CSE 250 Data Structures

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Day 14
Induction Review, Stacks and Queues
Textbook Ch. 15

Announcements

- PA2 is actually up now
 - Tests due 10/10
 - Full project due 10/17

Recap

QuickSort

- Divide and Conquer sorting algorithm like MergeSort
 - All of the work for Merge Sort happened during the combine step
 - QuickSort attempts to move the work to the divide step
- **Divide:** Move small elements to the left, and big elements to the right
- Conquer: Recursively call QuickSort on left and right halves
- Combine: ...nothing

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Divide: Move *small* elements to the left and *big* elements to the right

How do we define what is big and what is small?

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Pick a pivot value

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[smaller than pivot], pivot, [larger than pivot]

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How do we define what is big and what is small?

Pick a pivot value

[smaller than pivot], pivot, [larger than pivot]

How do we pick a pivot?

[4, 1, 8, 13, 12, 6, 2, 14, 7, 9, 3, 5, 11, 10, 15]

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```

If our pivot was the median value, then our list would be split in half by the divide step, resulting in the same runtime as MergeSort $O(n\log(n))$.

But finding the median value is expensive...(it also costs $n\log(n)$).

So what if we pick one randomly instead?

Expected Value

If I roll a 6-sided die, the probability of a particular side being rolled is 1/8

If X is a random variable representing this die roll, then the expected value of X is:

$$E[X] = \frac{1}{6} \cdot 1 + \frac{1}{6} \cdot 2 + \frac{1}{6} \cdot 3 + \frac{1}{6} \cdot 4 + \frac{1}{6} \cdot 5 + \frac{1}{6} \cdot 6$$

$$E[X] = \sum_{i=1}^{6} \frac{1}{6}i = 3.5$$

Expected Value

If I roll a 20-sided die, the probability of a particular side being rolled is 1/20

If X is a random variable representing this die roll, then the expected value of X is:

$$E[X] = \frac{1}{20} \cdot 1 + \frac{1}{20} \cdot 2 + \dots + \frac{1}{20} \cdot 20 = \sum_{i=1}^{20} \frac{1}{20}i$$

Expected Value

If I roll an n-sided die, the probability of a particular side being rolled is 1/n If X is a random variable representing this die roll, then the expected value of X is:

$$E[X] = \frac{1}{n} \cdot 1 + \frac{1}{n} \cdot 2 + \dots + \frac{1}{n} \cdot n = \sum_{i=1}^{n} \frac{1}{n}i$$

$$E[X] = \sum_{i} P_i \cdot X_i$$

Picking a pivot value randomly from the *n* elements of our sequence is the same as rolling an *n*-sided die.

There is a 1/n probability in any particular value being selected.

X = k means that X is the kth largest value, and the expected value of X corresponds to the median value.

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq 1 \\ T(0) + T(n-1) + \Theta(n) & \text{if } n > 1 \land X = 1 \\ T(1) + T(n-2) + \Theta(n) & \text{if } n > 1 \land X = 2 \\ T(2) + T(n-3) + \Theta(n) & \text{if } n > 1 \land X = 3 \\ \dots & \\ T(n-2) + T(1) + \Theta(n) & \text{if } n > 1 \land X = n-1 \\ T(n-1) + T(0) + \Theta(n) & \text{if } n > 1 \land X = n \end{cases}$$

$$E[T(n)] = \begin{cases} \Theta(1) & \text{if } n \leq 1 \\ E[T(X-1) + T(n-X)] + \Theta(n) & \text{otherwise} \end{cases}$$

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Expected value of two independent events can be split up

$$E[T(n)] = \begin{cases} \Theta(1) & \text{if } n \leq 1 \\ E[T(X-1)] + E[T(n-X)] + \Theta(n) & \text{otherwise} \end{cases}$$

$$E[T(n)] = \begin{cases} \Theta(1) & \text{if } n \leq 1 \\ E[T(X-1)] + E[T(n-X)] + \Theta(n) & \text{otherwise} \end{cases}$$

How are these two terms related?

$$E[T(X-1)]$$

$$E[T(X-1)]$$

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$

$$E[T(X-1)]$$

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$

$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(i-1)$$

$$E[T(X-1)]$$

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$

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$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(n-i)$$

$$E[T(X-1)]$$

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$

$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(i-1)$$

$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(n-i) = E[T(n-X)]$$

$$E[T(X-1)]$$

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$

$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(i-1)$$

$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(n-i) = E[T(n-X)]$$

$$E[T(n)] = \begin{cases} \Theta(1) & \text{if } n \leq 1 \\ 2E[T(X-1)] + \Theta(n) & \text{otherwise} \end{cases}$$

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Each T(X-1) is independent, so the expected values can be split out

$$E[T(n)] = \begin{cases} \Theta(1) & \text{if } n \leq 1\\ \frac{2}{n} \left(\sum_{i=0}^{n-1} E[T(i)] \right) + \Theta(n) & \text{otherwise} \end{cases}$$

Inductive Proof

Back to Induction

Hypothesis: $E[T(n)] \in O(n \log(n))$

Base Case: $E[T(2)] \le c (2 \log(2))$

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$$E[T(2)] \le c \ (2 \log(2))$$

 $2 \cdot E_i[T(i-1)] + 2c_1 \le 2c$

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$$2 \cdot E_{i}[T(i-1)] + 2c_{1} \le 2c$$
$$2 \cdot (T(0)/2 + T(1)/2) + 2c_{1} \le 2c$$

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Base Case:
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 $2 \cdot E_i[T(i-1)] + 2c_1 \le 2c$
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 $T(0) + T(1) + 2c_1 \le 2c$
 $2c_0 + 2c_1 \le 2c$

Base Case:
$$E[T(2)] \le c \ (2 \log(2))$$

 $2 \cdot E_i[T(i-1)] + 2c_1 \le 2c$
 $2 \cdot (T(0)/2 + T(1)/2) + 2c_1 \le 2c$
 $T(0) + T(1) + 2c_1 \le 2c$
 $2c_0 + 2c_1 \le 2c$
True for any $c \ge c_0 + c_1$

Assume: $E[T(n')] \le c (n' \log(n'))$ for all n' < n

Show: $E[T(n)] \le c (n \log(n))$

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Show:
$$E[T(n)] \le c \ (n \log(n))$$

$$\frac{2}{n} \left(\sum_{i=0}^{n-1} E[T(i)] \right) + c_1 \le cn \log(n)$$

Assume: $E[T(n')] \le c (n' \log(n'))$ for **all** n' < n

Show: $E[T(n)] \le c (n \log(n))$

Our *i* here is always less than *n*, so we can use our assumption to substitute

$$\frac{2}{n} \left(\sum_{i=0}^{n-1} E[T[i]] \right) + c_1 \le cn \log(n)$$

$$\frac{2}{n} \left(\sum_{i=0}^{n-1} ci \log(i) \right) + c_1 \le cn \log(n)$$

Assume: $E[T(n')] \le c (n' \log(n'))$ for all n' < n

Show: $E[T(n)] \le c \ (n \log(n))$

$$\frac{2}{n} \left(\sum_{i=0}^{n-1} E[T(i)] \right) + c_1 \le cn \log(n)$$

$$\frac{2}{n} \left(\sum_{i=0}^{n-1} ci \log(i) \right) + c_1 \le cn \log(n)$$

$$c\frac{2}{n}\left(\sum_{i=0}^{n-1} i\log(n)\right) + c_1 \le cn\log(n)$$

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$$c\frac{2\log(n)}{n}\left(\sum_{i=0}^{n-1}i\right) + c_1 \le cn\log(n)$$

$$c\frac{2}{n}\left(\sum_{i=0}^{n-1} i\log(n)\right) + c_1 \le cn\log(n)$$

$$c\frac{2\log(n)}{n}\left(\sum_{i=0}^{n-1}i\right) + c_1 \le cn\log(n)$$

$$c\frac{2\log(n)}{n}\left(\frac{(n-1)(n-1+1)}{2}\right) + c_1 \le cn\log(n)$$

$$c\frac{2}{n}\left(\sum_{i=0}^{n-1} i\log(n)\right) + c_1 \le cn\log(n)$$

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$$c\frac{2\log(n)}{n}\left(\frac{(n-1)(n-1+1)}{2}\right) + c_1 \le cn\log(n)$$

$$c\frac{\log(n)}{n}\left(n^2 - n\right) + c_1 \le cn\log(n)$$

$$c\frac{2}{n}\left(\sum_{i=0}^{n-1} i\log(n)\right) + c_1 \le cn\log(n)$$

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$$c\frac{2\log(n)}{n}\left(\frac{(n-1)(n-1+1)}{2}\right) + c_1 \le cn\log(n)$$

$$c\frac{\log(n)}{n}\left(n^2 - n\right) + c_1 \le cn\log(n)$$

$$cn\log(n) - c\log(n) + c_1 \le cn\log(n)$$

$$c\frac{2}{n}\left(\sum_{i=0}^{n-1} i\log(n)\right) + c_1 \le cn\log(n)$$

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$$cn\log(n) - c\log(n) + c_1 \le cn\log(n)$$

$$c_1 \le c\log(n)$$

QuickSort

So…is QuickSort $O(n \log(n))$ …?

No!

What guarantees do you get?

If f(n) is a Tight Bound

The algorithm always runs in cf(n) steps

If f(n) is a Worst-Case Bound

The algorithm always runs in at most cf(n)

If f(n) is an Amortized Worst-Case Bound

n invocations of the algorithm **always** run in cnf(n) steps

If f(n) is an Average Bound

...we don't have any guarantees

mutable.SeqADT

```
mutable.IndexedSeq (ie Array)
Efficiency apply(), update()
mutable.Buffer (ie ArrayBuffer, ListBuffer)
Efficiency apply(), update(), append()
```

A stack of objects on top of one another

Push Put a new object on top of the stack

Pop Remove the object on top of the stack

Top Peek at what's on top of the stack

```
s.push("Bob")
```

"Bob"

```
s.push("Bob")
s.push("Mary")
```

"Mary"

"Bob"

```
s.push("Bob")
s.push("Mary")
s.push("Sue")
```

"Sue"
"Mary"
"Bob"

```
s.push("Bob")
s.push("Mary")
s.push("Sue")
s.pop()
```

"Mary" "Bob"

```
s.push("Bob")
s.push("Mary")
s.push("Sue")
                                            "Steve"
s.pop()
                                             "Mary"
s.push("Steve")
                                             "Bob"
```

```
s.push("Bob")
s.push("Mary")
s.push("Sue")
s.pop()
                                            "Mary"
s.push("Steve")
                                            "Bob"
s.pop()
```

Stacks in Practice

- Storing function variables in a "call stack"
- Certain types of parsers ("context free")
- Backtracking search
- Reversing Sequences

```
trait Stack[A] {
  def push(element: A): Unit
  def top: A
  def pop: A
}
```

```
class ListStack[A] extends Stack[A] {
     val store = new SinglyLinkedList()
      def push(element: A): Unit =
       _store.prepend(element)
      def top: A =
       _store.head
      def pop: A =
       _store.remove(0)
```

```
class ListStack[A] extends Stack[A] {
      val store = new SinglyLinkedList()
      def push(element: A): Unit =
        _store.prepend(element)
      def top: A =
       _store.head
                                            What is the runtime?
      def pop: A =
       _store.remove(0)
```

```
class ListStack[A] extends Stack[A] {
      val store = new SinglyLinkedList()
      def push(element: A): Unit =
                                        \Theta(1)
        _store.prepend(element)
                        \Theta(1)
      def top: A =
        _store.head
                                               What is the runtime?
      def pop: A =
                            \Theta(1)
        _store.remove(0)
```

```
class ArrayBufferStack[A] extends Stack[A] {
     val store = new ArrayBuffer()
      def push(element: A): Unit =
       _store.append(element)
      def top: A =
       _store.last
      def pop: A =
       _store.remove(store.length-1)
```

```
class ArrayBufferStack[A] extends Stack[A] {
      val store = new ArrayBuffer()
      def push(element: A): Unit =
        _store.append(element)
      def top: A =
        _store.last
                                            What is the runtime?
      def pop: A =
       _store.remove(store.length-1)
```

```
class ArrayBufferStack[A] extends Stack[A] {
      val store = new ArrayBuffer()
      def push(element: A): Unit =
                                       Amortized O(1)
        store.append(element)
                     \Theta(1)
      def top: A =
        _store.last
                                              What is the runtime?
                       \Theta(1)
      def pop: A =
        _store.remove(store.length-1)
```

Stacks in Scala

Scala's **Stack** implementation is based on **ArrayBuffer**; Keeping memory together is worth the overhead of amortized *O*(1).

Outside of the US, "queueing" is lining up, ie at Starbucks

Enqueue Put a new object at the end of the queue

Dequeue Remove the next object in the queue

Head Peek at the next object in the queue

Front

enqueue ("Bob")

Front

"Bob"

```
enqueue("Bob")
enqueue("Mary")
```

Front

"Bob"

"Mary"

```
enqueue ("Bob")
enqueue ("Mary")
enqueue ("Sue")
```

Front

"Bob"

"Mary"

"Sue"

```
enqueue ("Bob")
enqueue ("Mary")
enqueue ("Sue")
dequeue ()
```

Front

"Mary"

"Sue"

```
enqueue ("Bob")
enqueue ("Mary")
enqueue ("Sue")
dequeue ()
enqueue ("Steve")
```

Front

"Mary"

"Sue"

"Steve"

```
enqueue ("Bob")
enqueue ("Mary"
enqueue ("Sue")
dequeue()
enqueue ("Steve")
dequeue()
 Front
                                                               Back
                   "Sue"
                            "Steve"
```

Queues vs Stacks

Queue First in, First Out (FIFO)

Statcks Last in, First Out (LIFO / FILO)

Queues in Practice

- Delivering network packets, emails, twitter/tiktok/instagram
- Scheduling CPU cycles
- Deferring long-running tasks

```
trait Queue[A] {
  def enqueue(element: A): Unit
  def dequeue: A
  def head: A
}
```

```
class ListQueue[A] extends Queue[A] {
     val store = new DoublyLinkedList()
      def enqueue(element: A): Unit =
       _store.append(element)
      def head: A =
       _store.head
      def dequeue: A =
       _store.remove(0)
```

```
class ListQueue[A] extends Queue[A] {
      val store = new DoublyLinkedList()
      def enqueue(element: A): Unit =
        store.append(element)
      def head: A =
       _store.head
                                            What is the runtime?
      def dequeue: A =
       _store.remove(0)
```

```
class ListQueue[A] extends Queue[A] {
      val store = new DoublyLinkedList()
      def enqueue(element: A): Unit =
        store.append(element)
                       \Theta(1)
      def head: A =
        _store.head
                                              What is the runtime?
      def dequeue: A =
                            \Theta(1)
        _store.remove(0)
```

Thought Experiment: How can we use an array to build a queue?