Data Structures and Algorithms Running time, Divide and Conquer February 02, 2016

Quicksort

In quick sort, we first decide on the pivot. This could be the element at any location in the input array. The function Partition is then invoked. Partition accomplishes the following: it places the pivot in the location that it should be in the output and places all elements that are at most the pivot to the left of the pivot and all elements greater than the pivot to its right. Then we recurse on both parts. The pseudocode for Quick Sort is as follows.

```
QSort(A[lo..hi])
  if hi <= lo then
    return
  else
    pivotIndex = floor((lo+hi)/2) (this could have been any location)
    loc = Partition(A, lo, hi, pIndex)
    QSort(A[lo..loc-1])
    QSort(A[loc+1..hi])</pre>
```

One possible implementation of the function Partition is as follows.

```
Partition(A, lo, hi, pIndex)
  pivot = A[pIndex]
  swap(A, pivotIndex, hi)
  left = lo
  right = hi-1
  while left <= right do
    if (A[left] <= pivot) then
      left = left + 1
    else
      swap(A, left, right)
      right = right - 1
  swap(A, left, hi)
  return left</pre>
```

The worst case running time of the algorithm is given by

$$T(n) = \begin{cases} 1, & n = 1 \\ T(n-1) + cn, & n \ge 2 \end{cases}$$

Hence the worst case running time of QSort is $\Theta(n^2)$.

An instance where the quick sort algorithm in which the pivot is always the first element in the input array performs poorly is an array in descending order of its elements.

Randomized Quick Sort

In randomized version of quick sort, we pick a pivot uniformly at random from all possibilities. We will now show that the expected number of comparisons made in randomized quick sort equals $2n \ln n + O(n)$.

Theorem. For any input, the expected number of comparisons made by randomized quick sort is $2n \ln n + O(n)$.

Proof. Let y_1, y_2, \ldots, y_n be the input elements in sorted order. Let X be a random variable denoting the total number of pair-wise comparisons made between elements of the input array A. Let X_{ij} be a random variable denoting the total number of times elements y_i and y_j are compared during the algorithm. Observe the following.

- comparisons between elements in the input array are done only in the function Partition.
- there are n-1 distinct pivots chosen in the algorithm and hence n-1 calls to Partition.
- two elements are compared only if one of them is a pivot.

Let X_{ij}^k be a random variable that is 1, if elements y_i and y_j are compared in the kth call to Partition. Note that

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

By the linearity of expectation, we have

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

We will now calculate $\mathbf{E}[X_{ij}]$. By the linearity of expectation, we have

$$\mathbf{E}[X_{ij}] = \sum_{k=1}^{n-1} \mathbf{E}[X_{ij}^k] = \mathbf{E}[X] = \sum_{k=1}^{n-1} \Pr[X_{ij}^k = 1]$$

Let z be the first call to Partition during which one of the elements from $y_i, y_{i+1}, \ldots, y_n$ is used as the pivot. From our observations above, note that for all h and all ℓ such that $1 \le h < z$ and $z < \ell < n$, $X_{ij}^h = 0$ and $X_{ij}^\ell = 0$. If one of y_i or y_j is chosen as the k^{th} pivot then clearly, $X_{ij}^k = 1$, otherwise, $X_{ij}^k = 0$, and y_i and y_j will be separated into different sublists and hence will never be compared again. Thus we have

$$\mathbf{E}[X_{ij}] = \Pr[X_{ij}^k = 1] = \frac{2}{i - i + 1}$$
(2)

Combining equations (??) and (??), we get

$$\mathbf{E}[X] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \mathbf{E}[X_{ij}]$$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{2}{j-i+1}$$

$$= \sum_{k=2}^{n} \frac{2(n-k+1)}{k}$$

$$= (n+1) \sum_{k=2}^{n} \frac{2}{k} - 2(n-1)$$

$$= 2(n+1) \sum_{k=1}^{n} \frac{1}{k} - 2(n-1) - 2(n+1)$$

$$= 2(n+1) \ln n + c - 4n \qquad (c \text{ is a constant less than 1})$$

$$= 2n \ln n + \Theta(n)$$

Example. Find the running time expressed by the following recurrence. Assume that n is a power of 3.

$$T(n) = \begin{cases} 5T(n/3) + n^2, & n \ge 2\\ 1, & n = 1 \end{cases}$$

$$T(n) = 5T(n/3) + n^{2}$$

$$= 5^{2}T(n/3^{2}) + 5(n/3)^{2} + n^{2}$$

$$= 5^{3}T(n/3^{3}) + 5^{2}(n/3^{2})^{2} + 5n^{2}/3^{2} + n^{2}$$
...
...
$$= 5^{k}T(n/3^{k}) + n^{2}\left(1 + \frac{5}{9} + \left(\frac{5}{9}\right)^{2} + \dots + \left(\frac{5}{9}\right)^{k-1}\right)$$

The recursion bottoms out when $n/3^k = 1$, i.e., $k = \log_3 n$. Thus, we get

$$T(n) = 5^{\log_3 n} T(1) + n^2 \sum_{i=0}^{k-1} (5/9)^i$$

$$= 5^{\log_3 n} + n^2 \left(\frac{(5/9)^{\log_3 n} - 1}{(5/9) - 1} \right)$$

$$= 5^{\log_5 n/\log_5 3} + \frac{9n^2}{4} \left(1 - (n^{\log_3 5}/n^2) \right)$$

$$= n^{\log_3 5} + \frac{9n^2}{4} - \frac{9n^{\log_3 5}}{4}$$

$$= \frac{9n^2}{4} - \frac{5n^{\log_3 5}}{4}$$

$$= \Theta(n^2)$$

Example. Solve the following recurrence. Assume that n is a power of 2.

$$T(n) = \begin{cases} 3T(n/2) + n, & n \ge 2\\ 1, & n=1 \end{cases}$$

Solution.

$$T(n) = 3T(n/2) + n$$

$$= 3^{2}T(n/2^{2}) + 3n/2 + n$$

$$= 3^{3}T(n/2^{3}) + (3/2)^{2}n + 3n/2 + n$$
...
...
$$= 3^{k}T(n/2^{k}) + n\sum_{i=0}^{k-1} (3/2)^{i}$$

The recursion bottoms out when $n/2^k = 1$, i.e., $k = \lg n$. Thus, we get

$$\begin{split} T(n) &= 3^{\log_2 n} T(1) + n \sum_{i=0}^{\lg n-1} (3/2)^i \\ &= 3^{\log_2 n} + n \left(\frac{(3/2)^{\lg n} - 1}{3/2 - 1} \right) \\ &= n^{\frac{\log_3 n}{\log_3 2}} + 2n((3/2)^{\log_{3/2} n/\log_{3/2} 2} - 1) \\ &= n^{\log_2 3} + 2n(n^{\lg(3/2)} - 1) \\ &= n^{\log_2 3} + 2n(n^{\lg 3 - \lg 2)} - 1) \\ &= n^{\log_2 3} + 2n(n^{\lg 3 - \lg 2)} - 1) \\ &= n^{\log_2 3} + 2n(n^{\lg 3 - 1}) - 2n \\ &= n^{\log_2 3} + 2n^{\lg 3} - 2n \\ &= 3n^{\log_2 3} - 2n \\ &= \Theta(n^{\lg 3}) \qquad \text{(can be argued by setting } c = 3 \text{ and } n_0 = 1) \end{split}$$