

# 1 Function Runtime Analysis

For comparing the asymptotic runtime of two functions  $f(n)$  and  $g(n)$ :

- $f(n) = O(g(n))$  if  $\lim_{x \rightarrow \infty} \frac{f(n)}{g(n)} < \infty$ . "Big O"
- $f(n) = \Omega(g(n))$  if  $\lim_{x \rightarrow \infty} \frac{f(n)}{g(n)} > 0$ . "Big Omega"
- $f(n) = o(g(n))$  if  $\lim_{x \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ . "Little O"
- $f(n) = \omega(g(n))$  if  $\lim_{x \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$ . "Little Omega"
- $f(n) = \Theta(g(n))$  if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ . "Theta"

# 2 Divide-and-Conquer Strategy

1. Break it into subproblems that are themselves smaller instances of the same type of problem.
2. Recursively solve these subproblems.
3. Appropriately combine their answers.

# 3 Master Theorem

For constants  $a > 0$ ,  $b > 1$ ,  $d \geq 0$ , the recurrence  $T(n) = aT(\frac{n}{b}) + O(n^d)$  solves to:

Crib Sheet		Textbook
$T(n) =$	$O(n^d)$ if $\frac{a}{b^d} < 1$ $O(n^d \log n)$ if $\frac{a}{b^d} = 1$ $O(n^{\log_b a})$ if $\frac{a}{b^d} > 1$	$T(n) =$ $O(n^d)$ if $d > \log_b a$ $O(n^d \log n)$ if $d = \log_b a$ $O(n^{\log_b a})$ if $d < \log_b a$

$a$  = branching factor, number of subproblems created

$n/b$  = size of subproblems

$O(n^d)$  time it takes to combine subproblem results

Height of recursion:  $\log_b n$  levels

Width of tree:  $n^{\log_b a}$

$k$ th level of the tree is made up of  $a^k$  subproblems, each of size  $n/b^k$

Total work at  $k$ th level:  $a^k \times O(\frac{n}{b^k})^d = O(n^d) \times (\frac{a}{b^d})^k$

Examples:  $T(n) = 7T(n/4) + n$ . Answer:  $\Theta(n^{\log_4(7)})$ . —  $T(n) = 7T(n/4) + n^2$ . Answer:  $\Theta(n^2)$  —  $T(n) = 64T(n/3) + \sqrt{n}$ . Answer:  $\Theta(n^{\log_3(64)})$

# 4 Big-O of Geometric Series

$$\sum_{i=0}^k a^i = \begin{cases} O(1) & \text{if } a < 1 \\ O(k) & \text{if } a = 1 \\ O(a^k) & \text{if } a > 1 \end{cases}$$

For  $a \neq 1$ ,  $1 + a + a^2 + \dots + a^n = \sum_{i=0}^n a^i = \frac{1-a^{n+1}}{1-a}$

# 5 Manipulating logs

1.  $\log_b n^c = c \log_b n$
2.  $\log_b(n \cdot m) = \log_b n + \log_b m$
3.  $\log_x y = \frac{\log_z y}{\log_z x}$  for any  $Z$

$$4. 2^{\log_3 n} = (3^{\log_3 2})^{\log_3 n} = (3^{\log_3 n})^{\log_3 2} = n^{\log_3 2}$$

$$5. \frac{d}{dx} \ln x = \frac{1}{x}$$

$$6. \frac{d}{dx} \log_a x = \frac{1}{x \ln a}$$

## 6 Proofs

### 1. Proof by Induction

Want to prove that some fact is true for all integers  $n$

### 2. Proof by Contradiction

If you want to prove that something is T, 1. Assume it is F, 2. Find a contradiction implied by assuming it is F

### 3. Contraposition

If you want to prove  $A \rightarrow B$ , you can prove  $\neg B \rightarrow \neg A$

### 4. Direct Proof

Logically straight forward proof

### 5. Counter example

In order to show that a blanket statement is false, you need to find only 1 counterexample that breaks the statement

#### Induction

Claim:  $f(n)$  returns  $n!$

Base Case: when  $n = 1$   $f(n)$  returns 1

Inductive step: Assume claim is true for all values  $< n$

When  $n > 1$   $f(n)$  returns  $n * f(n - 1)$  by the inductive hypothesis

$f(n - 1) = (n - 1)!$  therefore  $n * f(n - 1)$

$= n * (n - 1)! = n * (n - 1)(n - 2) \dots * 1 = n!$

#### Contradiction

Claim: there are infinitely many prime numbers

For sake of contradiction, assume there are a finite number of primes

Consider  $(p_1 * p_2 \dots * p_n + 1) = k$

1)  $k$  isn't divisible by  $p_i$ ,  $k \bmod p_i = 1$

Contradiction: 1)  $k$  is prime or 2)  $k$  is divisible by things not on our list  $1 \dots n$

## 7 Greedy Algorithms

1. Coins - counter example: den = 1, 10, 25. Greedy: 25, 1, 1, 1, 1, 1 Optimal: 10, 10, 10

2. Knapsack - Greedy Algo: (1) Calculate value/weight for each item. (2) Sort from highest to lowest ratio. (3) For each item in list, add it to bag if it fits. Counter Example: Bag size = 10. Item(weight, cost). Item0(6, 7000), Item1(5, 4000). Density:  $7000/6 = 1166$ .  $4000/5 = 800$ . Greedy: (6, 7000) Optimal: 2 times (5, 4000).

3. Activity Scheduling: - choose earliest ending time

**function** SCHEDULE(Array *Activities*[ $A_1, \dots, A_n$ ])

Sort *Activities* by increasing end times

*curend* = 0

*totevents* = 0

**for**  $i \in [1, \dots, n]$  **do**

**if**  $A_i.start \geq curend$  **then**

*curend* =  $A_i.end$

*totevents* ++

**return** TotEvents

▷ If TRUE, we are "choosing" this activity

Base case: say there was only one activity. Then our algorithm would choose that one activity on the one and only pass through the for loop. This is clearly optimal. Inductive step: say that we have  $n > 1$  activities total, and that this algorithm is optimal for up to  $[1, \dots, n - 1]$  activities.

First, we argue that  $A_1$  MUST be in an optimum solution. Consider an optimum solution that does not include  $A_1$ , but whose first element is some  $A_k$ . Since  $A_1$  ends before  $A_k$ , we could replace  $A_k$  with  $A_1$  without interfering with any of the activities that come after  $A_k$ . Therefore, we may as well choose  $A_1$  as the first element!