Divide and Conquer

Algorithms

Divide and Conquer

- Generic recipe for many solutions:
 - Divide the problem into two or more smaller instances of the same problem
 - Conquer the smaller instances using recursion (or a base case)
 - Combine the answers to solve the original problem

- Assume we want to multiply two n-bit integers with n a power of two
 - Divide: break the integers into two n/2-bit integers

$$x = 2^{\frac{n}{2}} x_L + x_R$$

$$y = 2^{\frac{n}{2}} y_L + y_R$$
 y_R

 Conquer: Solve the problem of multiplying of n/2 bit integers by recursion or a base case for n=1, n=2, or n=4

$$x = 2^{\frac{n}{2}} x_L + x_R \qquad \qquad x_R \qquad \qquad x_R$$

$$y = 2^{\frac{n}{2}} y_L + y_R \qquad \qquad y_L \qquad \qquad y_R$$

$$x_L \cdot y_L \quad x_L \cdot y_R \quad x_R \cdot y_L \quad x_R \cdot y_R$$

- Now combine:
 - In the naïve way:

$$x \cdot y = (x_L \cdot 2^{\frac{n}{2}} + x_R) \cdot (y_L \cdot 2^{\frac{n}{2}} + y_R)$$

$$= x_L \cdot y_L \cdot 2^n + (x_L \cdot y_R + x_R \cdot y_L) \cdot 2^{\frac{n}{2}} + x_R \cdot y_R$$

$$x \cdot y = (x_L 2^{\frac{n}{2}} + x_R) \cdot (y_L 2^{\frac{n}{2}} + y_R)$$
$$= x_L \cdot y_L 2^n + (x_L \cdot y_R + x_R \cdot y_L) \cdot 2^{\frac{n}{2}} + x_R \cdot y_R$$

- We count the number of multiplications
 - Multiplying by powers of 2 is just shifting, so they do not count
 - T(n) number of bit multiplications for integers with 2^n bits:

$$T(0) = 1$$

Recursion:

$$T(n+1) = 4T(n)$$

Solving the recursion

$$T(0) = 1$$

$$T(n+1) = 4T(n)$$

• Intuition:

$$T(n) = 4T(n-1) = 4^2T(n-2) = 4^3T(n-3) = \dots = 4^nT(0) = 4^n$$

- Proposition: $T(n) = 4^n$
- Proof by induction:
 - Induction base:

$$T(0) = 1 = 4^0$$

- Induction step: Assume $T(n) = 4^{n-1}$. Show $T(n+1) = 4^n$
 - Proof:

$$T(n) = 4T(n-1)$$
 Recursion Equation
= $4 \times 4^{n-1}$ Induction Assumption
= 4^n

- Since the number of bits is $m=2^n$
 - Number of multiplications is

$$S(m) = T(n) = 4^n = (2^n)^n = m^2$$

This is not better than normal multiplication

Now combine:

• Instead:
$$x \cdot y = (x_L 2^{\frac{n}{2}} + x_R) \cdot (y_L 2^{\frac{n}{2}} + y_R)$$

= $x_L \cdot y_L \cdot 2^n + (x_L \cdot y_R + x_R \cdot y_L) \cdot 2^{\frac{n}{2}} + x_R \cdot y_R$

- Use $(x_L \cdot y_R + x_R \cdot y_L) = (x_L + x_R) \cdot (y_L + y_R) x_L \cdot y_L x_R \cdot y_R$
- This reuses two multiplications that are already used

We need to deal with the potential overflow in calculating

$$(x_L + x_R) \cdot (y_L + y_R)$$

- Now, we only do three multiplications of 2^n bit numbers in order to multiply two 2^{n+1} bit numbers
- The recursion becomes

$$T(0) = 1$$
 $T(n + 1) = 3T(n)$

- Solving the recurrence T(0) = 1 T(n + 1) = 3T(n)
 - Heuristics:

$$T(n) = 3T(n-1) = 3^2T(n-2) = \dots = 3^nT(0) = 3^n$$

As before prove exactly using induction

• The multiplication of two $m=2^n$ -bit numbers takes

$$S(m) = T(n)$$

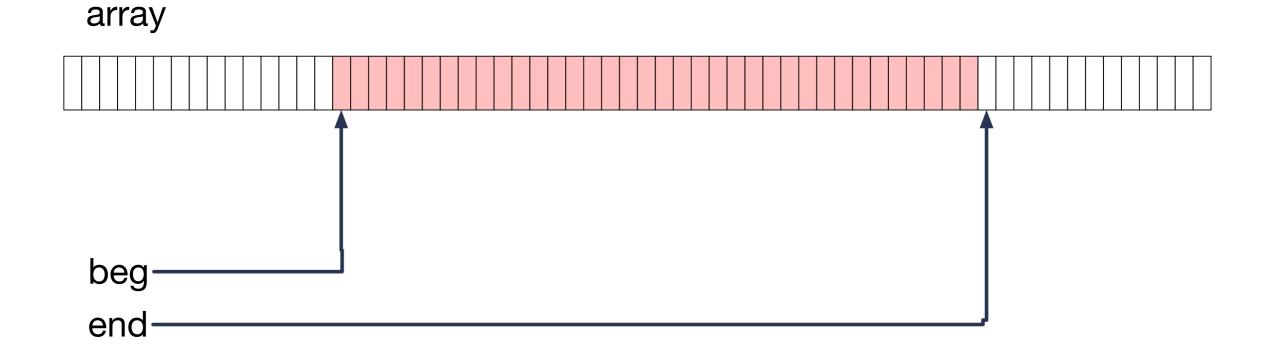
 $= 3^n$
 $= 3^{\log_2(m)}$
 $= \exp(\log(3^{\log_2(m)}))$
 $= \exp(\log_2 m \log 3)$
 $= \exp(\log m \log 3 \frac{1}{\log 2})$
 $= \exp(\log(m^{\log_2 3})$
 $= m^{\log_2 3}$

• This way, multiplication of m-bit numbers takes $m^{1.58496}$ bit multiplications

- Can be used for arbitrary length integer multiplication
- Base case is 32 or 64 bits

But can still do better using Fast Fourier Transformation

- Given an array of ordered integers, a pointer to the beginning and to the end of a portion of the array, decide whether an element is in the slice
- Search (array, beg, end, element)



- Divide: Determine the middle element. This divides the array into two subsets
- Conquer: Compare the element with the middle element.
 If it is smaller, find out whether the element is in the left half, otherwise, whether the element is in the right half
- Combine: Just return the answer to the one question

```
def binary search (array, beg, end, key):
    if beg >= end:
        return False
    mid = (beq+end)//2
    if array[mid] == key:
        return True
    elif array[mid] > key:
        return binary search (array, beg, mid, key)
    else:
        return binary search(array, mid+1, end, key)
test = [2, 3, 5, 6, 12, 15, 17, 19, 21, 23, 27, 29,
        31, 33, 35, 39, 41]
print(binary search(test, 0, len(test), 21))
print(binary search(test, 0, len(test), 22))
```

- Let T(n) be the runtime of binary_search on a subarray with n elements
- Recursion: There is a constant c such that

$$T(1) \le c$$

$$T(n) \le T(n//2) + c$$

Solving the recursion

$$T(n) \leq T(n//2) + c$$

 $\leq T(n//4) + 2c$
 \cdots
 $\leq T(n//2^m) + mc$

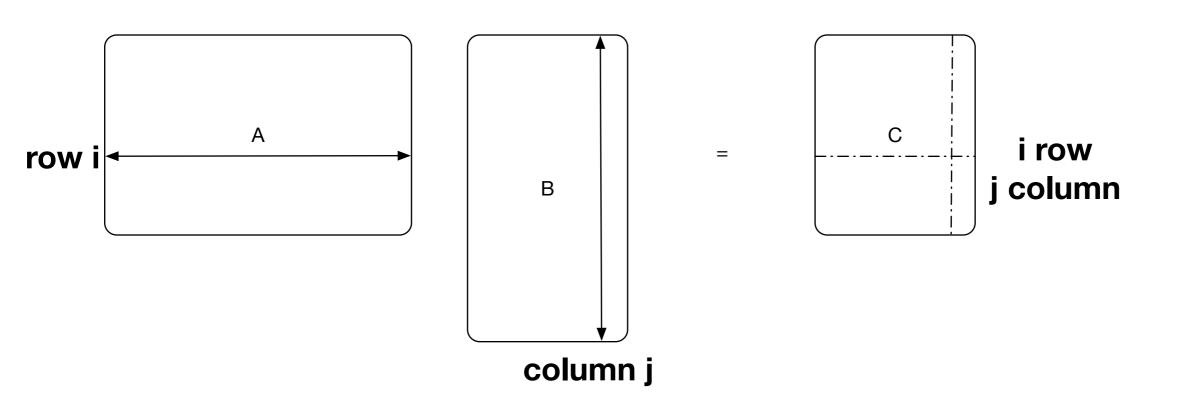
• If $m \ge \log_2 n$ then $T(n) \le T(1) + mc = (m+1)c$

• With other words, binary search on *n* elements takes time

$$\propto \log_2(n)$$

Definition of Matrix Multiplication

$$(a_{i,j})_{\substack{1 \le i \le m \\ 1 \le j \le n}} \cdot (b_{j,k})_{\substack{1 \le j \le n \\ 1 \le k \le p}} = (\sum_{j=1}^n a_{i,j}b_{j,k})_{\substack{1 \le i \le m \\ 1 \le k \le p}}$$



- Cost of definition:
- n^2 multiplications for all mk elements in the product
 - Square $n \times n$ matrices: n^4 elements

- Divide and conquer: Assume $n = 2^r$ is a power of two.
- We can use the following theorem:
 - Break each matrix into four submatrices of size $2^{r-1} \times 2^{r-1}$ and calculate

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \cdot \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} = \begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{21} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{21} + A_{22}B_{22} \end{pmatrix}$$

- As is, a divide and conquer algorithm gives us 8 multiplication of matrices half the size.
- Let m(n) be the number of multiplications needed to multiply two $2^n \times 2^n$ matrices using divide and conquer
- Obviously: m(1) = 1
- Recursion: m(n + 1) = 8m(n)

- Claim: $m(n) = 2^{3n}$
- Proof: Induction base: $m(0) = 1 = 2^{3 \cdot 0}$
- Induction step:
 - Hypothesis: $m(n) = 2^{3n}$
 - To show: $m(n + 1) = 2^{3(n+1)}$
 - Proof:

$$m(n + 1) = 8m(n) = 8 \cdot 2^{3n} = 2^3 \cdot 2^{3n} = 2^{3n+3} = 2^{3(n+1)}$$

- Strassen: Can use 7 matrix multiplications to calculate all eight products
 - $\mathbf{M}_1 := (\mathbf{A}_{1,1} + \mathbf{A}_{2,2})(\mathbf{B}_{1,1} + \mathbf{B}_{2,2})$
 - $\mathbf{M}_2 := (\mathbf{A}_{2,1} + \mathbf{A}_{2,2})\mathbf{B}_{1,1}$
 - $\mathbf{M}_3 := \mathbf{A}_{1,1}(\mathbf{B}_{1,2} \mathbf{B}_{2,2})$
 - $\mathbf{M}_4 := \mathbf{A}_{2,2}(\mathbf{B}_{2,1} \mathbf{B}_{1,1})$
 - $\mathbf{M}_5 := (\mathbf{A}_{1,1} + \mathbf{A}_{1,2})\mathbf{B}_{2,2}$
 - $\mathbf{M}_6 := (\mathbf{A}_{2,1} \mathbf{A}_{1,1})(\mathbf{B}_{1,1} + \mathbf{B}_{1,2})$
 - $\mathbf{M}_7 := (\mathbf{A}_{1,2} \mathbf{A}_{2,2})(\mathbf{B}_{2,1} + \mathbf{B}_{2,2})$

- Then can get all the submatrices on the right:
- $C_{1,1} = M_1 + M_4 M_5 + M_7$
- $C_{1,2} = M_3 + M_5$
- $C_{2,1} = M_2 + M_4$
- $C_{2,2} = M_1 M_2 + M_3 + M_6$

- Now the recurrence becomes
 - m(n + 1) = 7m(n), m(0) = 1
- which is obviously solved by
 - $m(n) = 7^n$.

- Remember that the size of the matrix was $2^n \times 2^n$.
- Thus, if M(n) is the number of multiplications for an $n \times n$ matrix with power of 2 rows, then
 - $M(n) = m(\log_2(n)) = 7^{\log_2(n)}$
- Since

$$\log_2(7^{\log_2(n)}) = \log_2(n)\log_2(7) = \log_2(7)\log_2(n) = \log_2(n^{\log_2(7)})$$

$$M(n) = n^{\log_2(7)} \approx n^{2.80735}$$

- The algorithm can be extended for matrices that
 - have number of rows = number of columns not a power of 2
 - are not square