# ساختمان داده و الگوريتم ها (CE203)

جلسه یازدهم: مرتب سازی خطی

سجاد شیرعلی شهرضا پاییز 1400 شنبه، 8 آبان 1400

## اطلاع رساىي

- بخش مرتبط کتاب برای این جلسه: 8.1
  امتحانک دوم:
  دوشنبه این هفته (پس فردا)، 10 آبان 1400
  در طی ساعت کلاس به صورت برخط (مشابه امتحانک اول)

مرتب سازی خطی

الگوریتم های مرتب سازی که بر مبنای مقایسه نیستند!

#### A NEW MODEL OF COMPUTATION

The elements we're working with have meaningful values.

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#### Now (examples):

مرتب سازی شمارشی

#### We assume that there are only k different possible values in the array (and we know these k values in advance)

For example: elements are integers in {10, 20, 30, 40, 50, 60}

Input:

30 50 20 30 10 60 50 20

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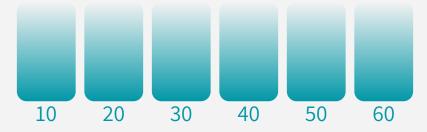
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20

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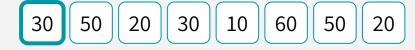
**Buckets:** 



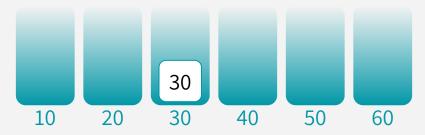
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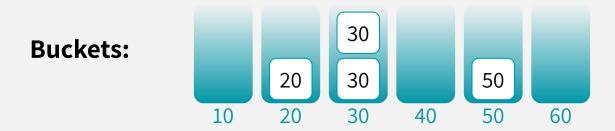
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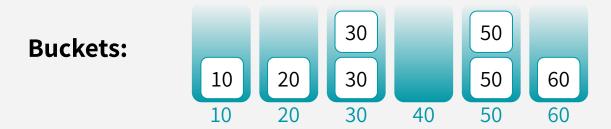
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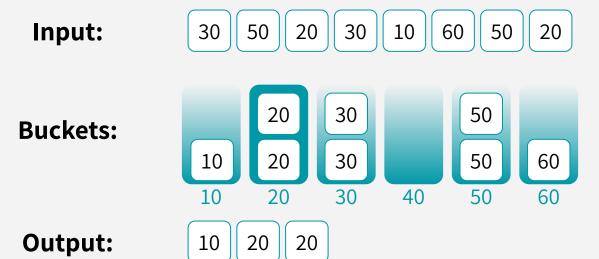
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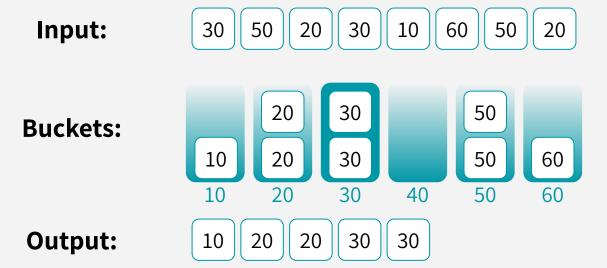
Buckets: 20 30 50 60 10 20 30 40 50 60

Output: 10

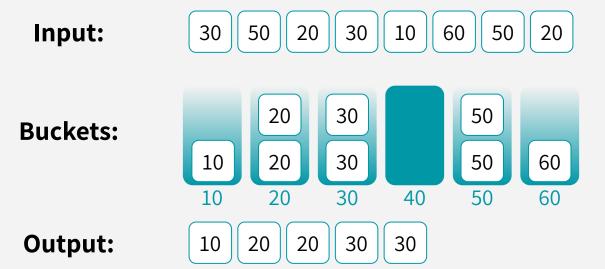
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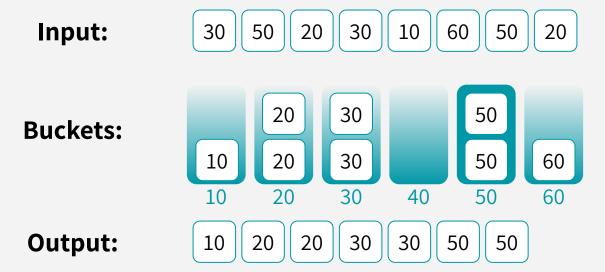
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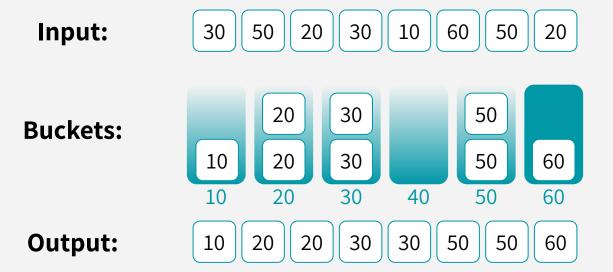
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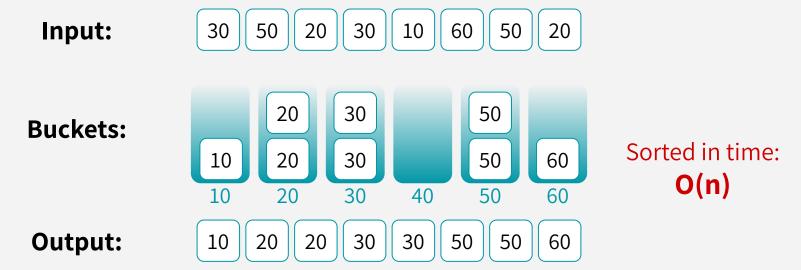
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#### **Assumptions:**

We are able to know what bucket to put something in.

We know what values might show up ahead of time.

There aren't too many such values.

If there are too many possible values that could show up, then we need a bucket per value...

This can easily amount to a lot of space.



# مرتب سازی مبنایی

الگوریتم مرتب سازی برای اعداد صحیح کوچکتر از M (و یا در حالت کلی تر، برای مرتب سازی رشته ها)

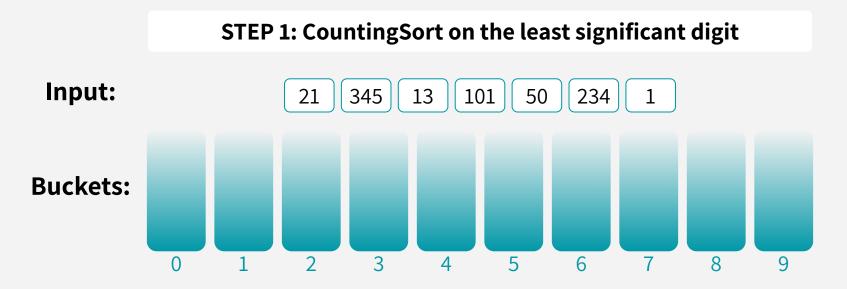
For sorting integers where the maximum value of any integer is M. (This can be generalized to lexicographically sorting strings as well)

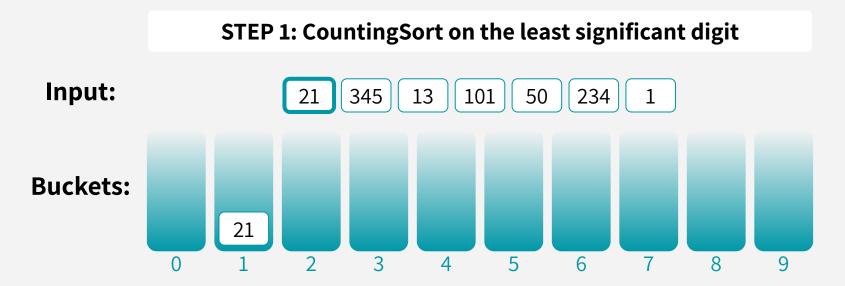
#### **IDEA:**

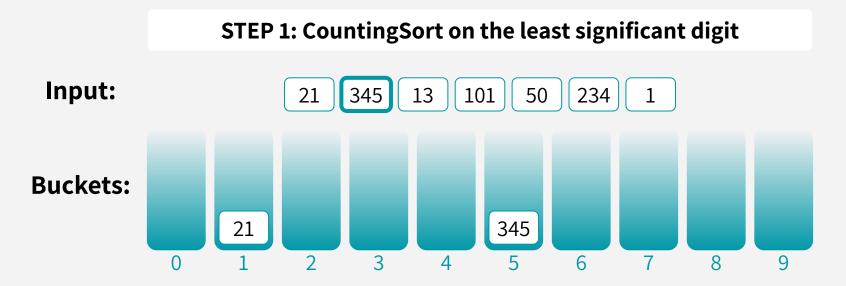
Perform CountingSort on the least-significant digit first, then perform CountingSort on the next least-significant, and so on...

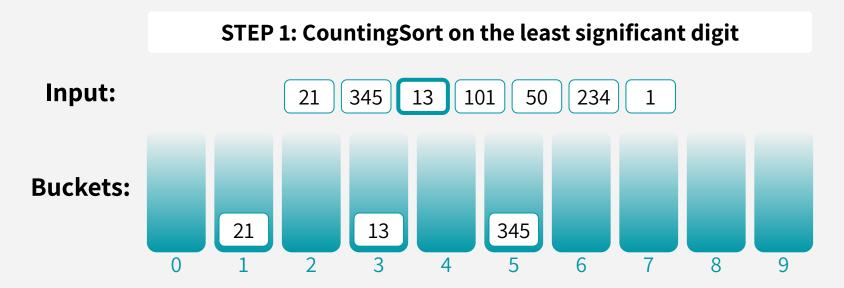
Instead of a bucket per possible value, we just need to maintain a bucket per possible value that a single digit (or character) can take on!

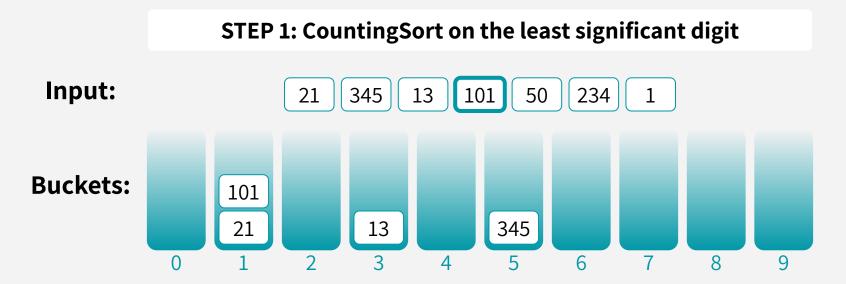
e.g. 10 buckets labeled 0, 1, ..., 9

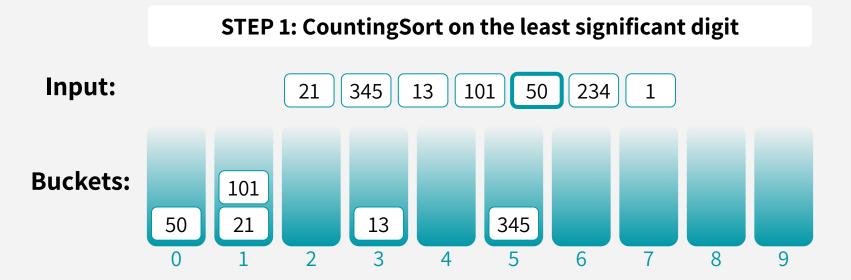


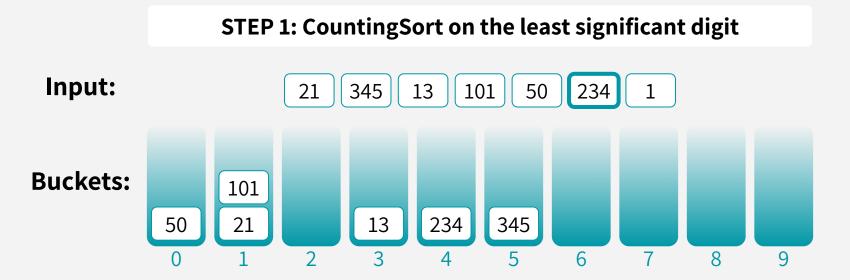


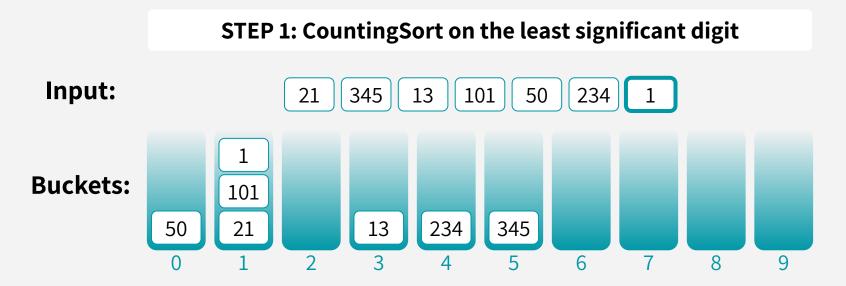


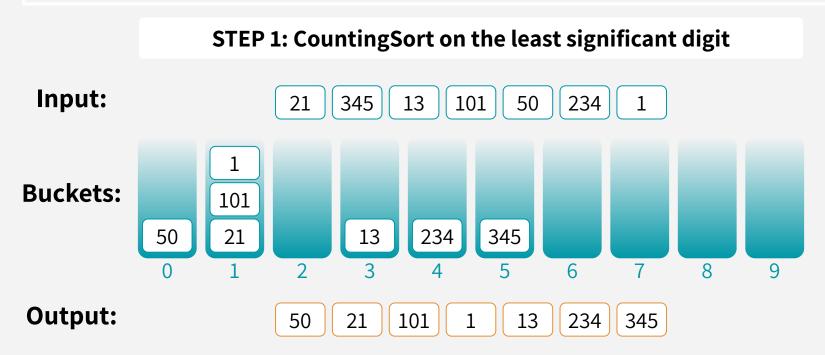












When creating the output list, make sure bucket items exit in FIFO order (i.e. use a *stable* implementation of CountingSort, where buckets are FIFO queues)

# QUICK ASIDE: STABLE SORTING

We say a sorting algorithm is STABLE if two objects with equal values appear in the same order in the sorted output as they appear in the input.

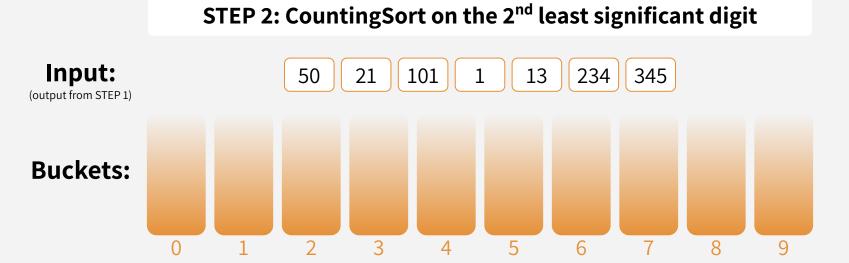
Input:

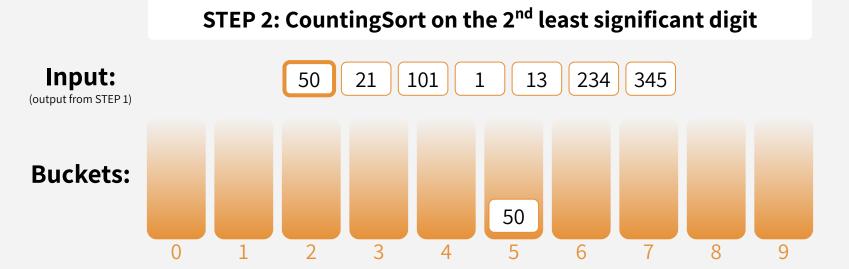
1 2 1 3 2

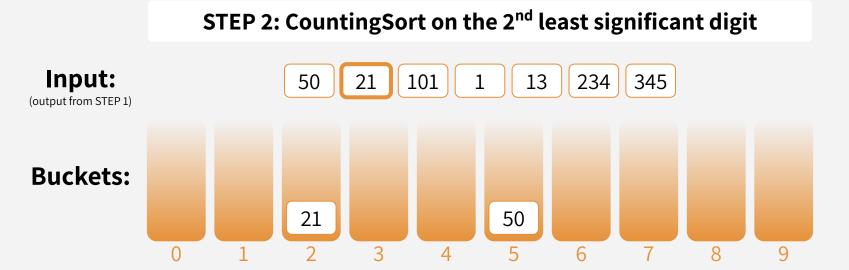
Sorted Output: (if algorithm is stable)

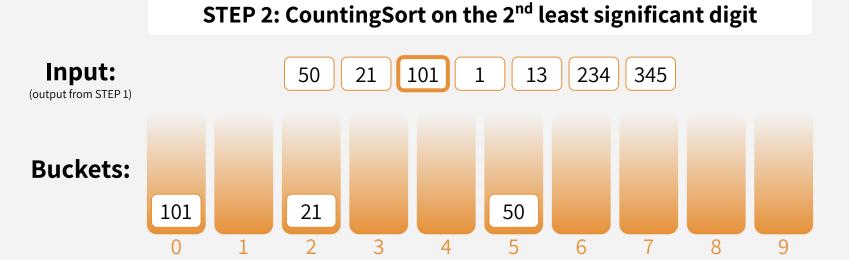
The red 1 appeared before the green 1 in the input, so they have to also appear in this order in the output!

The yellow 2 appeared before the purple 2 in the input, so they have to also appear in this order in the output!

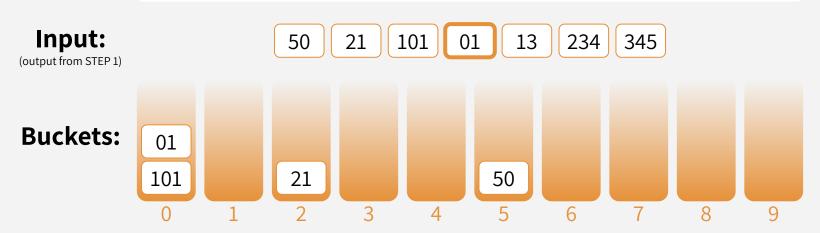




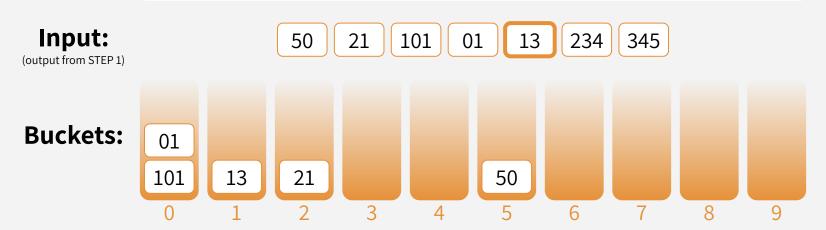




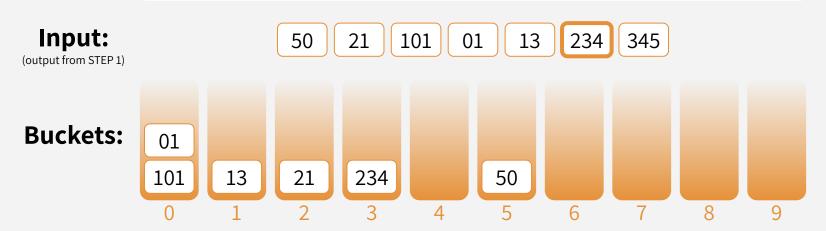




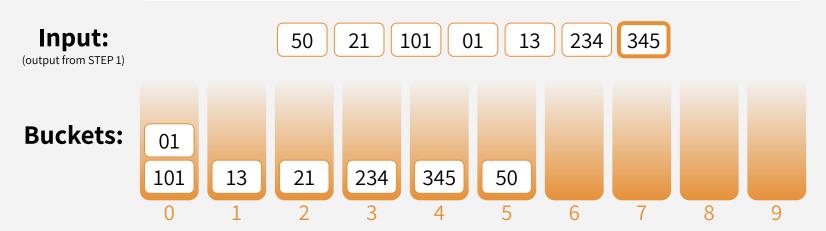




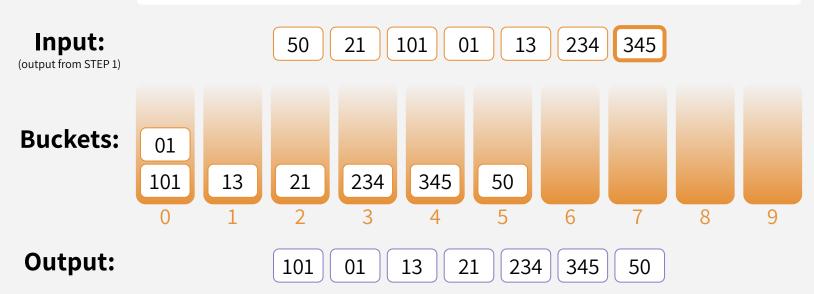






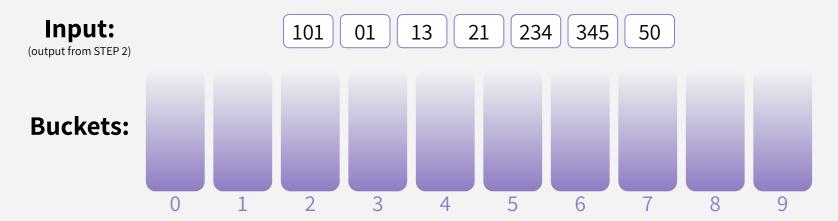




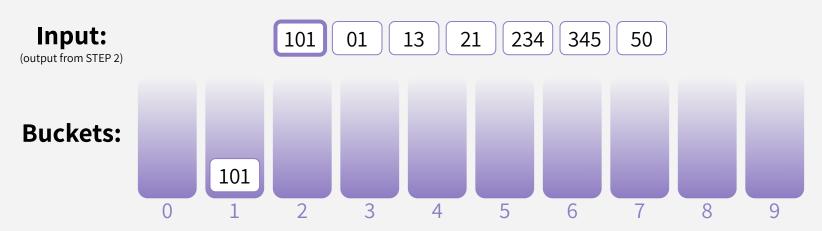


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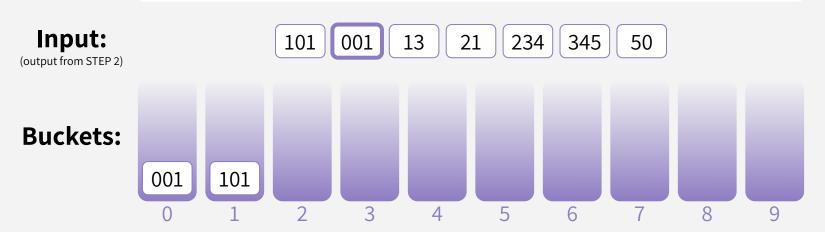




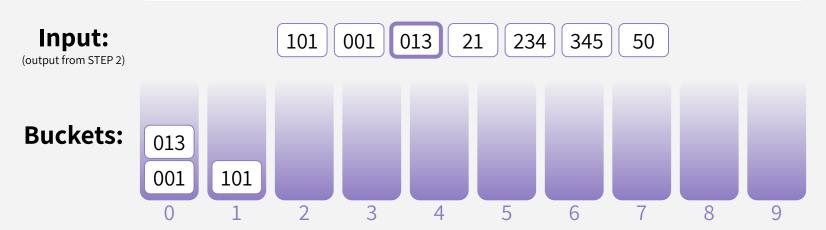


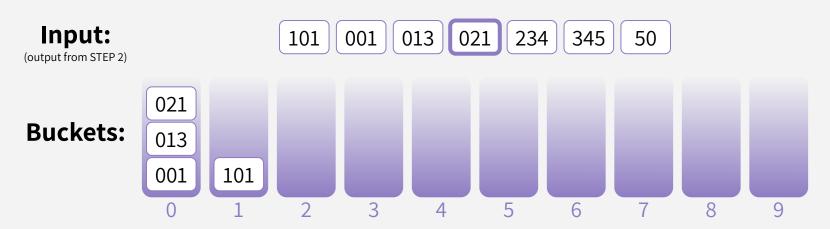


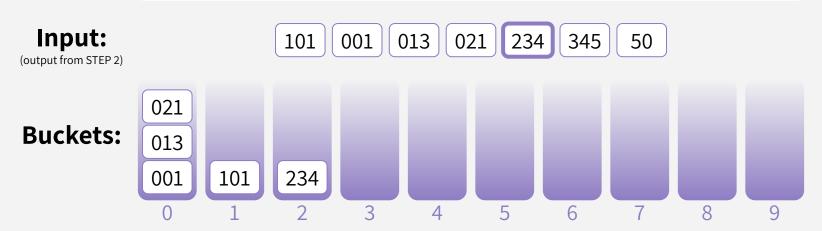


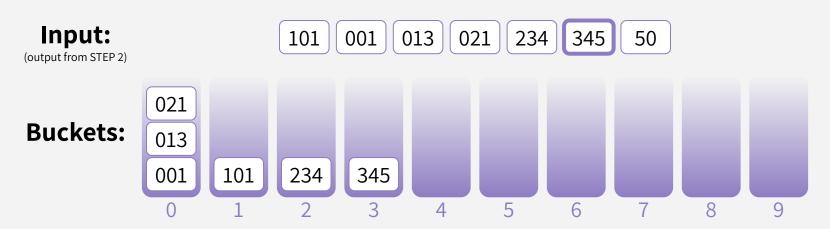


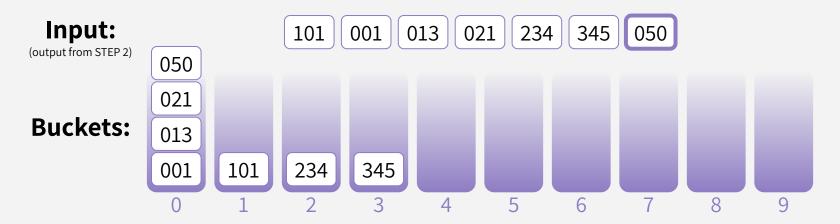




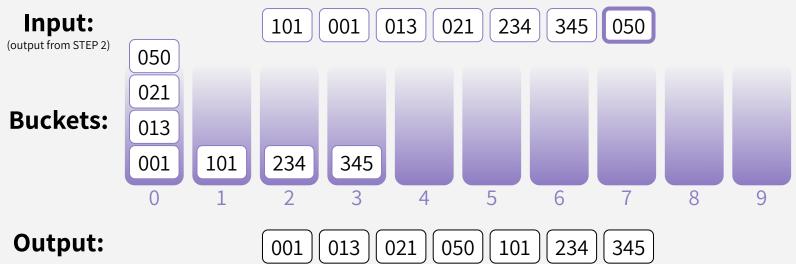












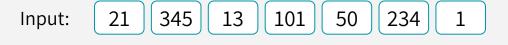
It worked! But why does it work???



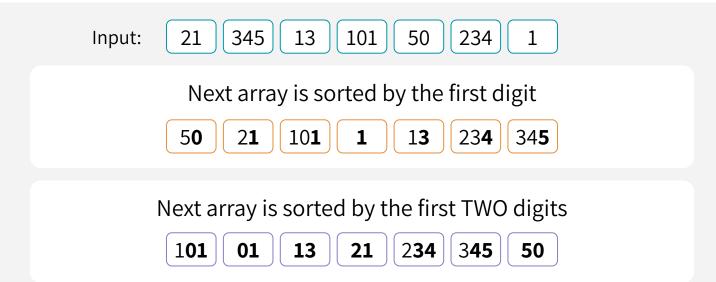
# درستی مرتب سازی مبنایی

چرا اعداد مرتب شدند؟

Input: 21 345 13 101 50 234 1



Next array is sorted by the first digit





Next array is sorted by the first digit

50 
$$(21)(101)(1)(13)(234)(345)$$

Next array is sorted by the first TWO digits

Next array is sorted by the first THREE digits (aka fully sorted)

$$igg(001igg)igg(013igg)igg(021igg)igg(050igg)igg(101igg)igg(234igg)igg(345igg)$$

#### **Proof by Induction!**

We'll perform induction on the number of iterations, and we'll use weak induction here:

#### ITERATIVE ALGORITHMS

- 1. **Inductive hypothesis**: some state/condition will always hold throughout your algorithm by any iteration **i**
- 2. **Base case**: show IH holds for iteration 0 (i.e. start of algorithm)
- 3. **Inductive step**: Assume IH holds for  $k \Rightarrow \text{prove } k+1$
- 4. **Conclusion**: IH holds for i = # total iterations  $\Rightarrow$  yay!

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#### **INDUCTIVE STEP** (weak induction)

Let k be an integer, where  $0 < k \le d$  (d is the number of digits). Assume that the IH holds for i = k-1, so the array is already sorted by the first k-1 least-significant digits. We need to show that after the k-th iteration, the array is sorted by the first k least-sig. digits.

At a high level, since the "buckets as FIFO-queue" implementation of CountingSort is *stable*, elements that get placed in the same bucket during this k-th round of CountingSort still maintain their previous relative ordering, so they are *still* in order of their k-1 least-sig. digits. Since this k-th round CountingSort sorts A by the k-th digit of the elements, this ultimately means that the elements are going to be sorted by their k least-significant digits.

This can be made

more rigorous!

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#### CONCLUSION

By induction, we conclude that the IH holds for all  $0 \le i \le d$ . In particular, it holds for i = d, so after the last iteration, the array is sorted by all the digits. Hence, it is sorted!

more rigorous!



# زمان اجرای مرتب سازی مبنایی

چقدر طول می کشد؟

### RADIX SORT RUNTIME

Suppose we are sorting **n** (up-to-)**d**-digit numbers in base 10 (e.g. n = 7, d = 3):

21 345 13 101 50 234 1

How many iterations are there?

How long does each iteration take?

What is the total running time?

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```
For example, if M = 1234:

log_{10} 1234J + 1

= 3 + 1 = 4
```

```
How many iterations are there?
d = Llog<sub>10</sub> MJ + 1 iterations
```

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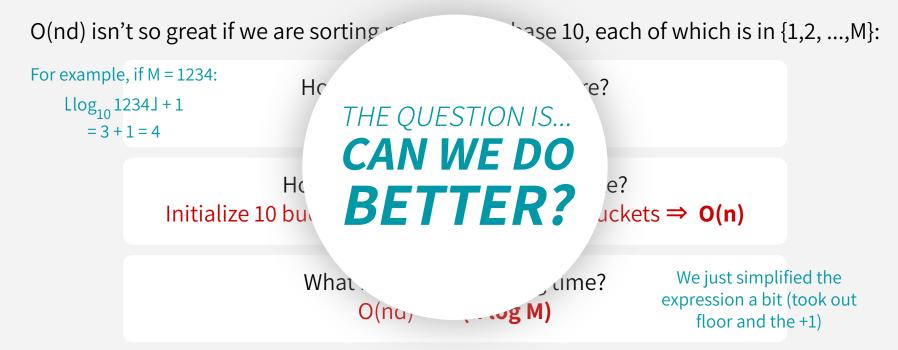
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 iterations

How long does each iteration take?

Initialize r buckets + put n numbers in r buckets  $\Rightarrow$  O(n + r)

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Bigger base r ⇒ fewer iterations, but more buckets to initialize!

A reasonable sweet spot: **let** r = n

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What is the total running time?

$$O(d \cdot n) = O((Llog_n MJ + 1) \cdot n)$$

This term is a constant!

If  $M \le n^C$  for some constant c, then  $O((\lfloor \log_n M \rfloor + 1) \cdot n) = O(n)$ 

A reasonable sweet spot: **let** r = n

This means that the running time of RadixSort using a base of  $\mathbf{r} = \mathbf{n}$  (instead of base 10 from earlier examples) depends on how big M is in terms of n. The formula is:

O( 
$$(L\log_n M \rfloor + 1) \cdot n$$
)

This is O(n) when  $M \le n^{C}$ .

The number of buckets need is r = n.

If  $M \le n^C$  for some constant c, then  $O((\lfloor \log_n M \rfloor + 1) \cdot n) = O(n)$ 

## RADIX SORT RECAP

Radix Sort can sort **n integers of size at most n^{100}** (or  $n^{C}$  for any constant c) in time **O(n)**.

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It matters how you pick the base! In general, if you have **n** elements, **M** = max size of any element, and **r** is the base:

Runtime of Radix Sort =  $O((Llog_r MJ + 1) \cdot n)$ 

#### WHY BOTHER WITH COMPARISON-BASED SORTING?

Comparison-based sorting algorithms can handle arbitrary comparable elements! And with numbers, it can handle sorting with high precision & arbitrarily large values:



Radix Sort requires us to look at all digits, which is problematic —  $\pi$  and e both have infinitely many! And n<sup>n</sup> is big enough to make Radix Sort slow...

Radix Sort is also not in place (you need those buckets!), so it could require more space.

