CS 457, Fall 2019

Drexel University, Department of Computer Science

Lecture 10

Midterm

- In-class midterm on Wednesday
- You can bring a handwritten cheat-sheet
- Tested material can be from anything that we covered so far:
 - Chapters: 1, 2, 3, 4 (except 4.2 and 4.6), 5 (except 5.3), 7, 8 (only 8.4), 9, and 11 (only 11.1 and 11.2)
 - You can disregard the formal proofs of correctness that appear in some of these sections
- Preparation:
 - All of the homework problems and solutions closely
 - All of the problems discussed during the recitation
 - Other problems from the textbook

Sorting in Linear Time

- Assume that the input is drawn from uniform distribution from interval [0, 1)
- Can we use this information to get a faster algorithm, in the worst case sense?
- Can we use this information to get a faster algorithm, in the average case sense?

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BUCKET-SORT (A)

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

2 n = A.length

3 for i = 0 to n - 1

4 make B[i] an empty list

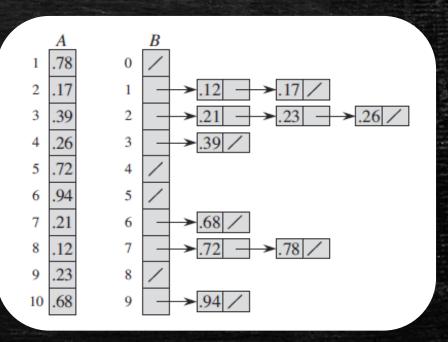
5 for i = 1 to n

6 insert A[i] into list B[\lfloor nA[i] \rfloor]

7 for i = 0 to n - 1

8 sort list B[i] with insertion sort

9 concatenate the lists B[0], B[1], \ldots, B[n-1] together in order
```



- Let n_i be the number of elements in B[i] (random variable)
- What is the running time, T(n) of bucket sort as a function of n_i ?

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

• If we let $X_{ij} = \mathbb{I}\{A[j] \text{ falls in bucket } i\}$, then $n_i = \sum_{j=1}^n X_{ij}$

$$E[T(n)] = E\left[\Theta(n) + \sum_{i=0}^{n-1} O(n_i^2)\right]$$

$$= \Theta(n) + \sum_{i=0}^{n-1} E\left[O(n_i^2)\right] \text{ (by linearity of expectation)}$$

$$= \Theta(n) + \sum_{i=0}^{n-1} O\left(E\left[n_i^2\right]\right) \text{ (by equation (C.22))}.$$

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- What is the running time, T(n) of bucket sort as a function of n_i ?
 - $T(n) = \Theta(n) + \sum_{i=0}^{n-1} O(n_i^2)$
- If we let $X_{ij}=\mathbb{I}\{A[j] \text{ falls in bucket } i\}$, then $n_i=\sum_{j=1}^n X_{ij}$

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- So, it suffices to show that $\sum_{i=0}^{n-1} O(\mathrm{E}[n_i^2])$ is $\Theta(n)$
- What is the value of $E[n_i^2]$?

$$E[n_{i}^{2}] = E\left[\left(\sum_{j=1}^{n} X_{ij}\right)^{2}\right]$$

$$= E\left[\sum_{j=1}^{n} \sum_{k=1}^{n} X_{ij} X_{ik}\right]$$

$$= E\left[\sum_{j=1}^{n} X_{ij}^{2} + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} X_{ij} X_{ik}\right]$$

$$= \sum_{j=1}^{n} E[X_{ij}^{2}] + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} E[X_{ij} X_{ik}],$$

- So, it suffices to show that $\sum_{i=0}^{n-1} O(\mathrm{E}[n_i^2])$ is $\Theta(n)$
- What is the value of $\mathrm{E}[n_i^2]$?

$$E[n_{i}^{2}] = E\left[\left(\sum_{j=1}^{n} X_{ij}\right)^{2}\right]$$

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$$= \sum_{j=1}^{n} E[X_{ij}^{2}] + \sum_{1 \leq j \leq n} \sum_{\substack{1 \leq k \leq n \\ k \neq j}} E[X_{ij} X_{ik}],$$

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

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

Simple Indicator Variables Example

Randomized-Loops (n)

```
1. c=0

2. for x=1 to n

3. k=\operatorname{Random}(1,n) // Random number from 1 to n

4. for y=1 to k

5. c=c+1

6. return c
```

- What is the expected running time of Randomized-Loops(n)?
 - Let $X_{ij} = \mathbb{I}\{\text{The value of } k \text{ in the } i\text{--th outer loop iteration is equal to } j\}$
 - Then we wish to compute expected value of $X = \sum_{i=1}^n \sum_{j=1}^n j X_{ij}$
 - Taking the expectation, we get

$$\mathbf{E}[X] = \mathbf{E}\left[\sum_{i=1}^{n} \sum_{j=1}^{n} j \ X_{ij}\right] = \sum_{i=1}^{n} \sum_{j=1}^{n} j \ \mathbf{E}[X_{ij}] = \sum_{i=1}^{n} \sum_{j=1}^{n} \frac{j}{n} = \sum_{i=1}^{n} \frac{n+1}{2} = \frac{n(n+1)}{2}$$

Today's Lecture

- Binary trees
 - Heaps and search trees
 - Structural induction

Binary Heaps

What is a binary heap? How is it stored in memory?

Heap data structure:

```
PARENT (i)

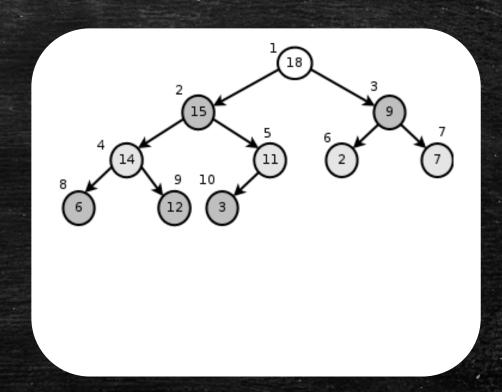
return \lfloor i/2 \rfloor

LEFT (i)

return 2i

RIGHT (i)

return 2i + 1
```



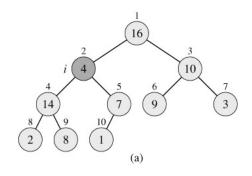
Max-heap property: $A[PARENT(i)] \ge A[i]$

Heapsort

Max-Heapify (A, i)

```
l = left(i)
         r = right(i)
2.
         if l \le A.heap-size and A[l] > A[i]
3.
                   largest = l
4.
         else largest = i
5.
         if r \le A.heap-size and A[r] > A[largest]
6.
                    largest = r
7.
8.
         if largest \neq i
                   exchange A[i] with A[largest]
9.
                   Max-Heapify (A, largest)
10.
```

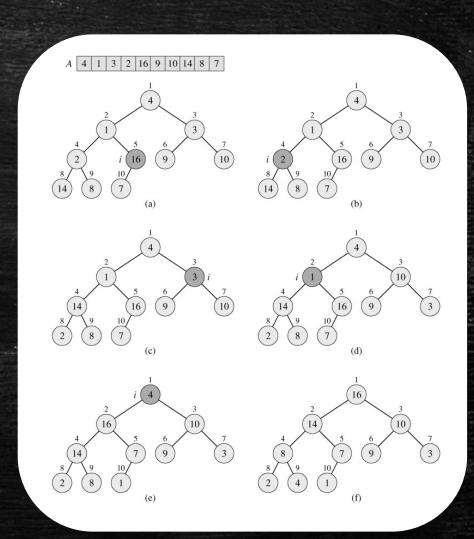
Call to Max-Heapify (A,2)



Heapsort

Build-Max-Heap (A)

- 1. A.heap-size = A.length
- for i = [A.length/2] down to 1
- 3. Max-Heapify(A, i)



Heapsort

Heapsort(A)

- Build-Max-Heap(A)
- 2. **for** i = A.length **down to** 2
- 3. exchange A[1] with A[i]
- 4. A.heap-size = A.heap-size -1
- 5. Max-Heapify(A,1)

