form we like; the eventual success justifies a good choice.

For our choice to be good, we need to be able to find a value of the constant $D$ such that

$$An^\alpha - Dn \geq 3 \left( A(n/2)^\alpha - Dn/2 \right) + Cn$$

holds for all $n$.

Observing that $n^\alpha = 3(n/2)^\alpha$ (this equality motivated the choice of the exponent $\alpha$), our inequality reduces to

$$-Dn \geq -3Dn/2 + Cn,$$

i.e.,

$$Dn/2 \geq Cn.$$  

Let us therefore choose $D := 2C$, and the required inequality is satisfied. Note that this holds \textit{regardless of the value of $A$}. We now set the value of $A$ sufficiently large such that $g(n) > 0$ for all $n \geq 1$. \textit{Exercise}. Prove that such a choice of $A$ is possible. (Use the fact that $n = o(n^\alpha)$.)

Therefore $B(n) = O(g(n)) = O(An^\alpha - Dn) = O(n^\alpha)$. This concludes the proof of Theorem 3. Q. E. D.