

Counting Sort

- Assumes each element is an integer in the range 0 to k for some integer k
- When $k = O(n)$, the sort runs in $\Theta(n)$ time
- Determines for each element x the number of elements less than x
 - Use this to place x directly in its position
- Requires extra storage

k must be reasonably sized since we need an extra array of size $k+1$. We also need an extra array of size n

COUNTING-SORT(A, B, n, k)

let $C[0..k]$ be a new array

for $i = 0$ to k
 $C[i] = 0$ $\Theta(k)$

for $j = 1$ to n

$C[A[j]] = C[A[j]] + 1$ $\Theta(n)$

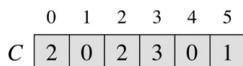
for $i = 1$ to k

$C[i] = C[i] + C[i-1]$ $\Theta(k)$

for $j = n$ downto 1

$B[C[A[j]]] = A[j]$ $\Theta(n)$

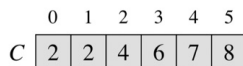
$C[A[j]] = C[A[j]] - 1$



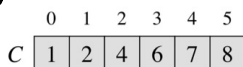
(a)



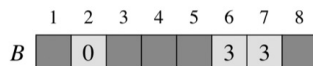
(b)



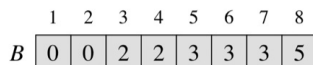
(c)



(d)



(e)



(f)

$$T(n) = \Theta(k) + \Theta(n) + \Theta(k) + \Theta(n) = \Theta(n + k)$$

if $k = O(n)$ then $T(n) = O(n)$

Sorting Stability

- A sort is stable if the items with the same value appear in the same order in the output as they do in the input
- Stability matters when the item's value is a key and there is additional data associated with the item (satellite data)

Radix Sort

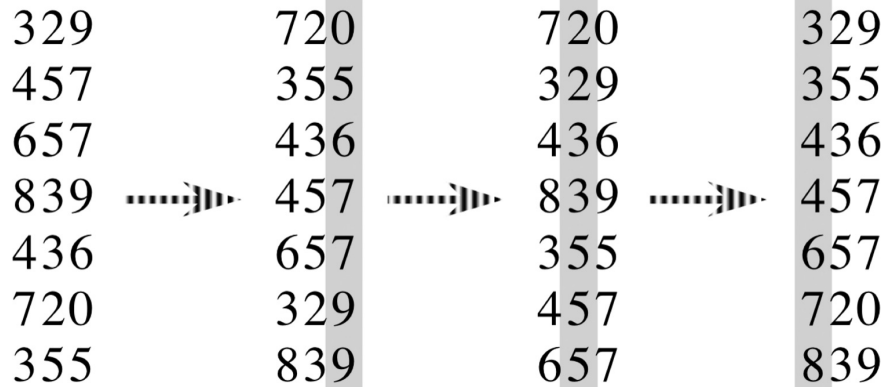
- Works for items with a fixed number of digits or characters or fields
- Starting with the least significant digit, sort the items by that digit using a stable sort, then work your way through the rest of the digits in increasing significance

RADIX-SORT(A, d)

for $i = 1$ to d

use a stable sort to sort array A on digit i

digit 1 is least significant
" 2 is most significant



Lemma 8.4 in the book

Given n d -digit numbers in which each digit can take on up to k possible values, Radix Sort correctly sorts these numbers in $\Theta(d(n+k))$ if using a $\Theta(n+k)$ algorithm like counting sort to sort the digits

If d is constant and $k = O(n)$, radix sort is $\Theta(n)$ time

- Works for unsigned binary numbers with a fixed number of bits

- "digits" can be bits or bytes

- Requires overhead since counting sort is not in-place

- Quicksort is still often faster

Bucket Sort

- Assumes input is drawn from a uniform distribution
- Each element is from the interval $[0, 1)$ (could be 0 but not 1)
- To sort n items, divide the interval into n equal sized buckets
- Distribute the n items into the buckets
- Sort each bucket
- Each bucket will have very few elements (1 on average)

BUCKET-SORT(A, n)

let $B[0..n-1]$ be a new array

for $i = 0$ **to** $n - 1$

 make $B[i]$ an empty list

for $i = 1$ **to** n

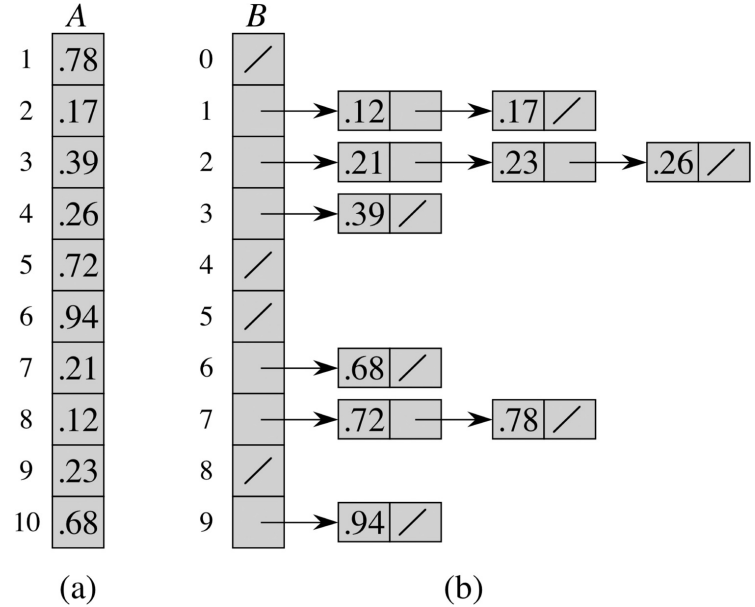
 insert $A[i]$ into list $B[\lfloor n \cdot A[i] \rfloor]$

for $i = 0$ **to** $n - 1$

 sort list $B[i]$ with insertion sort

concatenate lists $B[0], B[1], \dots, B[n-1]$ together in order

return the concatenated lists



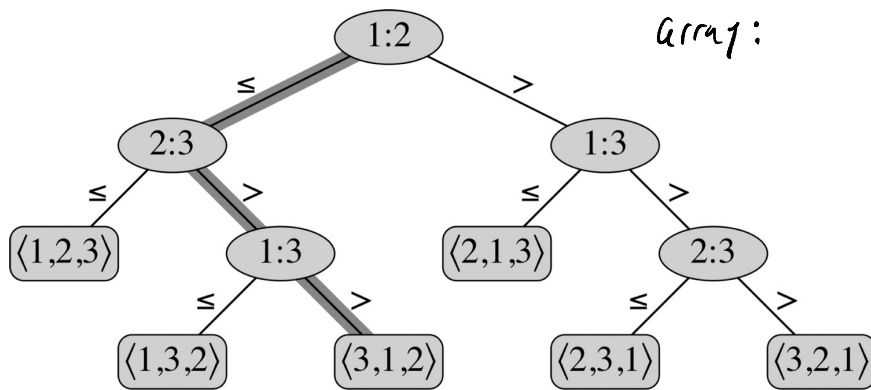
- Insertion sort is quadratic ($\Theta(n^2)$), but we expect that the sum of squares of the bucket sizes is linear in the number of elements
- Average case is $\Theta(n)$
- See the book for proof

- Previous sorts we looked at were comparison sorts

- insertion, merge, heap, quick

- The worst case running time for any comparison sort is $\Omega(n \lg n)$

Decision tree for arriving at all possible permutations of a 3-element



There are $n!$ permutations of an n -element list

This is a binary tree, so height $h \geq \lg(n!)$

$$h = \Omega(n \lg n) \quad (\text{See equation 3.19 in the book})$$

A comparison sorting algorithm that is $\Theta(n \lg n)$ time
is asymptotically optimal