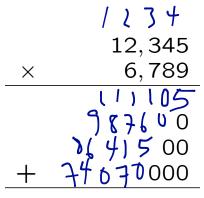
## **Binary Multiplication**

Q: How do we *multiply* two numbers?

eg.



83, 810, 205

	10111
×	– 10101 د ا
	12/0/11
	>000000
	60000
+	0000
)	1100011

Padymuldiply, add.

Consider the following algorithm to multiply two binary numbers.

```
PRECONDITION:
  x and y are binary bit arrays.
POSTCONDITION:
  Returns result a binary bit array equal to
  the product of x and y.
def MULTIPLY(x, y):
  result = [0];
  for i in range(len(y)-1, -1, -1):
    if y[i] == 1:
      result = BINARY_ADD(result, x)
      x.append(0) #add a 0 to the end of x
  return result
```

**Q:** If we *measure complexity* by the number of *bit operations*, what is the *worst case complexity* of MULTIPLY?

loop n times, each loop does an n-bit binary add so EO(n2)

Q: Is there a more efficient way to implement the multiplication?

yes.

### **Divide and Conquer Multiplication**

Notice that

$$010111 = 010000 + 111 = 010 \cdot 2^{n/2} + 111$$

 $011101 = 011000 + 101 = 011 \cdot 2^{n/2} + 101$ 

i.e., we can *split* a binary number into n/2 high bits and n/2 low bits.

**Q:** What is  $010111 \times 011101$  written in terms of *high bits* and *low bits*?

$$f''_{2}(10 \times 101 \times 101) = 101110 \times 1010 \times 1010)$$

$$f'_{2}(10 \times 101 \times 101) + 101 \times 101)$$

$$f(10 \times 101)$$

Q: What is the complexity of multiplying a number x by  $2^{n/2}$ ?  $just a shift left by \eta_2 05.50 O(\eta_2)$ 

In general,  $x \times y$  in terms of *low bits*  $\{x_l, y_l\}$  and *high bits*  $\{x_h, y_h\}$  is:

$$x \times y = \chi_{n} y_{n} 2^{n} + (\chi_{n} y_{\ell} + \chi_{\ell} y_{n}) \cdot 2^{n/2} + \chi_{\ell} y_{\ell}$$

So we can define a *recursive, divide and conquer, multiplication* algorithm.

#### PRECONDITION:

x and y are both binary bit vectors of length n (n a power of 2).

#### POSTCONDITION:

```
Returns a binary bit vector equal to
the product of x and y.
```

```
REC_MULTIPLY (x, y):

if (len(x) == 1):

return ([x[0]*y[0]])

xh = x[n/2:n] X \_ 0 : n/2]

xl = x[0:n/2] X \_ n] y \_ 0 : n/2]

yh = x[n/2:n] y \_ 0 : n/2]

yl = x[0:n/2] y \_ 0 : n/2]

a = Rec_MULTIPLY(xh, yh)

b = Rec_MULTIPLY(xh, yh)

c = Rec_MULTIPLY(xh, yh)

d = Rec_MULTIPLY(xl, yh)

d = Rec_MULTIPLY(xl, yh)

d = Rec_MULTIPLY(xl, yl)

b = BINARY_ADD(b, c)

a = shift(a, n)

b = shift(b, n/2)

return BINARY_ADD(a, b, d)

end Rec_MULTIPLY
```

Q: What is the recurrence relation for the complexity of REC\_MULTIPLY?

$$T(n) = 4T(\frac{n}{2}) + O(n) \qquad T(l) = C$$
Q: What is the *worst case complexity* of REC\_MULTIPLY?
$$O(n^2)$$

This is a bit disappointing...

## A Better Divide and Conquer Multiplication Algorithm

Recall we want to compute:

$$x_h y_h \cdot 2^n + (x_l y_h + x_h y_l) \cdot 2^{n/2} + x_l y_l$$

### **Observation** [Gauss]

$$x_l y_h + x_h y_l = (x_h + x_l)(y_h + y_l) - x_h y_h - x_l y_l$$

Q: Why is this true?  

$$(x_{h} + x_{l})(y_{h} + y_{l}) = X_{h} y_{h} + (X_{l} y_{h} + X_{h} y_{l}) + X_{l} y_{l}$$
Q: How does this help us?  
by reducing our recursive calls  
to 3 times.  
1. X<sub>h</sub> y<sub>h</sub>  
2. X<sub>l</sub> y<sub>l</sub>  
3.  $(X_{h} + X_{l})(Y_{h} + Y_{l})$   
Therefore,

$$xy = \chi_{h} y_{h} \cdot 2^{+} \left[ \left( \chi_{h} + \chi_{\ell} \right) \left( y_{h} + y_{\ell} \right) - \chi_{h} y_{h} - \chi_{\ell} y_{\ell} \right] \cdot 2^{+} \chi_{\ell} y_{\ell}$$

leading to a new *divided and conquer multiplication* algorithm:

~

## **Recursive Multiply – Take 2**

#### PRECONDITION:

```
x and y are both binary bit arrays of length n, n a power of 2.
```

#### POSTCONDITION:

```
Returns a binary bit array equal to the product of x and y.
```

```
REC_MULTIPLY_2(x, y):

if (len(x) == 1):

return (x[0]*y[0])

xh = x[n/2:n] (0: ^2)

xl = x[0:n/2] (12: n]

yh = x[n/2:n] y[0: n/2]

yl = x[0:n/2] (2: n/2)

yl = x[0:n/2] (2: n/2)

p1 = REC_MULTIPLY_2(xh, yh)

p2 = REC_MULTIPLY_2(xh+xl, yh+yl)

p3 = REC_MULTIPLY_2(xl, yl)

p2 = BINARY_ADD(p2, -p1, -p3)

p2 = SHIFT(p2, n/2)

p1 = SHIFT(p1, n)

return BINARY_ADD(p1, p2, p3)
```

Q: What is the *recurrence relation* for REC\_MULTIPLY\_2? T(n) = 3T(n/2) + O(n) T(1) = O(n) - C. Q: Is this really any better than T(n) = 4T(n/2) + O(n)? A: See Assignment 1!

# **Program Correctness – Chapter 2**

Proving program correctness really means proving

If some *condition* P holds at the *start* of the execution of a program, then

- the program will terminate
- some *condition* Q will hold at the end.

Condition *P* is called a *precondition*.

Condition *Q* is called a *postcondition*.

Think of this as a contract, if the *precondition is satisfied* then the progam is required to *meet the postcondition*.

**Note:** we are not concerned with *runtime errors* (e.g. overflow, division by zero). They are easier to spot.

Two cases we will consider:

- recursive programs (programs with recursive methods)
- iterative programs (programs with loops)

## **The Correctness of Recursive Programs**

Read the book, pages 47–53.

In this section, we consider how to prove correct programs that contain *recursive methods*.

We do this by using

*simple or complete induction* over the *arguments* to the recursive method.

## How to do the proof

To prove a recursive program correct (for a given precondition and a postcondition) we typically

- 1. prove the recursive *method* totally correct
- 2. prove the main *program* totally correct

## A first example

```
public class EXP
{
    int Expo(u,v) {
        if v == 0 return 1;
        else if v is even
            return SQUARE(Expo(u,v DIV 2));
        else
            return u*(SQUARE(Expo(u,v DIV 2)));
    }
    int SQUARE(x) {
        return x*x;
    }
    void main() {
        z = Expo(x,y);
    }
    *Note DIV truncates the decimals
    *Note DIV truncates the decimals
```

Note: the *main program* here does nothing but call the method on x and y

**Lemma:** For all  $m, n \in \mathbb{N}$ , the method Expo(m, n) terminates and returns the value  $m^n$ .

Proof: next page

**Theorem:** The program *EXP* is (totally) correct for *precondition*  $x, y \in \mathbb{N}$  and *postcondition*  $z = x^y$ .

**Proof:** immediate from the lemma.

**Lemma:** For all  $m, n \in \mathbb{N}$ , the method  $\mathbf{Expo}(m, n)$  terminates and returns the value  $m^n$ .

**Proof:** We prove by *complete induction* that P(n) holds for all  $n \in \mathbb{N}$ , where P(n) is for all  $n \in \mathbb{N}$ ,  $E \neq 0$  (m, n) (et n ( 15 m)).

Assume that  $n \in \mathbb{N}$ , and that P(i) holds for all  $i \in \mathbb{N}$ ,  $0 \le i < n$ . So, we have that for all i < n,  $E \neq p \circ (m, i)$  terminates and returns  $m^{i}$  (\*)

To prove P(n), there are three cases to consider:

**Case 1:** n = 0. For any m, **EXPO**(m, 0) terminates and returns  $1 = m^0$ .

Case 2: n > 0 n is odd.

From the code, for any m, when n > 0 and n is *odd*,

- **EXPO**(m, n) works by first calling **EXPO**(m, n DIV 2),
- then calling **SQUARE** on the result,
- and finally *multiplying* that result by *m*.

**Q.**Why is this correct?

induction hypothesis **Case 2 cont.** Since *n* is odd,  $\bigwedge_{n \to \infty}^{n-1} \prec \bigwedge_{n \to \infty}^{n}$ 

$$\therefore$$
 I. H holds,  $\binom{n-1}{2}$   
So we can apply (\*),  $\binom{n-1}{2}$ 

The method **SQUARE** always terminates, and *returns* 

 $\binom{\binom{n-1}{2}}{2} = m^{-1}$ 

Therefore, Expo(m, n) terminates and *returns* 

 $m \cdot m^{-1} = m^{-1}$ 

**Case 3:** n > 0 *n* is even.

### similar to previous case

We conclude that the lemma holds for all n and m.