

DESIGN AND ANALYSIS OF ALGORITHMS

Lecture 6: Linear Time Sorting



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Time complexities of Comparison Sorts

Sorting Algorithms	Best Case	Average Case	Worst Case
Insertion Sort	$O(n)$	$O(n^2)$	$O(n^2)$
Merge Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$
Quick Sort	$O(n \log n)$	$O(n \log n)$ (Randomized Quick sort)	$O(n^2)$
Heap Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$

HOW FAST CAN WE SORT?

All the sorting algorithms we have seen so far are *comparison sorts*: only use comparisons to determine the relative order of elements.

- *E.g.*, insertion sort, merge sort, quicksort, heapsort.

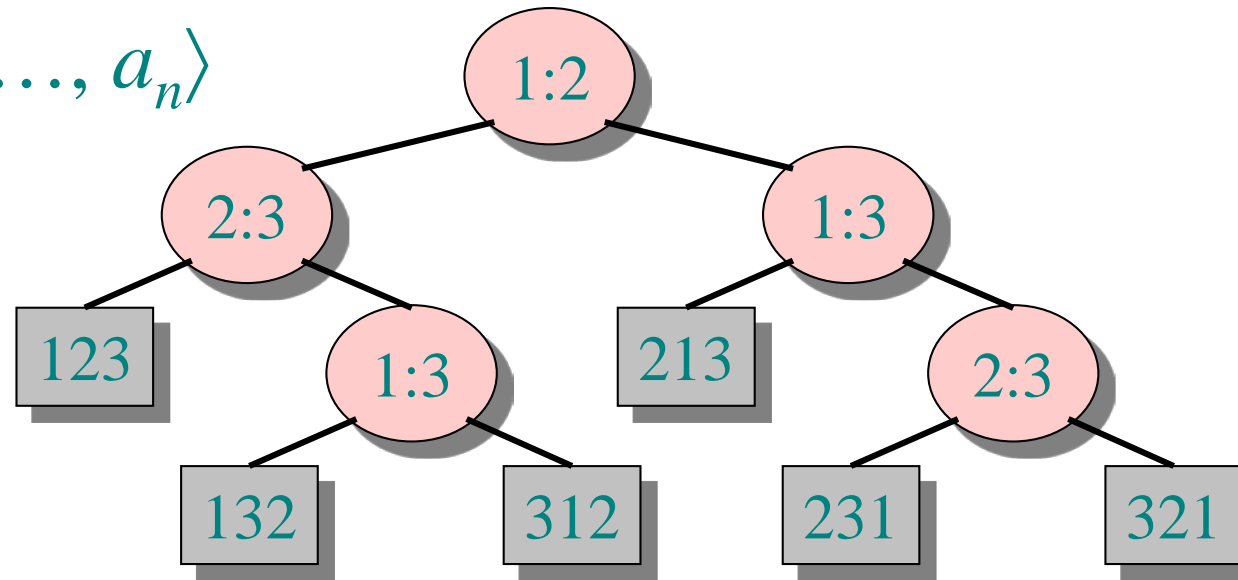
The best worst-case running time that we've seen for comparison sorting is $O(n \lg n)$.

Is $O(n \lg n)$ the best we can do?

Decision trees can help us answer this question.

DECISION-TREE EXAMPLE

Sort $\langle a_1, a_2, \dots, a_n \rangle$

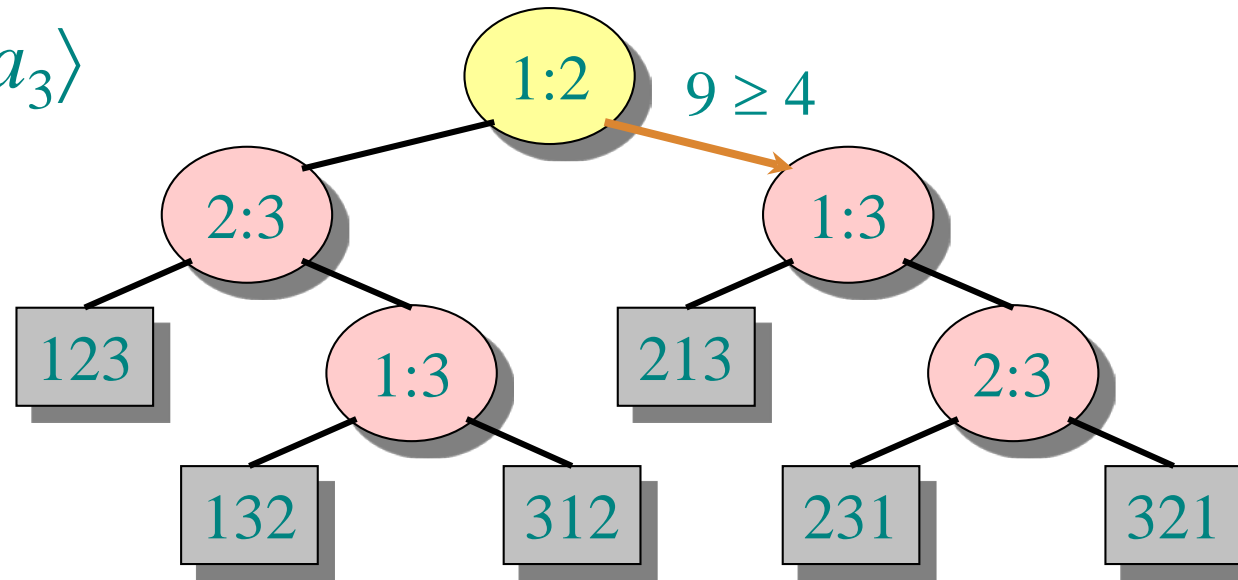


Each internal node is labeled $i:j$ for $i, j \in \{1, 2, \dots, n\}$.

- The left subtree shows subsequent comparisons if $a_i \leq a_j$.
- The right subtree shows subsequent comparisons if $a_i \geq a_j$.

DECISION-TREE EXAMPLE

Sort $\langle a_1, a_2, a_3 \rangle$
 $= \langle 9, 4, 6 \rangle$:

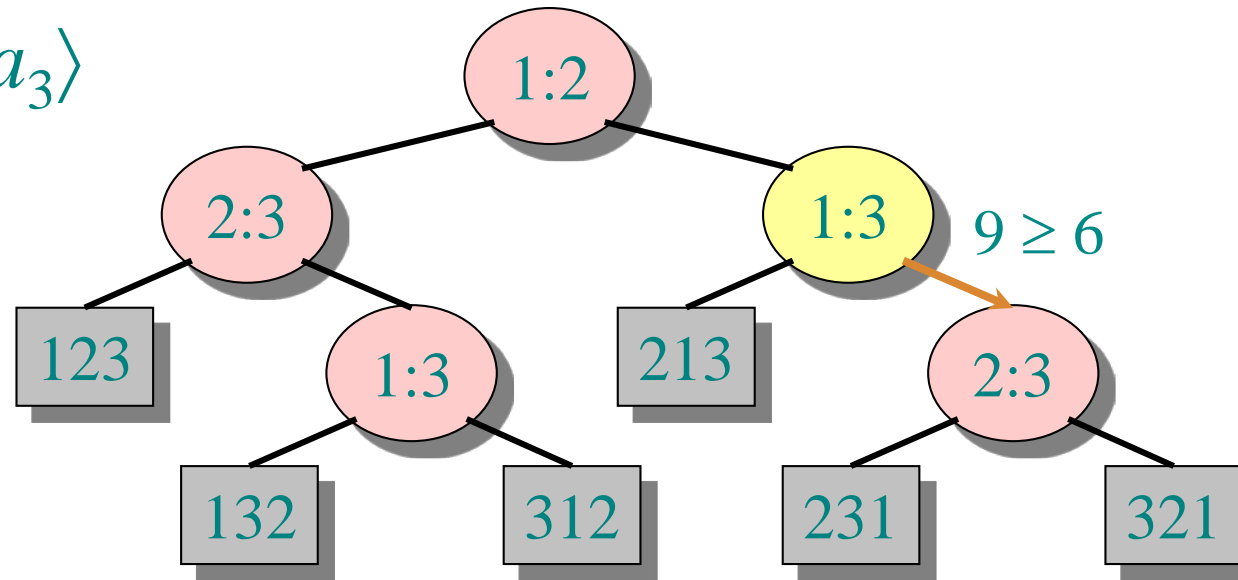


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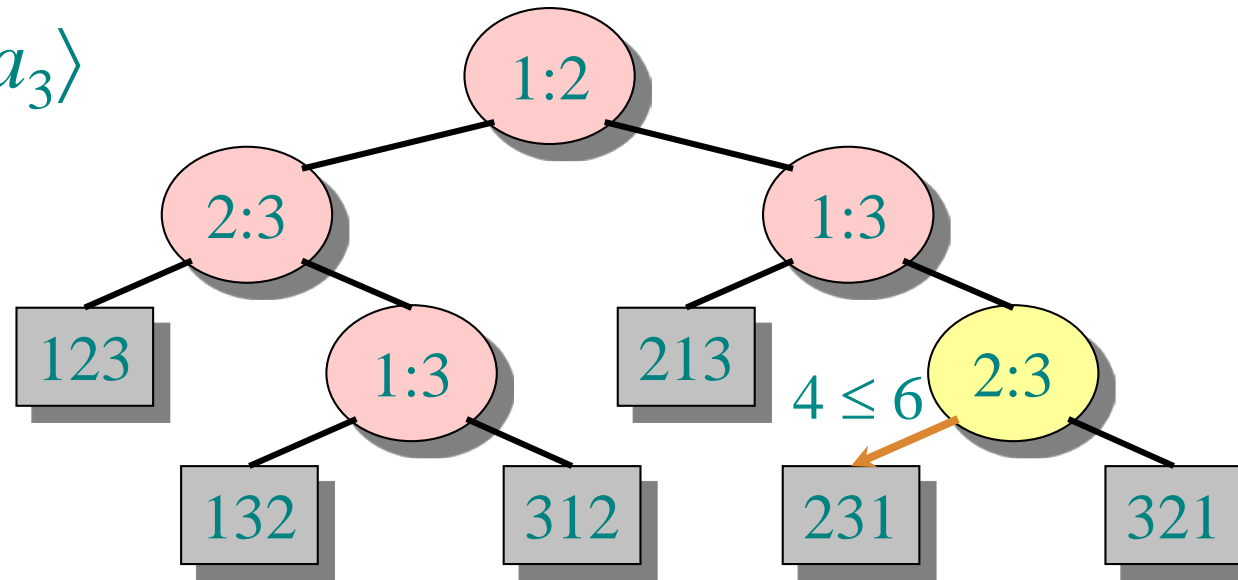


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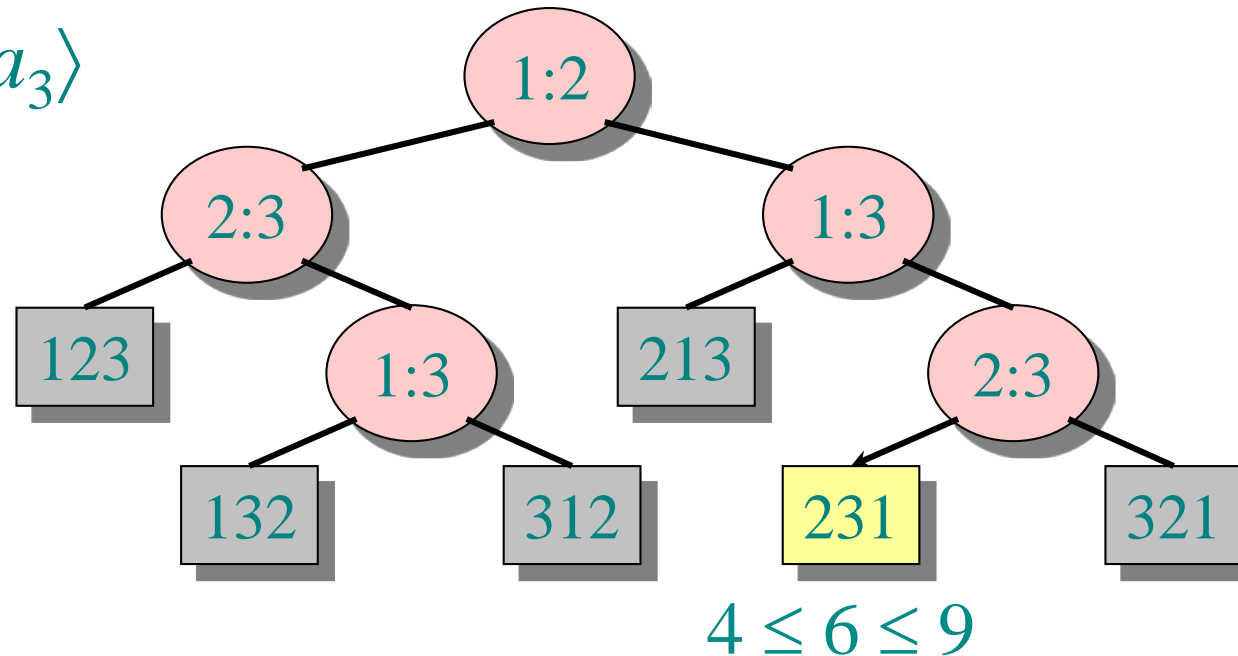


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- The left subtree shows subsequent comparisons if $a_i \leq a_j$.
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DECISION-TREE EXAMPLE

Sort $\langle a_1, a_2, a_3 \rangle$
 $= \langle 9, 4, 6 \rangle$:



Each leaf contains a permutation $\langle \pi(1), \pi(2), \dots, \pi(n) \rangle$ to indicate that the ordering $a_{\pi(1)} \leq a_{\pi(2)} \leq \dots \leq a_{\pi(n)}$ has been established.

DECISION-TREE MODEL

A decision tree can model the execution of any comparison sort:

- One tree for each input size n .
- View the algorithm as splitting whenever it compares two elements.
- The tree contains the comparisons along all possible instruction traces.
- The running time of the algorithm = the length of the path taken.
- Worst-case running time = height of tree.

LOWER BOUND FOR DECISION-TREE SORTING

Theorem. Any decision tree that can sort n elements must have height $\Omega(n \lg n)$.

Proof. The tree must contain $\geq n!$ leaves, since there are $n!$ possible permutations. A height- h binary tree has $\leq 2^h$ leaves. Thus, $n! \leq 2^h$.

$$\begin{aligned} \therefore h &\geq \lg(n!) && (\lg \text{ is mono. increasing}) \\ &\geq \lg((n/e)^n) && (\text{Stirling's formula}) \\ &= n \lg n - n \lg e \\ &= \Omega(n \lg n). \end{aligned}$$

LOWER BOUND FOR COMPARISON SORTING

Corollary. Heapsort and merge sort are asymptotically optimal comparison sorting algorithms.

SORTING IN LINEAR TIME

Counting sort: No comparisons between elements.

- *Input:* $A[1 \dots n]$, where $A[j] \in \{1, 2, \dots, k\}$.
- *Output:* $B[1 \dots n]$, sorted.
- *Auxiliary storage:* $C[1 \dots k]$.

COUNTING SORT

for $i \leftarrow 1$ **to** k

do $C[i] \leftarrow 0$

for $j \leftarrow 1$ **to** n

do $C[A[j]] \leftarrow C[A[j]] + 1$ $\triangleright C[i] = |\{\text{key} = i\}|$

for $i \leftarrow 2$ **to** k

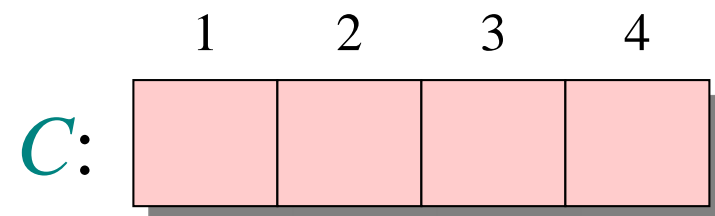
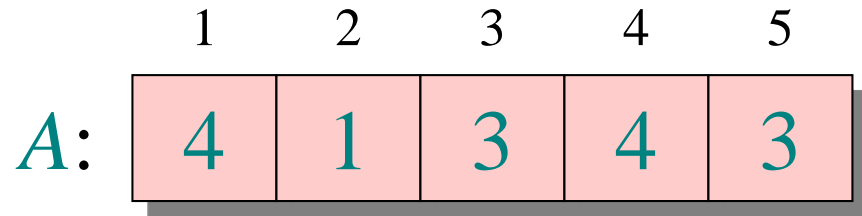
do $C[i] \leftarrow C[i] + C[i-1]$ $\triangleright C[i] = |\{\text{key} \leq i\}|$

for $j \leftarrow n$ **downto** 1

do $B[C[A[j]]] \leftarrow A[j]$

$C[A[j]] \leftarrow C[A[j]] - 1$

COUNTING-SORT EXAMPLE



LOOP 1

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	0	0	0	0

for $i \leftarrow 1$ **to** k
 do $C[i] \leftarrow 0$

LOOP 2

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

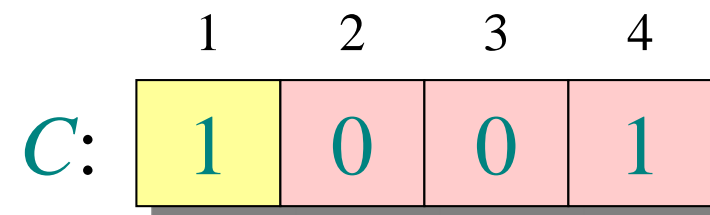
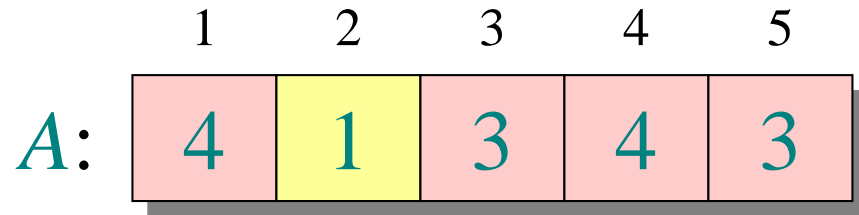
	1	2	3	4
<i>C</i> :	0	0	0	1

<i>B</i> :					
------------	--	--	--	--	--

for $j \leftarrow 1$ **to** n

do $C[A[j]] \leftarrow C[A[j]] + 1$ $\triangleright C[i] = |\{\text{key} = i\}|$

LOOP 2



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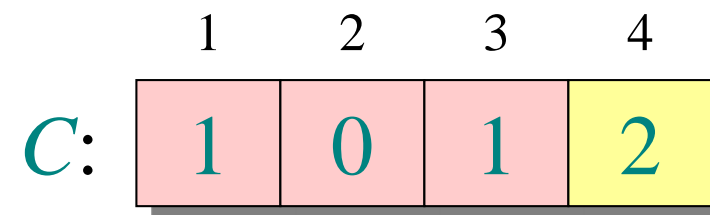
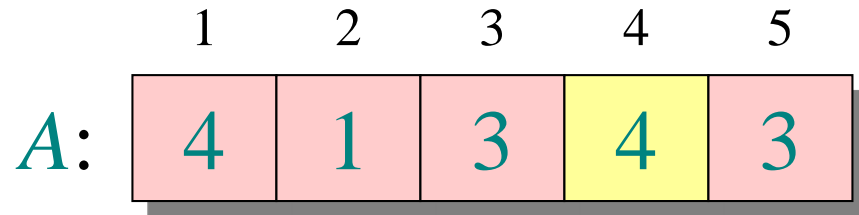
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LOOP 2

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

	1	2	3	4
<i>C</i> :	1	0	2	2

<i>B</i> :					
------------	--	--	--	--	--

for $j \leftarrow 1$ **to** n

do $C[A[j]] \leftarrow C[A[j]] + 1$ $\triangleright C[i] = |\{\text{key} = i\}|$

LOOP 3

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	1	0	2	2

<i>C'</i> :	1	1	2	2
-------------	---	---	---	---

for $i \leftarrow 2$ **to** k

do $C[i] \leftarrow C[i] + C[i-1]$

▷ $C[i] = |\{\text{key} \leq i\}|$

LOOP 3

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	1	0	2	2

<i>C'</i> :	1	1	3	2
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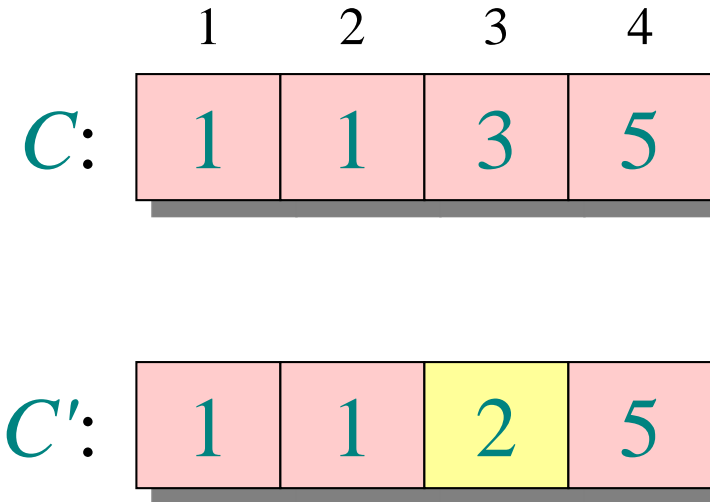
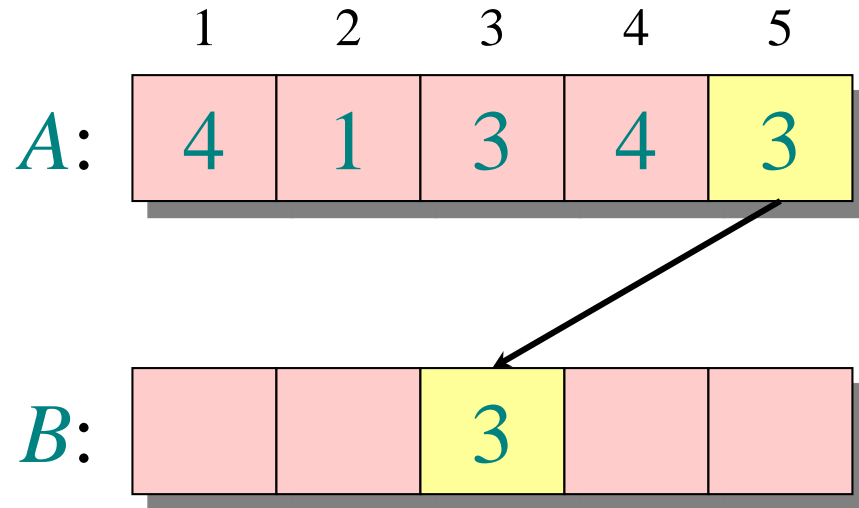
<i>C'</i> :	1	1	3	5
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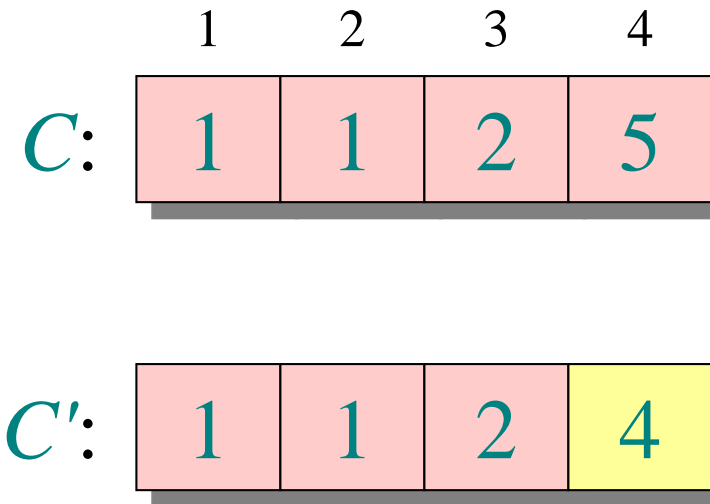
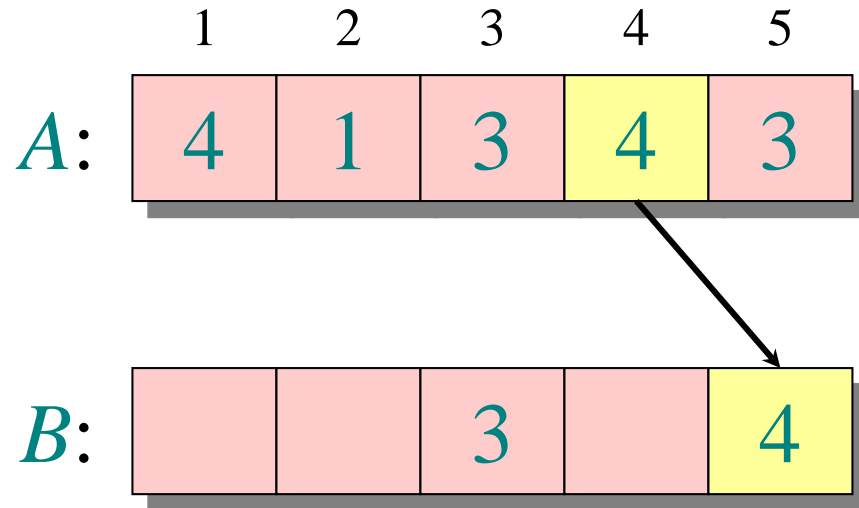
▷ $C[i] = |\{\text{key} \leq i\}|$

LOOP 4



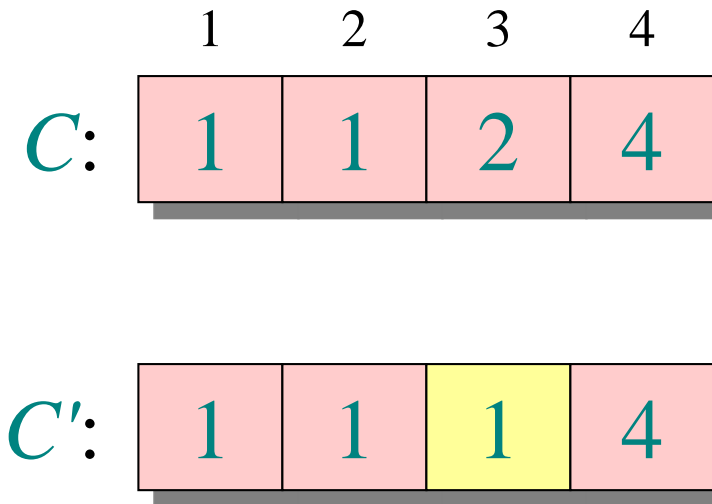
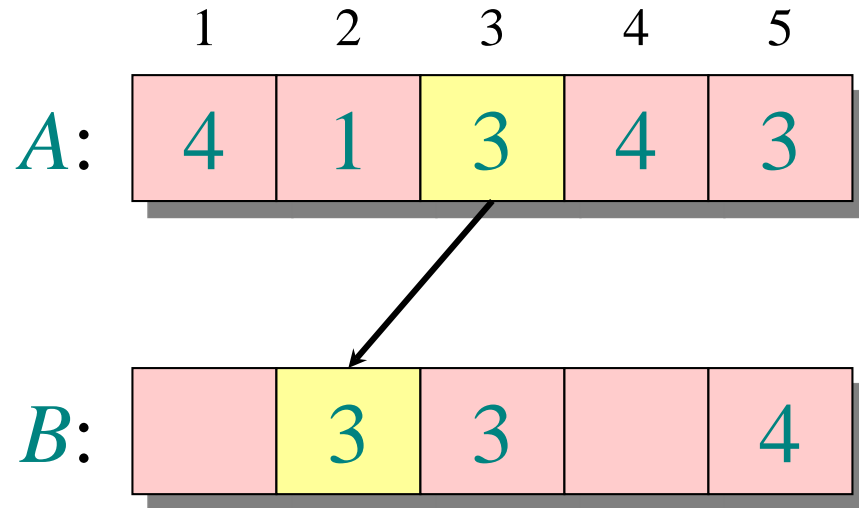
```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```


LOOP 4



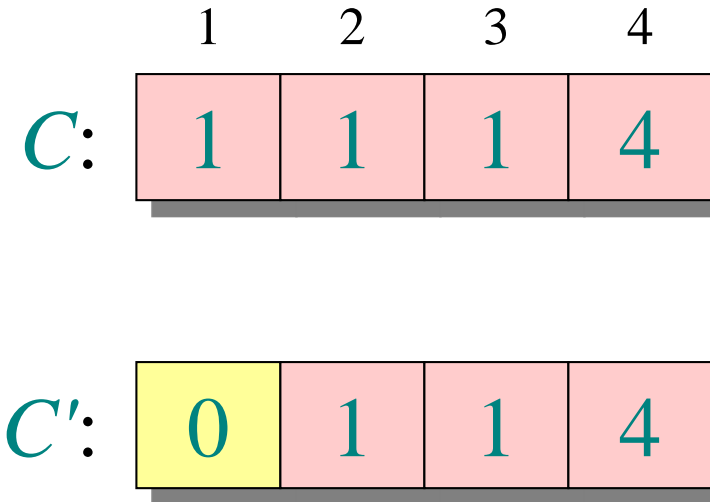
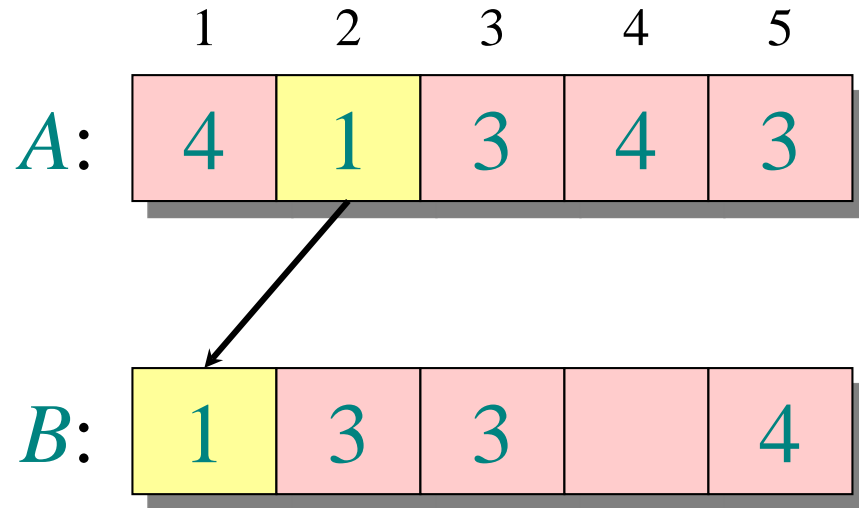
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LOOP 4



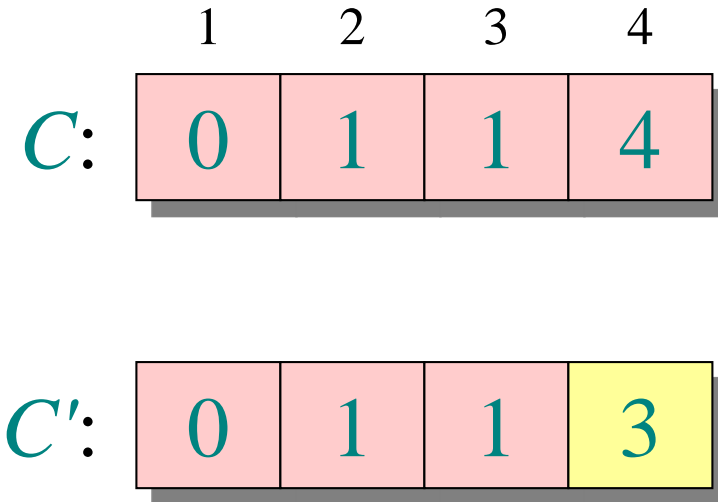
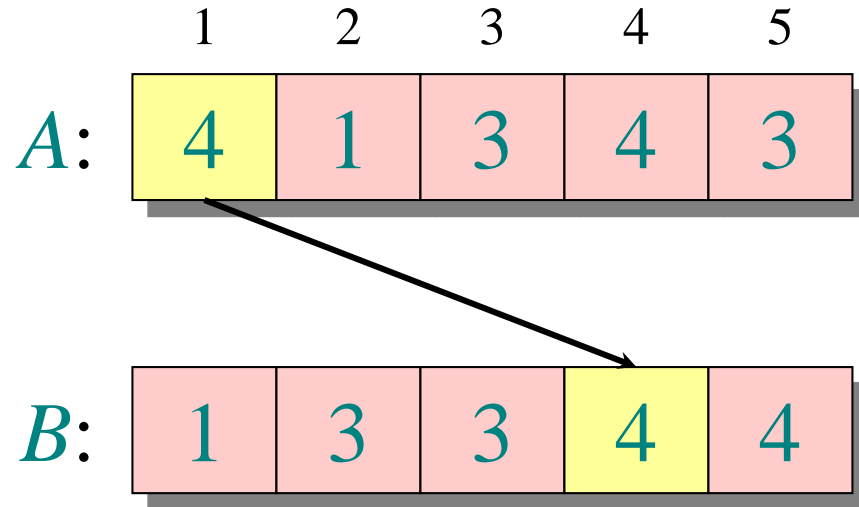
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      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```

ANALYSIS

$\Theta(k)$ { **for** $i \leftarrow 1$ **to** k
 do $C[i] \leftarrow 0$

$\Theta(n)$ { **for** $j \leftarrow 1$ **to** n
 do $C[A[j]] \leftarrow C[A[j]] + 1$

$\Theta(k)$ { **for** $i \leftarrow 2$ **to** k
 do $C[i] \leftarrow C[i] + C[i-1]$

$\Theta(n)$ { **for** $j \leftarrow n$ **downto** 1
 do $B[C[A[j]]] \leftarrow A[j]$
 $C[A[j]] \leftarrow C[A[j]] - 1$

$\Theta(n + k)$

RUNNING TIME

If $k = O(n)$, then counting sort takes $\Theta(n)$ time.

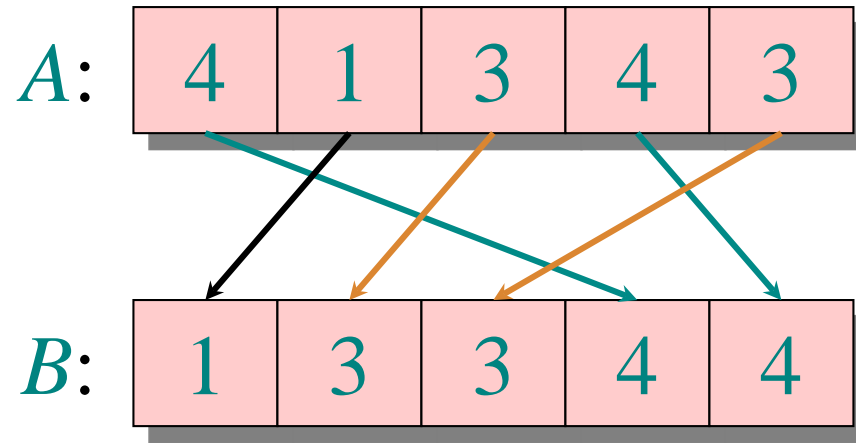
- But, sorting takes $\Omega(n \lg n)$ time!
- Where's the fallacy?

Answer:

- *Comparison sorting* takes $\Omega(n \lg n)$ time.
- Counting sort is not a *comparison sort*.
- In fact, not a single comparison between elements occurs!

STABLE SORTING

Counting sort is a *stable* sort: it preserves the input order among equal elements.



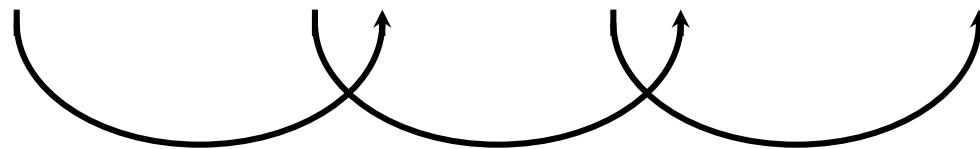
Exercise: What other sorts have this property?

RADIX SORT

- Digit-by-digit sort.
- Original idea: sort on most-significant digit first (Bad!!!).
- Good idea: Sort on *least-significant digit first* with auxiliary *stable* sort.

OPERATION OF RADIX SORT

3 2 9	7 2 0	7 2 0	3 2 9
4 5 7	3 5 5	3 2 9	3 5 5
6 5 7	4 3 6	4 3 6	4 3 6
8 3 9	4 5 7	8 3 9	4 5 7
4 3 6	6 5 7	3 5 5	6 5 7
7 2 0	3 2 9	4 5 7	7 2 0
3 5 5	8 3 9	6 5 7	8 3 9

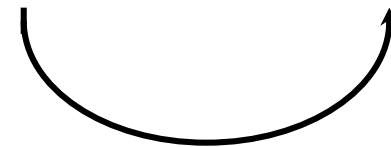


CORRECTNESS OF RADIX SORT

Induction on digit position

- Assume that the numbers are sorted by their low-order $t - 1$ digits.
- Sort on digit t

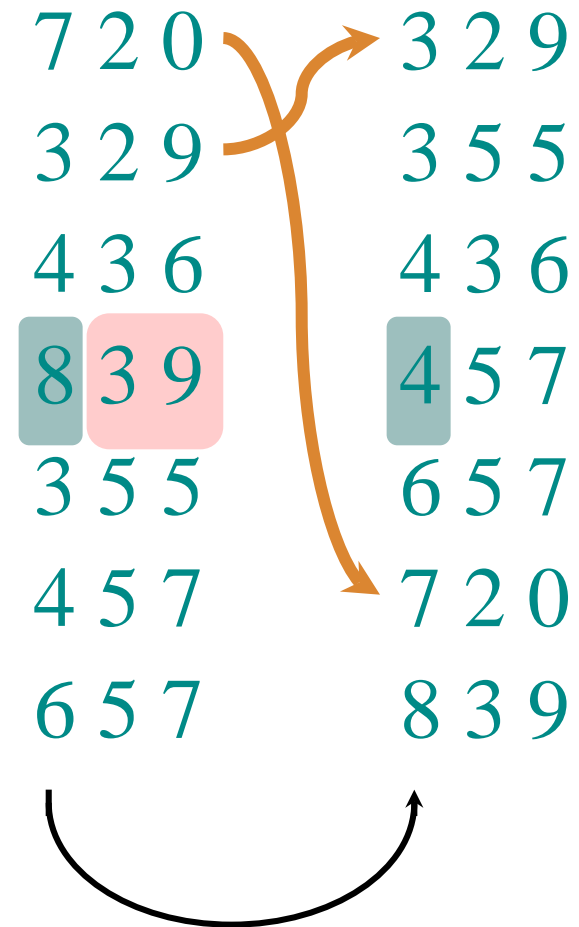
7 2 0	3 2 9
3 2 9	3 5 5
4 3 6	4 3 6
8 3 9	4 5 7
3 5 5	6 5 7
4 5 7	7 2 0
6 5 7	8 3 9



CORRECTNESS OF RADIX SORT

Induction on digit position

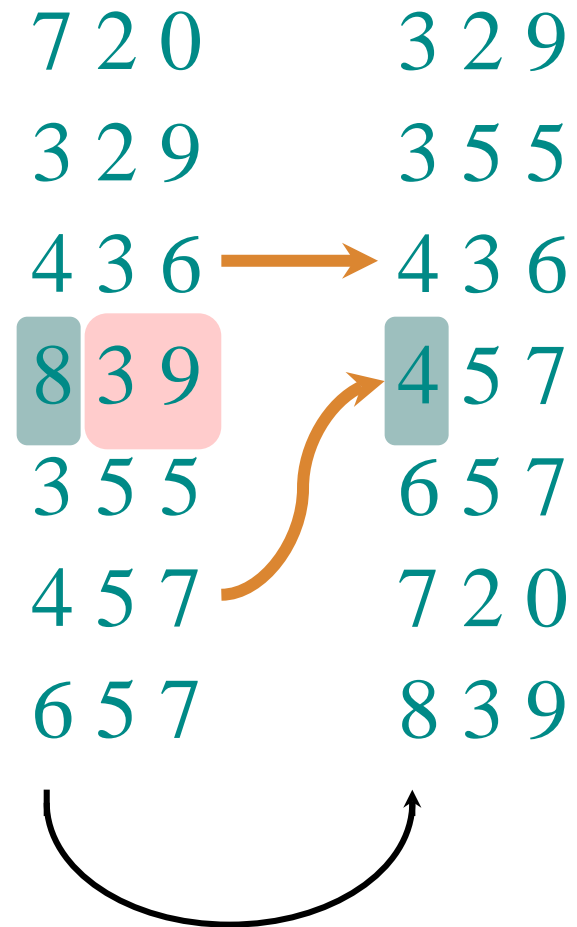
- Assume that the numbers are sorted by their low-order $t - 1$ digits.
- Sort on digit t
 - Two numbers that differ in digit t are correctly sorted.



CORRECTNESS OF RADIX SORT

Induction on digit position

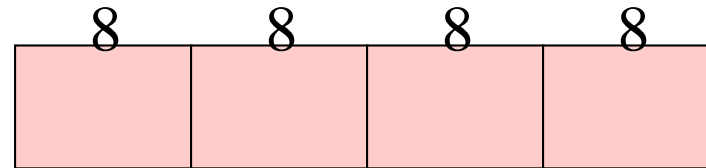
- Assume that the numbers are sorted by their low-order $t - 1$ digits.
- Sort on digit t
 - Two numbers that differ in digit t are correctly sorted.
 - Two numbers equal in digit t are put in the same order as the input \Rightarrow correct order.



ANALYSIS OF RADIX SORT

- Assume counting sort is the auxiliary stable sort.
- Sort n computer words of b bits each.
- Each word can be viewed as having b/r base- 2^r digits.

Example: 32-bit word



$r = 8 \Rightarrow b/r = 4$ passes of counting sort on base- 2^8 digits; or $r = 16 \Rightarrow b/r = 2$ passes of counting sort on base- 2^{16} digits.

How many passes should we make?

ANALYSIS (CONTINUED)

Recall: Counting sort takes $\Theta(n + k)$ time to sort n numbers in the range from 0 to $k - 1$.

If each b -bit word is broken into b/r equal pieces, each pass of counting sort takes $\Theta(n + 2^r)$ time. Since there are b/r passes, we have

$$T(n, b) = \Theta\left(\frac{b}{r}(n + 2^r)\right).$$

Choose r to minimize $T(n, b)$:

- Increasing r means fewer passes, but as $r \gg \lg n$, the time grows exponentially.

CHOOSING R

$$T(n, b) = \Theta\left(\frac{b}{r} (n + 2^r)\right)$$

Minimize $T(n, b)$ by differentiating and setting to 0.

Or, just observe that we don't want $2^r \gg n$, and there's no harm asymptotically in choosing r as large as possible subject to this constraint.

Choosing $r = \lg n$ implies $T(n, b) = \Theta(bn/\lg n)$.

- For numbers in the range from 0 to $n^d - 1$, we have $b = d \lg n \Rightarrow$ radix sort runs in $\Theta(dn)$ time.

BUCKET SORT

A 1 .78

2 .17

3 .39

4 .26

5 .72

6 .94

7 .21

8 .12

9 .23

10 .68

B 0 /

1 →

.12	—
-----	---

 →

.17	/
-----	---

2 →

.21	—
-----	---

 →

.23	—
-----	---

 →

.26	/
-----	---

3 →

.39	/
-----	---

4 /

5 /

6 →

.68	/
-----	---

7 →

.72	—
-----	---

 →

.78	/
-----	---

8 /

9 →

.94	/
-----	---

BUCKET SORT

Idea :

- Divide the interval $[0, n)$ into n equal – sized subintervals or buckets.
- Distribute the n input numbers into the buckets.

Since the inputs are assumed to be uniformly distributed over $[0,1)$, many numbers don't fall into each bucket.

To produce the output , simply sort the numbers in each bucket and then go through the buckets , in order , listing the elements in each.

Pseudocode for Bucket Code

Bucket Sort (A)

1. $n \leftarrow \text{length}(A)$
2. for $i \leftarrow 1$ to n
3. do insert $A[i]$ into list $B[\lfloor nA[i] \rfloor]$
4. for $i \leftarrow 0$ to $n-1$
5. do sort list $B[i]$ with insertion sort
6. Concatenate the list $B[0] \dots\dots\dots B[n-1]$ together in order .

ANALYSIS OF RUNNING TIME

- Observe that all lines except line 5 takes $O(n)$ time in worst case.
- We need to balanced that the total time taken by n calls to intersection sort in line 5

Let n_i be the random variables denoting the number of elements placed in bucket $B[i]$

So the running time of bucket sort is

$$T(n) = \Theta(n) + \sum_{i=0}^{n-1} O(n_i^2) .$$

ANALYSIS CONT.

Taking expectations of both sides and using linearity of expectation, we have

$$\begin{aligned} \mathbb{E}[T(n)] &= \mathbb{E}\left[\Theta(n) + \sum_{i=0}^{n-1} O(n_i^2)\right] \\ &= \Theta(n) + \sum_{i=0}^{n-1} \mathbb{E}[O(n_i^2)] \quad (\text{by linearity of expectation}) \\ &= \Theta(n) + \sum_{i=0}^{n-1} O(\mathbb{E}[n_i^2]) \quad \dots\dots\dots (1) \end{aligned}$$

ANALYSIS CONT.

we claim that

$$E(n_i^2) = 2 - (1/n) \dots\dots\dots (2)$$

Define

$$X_{ij} = I\{A[j] \text{ falls in bucket } i\}$$

for $i=0,1,\dots,n-1$, $j=1,2,\dots,n$

$$n_i = \sum_{j=1}^n X_{ij}$$

ANALYSIS CONT.

To compute $E[n_i^2]$, we expand the square and regroup terms:

$$\begin{aligned} E[n_i^2] &= E\left[\left(\sum_{j=1}^n X_{ij}\right)^2\right] \\ &= E\left[\sum_{j=1}^n \sum_{k=1}^n X_{ij} X_{ik}\right] \\ &= E\left[\sum_{j=1}^n X_{ij}^2 + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} X_{ij} X_{ik}\right] \\ &= \sum_{j=1}^n E[X_{ij}^2] + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} E[X_{ij} X_{ik}] , \end{aligned}$$

ANALYSIS CONT.

As, Indicator variables X_{ij} is 1 with probability $1/n$ and 0 otherwise

$$\begin{aligned}\text{SO } E[X_{ij}^2] &= 1 \cdot \frac{1}{n} + 0 \cdot \left(1 - \frac{1}{n}\right) \\ &= \frac{1}{n}.\end{aligned}$$

When $k \neq j$, the variables X_{ij} and X_{ik} are independent, and hence

$$\begin{aligned}E[X_{ij}X_{ik}] &= E[X_{ij}]E[X_{ik}] \\ &= \frac{1}{n} \cdot \frac{1}{n} \\ &= \frac{1}{n^2}.\end{aligned}$$

Substituting these two expected values in equation (8.3), we obtain

$$\begin{aligned}E[n_i^2] &= \sum_{j=1}^n \frac{1}{n} + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} \frac{1}{n^2} \\ &= n \cdot \frac{1}{n} + n(n-1) \cdot \frac{1}{n^2} \\ &= 1 + \frac{n-1}{n} \\ &= 2 - \frac{1}{n}, \quad \text{which proves} \\ &\quad (2)\end{aligned}$$

ANALYSIS CONT.

Using the expected value in (1)

we can say that the running time of bucket sort is expected to be

$$T(n) = \Theta(n) + n.O(2^{-(1/n)}) = \Theta(n)$$

thus, the entire bucket algorithm runs in *linear* expected time.

