Algorithms Chapter 8 Sorting in Linear Time

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Outline

- Lower bounds for sorting
- Counting sort
- Radix sort
- Bucket sort

Overview

▶ Sort *n* numbers in *O*(*n* lg *n*) time

- Merge sort and heapsort achieve this upper bound in the worst case.
- Quicksort achieves it on average.
- For each of these algorithms, we can produce a sequence of n input numbers that causes the algorithm to run in $\Theta(n \lg n)$ time.

Comparison sorting

- ▶ The only operation that may be used to gain order information about a sequence is comparison of pairs of elements.
- ▶ All sorts seen so far are comparison sorts: insertion sort, selection sort, merge sort, quicksort, heapsort.

Lower bounds for sorting

Lower bounds

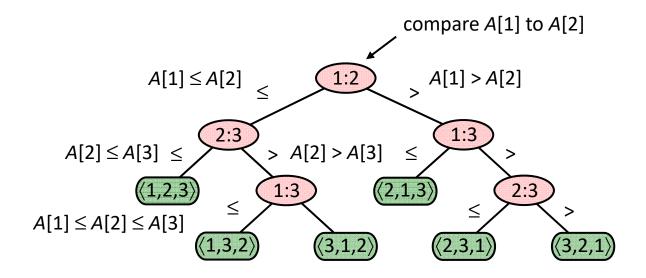
- $ightharpoonup \Omega(n)$ to examine all the input.
- All sorts seen so far are $\Omega(n \lg n)$.
- We'll show that $\Omega(n \lg n)$ is a lower bound for comparison sorts.

Decision tree

- Abstraction of any comparison sort.
- A full binary tree.
- Represents comparisons made by
 - a specific sorting algorithm
 - ▶ on inputs of a given size.
- Control, data movement, and all other aspects of the algorithm are ignored.

Decision tree

For insertion sort on 3 elements:



- ▶ How many leaves on the decision tree?
 - ▶ There are $\ge n!$ leaves, because every permutation appears at least once.

Properties of decision trees_{1/3}

- ▶ **Lemma 1** Any binary tree of height h has $\leq 2^h$ leaves.
 - ▶ *Proof*: By induction on *h*.
 - Basis:
 - ▶ h = 0. Tree is just one node, which is a leaf. $2^h = 1$.
 - Inductive step:
 - Assume true for height = h 1.
 - ▶ Extend tree of height h-1 by making as many new leaves as possible.
 - Each leaf becomes parent to two new leaves.

Properties of decision trees_{2/3}

▶ **Theorem 1** Any decision tree that sorts n elements has height $\Omega(n \lg n)$.

Proof:

- ▶ $\ell \ge n!$, where $\ell = \#$ of leaves.
- ▶ By lemma 1, $n! \le \ell \le 2^h$ or $2^h \ge n!$.
- ▶ Take logs: $h \ge \lg(n!)$.
- Use Stirling's approximation: $n! > (n/e)^n$

```
h \ge \lg(n!)
> \lg(n/e)^n
= n\lg(n/e)
= n\lg n - n\lg e
= \Omega(n\lg n).
```

Properties of decision trees_{3/3}

Corollary 1 Heapsort and merge sort are asymptotically optimal comparison sorts.

Proof:

The $O(n \lg n)$ upper bounds on the running times for heapsort and merge sort match the $\Omega(n \lg n)$ worst-case lower bound from Theorem 1.

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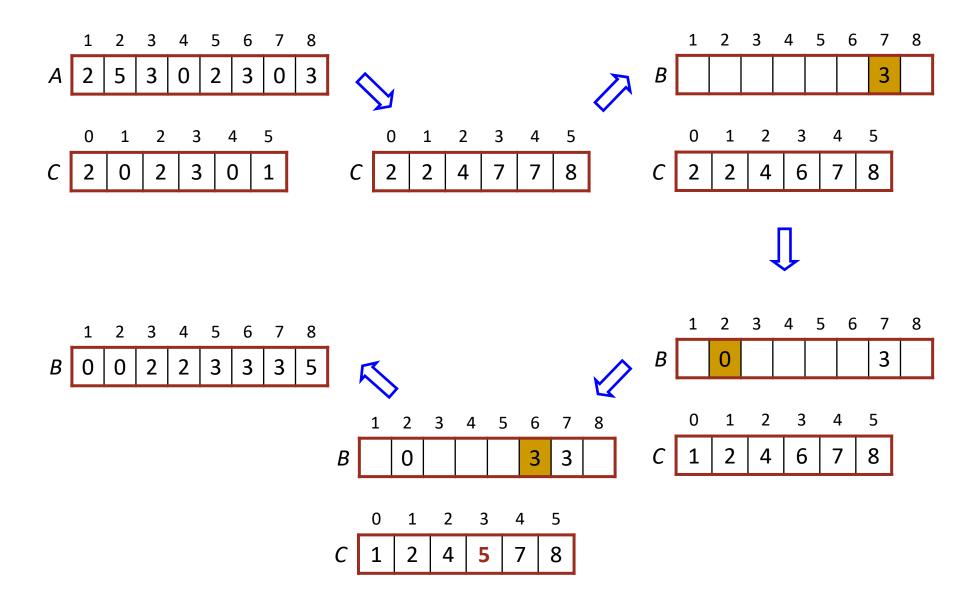
Counting sort

- Non-comparison sorts.
- ▶ Depends on a **key assumption**: numbers to be sorted are integers in $\{0, 1, ..., k\}$.
- ▶ Input: A[1..n], where $A[j] \in \{0, 1, ..., k\}$ for j = 1, 2, ..., n. Array A and values n and k are given as parameters.
- Output: B[1..n], sorted. B is assumed to be already allocated and is given as a parameter.
- **▶ Auxiliary storage:** *C*[0 . . *k*].
- ▶ Worst-case running time: $\Theta(n+k)$.

The Counting-Sort procedure

```
COUNTING-SORT(A, B, k)
        for i \leftarrow 0 to k
              do C[i] \leftarrow 0
      for j \leftarrow 1 to length[A]
do C[A[j]] \leftarrow C[A[j]] + 1 \Theta(n)
       /* C[i] now contains the number of elements equal to i. */
        for i \leftarrow 1 to k
     \operatorname{do} C[i] \leftarrow C[i] + C[i-1] \quad \right\} \Theta(k)
       /* C[i] now contains the number of elements less than or equal to i. */
        for j \leftarrow length[A] downto 1
            \operatorname{do} B[C[A[j]]] \leftarrow A[j] \qquad \bigg\} \Theta(n)
10.
                  C[A[j]] \leftarrow C[A[j]] - 1
11.
```

▶ The running time: $\Theta(n+k)$.



Properties of counting sort

- A sorting algorithm is said to be **stable** if keys with same value appear in same order in output as they did in input.
- Counting sort is stable because of how the last loop works.
- Counting sort will be used in radix sort.

Outline

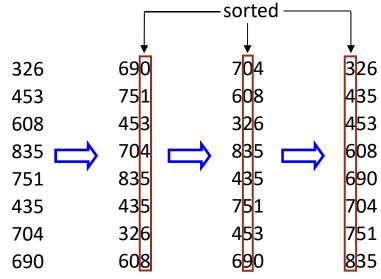
- Lower bounds for sorting
- Counting sort
- **▶** Radix sort
- Bucket sort

Radix sort

▶ Key idea: Sort least significant digits first.

RADIX-SORT(A, d)

- 1. for $i \leftarrow 1$ to d
- **do** use a stable sort to sort array A on digit i
- An example:



Correctness of radix sort

- Proof: By induction on number of passes (i in pseudocode).
- **Basis:**
 - i = 1. There is only one digit, so sorting on that digit sorts the array.
- **▶** Inductive step:
 - Assume digits 1, 2,..., i 1 are sorted.
 - ▶ Show that a stable sort on digit *i* leaves digits 1, 2,..., *i* sorted:
 - If 2 digits in position i are different, ordering by position i is correct, and positions 1,..., i − 1 are irrelevant.
 - If 2 digits in position *i* are equal, numbers are already in the right order (by inductive hypothesis). The stable sort on digit *i* leaves them in the right order.

Time complexity of radix sort

- Assume that we use counting sort as the intermediate sort.
- ▶ When each digit is in the range 0 to k-1, each pass over n d-digit number takes time $\Theta(n + k)$.
- ▶ There are d passes, so the total time for radix sort is $\Theta(d(n+k))$.
- If k = O(n), time = $\Theta(dn)$.
- **Lemma 2:** Given n d-digit numbers in which each digit can take on up to k possible values, RADIXSORT correctly sorts these numbers in $\Theta(d(n+k))$ time.

Break each key into digits

▶ **Lemma 3:** Given n b-bit numbers and any positive integer $r \le b$, RADIX-SORT correctly sorts these numbers in $\Theta((b/r)(n + 2^r))$ time.

Proof

- We view each key as having $d = \lceil b/r \rceil$ digits of r bits each.
- ▶ Each digit is an integer in the range 0 to $2^r 1$, so that we can use counting sort with $k = 2^r$.
- ▶ Each pass of counting sort takes time $\Theta(n+k) = \Theta(n+2^r)$.
- ▶ A total running time of $\Theta(d(n+2^r)) = \Theta((b/r)(n+2^r))$.

For example:

- ▶ 32-bit words, 8-bit digits.
- $b = 32, r = 8, d = 32/8 = 4, k = 2^8 1 = 255.$

The main reason

- How does radix sort violate the ground rules for a comparison sort?
 - ▶ Using counting sort allows us to gain information about keys by means other than directly comparing 2 keys.
 - Used keys as array indices.

Outline

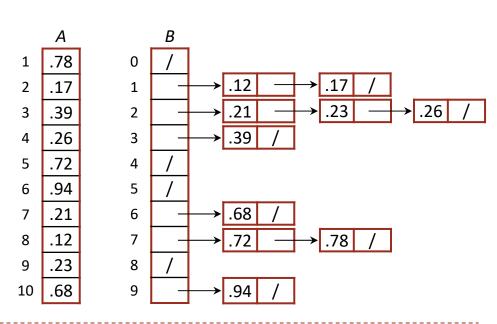
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Bucket sort

Assumes the input is generated by a random process that distributes elements uniformly over [0, 1).

Key idea:

- ▶ Divide [0, 1) into *n* equal-sized **buckets**.
- ▶ Distribute the *n* input values into the buckets.
- Sort each bucket.
- Then go through buckets in order, listing elements in each one.



The Bucket Sort procedure

- ▶ **Input:** A[1...n], where $0 \le A[i] < 1$ for all i.
- ▶ Auxiliary array: B[0..n-1] of linked lists, each list initially empty.

```
BUCKET-SORT(A, n)

1. for i \leftarrow 1 to n

2. do insert A[i] into list B[\lfloor n \cdot A[i] \rfloor]

3. for i \leftarrow 0 to n - 1

4. do sort list B[i] with insertion sort

5. concatenate lists B[0], B[1], ..., B[n-1] together in order

6. return the concatenated lists
```

Correctness of bucket sort

- Consider A[i], A[j]. Assume without loss of generality that $A[i] \le A[j]$.
- ▶ Then $\lfloor n \cdot A[i] \rfloor \leq \lfloor n \cdot A[j] \rfloor$.
- So A[i] is placed into the same bucket as A[j] or into a bucket with a lower index.
- If same bucket, insertion sort fixes up.
- ▶ If earlier bucket, concatenation of lists fixes up.

Time complexity of bucket sort

- Relies on no bucket getting too many values.
- All lines of algorithm except insertion sorting take $\Theta(n)$ altogether.
- Intuitively, if each bucket gets a constant number of elements, it takes O(1) time to sort each bucket $\rightarrow O(n)$ sort time for all buckets.
- We "expect" each bucket to have few elements, since the average is 1 element per bucket.

Time complexity of bucket sort

- ▶ Define a random variable: n_i = the number of elements placed in bucket B[i].
- Because insertion sort runs in quadratic time, bucket sort time is $T(n) = \theta(n) + \sum_{i=0}^{n-1} O(n_i^2)$.
- ▶ Take expectations of both sides:

$$E[T(n)] = E\left[\theta(n) + \sum_{i=0}^{n-1} O(n_i^2)\right]$$

$$= \theta(n) + \sum_{i=0}^{n-1} E[O(n_i^2)]$$
Therefore, $E[T(n)] = \theta(n) + \sum_{i=0}^{n-1} O(2-1/n)$

$$= \theta(n) + \sum_{i=0}^{n-1} O(E[n_i^2])$$

$$= \theta(n).$$

$$E[aX] = aE[X]$$
Claim that $E[n_i^2] = 2 - 1/n$ for $0 \le i \le n - 1$.

$$E[aX] = aE[X]$$

Proof of claim

► Claim: $E[n_i^2] = 2 - 1/n$ for $0 \le i \le n - 1$.

Proof

- ▶ $Pr{A[j] \text{ falls in bucket } i} = p = 1/n$.
- The probability that $n_i = k$ follows the binomial distribution b(k; n, p).
- ▶ So, $E[n_i] = np = 1$ and variance $Var[n_i] = np(1-p) = 1 1/n$.
- For any random variable X, we have $E[n_i^2] = Var[n_i] + E^2[n_i]$

$$= 1 - \frac{1}{n} + 1^{2}$$
$$= 2 - \frac{1}{n}.$$

Notes

- Again, not a comparison sort. Used a function of key values to index into an array.
- This is a probabilistic analysis. We used probability to analyze an algorithm whose running time depends on the distribution of inputs.
- ▶ Different from a randomized algorithm, where we use randomization to impose a distribution.
- ▶ With bucket sort, if the input isn't drawn from a uniform distribution on [0, 1), all bets are off (performance-wise, but the algorithm is still correct).