

Lecture 10: Multiplication

COSC 311 *Algorithms*, Fall 2022

Announcements

1. Homework 2 Finalized Today (1 additional question)
2. No reading/lecture ticket for Monday
3. Thoughts on Reading

Overview

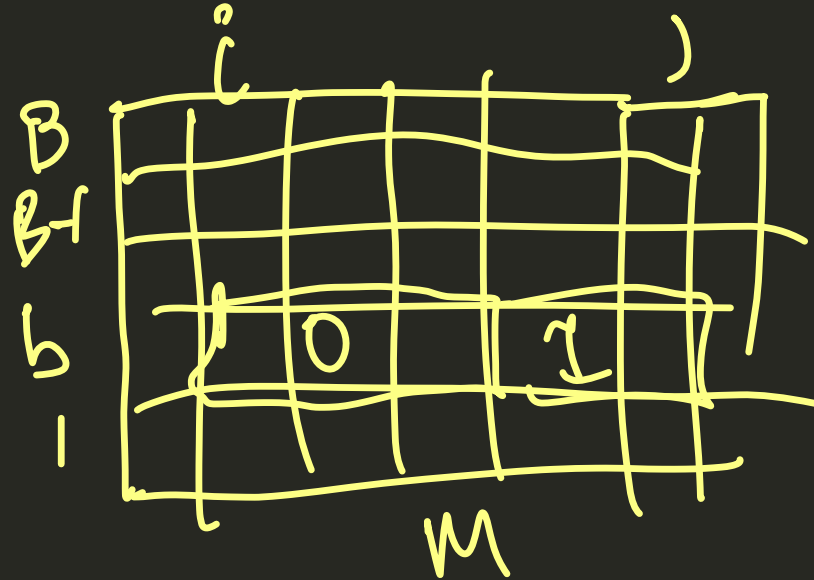
1. Recap of Binary Radix Sort
2. Binary Arithmetic
3. Multiplication via Divide and Conquer

Last Time

Binary Radix Sort

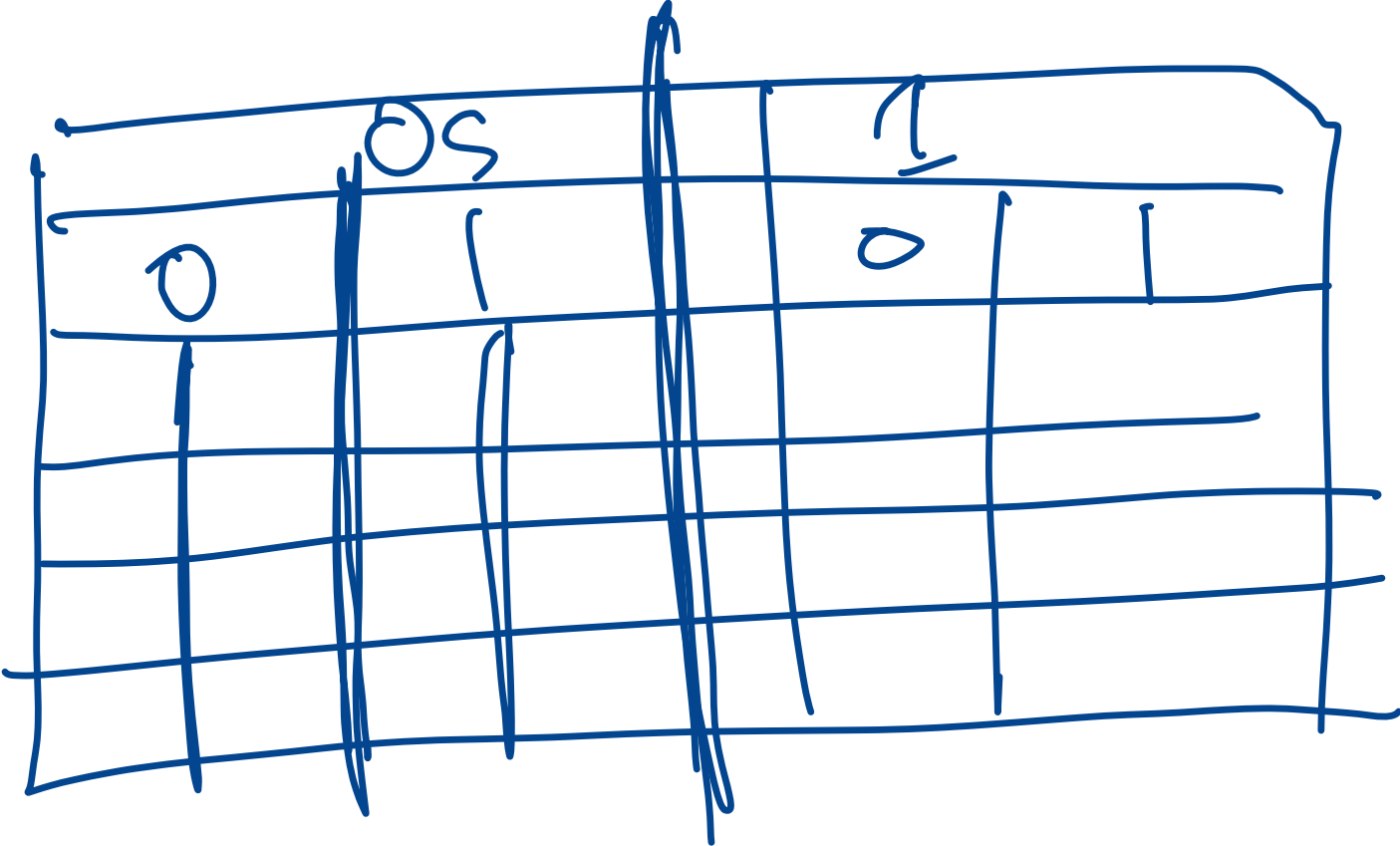
```
RadixSort(a, B):                                     # B is number of bits  
  RadixSort(a, 1, size(a)+1, B)
```

```
RadixSort(a, i, j, b):  
  if j - i <= 1 then  
    return  
  endif  
  m <- BitSplit(a, i, j, b)  
  RadixSort(a, i, m, b-1)  
  RadixSort(a, m, j, b-1)
```

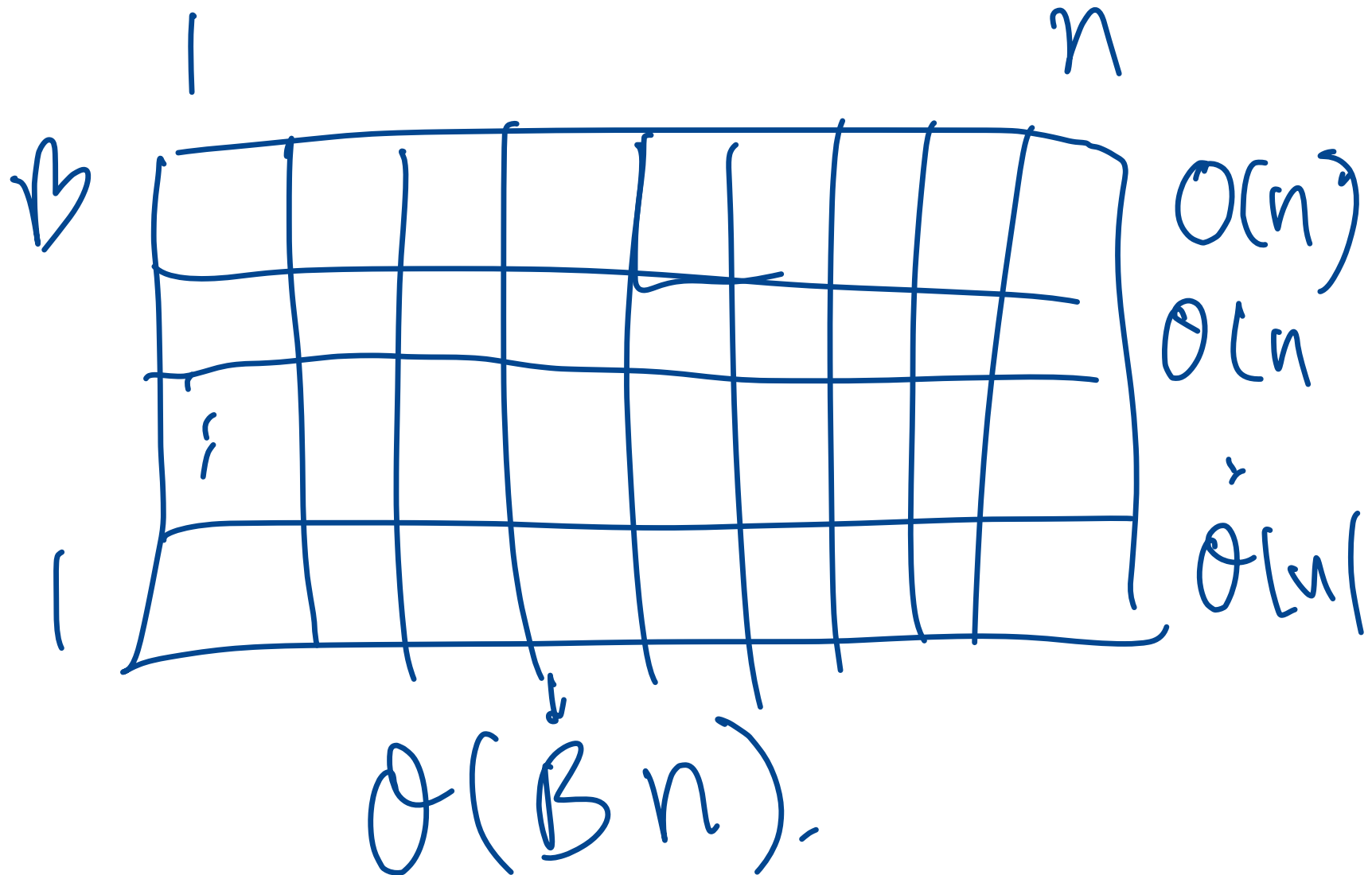


Illustration

Why Does RadixSort Work?



What is RadixSort Running Time?



Conclusion

Lower Bound. Any algorithm that sorts all permutations of size n using only compare and swap operations requires $\Omega(n \log n)$ comparisons.

Caveat. If values are all represented with B bits, then RadixSort sorts n using $O(Bn)$ bit-wise comparisons.

$\ll \log n$

Binary Arithmetic

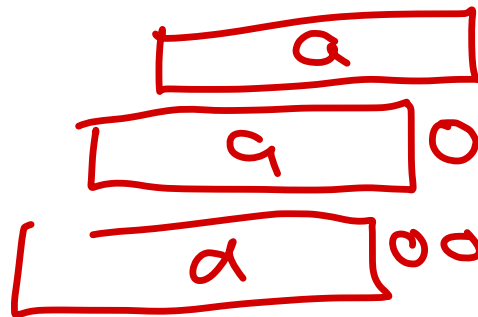
Multiplication in Binary

Example. Compute $10110 * 1011$.

$$\begin{array}{r} 10110 \\ \times 1011 \\ \hline 10110 \\ + 101100 \\ \hline 1000010 \\ + 10110000 \\ \hline 1110010 \end{array}$$

Multiplication Procedure

```
Multiply(a, b):  
product ← 0  
shifted ← a # copy of a we will shift  
for i = 1 to size b do  
  if b[i] = 1 do  
    product ← Add(product, shifted)  
  endif  
  shifted ←<< 1  
endfor
```



running total

Question

a, b $w \begin{matrix} \nearrow \log N \\ \text{bits} \end{matrix}$

```
Multiply(a, b):  
product ← 0  
shifted ← a # copy of a we will shift  
for i = 1 to size b do  
  if b[i] = 1 do  
    product ← Add(product, shifted)  
  endif  
  shifted ← shifted << 1  
endfor
```

n
iter.

$O(n)$

If a and b are represented with n bits, what is the running time of $\text{Multiply}(a, b)$?

$O(n^2)$? :

$n \cdot O(n) = O(n^2)$

$$\begin{aligned} a, b &\leq N \leq 2^n - 1 \\ a &= a_n a_{n-1} \dots a_1 \\ &\leq 2^{n-1} + 2^{n-2} + \dots + 2^0 \\ &= 2^n - 1 \end{aligned}$$

Another Question

Why did we previously assume arithmetic takes $O(1)$ time?

int in Java uses 32 bits

Multiplication via Divide and Conquer

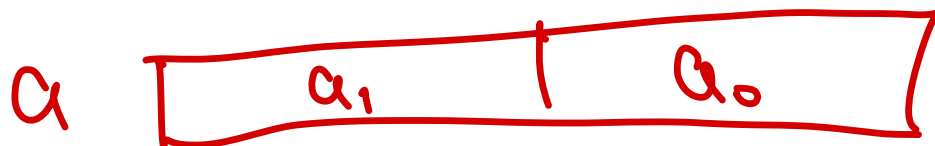
Idea. Break numbers up into parts

- Assume a and b are both represented with $n = 2B$ bits, n power of 2

• Write:

- $a = a_1 a_0 = a_1 2^B + a_0$

- $b = b_1 b_0 = b_1 2^B + b_0$



$$a = \overbrace{a_1}^{a_1 \cdot 2^B} \overbrace{0 \dots 0} + \overbrace{0 \dots 0}^{a_0}$$

• Rewrite multiplication

- $\boxed{ab} = \underbrace{(a_1 2^B + a_0)}_{\text{mult size } n} \underbrace{(b_1 2^B + b_0)}_{\text{mult size } n} = \boxed{a_1 b_1} 2^{2B} + \boxed{a_1 b_0 + a_0 b_1} 2^B + \boxed{a_0 b_0}$

↑
mult
size
 n

4 mult of size $n/2$

Does This Help?

$$ab = (a_1 2^B + a_0)(b_1 2^B + b_0) = a_1 b_1 2^{2B} + (a_1 b_0 + a_0 b_1) 2^B +$$

- Replaced 1 product with n bit numbers with 4 products of $n/2$ bits

Not yet...

A Magic Trick

$$ab = \boxed{a_1 b_1} 2^{2B} + \boxed{(a_1 b_0 + a_0 b_1)} 2^B + \boxed{a_0 b_0}$$

• Define: c_2

$$c_2 = a_1 b_1 \quad c_1 = a_1 b_0 + a_0 b_1 \quad c_0 = a_0 b_0$$

• Now: $ab = \boxed{c_2} 2^{2B} + c_1 2^B + \boxed{c_0}$

• Consider:

$$\underline{c^*} = \underline{(a_1 + a_0)(b_1 + b_0)} = \underline{a_1 b_1} + \underline{a_1 b_0} + \underline{a_0 b_1} + \underline{a_0 b_0}$$

Question. How do c_0, c_1, c_2 relate to c^* ?

$$\underline{c^* = c_2 + c_1 + c_0}$$

$$\Rightarrow c_1 = c^* - c_2 - c_0$$

$\frac{3}{4} n^2$

n^2 ops for
 g.s. mult
 \downarrow
 $3 \cdot \left(\frac{n}{2}\right)^2 + O(n)$
 mult of size $n/2$

Counting Products

- Standard multiplication:
 - $c_2 = a_1 b_1$
 - $c_1 = a_1 b_0 + a_0 b_1$
 - $c_0 = a_0 b_0$
- Tricky multiplication
 - $c_2 = a_1 b_1$
 - $c_0 = a_0 b_0$
 - $c^* = (a_1 + a_0)(b_1 + b_0)$
 - $c_1 = c^* - c_2 - c_0$

Progress?

By using c^* to compute ab :

- $c_2 = a_1 b_1$
- $c_0 = a_0 b_0$
- $c^* = (a_1 + a_0)(b_1 + b_0)$

Compute

- $ab = c_2 2^{2B} + (c^* - c_2 - c_0) 2^B + c_0$

Computing ab uses:

- 3 multiplications of size $n/2$
- $O(1)$ additions/subtractions/shifts of size $O(n)$

Karatsuba Multiplication

```
KMult(a, b):  
  n <- size(a) (= size(b))  
  if n = 1 then return a*b  
  a = a1 a0  
  b = b1 b0  
  c2 <- KMult(a1, b1)  
  c0 <- KMult(a0, b0)  
  c <- KMult(a1 + a0, b1 + b0)  
  return (c2 << n) + ((c - c2 - c0) << (n/2)) + c0
```

Karatsuba Recursion Tree

Efficiency of Karatsuba

At depth k :

- 3^k calls to `KMult`
- size of each call is $n/2^k$
- depth of recursion is $\log n$

Total running time:

- $O(n) + \frac{3}{2}O(n) + \left(\frac{3}{2}\right)^2 O(n) + \dots + \left(\frac{3}{2}\right)^{\log n} O(n)$

Can show:

- This expression is $O(3^{\log n})$

Simplify:

Final Running Time

Result. The running time of Karatsuba multiplication is $O(n^{\log 3}) \approx O(n^{1.58})$

- when n is reasonably large, $n^{1.58} \ll n^2$
- E.g., $1,000^2 = 1,000,000$ vs $1,000^{1.58} \approx 55,000$

Next Time

- More Divide and Conquer
- Solving General Recurrences