

HW2 – Intro to Algorithms
John Carroll

Exercises 3.2-2 and 3.2-3, Problem 3-1 parts a-c, Exercises 4.2-2 and 4.2-4, Problem 4-1 a-f

1) Exercise 3.2-2

Prove equation (3.16).

Ans: Let's have $\log_a b^c = c \log_a b$. (Equation *)

Claim 1: Let $\log_a b = x$. Therefore, $b = a^x$. Then,

$$\log_a b^c = \log_a (a^x)^c = \log_a a^{xc} = xc = \log_a b \cdot c.$$

□

Proof: Let $\log_b a = x$. Therefore, $a = b^x$. Then, using Equation *,

$$a^{\log_b c} = (b^x)^{\log_b c} = b^{x \log_b c} = b^{\log_b c^x} = c^x = c^{\log_b a}.$$

□

2) Exercise 3.2-3

Prove equation (3.19). Also prove that $n! = \omega(2^n)$ and $n! = o(n^n)$.

Ans:

$$\begin{aligned} n! &= o(n^n), \\ n! &= \omega(2^n), \\ \lg(n!) &= \Theta(n \lg n), \end{aligned}$$

Obviously $n! \leq n^n$, so we know that $\log n!$ is $O(n \log n)$. A lower bound for the factorial function would be found using the following:

$$\begin{aligned} n! &= n \times (n-1) \times \dots \times \frac{n}{2} \times \left(\frac{n}{2}-1\right) \times \dots \times 2 \times 1 \\ &\geq \frac{n}{2} \times \frac{n}{2} \times \dots \times \frac{n}{2} \times 1 \times \dots \times 1 \times 1 \\ &= \left(\frac{n}{2}\right)^{n/2} \end{aligned}$$

Therefore

$$\log n! \geq \log \left(\frac{n}{2}\right)^{\frac{n}{2}} = \left(\frac{n}{2}\right) \log \left(\frac{n}{2}\right).$$

In other words, $\log n!$ is in $\Omega(n \log n)$. Thus, $\log n! = \Theta(n \log n)$.

The equation can be proved by using Stirling's approximation:

The equation holds for all $n \geq 1$:

$$\begin{aligned}
 n! &= \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \cdot e^{\alpha_n} \\
 &= \sqrt{2\pi} \cdot n^{1/2} \cdot \frac{n^n}{e^n} \cdot e^{\alpha_n} \\
 &= \sqrt{2\pi} \cdot n^{1/2} \cdot n^n \cdot \frac{1}{e^n} \cdot e^{\alpha_n} \quad \left[\text{Since, } x^a + x^b = x^{a+b} \right] \\
 &= \sqrt{2\pi} \cdot n^{1/2+n} \cdot e^{\alpha_n - n} \quad \left[\text{Since, } x^a + x^b = x^{a+b}, \frac{x^a}{x^b} = x^{a-b} \right] \\
 n! &= \sqrt{2\pi} \cdot n^{n+1/2} \cdot e^{-n} \quad \left[\text{Since, } e^{\alpha_n} = 0 \text{ constant} \right]
 \end{aligned}$$

Apply log both sides then

$$\begin{aligned}
 \lg n! &= \lg(\sqrt{2\pi}) + \lg n^{n+1/2} + \lg e^{\alpha_n - n} \\
 &= \lg(\sqrt{2\pi}) + (n+1/2) \lg n + (\alpha_n - n) \lg e \\
 &= \lg(\sqrt{2\pi}) + n \lg n + 1/2 \lg n - n \quad \therefore \log_e e = 1 \\
 &\approx n \lg n - n \\
 &= \Theta(n \lg n)
 \end{aligned}$$

$$n! = o(n^n)$$

Apply limit theorem then we get

$$\begin{aligned}
 \lim_{n \rightarrow \infty} \frac{n!}{n^n} &= \lim_{n \rightarrow \infty} \frac{\sqrt{2\pi n} \left(\frac{n}{e}\right)^n}{n^n} \\
 &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \frac{n^n}{n^n e^n} \\
 &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \left(\frac{n}{ne}\right)^n \\
 &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \left(\frac{1}{e}\right)^n \\
 &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \left(\frac{1}{e^n}\right) \\
 &= \sqrt{2\pi n} \lim_{n \rightarrow \infty} \left(\frac{1}{e^n}\right) \\
 &= \sqrt{2\pi n} \lim_{n \rightarrow \infty} \left(\frac{1}{e^\infty}\right) \\
 &= \sqrt{2\pi n} \cdot \left(\frac{1}{\infty}\right) \\
 &= 0
 \end{aligned}$$

Since the limit is equal to zero, $\log_2 n$ has a smaller order of growth than \sqrt{n} . So, $\lim_{n \rightarrow \infty} \frac{n!}{n^n} = 0$, we can use the little-oh notation.

Thus, $n! \in o(n^n)$

The given equation:

$$n! = \omega(2^n)$$

Apply Limit theorem then we get

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{n!}{2^n} &= \lim_{n \rightarrow \infty} \frac{\sqrt{2\pi n} \left(\frac{n}{e}\right)^n}{2^n} \\ &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \frac{n^n}{2^n e^n} \\ &= \lim_{n \rightarrow \infty} \sqrt{2\pi n} \left(\frac{n}{2e}\right)^n \\ &= \infty \end{aligned}$$

Thus 2^n grows very fast, $n!$ grows still faster. We can write symbolically that $n! \in \omega(2^n)$

3) Problem 3-1 a-c

3-1 Asymptotic behavior of polynomials

Let

$$p(n) = \sum_{i=0}^d a_i n^i,$$

where $a_d > 0$, be a degree- d polynomial in n , and let k be a constant. Use the definitions of the asymptotic notations to prove the following properties.

a) If $k \geq d$, then $p(n) = O(n^k)$.

Ans:

For a given function $m g(n)$, we denote $O(g(n))$ is the set of functions given by

$$O(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq c g(n) \text{ for all } n \geq n_0\}$$

From the given statement $f(n) = p(n)$ and $g(n) = n^k$, where k is a constant.

Given $k \geq d$

$$\Rightarrow n^d \leq n^k$$

$$\Rightarrow c_1 n^d \leq c_2 n^k$$

From (1) and (2) we have

$$0 \leq p(n) \leq c_2 n^d \leq c_2 n^k \text{ for all } n \geq n_0$$

Hence from the definition, $\boxed{p(n) = O(n^k)}$.

b) If $k \leq d$, then $p(n) = \Omega(n^k)$.

Ans:

For a given function $g(n)$, we denote $\Omega(g(n))$ is the set of functions given by

$$\Omega(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq c g(n) \leq f(n) \text{ for all } n \geq n_0\}$$

From the given statement $f(n) = p(n)$ and $g(n) = n^k$, where k is a constant.

Given $k \leq d$

$$\Rightarrow n^k \leq n^d$$

$$\Rightarrow c_1 n^k \leq c_1 n^d \quad \dots\dots(3)$$

From (1) and (3) we have

$$0 \leq c_1 n^k \leq c_1 n^d \leq p(n) \text{ for all } n \geq n_0$$

Hence from the definition, $\boxed{p(n) = \Omega(n^k)}$.

c) If $k = d$, then $p(n) = \Theta(n^k)$.

Ans:

For a given function $g(n)$, we denote $\Theta(g(n))$ is the set of functions given by

$$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2 \text{ and } n_0 \text{ such that}$$
$$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}$$

From the given statement $f(n) = p(n)$ and $g(n) = n^k$, where k is a constant.

From (1) we have

$$0 \leq c_1 n^d \leq p(n) \leq c_2 n^d \text{ for all } n \geq n_0$$

Hence from the definition, $\boxed{p(n) = \Theta(n^k)}$.

4) Exercise 4.2-2

Write pseudocode for Strassen's algorithm.

Ans:

def Matrix(a,b):

 result = []

 for i in range(0,len(a)):

 new_array = []

 result.extend(new_array)

 for j in range(0,len(b[0])):

 ssum = 0

 for k in range(0,len(a[0])):

 ssum += a[i][k] * b[k][j]

 result[i][j] = ssum

 return result

5) Exercise 4.2-4

What is the largest k such that if you can multiply 3×3 matrices using k multiplications (not assuming commutativity of multiplication), then you can multiply $n \times n$ matrices in time $O(n^{\lg 7})$? What would the running time of this algorithm be?

Ans:

Strassen's algorithm takes the approach of a recursive multiply with a base condition of 2×2 matrices. We are asked to apply an algorithm using a base condition of a 3×3 matrix and told it will take k multiplications.

Consider the comparative recursions:

$$\begin{array}{ll} \text{Strassen:} & T(n) = 7T(n/2) + \Theta(n^2) \\ 3 \times 3: & T(n) = kT(n/3) + \Theta(n^2) \end{array}$$

As the hint points out, case 1 of the Master Theorem applies and the recursive term dominates. Concentrating on the 3×3 recursion, we want to solve for k such that the number of multiplies will be less than $n^{\lg 7}$. We do so as follows:

$$\Theta(n^{\lg 7}) \geq \Theta(n^{\log_3 k}), \text{ so}$$

$$\begin{aligned} n^{\lg 7} &\geq n^{\log_3 k} \\ \lg 7 &\geq \log_3 k \end{aligned}$$

Utilizing maple as suggested to solve for k we find that $21.8499 \geq k$. Therefore the largest k possible, while still doing better than $O(n^{\lg 7})$ using this 3×3 method is 21.

Plugging 21 back into the recurrence and solving using case 1 of the Master Theorem provides us with a running time of:

$$\Theta(n^{\log_3(21)}) \approx \Theta(n^{2.7712})$$

6) Problem 4-1 a-f

Give asymptotic upper and lower bounds for $T(n)$ in each of the following recurrences. Assume that $T(n)$ is constant for $n \leq 2$. Make your bounds as tight as possible, and justify your answers.

Case 3 of the master theorem is applied for parts a, b and d.

a) $T(n) = 2T(n/2) + n^4.$

Ans:

Consider the recurrence relation, $T(n) = 2T(n/2) + n^4.$

Here $a=2$, $b=2$ and $f(n) = n^4.$

$$\begin{aligned} n^{\log_b a} &= n^{\log_2 2} \\ &= n^{\log 1} \\ &= n \quad (\text{since } \log 1 = 1) \end{aligned}$$

Apply case 3 of master theorem (refer theorem 4.1).

$$\begin{aligned} f(n) &= n^4 \\ &= \Omega(n^{\log_b a + \epsilon}) \\ &= \Omega(n^{\log_2 2 + 2}) \end{aligned}$$

Substitute the values of a , b , and k to find the value of $\frac{a}{b^k}$

$$\begin{aligned} \frac{a}{b^k} &= \frac{2}{2^4} \\ &= \frac{2}{16} \\ &= \frac{1}{8} < 1 \end{aligned}$$

Thus running time is $\boxed{\Theta(n^4)}$.

b) $T(n) = T(7n/10) + n.$

Ans:

Consider the recurrence relation, $T(n) = T(7n/10) + n.$

Here, $a = 1$; $b = 10/7 = 1.42$ and $f(n) = n.$

$$\begin{aligned} n^{\log_b a} &= n^{\log_{10/7} 1} \\ &= n^0 \\ &= 1 \end{aligned}$$

Apply case 3 of master theorem (refer theorem 4.1).

$$f(n) = n = \Omega(n^{\log_{10/7} 1 + 1})$$

Substitute the values of a , b , and k to find the value of $\frac{a}{b^k}$.

$$\begin{aligned} \frac{a}{b^k} &= \frac{1}{\left(\frac{10}{7}\right)^1} \\ &= \frac{7}{10} < 1 \end{aligned}$$

Thus, the running time is $T(n) = \boxed{\Theta(n)}$.

c) $T(n) = 16T(n/4) + n^2.$

Ans:

Consider the recurrence relation, $T(n) = 16T(n/4) + n^2.$

Here, $a = 16$; $b = 4$ and $f(n) = n^2.$

$$\begin{aligned} n^{\log_b a} &= n^{\log_4 16} \\ &= n^{\log_4 4^2} \\ &= n^{2 \log_4 4} \quad (\text{since } \log_b a = 1) \\ &= n^2 \end{aligned}$$

Apply case 2 of master theorem (refer theorem 4.1).

$$\begin{aligned} f(n) &= \Theta(n^{\log_b a}) \\ &= \Theta(n^{\log_4 16}) \\ &= \Theta(n^{\log_4 4^2}) \\ &= \Theta(n^{2 \log_4 4}) \\ &= \Theta(n^2) \end{aligned}$$

Thus, the running time $T(n) = \Theta(n^2 \log n).$

d) $T(n) = 7T(n/3) + n^2$.

Ans:

Consider the recurrence relation, $T(n) = 7T(n/3) + n^2$.

Here, $a = 7$, $b = 3$ and $f(n) = n^2$.

$$n^{\log_b a} = n^{\log_3 7}$$

The value of $n^{\log_3 7}$ is between 1 and 2.

$$1 < n^{\log_3 7} < 2$$

$$\begin{aligned} f(n) &= n^2 \\ &= \Omega(n^{\log_b a + \epsilon}) \\ &= \Omega(n^{\log_3 7 + \epsilon}) \text{ for some constant } \epsilon > 0 \end{aligned}$$

Also, $\frac{a}{b^k} = \frac{7}{3^2}$

$$= \frac{7}{9} < 1$$

According to case 3 of master theorem (refer theorem 4.1), the running time $T(n) = \Theta(n^2)$.

e) $T(n) = 7T(n/2) + n^2$.

Ans:

Consider the recurrence relation, $T(n) = 7T(n/2) + n^2$.

Here, $a = 7$, $b = 2$ and $f(n) = n^2 = n^{\log_2 7}$.

The value of $n^{\log_2 7}$ is between 1 and 2

$$2 < n^{\log_2 7} < 3$$

So, we have $f(n) = n^2$.

$$\begin{aligned} &= O(n^{\log_b a - \epsilon}) \\ &= O(n^{\log_2 7 - \epsilon}) \text{ for some constant } \epsilon > 0 \end{aligned}$$

According to case 1 of master theorem (refer theorem 4.1), the running time $T(n) = \Theta(n^{\log_2 7})$.

f) $T(n) = T(n-2) + n^2.$

Ans:

Consider the recurrence relation, $T(n) = 2T(n/4) + \sqrt{n}.$

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

$$\log_b a = \log_4 2 = \sqrt{n}$$

Apply case 2 of master theorem (refer theorem 4.1).

$$\sqrt{n} = \Theta\left(n^{\log_4 2}\right) = \Theta\left(n^{\log_4 2}\right)$$

$$\begin{aligned} T(n) &= \Theta\left(n^{\log_4 2} \lg n\right) \\ &= \Theta\left(n^{\log_4 2} \lg n\right) \\ &= \Theta\left(\sqrt{n} \lg n\right) \end{aligned}$$

Thus, the running time $T(n) = \Theta\left(\sqrt{n} \lg n\right).$