CIS 575. Introduction to Algorithm Analysis Material for February 26, 2024

Multiplying Large Integers

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1 Multiplying Large Integers

We shall consider the task

Multiply two *n*-bit positive integers

One may ask: given that all computers provide multiplication at machine code level, why do we bother about this? But recall that a given computer model uses only a certain number of bits to represent each integer; if 2n is bigger than that number then we cannot directly use the multiplication operation, but may have to implement our own. For that purpose, the divide \mathcal{E} conquer paradigm can be used, as we shall now show. To make the examples more appealing, we shall use base 10 rather than base 2 to represent integers, and thus consider n-digit integers.

First recall (from primary school!) that if we know how to multiply 1-digit integers then we can also multiply 2-digit integers: to multiply say 47 and 23, we perform 4 multiplications of 1-digit integers, suitably align the results, and add them:

Similarly, if we know how to multiply 2-digit integers then we can also multiply 4-digit integers, as illustrated by the following computation of **2043** times **2512**:

It should be obvious how to generalize this so as to write an algorithm that multiplies two n-digit integers by 4 multiplications of $\frac{n}{2}$ -digit integers. (By adding leading zeros we can

assume that the number of digits in an integer is a power of 2.) Since addition can be done in time $\Theta(n)$, for the running time T(n) we get the recurrence

$$T(n) = 4T(\frac{n}{2}) + \Theta(n)$$

and thus (as $\log_2(4) = 2$) we have $T(n) \in \Theta(n^2)$.

But we can do better: in the second example, multiplying **2043** and **2512**, we can do a little conjuring trick:

While we have increased the number of additions, and thus likely also increased the running time for integers with only few digits, there are now only 3 multiplications¹ and this will improve asymptotic running time: we get the recurrence

$$T(n) = 3T(\frac{n}{2}) + \Theta(n) \tag{1}$$

which has solution $T(n) \in \Theta(n^{\lg(3)})$.

To understand why our magic works, observe that: if we have two integers P and Q where

$$P = w \cdot 10^n + y$$
$$Q = x \cdot 10^n + z$$

then

$$P \cdot Q = wx \cdot 10^{2n} + (wz + yx) \cdot 10^n + yz$$

but

$$wz + yx = (w+y)(x+z) - wx - yz$$

and thus the only multiplications needed are: $\mathbf{w}\mathbf{x}$, $\mathbf{y}\mathbf{z}$, and $(\mathbf{w} + \mathbf{y})(\mathbf{x} + \mathbf{z})$.

It is actually possible to get an even better asymptotic running time (at the price of running slower for "small" numbers), as demonstrated in Exercise 10.4 in the *Howell* textbook (where the context is the multiplication of polynomials). For any $k \geq 2$, it is possible to get the the recurrence (when k = 2 this is just (1))

$$T(n) = (2k - 1)T(\frac{n}{k}) + \Theta(n)$$

which has solution $T(n) \in \Theta(n^{\log_k(2k-1)})$ where the exponent $\log_k(2k-1)$ will approach 1 as k approaches infinity.

We can thus get as close to linear time as we want. But we cannot do better than linear time: any algorithm has to generate at least n digits, and hence must have running time in $\Omega(n)$.

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One of which could have involved 3-digit numbers, if we had chosen say 6043 rather than 2043.