## Lecture 11: Midterm Review

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bit.ly/cs3000syllabus

## Business

- Homework 2 deadline has passed
- Solutions released on Canvas as of this morning
- Will be graded by early next week, please be patient
- Do not try to grade yourself in detail
- Do not ask us "how many points will I get for..."
- Just wait and you can ask for clarification/modification after grading is done!
- Midterm 1 to be released TONIGHT 8PM and due on Friday 8PM (Boston times)
- Some review questions answered today!


## This week

## Today:

- First half-ish: Continue with graph algorithms
- Second half-ish: Answers to student-submitted questions
- Form link sent out on Piazza

Thursday:

- No class while midterm exam is out


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## Midterm basic info

Reminder:

- Absolutely NO collaboration of any kind or use of the internet to find solutions is allowed
- You can use the Erickson book or the CLR book, as well as the lectures, slides, and any notes you have taken


## Topics:

- Asymptotic order of growth
- Recurrence relations
- Proof by induction
- Recursive algorithms
- Dynamic programming

If you understand the solutions to the homework problems and have followed along with the lectures, you will do fine!

## Midterm basic info

You should direct any midterm related questions to Piazza

- Ask all questions privately to all of the instructors
- If we think the answer is relevant for everyone, we will make it public or write it in a note.
- You should not publicly post anything about the exam anywhere, including on Piazza.

We will also be holding office hours as scheduled, however do not expect nearly as many hints as we give for the homework assignments!

- We are likely only going to answer clarifying questions
- Ask clarifying questions on Piazza BEFORE attending office hours
- Often writing out your question helps you answer it yourself!


## Format of Today's Review

You have submit questions over the last few days via a Google Form

I have aggregated your questions and chosen some problems to go over

As we are going over the problems, you can ask me questions in the chat to help clarify

If we run in to a "Tim doesn't have a good answer" type situation, I will post a Piazza note after lecture with a better answer

## Constant time

When we say an operation takes "constant time", we literally mean there is some constant $c$ that does not depend on $n$ that describes the number of instructions it takes to do that operation. We write it as $O$ (1).

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## When we say an operation takes "constant time", we literally mean there is some constant $c$ that does not depend on $n$ that describes the number of instructions it takes to do that operation. We write it as $O$ (1).

```
INSERTION-SORT (A)
for j }\leftarrow2\mathrm{ to length[A]
do key \leftarrowA[j]
    \Delta Insert A[j] into the sorted
        sequence }A[1 j - 1]
        i}\leftarrowj - 1
        while i > 0 and A[i] > key
            do }A[i+1]\leftarrowA[i
            i}\leftarrow i - 1
            A[i + 1] \leftarrow key
\begin{tabular}{ll} 
cost & times \\
\(C_{1}\) & \(n\) \\
\(C_{2}\) & \(n-1\) \\
& \\
0 & \(n-1\) \\
\(C_{4}\) & \(n-1\) \\
\(C_{5}\) & \(\sum_{j=2}^{n} t_{j}\) \\
\(C_{6}\) & \(\sum_{j=2}^{n}\left(t_{j}-1\right)\) \\
\(C_{7}\) & \(\sum_{j=2}^{n}\left(t_{j}-1\right)\) \\
\(C_{8}\) & \(n-1\)
\end{tabular}
```

The running time of the algorithm is the sum of running times for each statement executed; a statement that takes $c_{i}$ steps to execute and is executed $n$ times will contribute $c_{i} n$ to the total running time. ${ }^{[5]}$ To compute $T(n)$, the running time of INSERTION-SORT, we sum the products of the cost and times columns, obtaining
"Recurrences": what do words mean?

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When we are describing a solution to a problem in terms of subproblems, we have a recursive relationship like:

$$
f_{n}\left\{\begin{array}{cll}
0 & \text { if } n=0 \\
1 & \text { if } n=1 \\
f_{n-1}+f_{n-2} & \text { otherwise }
\end{array} \quad \text { SubsetSumS }(i, t)=\left\{\begin{array}{cl}
\text { True } & \text { if } t=0 \\
\text { False } & \text { if } i>n \\
\operatorname{SubSum}(i+1, t) & \text { if } t<X[i] \\
\operatorname{SubSum}(i+1, t) \vee \operatorname{SubSum}(i+1, t-X[i]) & \text { otherwise }
\end{array}\right.\right.
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When we describe the runtime of an algorithm, we have a recursive relationship like:

$$
T(n)=T\left(\frac{n}{2}\right)+3 \quad T(n)=\sum_{i=0}^{n-1} T(i)+3
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T(n)=T\left(\frac{n}{2}\right)+3 \quad T(n)=\sum_{i=0}^{n-1} T(i)+3
$$

This is a "recurrence relation"

## Running time of recursive algorithms

To get the running time of a recursive algorithm, we write a recurrence relation describing the time to run the algorithm on an input of size $n$ in terms of the running time on inputs smaller than $n$.

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To get the running time of a recursive algorithm, we write a recurrence relation describing the time to run the algorithm on an input of size $n$ in terms of the running time on inputs smaller than $n$.

Let's look back at how we wrote the recurrence relations for binary search and merge sort.

## Binary Search Recurrence Relation

What does the recurrence relation look like for binary search?

```
StartSearch(A,t):
    // A[1:n] sorted in ascending order
    Return Search(A,1,n,t)
Search(A,l,r,t):
    If(\ell> r): return FALSE
    m}\leftarrow\ell+\lfloor\frac{r-\ell}{2}
    If(A[m] = t): return m
    ElseIf(A[m] > t): return Search(A,l,m-1,t)
    Else: return Search(A,m+1,r,t)
```


## MergeSort: Runtime Analysis

```
MergeSort(A[1..n]):
    if n>1
        m\leftarrow\lfloorn/2\rfloor
        MergeSort(A[1..m])
        MergeSort(A[m+1..n])〈<Recurse!\\rangle
        Merge(A[1..n],m)
```

| Merge $(A[1 . . n], m):$ |
| :--- |
| $i \leftarrow 1 ; j \leftarrow m+1$ |
| for $k \leftarrow 1$ to $n$ |
| if $j>n$ |
| $B[k] \leftarrow A[i] ; i \leftarrow i+1$ |
| else if $i>m$ |
| $B[k] \leftarrow A[j] ; j \leftarrow j+1$ |
| else if $A[i]<A[j]$ |
| $B[k] \leftarrow A[i] ; i \leftarrow i+1$ |
| else $B[k] \leftarrow A[j] ; j \leftarrow j+1$ |
| $B[k]$ |
| for $k \leftarrow 1$ to $n$ |
| $A[k] \leftarrow B[k]$ |

Merge $(A[1 . . n], m)$ :
$i \leftarrow 1 ; j \leftarrow m+1$
for $k \leftarrow 1$ to $n$
if $j>n$
$B[k] \leftarrow A[i] ; i \leftarrow i+1$
else if $i>m$
$B[k] \leftarrow A[j] ; j \leftarrow j+1$
else if $A[i]<A[j]$
$B[k] \leftarrow A[i] ; i \leftarrow i+1$
else
$B[k] \leftarrow A[j] ; j \leftarrow j+1$
$k \leftarrow 1$ to $n$
$A[k] \leftarrow B[k]$

Let's write down a recurrence relation that describes the runtime:

Difference between the two

$$
T(n)=T\left(\frac{n}{2}\right)+O(1)
$$

$$
T(n)=T\left(\frac{n}{2}\right)+O(n)
$$

## Solving Recurrence Relations

3 Methods:

1. Master theorem (if applicable)
2. Writing a few values $\rightarrow$ guess and check
3. Recursion Trees

## Master Theorem

- Recipe for recurrences of the form:
- $T(n)=\boldsymbol{a} \cdot T(n / b)+C n^{d}$
- Three cases:
- $\left(\frac{a}{b^{d}}\right)>1: T(n)=\Theta\left(n^{\log _{b} a}\right)$
- $\left(\frac{a}{b^{d}}\right)=1: T(n)=\Theta\left(n^{d} \log n\right)$
- $\left(\frac{a}{b^{d}}\right)<1: T(n)=\Theta\left(n^{d}\right)$

$$
\left(\frac{1}{2^{0}}\right)=1
$$

So

$$
T(n)=\Theta\left(n^{0} \log n\right)
$$

and we get

$$
T(n)=\Theta(\log n)
$$

Solving $T(n)=2 T(n-1)+1$

$$
\begin{gathered}
T(n)=2 T(n-1)+1 \\
T(0)=0
\end{gathered}
$$

## Solving $T(n)=2 T(n-1)+1$

$$
\begin{gathered}
T(n)=2 T(n-1)+1 \\
T(0)=0 \\
T(1)=2 T(0)+1=1
\end{gathered}
$$

## Solving $T(n)=2 T(n-1)+1$

$$
\begin{gathered}
T(n)=2 T(n-1)+1 \\
T(0)=0 \\
T(1)=2 T(0)+1=1 \\
T(2)=2 T(1)+1=2+1=3
\end{gathered}
$$

## Solving $T(n)=2 T(n-1)+1$

$$
\begin{gathered}
T(n)=2 T(n-1)+1, \\
T(0)=0 \\
T(1)=2 T(0)+1=1 \\
T(2)=2 T(1)+1=2+1=3 \\
T(3)=2 \cdot 3+1=7
\end{gathered}
$$

## Solving $T(n)=2 T(n-1)+1$

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\begin{gathered}
T(n)=2 T(n-1)+1, \\
T(0)=0 \\
T(1)=2 T(0)+1=1 \\
T(2)=2 T(1)+1=2+1=3 \\
T(3)=2 \cdot 3+1=7 \\
T(4)=2 \cdot 7+1=15 \\
T(5)=2 \cdot 15+1=31
\end{gathered}
$$

There is a pattern here!

## Solving $T(n)=2 T(n-1)+1$

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There is a pattern here!
We can use it to "guess" the running time.

What is our guess?

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\end{gathered}
$$

There is a pattern here!

We can use it to "guess" the running time.

What is our guess?
Now we need to prove it!

Proving $T(n)=2 T(n-1)+1=O\left(2^{n}\right)$

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\begin{gathered}
T(n)=2 T(n-1)+1 \\
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$$
\begin{gathered}
T(n)=2 T(n-1)+1 \\
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$$

Assume for induction that

$$
T(n)=2 T(n-1)+1 \leq c 2^{n}+1 \quad \text { (for some constant } c \text { ) }
$$

We will show that $\quad T(n+1)=2 T(n+1-1)+1 \leq c 2^{n+1}+1$

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$$
\text { We have } \quad 2 T(n) \leq c 2^{n+1}
$$

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We have

$$
2 T(n) \leq c 2^{n+1}
$$

$$
2 \cdot 2^{n} c \leq c 2^{n+1} \quad \text { By the inductive hypothesis }
$$

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We will show that

$$
\begin{aligned}
& T(n+1)=2 T(n+1-1)+1 \leq c 2^{n+1}+1 \\
& \text { We have } \\
& \qquad \begin{array}{c}
2 T(n) \leq c 2^{n+1} \\
2 \cdot 2^{n} c \leq c 2^{n+1} \quad \text { By the inductive hypothesis } \\
2^{n+1} \leq 2^{n+1}
\end{array}
\end{aligned}
$$

## Recursion Tree: MOMSelect Running Time

```
MOMSelect(A[1..n], k):
    If n <= 25:
        return median(A)
    Else:
        mom \leftarrow MOM(A[1..n])
        r}\leftarrow\mathrm{ Partition(A, mom)
        If k < r:
        Return MOMSelect(A[1..r], k)
        ElseIf k > r:
            Return MOMSelect(A[r+1..n], k-r)
        Else:
            Return A[r]
```

What is a recurrence relation for MOMSelect?

$$
T(n)=T(\text { Selection })+T(M O M)+f(o p s \text { per step })
$$

## Recursion Tree

$$
T(n)=T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n)
$$



Since the work at each level is decreasing exponentially, the $\mathrm{O}(\mathrm{n})$ term dominates!

## Proof by induction

$$
\begin{aligned}
& T(n)=T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n) \\
& T(1)=1
\end{aligned}
$$

We want to show that

$$
T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n) \leq O(n), \text { meaning }
$$

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\begin{gathered}
T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n) \leq O(n), \text { meaning } \\
T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+n \leq C n(\text { for some } C)
\end{gathered}
$$

## Proof by induction

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\end{gathered}
$$

By induction, since $\frac{1}{5} n<\frac{7}{10} n<n$, we have

$$
C \frac{7 n}{10}+C \frac{n}{5}+n
$$

## Proof by induction

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\begin{aligned}
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$$
C \frac{7 n}{10}+C \frac{n}{5}+n
$$

Pulling out $n$, we get

$$
n\left(C \frac{7}{10}+C \frac{1}{5}+1\right)
$$

$$
n\left(C \frac{9}{10}+1\right)
$$

$$
\leq C n
$$

## Proof by induction

$$
\begin{aligned}
& T(n)=T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n) \\
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$$
n\left(C \frac{7}{10}+C \frac{1}{5}+1\right)
$$

$$
n\left(C \frac{9}{10}+1\right)
$$

For which values of C ?

$$
\begin{gathered}
C \frac{9}{10}+1 \leq C \\
9 C+10 \leq 10 C \\
C \geq 10
\end{gathered}
$$

$$
\leq C n
$$

## Proof by induction

$$
\begin{aligned}
& T(n)=T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+O(n) \\
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We want to show that

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T\left(\frac{7 n}{10}\right)+T\left(\frac{n}{5}\right)+n \leq C n(\text { for some } C)
\end{gathered}
$$

By induction, since $\frac{1}{5} n<\frac{7}{10} n<n$, we have

$$
C \frac{7 n}{10}+C \frac{n}{5}+n
$$

Pulling out $n$, we get

$$
n\left(C \frac{7}{10}+C \frac{1}{5}+1\right)
$$

For which values of C ?

$$
\begin{gathered}
C \frac{9}{10}+1 \leq C \\
9 C+10 \leq 10 C \\
C \geq 10
\end{gathered}
$$

$$
n\left(C \frac{9}{10}+1\right)
$$

$$
\leq C n \quad(\text { as long as } C \geq 10)
$$

## Reporting running times: tight vs. loose bounds

- Usually in this class it should be pretty clear which bounds are appropriate from context
- For recursive algorithms, if you write down a recurrence relation, you should be able to find a bound for it
- If you think there is a tighter bound than the one you've written, you should try to prove it
- But you don't want to be doing original research, so keep it simple!


## Backtracking

- So far, we have seen cases where the next recursive call is clear
- In MergeSort, we need both left and right subarrays to be sorted
- In MOMSelect and BinarySearch, we guarantee the value we are looking for is in a specific subarray
- What if we can't tell which decision to make?
- Enter backtracking: When we need to make a decision, try one small step in all directions and evaluate all outcomes.


## N Queens

Problem statement: Given an $n \times n$ dimensional chessboard, place $n$ queens on the board such that none can attack each other.


Idea: Incrementally build a solution by placing one queen at a time!

Figure 2.1. Gauss's first solution to the 8 queens problem, represented by the array $[5,7,1,4,2,8,6,3]$
Given an arbitrary $n$, how can we decide where to place queens?

## N Queens

Idea: Incrementally build a solution by placing one queen at a time!

```
PlaceQueens(Q[1..n], r):
    If r = n+1:
        print Q[1..n]
    Else:
        for }j\leftarrow1\mathrm{ to n:
        legal }\leftarrow Tru
        for i}\leftarrow1\mathrm{ tor - 1:
            if(Q[i]=j) or
                (Q[i]=j+r-i) or
                (Q[i] = j - r):
                        legal }\leftarrow Fals
        if legal:
            Q[r]}\leftarrow
        PlaceQueens(Q[1..n],r+1)
```



Figure 2.3. The complete recursion tree of Gauss and Laquière's algorithm for the 4 queens problem.

## N Queens

Idea: Incrementally build a solution by placing one queen at a time!

```
PlaceQueens(Q[1..n], r):
    If r = n+1:
        print Q[1..n]
    Else:
        for }j\leftarrow1\mathrm{ to n:
        Q[r] = j
        if CheckLegal(Q[1..r]):
            PlaceQueens(Q[1..n], r+1)
```



## Backtracking pattern

Idea: Incrementally build a solution by placing one queen at a time!

- Appropriate when a sequence of incremental decisions can enumerate solutions
- Solution is often itself a sequence, e.g. $\mathrm{Q}[1 . . \mathrm{n}]$ is a sequence of queens placed in rows 1..n
- Exactly 1 decision is made at every step
- We usually need some information about previous decisions, but this should be as small as possible
- Problem is solved by recursive brute force, meaning we do not "prune" decisions that are obviously bad (leaves in the tree)


Figure 2.3. The complete recursion tree of Gauss and Laquière's algorithm for the 4 queens problem.

## Proof of correctness for dynamic programming

We prove a dynamic programming algorithm is correct by showing that the recursive specification is correct by induction.

- If it is easier for you to think through, you can write the recursive algorithm in pseudocode and prove it from there (since they are equivalent)

If we know the recursive specification is correct, then it is straightforward to explain that the iterative algorithm is also correct, since all it does is realize the recursive specification!

## General Notes and Strategies

- I am not trying to trick you, some questions have simple answers
- Recall what we have already done and start thinking from there
- Make sure you understand the problem before you start working on the solution
- If you are feeling frustrated, ask a question and/or take a break!
- Please sleep. You will not do better if you pull an all-nighter and there are no points for "most time spent staring at blank page"

Wrap Up

## Good luck!

