算法设计与分析 Algorithms Design & Analysis

第六讲: 随机化算法

随机化算法(Randomized Algorithms)

- Average-case analysis of deterministic algorithms(确定性算法的平均情况分析)
- Expected running time of randomized algorithms(随机算法的期望运行时间)

Introduction: 雇佣问题(The Hiring Problem)

- Problem(问题描述)
 - Use a job agency to hire a new office assistant(通过中介招聘新职员)
 - Every day we interview a new candidate provided by the agency(一天面试一个中介 介绍的候选人)
 - If the candidate is better qualified than our current office assistant, we fire the current assistant and hire the candidate(优胜劣汰)

The Hiring Problem—An Introduction

代价模型(A Cost Model)

- Every candidate we interview costs us c_i dollars(面试代价)
 - c_1 dollars fee to be paid to the job agency(中介 费用)
 - 。 c_2 dollars travel expenses of the candidate(候选人路费等等) …
- If we hire the candidate, we have to pay c_h extra dollars(雇佣代价)
 - c_3 dollars additional charge to be paid to the job agency(因雇佣付给中介的额外费用)
 - c_4 dollars internal administration(雇佣导致的内部管理费用)

代价模型(A Cost Model)

• **Question**: How much should we expect to pay when selecting an office assistant out of *n* candidates? (*n* 选一的 期望支出为多少?)

■ General answer(问题解答):

- $c_i \cdot n + c_h \cdot m$

■ Best case(最好情况):

- □ The first candidate is the best(第一个最优)
- Then our cost is
- $\Box c_i \cdot n + c_h$

■ Worst case(最坏情况):

- The candidates show up in increasing order of qualification(越来越好)
- □ Then we hire every candidate(每次面试的结果都是被招聘)
- The cost is
- $c_i \cdot n + c_h \cdot n$
- Abstract problem: Compute the expected number of candidates we would hire.(问题:期望的m 是什么?)

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概率分析(Probability Analysis)

- To compute expectations, we need a probability distribution over all possible inputs.
- (欲求期望的*m*,则需要了解输入的分布)

■ For the hiring problem(雇佣问题概率分析):

- Assume that no two candidates are equally qualified(互异)
- □ Then we can rank the candidates 1, 2, ···, n(排序)
- We assume that every possible order (permutation) of the candidates occurs with equal probability(每种排列的概 率相等)
- This is called a uniformly random permutation.(称为一致性随机排列)

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期望雇佣人数(The Expected Number of Hired Candidates)

- The number m of candidates we hire is a random variable that depends on the given input permutation. (与输入的排列有 关)
- Our goal is to compute E(m), the expected value of m.(目标: E(m))

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指示变量(Indicator Variables)

• For an event E, we can define an indicator variable $X_F = I(E)$: (定义)

$$I(E) = \begin{cases} 1 & \text{if event } E \text{ occurs} \\ 0 & \text{if event } E \text{ does not occur} \end{cases}$$

- Indicator variables help to convert between probabilities and expectations.(作用: 概率和期望值 之间的转换,见如下定理)
- **Lemma:** For an event E and its indicator variable $X_F = I(E)$, $Pr(E) = E(X_F)$.

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雇佣问题平均情况分析(Average-Case Analysis of Hiring Algorithm)

- Events: E₁, E₂, ···, E_n (事件)
- E_i = we hire the *i*-th candidate(第*i*个受聘用)
- Indicator variables: X_1 , X_2 , …, X_n (指示变量)
- $X_i = I(E_i)$
- Then

$$m = \sum_{i=1}^{n} X$$

$$E(X_i) = \frac{1}{i}$$

$$E(m) = E\left(\sum_{i=1}^{n} X_i\right)$$

■Lemma:

Assuming that the order in which the candidates are presented is a uniformly random permutation, the expected cost for hiring a new assistant out of n candidates is $\Theta(c_i \cdot n + c_h \cdot \lg n)$. (期望雇佣代价)

平均情况分析的缺限(The Flaw of Average-Case Analysis)

- Adversarial Behaviour
- The hiring algorithm is deterministic(算法确定)
- Every deterministic algorithm has an input that elicits its worst-case behaviour(有最坏的输入)
- An adversary (the job agency) may always provide us with this worst-case input(可能为不良中介利用)
- The big flaw of average-case analysis: (严重情况)
- The input distribution we assume may or may not be correct. (假设分布错误)

解决办法-随机化(Randomization-Beating the Adversary)

- Modification of our hiring algorithm(改进雇佣算
 - Get the complete list of candidates from the agency(从中介获得所有候选人名单)
 - Permute them in a uniformly random order(产生输入的一致性随机排列)
 - Interview them in this order using our old strategy(再面试)
- Expected cost(期望代价): Θ(c_i·n + c_n·lg n)

解决办法-随机化(Randomization-Beating the Adversary)

- Big difference(不同之处):
 - □ No more questionable assumptions(没有争议性假
 - No input is guaranteed to elicit a worst-case behaviour(不存在明确的导致最坏情况的输入)
 - □ The adversary is powerless(削弱了对手的操控能力)
 - □ 如何实现?

·致随机排列生成(Generating Uniformly Random Permutations)

- Tool: Random number generator(随机 数生成器)
 - □ random(a, b) generates random integer between a and b in constant time. ($\stackrel{\circ}{\vdash}$ $\pm (a,$ b)之间的随机整数)
- Question: How fast can we generate a random permutation of *n* elements so that every permutation is equally likely?(如何得 到n个元素的一致性随机排列,方法的效率如何?两 种方法)

排序法排列(Permuting by Sorting)

- Permute-By-Sorting(A)
- Allocate an array B of size n (数组)
- for i = 1..n**2**
- **do** $B[i] \leftarrow \text{random}(1, n^3)$
- Sort arrays A, using B as sort keys(排序,以B为

a b c d e f g h i j i b d j e g c a h f

340 23 223 55 99 722 182 500 15 88 15 23 55 88 99 182 223 340 500 722

Lemma: Procedure Permute-By-Sorting takes O(n lg n) time.(开销)

Lemma: Procedure Permute-By-Sorting produces a uniformly random permutation of array A.(一致性随机排列)

换位法排列(Permuting in Place)

- Permute-In-Place(A)
- \blacksquare 1 **for** i = 1..n 1
- 2 do $j \leftarrow \text{random}(i, n)$
- 3 swap A[/] ↔ A[/] (交换)

Lemma: Procedure Permute-In-Place takes linear time O(n). 开销O(n).

Lemma: Procedure Permute-In-Place produces a uniformly random permutation. .(一致性随机排列)

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快速排序(Quick Sort)

- Based on the three-step process of divideand-conquer(分治).
- To sort A[p . . r]:
 - Divide: Partition A[p..r], into two (possibly empty) subarrays A[p..q-1] and A[q+1..n], such that each element in the first subarray A[p..q-1] is ≤ A[q] and A[q] is ≤ each element in the second subarray A[q+1..n].(根据选定某个中心,将A分成两个部分)
 - □ **Conquer:** Sort the two subarrays by recursive calls to QUICKSORT.(每一部分用QUICKSORT嵌套处理)
 - Combine: No work is needed to combine the subarrays, because they are sorted in place.(组合)

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QUICKSORT(A, p, r)

1 if p < r

2 then q \leftarrow \text{PARTITION}(A, p, r)

3 QUICKSORT(A, p, q - 1)

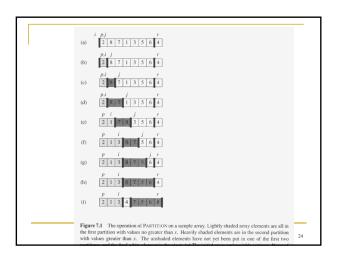
4 QUICKSORT(A, q + 1, r)
```

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Paratition(A,p,r)

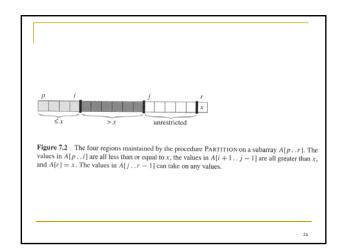
 Perform the divide step by a procedure PARTITION, which returns the index q that marks the position separating the subarrays. (分割,返回分离点的位置q)

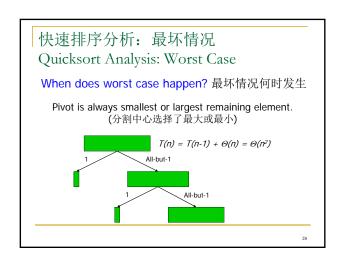
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PARTITION(A, p, r)
1
    x \leftarrow A[r]
2
    i \leftarrow p-1
    for j \leftarrow p to r-1
3
4
           do if A[j] \leq x
5
                  then i \leftarrow i + 1
6
                        exchange A[i] \leftrightarrow A[j]
7
    exchange A[i+1] \leftrightarrow A[r]
    return i + 1
```



Partition(A,p,r)解释

- PARTITION always selects the last element *A*[*r*] in the subarray *A*[*p* .. *r*] as the *pivot*—the element around which to partition.(选择最后一个元素作为分割中心)
- As the procedure executes, the array is partitioned into four regions, some of which may be empty(分割过程出现 四个区域)
 - 1. All entries in $A[p .. i] \le pivot$.
 - 2. All entries in A[i+1 ... j-1] > pivot.
 - 3. *A*[*j* . . *r*-1]
 - 4. A[r] = pivot.





最好情况(Best Case) When does best case happen? (最好情况何时发生) Pivot is always median element. (分割中心每次都选择在中值) T(n) = 2×T(n/2) + Θ(n) = Θ(n lg n) Half Half Half Half

平均情况(Average Case) Average case is more like the best case than the worst case. (平均情况接近最好情况) 如何实现?


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快速排序运行时间(The Running Time of Quicksort)

Quicksort(A, p, r)

if p < r

then swap A[r] \leftrightarrow A[random(p, r)]

q \leftarrow Partition(A, p, r)

then swap A[r] \leftrightarrow A[random(p, r)]

q \leftarrow Partition(A, p, r)

then swap A[r] \leftrightarrow A[random(p, r)]

Quicksort(A, p, q - 1)

Quicksort(A, q + 1, r)

Quicksort(A, q + 1, r)

Lemma: Let X be the total number of comparisons performed by Quicksort. Then the running time is O(n + X). (X: Quicksort f the f then f
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比较次数计算(Counting Comparisons)

- Let $Z_{ij} = \{ z_i, z_{i+1}, \dots, z_j \}$
- Observation: Any two elements z_i and z_j are compared at most once. (任意两个元素至多比较一次,指示变量为)

$$X_{ij} = \begin{cases} 1 & \text{if } z_i \text{ and } z_j \text{ are compared} \\ 0 & \text{if } z_i \text{ and } z_j \text{ are not compared} \end{cases}$$

Then
$$X = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}$$

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The Expected Number of Comparisons

$$E(X) = E\left(\sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}\right)$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} E(X_{ij})$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} Pr(z_i \text{ and } z_j \text{ are compared})$$

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两个元素的比较概率(The Probability of Comparing Two Elements)

- If this pivot is neither z_i nor z_j, then z_i and z_j will never be compared.
- If this pivot is z_i or z_j, then z_i and z_j are compared when partitioning the chunk containing Z_{ij} around the pivot.

$$Pr(z_i \text{ and } z_i \text{ are compared}) =$$

$$Pr(z_i \text{ or } z_j \text{ is the first pivot chosen from } Z_{ij}) = \frac{2}{i-i+1}$$

期望比较次数(The Expected Number of Comparisons)

$$E(X) = E\left(\sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}\right)$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} E(X_{ij})$$

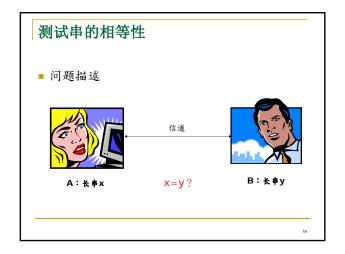
$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} Pr(z_i \text{ and } z_j \text{ are compared})$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{2}{j-i+1}$$
Lemma: The expected running time of random Quicksort is $O(n \lg n)$.
$$= \sum_{i=1}^{n-1} O(\lg n) = O(n \lg n)$$

随机算法分类

- Las Vegas算法: 随机算法总是或者给出正确的解,或者无解。
- Monte Carlo算法: 居多能够给出正确的解, 偶尔产生错误的解。可以采取措施使产生错误 解的概率降低到任意的程度。

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测试方法

- A将x发送给B;或B将y发送给A,然后判断x=y? 缺点:资源浪费
- A从x中取出一个短得多的串作为x的"指纹" 发送给B,B用同样的方法获得y的"指纹", 如果两者的"指纹"相同,则认为x=y,否则 x=y 不成立。

特点: 节约了资源, 但理论上不能100%正确。

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"指纹"生成

■ 对于一个串w,设I(w)是比特串w表示的一个整数,一种产生"指纹"的方法是选择一个素数p,通过"指纹"函数产生。

$$I_{p}(x) = I(x) \pmod{p}$$

 $I_{p}(x) \neq I_{p}(y)$, 则 $x \neq y$ $I_{p}(x) = I_{p}(y)$, 则x = y (?)

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测试串相等性算法

- 1、A从小于M的素数集中随机选择p;
- 2、A将p和I_n(x) 发送给B;
- 3、B检查是否 $I_p(x) = I_p(y)$, 确定x和y是否相等。

失败概率分析

■ 失败概率是:

 $\frac{\left|\left\{p\mid \bar{x}\otimes p<2^{n}\, \exists p\, \bar{x}\, \hat{x}(x)-I(y)\right\}\right|}{\pi(M)}\leq \frac{\pi(n)}{\pi(M)}$

n 是x的二进制的表示形式的位数 问题:

 π (n)是小于 n 的不同素数的个数 $\frac{\text{如何降低失败的概率?}}{\text{position}}$

 π (n) \rightarrow n/ln (n)

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