

0.1 Lecture 4 - Master theorem

0.1.1 intro

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

1. $f(n) = n^{c+\epsilon}$ then $T(n) = \Theta(n^c)$
2. $f(n) = n^c$ then $T(n) = O(n^c)$
3. $f(n) = n^{c-\epsilon}$, then $T(n) = \Omega(f(n))$

0.1.2 inequality

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

1. $f(n) = O(n^{c+\epsilon})$ then $T(n) = O(n^c)$
2. $f(n) = O(n^c)$ then $T(n) = O(n^c)$
3. $f(n) = O(n^{c-\epsilon})$, then $T(n) = O(f(n))$

Remark 0.1 — when $f(n) = \Theta(n)$, we will split the cases by checking $c > 1$, $c = 1$ and $c < 1$

Proof when $f(n) = O(n)$

1.

$$\begin{aligned}
 T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
 &= a \cdot \left(T\left(\frac{n}{b^2}\right) + \left(\frac{n}{b}\right)\right) + f(n) \\
 &= \dots \\
 &= a^i T(1) + a^{i-1} \left(\frac{n}{b^i}\right) + \dots + kn \\
 &= n^c + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i && \text{case 1: } c > 1 \\
 &= n^c + kn \cdot \frac{\left(\frac{a}{b}\right)^{\log_b n} - 1}{\frac{a}{b} - 1} \\
 &\leq n^c + kn \cdot \frac{\left(\frac{a}{b}\right)^{\log_b n}}{\frac{a}{b} - 1} \\
 &= n^c + k \cdot \frac{n^{\log_b a}}{\frac{a}{b} - 1} \\
 &= O(n^c)
 \end{aligned}$$

2.

$$\begin{aligned}
T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
&= n + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i && \text{case 2: } c = 1 \\
&= n + kn \cdot (\log_b n - 1) \\
&= O(n \log n)
\end{aligned}$$

3.

$$\begin{aligned}
T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
&= n^c + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i && \text{case 3: } c < 1 \\
&= n^c + kn \cdot O(1) && \text{decreasing geometric series} \\
&= O(n)
\end{aligned}$$

0.2 Lecture 5 - Randomized algorithms

Exercise 0.2

$$\begin{aligned}
S &= \sum_{i=1}^{\infty} i(1-p)^{i-1} = \sum_{j=0}^{\infty} (j+1)(k)^j \\
kS &= \sum_{j=1}^{\infty} (j+1)(k)^j && (1)
\end{aligned}$$

$$S - kS = 1 + k + k^2 + k^3 + \dots = \frac{1}{1-k} \quad (2)$$

$$S = \frac{1}{(1-k)^2}$$

Exercise 0.3 (Coupon collector)

$$\begin{aligned}
 P(i) &= \frac{n-i}{n} \\
 E[x_i] &= \frac{1}{P(i)} \\
 E[X] &= \sum_{i=1}^{n-1} E[x_i] \\
 &= n \sum_{i=1}^n \frac{1}{i} \\
 &= \Theta(n \log n)
 \end{aligned}$$

1. hiring problem
2. birthday paradox
3. generate random permutation (with proof)

0.3 Lecture 6 - Quick Sort**0.3.1 Intro to algo**

Algorithm 1 Quicksort

Input: A, p, q, r **Output:** 1

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1: if  $p \geq r$  then
2:   return
3: end if
4:  $q = \text{PARTITION}(A, p, r)$ 
5:  $\text{QUICKSORT}(A, p, q - 1)$ 
6:  $\text{QUICKSORT}(A, q + 1, r)$ 
7: return state

```

Algorithm 2 Partition

Input: A, p, r **Output:** q

```

1:  $i \leftarrow p - 1$ 
2: for  $j \leftarrow p$  to  $r - 1$  do
3:   if  $A[j] \leq x$  then
4:      $i \leftarrow i + 1$ 
5:     SWAP( $A[i], A[j]$ )
6:   end if
7: end for
8: return  $i$ 

```

0.3.2 Run time analysis

1. best case: $O(\log n)$
2. worse case: $O(n^2)$

Average run-time of quicksort. To attain average case, randomly select the pivot. Let $p(x_{i,j})$ be the probability that z_i and z_j are compared

$$p(x_{i,j}) = \frac{1}{j - i + 1}$$

$$E[X] = \sum_{i=1}^n 2 \cdot p(x_{i,j}) \quad z_i \text{ and } z_j \text{ are symmetric}$$

$$= O(n \log n)$$

Exercise 0.4 (k-smallest element in union of two sorted lists)