# **lesson13 Randomized Algorithms**

### 1.introduction

The algorithm behaves randomly – make random decisions as the algorithm processes the worst-case input.

key words: efficient and deterministic

**application**: Symmetry-breaking among processes in a distributed system(分布式系统中进程之间的对称性破坏)

为什么随机化? 因为简单

概率 一些复习:

$$Pr[A] := \text{the } \textit{probability} \text{ of the even } A$$
 $\overline{A} := \text{the } \textit{complementary} \text{ of the even } A \text{ ($A$ did not occur)}$ 

$$Pr[A] + Pr[\overline{A}] = 1$$

$$E[X] := \text{the } \textit{expectation} \text{ (the "average value") of the random variable } X$$

$$E[X] = \sum_{i=0}^{\infty} j \cdot Pr[X = j]$$

## 2.hiring problem

雇佣问题:

#### **[Example]** The Hiring Problem

- Fire an office assistant from headhunter
- ${\mathscr F}$  Interview a different applicant per day for N days
- Finterviewing Cost =  $C_i$  << Hiring Cost =  $C_h$
- Analyze interview & hiring cost instead of running time

Assume M people are hired.

Total Cost:  $O(NC_i + MC_h)$ 

naive solution:

```
int Hiring ( EventType C[ ], int N )
{    /* candidate 0 is a least-qualified dummy candidate */
    int Best = 0;
    int BestQ = the quality of candidate 0;
    for ( i=1; i<=N; i++ ) {
        Qi = interview( i ); /* Ci */
        if ( Qi > BestQ ) {
            BestQ = Qi;
            Best = i;
            hire( i ); /* Ch */
        }
}
```

```
}
return Best;
}
```

worst case: The candidates come in increasing quality order 时间复杂度O(N\*Ch)

best case: The candidates come in decreasing quality order random case:

X; number of hires

Xi = SI, hired

X = 
$$\frac{1}{1}$$
Xi

V, not hired)

E(x) =  $\frac{1}{1}$ E(xi)

assumption; any of first i candidates
is equally likely to be the best-qualified no far.

 $\Rightarrow E(xi) = 1/i$ 

E(x) =  $\frac{1}{1}$ I/i = In N + O Ci)

 $\Rightarrow$  O(Ch·In N + Ci·N)

## randomized algorithm

```
int RandomizedHiring ( EventType C[ ], int N )
{    /* candidate 0 is a least-qualified dummy candidate */
    int Best = 0;
    int BestQ = the quality of candidate 0;
//对输入进行随机排列! !! 优点: 不用担心输入的好坏: 缺点: 费时间
    randomly permute the list of candidates;

for ( i=1; i<=N; i++ ) {
        Qi = interview( i ); /* Ci */
        if ( Qi > BestQ ) {
            BestQ = Qi;
            Best = i;
            hire( i ); /* Ch */
        }
    }
}
```

### Radomized Permutation(排列) Algorithm

Target: Permute array A[]

Assign each element A[i] a random priority P[i], and sort

```
void PermuteBySorting ( ElemType A[ ], int N )
{
   for ( i=1; i<=N; i++ )
        A[i].P = 1 + rand()%(N3);
/* makes it more likely that all priorities are unique */
        Sort A, using P as the sort keys;
}</pre>
```

claim: : Permute By Sorting produces a uniform(N!个排列的发生可能性相同) random permutation of the input, assuming all priorities are distinct.

offline algorithm interview之前,得到priority

### Online Hiring Algorithm - hire only once

只雇佣一个,前k个,用来确定 BestQ,后面一旦遇到Qi大于 BestQ,就直接雇佣,退出;

两个问题:

What is the probability we hire the best qualified candidate for a given k?

$$\frac{1}{2} \times \frac{1}{2} = \frac{1}{2} \times \frac{1$$

What is the best value of k to maximize above probability?

$$f(k) = \frac{k}{N} \ln(\frac{N}{K}) \neq k \text{ so thox mox} (f(k))$$

$$f(k) = \frac{1}{N} \ln(\frac{N}{K}) + \frac{k}{N} \cdot \frac{k}{N} \cdot \frac{N}{K^2} = \frac{1}{N} (\ln \frac{N}{K} - 1)$$

$$k = \frac{N}{e} \text{ if, } f(k) \text{ forms} = \frac{1}{e}$$

so the probability of hiring the best is 1/e

## 3.quick sort

## **Deterministic Quicksort**

- $\mathcal{P}(N^2)$  worst-case running time
- $\Theta(N \log N)$  average case running time, assuming every input permutation is equally likely

random: How about choosing the pivot uniformly at random?

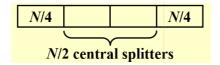
- **Central splitter**: = the pivot that divides the set so that each side contains\*\* **at least** n/4
- Modified Quicksort := always select a central splitter before recursions

Quicksort主要包含以下3步:

- 1. 从数组中取出一个元素,叫做主元 (pivot)
- 2. 重排序数组,使得所有小于pivot的元素在它前面,所有大于pivot的元素在它后面,等于pivot的元素放在哪面都行。这样的划分以后,pivot的位置已经排好了。这个过程叫做partition操作
- 3. 递归地应用步骤2到小于pivot的子数组和大于pivot的子数组

#### claim1:

The expected number of iterations needed until we find a central splitter is at most 2.



找到的概率50%嘛

#### run-time

定义: Type j: the subproblem S is of type j if  $N\left(\frac{3}{4}\right)^{j+1} \le |S| \le N\left(\frac{3}{4}\right)^{j}$ 

3/4: 如果pivot刚好选在两个1/4,3/4分界点上,那么对应最大子集

下界是由他自己得到的最大子问题, 所以肯定比他的子问题大;

claim2: Claim: There are at most  $\left(\frac{4}{3}\right)^{j+1}$  subproblems of type j.

最多有那么多个子问题,每个子问题size最小是他的倒数,那么相乘后肯定要满足不等式左边; 所以时间期望 = 子问题内部遍历调整的时间 \* 子问题的数量;

$$E[T_{type j}] = O(N\left(\frac{3}{4}\right)^{j}) \times \left(\frac{4}{3}\right)^{j+1} = O(N)$$

再考虑j的不同取值,  $\log_{4/3} N = O(\log N)$ 

所以,总时间  $O(N\log N)$