

Algorithms with numbers (1)

CISC5835, Computer Algorithms

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Acknowledgement

- The set of slides have used materials from the following resources
 - Slides for textbook by Dr. Y. Chen from Shanghai Jiaotong Univ.
 - Slides from Dr. M. Nicolescu from UNR
 - Slides sets by Dr. K. Wayne from Princeton
 - which in turn have borrowed materials from other resources
 - Other online resources

Outline

- Motivation
- Algorithm for integer addition
- Algorithms for multiplication
 - grade-school algorithm
 - recursive algorithm
 - divide-and-conquer algorithm
- Division
- Exponentiation

Algorithms for integer arithmetics

- We study adding/multiplying two integers
 - earliest algorithms!
 - mostly what you learned in grade school!
- Analyze running time of these algorithms by counting number of elementary operations on individual bits when adding/multiplying two N-bits long ints (so called **bit complexity**)
 - **input size N: the length of operands**
 - for example, to add two N-bits integer numbers, we need to $O(n)$ bit operations (such as adding three bits together).

Practical consideration

- But, why bother?
 - Given that with a single (machine) instruction, one can add/subtract/multiply integers whose size in bits is within word length of computer – 32, or 64.
 - i.e., they are implemented in hardware
- **Bit complexity** of arithmetic operations algorithms captures amount of hardware (transistors and wires) necessary for implementing algorithm using digital logic circuit.
 - e.g., number of logic gates needed ...

Support for Big Integer

- What if we need to handle numbers that are several thousand bits long?
 - need to implement arithmetic operations of large integers in software.
- Ex: Use an array of ints to store the (decimal or binary) digits of integer,
 - `int digits[3]={2,4,6}; //represents 642`
 - `int digits1[10]={3,4,5,7,0,7,8}; //represent 8707543`
 - `int bindigits[4]={1,0,1,0}; //represent 0101, i.e., 3`
- Algorithms studied here are presented assuming base 2
 - those for other base (e.g., base 10) are similar

Support for Big Integer*

- But notice that each int variable can store up to 64 bits, and we can add/subtract/multiply 64-bits ints in one machine instruction
- To save space and time, one could divide big integer into chunks of 63 bits long, and store each chunk in a int

• 101100... 10 1111110...101110 111111...0000110011110111
63 bits 63 bits 63 bits

- `int chunks[3]={32, 121254, 145246};`
`//represent a value of`

$$32 \times 2^{126} + 121254 \times 2^{63} + 145246$$

When adding two numbers, adding corresponding chunks together, carry over are added to the next chunks...

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Adding two binary number

Carry:	1			1	1	1		
		1	1	0	1	0	1	(53)
		1	0	0	0	1	1	(35)
<hr/>								
	1	0	1	1	0	0	0	(88)

Algorithm for adding integers

- Sum of any three single-digit numbers is at most two digits long. (holds for any base)
 - In binary the largest possible sum of three single-bit numbers is 3, which is a 2-bit number.
 - In decimal, the max possible sum of three single digit numbers is 27 (9+9+9), which is a 2-digit number
- **Algorithms for addition (in any base):**
 - align their right-hand ends,
 - perform a single right-to-left pass
 - the sum is computed digit by digit, maintaining overflow as a **carry**
 - since we know each individual sum is a two-digit number, the carry is always a single digit, and so at any given step, three single-digit numbers are added)

Question

Given two binary numbers x and y , how long does our algorithm take to add them?

We want the answer expressed as a **function of the size of the input**: the number of bits of x and y .

Suppose x and y are each n bits long. Then the sum of x and y is $n + 1$ bits at most, and each individual bit of this sum gets computed in a fixed amount of time.

The total running time for the addition algorithm is therefore of the form $c_0 + c_1 n$, where c_0 and c_1 are some constants, i.e., $O(n)$.

Question

Is there a *faster* algorithm?

In order to add two n -bit numbers we must at least read them and write down the answer, and even that requires n operations.

So the addition algorithm is *optimal*, up to multiplicative constants!

Ubiquitous $\log_2 N$

- $\log_2 N$ is the power to which you need to **raise 2** in order to obtain N .
 - e.g., $\log_2 8 = 3$ (as $2^3 = 8$), $\log_2 1024 = 10$ (as $2^{10} = 1024$).
- Going backward, it can be seen as the number of times you must **halve N** to get down to 1, more precisely: $\lceil \log_2 N \rceil$
 - e.g., $N=10$, $\lceil \log_2 10 \rceil = 4$, $N=8$, $\lceil \log_2 8 \rceil = 3$
- It is the number of bits in binary representation of N , more precisely: $\lceil \log_2(N + 1) \rceil$
 - e.g., hw1 questions
- It is the depth of a complete binary tree with N nodes, more precisely: $\lceil \log_2 N \rceil$
 - height of a heap with N nodes ...
- It is even the sum $1 + 1/2 + 1/3 + \dots + 1/N$, to within a constant factor.

Multiplication in base 2

				1	1	0	1	(binary 13)
			×	1	0	1	1	(binary 11)
<hr/>								
				1	1	0	1	(1101 times 1)
			1	1	0	1		(1101 times 1, shifted once)
		0	0	0	0			(1101 times 0, shifted twice)
+	1	1	0	1				(1101 times 1, shifted thrice)
<hr/>								
1	0	0	0	1	1	1	1	(binary 143)

The grade-school algorithm for multiplying two numbers x and y:

- create an array of **intermediate sums**, each representing the product of x by a single digit of y.
- these values are appropriately **left-shifted** and then added up.

Multiplication in base 2

$$\begin{array}{rcccccccc} & & & & & 1 & 1 & 0 & 1 & & \text{(binary 13)} \\ & & & & & \times & 1 & 0 & 1 & 1 & \text{(binary 11)} \\ \hline & & & & & & 1 & 1 & 0 & 1 & \text{(1101 times 1)} \\ & & & & & 1 & 1 & 0 & 1 & & \text{(1101 times 1, shifted once)} \\ & & & 0 & 0 & 0 & 0 & & & & \text{(1101 times 0, shifted twice)} \\ + & & 1 & 1 & 0 & 1 & & & & & \text{(1101 times 1, shifted thrice)} \\ \hline 1 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & & & \text{(binary 143)} \end{array}$$

If x and y are both n bits, then there are n intermediate rows, with lengths of up to $2n$ bits.

The total time taken to add up these rows, doing two numbers at a time, is

$$\underbrace{O(n) + \dots + O(n)}_{n - 1 \text{ times}}$$

which is $O(n^2)$.

Multiplication: top-down approach

```
MULTIPLY(x, y)  
// Two n-bit integers x and y, where  $y \geq 0$ .  
1.  if  $y = 0$  then return 0  
2.   $z = \text{MULTIPLY}(x, \lfloor y/2 \rfloor)$   
3.  if y is even then return 2 $z$   
4.           else return  $x + 2z$ 
```

- Totally, n recursive calls, because at each call y is halved (i.e., n decreases by 1)
- In each recursive call: a division by 2 (right shift), a test for odd/even (looking up the last bit); a multiplication by 2 (left shift); and possibly one addition \Rightarrow a total of $O(n)$ bit operations.
- The total time taken is thus $O(n^2)$.

Divide-and-conquer

Suppose x and y are two n -bit integers, and assume for convenience that n is a *power of 2*.

Lemma

For every n there exists an n' with $n \leq n' \leq 2n$ such that n' a power of 2.

As a first step toward multiplying x and y , we split each of them into their **left and right halves**, which are $n/2$ bits long:

$$x = \begin{array}{|c|c|} \hline x_L & x_R \\ \hline \end{array} = 2^{n/2}x_L + x_R$$
$$y = \begin{array}{|c|c|} \hline y_L & y_R \\ \hline \end{array} = 2^{n/2}y_L + y_R.$$

$$xy = (2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R) = 2^n x_L y_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R.$$

The additions take linear time, as do the multiplications by powers of 2. The significant operations are the four $n/2$ -bit **multiplications**; these we can handle by *four recursive calls*.

Running time

$$xy = (2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R) = 2^n x_L y_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R.$$

Our method for multiplying n-bit numbers

1. making recursive calls to multiply these four pairs of n/2-bit numbers,
2. evaluates the above expression in $O(n)$ time

Writing $T(n)$ for the overall running time on n-bit inputs, we get the recurrence relation:

$$T(n) = 4T(n/2) + O(n)$$

By master theorem, $T(n) = O(n^2)$

Can we do better?

$$xy = (2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R) = 2^n x_L y_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R.$$

By **Gauss's** trick, three multiplications, $x_L y_L$, $x_R y_R$, and $(x_L + x_R)(y_L + y_R)$, suffice, as

$$x_L y_R + x_R y_L = (x_L + x_R)(y_L + y_R) - x_L y_L - x_R y_R.$$

The recurrence relation:

$$T(n) = 3T(n/2) + O(n)$$

By Master Theorem: $T(n) = n^{\log_2 3}$

Integer Division

DIVIDE(x, y)

// Two n -bit integers x and y , where $y \geq 1$.

1. **if** $x = 0$ **then** return $(q, r) = (0, 0)$
2. $(q, r) = \text{DIVIDE}(\lfloor x/2 \rfloor, y)$
3. $q = 2 \cdot q, r = 2 \cdot r$
4. **if** x is odd **then** $r = r + 1$
5. **if** $r \geq y$ **then** $r = r - y, q = q + 1$
6. return (q, r)

Readings



- Chapter 1.1