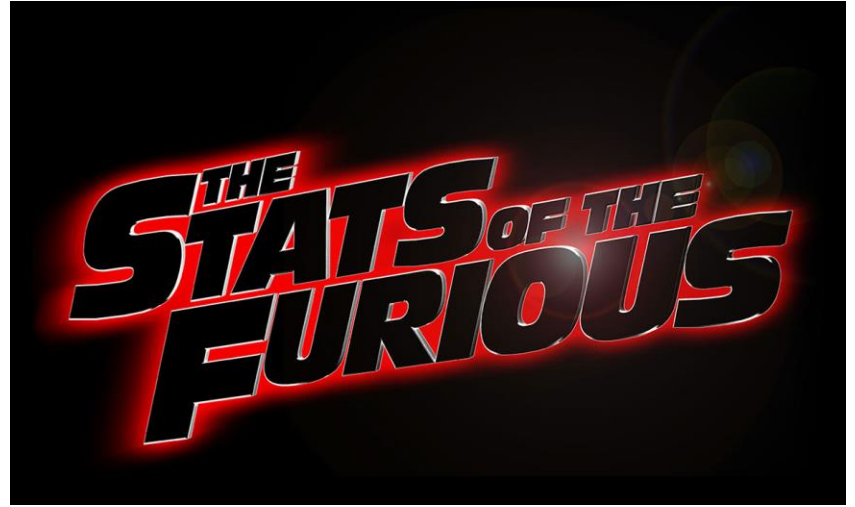


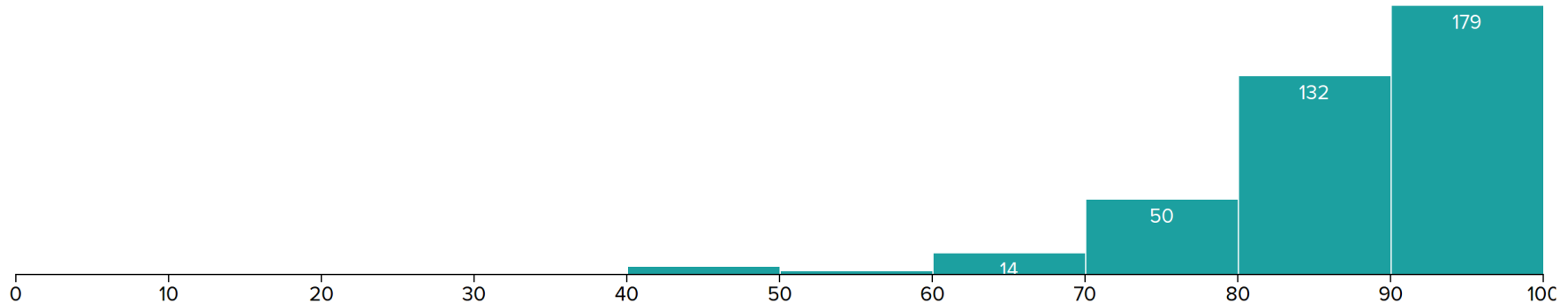
Lecture 6: How Fast Can We Sort?



<https://www.bloomberg.com/graphics/2017-fast-and-furious/>

Announcements

- Exam 1 graded



MINIMUM

45.0

MEDIAN

89.0

MAXIMUM

100.0

MEAN

86.62

STD DEV

10.06

Announcements

- Exam 1 graded
 - Regrade requests due by 3/3
- Lab 6
 - Out Wednesday, due 3/8
 - Practice with recurrences, sorting, searching
 - Will not have a coding portion or pre-lab (wait for Lab 7)

What Do We Know About Sorting?

- We know a couple of worst-case $\Theta(n \log n)$ algorithms
- HeapSort
 - Insert all inputs into a heap
 - Extract in sorted order
 - Lab 3 unit test did this
- MergeSort (Thursday's studio!)
 - Based on linear merge of two sorted arrays
 - Divide-and-conquer algorithm

What Other Sorting Algorithms Exist?

- BubbleSort – $\Theta(n^2)$
- InsertionSort – $\Theta(n^2)$
- ShellSort – $\Theta(n^2)$, or $\Theta(n^{4/3})$, or $\Theta(n \log^2 n)$, or ... [many different variants]
- QuickSort – $\Theta(n \log n)$ [*if we work at it; see 347*]
- ...
- See ["The Sounds of Sorting" website](#) for audio/visual intuition

How Fast *Can We Sort*?

- Multiple worst-case $\Theta(n \log n)$ time algorithms
- All the others we listed are slower!
- Is there a faster sorting algorithm?

To answer, we need to be more precise about what “sorting algorithm” means...

What is a Sorting Algorithm Allowed to Do?

- Computers are not infinitely powerful...
- They can do only **limited work in constant time.**
- In particular, they can make **limited decisions about their inputs** in constant time.

What is a Sorting Algorithm Allowed to Do?

- Computers are not infinitely powerful

- They can't do things in

- In particular, they can't

about

“Model of Computation” –
which operations can your
computer do in constant
time?

time.

ons

Limited Decisions for Sorting

- All the sorting algorithms we listed work on any **Comparable** data type.
- The only way they inspect the input is by comparing pairs of elements to each other!
- **Can answer “Is $x > y$?” in constant time.**

Limited Decisions for Sorting

- All the
Com
- The
elem
- Can

Any sorting algorithm that inspects its input only via pairwise comparisons is called a “*comparison sort*.”

g two

An Aside on Comparisons

- If we can test “ $x > y$ ” ...
- We can also test “ $x \leq y$ ” (NOT $x > y$)
- Hence, we can test “ $x = y$ ” ($x \leq y$ AND $y \leq x$),
“ $x \geq y$ ” ($x = y$ OR $x > y$), and “ $x < y$ ” (NOT $x \geq y$)

We can implement all ordered comparisons in $O(1)$ >'s.

Reformulating the Question

- **How many comparisons** do we need to sort an input array of size n ?
- If each comparison takes **constant time**, and comparison is the **dominant cost** of sorting...
- ...then # of comparisons gives time complexity of sorting.

What We Know

- We know of algorithms that use $\Theta(n \log n)$ comparisons to sort an array of size n .
- Hence, # of required comparisons for *fastest possible algorithm* is $???(n \log n)$

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What We Know

- We know of algorithms that use $\Theta(n \log n)$ comparisons to sort an array of size n .
- Hence, # of required comparisons for *fastest possible algorithm* is $O(n \log n)$
- Any *fixed* sorting algorithm gives **upper bound** on cost of fastest possible algorithm.

What We Want

- Is there an $f(n)$ for which every comparison sort requires $\Omega(f(n))$ comparisons to sort an array of size n ?
- That is, can we find an asymptotic **lower bound** on cost of *any* comparison sort?

A Trivial Lower Bound

- Claim: every comparison sort takes time $\Omega(n)$.

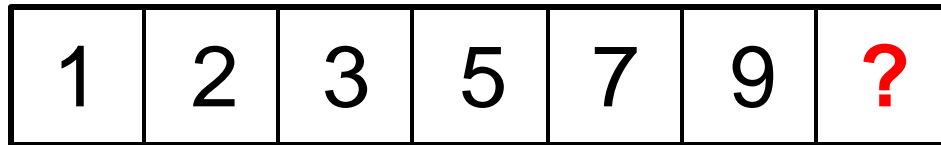
A Trivial Lower Bound

- Claim: every comparison sort takes time $\Omega(n)$.
- Pf: a correct sorting algorithm must inspect every element of its input array *at least once*.

3	1	7	2	5	9	?
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A Trivial Lower Bound

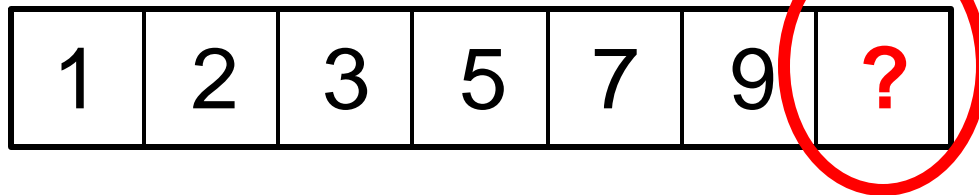
- Claim: every comparison sort takes time $\Omega(n)$.
- Pf: a correct sorting algorithm must inspect every element of its input array *at least once*.



A Trivial Lower Bound

- Claim: every comparison-based sorting algorithm requires $\Omega(n \log n)$ comparisons in the worst case.
- Pf: a correct sorting algorithm must compare every element of its input array at least once.

We have no idea what this value is, so cannot determine correct place for it in order.



A Trivial Lower Bound

- Claim: every comparison sort takes time $\Omega(n)$.
- Pf: a correct sorting algorithm must inspect every element of its input array *at least once*.
- Each comparison inspects only 2 elements, so we need at least ??? comparisons.

A Trivial Lower Bound

- Claim: every comparison sort takes time $\Omega(n)$.
- Pf: a correct sorting algorithm must inspect every element of its input array *at least once*.
- Each comparison inspects only 2 elements, so we need at least $n/2$ comparisons. **QED**

Can We Improve This Lower Bound?

- Yes, but it will take a bit more work.
- Need a way to represent **any** possible comparison sort

Can We Improve This Lower Bound?

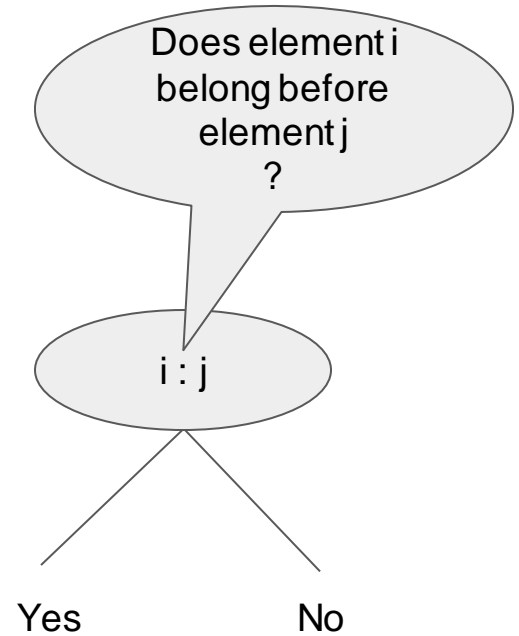
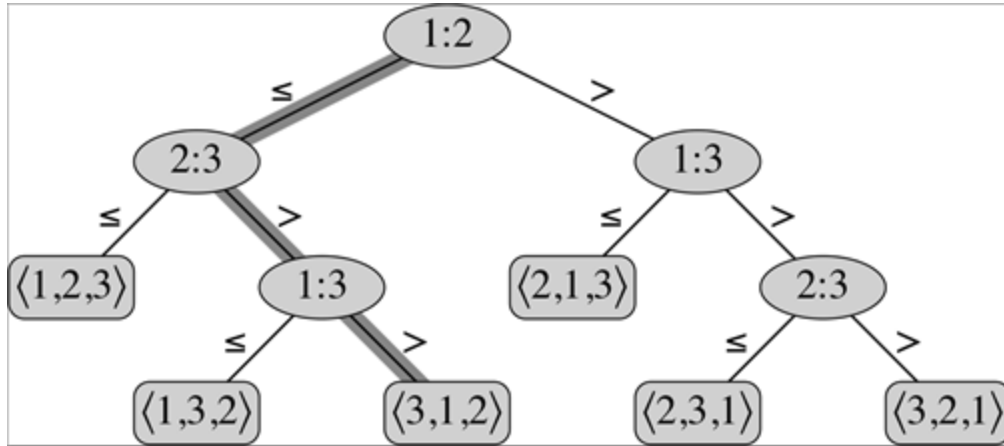
- Yes, but it will take a bit more work.
- Need a way to represent **any** possible comparison sort
- *(Even algorithms we have never imagined!)*
- **Will use properties of representation to prove bound.**

A New Way to Represent Algorithms

- Given an input array of size n
- Any fixed sorting algorithm compares elements according to some logic.
- Choice of later comparisons might depend on results of earlier ones.
- Will use a tree to encode logic of comparison sequence.

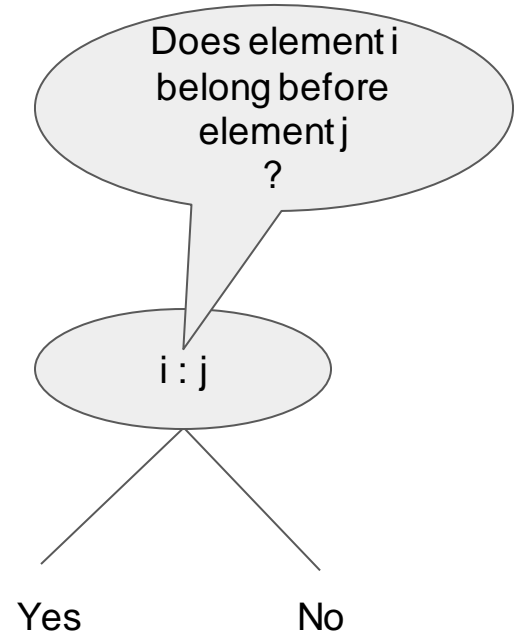
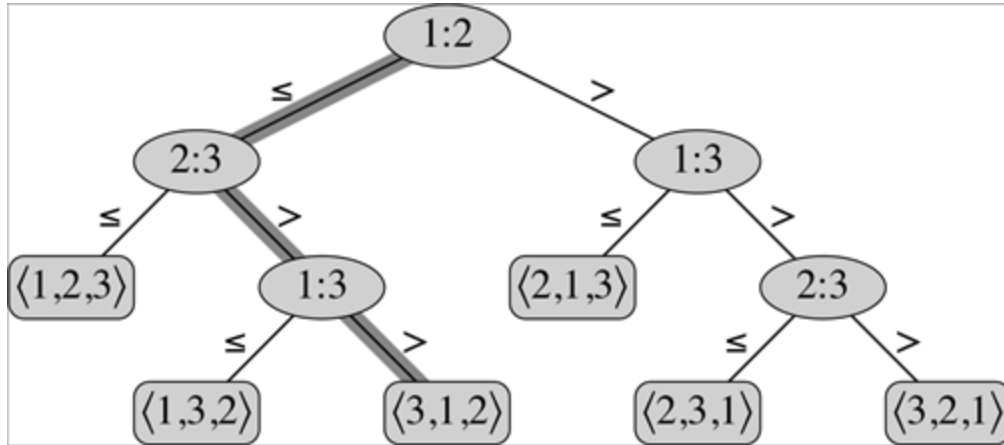
Decision tree for sorting using comparisons

Sorting 3 elements (Figure 8.1 from text)



One Possible decision tree for sorting using comparisons

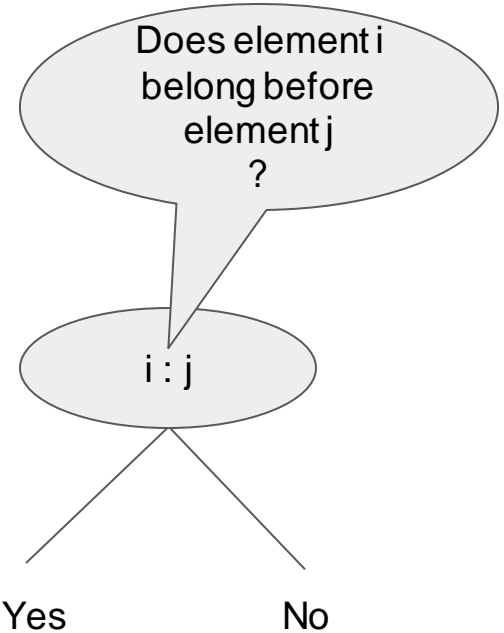
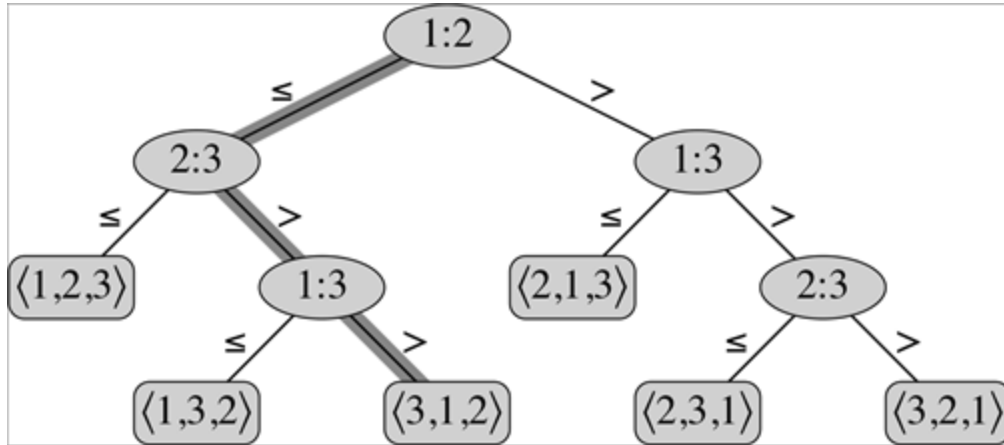
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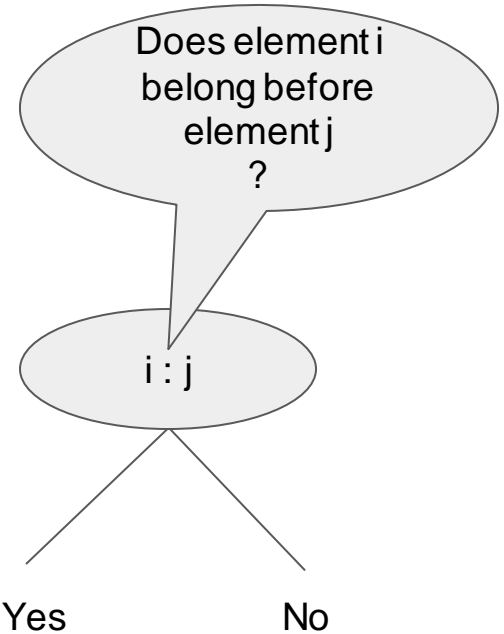
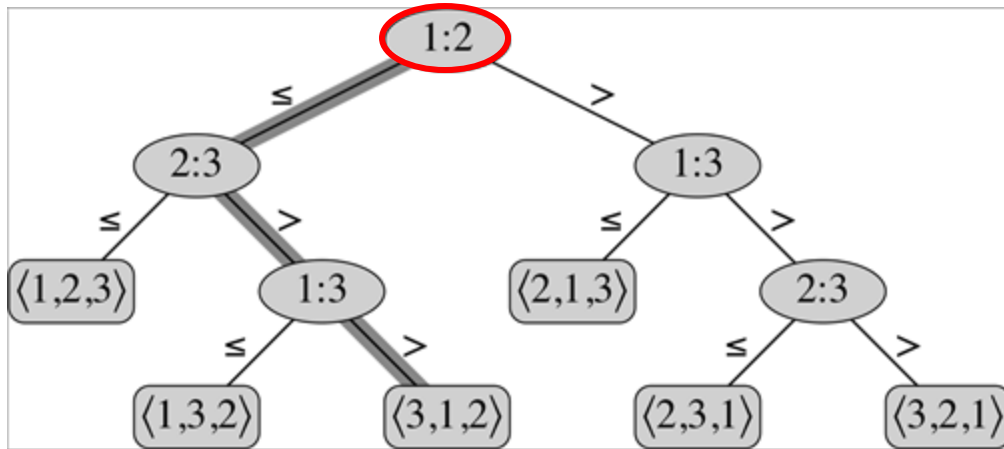
75 99 50



Decision tree for sorting using comparisons

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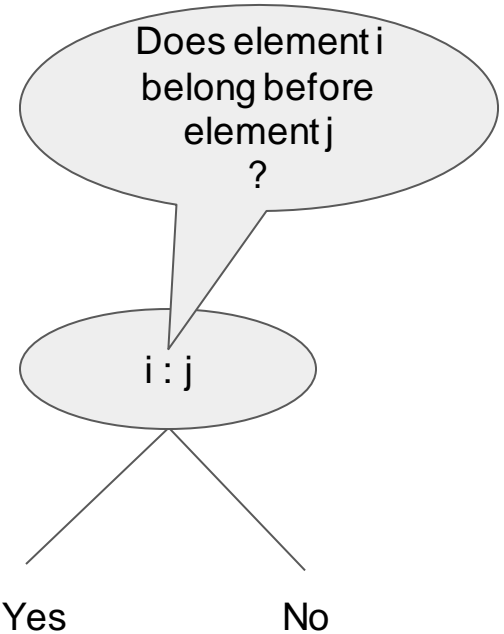
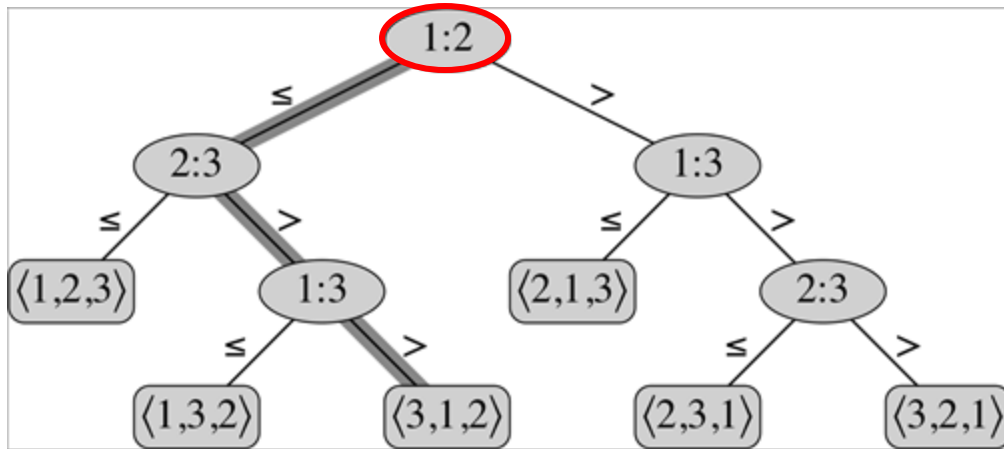
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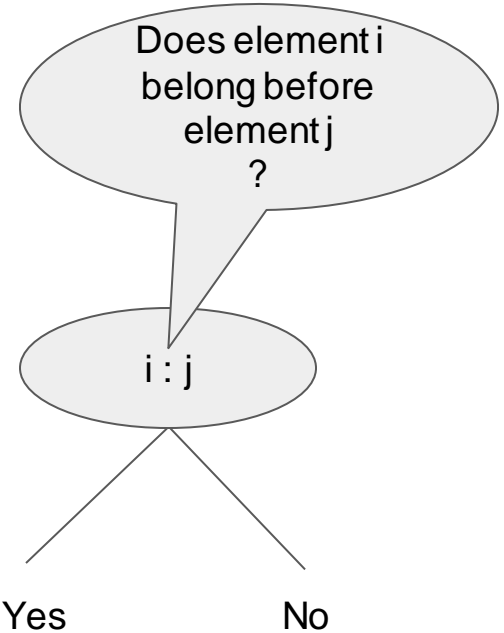
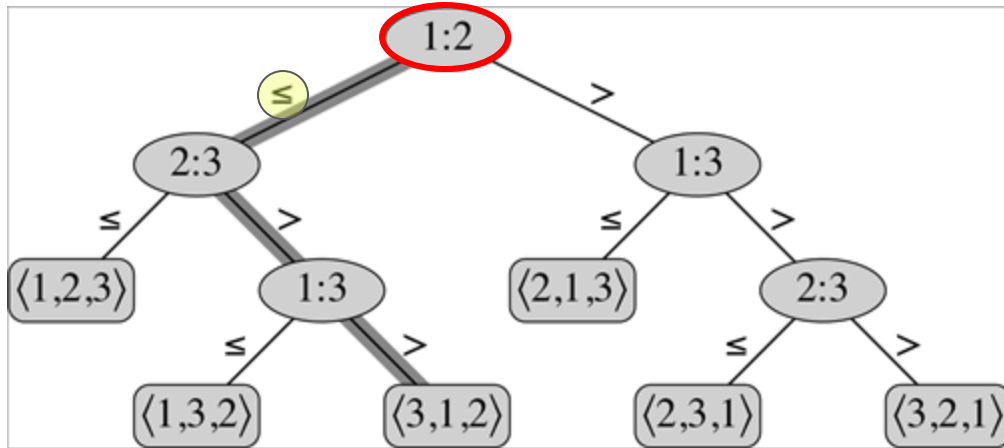
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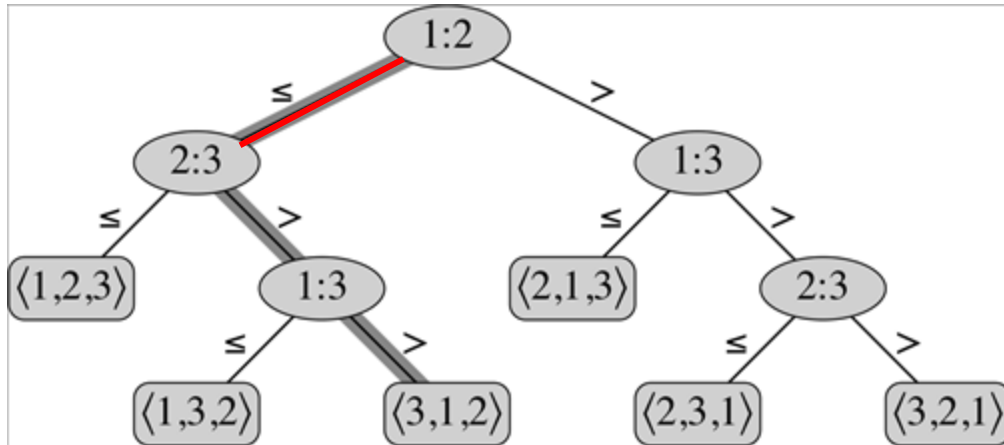
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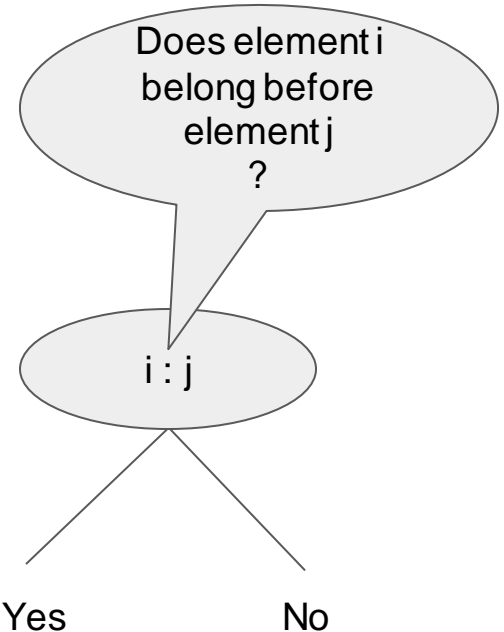
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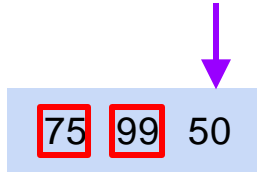
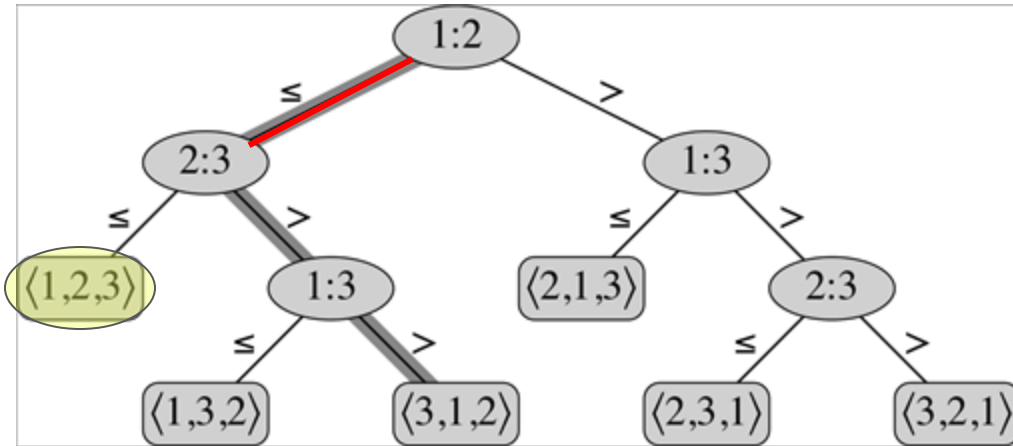


This leaves 3 places 50 could go

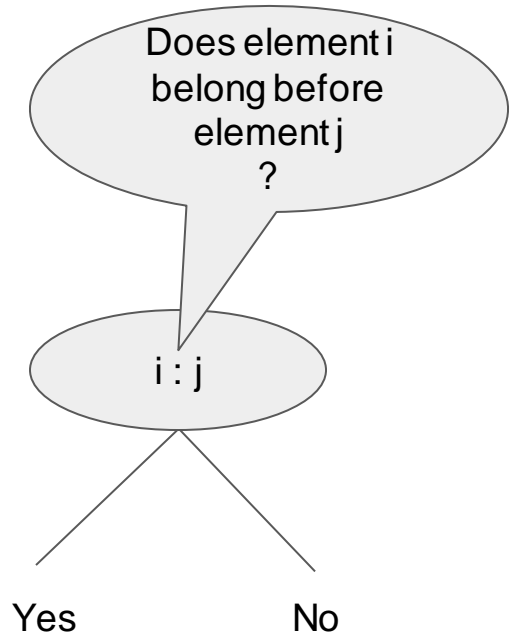


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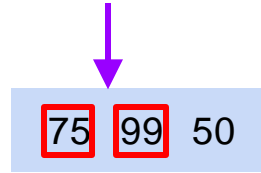
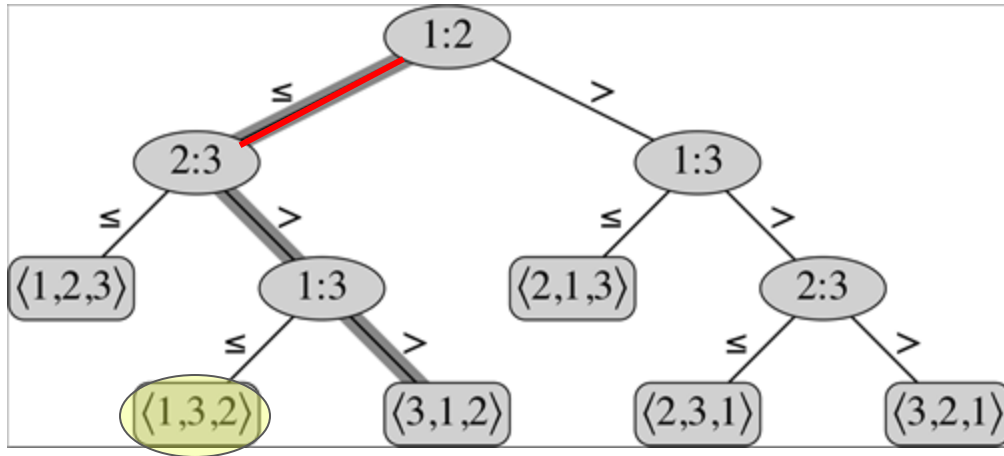


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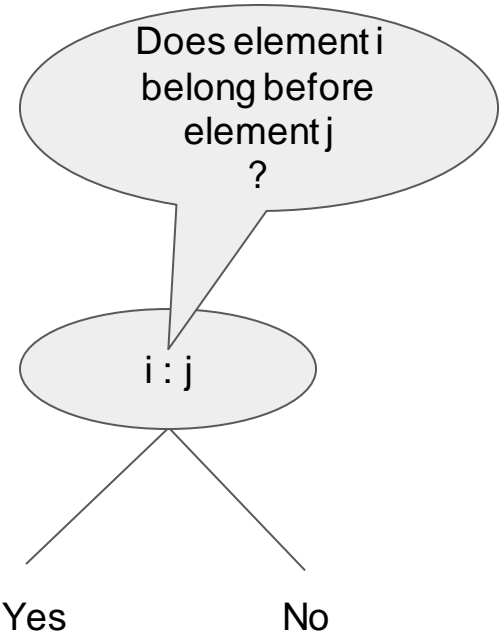


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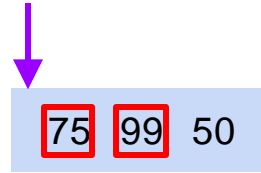
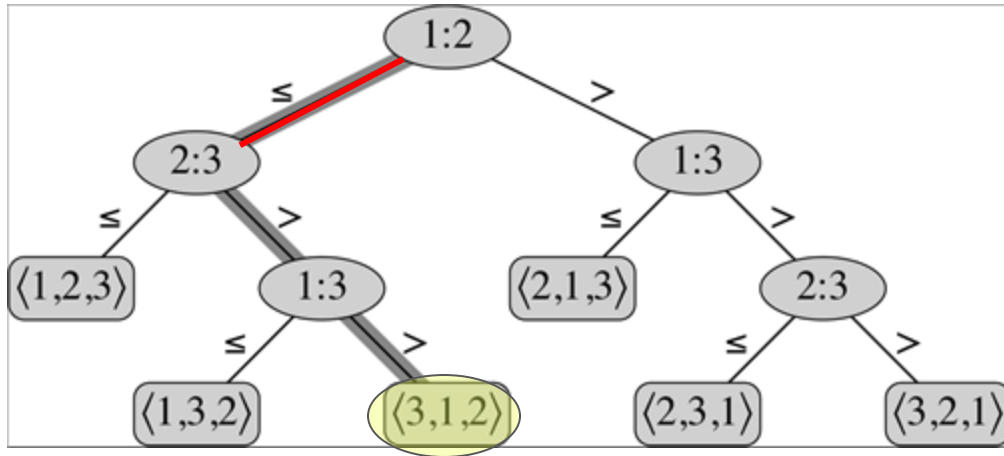


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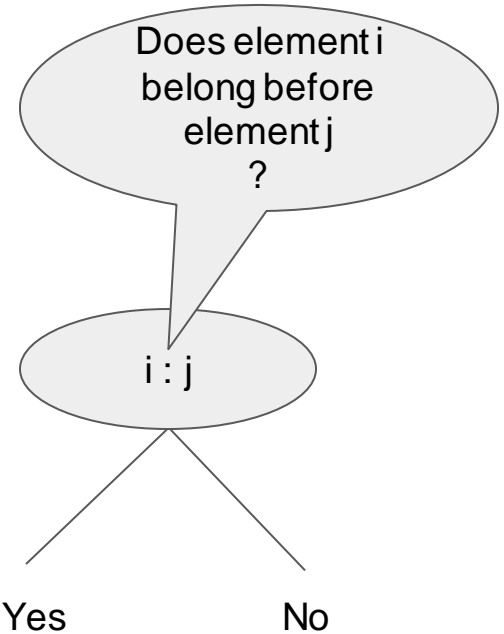


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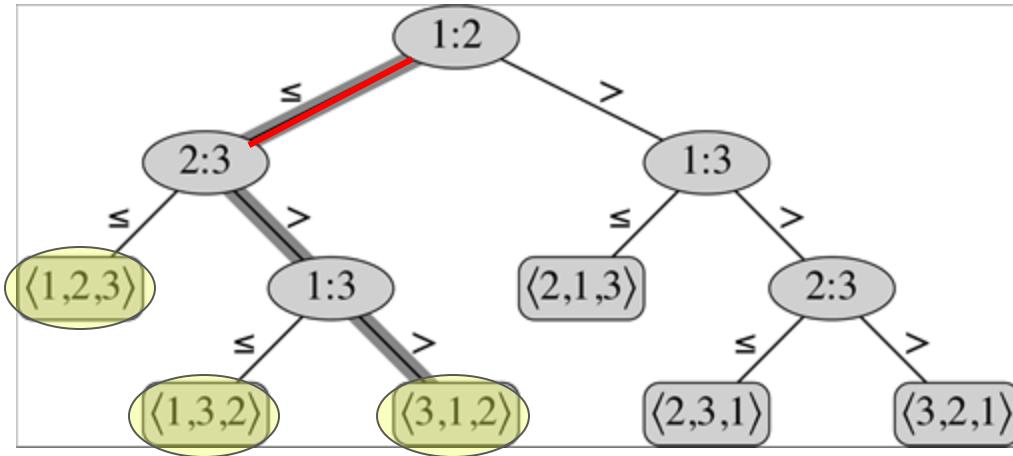
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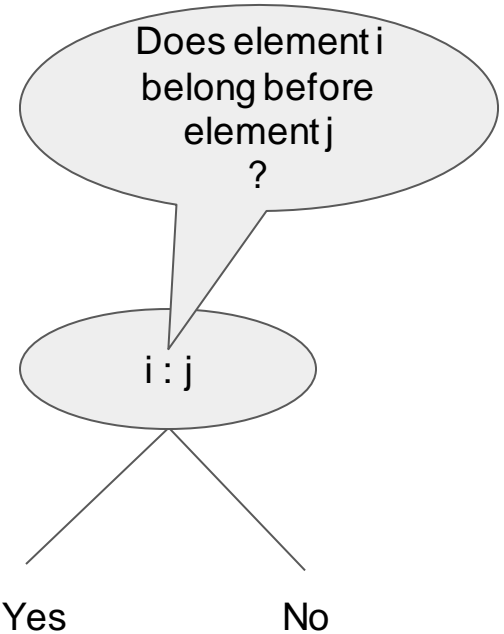
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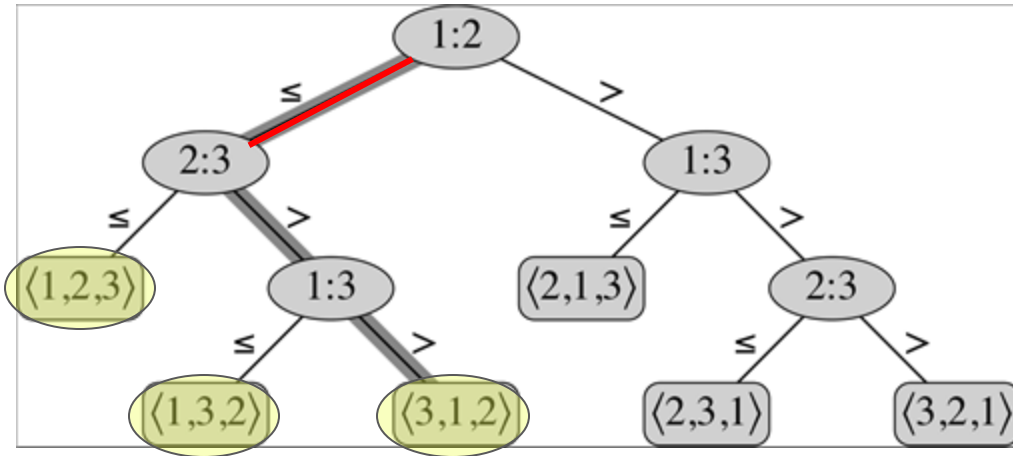
We cannot figure this out with just one comparison



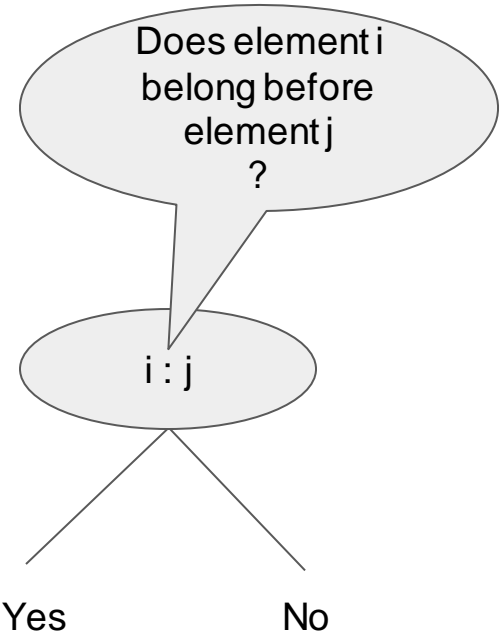
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Try to do it with just one

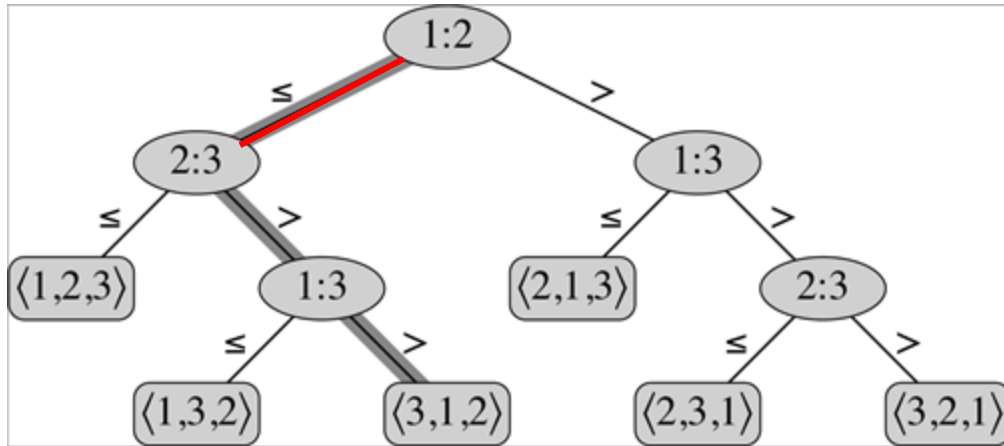


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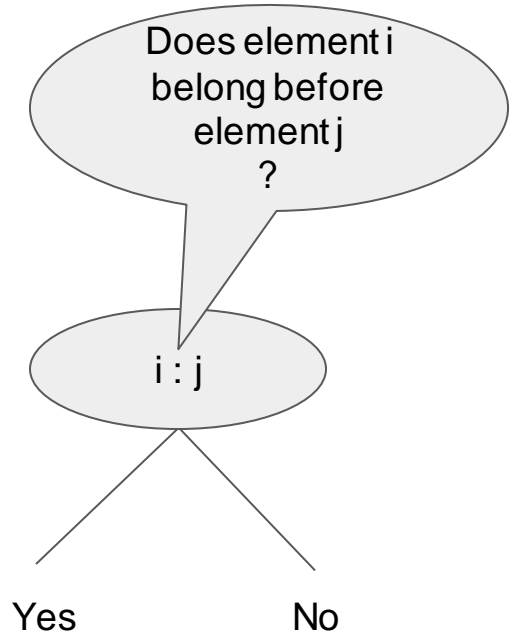
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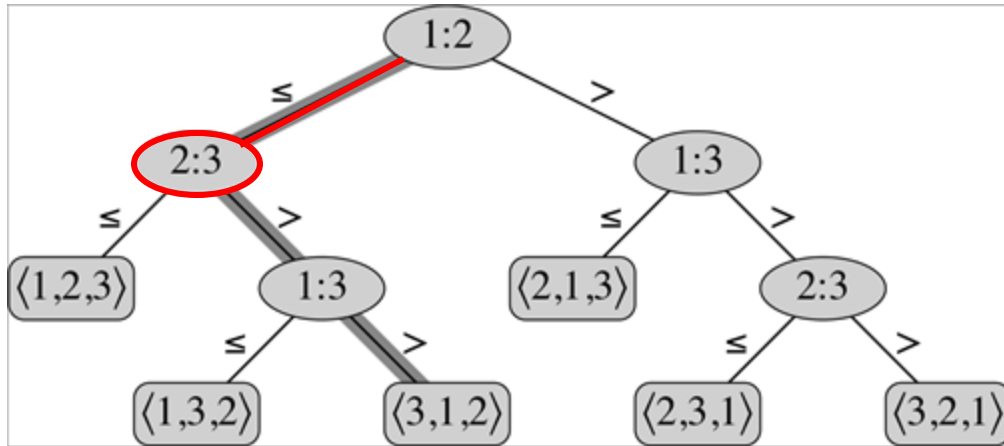


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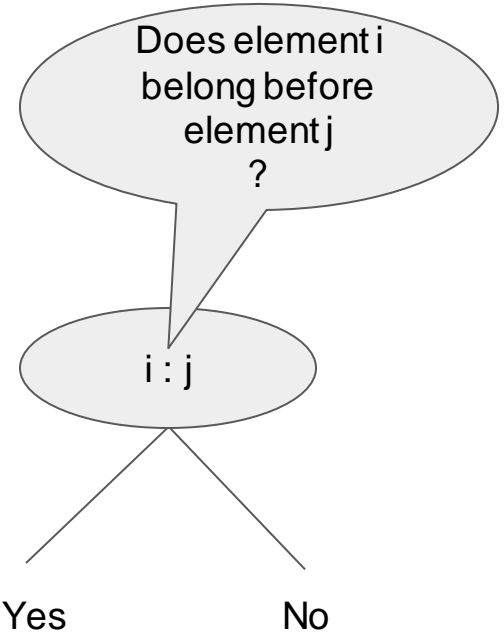
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75 **99** **50**

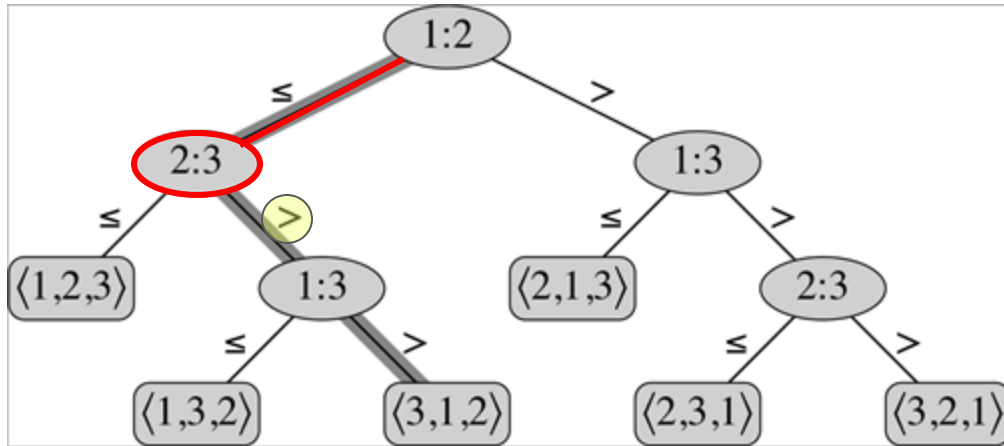
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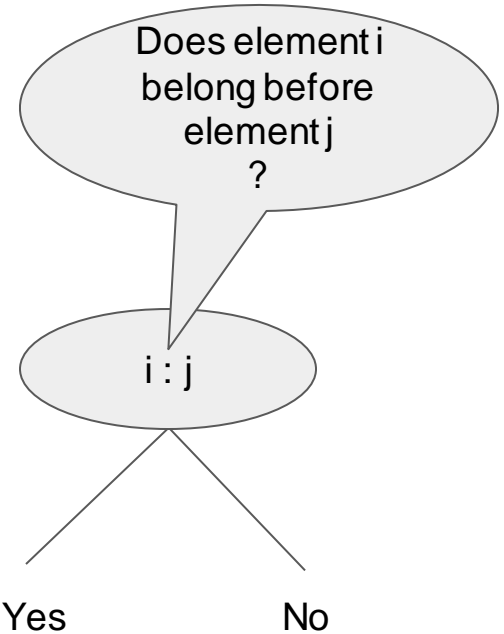
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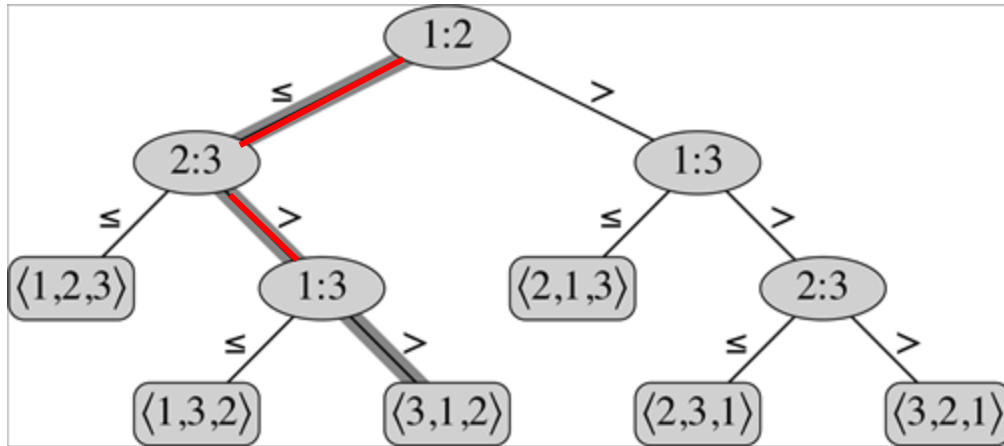
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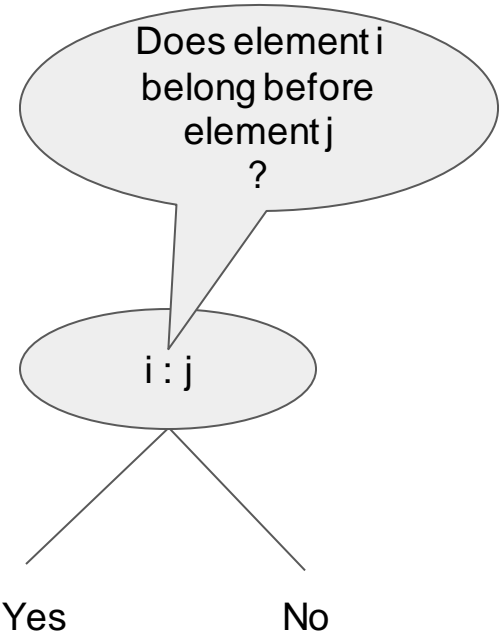
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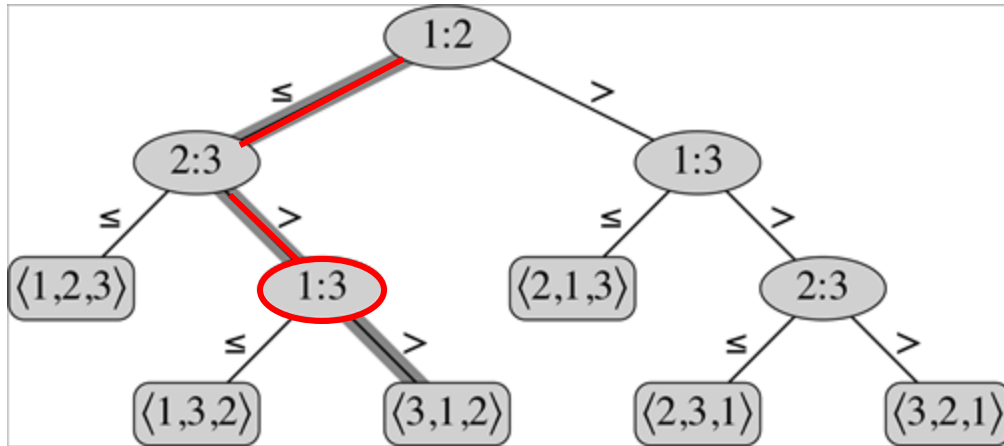
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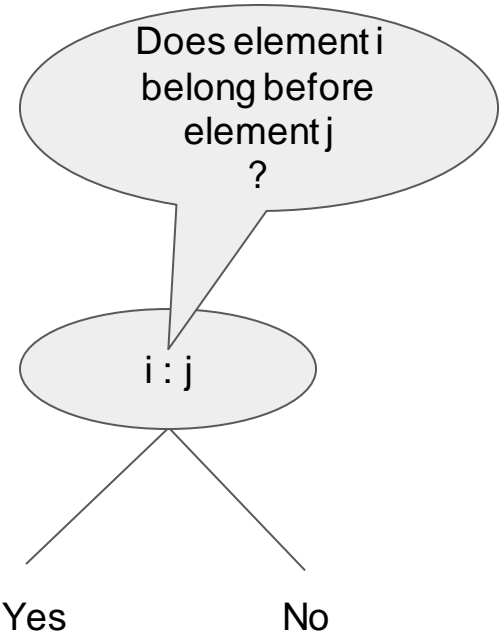
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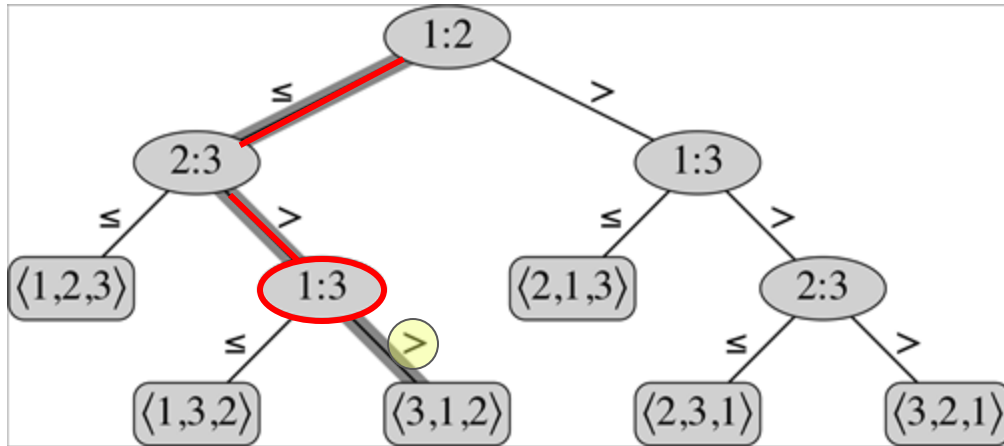
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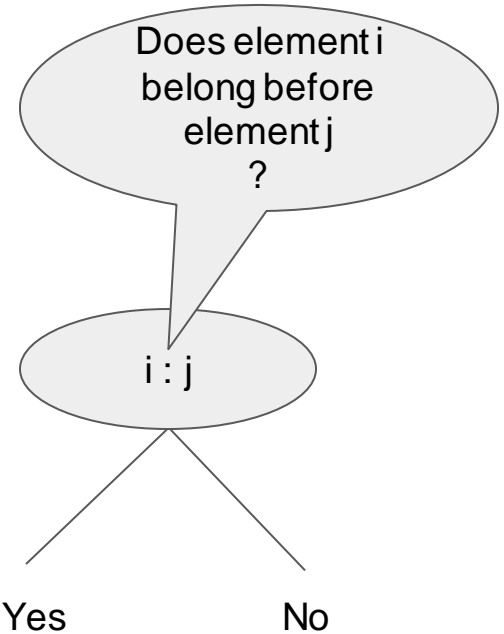
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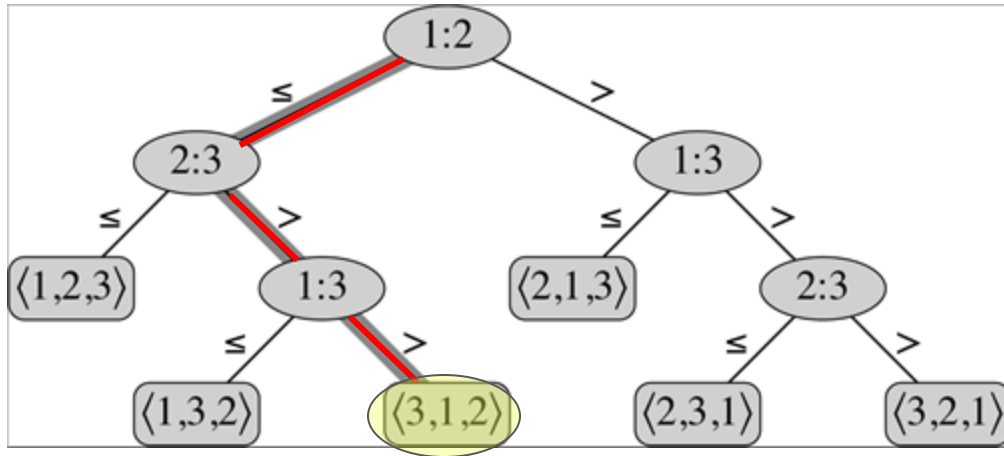
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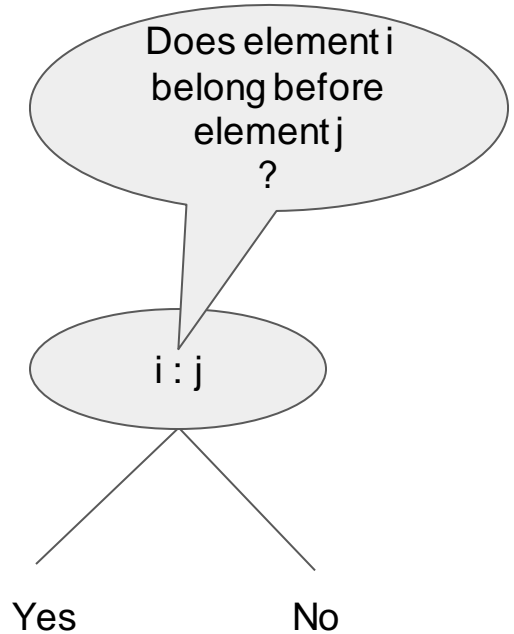
Decision tree for sorting using comparisons

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75 99 50



We find our answer here



How to Read a Decision Tree

- For fixed input size n ...
- Start at root
- Do specified operation at each internal node and follow edge based on outcome
- Leaf reached represents answer

Decision trees \leftrightarrow Sorting Algorithms

- For any comparison sort, we can construct its decision tree on inputs of size n
- (Just trace what the code does for every possible input of size n)
- For any decision tree representing a correct comparison sort, we can derive equivalent code.
- (Follow the tree as shown above.)

Decision trees \leftrightarrow Sorting Algorithms

- For any comparison sort, we can construct its decision tree on inputs of size n .
- (Just take the decision tree of the algorithm.)
- For any decision tree on inputs of size n , we can derive a comparison sort that runs in time $O(n \log n)$.
- (Follow the tree as shown above.)

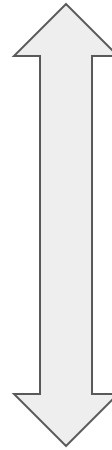
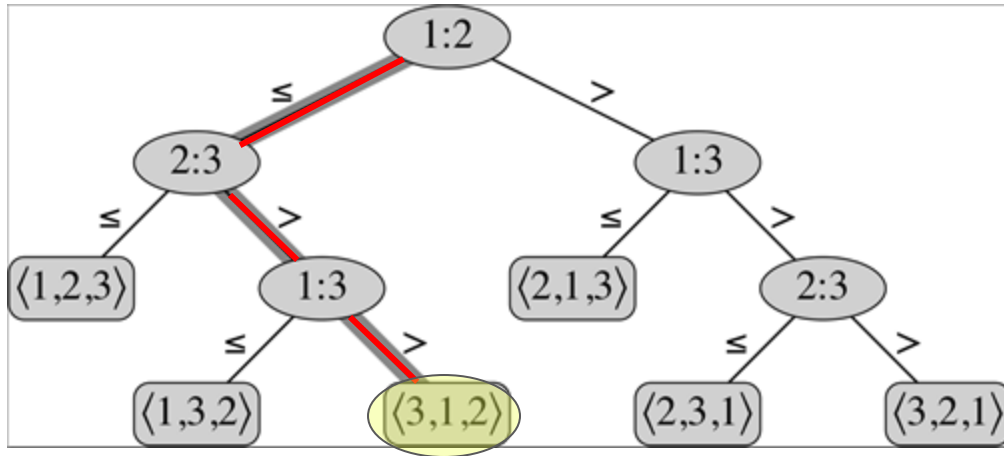
Hence, a lower bound on running time for **all** decision trees holds for **all** comparison sorts, and vice versa.

What's the Running Time of a Decision Tree?

- As many comparisons as it takes to get from root to leaf...
- ... in the worst case → maximum depth
- Hence, running time of a decision tree is its height.

Decision tree for sorting using comparisons

Sorting 3 elements (Figure 8.1 from text)



height = 3

Generic Lower Bound Argument

- Suppose **every** decision tree for a problem of size n has at least $t(n)$ leaves.
- Moreover, the operation labeling each internal node has at most w possible outcomes.
- **Claim:** the problem requires at least $\log_w t(n)$ operations to solve.

Generic Lower Bound Argument

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- **Claim:** the problem requires at least $\log_w t(n)$ operations to solve.
- **Pf:** Tree starts with one root
- Every level of tree increases # nodes by a factor $\leq w$
- Need enough levels h s.t. $w^h \geq t(n)$.
- Hence, $h \geq \log_w t(n)$. QED

Application to Comparison Sorting

- Every node of the tree is a comparison using $>$.
- Hence, $w = ???$.

Application to Comparison Sorting

- Every node of the tree is a comparison using $>$.
- Hence, $w = 2$. [# of outcomes for “Is $x > y$?”]
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Application to Comparison Sorting

- Every node of the tree is a comparison using $>$.
- Hence, $w = 2$. [# of outcomes for “Is $x > y$?”]
- Every leaf of the tree is a possible sorted order of n elements.
- Hence, $t(n) = n!$ [Really?]

$$t(n) = n!$$

- When $n = 1$, only one sorted order.

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We applied the inductive hypothesis.

$t(n) = n!$

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- For $n > 1$, assume all elements of input distinct.
- n possibilities for which element goes first in output.
- Given 1st choice, # of orders of remaining $n-1$ elts is $(n-1)!$
- Hence, $n \times (n-1)! = n!$ possible orders of n elements. QED₆₂

Summary

- For comparison sorting, $w = 2$, $t(n) = n!$
- Hence, by our general theorem, sorting an array of size n requires at least $\log_2 n!$ comparisons in the worst case.

Summary

- For comparison sorting, $w = 2$, $t(n) = n!$
- Hence, by our general theorem, sorting an array of size n requires **at least $\log_2 n!$** comparisons in the worst case.
- ***Wait, how big is $\log n!$???***

Bounding $\log(n!)$

- $\log(n!) = \log(n \times (n-1) \times (n-2) \times \dots \times 1)$

Bounding $\log(n!)$

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 $= \log n + \log (n-1) + \log (n-2) + \dots + \log(2) + \log(1)$

Bounding $\log(n!)$

- $\log(n!) = \log(n \times (n-1) \times (n-2) \times \dots \times 1)$
 $= \log(n) + \log(n-1) + \log(n-2) + \dots + \log(2) + \log(1)$

 $\leq \log(n) + \log(n) + \log(n) + \dots + \log(n) + \log(n)$

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So $\log(n!) = O(n \log n)$ <-- Upper Bound

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So $\log(n!) = O(n \log n)$ <-- Upper Bound

--look familiar? bound on n heap operations!

Bounding $\log(n!)$: lower bound*

- $\log(n!) = \log(n \times (n-1) \times (n-2) \times \dots \times 1)$
 $= \log(n) + \log(n-1) + \log(n-2) + \dots + \log(2) + \log(1)$

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So $\log(n!) = \Omega(n \log n)$ <-- Lower bound

Bounding $\log(n!)$

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$$\log(n!) = O(n \log n) \quad \leftarrow \text{Upper bound}$$

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$$\log(n!) = \Theta(n \log n)$$

Summary

- For comparison sorting, $w = 2$, $t(n) = n!$
- Hence, by our general theorem, sorting an array of size n requires **at least $\log_2 n!$** comparisons in the worst case.

Summary

- For comparison sorting, $w = 2$, $t(n) = n!$
- Hence, by our general theorem, sorting an array of size n requires $\Omega(n \log n)$ comparisons in the worst case.
- **→ MergeSort and HeapSort are asymptotically optimal comparison sorts!**

OK, so it's **impossible**
to sort in time less
than $n \log n$...

OK, so it's **impossible**
to sort in time less
than $n \log n$... using
comparisons

Next, let's see **how** we
can sort in time less
than $n \log n$.

Breaking the $n \log n$ barrier

- Our lower bound is for **comparison sorts**, which work on items from any totally ordered set.
- To sort faster, we need to be able to inspect input using ops other than comparisons.
- Will limit attention to sorting **integers**.

Counting Sort

- Assume our inputs are n integers in range $[0, k)$.
- Count how often each value occurs in input.
- Write that many values to output.

Counting Sort ($k = 5$)

[1 4 2 0 1 3 0 1 2 3]

Value	Count
0	
1	
2	
3	
4	

Counting Sort (k = 5)

[1 4 2 0 1 3 0 1 2 3]



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Counting Sort (k = 5)

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[]

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Counting Sort (k = 5)

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Counting Sort (k = 5)

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[0 0 1 1 1]

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$$\text{Cost} = \Theta(n + k)$$

Fun Facts About Counting Sort

- Counting sort is a “linear-time” sort (in n)
 - Still depends on k
- Can extend to sort arbitrary items with **integer keys**
- Highly efficient when max value **k** is small vs n

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- Counting sort is a “linear-time” sort (in n)
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- Can extend to sort arbitrary items with **integer keys**
- Highly efficient when max value **k** is small vs n
- **But what if k is large?**

Radix Sort

- Divide each input integer into d digits
- Digits may be in any base k ; we'll use base 10 in example
- Sort using d successive passes of counting sort
- j th pass uses j th digit of each input as sorting key

Radix Sort – Key Requirements

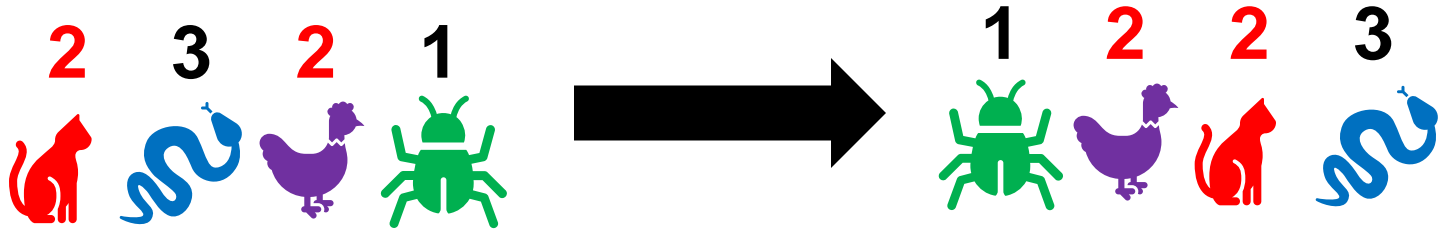
- Sort using d successive passes of counting sort.
- We sort by **least significant** digit first.
- Sort in each pass must be **stable** – never inverts order of two inputs with the same key.



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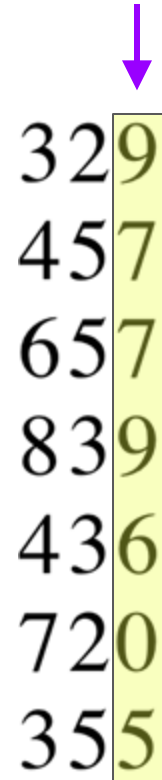
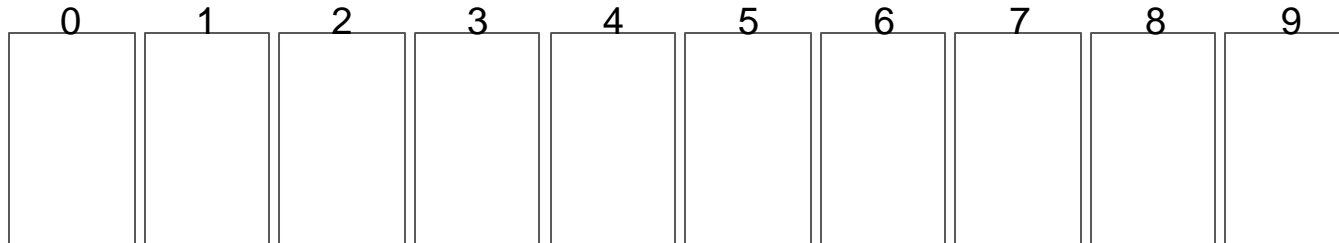
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NOT STABLE!



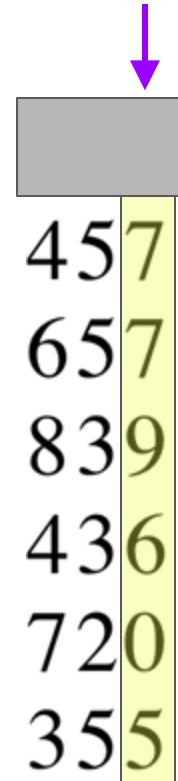
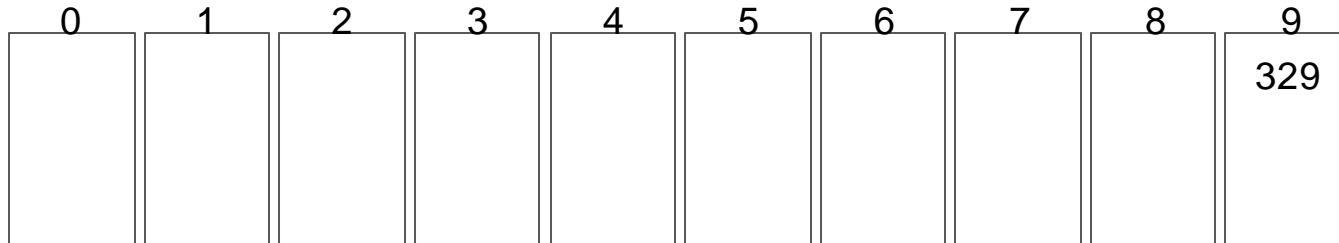
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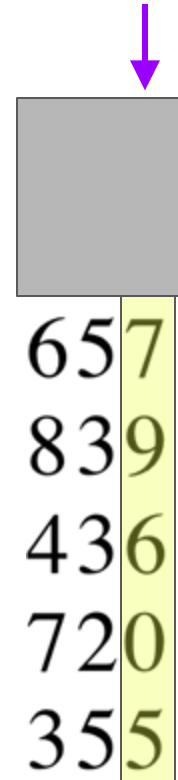
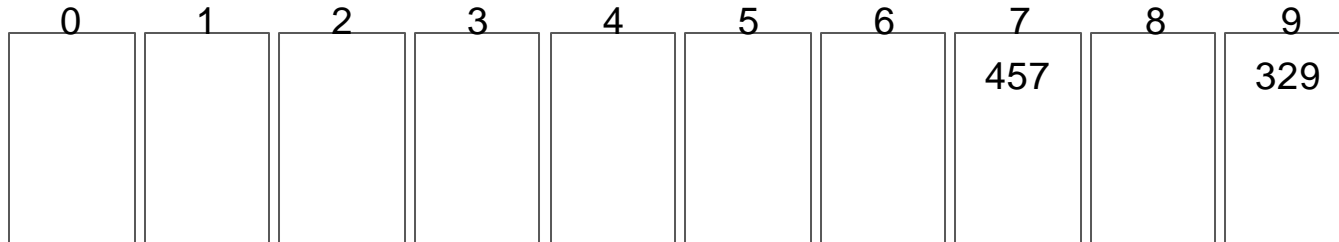
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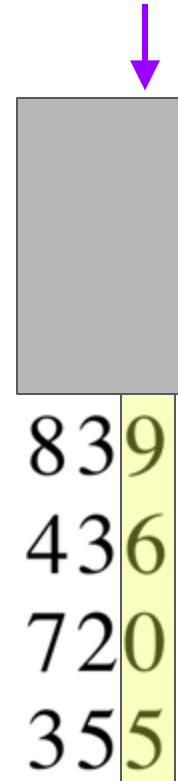
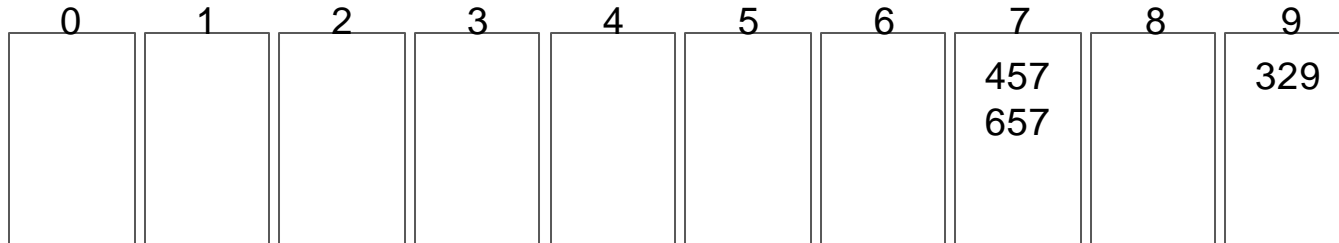
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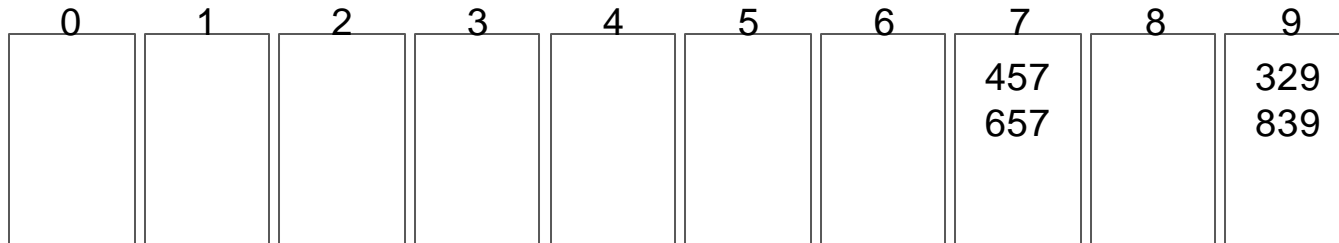
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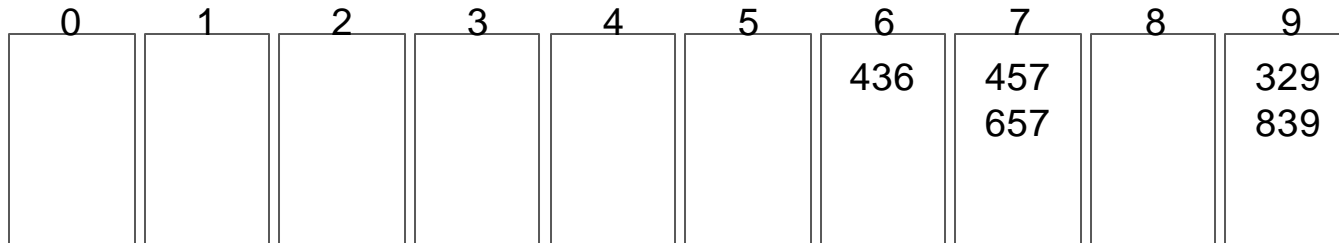
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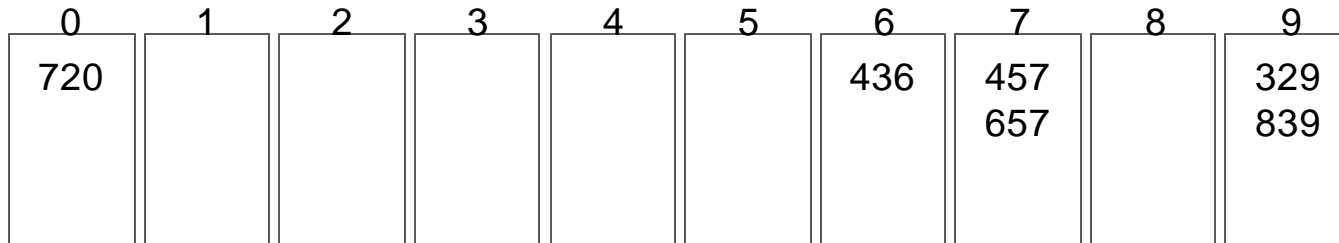
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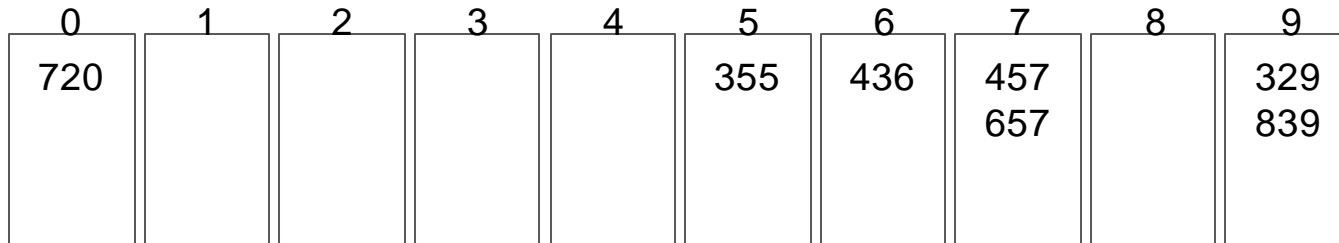
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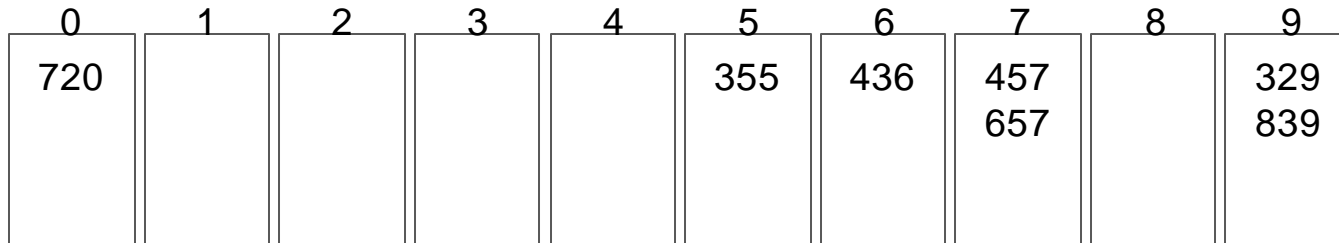
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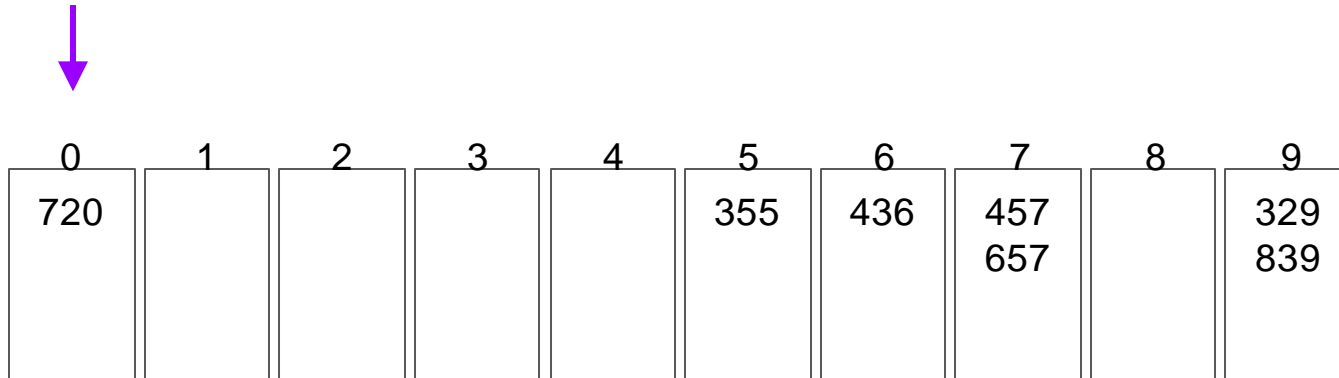
We next recreate the list by sweeping the bins



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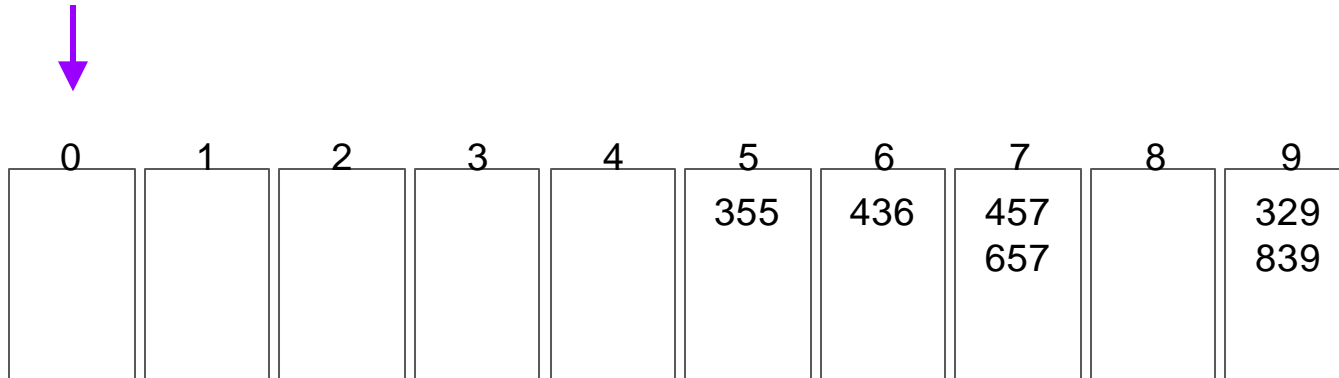
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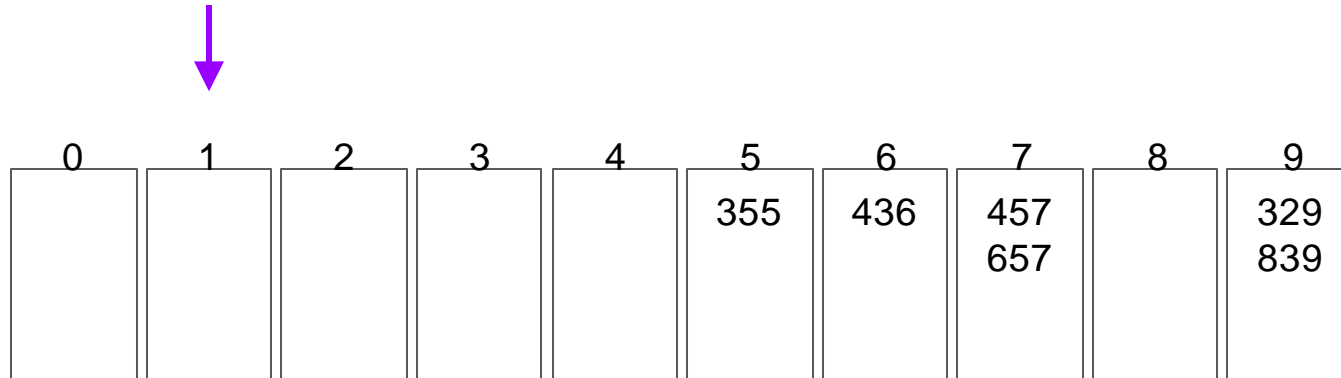
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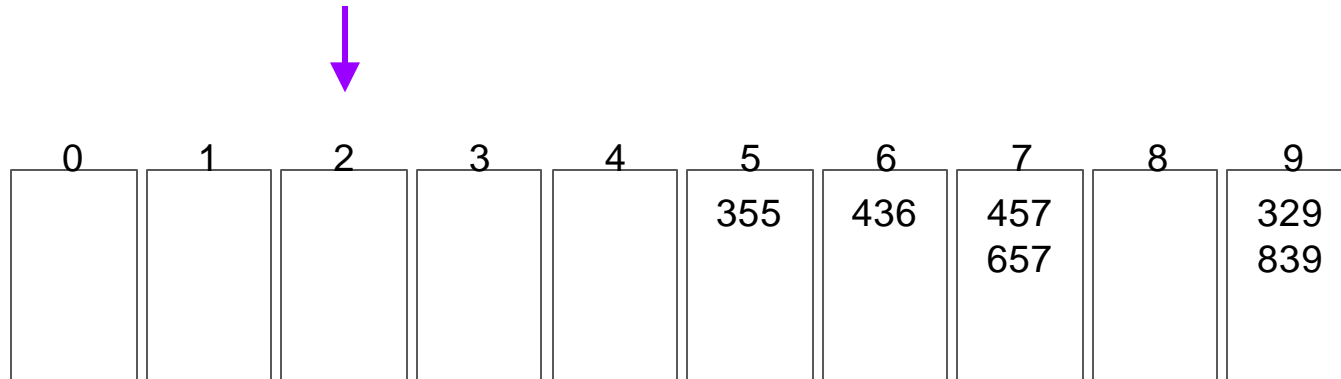
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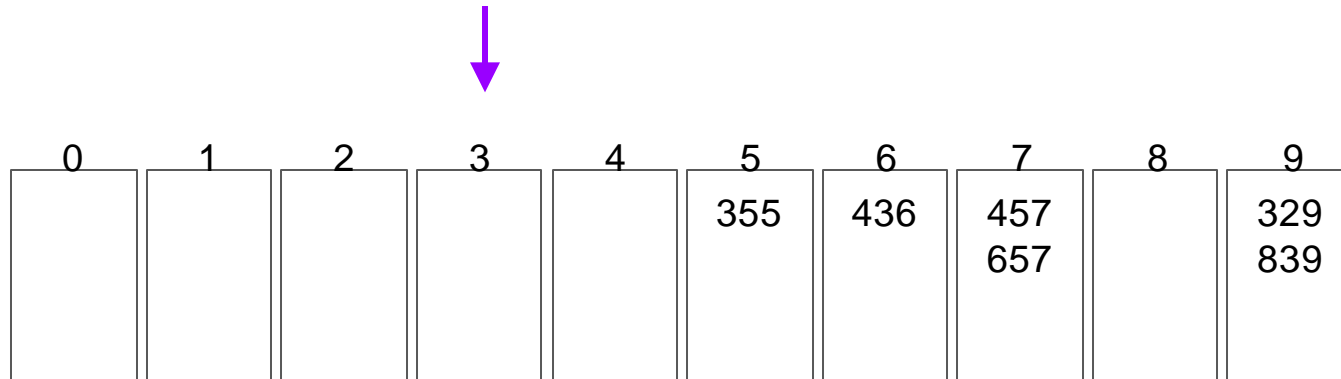
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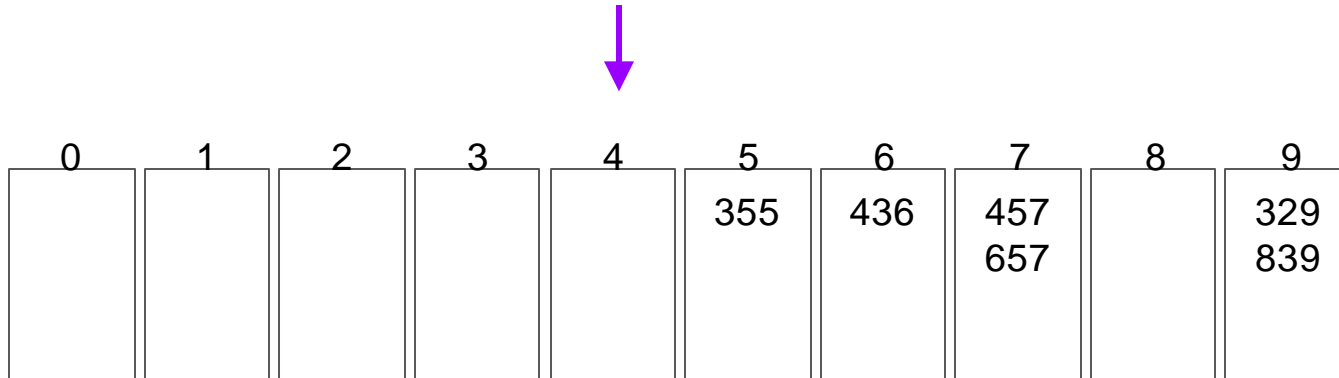
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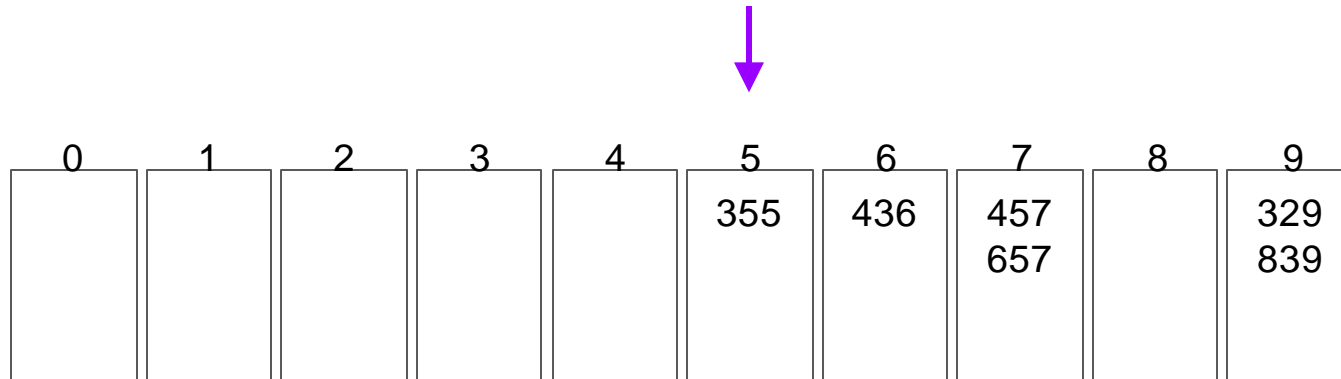
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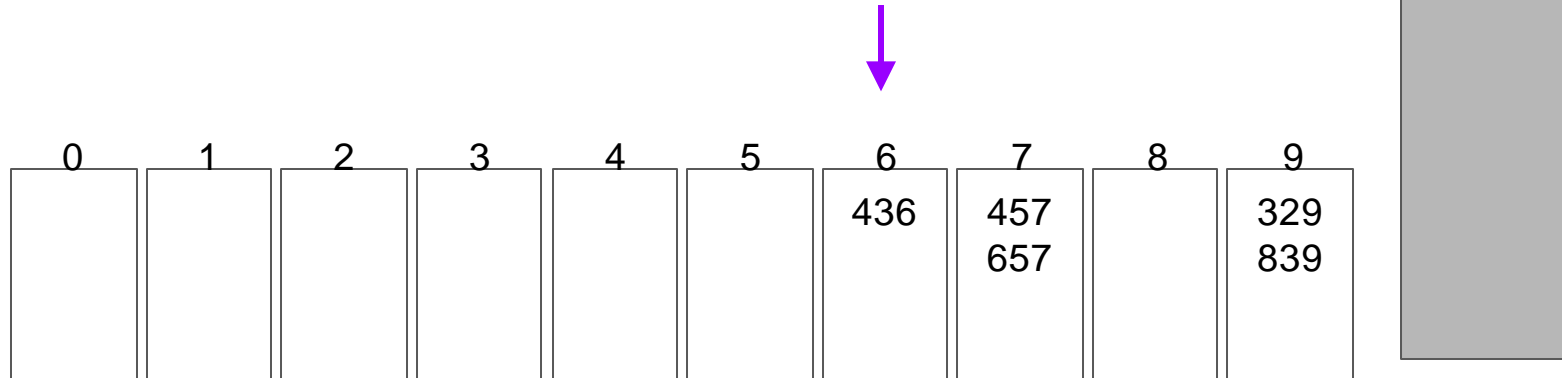
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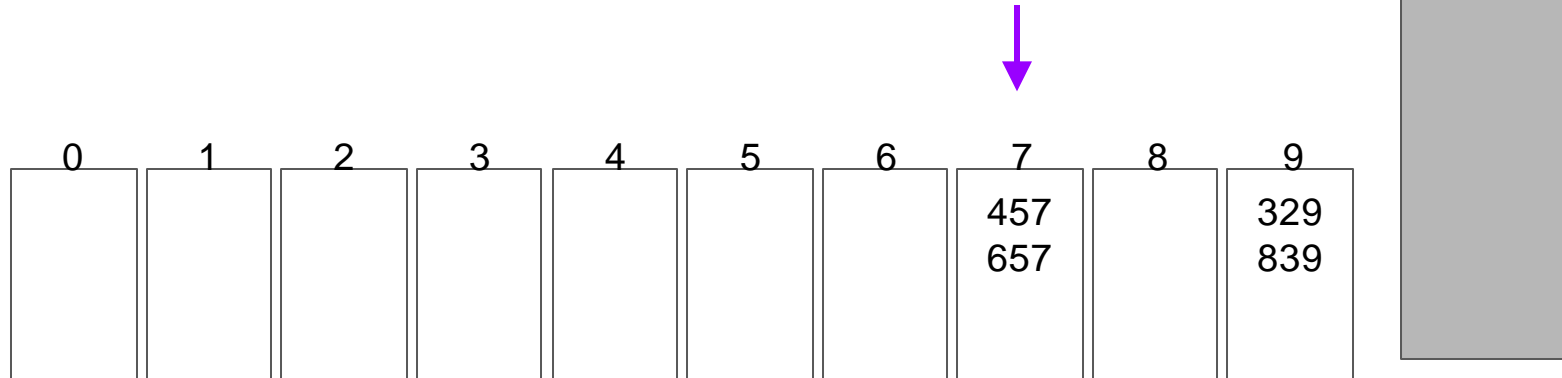
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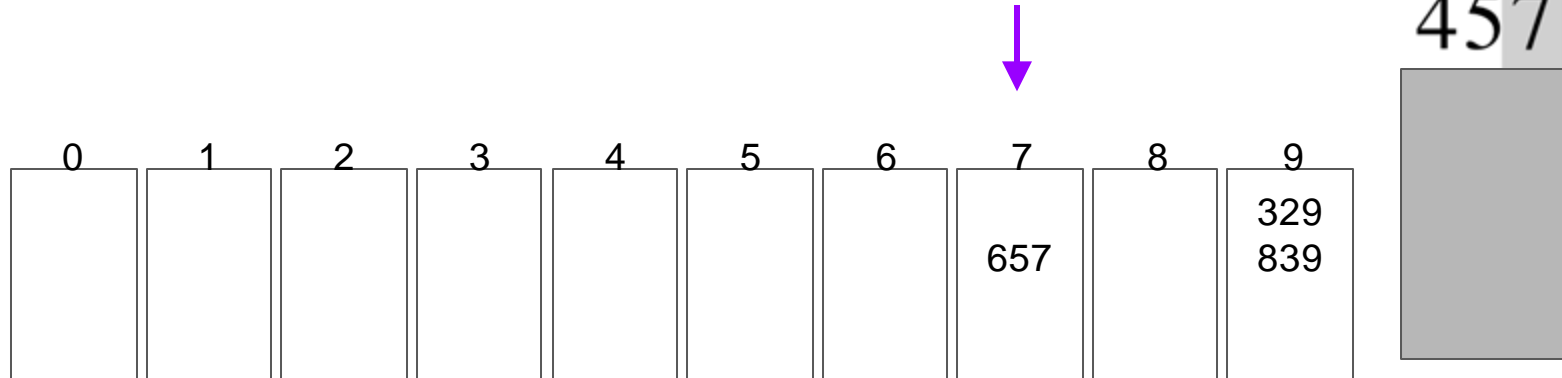
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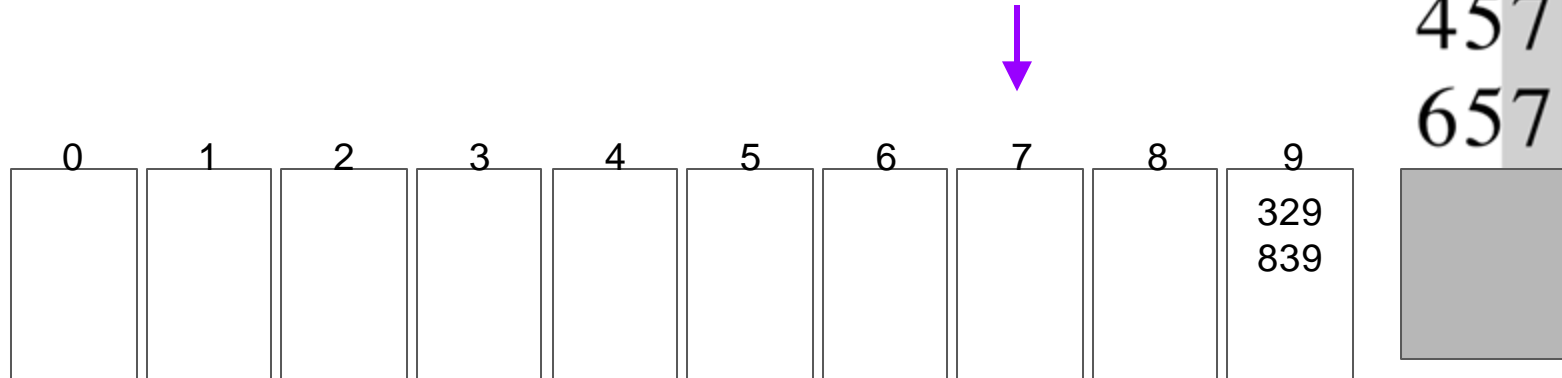
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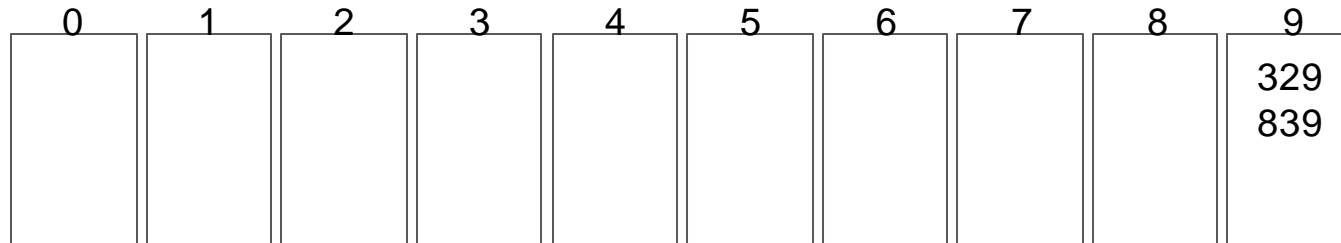
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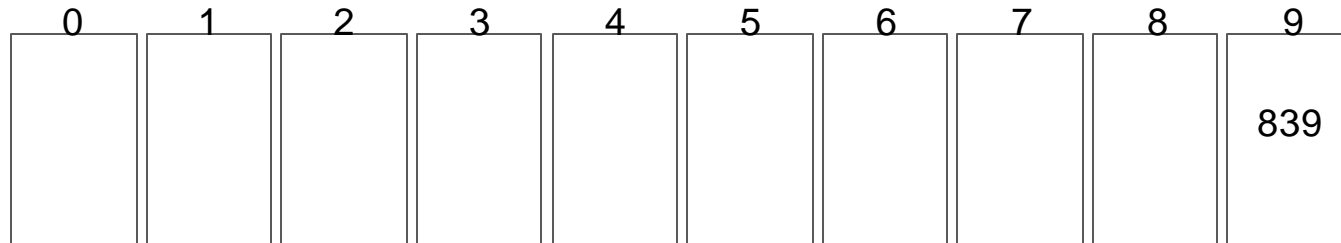
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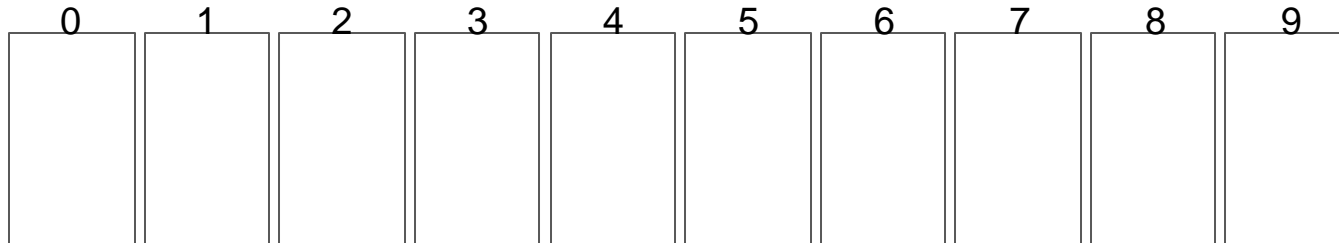
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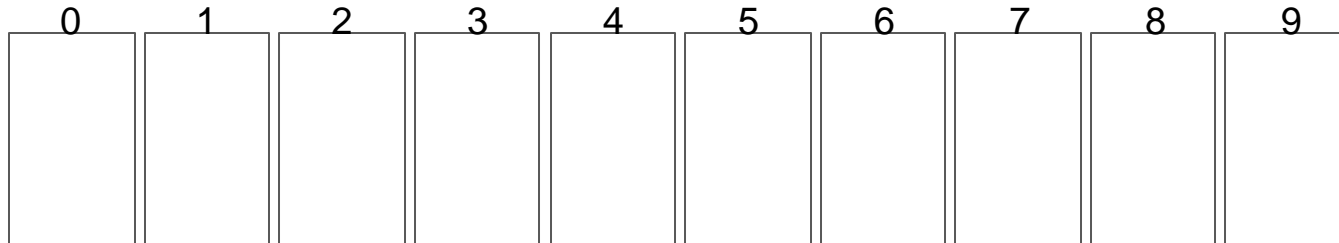
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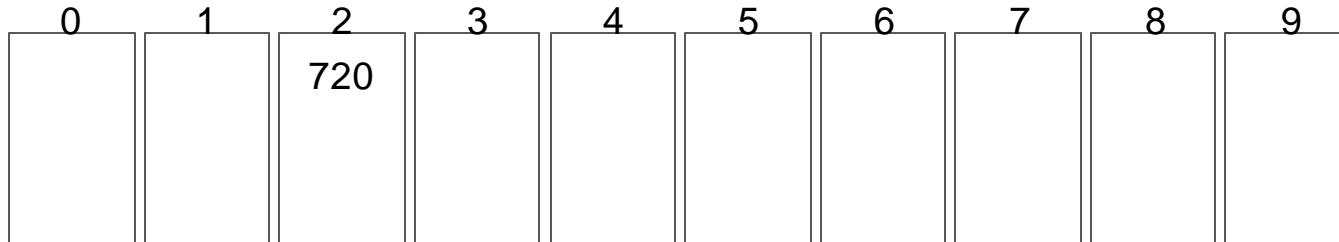
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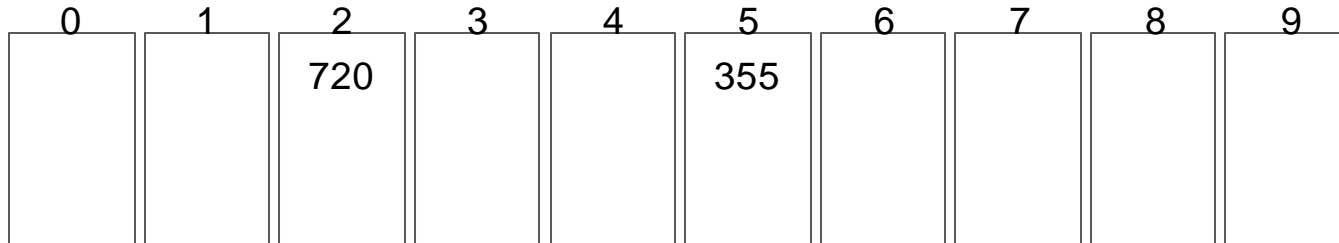
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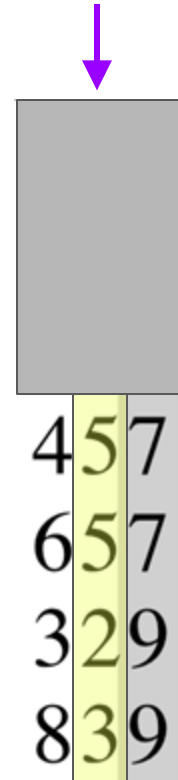
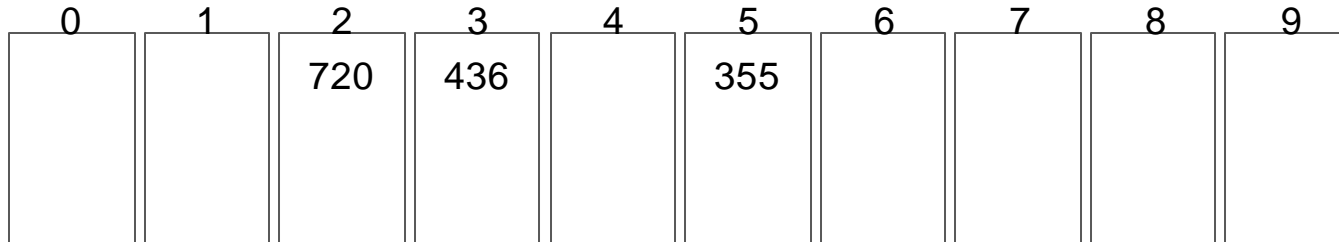
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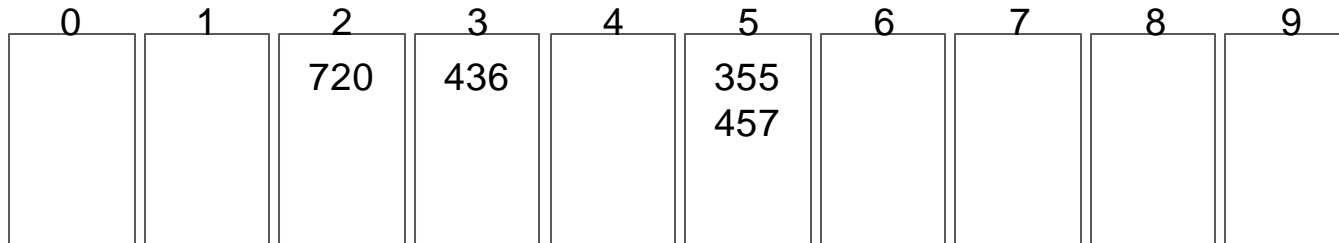
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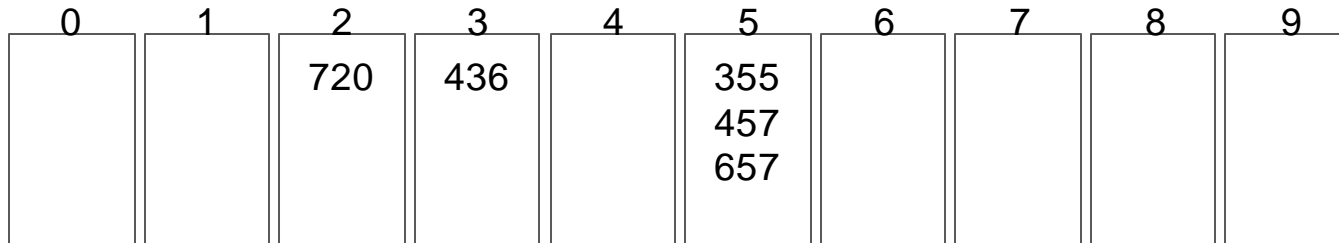
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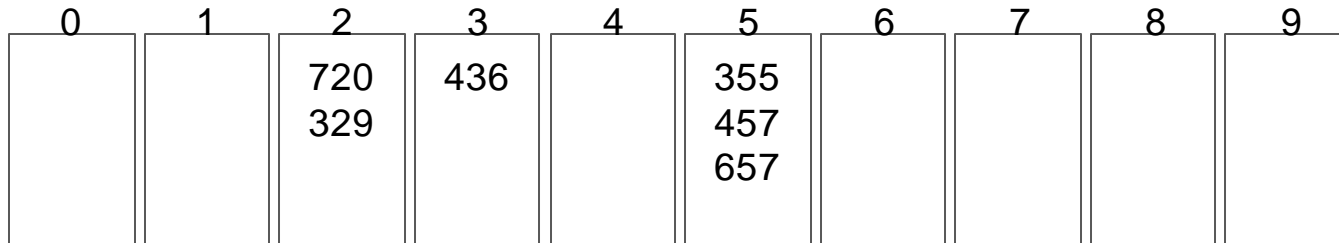
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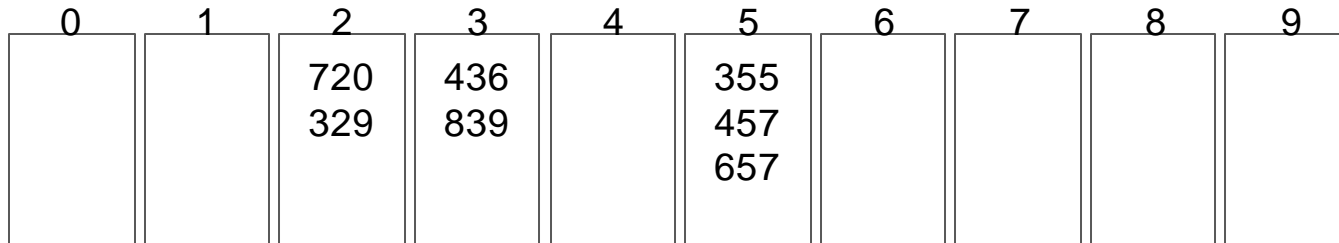
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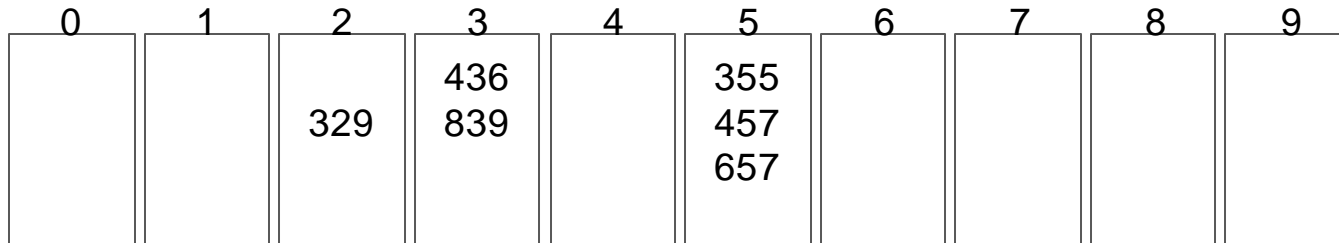
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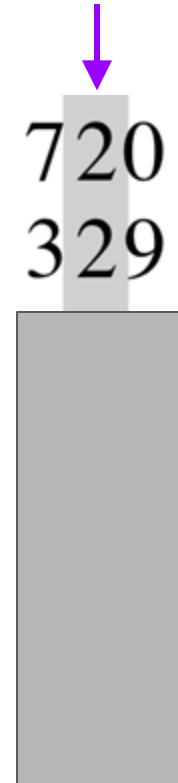
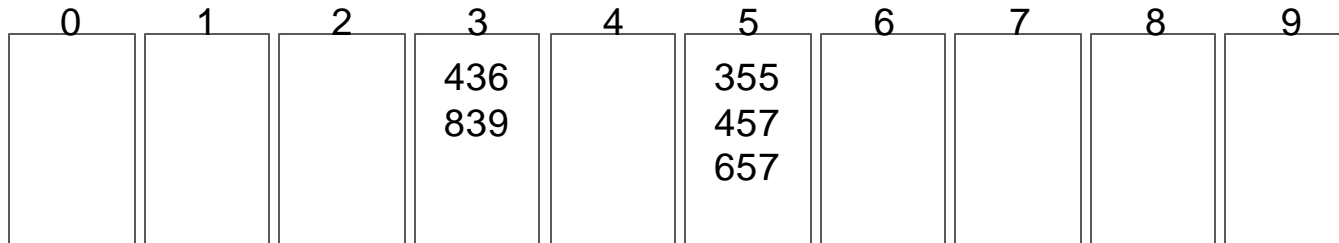
Radix Sort

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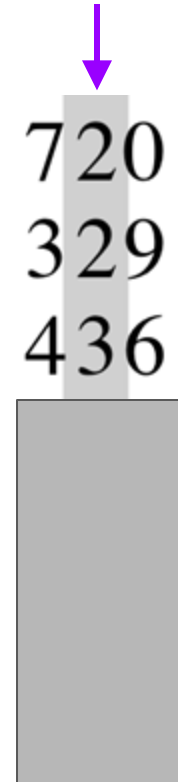
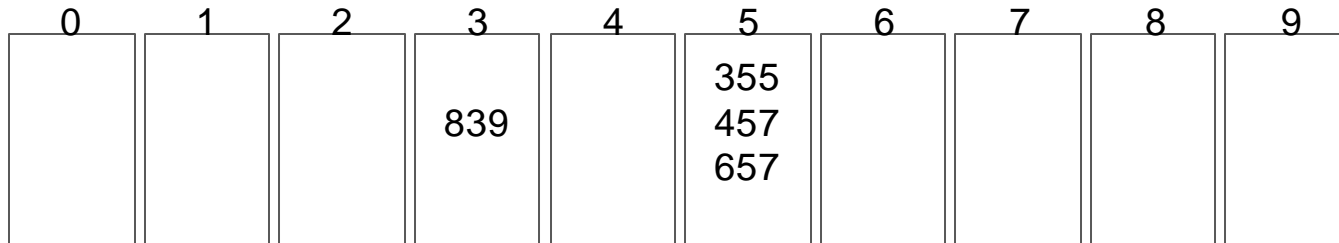
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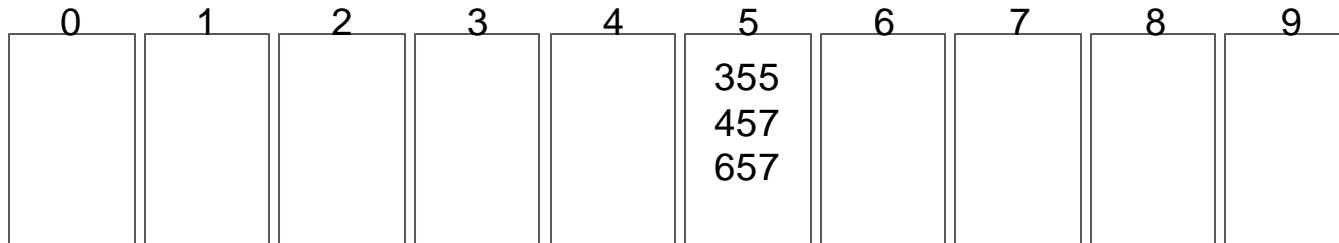
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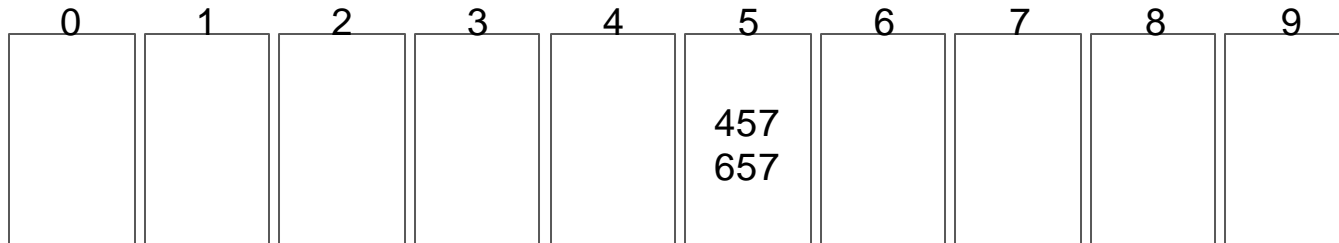
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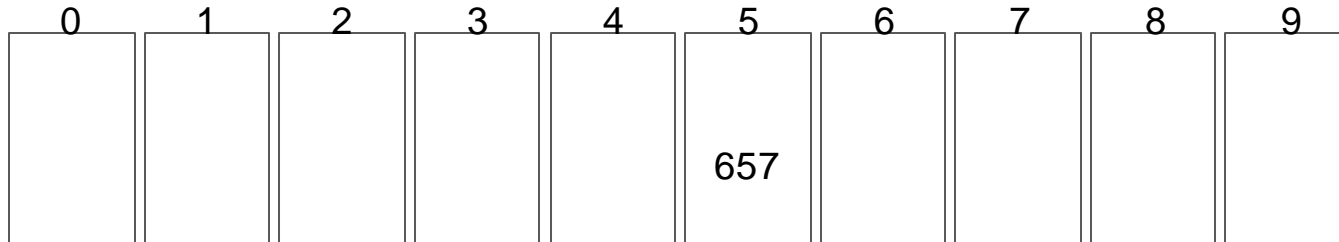
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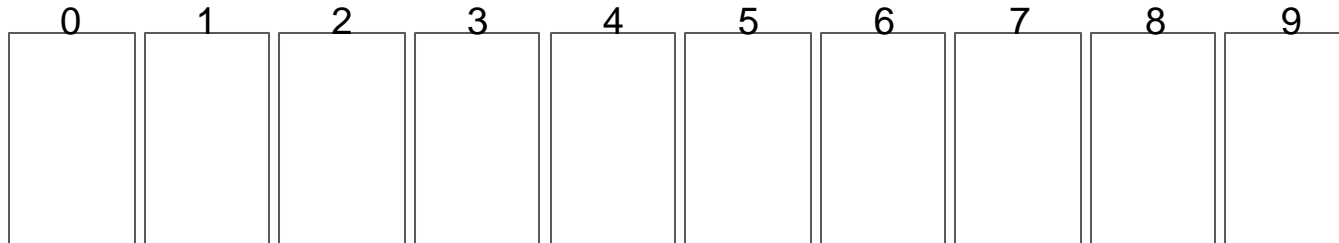
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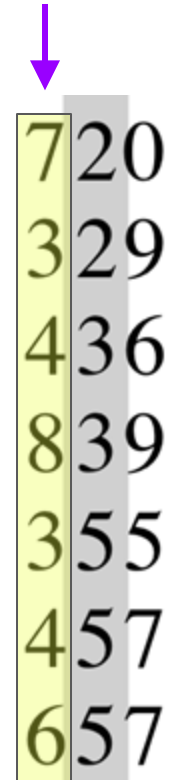
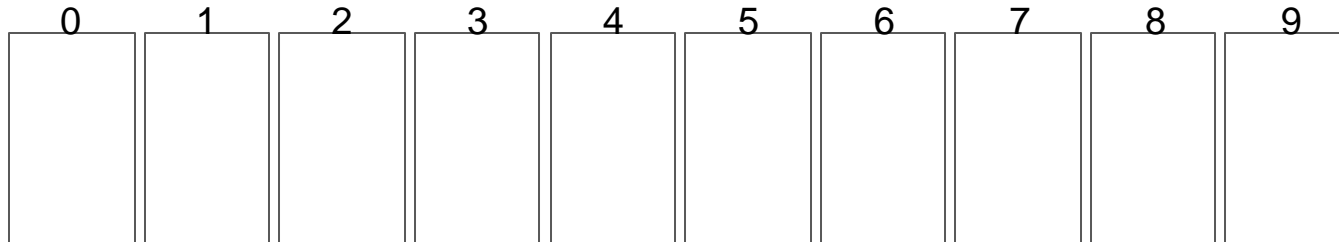
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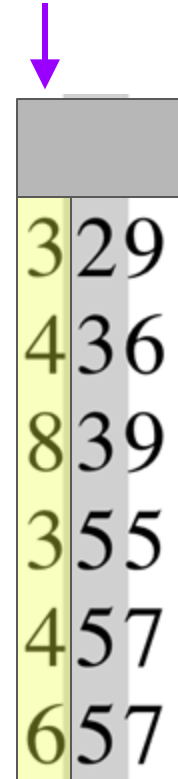
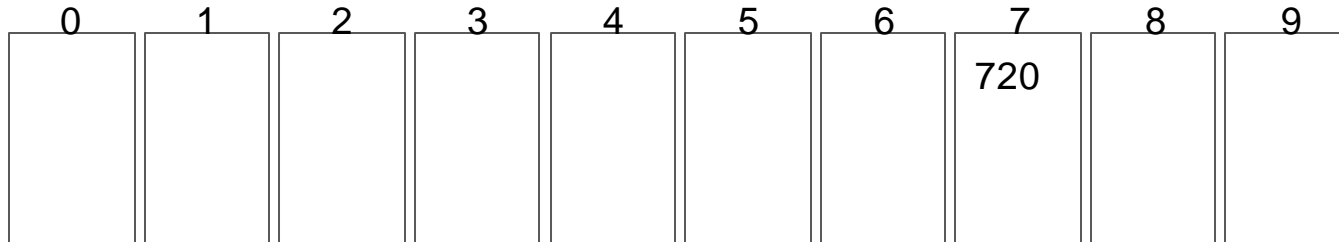
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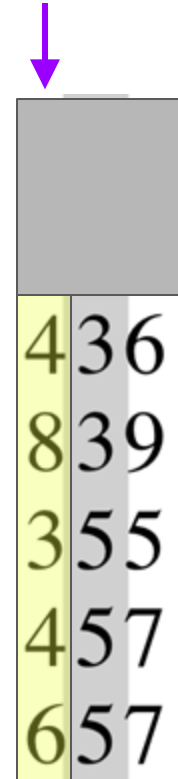
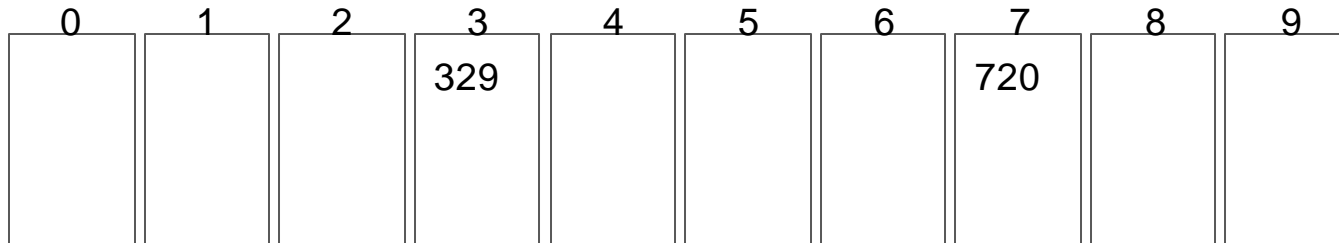
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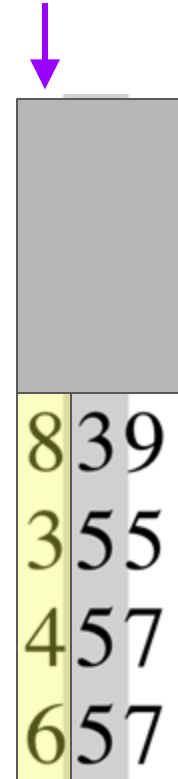
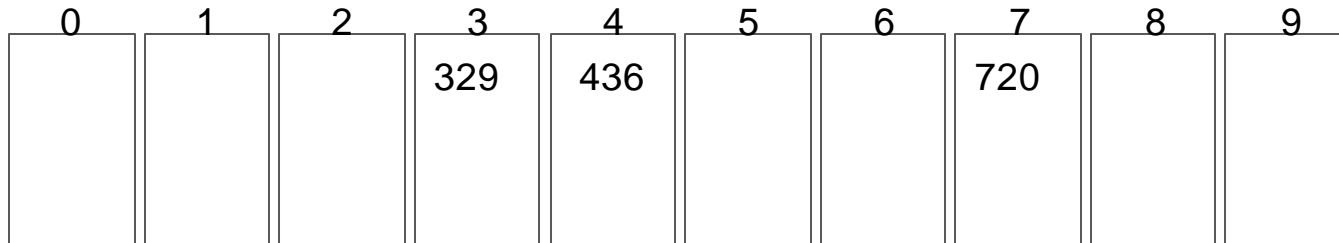
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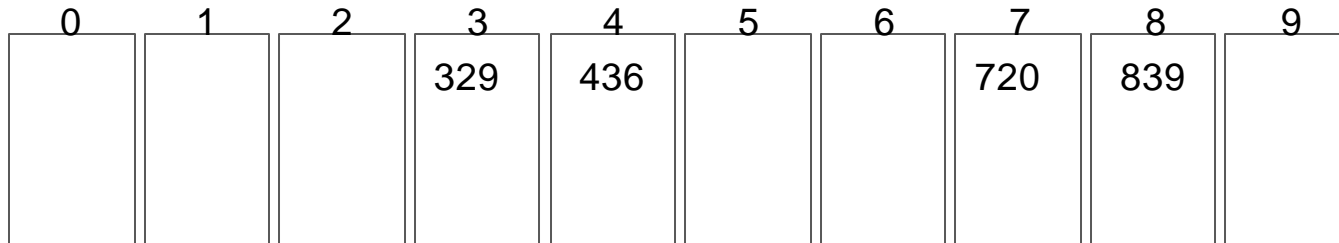
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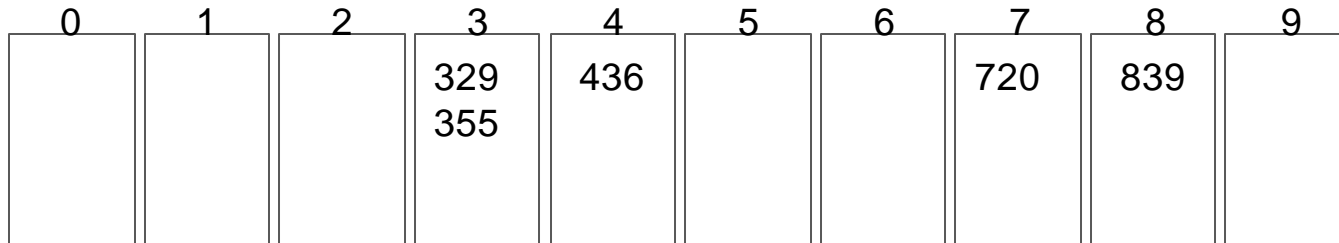
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Radix Sort

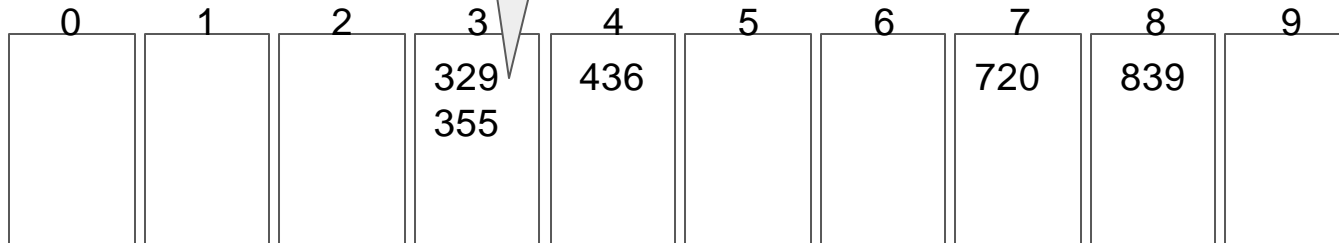
- From left to right, significant digit j

- C

These end up in the right order because....

position j

- A_j in bucket e_j



Radix Sort

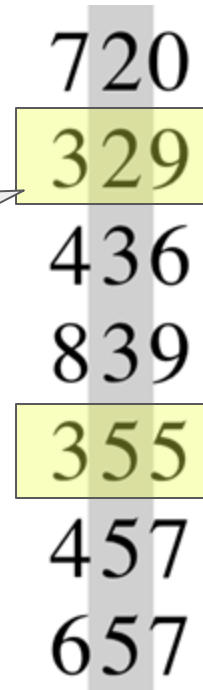
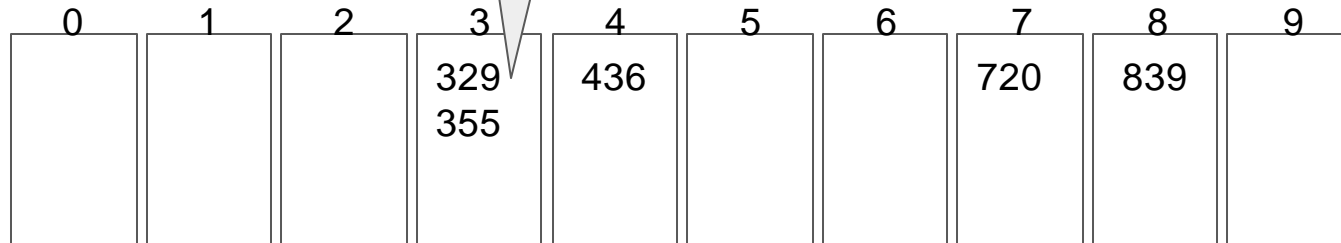
- From least significant digit j

- C

These end up in the right order because....

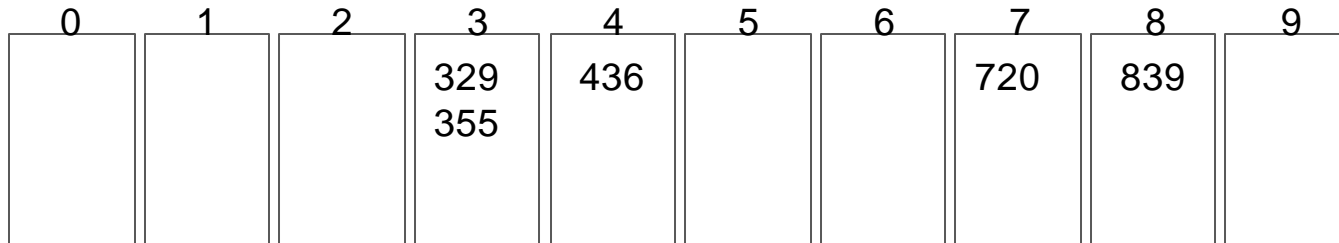
position j

The sort is *stable*: the original order of elements is preserved within a bin



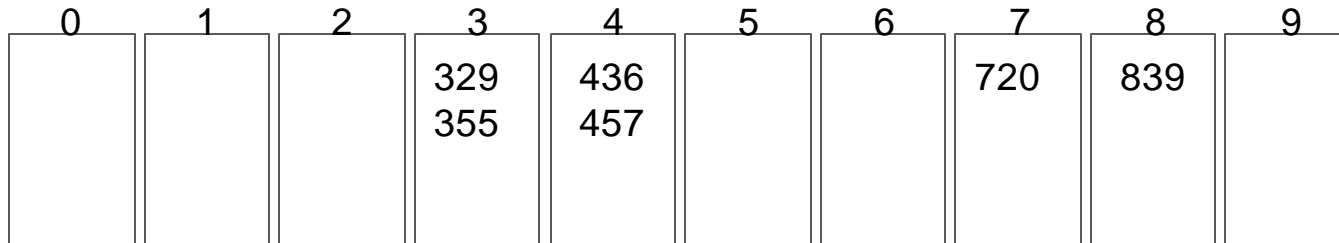
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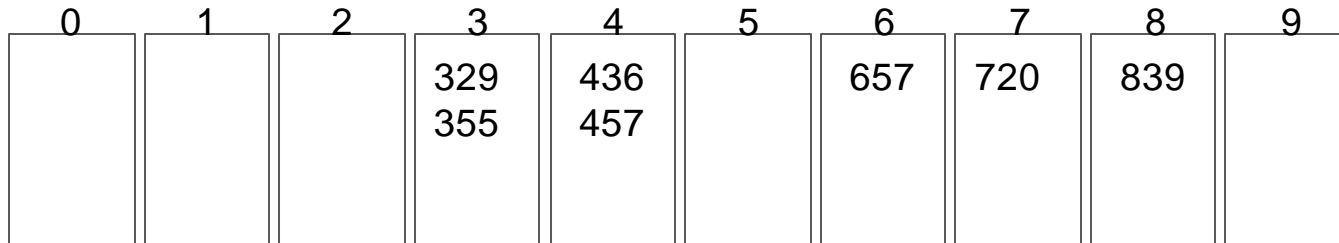
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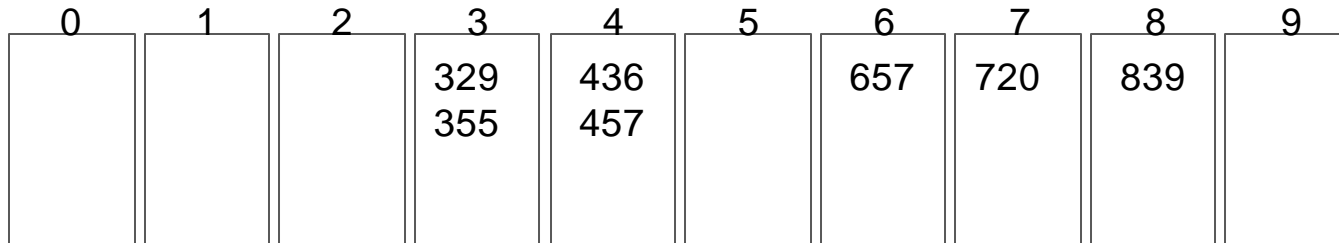
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Radix Sort

- Alternative visualization: [Radix sort on the playground](#)

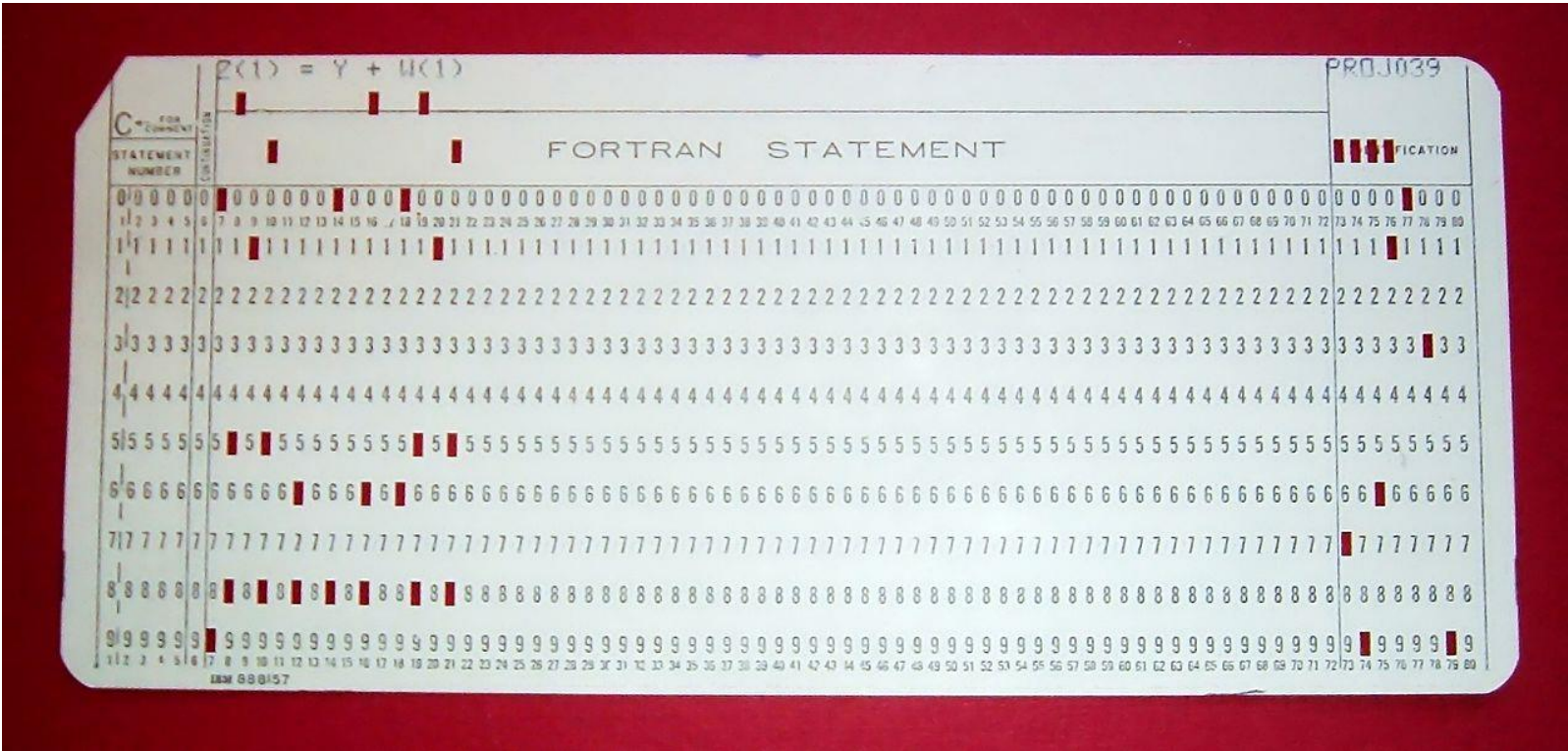
Why Does Radix Sort Work?

- **Invariant** – after j sorting passes, input is sorted by its j th least significant digits. *[Prove inductively on j]*
- Stability is needed to show that invariant holds for inputs with equal-valued j th digits.
- *(Proof left as exercise – see Lab 6.)*

What Does Radix Sort Cost?

- d passes of counting sort
- Each pass takes time $\Theta(n + k)$
 - Why?
- Hence, total time is $\Theta(d(n+k))$

Application: Sorting Punch Cards



Whoops! We Dropped our Deck!



Application: Sorting Punch Cards



https://en.wikipedia.org/wiki/IBM_card_sorter

Sorting

End of notes