CSE 250 Data Structures

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Stacks and Queues Textbook Ch. 15

Announcements

- WA1 due tonight @ Midnight. Check your submissions!
- PA2 will be released by the end of the week (hopefully tonight) so keep an eye on Piazza

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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Pick a pivot value

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[smaller than pivot], pivot, [larger than pivot] How do we pick a pivot?

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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 ¹/₆ 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:

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

00

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:

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

00

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

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.

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

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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?

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

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

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

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

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

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

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

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

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

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

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

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

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

$$= \sum_{i=1}^{n} P_i \cdot T(X_i - 1)$$
They are equivalent!!
$$= \sum_{i=1}^{n} \frac{1}{n} \cdot T(i - 1)$$

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

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

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

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

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

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

Assume: $E[T(n')] \le c (n' \log(n'))$ for all n' < nShow: $E[T(n)] \le c (n \log(n))$ $\frac{2}{n} \left(\sum_{i=0}^{n-1} E[T(i)] \right) + c_1 \le cn \log(n)$

Assume: $E[T(n')] \le c (n' \log(n'))$ for **all** n' < n**Show:** $E[T(n)] \le c (n \log(n))$ $\frac{2}{n} \left(\sum_{i=1}^{n-1} E[T(i)] \right) + c_1 \le cn \log(n)$ $\frac{2}{n} \left(\sum_{i=0}^{n-1} ci \log(i) \right) + c_1 \le cn \log(n)$

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

Assume: $E[T(n')] \le c (n' \log(n'))$ for all n' < n**Show:** $E[T(n)] \le c (n \log(n))$ $\frac{2}{n} \left(\sum_{i=0}^{n-1} E[T(i)] \right) + c_1 \le cn \log(n)$ $\frac{2}{n} \left(\sum_{i=0}^{n-1} ci \log(i) \right) + c_1 \le cn \log(n)$ $2\left(\frac{n-1}{2}\right)$

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

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

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

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

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

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$$c\frac{\log(n)}{n}\left(n^{2}-n\right) + c_{1} \leq cn\log(n)$$

$$cn\log(n) - c\log(n) + c_{1} \leq cn\log(n)$$

$$c_{1} \leq c\log(n)$$

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

Analysis Tools/Techniques	ADTs	Data Structures
Asymptotic Analysis, (Unqualified) Runtime Bounds		
	Seq	Array
Amortized Runtime	Seq, Buffer	ArrayBuffer
	Seq	Linked Lists
Recursive analysis, divide and conquer, Average/Expected Runtime		
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We are here \rightarrow

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Looking Ahead...

Analysis Tools/Techniques	ADTs	Data Structures
	Stack, Queue	
	Graphs	EdgeList, Adjacency List, Adjacency Matrix
	Heaps, Trees	BST, AVL Tree, Red-Black Tree
	HashTables	
Miscellaneous		

mutable.SeqADT

mutable.IndexedSeq (ie Array)
Efficiency apply(), update()

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

Stacks

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

Stacks

s.push("Bob")

"Bob"

Stacks

s.push("Bob")

s.push("Mary")


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



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

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

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

```
s.pop()
```





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

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

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



s.push("Bob")

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 =
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def top: A =
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class ListStack[A] extends Stack[A] {
 val _store = new SinglyLinkedList()

def push(element: A): Unit =
______store.prepend(element)

```
def top: A = \Theta(1)
_store.head
```

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
```

```
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)

```
def top: A = \Theta(1)
__store.last____
```

def pop: A = O(1)
_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 queueDequeue Remove the next object in the queueHead Peek at the next object in the queue

Front



enqueue("Bob")

Front



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

Front



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





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

Front



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

Front





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

Front





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()

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def enqueue(element: A): Unit =
   ____store.append(element)
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def head: A =
  _store.head
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def dequeue: A =
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```

class ListQueue[A] extends Queue[A] {
 val _store = new DoublyLinkedList()

def enqueue(element: A): Unit = O(1)
 store.append(element)

def head: $A = \Theta(1)$ _store.head

```
def dequeue: A = _____store.remove(0) \Theta(1)
```



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