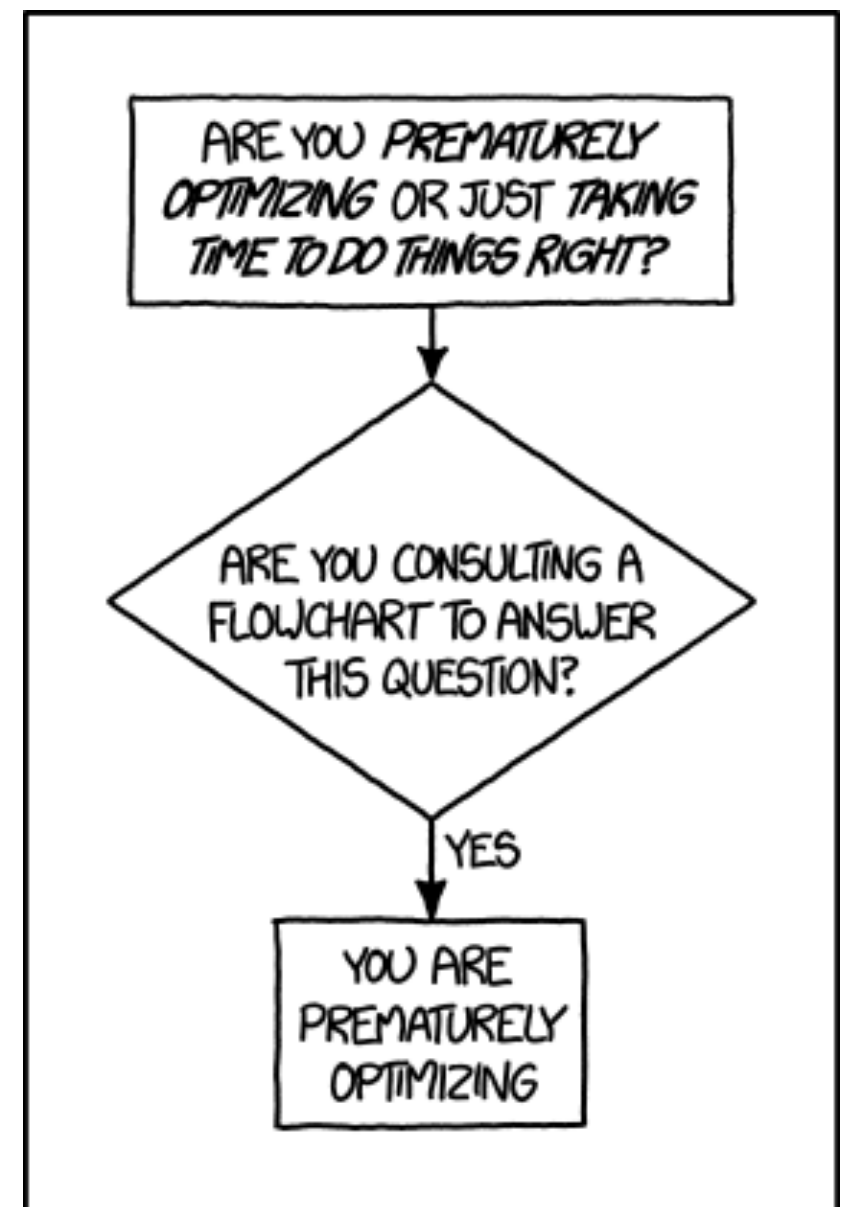


# Applications of Mathematics in Computer Science (MACS)

## LECTURE #4: ALGORITHMIC DIFFERENTIATION



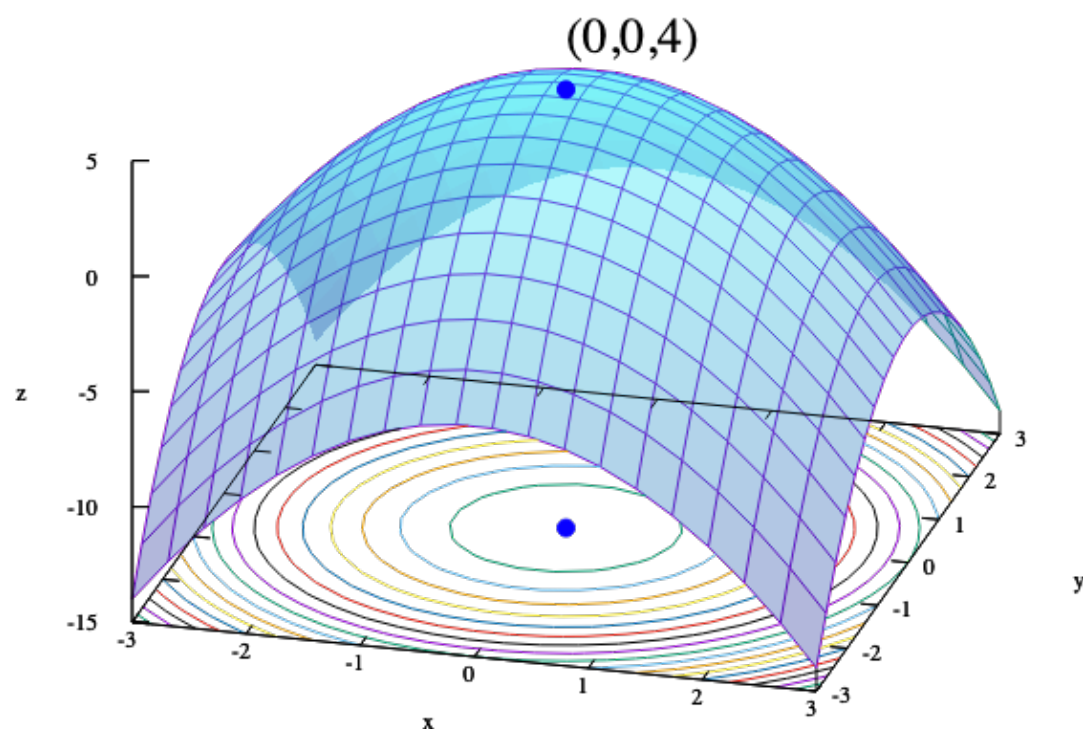
# The CS Problem:

Optimization: finding the **best** solution

- **shortest** path in a graph
- **Most economical** packaging
- **Most effective** vaccination strategy

# Optimization

- **Given:** a function  $f : A \rightarrow \mathbb{R}$
- **Find:**  $x_0 \in A$  such that:
  - $f(x_0) \leq f(x)$  for all  $x \in A$  (minimization)
  - or
  - $f(x) \leq f(x_0)$  for all  $x \in A$  (maximization)



$$z = -(x^2 + y^2) + 4$$

$$\max_{(x,y)} z = (0,0)$$

# How to solve an optimization problem?

## Stochastic optimization

try many values randomly



# How to solve an optimization problem?

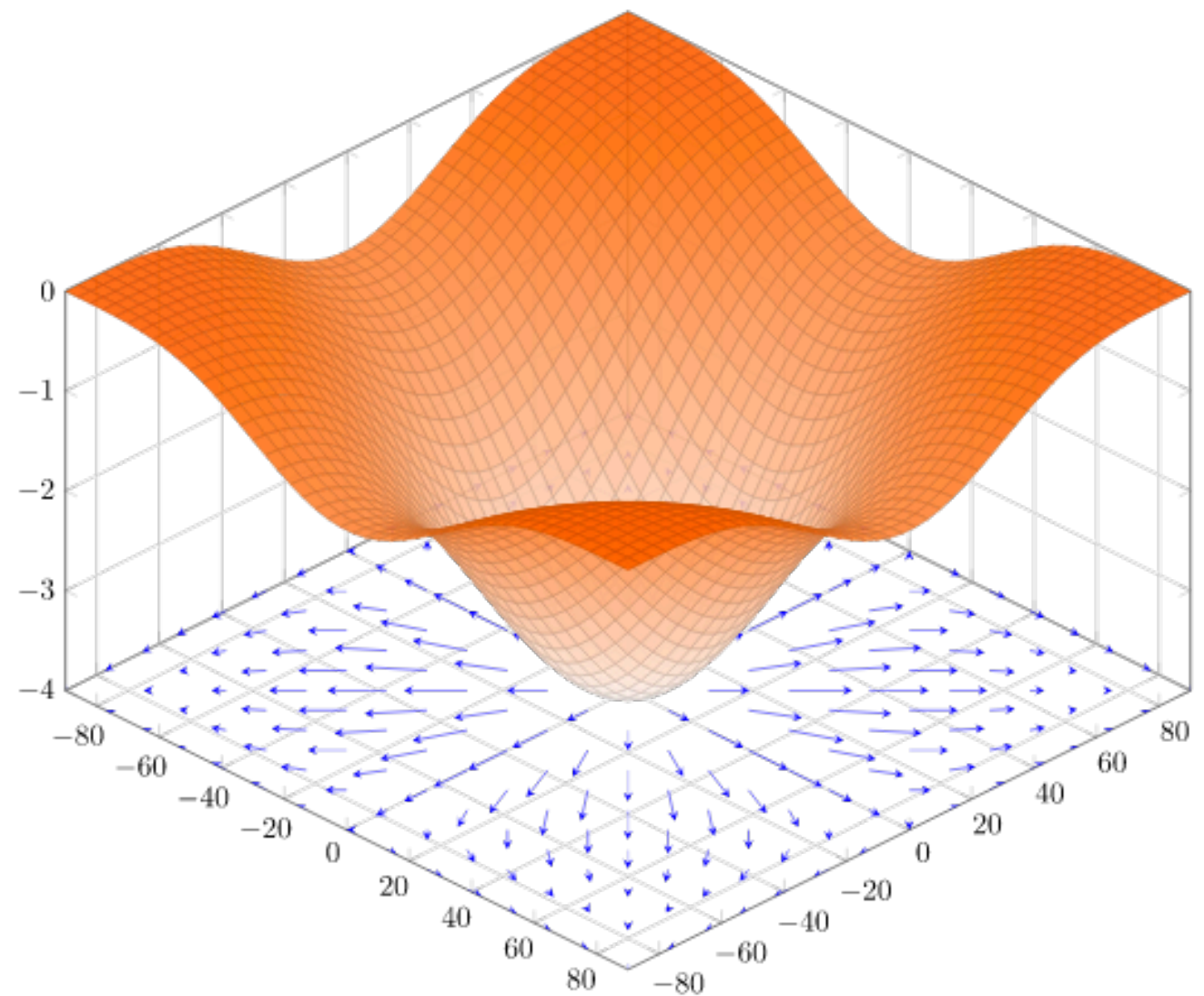
## Gradient-based optimization

Differentiation based

Gradient:

direction of fastest ascent

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \end{pmatrix}$$



# How to solve an optimization problem?

## Gradient descent

With  $\nabla f(x, y, \dots)$  we can optimize much faster

Repeat:

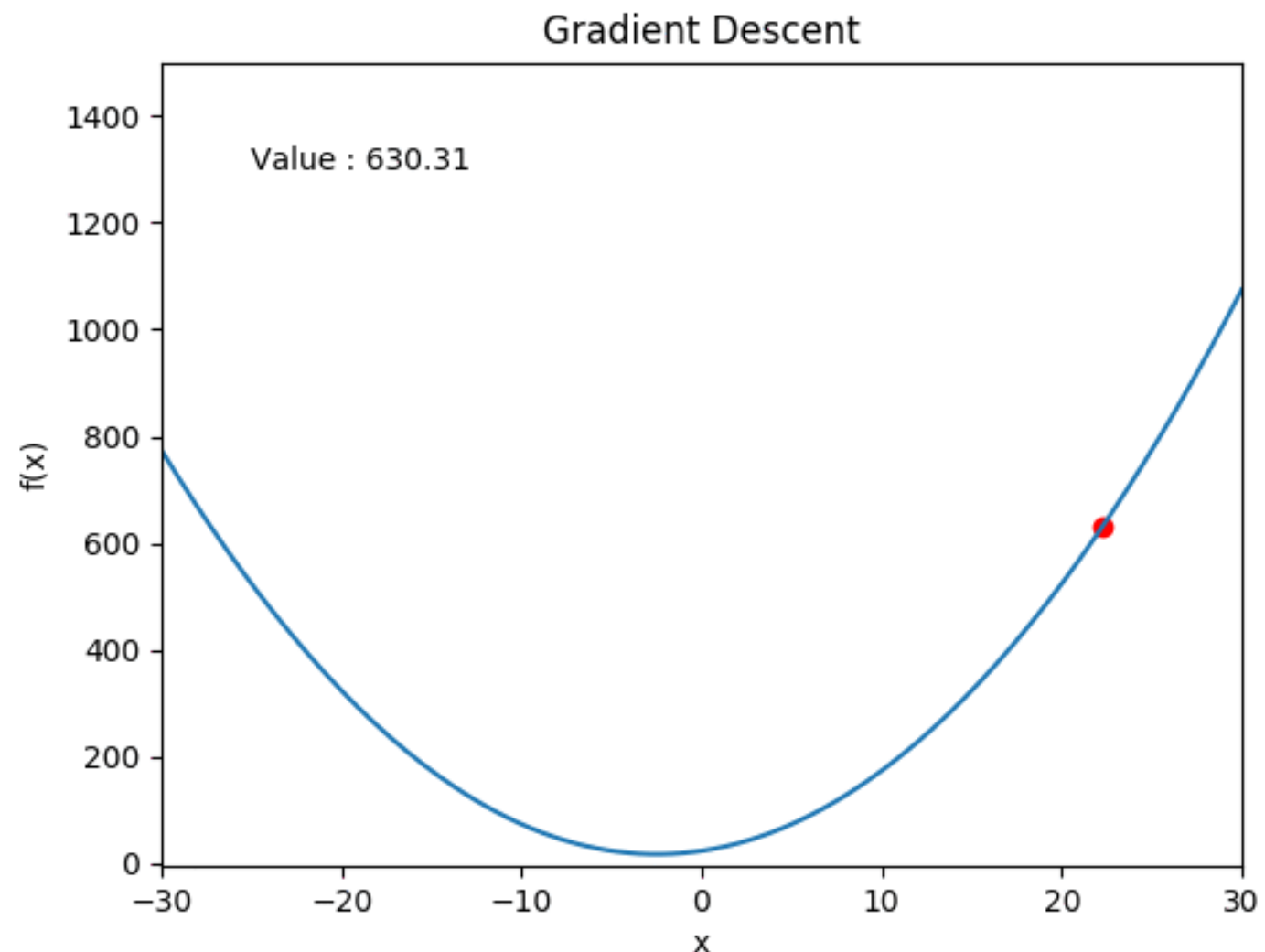
- $x \leftarrow x - \delta \nabla f(x)$

- $\delta \leftarrow \gamma \delta$

for:

- $\delta$  is small

- $\gamma = 1 - \varepsilon$



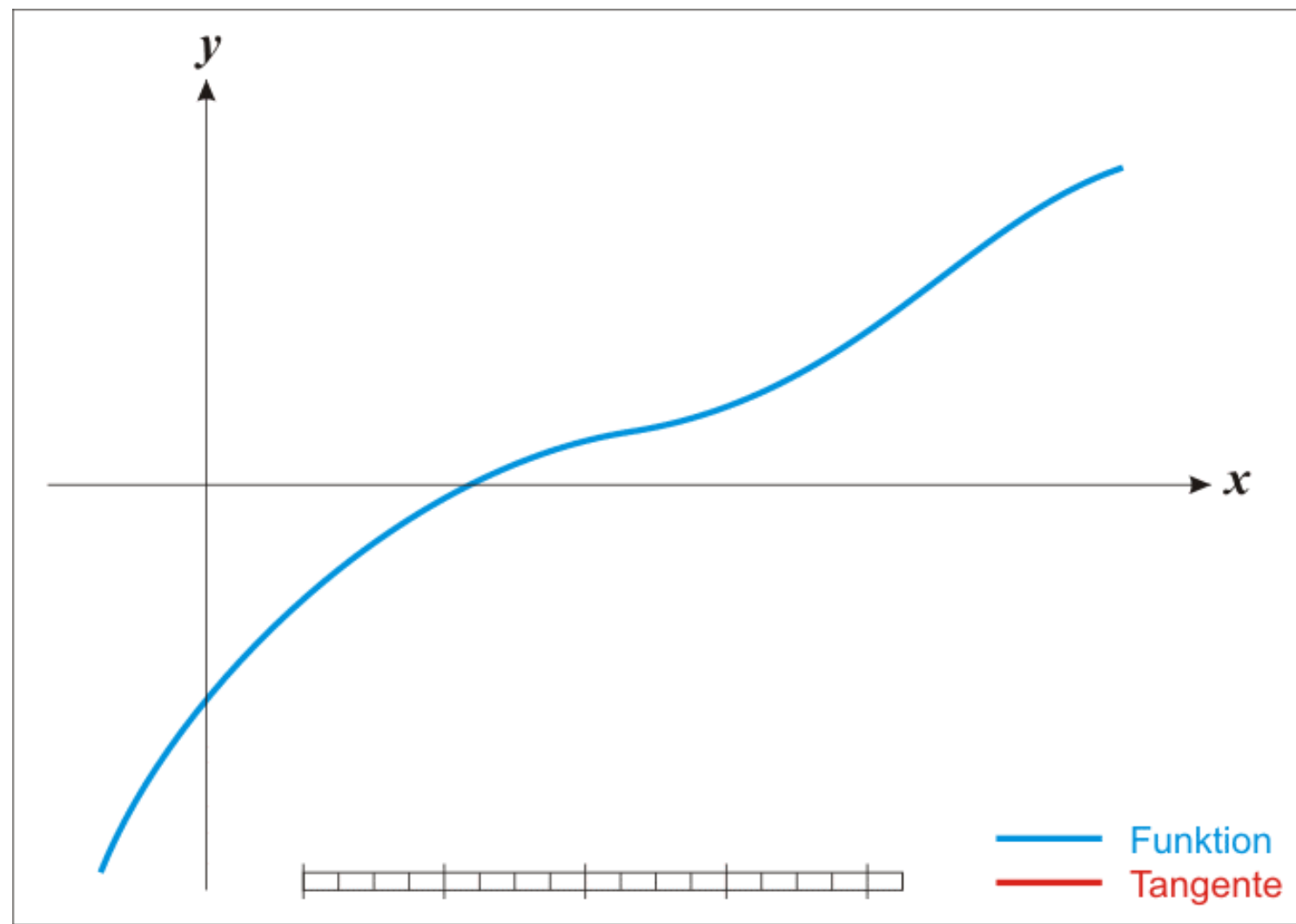
# How to solve an optimization problem?

## Newton's method

Repeat:

$$\bullet \quad x \leftarrow x - \frac{f'(x)}{f''(x)}$$

converges faster  
than Gradient  
Descent



But how do we obtain  
derivatives?

$$\nabla f(x)$$

$$f'(x)$$

$$f''(x)$$



The math technique:  
Differentiation

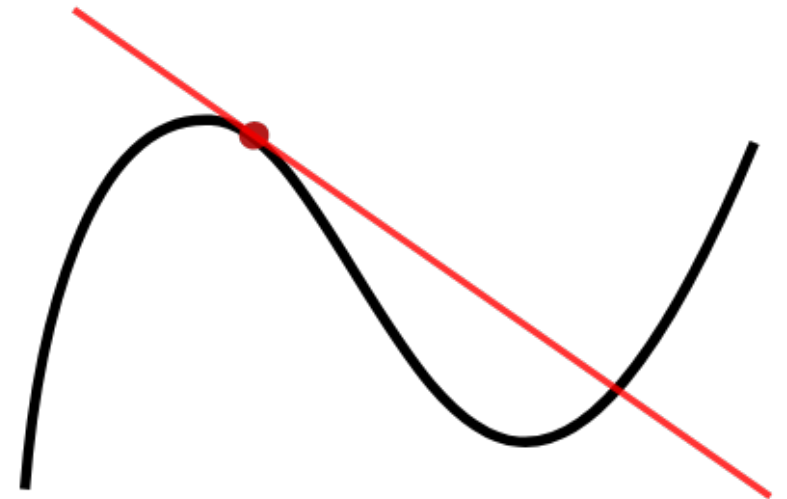
# Differentiation

Slope of the function's tangent

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

Leibniz's notation:  $\frac{df}{dx}$

Partial derivatives:  $\frac{\partial f(x_1, x_2, \dots, x_n)}{\partial x_i}$



# Differentiation Rules

- **Constants:** if  $f(x) = C$  then  $f'(x) = 0$
- **Sum:**  $(f(x) + g(x))' = f'(x) + g'(x)$
- **Product:**  $(f(x) \cdot g(x))' = f'(x)g(x) + f(x)g'(x)$
- **Quotient:**  $\left(\frac{1}{f(x)}\right)' = -\frac{f'(x)}{f(x)^2}$
- **Chain rule:**  $f(g(x))' = f'(g(x))g'(x)$
- **Trigo:**  $\sin(x)' = \cos(x)$  ,  $\cos(x)' = -\sin(x)$

# Differentiation Examples

$$3' = 0$$

$$x' = 1$$

$$(x^2)' = (xx)' = x'x + xx' = 2x$$

$$\tan(x)' = \left( \frac{\sin(x)}{\cos(x)} \right)' = \frac{\sin^2(x) + \cos^2(x)}{\cos^2(x)} = \frac{1}{\cos^2(x)}$$

But how do computers  
differentiate?

# Numerical differentiation

1. Choose  $h$
2. Compute  $f(x)$
3. Compute  $f(x + h)$
4. Return  $\frac{f(x + h) - f(x)}{h}$

Problems:

- $h$  too large — truncation error
- $h$  too small — roundoff error

# Symbolic differentiation

1. Represent function symbolically.
2. Apply differential rules.

Problems:

Cannot handle complex constructors  
(loops/conditional/recursion)

Code swell:  $\left( \frac{\log(x) + \exp(x)}{\log(x)\exp(x)} \right)' =$

$$-\frac{\exp(-x)(\log(x) + \exp(x))}{\log(x)} - \frac{\exp(-x)(\log(x) + \exp(x))}{x \log^2(x)} + \frac{\exp(-x)(\exp(x) + \frac{1}{x})}{\log(x)}$$

# Algorithmic differentiation

- NOT symbolic differentiation
- NOT numerical differentiation
- differentiates ANY code
- Computation costs of  $f'(x)$  and  $f(x)$  are similar



# Algorithmic differentiation

```
def f(a, b):  
    c = a*b  
    d = sin(c)  
    return d
```

```
def f(a, da, b, db):  
    c, dc = a*b, da*b + a*db  
    d, dd = sin(c), dc * cos(c)  
    return d, dd
```

# Program trace

**f**

```
def f(a, b):  
    c = a*b  
    if c > 0:  
        d = log(c)  
    else:  
        d = sin(c)  
    return d
```

**f(2, 3)**

a=2, b=3

c=a\*b=6

d=log(c)=1.791

return d=1.791

**f(2, 3), df(2,3)/da**

a=2, b=3, da=1, db=0

c=a\*b=6, dc=da\*b + a\*db=3

d=log(c)=1.791, dd=dc\*(1/c)=0.5

return d=1.791, dd=0.5

Let's try...