# CSC373 Worksheet 1 Solution

# July 23, 2020

```
1_1
       Strassen_Algorithm(A,B):
           n = A.rows
           let C be a new n x n matrix
 3
           if n == 1
               C_{11} = A_{11} * B_{11}
6
           else partition as in step 3 of strassen's algorithm
8
               p1 = Strassen_Algorithm(A_11, B_12) -
                     Strassen_Algorithm(A_11, B_22)
11
12
13
               p2 = Strassen_Algorithm(A_11, B_22) +
                     Strassen_Algorithm(A_12, B_22)
14
15
               p3 = Strassen_Algorithm(A_21, B_11) +
16
                     Strassen_Algorithm(A_22, B_11)
17
               p4 = Strassen_Algorithm(A_22, B_21) -
19
                     Strassen_Algorithm(A_22, B_11)
20
21
               p5 = Strassen_Algorithm(A_11, B_11) +
                     Strassen_Algorithm(A_11, B_22) +
23
                     Strassen\_Algorithm(A\_22, B\_11) +
24
                     Strassen_Algorithm(A_22, B_22)
25
               p6 = Strassen_Algorithm(A_12, B_21) +
27
                     Strassen_Algorithm(A_12, B_22) -
28
                     Strassen_Algorithm(A_22, B_21) -
29
                     Strassen_Algorithm(A_22, B_22)
30
31
               p7 = Strassen_Algorithm(A_11, B_11) +
32
                     {\tt Strassen\_Algorithm(A\_11,\ B\_12)}
33
                     Strassen_Algorithm(A_21, B_11)
34
                     Strassen_Algorithm(A_21, B_12)
35
36
               C_{11} = p5 + p4 - p2 + p6
               C_{12} = p1 + p2
38
               C_{21} = p3 + p4
```

### Notes:

- Strassen's method for matrix multiplication
  - Reduces the time complexity of matrix multiplication from  $O(n^3)$  to  $O(n^{\log_2 7}) = O(n^{2.81})$
  - Has four steps
    - 1) Divide the input matrices A and B and output matrix C into  $n/2 \times n/2$  submatrices

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}, \quad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}, \quad C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix},$$

2) Create 10 matrices,  $S_1, S_2, ..., S_{10}$  each of which is  $n/2 \times n/2$  and is the sum or difference of two matrices created in step 1

$$S_1 = B_{12} - B_{22}$$

$$S_2 = A_{11} + A_{12}$$

$$S_3 = A_{21} + A_{22}$$

$$S_4 = B_{21} - B_{11}$$

$$S_5 = A_{11} + A_{22}$$

$$S_6 = B_{11} + B_{22}$$

$$S_7 = A_{12} - A_{22}$$

$$S_8 = B_{21} + B_{22}$$

$$S_9 = A_{11} - A_{21}$$

$$S_{10} = B_{11} + B_{12}$$

3) Recursively multiply  $n/2 \times n/2$  matrices seven times to compute the following  $n/2 \times n/2$  matrices

$$\begin{split} P_1 &= A_{11} \cdot S_1 = A_{11} \cdot B_{12} - A_{11} \cdot B_{22} \\ P_2 &= S_2 \cdot B_{22} = A_{11} \cdot B_{22} + A_{12} \cdot B_{22} \\ P_3 &= S_3 \cdot B_{11} = A_{21} \cdot B_{11} + A_{22} \cdot B_{11} \\ P_4 &= A_{22} \cdot S_4 = A_{21} \cdot B_{11} + A_{22} \cdot B_{11} \\ P_5 &= S_5 \cdot S_6 = A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22} \\ P_6 &= S_7 \cdot S_8 = A_{12} \cdot B_{21} + A_{12} \cdot B_{22} - A_{22} \cdot B_{21} - A_{22} \cdot B_{22} \\ P_7 &= S_9 \cdot S_{10} = A_{11} \cdot B_{11} + A_{11} \cdot B_{12} - A_{21} \cdot B_{11} - A_{21} \cdot B_{12} \end{split}$$

4) Construct the four  $n/2 \times n/2$  submatrices of the product C

$$C_{11} = P_5 + P_4 - P_2 + P_6 = A_{11} \cdot B_{11} + A_{12} + B_{12}$$

$$C_{12} = P_1 + P_2 = A_{11} \cdot B_{12} + A_{12} \cdot B_{22}$$

$$C_{21} = P_3 + P_4 = A_2 \cdot B_{11} + A_{22} \cdot B_{21}$$

$$C_{22} = P_5 + P_1 - P_3 - P_7 = A_{22} \cdot B_{22} + A_{21} \cdot B_{12}$$

Example: Use Strassen's algorithm to compute the matrix product

$$\begin{pmatrix} 1 & 3 \\ 7 & 5 \end{pmatrix} \begin{pmatrix} 6 & 8 \\ 4 & 2 \end{pmatrix}$$

#### \* STEP 1

$$A_{11} = 1, A_{12} = 3, A_{21} = 7, A_{22} = 5$$
  
 $B_{11} = 6, B_{12} = 8, B_{21} = 4, B_{22} = 2$ 

#### \* STEP 2

$$S_{1} = B_{12} - B_{22} = 4 - 2 = 2$$

$$S_{2} = A_{11} + A_{12} = 1 + 3 = 4$$

$$S_{3} = A_{21} + A_{22} = 7 + 5 = 12$$

$$S_{4} = B_{21} - B_{11} = 4 - 6 = -2$$

$$S_{5} = A_{11} + A_{22} = 1 + 5 = 6$$

$$S_{6} = B_{11} + B_{22} = 6 + 2 = 8$$

$$S_{7} = A_{12} - A_{22} = 3 - 5 = -2$$

$$S_{8} = B_{21} + B_{22} = 8 + 2 = 10$$

$$S_{9} = A_{11} - A_{21} = 3 - 5 = -2$$

$$S_{10} = B_{11} + B_{12} = 6 + 4 = 10$$

#### \* STEP 3

$$P_{1} = A_{11} \cdot S_{1} = A_{11} \cdot B_{12} - A_{11} \cdot B_{22} = 1 \cdot 4 - 1 \cdot 2 = 2$$

$$P_{2} = S_{2} \cdot B_{22} = A_{11} \cdot B_{22} + A_{12} \cdot B_{22} = 1 \cdot 2 + 3 \cdot 2 = 8$$

$$P_{3} = S_{3} \cdot B_{11} = A_{21} \cdot B_{11} + A_{22} \cdot B_{11} = 6 \cdot 7 + 6 \cdot 5 = 72$$

$$P_{4} = A_{22} \cdot S_{4} = A_{22} \cdot B_{21} - A_{22} \cdot B_{11} = 5 \cdot 4 - 5 \cdot 6 = -10$$

$$P_{5} = S_{5} \cdot S_{6} = A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22} = 48$$

$$P_{6} = S_{7} \cdot S_{8} = A_{12} \cdot B_{21} + A_{12} \cdot B_{22} - A_{22} \cdot B_{21} - A_{22} \cdot B_{22} = -20$$

$$P_{7} = S_{9} \cdot S_{10} = A_{11} \cdot B_{11} + A_{11} \cdot B_{12} - A_{21} \cdot B_{11} - A_{21} \cdot B_{12} = -20$$

\* STEP 4

$$C_{11} = P_5 + P_4 - P_2 + P_6 = 48 - 10 - 8 - 20 = 10$$
  
 $C_{12} = P_1 + P_2 = 10$   
 $C_{21} = P_3 + P_4 = 62$   
 $C_{22} = P_5 + P_1 - P_3 - P_7 = 48 + 2 - 72 + 20 = -2$ 

- Is not preferred in practical purposes
  - 1) The constants used in Strassen's method are high and for a typical application Naive method works better.
  - 2) For Sparse matrices, there are better methods especially designed for them.
  - 3) The submatrices in recursion take extra space.
  - 4) Because of the limited precision of computer arithmetic on noninteger values, larger errors accumulate in Strassen's algorithm than in Naive Method

#### References:

- 1) GeeksForGeeks, Divide and Conquer Set 5 (Strassen's Matrix Multiplication), link
- Regular matrix multiplication

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- The master method for solving recurrences
  - provides 'cookbook' method for solving recurrences of the form

$$T(n) = aT(n/b) + f(n)$$

- depends on the following theorem
  - \* Let  $a \leq 1$  and b > 1 be constants, let f(n) be a function and let T(n) be defined on the nonnegative integers by the recurrence

$$T(n) = aT(n/b) + f(n)$$

where we interpret n/b to mean either  $\lfloor n/b \rfloor$  or  $\lceil n/b \rceil$ . Then T(n) has the following asymptotic bounds:

- 1. If  $f(n) = O(n^{\log_b a \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$
- 2. If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$
- 3. If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(n/b) \le cf(n)$  for some constant c < 1, and all sufficiently large n, then  $T(n) = \Theta(f(n))$ .

## Example:

$$T(n) = 9T(n/3) + n$$

Here, 
$$a = 9$$
,  $b = 3$ , and  $f(n) = n = O(n^{\log_3 9 - 1})$  where  $\epsilon = 1$ .

Thus, 
$$T(n) = \Theta(n^{\log_3 9})$$
 or  $T(n) = \Theta(n^2)$ 

### Example 2:

$$T(n) = T(2n/3) + 1$$

Here, 
$$a = 1$$
,  $b = 3/2$ ,  $f(n) = 1 = \Theta(n^{\log_{3/2} 1})$ .

Thus, 
$$T(n) = \theta(\lg n)$$

#### Example 3:

$$T(n) = T(n/4) + n \lg n$$

Here  $a=1,\ b=4,$  and  $f(n)=n\lg n$  has asymptotic lower bound of  $f(n)=\Omega(n^{\log_4 3+\epsilon})=\Omega(n)$  where  $\epsilon\approx 0.2$ 

Furthermore,

$$af(n/b) = (3n/4) \lg n/4$$

$$= (3/4)n \lg n/4$$

$$= (3/4)n \lg n/4$$

$$= 3/4n \lg n - \lg 4$$

$$< 3/4n \lg n$$

$$= cf(n)$$

where c = 3/4.

Thus,  $T(n) = \Theta(n \lg n)$ 

# Example 4:

$$T(n) = 2T(n/2) + n \lg n$$

Here, 
$$a = 2$$
,  $b = 2$ ,  $f(n) = n \lg n$ .

2. Let  $n = 3^m$  where m is an element of  $\mathbb{Z}^+ \cup \{0\}$ 

Then we know the time it takes to multiply  $n \times n$  matrices in  $3 \times 3$  matrices is  $T(n) = kT(\frac{n}{3}) + \Theta(n^2)$ .

Now, I need to look for the upper bound of k in  $T(n) = \Theta(n^{\log_3 k})$  satisfying  $O(n^{\lg 7}) \approx O(n^{2.81})$ .

And using master's master's theorem, we can write that the upper limit of k is 21.

## Improved Solution:

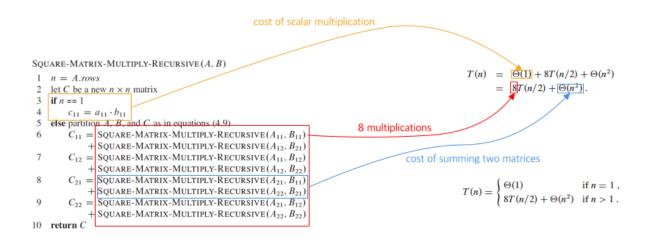
Let  $n = 3^m$  where m is an element of  $\mathbb{Z}^+ \cup \{0\}$ 

Then we know the time it takes to multiply  $n \times n$  matrices in  $3 \times 3$  matrices is  $T(n) = kT(\frac{n}{3}) + \Theta(n^2)$ .

Now, I need to look for the upper bound of k in  $T(n) = \Theta(n^{\log_3 k})$  satisfying  $O(n^{\lg 7}) \approx O(n^{2.81})$ .

And using master's master's theorem, we can write that the upper limit of k is 21 (Following the first condition  $f(n) = \mathcal{O}(n^{\log_3 k - \epsilon})$  where  $\epsilon \approx 0.81$ ).

#### Notes:



- T(n) represents the time it takes to multiply two  $n \times n$  matrices.
- At base case scalar multiplication is performed. So,  $T(1) = \Theta(1)$ .
- 8 represents the number of recursive calls on the function SQUARE-MATRIX-MULTIPLY-RECURSIVE
- $\Theta(n^2)$  represents the addition of two  $\frac{n}{2} \times \frac{n}{2}$  matrices
- 3.  $68 \times 68$  matrices using 132, 464 multiplications:
  - Has recurrence of form  $T(n) = 132,464T(\frac{n}{68}) + \Theta(n^2)$
  - Has  $a = 132, 464, b = 68, f(n) = \Theta(n^2)$
  - Since  $f(n) = \Theta(n^{\log_b a \epsilon})$  where  $\epsilon \approx 0.80$ , case 1 of master's theorem applies and  $T(n) = \Theta(n^{\log_b a}) \approx \Theta(n^{2.80})$
  - $70 \times 70$  matrices using 143,640 multiplications:

- Has recurrence of form  $T(n) = 143,640T(\frac{n}{70}) + \Theta(n^2)$
- Has  $a = 143,640, b = 70, f(n) = \Theta(n^2)$
- Since  $f(n) = \Theta(n^{\log_b a \epsilon})$  where  $\epsilon \approx 0.80$ , case 1 of master's theorem applies and  $T(n) = \Theta(n^{\log_b a}) \approx \Theta(n^{2.80})$
- $72 \times 72$  matrices using 155, 424 multiplications:
  - Has recurrence of form  $T(n) = 155,424(\frac{n}{72}) + \Theta(n^2)$
  - Has  $a = 155, 424, b = 72, f(n) = \Theta(n^2)$
  - Since  $f(n) = \Theta(n^{\log_b a \epsilon})$  where  $\epsilon \approx 0.80$ , case 1 of master's theorem applies and  $T(n) = \Theta(n^{\log_b a}) \approx \Theta(n^{2.80})$

They all have the same asymptotic running time (오잉?!).

In comparison to Strassen method (which has  $\Theta(n^{\lg 7}) \approx \Theta(n^{2.81})$ ), the above three divide and conquer algorithms are a bit faster.

#### **Correct Solution:**

We need to find the divide and conquer method that yields the best asymptotic running time.

Using master's method, we have:

- $T(n) = 132,464T(\frac{n}{68}) + \Theta(n^2) \rightarrow T(n) \approx \Theta(n^{2.7951284873613815})$
- $T(n) = 143,640T(\frac{n}{70}) + \Theta(n^2) \rightarrow T(n) \approx \Theta(n^{2.795122689748337})$
- $T(n) = 155,424T(\frac{n}{72}) + \Theta(n^2) \to T(n) \approx \Theta(n^{2.795147391093449})$

Based on the above, the second method  $T(n)=143,640T(\frac{n}{70})$  has the best asymptotic running time.

In comparison to Strassen method (which has  $\Theta(n^{\lg 7}) \approx \Theta(n^{2.81})$ ), the above three divide and conquer algorithm is a bit faster.

4.

5. a) Here, a = 2, b = 2, f(n) = 4.

Since  $f(n) = n^{\log_2 2+3} = n^{\log_b a+\epsilon}$  where  $\epsilon = 3$ , and  $af(\frac{n}{b}) = 2\left(\frac{n^4}{16}\right) = \frac{n^4}{8} \le cn^4$  where  $c = \frac{1}{8}$  for sufficiently large n, the case 3 of master's theorem applies.

Thus, T(n) has upper bound of  $\mathcal{O}(n^4)$  and lower bounds of  $\Omega(n^4)$ , or  $\Theta(n^4)$ .

b) Here  $a = 1, b = \frac{10}{7}, f(n) = n$ .

Since  $f(n) = 1 = n^{0+1} = n^{\log_{10/7}(1)+1} = n^{\log_b(a)+\epsilon}$ , where  $\epsilon = 1$ , and  $af(\frac{n}{b}) = \frac{7n}{10} \le cn^4$  where  $c = \frac{7}{10}$  for sufficiently large n, the case 3 of master's theorem applies.

Thus, T(n) has upper bound of  $\mathcal{O}(n)$  and lower bounds of  $\Omega(n^4)$ , or  $\Theta(n)$ .

c) Here we have  $a = 16, b = 4, f(n) = n^2$ .

Since  $f(n) = n^2 = n^{\log_4 16} = n^{\log_b a}$ , case 2 of master's theorem applies.

Thus, T(n) has upper bound of  $\mathcal{O}(n^2 \lg n)$  and lower bounds of  $\Omega(n^2 \lg n)$ .

d) Here we have  $a = 7, b = 3, f(n) = n^2$ .

Since  $n^2 = n^{\log_3 7 + \epsilon} = n^{\log_b a + \epsilon}$  where  $\epsilon \approx 0.23$ , and  $af(\frac{n}{b}) \leq cn^2$  where  $c = \frac{7}{9}$ , the case 3 of master's theorem applies.

Thus, T(n) has upper bound of  $\mathcal{O}(n^2)$  and lower bounds of  $\Omega(n^2)$ , or  $\Theta(n^2)$ .

e) Here we have  $a = 7, b = 2, f(n) = n^2$ .

Since  $f(n) = n^2 = n^{\log_2(7) - \epsilon}$ , where  $\epsilon \approx 0.81$ , case 1 of master theorem applies.

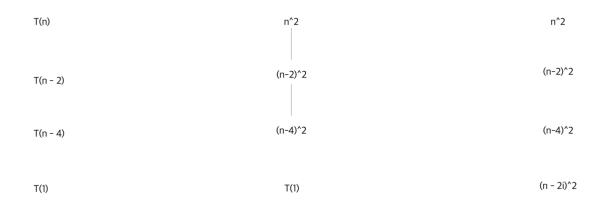
Thus, T(n) has upper bound of  $\mathcal{O}(n^{\log_2 7})$  and lower bound of  $\Omega(n^{\log_2 7})$ , or  $\Theta(n^{\log_2 7})$ .

f) Here we have  $a = 2, b = 4, f(n) = \sqrt{(n)}$ .

Since  $f(n) = \sqrt{n} = n^{\log_4 2} = n^{\log_b a}$ , case 2 of master's theorem applies.

Thus T(n) has upper bound of  $\mathcal{O}(\sqrt{n} \lg n)$ , and lower bound of  $\Omega(\sqrt{n} \lg n)$ , or  $\Theta(\sqrt{n} \lg n)$ .

#### g) Solution:



Using recurrence tree method, we can see that the tree has depth of  $\frac{n}{2}$ , level cost of  $(n-2i)^2$  where i=0,1,...,n-1, and bottom level cost of T(1) or  $\Theta(1)$ .

So, the total cost of T(n) is:

$$T(n) = \sum_{i=0}^{\frac{n}{2}-1} (n-2i) + \Theta(1)$$
 (1)

$$= \frac{n^2}{2} - 2\sum_{i=0}^{\frac{n}{2}-1} i + \Theta(1) \tag{2}$$

$$=\frac{n^2}{2} - \left(\frac{n}{2}\right)\left(\frac{n}{2} - 1\right) + \Theta(1) \tag{3}$$

$$= \left(\frac{n^2}{2}\right) - \left(\frac{n^2}{4} - \frac{n}{2}\right) + \Theta(1) \tag{4}$$

$$= \frac{n^2}{4} + \frac{n}{2} + \Theta(1) \tag{5}$$

$$=\Theta(n^2) \tag{6}$$

And to verify T(n), I will use subtitution method.

Let the guess be  $T(n) \leq cn^3$ .

Then,

$$T(n) = T(n-2) + n^2 (7)$$

$$\leq c(n-2)^3 + n^2 \tag{8}$$

$$= c(n^3 - 6n^2 + 12n - 8) + n^2 (9)$$

$$= c(n^3 - 5n^2 + 12n - 8) - n^2(c - 1)$$
(10)

$$\leq c(n^3 + 12n - 8) - n^2(c - 1) \tag{11}$$

$$= cn^3 - n^2(c-1)$$
 [Since  $n^3$  dominates  $n$ ] (12)

$$\leq cn^3$$
(13)

and the boundary holds as long as  $c \geq 1$ .