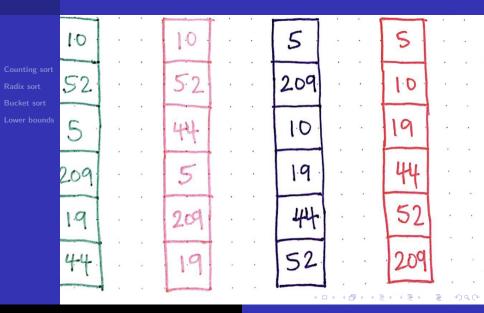
Fast Sorting Algorithms



Sorting algorithms on values in a known range

Counting sort

CLRS Ch.8

- Counting sort
- Radix sort
- Bucket sort
- Lower bounds for general sorting
- The algorithms will sort an array A[n] of non-negative integers in the range [0, r].

Sorting algorithms on values in a known range

Counting sort Radix sort Bucket sort

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- The algorithms will sort an array A[n] of non-negative integers in the range [0, r].
- The complexity of the algorithms depends on both n and r.
- For some values of r, the algorithms have cost O(n) or $o(n \log n)$.

The counting sort algorithm,

Counting sort

Radix sort

Bucket sort

Lower bound

Counting sort

Radix sort

Lower bound

The counting sort algorithm,

• consider all possible values $i \in [0, r]$.

Counting sort

Radix sort Bucket sort Lower bound The counting sort algorithm,

- consider all possible values $i \in [0, r]$.
- For each of them, count how many elements in *A* are smaller or equal to *i*.

Counting sort
Radix sort
Bucket sort

The counting sort algorithm,

- consider all possible values $i \in [0, r]$.
- For each of them, count how many elements in *A* are smaller or equal to *i*.
- Use this information to place the elements in the right order.

Counting sort
Radix sort
Bucket sort

The counting sort algorithm,

- consider all possible values $i \in [0, r]$.
- For each of them, count how many elements in *A* are smaller or equal to *i*.
- Use this information to place the elements in the right order.
- The input A[n], is an array of integers in the range [0, r].
- Uses: B[n] (output) and C[r+1] (internal).

Counting sort: Algorithm

Counting sort
Radix sort
Bucket sort

```
CountingSort (A, r)
for i = 0 to r do
  C[i] = 0
for i = 0 to n - 1 do
  C[A[i]] = C[A[i]] + 1
                                      \{C[j] = |\{i \mid A[i] = j\}|\}
for i = 1 to r do
  C[i] = C[i] + C[i-1]
                                       \{C[j] = |\{i \mid A[i] \le j\}|\}
for i = n - 1 downto 0 do
  B[C[A[i]]] = A[i];
  C[A[i]] = C[A[i]] - 1
                                { C holds the sorted elements}
```

Counting sort: Cost

Counting sort
Radix sort
Bucket sort

Jucket sort

CountingSort
$$(A, r)$$

for $i = 0$ to r do
 $C[i] = 0$ $\{O(r)\}$
for $i = 0$ to $n - 1$ do
 $C[A[i]] = C[A[i]] + 1$ $\{O(n)\}$
for $i = 0$ to r do
do $C[i] = C[i] + C[i - 1]$ $\{O(r)\}$
for $i = n - 1$ downto 0 do
 $B[C[A[i]] - 1] = A[i];$
 $C[A[i]] = C[A[i]] - 1$ $\{O(n)\}$

$$T(n) = O(n+r),$$

Counting sort: Cost

Counting sort
Radix sort
Bucket sort
Lower bounds

```
CountingSort (A, r)
for i = 0 to r do
  C[i] = 0
                                                      \{O(r)\}
for i = 0 to n - 1 do
  C[A[i]] = C[A[i]] + 1
                                                     \{O(n)\}
for i = 0 to r do
  do C[i] = C[i] + C[i-1]
                                                     \{O(r)\}
for i = n - 1 downto 0 do
  B[C[A[i]] - 1] = A[i]:
  C[A[i]] = C[A[i]] - 1
                                                     \{O(n)\}
```

$$T(n) = O(n+r)$$
, for $r = O(n)$, $T(n) = O(n)$.

Counting sort: stability

Counting sort
Radix sort
Bucket sort

An important property of counting sort is that it is stable: numbers with the same value appear in the output in the same order as they do in the input.

Radix sort: What does radix mean?

Radix sort

Radix means the base in which we express an integer

Radix 10=Decimal; Radix 2= Binary; Radix 16=Hexadecimal; Radix 20 (The Maya numerical system)



0	1	2	3	4
5	6	7 ••	8 •••	9

Binary	Hex	Decimal
0000	0	0
0001	1	1
0010	2	2
0011	3	2 3 4
0100	4	4
0101	5	5
0110	6	6
0111	7	7
1000	8	8
1001	9	9
1010	A	10
1011	В	11
1100	C	12
1101	D	13
1110	E	14
1111	F	15

Radix Change: Example

Radix sort

Bucket sort

- To convert an integer from binary to decimal: $1011 = 1 \times 2^3 + 0 \times 2^2 + 1 \times 2^1 + 1 \times 2^0 = 11$
- To convert an integer from decimal to binary: Repeatedly dividing the enter by 2, will give a result plus a remainder:

$$19 \Rightarrow \underbrace{19/2}_{1} \underbrace{9/2}_{1} \underbrace{4/2}_{0} \underbrace{2/2}_{01} = \underbrace{10011}_{0}$$

■ To transform an integer radix 16 to decimal: $(4CF5)_{16} = (4 \times 16^3 + 12 \times 16^2 + 15 \times 16^1 + 5 \times 16^0) = 19701$

- To convert $(4CF5)_{16}$ into binary you have to expand each digit to its binary representation. In the above example, $(4CF5)_{16}$ in binary is 0011110011110101
- To convert an integer in binary to radix 16: Make groups of 4 from left to right and replace by the corresponding digit 1101010010100010000101101111110100 in HEX is 1A9442DF4

RADIX LSD algorithm

Given an array A with n numbers, each one with d digits in base b the Radix Least Significant Digit, algorithm is

Radix sort
Bucket sort

Bucket sort

RADIX LSD (A, d, b) **for** i = 1 to d **do**

Use a stable sorting algorithm to sort A according to the i-th digit values.

The values to sort are in the range $[0, 2^d)$.

Counting sort

Radix sort

Bucket sort

	329		72 <mark>0</mark>
	475		47 <mark>5</mark>
	657		35 <mark>5</mark>
Radix sort	839	\Rightarrow	43 <mark>6</mark>
	436		65 <mark>7</mark>
	720		32 <mark>9</mark>
	355		83 <mark>9</mark>

	329		72 <mark>0</mark>		7 <mark>2</mark> 0
	475		47 <mark>5</mark>		3 <mark>2</mark> 9
	657		35 <mark>5</mark>		4 <mark>3</mark> 6
Radix sort	839	\Rightarrow	436	\Rightarrow	839
	436		65 <mark>7</mark>		3 <mark>5</mark> 5
	720		32 <mark>9</mark>		6 <mark>5</mark> 7
	355		83 <mark>9</mark>		4 <mark>7</mark> 5

29
55
36
75
57
20
39

Theorem

RADIX LSD sorts correctly the n given numbers.

Radix sort

Ducket sort

Lower bound:

Counting sort

Radix sort

Bucket sort

Lower bound

Theorem

RADIX LSD sorts correctly the n given numbers.

Induction on *d*.

Theorem

Radix sort

RADIX LSD sorts correctly the n given numbers.

Induction on d.

Base: If d = 1 the stable sorting algorithm sorts correctly.



Radix sort

Lower bound

Theorem

RADIX LSD sorts correctly the n given numbers.

Induction on d.

Base: If d = 1 the stable sorting algorithm sorts correctly.

IH: Assume that it is true for d-1 digits.

Looking at the the *d*-th digit, we have

Counting sort

Radix sort
Bucket sort

Theorem

RADIX LSD sorts correctly the n given numbers.

Induction on d.

Base: If d=1 the stable sorting algorithm sorts correctly. IH: Assume that it is true for d-1 digits. Looking at the d-th digit, we have

• if $a_d < b_d$, a < b and the algorithm places a before b,

Theorem

Radix sort

RADIX LSD sorts correctly the n given numbers.

Induction on d.

Base: If d=1 the stable sorting algorithm sorts correctly. IH: Assume that it is true for d-1 digits. Looking at the d-th digit, we have

- if $a_d < b_d$, a < b and the algorithm places a before b,
- if a_d = b_d, as we are using a stable sorting, a and b remain in the same order as in the previous step.
 By IH, they are already the correct one.

Time complexity

Radix sort

Bucket sort

Lower bound

Given n numbers, each number with at most d digits, and each digit in the range 0 to b, if we use counting sorting at each round of RADIX LSD:

$$T(n,d,b) = \Theta(d(n+b)).$$

Time complexity

Radix sort
Bucket sort

Given n numbers, each number with at most d digits, and each digit in the range 0 to b, if we use counting sorting at each round of RADIX LSD:

$$T(n,d,b) = \Theta(d(n+b)).$$

- Consider that each number has a value up to f(n).
- Then the number of digits is $d = \lceil \log_b f(n) \rceil$, so $T(n,b) = \Theta(\log_b f(n)(n+b))$,
- if $\log_b f(n) = \omega(1)$, $T(n, b) = \omega(n)$ and RADIX is not linear.

RADIX: selecting the base

Counting sort

Radix sort

Bucket sort

Can we tune the parameters?

- Yes, in some cases, we can select the best radix to express the input values.
- For numbers in binary, we can select as new radix \hat{b} a power of 2. This simplifies the computation as we have only to look to pieces of bits to change from one representation to anoter.
- For ex., if we have numbers of d=64 bits (b=2), and take the new radix to be $\hat{b}=2^8$, we have $\hat{d}=4$ new digits per number.

11001010 00110101011101001110000

RADIX: selecting the base

Counting sort

Radix sort

Bucket sort

Given n, d-bits integers, we want to choose an integer e, 1 < e < d to use as new radix $\hat{b} = 2^e$.

- In the new radix, the number of digits is $\hat{d} = \lceil d/e \rceil$ digits,
- Running RADIX LSD with base 2^e has cost the new $\hat{d} = \lceil d/e \rceil$ digits,

$$T(n) = \Theta(\hat{d}(n+2^e)) = \Theta((d/e)(n+2^e)).$$

- The best choice for e is roughly $\lceil \lg n \rceil$.
 - Then, $2^e = O(n)$.
 - So, the cost is, $O(\frac{d}{\lg n}n)$.
 - Which provides, linear cost if $\frac{d}{\lg n} = O(1)$.

Bucket sort

Counting so

Bucket sort

ower bound

- Suppose the values to sort are in the range $[0 \dots m-1]$.
- The algorithm starts with m empty buckets numbered 0 to m-1.
- Scan the list and place element s[i] in bucket s[i].
- Output the buckets in order.

Bucket sort

Radix sort
Bucket sort

- Suppose the values to sort are in the range $[0 \dots m-1]$.
- The algorithm starts with m empty buckets numbered 0 to m-1.
- Scan the list and place element s[i] in bucket s[i].
- Output the buckets in order.
- It needs an array of buckets.
- The values in the list to be sorted are the indexes to the buckets.
- No comparisons are done.

4	2	1	2	0	4	1	2	2	3	4	1	0	
---	---	---	---	---	---	---	---	---	---	---	---	---	--

Bucket sort

4 2 1 2 0 4 1 2 2 3 4 1 0

 $\begin{vmatrix} 1 & 2 & 4 \\ 0 & 1 & 2 \\ 0 & 1 & 2 & 3 \end{vmatrix}$

Radix sort
Bucket sort

Lower bounds

4 2 1 2 0 4 1 2 2 3 4 1 0

		2		
	1	2 2 2		4
0	1	2		4
0	1	2	3	4

Bucket sort

0	0	1	1	1	2	2	2	2	3	4	4	4
l .				l .	l		l		1			

Bucket sort: values or keys?

Bucket sort

- When sorting values, each bucket can be just a counter.
 - When sorting entries according to keys, a bucket is a queue.



Bucket sort: complexity

Counting sor

Radix sort

Bucket sort

ower bound

■ Bucket initialization: O(m)

■ From array to buckets: O(n)

■ From buckets to array: O(m+n)

■ Total cost is O(n+m)

When m is small compared to n, Bucket sort is O(n)

Bucket sort: extensions

Counting sor

Radix sort

Bucket sort

- In the algorithm each bucket contains elements with the same key.
- The algorithm can be implemented in such a way that buckets hold elements with different keys.
- In such a case we have to take care of the additional cost of sorting the elements in each bucket.
- A typical implementation divides the range of values, form lowest to highest key value, into n equal sized ranks. In the worst case the algorithm has cost $O(n \lg n)$ and average cost O(n).

A bit of history

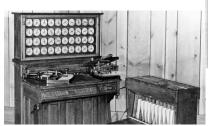
Radix sort

Bucket sort

Radix and counting sort ideas are due to Herman Hollerith.

In 1890 he invented the card sorter that, for ex., allowed to process the US census in 5 weeks, using punching cards.







Counting sort Radix sort Bucket sort

Lower bounds

A problem has a time upper bound T(n) if there is an algorithm A such that, for any input x of size n, A(x) gives the correct answer in $\leq T(n)$ steps.

Counting sort Radix sort Bucket sort

- A problem has a time upper bound T(n) if there is an algorithm \mathcal{A} such that, for any input x of size n, A(x) gives the correct answer in $\leq T(n)$ steps.
- A problem has a time lower bound L(n) if there is NO algorithm which solves the problem in time < L(n), for any input e of size n.

Counting sort
Radix sort
Bucket sort
Lower bounds

- A problem has a time upper bound T(n) if there is an algorithm \mathcal{A} such that, for any input x of size n, A(x) gives the correct answer in $\leq T(n)$ steps.
- A problem has a time lower bound L(n) if there is NO algorithm which solves the problem in time < L(n), for any input e of size n.
- Lower bounds are hard to prove, as we have to consider every possible algorithm.

Counting sor Radix sort Bucket sort

- Upper bound: $\exists A, \forall x \ t_A(x) \leq T(|x|)$,
- Lower bound: $\forall A, \exists x \ t_A(x) \geq L(|x|)$,

Counting sort Radix sort Bucket sort

Lower bounds

- Upper bound: $\exists A, \forall x \ t_A(x) \leq T(|x|)$,
- Lower bound: $\forall A, \exists x \ t_A(x) \geq L(|x|)$,

To prove an upper bound: produce an A so that the bound holds for any input x (n = |x|).

Counting sort Radix sort Bucket sort

Lower bounds

- Upper bound: $\exists A, \forall x \ t_A(x) \leq T(|x|)$,
- Lower bound: $\forall A, \exists x \ t_A(x) \geq L(|x|)$,

To prove an upper bound: produce an A so that the bound holds for any input x (n = |x|).

To prove a lower bound, show that for any possible algorithm, the time on one input is greater than or equal to the lower bound.

Lower bound for **comparison based** sorting algorithm.

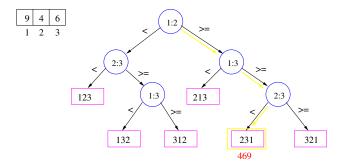
Radix sort
Bucket sort
Lower bounds

To prove the lower bound, we consider binary decision trees a way to represent the comparisons made by a sorting algorithm to distinguish the possible inputs of size n.

- each leaf represents one of the n! possible permutations $(a_{\pi(1)}, a_{\pi(2)}, \dots, a_{\pi(n)})$. The tree has exactly n! leaves.
- each internal node is labeled by a comparison $a_i : a_j$, the leaves in the left subtree verify $a_i < a_j$ and the ones in the right subtree verify $a_i \ge a_j$.

An example of binary decision tree for n = 3

Radix sort
Bucket sort
Lower bounds



Theorem

For any comparison sort algorithm that sorts n elements, there is an input in which it has to perform $\Omega(n \lg n)$ comparisons.

Counting sort

Theorem

For any comparison sort algorithm that sorts n elements, there is an input in which it has to perform $\Omega(n \lg n)$ comparisons.

Proof.

Equivalent to prove: Any decision tree that sorts n elements must have height $\Omega(n \lg n)$.

Bucket sort

Lower bounds

Theorem

For any comparison sort algorithm that sorts n elements, there is an input in which it has to perform $\Omega(n \lg n)$ comparisons.

Proof.

- Equivalent to prove: Any decision tree that sorts n elements must have height $\Omega(n \lg n)$.
- Let *h* the height of a decision tree with *n*! leaves,

$$n! \le 2^h \Rightarrow h \ge \lg(n!) > \lg(\frac{n}{e})^n = \Omega(n \lg n).$$