

## Lecture 9: Linear Sorting

### Review

- Comparison search lower bound: any decision tree with  $n + 1$  leaves has height  $\geq \lceil \lg n \rceil - 1$
- Can do faster using random access indexing: an operation with linear branching factor!
- **Direct access array** is fast, but may use a lot of space ( $\Theta(u)$ )
- Solve space problem by mapping (**hashing**) key space  $u$  down to  $m = \Theta(n)$
- **Hash tables** give **expected**  $O(1)$  time operations, **amortized** if dynamic
- Expectation input-independent: choose hash function randomly from **universal** hash family
- Data structure overview!
- Last time we achieved faster find. Can we also achieve faster sort?

Set Interface  Data Structure Implementation	Operation, Worst Case $O(\cdot)$						Space  $\sim \times n$
	Set	Dynamic (D)		Order (O)		D + O	
	find(k)	insert(x)	delete(k)	find_ next(k)	find_ max()	delete_ max()	
Unsorted Array	$n$	$n$	$n$	$n$	$n$	$n$	1
Linked List	$n$	1	$n$	$n$	$n$	$n$	3
Dynamic Array	$n$	$1_{(a)}$	$n$	$n$	$n$	$n$	4
Sorted Array	$\lg n$	$n$	$n$	$\lg n$	1	$n$	1
Max-Heap	$n$	$\lg n_{(a)}$	$n$	$n$	1	$\lg n$	1
Balanced BST	$\lg n$	$\lg n$	$\lg n$	$\lg n$	(1)	$\lg n$	5
Direct Access	1	1	1	$u$	$u$	$u$	$u/n$
Hash Table	$1_{(e)}$	$1_{(e,a)}$	$1_{(e,a)}$	$n$	$n$	$n$	4

## Comparison Sort Lower Bound

- Comparison model implies that algorithm decision tree is binary (constant branching factor)
  - Requires # leaves  $L \geq \#$  possible outputs
  - Tree height lower bounded by  $\Omega(\log L)$ , so worst-case running time is  $\Omega(\log L)$
  - To sort array of  $n$  elements, # outputs is  $n!$  permutations
  - Thus height lower bounded by  $\log(n!) \geq \log((n/2)^{n/2}) = \Omega(n \log n)$
  - So merge sort, heap sort, AVL sort are all optimal in comparison model
  - Can we exploit a direct access array to sort faster?
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## Direct Access Array Sort

- **Example:** [5, 2, 7, 0, 4]
- Suppose all keys are **unique** positive integers in range  $\{0, \dots, u-1\}$ , so  $n \leq u$
- Insert each item into a direct access array with size  $u$  in  $\Theta(n)$
- Return items in order they appear in direct access array in  $\Theta(u)$
- Running time is  $\Theta(u)$ , which is  $\Theta(n)$  if  $u = \Theta(n)$ . Yay!

```

1 def direct_access_sort(A):
2     "Sort A assuming items have distinct non-negative keys"
3     u = 1 + max([x.key for x in A])    # O(n) find maximum key
4     D = [None] * u                    # O(u) direct access array
5     for x in A:                        # O(n) insert items
6         D[x.key] = x
7     i = 0
8     for key in range(u):                # O(u) read out items in order
9         if D[key] is not None:
10             A[i] = D[key]
11             i += 1

```

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- What if keys are in larger range, like  $u = \Omega(n^2) < n^2$ ?
- **Idea!** Represent each key  $k$  by tuple  $(a, b)$  where  $k = an + b$  and  $0 \leq b < n$
- Specifically  $a = \lfloor k/n \rfloor < n$  and  $b = (k \bmod n)$  (just a 2-digit base- $n$  number!)
- This is a built-in Python operation  $(a, b) = \text{divmod}(k, n)$
- **Example:** [17, 3, 24, 23, 12]  $\implies$  [(3,2), (0,3), (4,4), (4,2), (2,2)]
- How can we sort tuples?

## Tuple Sort

- Item keys are tuples of equal length, i.e. item  $x.key = (x.k_1, x.k_2, x.k_2, \dots)$ .
- Want to sort on all entries **lexicographically**, so first key  $k_1$  is most significant
- How to sort? **Idea!** Use other **auxiliary sorting algorithms** to separately sort each key
- (Like sorting rows in a spreadsheet by multiple columns)
- What order to sort them in? Least significant to most significant!
- **Exercise:**  $[32, 03, 44, 42, 22] \implies [42, 22, 32, 03, 44] \implies [03, 22, 32, 42, 44]$

- **Idea!** Use tuple sort with **auxiliary direct access array sort** to sort tuples  $(a, b)$ .
- **Wrong!** Many integers could have the same  $a$  or  $b$  value, even if input keys distinct
- Need sort allowing **repeated keys** which preserves input orders
- Want sort to be **stable**: repeated keys appear in output in same order as input
- Direct access array sort cannot even sort arrays having repeated keys!
- Can we modify direct access array sort to admit multiple keys in a way that is stable?

## Counting Sort

- Instead of storing a single item at each array index, store a chain, just like hashing!
- For stability, chain data structure should remember the order in which items were added
- Use a **sequence** data structure which maintains insertion order (specifically, a **queue**)
- To insert item  $x$ , add to the end of the chain at index  $x.key$
- Then to sort, read through all chains in sequence order, returning items one by one

```

1 def counting_sort(A):
2     "Sort A assuming items have non-negative keys"
3     u = 1 + max([x.key for x in A])    # O(n) find maximum key
4     D = [[] for i in range(u)]         # O(u) direct access array of chains
5     for x in A:                         # O(n) insert into chain at x.key
6         D[x.key].append(x)
7     i = 0
8     for chain in D:                     # O(u) read out items in order
9         for x in chain:
10             A[i] = x
11             i += 1

```

## Radix Sort

- Use tuple sort with **auxiliary counting sort** to sort tuples  $(a, b)$
- Sort least significant key  $b$ , then most significant key  $a$
- Stability ensures previous sorts stay sorted
- Running time for this algorithm is  $O(2n) = O(n)$ . Yay!
- If every key  $< n^c$  for some positive  $c$ , each key has  $c$  digits when written in base  $n$
- A  $c$ -digit number can be written as a  $c$ -element tuple
- So tuple sort with **auxiliary counting sort** runs in  $O(cn)$  time
- If  $c$  is constant, and each key is  $\leq n^c$ , this sort is linear  $O(n)$ !

```

1 def radix_sort(A):
2     "Sort A assuming items have non-negative keys"
3     n = len(A)
4     u = 1 + max([x.key for x in A])          # O(n) find maximum key
5     c = 1 + (u.bit_length() // n.bit_length())
6     class Obj: pass
7     D = [Obj() for a in A]
8     for i in range(n):                       # O(nc) make digit tuples
9         D[i].digits = []
10        D[i].item = A[i]
11        high = A[i].key
12        for j in range(c):                   # O(c) make digit tuple
13            high, low = divmod(high, n)
14            D[i].digits.append(low)
15    for i in range(c):                       # O(nc) sort each digit
16        for j in range(n):                 # O(n) assign key i to tuples
17            D[j].key = D[j].digits[i]
18        counting_sort(D)                   # O(n) sort on digit i
19    for i in range(n):                       # O(n) output to A
20        A[i] = D[i].item

```

Algorithm	Time $O(\cdot)$	In-place?	Stable?	Comments
Insertion Sort	$n^2$	Y	Y	$O(nk)$ for $k$ -proximate
Selection Sort	$n^2$	Y	N	$O(n)$ swaps
Merge Sort	$n \lg n$	N	Y	stable, optimal comparison
Heap Sort	$n \lg n$	Y	N	low space, optimal comparison
AVL Sort	$n \lg n$	N	Y	good if also need dynamic
Counting Sort	$n$	N	Y	$u = \Theta(n)$ is domain of possible keys
Radix Sort	$cn$	N	Y	$u = \Theta(n^c)$ is domain of possible keys