

Algorithms - Spring 25

Pearson  
Beet tracking



# Recap

- HW due Friday

↳ If algorithm is requested, 3 parts:

- pseudocode (+ description)
- correctness
- Runtime

- Teams of up to 3  
(sign up on gradescope)

# Last time: Recursion Trees

$$\rightarrow T(n) = r T\left(\frac{n}{c}\right) + f(n)$$

$r, c$  constant

level 0

1

2

⋮  
level i

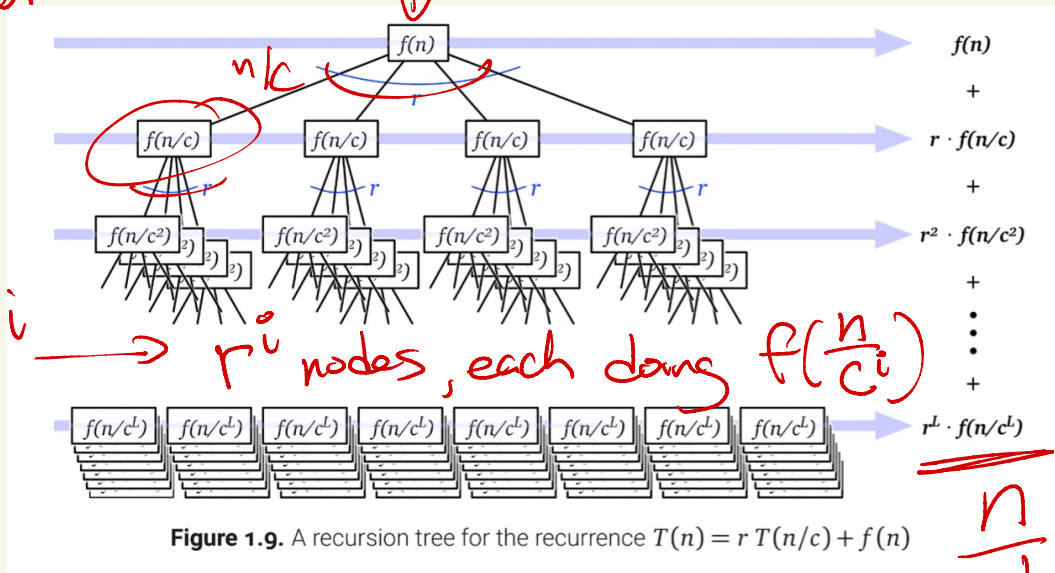


Figure 1.9. A recursion tree for the recurrence  $T(n) = r T(n/c) + f(n)$

To solve: Sum all work in the tree!

$$\sum_{\text{levels } i \text{ in tree}} (\text{work on level } i)$$

is this a geom. series?

$$= \sum_{i=0}^{\text{depth}} (\# \text{ nodes}) (\text{work per node})$$

$$\approx \sum_{i=0}^{\log n} r^i \cdot f\left(\frac{n}{c^i}\right)$$

# Master Theorem

$$T(n) = aT\left(\frac{n}{b}\right) + f(n) \Rightarrow$$

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \epsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ AND } af(n/b) < cf(n) \text{ for large } n \end{cases}$$

$f(n) \ll n^{\log_b a}$   
 $\epsilon > 0$   
 $c < 1$   
 $f(n) = n^{\log_b a}$   
 $f(n) \gg n^{\log_b a}$

Combining the three cases above gives us the following "master theorem".

**Theorem 1** The recurrence

$$\begin{aligned} T(n) &= aT(n/b) + cn^k \\ T(1) &= c, \end{aligned}$$

where  $a, b, c,$  and  $k$  are all constants, solves to:

$$\begin{aligned} T(n) &\in \Theta(n^k) \text{ if } a < b^k \\ T(n) &\in \Theta(n^k \log n) \text{ if } a = b^k \\ T(n) &\in \Theta(n^{\log_b a}) \text{ if } a > b^k \end{aligned}$$

## THEOREM 2

**MASTER THEOREM** Let  $f$  be an increasing function that satisfies the recurrence relation

$$f(n) = af(n/b) + cn^d \leftarrow \text{poly}$$

whenever  $n = b^k$ , where  $k$  is a positive integer,  $a \geq 1$ ,  $b$  is an integer greater than 1, and  $c$  and  $d$  are real numbers with  $c$  positive and  $d$  nonnegative. Then

$$f(n) \text{ is } \begin{cases} O(n^d) & \text{if } a < b^d, \\ O(n^d \log n) & \text{if } a = b^d, \\ O(n^{\log_b a}) & \text{if } a > b^d. \end{cases}$$

# Other examples

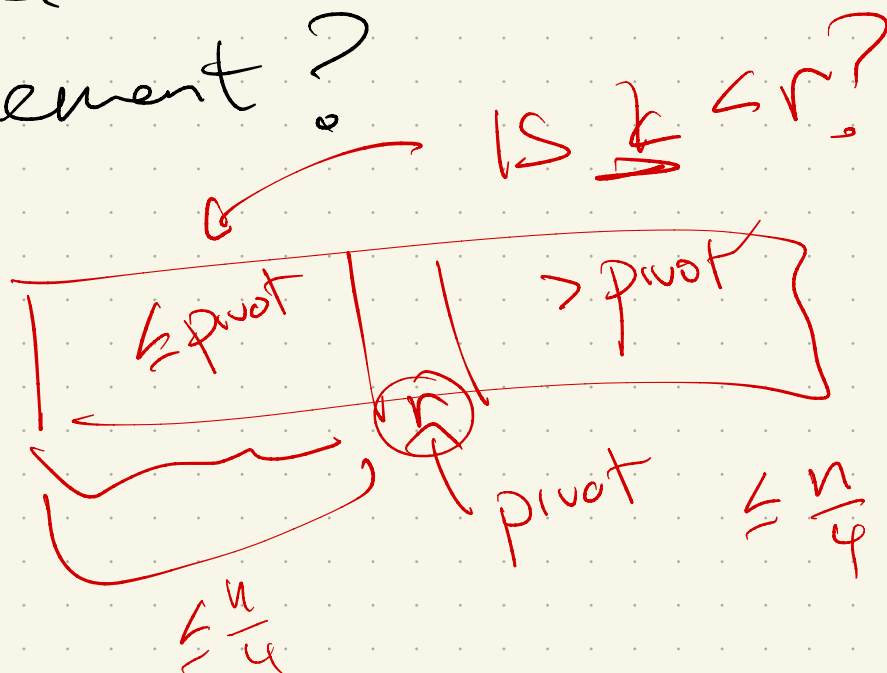
Medians: find "middle" element.

Two were covered:

```
QUICKSELECT(A[1..n], k):  
  if n = 1  
    return A[1]  
  else  
    Choose a pivot element A[p]  
    r ← PARTITION(A[1..n], p)  
    if k < r  
      return QUICKSELECT(A[1..r-1], k)  
    else if k > r  
      return QUICKSELECT(A[r+1..n], k-r)  
    else  
      return A[r]
```

Figure 1.12. Quickselect, or one-armed quicksort

Q: How do we know which side has the  $k^{\text{th}}$  element?



# Runtime

Still depends on pivot!

worst case:  
choose 1<sup>st</sup> or last  
element

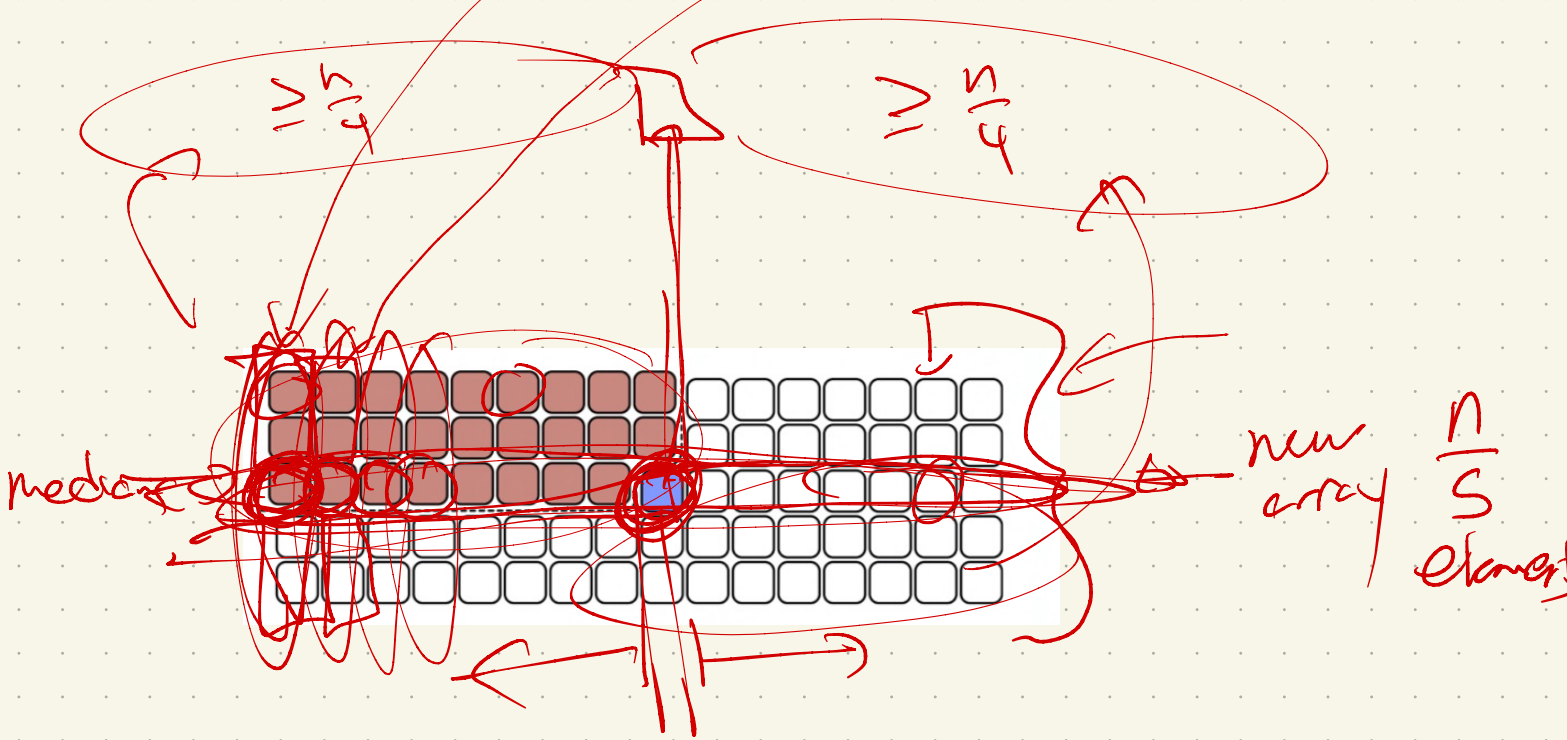
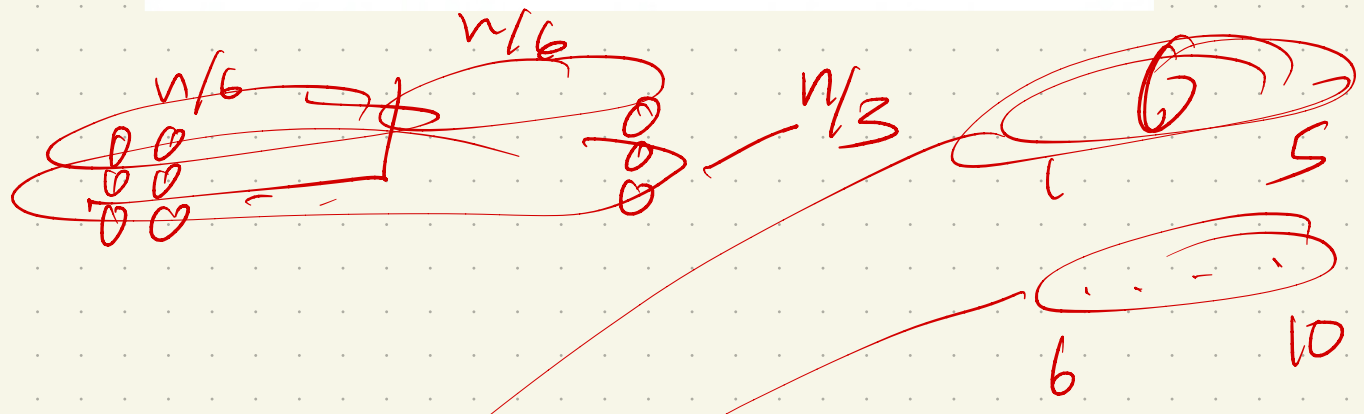
$$T(n) = \underbrace{(n-1)}_{\text{unrot}} + T(n-1)$$
$$= O(n^2)$$

# "Faster" version:

use for loop  
(still  $O(1)$  time)

```

MOMSELECT(A[1..n], k):
  if n ≤ 25  <<or whatever>>
    use brute force
  else
    m ← [n/5]
    for i ← 1 to m
      M[i] ← MEDIANOFFIVE(A[5i-4..5i])  <<Brute force!>>
    mom ← MOMSELECT(M[1..m], [m/2])  <<Recursion!>>
    r ← PARTITION(A[1..n], mom)
    if k < r
      return MOMSELECT(A[1..r-1], k)  <<Recursion!>>
    else if k > r
      return MOMSELECT(A[r+1..n], k-r)  <<Recursion!>>
    else
      return mom
  
```



# The recurrence:

```

MOMSELECT(A[1..n], k):
  if n ≤ 25  ((or whatever))
    use brute force
  else
    m ← ⌊n/5⌋
    for i ← 1 to m
      M[i] ← MEDIANOFIVE(A[5i-4..5i])  ((Brute force!))
    mom ← MOMSELECT(M[1..m], ⌊m/2⌋)  ((Recursion!))
    r ← PARTITION(A[1..n], mom)
    if k < r
      return MOMSELECT(A[1..r-1], k)  ((Recursion!))
    else if k > r
      return MOMSELECT(A[r+1..n], k-r)  ((Recursion!))
    else
      return mom
  
```

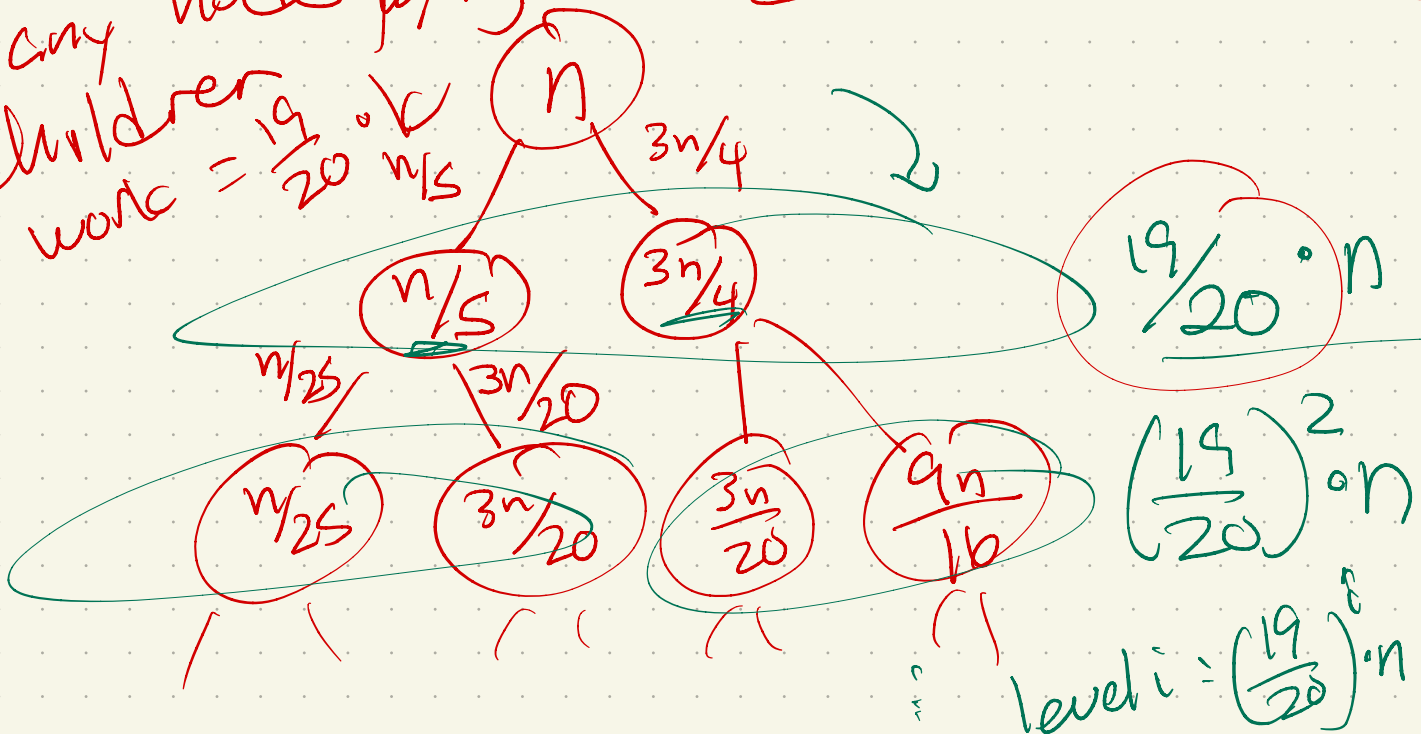
$O(n)$

$O(n)$

$T(n)$  = runtime for  $n$  elt list

$$\leq O(n) + T\left(\frac{n}{5}\right) + T\left(\frac{3n}{4}\right)$$

for any node  $p/k_j$   
 2 children  
 work =  $\frac{19}{20} \cdot k_j$





Tree has depth

$$\log_5 n \leq d \leq \log_{4/3} n$$

↑  
divide  
by 5  
(left branch)

↑  
multiply  
by  $3/4$   
↳ div. by  $4/3$

$$T(n) = \sum_{i=0}^{\log n} (\text{work per level})$$

$$\approx \sum_{i=0}^{\log n} \left(\frac{19}{20}\right)^i \cdot n$$

$r < 1$

(geom series idem.)

$$\leq n \sum_{i=0}^{\infty} \left(\frac{19}{20}\right)^i = n \left( \frac{1}{1 - \frac{19}{20}} \right) = O(n) \quad \square$$

Why? Recall quicksort:

$$T(n) = \max_r \left\{ \underbrace{T(r) + T(n-r)}_{\text{2 calls}} + \underbrace{O(n)}_{\text{pivot code}} \right\}$$

Instead of random pivot,  
call median:

$$T(n) = \underbrace{O(n)}_{\text{call mom}} + \underbrace{O(n)}_{\text{pivot}} + 2T\left(\frac{n}{2}\right)$$

"Quicksort" in  $n \log n$

Multiplication:  $2m$  digit #

Known "fact":

$$(10^m a + b)(10^m c + d) =$$

$$10^{2m} ac + 10^m (bc + ad) + bd$$

Example:

$$102568 \times 358691$$

$$= (102 \times 10^3 + 568) \times (358 \times 10^3 + 691)$$

$\Rightarrow$

↳ Why does this suggest recursion??

$\nwarrow$  multiply smaller #s!

# The algorithm:

```
SPLITMULTIPLY(x, y, n):  
  if n = 1  
    return x · y  
  else  
    m ← ⌊n/2⌋  
    a ← ⌊x/10m⌋; b ← x mod 10m    ⟨⟨x = 10ma + b⟩⟩  
    c ← ⌊y/10m⌋; d ← y mod 10m    ⟨⟨y = 10mc + d⟩⟩  
    e ← SPLITMULTIPLY(a, c, m) ← a · c  
    f ← SPLITMULTIPLY(b, d, m) ← b · d  
    g ← SPLITMULTIPLY(b, c, m) ← b · c  
    h ← SPLITMULTIPLY(a, d, m) ← a · d  
    return 102me + 10m(g + h) + f
```

additions + 0-padding

## Runtime:

$$T(m) = O(1) + 4T\left(\frac{m}{2}\right)$$

Master thm:  $f(n) = O(1) = n^0$   
 $d = 0$

$$a = 4$$

$$b = 2$$

$$a = 4 \Rightarrow b^d = 2^0 = 1$$

$$T(m) = m \log_2^4 = m^2$$

A better trick  $ac - bc + bd - ad$

$$\underbrace{ac + bd}_{e \quad f} - \underbrace{(a-b)(c-d)}_{bc + ad} \quad \text{Why?}$$

```

FASTMULTIPLY(x, y, n):
  if n = 1
    return x · y
  else
    m ← ⌊n/2⌋
    a ← ⌊x/10m⌋; b ← x mod 10m    ⟨⟨x = 10ma + b⟩⟩
    c ← ⌊y/10m⌋; d ← y mod 10m    ⟨⟨y = 10mc + d⟩⟩
    e ← FASTMULTIPLY(a, c, m)
    f ← FASTMULTIPLY(b, d, m)
    g ← FASTMULTIPLY(a - b, c - d, n)
    return 102me + 10m(e + f - g) + f
  
```

$bc + ad$

Runtime:

$$T(n) = O(n) + 3T\left(\frac{n}{2}\right)$$

$n^0, \text{ so } d=0$        $a=3$        $b=2$

MT:  $3 > 2^0$

$$T(n) = \Theta(n^{\log_2 3})$$

$1.6 \log_2 3 < 2$

(additions or bit tricks)

Exponentiation:

Still open!

(Amazing, right??)

The algorithms do very well:

- to compute  $a^n$ ,  
need  $O(\log n)$   
multiplications

However, doesn't achieve  
lowest possible for  
every value - it's just  
with a constant!

## Ch 2: Back tracking:

Many of you saw in AI,  
apparently!

(Don't worry if not...)

Why we discuss:

It's really recursion ~~etc~~  
(again)!

Also really a form of  
brute force:

try everything recursively,  
↓ see what works.

↳ dyn. programming

# N Queens

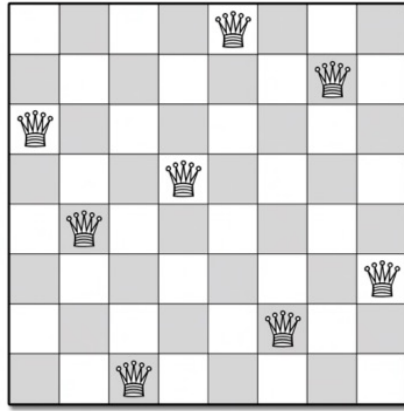


Figure 2.1. Gauss's first solution to the 8 queens problem, represented by the array [5, 7, 1, 4, 2, 8, 6, 3]

Issue: representation!

His choice: one per row,  
so store index of queen  
on rows in array.

Now, how to solve:

brute force! Place a  
queen + keep going.

If you get stuck,  
"unplace" last queen  
+ back up.



The tree (b/c pretty) :)

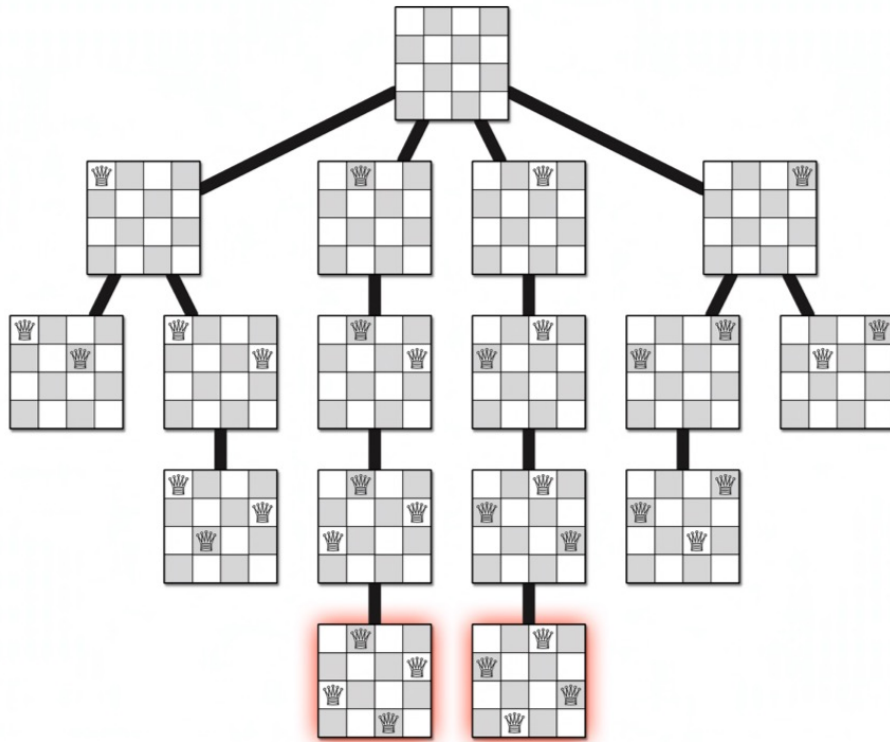


Figure 2.3. The complete recursion tree of Gauss and Laquière's algorithm for the 4 queens problem.

Problem (a hard part):  
Formalizing this in code.  
Sketch:

# Result:

```
PLACEQUEENS(Q[1..n], r):  
  if  $r = n + 1$   
    print Q[1..n]  
  else  
    for  $j \leftarrow 1$  to  $n$   
      legal  $\leftarrow$  TRUE  
      for  $i \leftarrow 1$  to  $r - 1$   
        if  $(Q[i] = j)$  or  $(Q[i] = j + r - i)$  or  $(Q[i] = j - r + i)$   
          legal  $\leftarrow$  FALSE  
      if legal  
        Q[r]  $\leftarrow$  j  
        PLACEQUEENS(Q[1..n], r + 1)    ⟨⟨Recursion!⟩⟩
```

**Figure 2.2.** Gauss and Laquière's backtracking algorithm for the  $n$  queens problem.

Runtime:

$$Q(n) =$$

# Game Trees:

a way to model moves in  
2-player games

Assume:

- No randomness so the game is just 2 people taking turns

Ex: Checkers, chess, Nim, Go  
— (not settlers of Catan!)

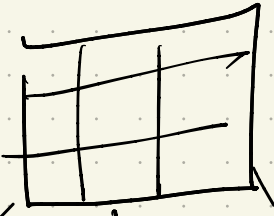
- "Perfect" players:

Makes rational decisions, +  
if there is a move to get  
them to a win state, they  
do it!

Idea: Track current state of the game, as play occurs

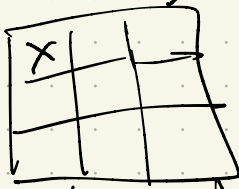
Tic-tac-toe

1st player:  
play an 'x'



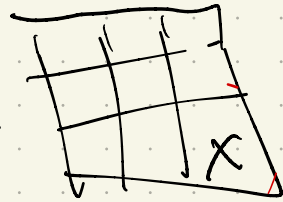
...

2nd player:  
put 'o'

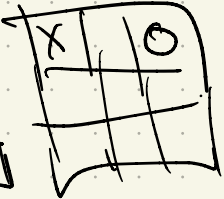
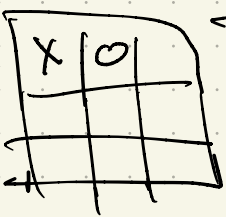


...

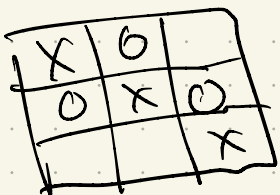
...



1st



1st player:  
put 'x'



leaf:  
good for player 1  
bad for player 2

Model every possible move.

A state is good for player 1 if they either have won, or could move to a bad state for player 2.

and bad if they have lost, or if all possible moves lead to a state that is good for player 2.

Think from the bottom up:

# Tic-tac-toe again:

2's turn 

X	O	X
O	X	

 good or bad?

1's turn 

X	O	X
O	X	O

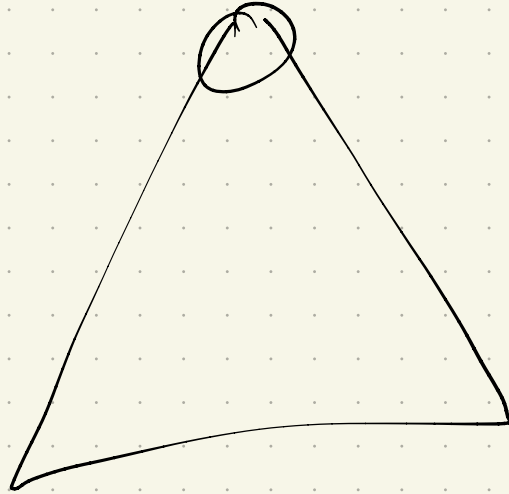
X	O	X
O	X	O
		X

  
good for 1  
bad for 2

This is  
good for 1.  
(He can move  
some where  
bad for 2)

So:

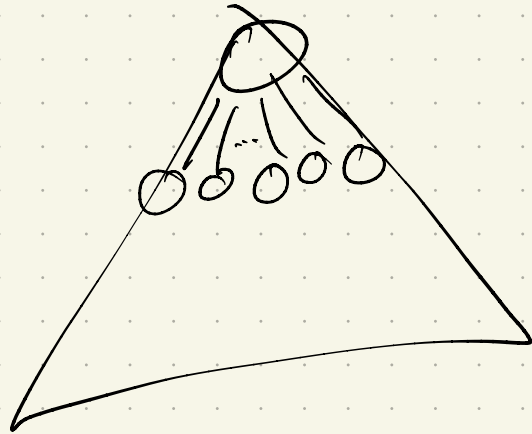
good:  
I have a  
child who  
other guy  
thinks is  
bad:



Result:

Bad

All →  
of these  
are good  
for other guy



Result:



## Downsides:

Game trees are HUGE!

Tic-tac-toe: over 200,000 leaves.

People can still "predict":  
we're good at inferring  
state/strategy intuitively,  
with practice

Computers have to search.

Hence - took 60 years to  
get a decent computer  
chess player! Need  
"heuristics" (aka guesses)  
to make it work.

Game theory — a bit more complicated.

Here, we assume clear win vs. lose

Game theory suggests more subtle possibilities, as well as simultaneous moves & "randomness".

#### Example: Odds and Evens

Consider the simple game called **odds and evens**. Suppose that player 1 takes evens and player 2 takes odds. Then, each player simultaneously shows either one finger or two fingers. If the number of fingers matches, then the result is *even*, and player 1 wins the bet (\$2). If the number of fingers does not match, then the result is *odd*, and player 2 wins the bet (\$2). Each player has two possible strategies: show one finger or show two fingers. The *payoff matrix* shown below represents the payoff to player 1.

*Payoff Matrix*

Strategy		Player 2	
		1	2
Player 1	1	2	-2
	2	-2	2

Even if both know outcomes, result is unclear!

## Example: Subset Sum

Given a set  $X$  of positive integers and a target value  $t$ , is there a subset of  $X$  which sums to  $t$ ?

Ex:  $X = \{8, 6, 7, 3, 10, 5, 9\}$

$$t = 15$$

How would we solve?

Consider one at a time:

$$X = \{8, 6, 7, 5, 3, 1, 9\}$$

Formalize this: recursion!

↳ base case?

Algorithm:

reset to use  
arrays.

⟨⟨Does any subset of  $X$  sum to  $T$ ?⟩⟩

SUBSETSUM( $X, T$ ):

if  $T = 0$

return TRUE

else if  $T < 0$  or  $X = \emptyset$

return FALSE

else

$x \leftarrow$  any element of  $X$

$with \leftarrow$  SUBSETSUM( $X \setminus \{x\}, T - x$ )   ⟨⟨Recurse!⟩⟩

$wout \leftarrow$  SUBSETSUM( $X \setminus \{x\}, T$ )   ⟨⟨Recurse!⟩⟩

return ( $with \vee wout$ )

⟨⟨Does any subset of  $X[1..i]$  sum to  $T$ ?⟩⟩

SUBSETSUM( $X, i, T$ ):

if  $T = 0$

return TRUE

else if  $T < 0$  or  $i = 0$

return FALSE

else

$with \leftarrow$  SUBSETSUM( $X, i - 1, T - X[i]$ )   ⟨⟨Recurse!⟩⟩

$wout \leftarrow$  SUBSETSUM( $X, i - 1, T$ )   ⟨⟨Recurse!⟩⟩

return ( $with \vee wout$ )

Correctness: inductive proof,  
on size of  $X, i$

Base cases:

$i = |X| = 0$  (so  $X = \{\}$ ):

Ind Hyp: works for  $X[1..n-1]$   
or smaller values of  $T$

Ind step: Full array  $X[1..n]$

Consider  $X[n]$ :

Runtime:

