Sorting atomic items

Chapter 5

Distribution based sorting paradigms

The distribution-based sorting

QuickSort is an in place algorithm, but...

Consider the stack for the recursive calls.

For balanced partitions: O(logn) space

Worst case of unbalanced partitions: $\Omega(n)$ calls, $\Theta(n)$ space!!

QuickSort is modified.

We can bound the recursive depth. Algorithm Bounded Based on the fact that: Quicksort does not depend on the order in which recursive calls are executed!

Small arrays can be better sorted with InsertionSort (when n is typically of the order of tens).

The modified version mixes one recursive call with an iterative while loop.

Algorithm Bounded(S, I, j)

Algorithm 5.6 The binary quick-sort with bounded recursive-depth:

```
1: while (j - i > n_0) do
         r = pick the position of a "good pivot";
2:
         swap S[r] with S[i];
3:
         p = Partition(S, i, j);
4:
        if (p \le \frac{i+j}{2}) then
                                                        n > no
5:
              BoundedQS(S, i, p - 1);
6:
              i = p + 1;
7:
         else
8:
              BoundedQS(S, p + 1, j);
9:
              j = p - 1;
10:
         end if
11:
12: end while

 INSERTIONSORT(S, i, j);

                                               n ≤ no
```

Algorithm Bounded(S, I, j)

- The recursive call is executed on the smaller part of the partition
- It drops the recursive call on the larger part of the partition in favor of another execution of the while-loop.
- Ex:

Partition

Only one recursive call on S(i,p-1); For the larger part S(i+1,j) we iterate to the while loop

- Technique: Elimination of Tail Recursion
- Bounded takes O(nlogn) time in average and O(logn) space.

- 2-level model: M = internal memory size; B = block size;
- Split the sequence S into $k = \Theta(M/B)$ sub-sequences using k-1 pivots.
- We would like balanced partitions, that is of $\Theta(n/k)$ items each.
- Select k-1 s_1 , s_2 , ... s_{k-1} "good pivots" is not a trivial task!

```
Let bucket Bi the portion of S between pivot s_{i-1} and s_i. We want guarantee that |Bi| = \Theta(n/k) for all buckets. So, at the first step of Multi-way QuickSort size of portions n/k at the second step the size of portions n/k^2 at the third step the size of portions n/k^3
```

Stop when $n/k^i \le M$: $n/M \le k^i$, i is the number of steps $i \ge \log_k n/M = \log_{M/B} n/M$

This number of steps is enough to have portions shorter than M, and sorted in internal memory!

Partition takes O(n/B) I/O's (dual to MS): 1 input block, k output blocks.

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Find k good pivots efficiently.

Randomized strategy called oversampling.

 $\Theta(ak)$ items are sampled, a parameter of the oversampling.

Algorithm 5.7 Selection of k-1 good pivots via oversampling

```
    Take (a + 1)k - 1 samples at random from the input sequence;
```

Sort them into an ordered sequence A;

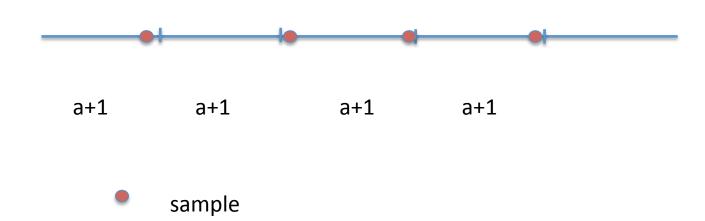
```
3: For i = 1, ..., k - 1, pick the pivot s_i = A[(a + 1)i];
```

return the pivots s_i;

Θ(ak) candidates Θ(ak)log(ak) time Select k-1 pivots evenly distributed

Balanced selection of $s_i = A[(a+1)i]$ should provide good pivots!!

• K=5

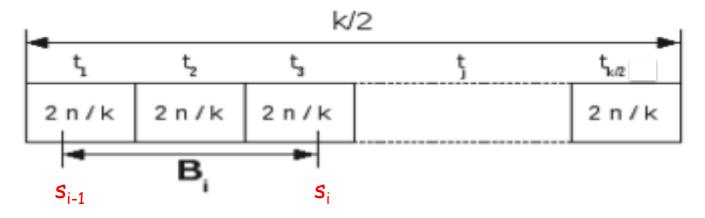


- The larger is a the closer to ⊖(n/k)
- If a=n/k the elements of set A cannot be sorted in M!
- If a = 0 the selection is fast, but unbalanced partitions are is more probable.
- Good choice for a: $\Theta(\log k)$. Pivot-selection costs $\Theta(k\log^2 k)$

Lemma. Let $k \ge 2$, a + 1 = 12 In k. A sample of (a+1)k-1 suffices to ensure that all buckets receive less than 4n/k elements, with probability at least $\frac{1}{2}$.

Proof. We find un upper bound to complement event: there exists 1 bucket containing more than 4n/k elements with probability at most $\frac{1}{2}$. Failure sampling.

Consider the sorted version of S, S'. Logically split S' in k/2 segments (t_1 , t_2 , ..., $t_{k/2}$) of 2n/k elements each.



- The event is that there exists a bucket Bi with more that 4n/k items. It spans more than one segment: pivots \mathbf{s}_{i-1} and \mathbf{s}_i fall outside \mathbf{t}_2 .
- In t_2 fall less than (a+1) samples (see selection algorithm: between 2 pivots there are a+1 samples, hence in t_2 there are less).
- Pr (exists B_i: |B_i| ≥ 4 n/k) ≤ Pr (exists t_j: contains < (a+1) samples)
 ≤ k/2 Pr (a specific segment contain < (a+1) samples)

Since k/2 is the number of segments.

- Pr (1 sample goes in a given segment) = (2n/k)/n = 2/kIf drawn uniformly at random from S (and S').
- Let X the number of samples going in a given segment, we want to compute:

$$Pr(X < a+1)$$

Observe: E(X) = ((a+1)k-1) × 2/k ≥ 2(a+1)-2/k, per k ≥ 2
 E(X) ≥ 2(a+1) - 1 ≥ 3/2(a+1) for all a ≥ 1
 a+1 ≤ 2/3 E(X) = (1-1/3) E(X)

By Chernoff bound

Pr
$$(X < (1 - \delta) E(X) \le e^{\delta^2/2}E(X)$$
)

Setting $\delta = 1/3$ and assume $a+1 = 12 \ln k$

Pr
$$(X < a+1) \le P(X \le (1-1/3)E(X)) \le e^{-E(X)/18} \le e^{-(a+1)/12} = e^{-lnk} = 1/k$$

Pr (X < a+1) ≤ 1/k

We have already derived:

- Pr (exists $B_i : |B_i| \ge 4 \text{ n/k}) \le k/2 \times Pr$ (a segment contain (a+1) samples)
- Pr (exists $B_i : |B_i| \ge 4 \text{ n/k}) \le 1/2$ complement event of the lemma
- All buckets receives less than 4n/k elements with probability > 1/2

Dual Pivot QuickSort

- Good strategy in practice no theoretical result.
- Empirical good results in average.



- p, q pivots $l, k, g indices \rightarrow 4 pieces$
- items smaller than p
- 2. items larger or equal to p and smaller or equal to q.
- 3. items not jet considered
- 4. items greater than q

Dual Pivot QuickSort

Similar to the 3-ways Partition: maintains the invariants.

- Items equal to the pivot are not treated separately.
- 2 indices move rightward, I and k, while g moves leftward.
- Termination: k ≥ g.
- For item k, compare S[k]: p,
 if S[k]

```
else if S[k] > q decrease g while S[g] > q and g \neq k
the last value of g : S[g] \leq q
exchange S[k] and S[g]
```

.....

The comparison with S[k] drives the phases possibly including a long shift to the left. The nesting of comparison is the key for the efficiency of the algorithm.

Dual Pivot Partition

```
5 12 9 12 13 15 17 19 12 26 18 22 20
                                                p=12 q= 17
                                    exchange 19 and 17, k++, g—
5 12 9 12 13 15 17 19 12 26 18 22 20
                                 I++ exchange 12 and 13, k++
                           k g
5 12 9 12 12 15 17 19 12 13 26 18 22 20
                                              no exchange
```

Dual Pivot QuickSort

You can find the complete code description and the visualization of the algorithm on youtube by searching for Dual pivot QuickSort.

Conclusions:

Even a very old, classic algorithm such as QuickSort can be speed up and innovated!