

# CSE 250: Induction

## Lecture 12

Sept 25, 2023

# Reminders

- WA2 due Sun, Oct 1 at 11:59 PM
  - Released Monday
- Midterm 1 on Oct 2 in-class
  - Covers: Asymptotics, Sequences/Lists, Arrays, Linked Lists, Recursion
  - Bounds: Tight Upper/Lower, Unqualified vs Amortized

# Sorted Lists

## Sorting an Array

- **Bubble Sort:**  $O(N^2)$
- **Merge Sort:**  $O(N \log(N))$  (probably?)

# Induction

- 1 Generate a hypothesis.
- 2 Prove the hypothesis for the base case.
- 3 Prove the hypothesis inductively.
  - Assume that the hypothesis is true for a small case (e.g.,  $N - 1$ ).
  - Prove that the hypothesis is true for a bigger case (e.g.,  $N$ ).

# Induction

## We showed there exists a $c$ such that...

- $T_f(1) \leq 1 \cdot c$  (Base Case)
- if  $T_f(N-1) \leq (N-1) \cdot c$  then  $T_f(N) \leq N \cdot c$  (Inductive Proof)

## So...

- 1  $T_f(1) \leq 1 \cdot c$  (Base Case)
- 2  $T_f(2) \leq 2 \cdot c$  (#1 + Inductive Proof)
- 3  $T_f(3) \leq 3 \cdot c$  (#2 + Inductive Proof)
- 4  $T_f(4) \leq 4 \cdot c$  (#3 + Inductive Proof)
- 5  $T_f(5) \leq 5 \cdot c$  (#4 + Inductive Proof)
- 6  $T_f(6) \leq 6 \cdot c$  (#5 + Inductive Proof)
- 7 ...

The proof holds for any  $N \geq 1 \rightarrow T(N) \in O(N)$  ✓

# Factorial

What is the complexity of Factorial?

```
1  public long factorial(long N)
2  {
3      if(N <= 1){ return 1; }
4      else { return N * factorial(N-1); }
5  }
```

**Answer:**  $O(N)^1$

**How much memory does it use?**

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<sup>1</sup>Technically it's  $\theta(N)$ , but we haven't proven  $T(N) \in \Omega(N)$

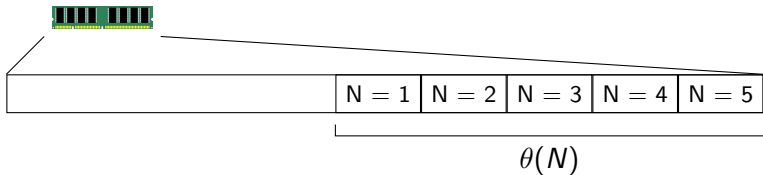
# Stack Frames

Every time you call a function, it allocates some memory for local variables (e.g.,  $N$ ).

This chunk of memory is called a **Stack Frame**.

This is where the term `StackOverflowError` comes from.

# Stack Frames





# Factorial (as a loop)

```
1  public long factorial(long N)
2  {
3      long total = 1;
4      for(long i = N; i > 0; i--)
5      {
6          total *= i
7      }
8      return total
9  }
```

# Factorial

**Why does this work?**

# Factorial (as a loop)

```
1  public long factorial(long N)
2  {
3      if(N <= 1){ return 1; }
4      else { return N * factorial(N-1); }
5  }
```

Each call to factorial only makes **one** recursive call.

# Factorial (as a loop)

- Is  $N > 1$ ? ← Requires stack frame
- Compute  $\text{arg} = N - 1$  ← Requires stack frame
- Call `factorial(arg)`
- Compute  $N \times \text{result}$  ← Requires stack frame
- Return

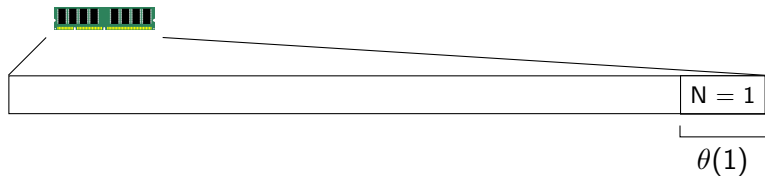
# Factorial (as a loop)

```
1 public long factorial(long N, long total)
2 {
3     if(N <= 1){ return total; }
4     else { return factorial(N-1, N * total); }
5 }
```

# Factorial (as a loop)

- Is  $N > 1$ ? ← Requires stack frame
- Compute  $\text{arg1} = N - 1$  ← Requires stack frame
- Compute  $\text{arg2} = N \times \text{total}$  ← Requires stack frame
- Call `factorial(arg1, arg2)`
- Return ← Stack frame unnecessary

# Stack Frames



# Tail Recursion

If the recursive call is the *last* operation before the return, most languages optimize the recursion away<sup>2</sup>.

This is called **Tail Recursion**

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<sup>2</sup>... but not Java



# Merge Sort

```
1  public ArrayList<E> mergeSort<E>(ArrayList<E> list)
2  {
3      if(list.size() > 2){
4          int splitIndex = input.size()/2;
5          ArrayList<E> left =
6              mergeSort(list.subList(0, splitIndex));
7          ArrayList<E> right =
8              mergeSort(list.subList(splitIndex, list.size()));
9          return merge(left, right);
10     } else {
11         if((list.size() == 2) && (list.get(0) > list.get(1))){
12             E tmp = list.get(0);
13             list.set(0, list.get(1));
14             list.set(1, tmp);
15         }
16         return list;
17     }
18 }
```

# Merge Sort

- If  $N > 2$ 
  - Split  $O(1)$
  - $2 \times \text{mergeSort}(\frac{N}{2})$   $2 \cdot T(\frac{N}{2})$
  - $\text{merge}(N)$   $O(N)$
- Otherwise
  - Sort 2 elements  $O(1)$

$$T_{\text{merge}}(N) = \begin{cases} O(N) + 2 \cdot T_{\text{merge}}(\frac{N}{2}) & \text{if } N > 2 \\ O(1) & \text{otherwise} \end{cases}$$

# Merge Sort

$$T_{\text{merge}}(N) = \begin{cases} O(N) + 2 \cdot T_{\text{merge}}(\frac{N}{2}) & \text{if } N > 2 \\ O(1) & \text{otherwise} \end{cases}$$

## Induction

- **Hypothesis:**  $T_{\text{merge}}(N) \in O(N \cdot \log(N))$ 
  - $T_{\text{merge}}(N) \leq c \cdot N \log(N)$
- **Base Step:**
  - Show  $T_{\text{merge}}(2) \leq c \cdot 2 \cdot \log(2) = c \cdot 2$
- **Inductive Step:**
  - Assume  $T_{\text{merge}}(\frac{N}{2}) \leq c \cdot \frac{N}{2} \cdot \log(\frac{N}{2})$
  - Show  $T_{\text{merge}}(N) \leq c \cdot N \cdot \log(N)$

# Base Step

$$T_{\text{merge}}(N) = \begin{cases} O(N) + 2 \cdot T_{\text{merge}}(\frac{N}{2}) & \text{if } N > 2 \\ O(1) & \text{otherwise} \end{cases}$$

$$T_{\text{merge}}(2) \stackrel{?}{\leq} c \cdot 2$$

$$O(1) \leq 2c$$

# Inductive Step

$$T_{\text{merge}}(N) = \begin{cases} O(N) + 2 \cdot T_{\text{merge}}\left(\frac{N}{2}\right) & \text{if } N > 2 \\ O(1) & \text{otherwise} \end{cases}$$

**Assume:**  $T_{\text{merge}}\left(\frac{N}{2}\right) \leq c \cdot \frac{N}{2} \cdot \log\left(\frac{N}{2}\right)$

$$T_{\text{merge}}(N) \stackrel{?}{\leq} c \cdot N \log(N)$$

$$O(N) + 2 \cdot T_{\text{merge}}\left(\frac{N}{2}\right) \stackrel{?}{\leq} c \cdot N \log(N)$$

$$O(N) + 2 \cdot c \cdot \frac{N}{2} \cdot \log\left(\frac{N}{2}\right) \stackrel{?}{\leq} c \cdot N \log(N)$$

# Inductive Step

$$O(N) + 2 \cdot c \cdot \frac{N}{2} \cdot \log\left(\frac{N}{2}\right) \stackrel{?}{\leq} c \cdot N \log(N)$$

$$O(N) + c \cdot N \cdot \log\left(\frac{N}{2}\right) \stackrel{?}{\leq} c \cdot N \log(N)$$

$$O(N) + c \cdot N \cdot (\log(N) - \log(2)) \stackrel{?}{\leq} c \cdot N \log(N)$$

$$O(N) + c \cdot N \cdot \log(N) \stackrel{?}{\leq} c \cdot N \log(N) + c \cdot N$$

$$O(N) \leq c \cdot N$$

# Merge Sort

So...  $T_{merge}(N) \in O(N \cdot \log(N))$

# Quicksort

## Merge Sort

- Split up the problem  $(O(1))$
- Merge the sub-solutions  $(O(N))$

## Quicksort

- Split up the problem  $(O(N))$
- Merge the sub-solutions  $(O(1))$



# Quicksort

- If the list has only 1 element, it's sorted. Return.  $O(1)$
- Find the **median** value ('pivot').  $T_{pivot}(N)O(N \log(N))$
- Create a list of elements  $<$  **medium**.  $O(N)$
- Create a list of elements  $>$  **medium**.  $O(N)$
- Sort each of the two lists  $2 \cdot T(\frac{N}{2})$
- Concatenate the lists  $O(N)$  or  $O(1)$

$$T_{merge}(N) = \begin{cases} 1 & \text{if } N \leq 1 \\ O(N) + 2T_{merge}(\frac{N}{2}) + T_{pivot}(N)N \log(N) & \text{otherwise} \end{cases}$$

# Quicksort

**Idea:** If we pick a pivot value at random, in expectation, half of the values will be lower.

# The Worst-Case Pivot

What is the worst case runtime?

# The Worst-Case Pivot

What if we always pick the worst pivot?

1 [ 8, 7, 6, 5, 4, 3, 2, 1 ]

2 [ 7, 6, 5, 4, 3, 2, 1 ], 8, []

3 [ 6, 5, 4, 3, 2, 1 ], 7, [], 8, []

4 [ 5, 4, 3, 2, 1 ], 6, [], 7, [], 8, []

...

- For each level,  $O(N)$  work
- At worst,  $O(N)$  levels

**Total:**  $T_{quicksort}(N) \in O(N^2)$

# Quicksort

# Expected Runtime

Is the worst case runtime representative?

**No!** (in typical cases, it will be faster)

**Is there something we can say about the runtime?**

# Quicksort

If we pick the  $X$ 'th largest element as a pivot, what is  $T_{qs}(N)$ ?

- $X = 1: \theta(N) + T_{qs}(0) + T_{qs}(N - 1)$
- $X = 2: \theta(N) + T_{qs}(1) + T_{qs}(N - 2)$
- $X = 3: \theta(N) + T_{qs}(2) + T_{qs}(N - 3)$
- $X = 4: \theta(N) + T_{qs}(3) + T_{qs}(N - 4)$
- ...
- $X = N - 1: \theta(N) + T_{qs}(N - 2) + T_{qs}(1)$
- $X = N: \theta(N) + T_{qs}(N - 1) + T_{qs}(0)$

$$T_{qs}(N) = \theta(N) + T_{qs}(X - 1) + T_{qs}(N - X) \quad (\text{for } X \in [1, N])$$

# Probabilities

What is the chance we (randomly) pick  $X = 1$ ?  $P[X = 1] = \frac{1}{N}$

What is the chance we pick  $X = \lfloor \frac{N}{2} \rfloor$ ?  $P[X = \lfloor \frac{N}{2} \rfloor] = \frac{1}{N}$

What is the chance we pick  $X = N$ ?  $P[X = N] = \frac{1}{N}$



# Probability

If I roll a d6 (a 6-sided die 🎲)  $k$  times,  
what is the average over all possible outcomes?

$k=1$ 

If I roll a d6 (a 6-sided die 🎲) 1 time...

Roll	Probability	Contribution
1	$\frac{1}{6}$	1
2	$\frac{1}{6}$	2
3	$\frac{1}{6}$	3
4	$\frac{1}{6}$	4
5	$\frac{1}{6}$	5
6	$\frac{1}{6}$	6

# Expectation

$$\frac{1}{6} \cdot 1 + \frac{1}{6} \cdot 2 + \frac{1}{6} \cdot 3 + \frac{1}{6} \cdot 4 + \frac{1}{6} \cdot 5 + \frac{1}{6} \cdot 6 = 3.5$$

$$= \sum_i \mathbf{Probability}_i \cdot \mathbf{Contribution}_i$$

If  $X$  is a variable representing the outcome of the roll, we call this number the **expectation** of  $X$ , or  $E[X]$ .

$$E[X] = \sum_i P_i \cdot X_i$$

$k = 2$ 

If I roll a d6 (a 6-sided die 🎲) 2 times...

Does one roll affect the outcome of the other?

**No:** Each roll is a *independent* event.

# Independent Events

If  $X$  and  $Y$  are random variables representing the outcome of each roll (i.e., independent random variables):

$$\begin{aligned} E[X + Y] &= E[X] + E[Y] \\ &= 3.5 + 3.6 = 7 \end{aligned}$$

# Back to Quicksort

$$T_{qs}(N) = \begin{cases} \theta(1) & \text{if } N \leq 1 \\ T_{qs}(X - 1) + T_{qs}(N - X) + \theta(N) & \text{otherwise} \end{cases}$$

... **but  $X$  is random!**

# Expected Runtimes

$$E[T_{qs}(N)] = \begin{cases} \theta(1) & \text{if } N \leq 1 \\ E[T_{qs}(X - 1) + T_{qs}(N - X) + \theta(N)] & \text{otherwise} \end{cases}$$

# Expected Runtimes

$$E[T_{qs}(N)] = \begin{cases} \theta(1) & \text{if } N \leq 1 \\ E[T_{qs}(X - 1)] + E[T_{qs}(N - X)] + E[\theta(N)] & \text{otherwise} \end{cases}$$



# Expected Runtimes

$$\begin{aligned} & E[T(X - 1)] \\ &= \sum_{i=1}^N P_i \cdot T(X_i - 1) \\ &= \sum_{i=1}^N \frac{1}{N} \cdot T(i - 1) \quad (T(0) \text{ up to } T(n - 1)) \\ &= \sum_{i=1}^N \frac{1}{N} \cdot T(n - i) \quad (T(n - 1) \text{ down to } T(0)) \\ &= E[T(N - X)] \end{aligned}$$

## Expected Runtimes

$$E[T_{qs}(N)] = \begin{cases} \theta(1) & \text{if } N \leq 1 \\ 2 \cdot E[T_{qs}(X - 1)] + E[T_{qs}(N - X)] + \theta(N) & \text{otherwise} \end{cases}$$

**The  $X$  we pick at each step is independent.**

# Expected Runtimes

$$E[T_{qs}(N)] = \begin{cases} \theta(1) & \text{if } N \leq 1 \\ 2 \cdot \left( \sum_{i=10}^{N-1} \frac{1}{N} E[T_{qs}(i-1)] \right) + \theta(N) & \text{otherwise} \end{cases}$$

Back to Induction! **Inductive Hypothesis:**

$$E[T_{qs}(N)] \in O(N \log(N))$$

$$E[T_{qs}(N)] \leq c \cdot N \cdot \log(N)$$

# Quicksort's Runtime

$$E[T_{qs}(N)] = \begin{cases} c_1 & \text{if } N \leq 1 \\ 2 \cdot \left( \sum_{i=0}^{N-1} \frac{1}{N} E[T_{qs}(i)] \right) + c_2 N & \text{otherwise} \end{cases}$$

## Base Case

$$E[T_{qs}(1)] \stackrel{?}{\leq} c \cdot 1 \cdot \log(1)$$

$$E[T_{qs}(1)] \stackrel{?}{\leq} c \cdot 1 \cdot 0$$

$$E[T_{qs}(1)] \not\leq 0 \quad \dots \text{oops}$$

# Quicksort's Expected Runtime

$$E[T_{qs}(N)] = \begin{cases} c_1 & \text{if } N \leq 1 \\ 2 \cdot \left( \sum_{i=0}^{N-1} \frac{1}{N} E[T_{qs}(i)] \right) + c_2 N & \text{otherwise} \end{cases}$$

**Base Case (with  $N_0 = 2$ )**

$$E[T_{qs}(2)] \stackrel{?}{\leq} c \cdot 2 \cdot \log(2)$$

$$2 \cdot \left( \sum_{i=0}^1 \frac{1}{2} E[T_{qs}(i)] \right) + 2c_2 \stackrel{?}{\leq} c \cdot 2 \cdot \log(2)$$

$$\left( \sum_{i=0}^1 \frac{1}{2} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c$$

# Quicksort's Expected Runtime

$$\left( \sum_{i=0}^1 \frac{1}{2} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c$$

$$\frac{1}{2} (c_1 + c_1) + c_2 \stackrel{?}{\leq} c$$

$$c_1 + c_2 \leq c$$

(true if we set  $c \geq c_1 + c_2$ )

# Quicksort's Expected Runtime

$$E[T_{qs}(N)] = \begin{cases} c_1 & \text{if } N \leq 1 \\ 2 \cdot \left( \sum_{i=0}^{N-1} \frac{1}{N} E[T_{qs}(i)] \right) + c_2 N & \text{otherwise} \end{cases}$$

## Inductive Step

Assume:  $E[T_{qs}(N')] \leq c \cdot N' \log(N')$  for all  $2 \leq N' \leq N$

Show:  $E[T_{qs}(N)] \leq c \cdot N \log(N)$

$$2 \cdot \left( \sum_{i=0}^{N-1} \frac{1}{N} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$2 \cdot \frac{1}{N} \left( \sum_{i=0}^{N-1} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

# Quicksort's Expected Runtime

$$2 \cdot \frac{1}{N} \left( \sum_{i=0}^{N-1} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$2 \cdot \frac{1}{N} \left( E[T_{qs}(0)] + E[T_{qs}(1)] + \sum_{i=2}^{N-1} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$2 \cdot \frac{1}{N} \left( 2 \cdot c_1 + \sum_{i=2}^{N-1} E[T_{qs}(i)] \right) + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2}{N} \left( \sum_{i=2}^{N-1} E[T_{qs}(i)] \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$



# Quicksort's Expected Runtime

$$\frac{2}{N} \left( \sum_{i=2}^{N-1} E[T_{qs}(i)] \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2}{N} \left( \sum_{i=2}^{N-1} c \cdot i \log(i) \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2c}{N} \left( \sum_{i=2}^{N-1} i \log(i) \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

The following left-hand-side is strictly bigger than the preceding.

$$\frac{2c}{N} \left( \sum_{i=1}^{N-1} i \log(N) \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

# Quicksort's Expected Runtime

$$\frac{2c}{N} \left( \sum_{i=1}^{N-1} i \log(N) \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2c \log(N)}{N} \left( \sum_{i=1}^{N-1} i \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2c \log(N)}{N} \left( \frac{(N-1) \cdot N}{2} \right) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$c \cdot N \log(N) - c \cdot \log(N) + \frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot N \log(N)$$

$$\frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot \log(N)$$

## Quicksort's Expected Runtime

$$\frac{2 \cdot c_1}{N} + c_2 \stackrel{?}{\leq} c \cdot \log(N)$$

The following left-hand-side is strictly bigger than the preceding.

$$2 \cdot c_1 + c_2 \stackrel{?}{\leq} c \cdot \log(N)$$

True for any  $N \geq 2$  if  $c \geq 2 \cdot c_1 + c_2$

## Quicksort's Expected Runtime

$$E[T_{qs}(N)] \in O(N \log(N))$$

So is Quicksort  $O(N \log(N))$ ? **No!**

Quicksort's **Expected** runtime is  $O(N \log(N))$

# Bound Guarantees

- $f(N)$  is a Worst-Case Bound  $(T(N) \in O(f(N)))$   
The algorithm **always** runs in at most  $c \cdot f(N)$  steps.
- $f(N)$  is an Amortized Worst-Case Bound  
 $N$  **invocations** of the algorithm **always** run in at most  $N \cdot c \cdot f(N)$  steps.
- $f(N)$  is an Expected Worst-Case Bound  $(E[T(N)] \in O(f(N)))$   
The algorithm is **statistically likely** to run in at most  $c \cdot f(N)$  steps.