# Exponentiation by Squaring

April 2, 2020

#### Concepts to take away

- Usefulness of divide and conquer algorithms
- How the exponentiation function is implemented
- How seemingly disjoint problems are connected

#### Short slide on Divide and Conquer

Reduce repeated work by recursively splitting the problem.

- 1. Splitting the problem
- 2. Combining the solutions of the two sub-problems

#### Introduction

$$2^{8} = \overbrace{2 \times 2 \times \cdots \times 2 \times 2}^{8}$$

$$a^{n} = \overbrace{a \times a \times \cdots \times a \times a}^{n}$$

$$def pow(a, n):$$
# ???

#### Iterative Solution

```
a^n = 1 \times \overbrace{a \times a \cdots a \times a}^n

def pow(a, n):
    ret = 1
    for i in range(n):
        ret *= a
    return ret
```

#### Iterative Solution

```
a^{n} = 1 \times \overbrace{a \times a \cdots a \times a}^{n}
def pow(a, n):
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```

There are n multiplications here so the function is proportional to input size n.

$$2^8 = \overbrace{2 \cdot 2 \cdot 2}^{8}$$

$$2^{8} = \underbrace{2 \cdot 2 \cdot 2}_{8}$$

$$2^{8} = \underbrace{2 \cdot 2 \cdot 2}_{4}$$

We can use this observation to reduce the computational speed!

## Developing the Equation

$$2^{8} = (2 \cdot 2)^{\frac{8}{2}} = 4^{4}$$
  $(x^{2})^{\frac{n}{2}} = x^{2 \cdot \frac{n}{2}} = x^{n}$ 

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=  $(4 \cdot 4)^{\frac{4}{2}} = 16^{2}$ 

## Developing the Equation

$$2^{8} = (2 \cdot 2)^{\frac{8}{2}} = 4^{4} \qquad (x^{2})^{\frac{n}{2}} = x^{2 \cdot \frac{n}{2}} = x^{n}$$
$$= (4 \cdot 4)^{\frac{4}{2}} = 16^{2}$$
$$= 16 \cdot 16 = 4096$$

Only 3 multiplications needed!

### What about odd exponents?

$$3^{15} = \overbrace{3 \times 3 \times \cdots \times 3 \times 3}^{14} \times 3$$

$$3^{15} = (3 \cdot 3)^{\frac{14}{2}} \cdot 3 = 9^{7} \cdot 3$$

$$(x^{2})^{\frac{n-1}{2}} \cdot x = x^{n}$$

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$$= (9 \cdot 9)^{\frac{6}{2}} \cdot 9 \cdot 3 = 81^{3} \cdot 9 \cdot 3$$

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$$= 81 \cdot 81 \cdot 81 \cdot 9 \cdot 3 = 14348907$$

$$(x^{2})^{\frac{n-1}{2}} \cdot x = x^{n}$$

### All together now

$$a^n = \begin{cases} (a^2)^{\frac{n}{2}} & n \text{ is even} \\ (a^2)^{\frac{n-1}{2}} \times a & n \text{ is odd} \end{cases}, \qquad a^0 = 1$$

Intuitively, n is being halved in every call - resulting in  $\log n$  multiplications to be done.

#### Recursive Code

```
a^{n} = \begin{cases} (a^{2})^{\frac{n}{2}} & n \text{ is even} \\ (a^{2})^{\frac{n-1}{2}} \times a & n \text{ is odd} \end{cases}, \quad a^{0} = 1
\text{def pow(a, n):} \\ \text{if n == 0: return 1} \\ \text{if n % 2 == 1:} \\ \text{return pow(a * a, n // 2) * a} \\ \text{else:} \\ \text{return pow(a * a, n / 2)}
```

### Fibonacci Sequence

$$f(n) = \begin{cases} 0 & n = 0 \\ 1 & n = 1 \\ f(n-1) + f(n-2) & n > 1 \end{cases}$$

```
def fib(n):
    if n <= 1: return n
    return fib(n - 1) + fib(n - 2)</pre>
```

The number of additions in this function is approximately  $2^n$ .\*

### Can we go faster?

```
def fib(n):
    y = 0
    x = 1
    for i in range(n):
        y, x = x, x + y
    return y
```

In this example, y holds the value of f(n), x holds the value of f(n+1). Each iteration updates x and y according to the equations above. This implementation involves n additions, which is  $<< 2^n$ .

#### Review of Matrices

A matrix is a 2-D rectangular array of numbers.

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

A vector is a 1-D rectangular array of numbers.

$$v = \begin{bmatrix} x \\ y \end{bmatrix}$$

You can multiply them together.

$$Av = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \end{bmatrix}$$
 Matrix \* Vector 
$$AB = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x & p \\ y & q \end{bmatrix} = \begin{bmatrix} ax + by & ap + bq \\ cx + dy & cp + dq \end{bmatrix}$$
 Matrix \* Matrix

## More Matrix Properties

$$A(BC) = (AB)C$$

$$A(Bv) = (AB)v$$

$$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$AI = IA = A$$

Associative Property
Associative Property
Identity Matrix

### Modelling Recursive Functions Using Matrices

$$\overbrace{\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}}^{T} \overbrace{\begin{bmatrix} x \\ y \end{bmatrix}}^{v^{[i]}} = \overbrace{\begin{bmatrix} x+y \\ x \end{bmatrix}}^{v^{[i+1]}}$$

We define a **transition matrix**,  $T = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$ , to go from the i iteration to i+1 iteration, or from  $v^{[i]}$  to  $v^{[i+1]}$ .

### Modelling Recursive Functions Using Matrices

$$\underbrace{\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}}_{v_i} \underbrace{\begin{bmatrix} x \\ y \end{bmatrix}}_{v_i} = \underbrace{\begin{bmatrix} x + y \\ x \end{bmatrix}}_{v_i}$$

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### Some examples

Start with 
$$T = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$
,  $v^{[0]} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ . 
$$v^{[1]} = Tv^{[0]} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1+0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$v^{[2]} = Tv^{[1]} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1+1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$
$$v^{[3]} = Tv^{[2]} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 2+1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$
$$v^{[4]} = Tv^{[3]} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 3+2 \\ 3 \end{bmatrix} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

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Highlighted numbers are f(n).

$$v^{[1]} = Tv^{[0]}$$

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 $v^{[2]} = Tv^{[1]} = T \cdot Tv^{[0]}$ 

$$v^{[1]} = Tv^{[0]}$$
 $v^{[2]} = Tv^{[1]} = T \cdot Tv^{[0]}$ 
 $v^{[3]} = Tv^{[2]} = T \cdot T \cdot Tv^{[0]}$ 

### Computing Fibonacci

```
f(n) = v_v^{[n]}, v^{[n]} = T^n v^{[0]}
def pow(a, n):
    if n == 0: return np.eye(2) # Identity
    if n % 2 == 1:
         return pow(a * a, n // 2) * a
    else:
         return pow(a * a, n / 2)
def fib(n):
    T = np.matrix([[1,1], [1,0]])
    v_0 = np.matrix([[1], [0]])
    T_n = pow(T, n)
    v n = T n * v 0
    return v_n[1,0] # Returning the second value
```

#### Computing Fibonacci

```
def fib(n):
    T = np.matrix([[1,1], [1,0]])
    v_0 = np.matrix([[1], [0]])
    T_n = pow(T, n) # Log n operation
    v_n = T_n * v_0
    return v_n[1,0] # Returning the second value
```

The number of multiplications for matrix multiplication is  $dim^3$  but dimension is constant here.

Thus, the computation is on order of  $\log n$ .

This strategy works for any linear recurrence!

## Extra: Deriving Binet's Formula

$$f(n) = \frac{\left(\frac{1+\sqrt{5}}{2}\right)^n - \left(\frac{1-\sqrt{5}}{2}\right)^n}{\sqrt{5}}$$

Diagonalization of  $T = PDP^{-1}$ .  $T^n = PD^nP^{-1}$ 

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \frac{1+\sqrt{5}}{2} & \frac{1-\sqrt{5}}{2} \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{1+\sqrt{5}}{2} & 0 \\ 0 & \frac{1-\sqrt{5}}{2} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{1-\sqrt{5}}{2\sqrt{5}} \\ -\frac{1}{\sqrt{5}} & \frac{1+\sqrt{5}}{2\sqrt{5}} \end{bmatrix} \\
\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n = \begin{bmatrix} \frac{1+\sqrt{5}}{2} & \frac{1-\sqrt{5}}{2} \\ 1 & 1 \end{bmatrix} \begin{bmatrix} (\frac{1+\sqrt{5}}{2})^n & 0 \\ 0 & (\frac{1-\sqrt{5}}{2})^n \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{1-\sqrt{5}}{2\sqrt{5}} \\ -\frac{1}{\sqrt{5}} & \frac{1+\sqrt{5}}{2\sqrt{5}} \end{bmatrix} \\
= \begin{bmatrix} \frac{(\frac{1+\sqrt{5}}{2})^{n+1} - (\frac{1-\sqrt{5}}{2})^{n+1}}{\sqrt{5}} & (\cdots) \\ \frac{(\frac{1+\sqrt{5}}{2})^n - (\frac{1-\sqrt{5}}{2})^n}{\sqrt{5}} & (\cdots) \\ \frac{1+\sqrt{5}}{2} & \frac{1-\sqrt{5}}{2} & (\cdots) \end{bmatrix} = \begin{bmatrix} f(n+1) & (\cdots) \\ f(n) & (\cdots) \end{bmatrix}$$