```
    Last day
of Midterm
2 Material
```


## 61A Lecture 23

Friday, October 19

## Trees with Internal Node Values

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Trees can have values at their roots as well as their leaves.

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class Tree(object):

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    def ___init__(self, entry, left=None, right=None):
    self.entry = entry
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| $\qquad$ init $\qquad$ (self, entry, left=None, right=None): self.entry = entry <br> self.left = left <br> Valid if left and right $\qquad$ are each either None or a Tree instance <br> fib_tree(n): <br> A valid tree cannot <br> be a subtree of itself (no cycles!) |  |
| :---: | :---: |
|  |  |
|  |  |

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    def __init__(self, entry, left=None, right=None):
        self.entry = entry
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        are each either None or
        a Tree instance
    A valid tree cannot
        be a subtree of
    itself (no cycles!)
    if n == 2:
        return Tree(1)
    left = fib_tree(n-2)
    right = fib_tree(n-1)
    return Tree(left.entry + right.entry, left, right)
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def fib_tree(n):
    if n == 1:
        return Tree(0)
    if n == 2:
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    left = fib_tree(n-2)
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factors = 0
for k in range(1, n+1):
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        factors += 1
    return factors
    sqrt_n = sqrt(n)
    k, factors = 1, 0
    while k < sqrt_n:
    if n % k == 0:
        factors += 2
        k += 1
    if k * k == n:
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Each step of evaluation has a set of active environments.
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Active environments:

- Environments for any statements currently being executed
- Parent environments of functions named in active environments


## Fibonacci Memory Consumption



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Has an active environment


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means that there are positive constants $k_{1}$ and $k_{2}$ such that

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k_{1} \cdot f(n) \leq R(n) \leq k_{2} \cdot f(n)
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for sufficiently large values of $\boldsymbol{n}$.

## Iteration vs Memoized Tree Recursion

Iterative and memoized implementations are not the same.

```
```

def fib_iter(n):

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prev, curr = 1, 0
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for $\quad$ in range $(\mathrm{n}-1)$ :
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for - in range(n-1):
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return curr

```
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```
@memo
```

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def fib(n):
def fib(n):
if n == 1:
if n == 1:
return 0
return 0
if n == 2:
if n == 2:
return 1
return 1
return fib(n-2) + fib(n-1)

```
```

    return fib(n-2) + fib(n-1)
    ```
```

Space

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def fib_iter(n):
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    for _ in range (n-1):
        prev, curr \(=\) curr, prev + curr
    return curr
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def fib(n):
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Time Space
\[
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    factors = 0
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sqrt_n = sqrt(n)
k , factors \(=1,0\)
while k < sqrt_n:
    if \(\mathrm{n} \% \mathrm{k}==0\) :
            factors += 2
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    if \(\mathrm{k} * \mathrm{k}=\mathrm{n}\) :
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b^{n}= \begin{cases}1 & \text { if } n=0 \\ \left(b^{\frac{1}{2} n}\right)^{2} & \text { if } n \text { is even } \\ b \cdot b^{n-1} & \text { if } n \text { is odd }\end{cases}
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Goal: one more multiplication lets us double the problem size.
```

def exp(b, n):
if n == 0:
return 1
def square(x):
return x*x
def square(x): return $x$ *x

```
    return b * \(\exp (b, n-1)\)
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def exp(b, n):
if n == 0:
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def square(x):
return x*x
def fast_exp(b, n):

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def square(x):
return x*x
def fast_exp(b, n):
if n == 0:
return 1

```
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def exp(b, n):
if n == 0:
return 1
b}={\begin{array}{ll}{1}\&{\mathrm{ if }n=0}<br>{b\cdot\mp@subsup{b}{}{n-1}}\&{\mathrm{ otherwise}}
return b * exp(b, n-1)
def square(x):
return x*x
def fast_exp(b, n):
b}={\begin{array}{ll}{1}\&{\mathrm{ if }n=0}<br>{(\mp@subsup{b}{}{\frac{1}{2}n}\mp@subsup{)}{}{2}}\&{\mathrm{ if }n\mathrm{ is even }}<br>{b\cdot\mp@subsup{b}{}{n-1}}\&{\mathrm{ if }n\mathrm{ is odd }}
if n == 0:
return 1
if n % 2 == 0:
return square(fast_exp(b, n//2))

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if n == 0:
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else:
return b * fast_exp(b, n-1)

```

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def exp(b, n):
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def square(x):
    return \(x^{*} x\)
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    if \(n\) \% 2 == 0:
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    else:
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Comparing orders of growth ( n is the problem size)

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\(\Theta\left(b^{n}\right)\)

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\(\Theta\left(b^{n}\right) \quad\) Exponential growth! Recursive fib takes
\(\Theta\left(\phi^{n}\right)\) steps, where \(\phi=\frac{1+\sqrt{5}}{2} \approx 1.61828\)

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Incrementing the problem scales \(R(n)\) by a factor.

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Incrementing the problem scales \(R(n)\) by a factor.
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Incrementing \(n\) increases \(R(n)\) by the problem size \(n\).
\(\Theta(n)\) Linear growth. Resources scale with the problem.
\(\Theta(\log n) \quad\) Logarithmic growth. These processes scale well.

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\(\Theta\left(n^{2}\right) \quad\) Quadratic growth. E.g., operations on all pairs.
Incrementing \(n\) increases \(R(n)\) by the problem size \(n\).
\(\Theta(n)\) Linear growth. Resources scale with the problem.
\(\Theta(\log n) \quad\) Logarithmic growth. These processes scale well.
Doubling the problem only increments \(R(n)\).

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