## Lecture 3: Sorting

## Set Interface (L03-L08)

| Container | build $(X)$ <br> len () | given an iterable $x$, build set from items in $X$ <br> return the number of stored items |
| :--- | :--- | :--- |
| Static | find $(k)$ | return the stored item with key $k$ |
| Dynamic | insert $(x)$ <br> delete $(k)$ | add $x$ to set (replace item with key $x . k e y ~ i f ~ o n e ~ a l r e a d y ~ e x i s t s) ~$ <br> remove and return the stored item with key $k$ |
| Order | iter_ord () <br> find_min () <br> find_max () <br> find_next $(k)$ <br> find_prev $(k)$ | return the stored items one-by-one in key order <br> return the stored item with smallest key <br> return the stored item with largest key <br> return the stored item with smallest key larger than $k$ <br> return the stored item with largest key smaller than $k$ |

- Storing items in an array in arbitrary order can implement a (not so efficient) set
- Stored items sorted increasing by key allows:
- faster find $\mathrm{min} / \mathrm{max}$ (at first and last index of array)
- faster finds via binary search: $O(\log n)$

| Set <br> Data Structure | Operations $O(\cdot)$ |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Container | Static | Dynamic |  |  |
|  | build(X) | find (k) | insert(x) <br> delete (k) | $\begin{aligned} & \text { find_min() } \\ & \text { find_max() } \end{aligned}$ | $\begin{aligned} & \text { find_prev (k) } \\ & \text { find_next (k) } \end{aligned}$ |
| Array | $n$ | $n$ | $n$ | $n$ | $n$ |
| Sorted Array | $n \log n$ | $\log n$ | $n$ | 1 | $\log n$ |

- But how to construct a sorted array efficiently?


## Sorting

- Given a sorted array, we can leverage binary search to make an efficient set data structure.
- Input: (static) array $A$ of $n$ numbers
- Output: (static) array $B$ which is a sorted permutation of $A$
- Permutation: array with same elements in a different order
- Sorted: $B[i-1] \leq B[i]$ for all $i \in\{1, \ldots, n\}$
- Example: $[8,2,4,9,3] \rightarrow[2,3,4,8,9]$
- A sort is destructive if it overwrites $A$ (instead of making a new array $B$ that is a sorted version of $A$ )
- A sort is in place if it uses $O(1)$ extra space (implies destructive: in place $\subseteq$ destructive)


## Permutation Sort

- There are $n$ ! permutations of $A$, at least one of which is sorted
- For each permutation, check whether sorted in $\Theta(n)$
- Example: $[2,3,1] \rightarrow\{[1,2,3],[1,3,2],[2,1,3],[2,3,1],[3,1,2],[3,2,1]\}$

```
def permutation_sort(A):
    '''Sort A'''
    for B in permutations(A): # O(n!)
        if is_sorted(B): # O(n)
            return B # O(1)
```

- permutation_sort analysis:
- Correct by case analysis: try all possibilities (Brute Force)
- Running time: $\Omega(n!\cdot n)$ which is exponential :(


## Solving Recurrences

- Substitution: Guess a solution, replace with representative function, recurrence holds true
- Recurrence Tree: Draw a tree representing the recursive calls and sum computation at nodes
- Master Theorem: A formula to solve many recurrences (R03)


## Selection Sort

- Find a largest number in prefix $\mathrm{A}[: \mathrm{i}+1]$ and swap it to $\mathrm{A}[\mathrm{i}]$
- Recursively sort prefix A [: i]
- Example: $[8,2,4,9,3],[8,2,4,3,9],[3,2,4,8,9],[3,2,4,8,9],[2,3,4,8,9]$

```
def selection_sort(A, i = None): # T(i)
    '''Sort A[:i + 1]'''
        if i is None: i = len(A) - 1 # O(1)
        if i > 0: # O(1)
            j = prefix_max(A, i) # S(i)
            A[i], A[j] = A[j], A[i] # O(1)
            selection_sort(A, i - 1) # T(i - 1)
def prefix_max(A, i): # S(i)
        '''Return index of maximum in A[:i + 1]'''
        if i > 0: # O(1)
            j = prefix_max(A, i - 1) # S(i - 1)
            if A[i] < A[j]: # O(1)
                return j # O(1)
    return i # O(1)
```

- prefix_max analysis:
- Base case: for $i=0$, array has one element, so index of max is $i$
- Induction: assume correct for $i$, maximum is either the maximum of A[:i] or A[i], returns correct index in either case.
- $S(1)=\Theta(1), S(n)=S(n-1)+\Theta(1)$
* Substitution: $S(n)=\Theta(n), \quad c n=\Theta(1)+c(n-1) \Longrightarrow 1=\Theta(1)$
* Recurrence tree: chain of $n$ nodes with $\Theta(1)$ work per node, $\sum_{i=0}^{n-1} 1=\Theta(n)$
- selection_sort analysis:
- Base case: for $i=0$, array has one element so is sorted
- Induction: assume correct for $i$, last number of a sorted output is a largest number of the array, and the algorithm puts one there; then $\mathrm{A}[$ : i] is sorted by induction
- $T(1)=\Theta(1), T(n)=T(n-1)+\Theta(n)$
* Substitution: $T(n)=\Theta\left(n^{2}\right), \quad c n^{2}=\Theta(n)+c(n-1)^{2} \Longrightarrow c(2 n-1)=\Theta(n)$
* Recurrence tree: chain of $n$ nodes with $\Theta(i)$ work per node, $\sum_{i=0}^{n-1} i=\Theta\left(n^{2}\right)$


## Insertion Sort

- Recursively sort prefix A[:i]
- Sort prefix A[:i + 1] assuming that prefix A[:i] is sorted by repeated swaps
- Example: $[8,2,4,9,3],[2,8,4,9,3],[2,4,8,9,3],[2,4,8,9,3],[2,3,4,8,9]$

```
def insertion_sort(A, i = None): # T(i)
        'r'Sort A[:i + 1]'r'
        if i is None: i = len(A) - 1 # O(1)
        if i > 0: # O(1)
            insertion_sort(A, i - 1) # T(i - 1)
            insert_last(A, i) # S(i)
def insert_last(A, i): # S(i)
        '''Sort A[:i + 1] assuming sorted A[:i]'''
        if i > O and A[i] < A[i - 1]: # O(1)
            A[i], A[i - 1] = A[i - 1], A[i] # O(1)
            insert_last(A, i - 1) # S(i - 1)
```

- insert_last analysis:
- Base case: for $i=0$, array has one element so is sorted
- Induction: assume correct for $i$, if A[i] >= A[i - 1], array is sorted; otherwise, swapping last two elements allows us to sort $\mathrm{A}[$ : i] by induction
- $S(1)=\Theta(1), S(n)=S(n-1)+\Theta(1) \Longrightarrow S(n)=\Theta(n)$
- insertion_sort analysis:
- Base case: for $i=0$, array has one element so is sorted
- Induction: assume correct for $i$, algorithm sorts A [: i] by induction, and then insert_ last correctly sorts the rest as proved above
- $T(1)=\Theta(1), T(n)=T(n-1)+\Theta(n) \Longrightarrow T(n)=\Theta\left(n^{2}\right)$


## Merge Sort

- Recursively sort first half and second half (may assume power of two)
- Merge sorted halves into one sorted list (two finger algorithm)
- Example: $[7,1,5,6,2,4,9,3],[1,7,5,6,2,4,3,9],[1,5,6,7,2,3,4,9],[1,2,3,4,5,6,7,9]$

```
def merge_sort(A, a = 0, b = None): # T (b - a = n)
    '''Sort A[a:b]'''
    if b is None: b = len(A) # O(1)
    if 1 < b - a: # O(1)
            c=(a+b + 1) // 2 # O(1)
            merge_sort(A, a, c) # T(n / 2)
            merge_sort(A, c, b) # T(n / 2)
            L, R = A[a:c], A[c:b] # O(n)
            merge(L, R, A, len(L), len(R), a, b) # S(n)
def merge(L, R, A, i, j, a, b): # S(b - a = n)
    '''Merge sorted L[:i] and R[:j] into A[a:b]'''
    if a < b: # O(1)
        if (j <= 0) or (i > 0 and L[i - 1] > R[j - 1]): # O(1)
            A[b - 1] = L[i - 1] # O(1)
                i = i - 1 # O(1)
            else: # O(1)
            A[b - 1] = R[j - 1] # O(1)
            j = j - 1 # O(1)
            merge(L, R, A, i, j, a, b - 1) # S(n - 1)
```

- merge analysis:
- Base case: for $n=0$, arrays are empty, so vacuously correct
- Induction: assume correct for $n$, item in $\mathrm{A}[\mathrm{r}$ ] must be a largest number from remaining prefixes of $L$ and $R$, and since they are sorted, taking largest of last items suffices; remainder is merged by induction
- $S(0)=\Theta(1), S(n)=S(n-1)+\Theta(1) \Longrightarrow S(n)=\Theta(n)$
- merge_sort analysis:
- Base case: for $n=1$, array has one element so is sorted
- Induction: assume correct for $k<n$, algorithm sorts smaller halves by induction, and then merge merges into a sorted array as proved above.
- $T(1)=\Theta(1), T(n)=2 T(n / 2)+\Theta(n)$
* Substitution: Guess $T(n)=\Theta(n \log n)$ $c n \log n=\Theta(n)+2 c(n / 2) \log (n / 2) \Longrightarrow c n \log (2)=\Theta(n)$
* Recurrence Tree: complete binary tree with depth $\log _{2} n$ and $n$ leaves, level $i$ has $2^{i}$ nodes with $O\left(n / 2^{i}\right)$ work each, total: $\sum_{i=0}^{\log _{2} n}\left(2^{i}\right)\left(n / 2^{i}\right)=\sum_{i=0}^{\log _{2} n} n=\Theta(n \log n)$

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Spring 2020
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