



DeepLearning.AI

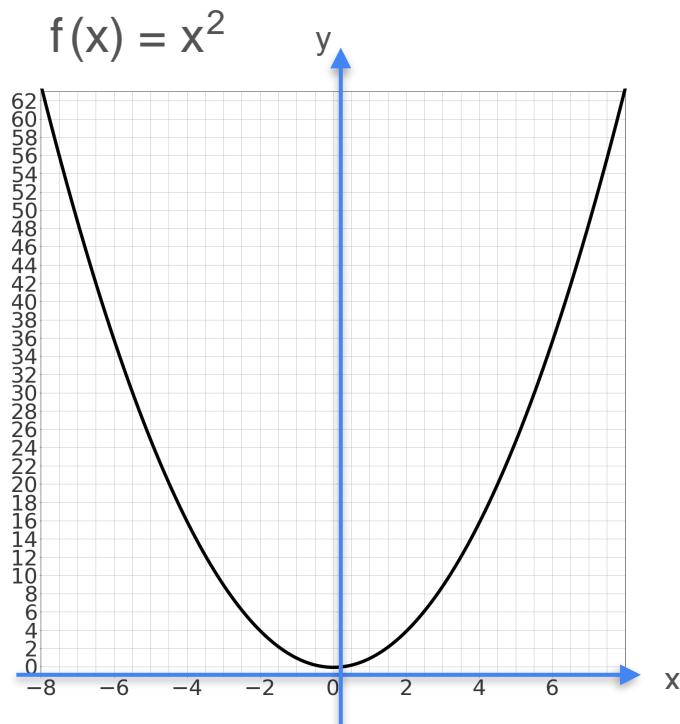
# Gradients and Gradient Descent

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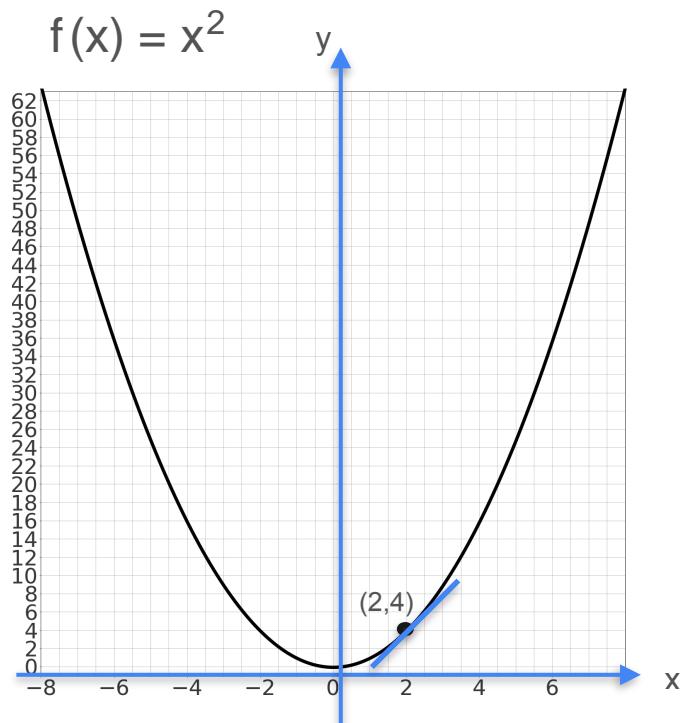
## Tangent planes

# Functions of Two Variables

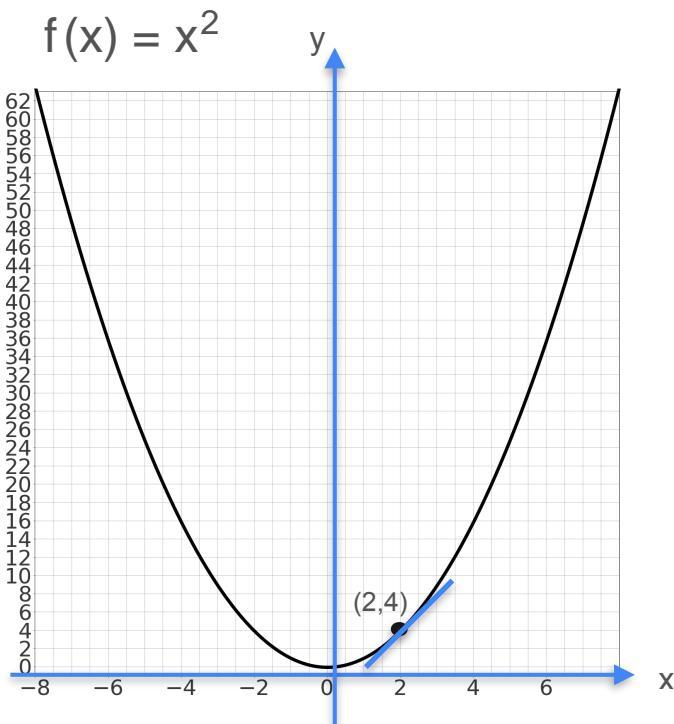
# Functions of Two Variables



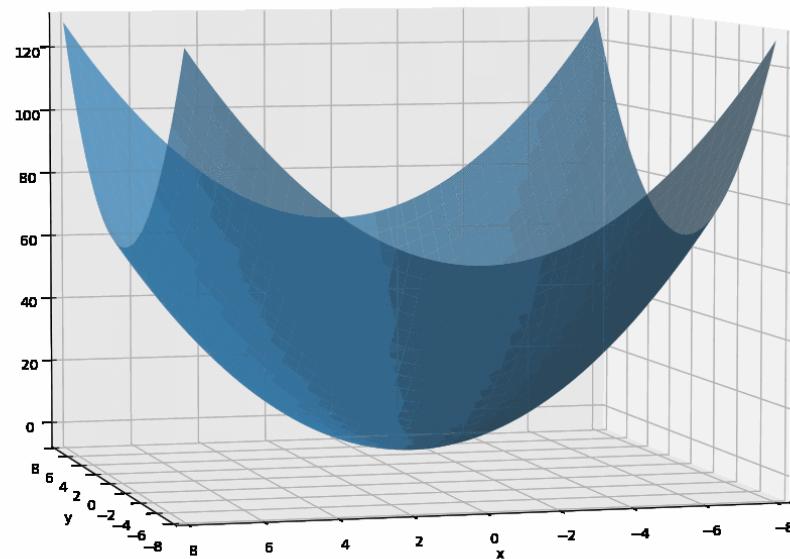
# Functions of Two Variables



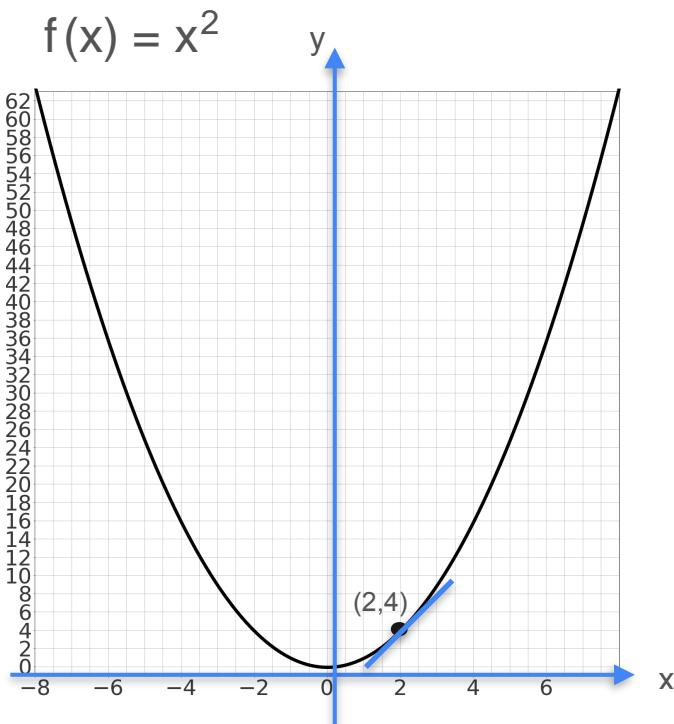
# Functions of Two Variables



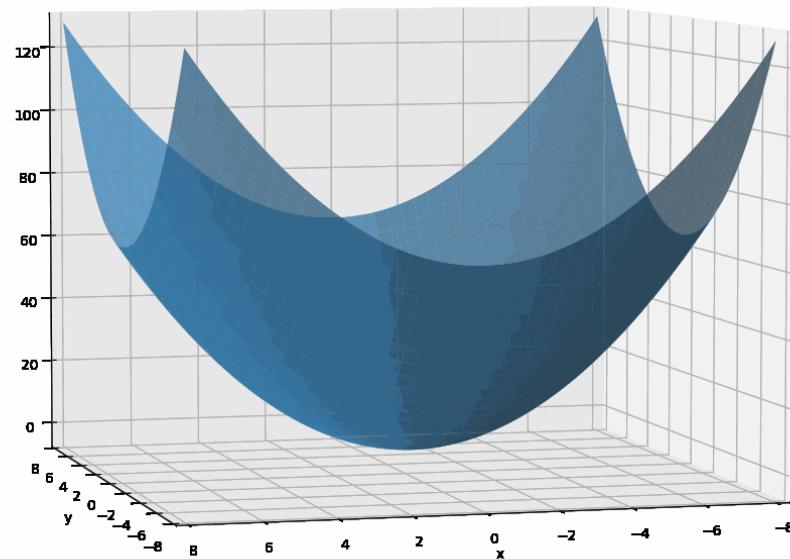
$$f(x, y) = x^2 + y^2$$



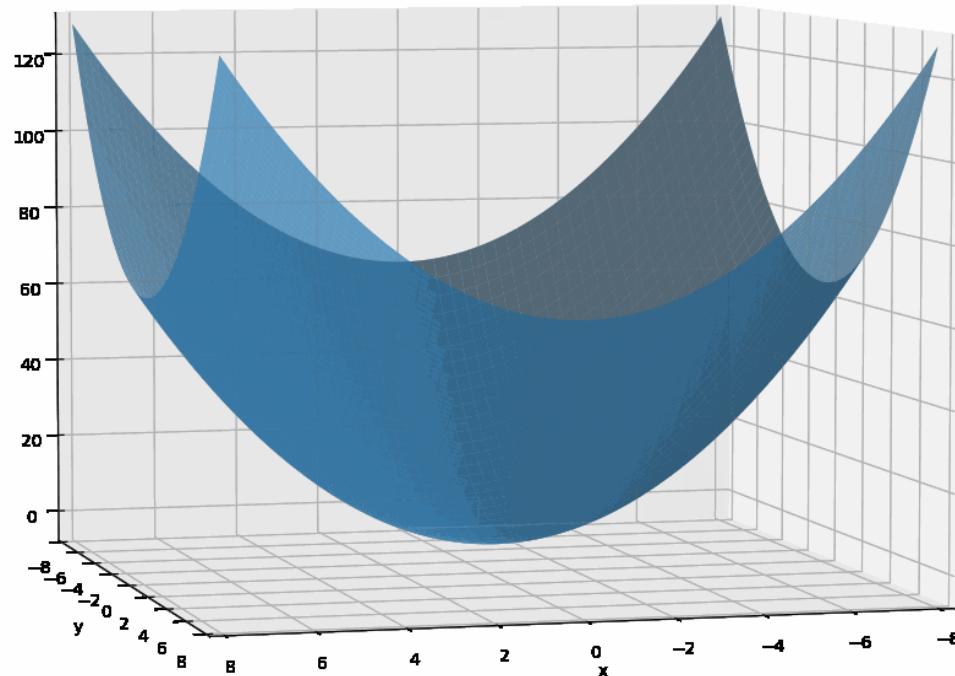
# Functions of Two Variables



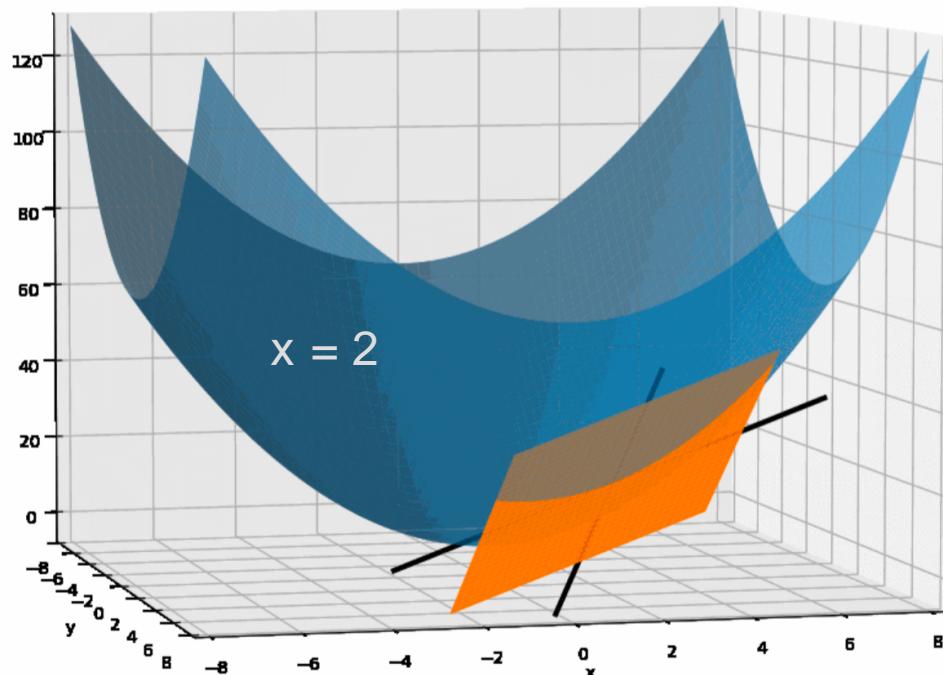
$$f(x, y) = x^2 + y^2$$



# Finding the Tangent Plane



# Finding the Tangent Plane



$$\text{Fix } y=4 \quad f(x,4) = x^2 + 4^2 \\ \frac{d}{dx} (f(x,4)) = 2x$$

$$\text{Fix } x=2 \quad f(2,y) = 2^2 + y^2 \\ \frac{d}{dy} (f(2,y)) = 2y$$

The tangent plane contains both tangent lines.

# Video 2: Introduction to Partial Derivatives

Example with the parabola, show tangent plane and slices



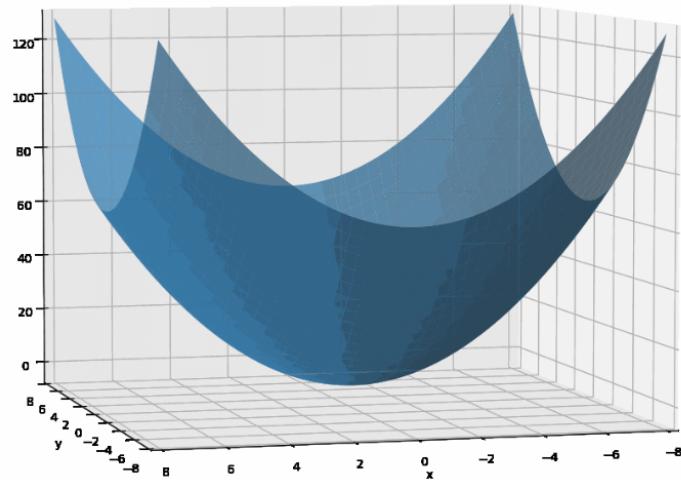
DeepLearning.AI

# Gradients and Gradient Descent

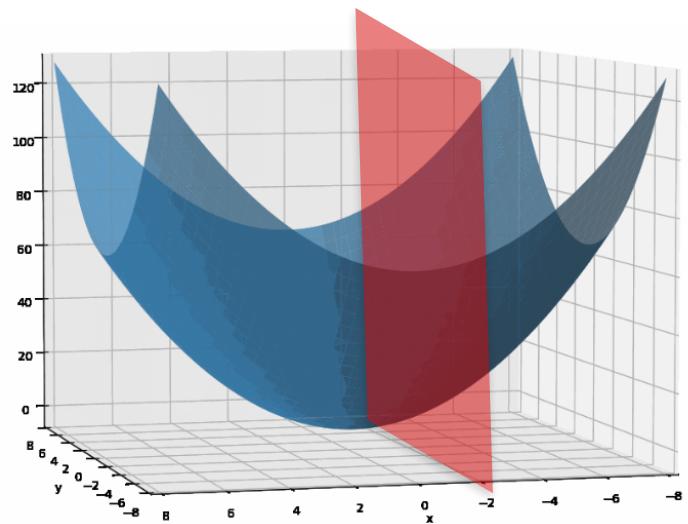
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## Partial derivatives - Part 1

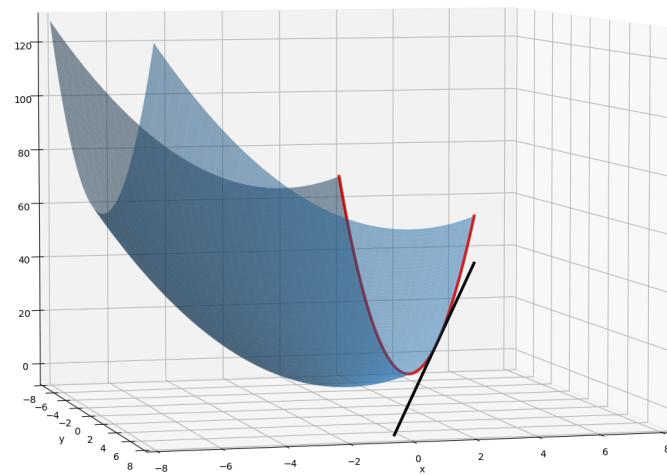
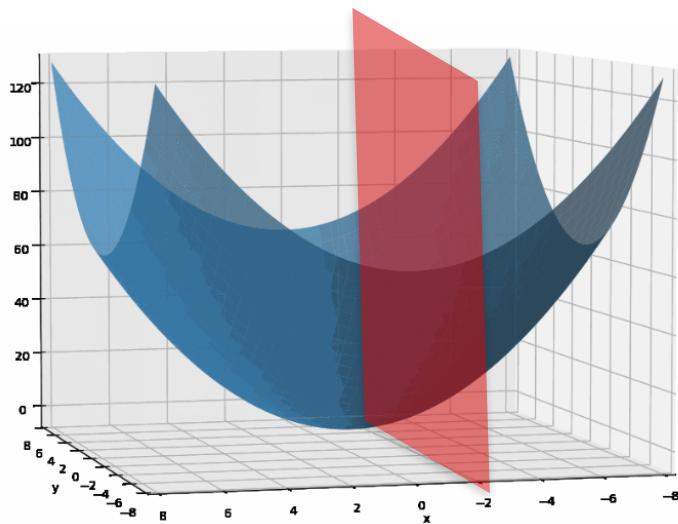
# Slicing the Space



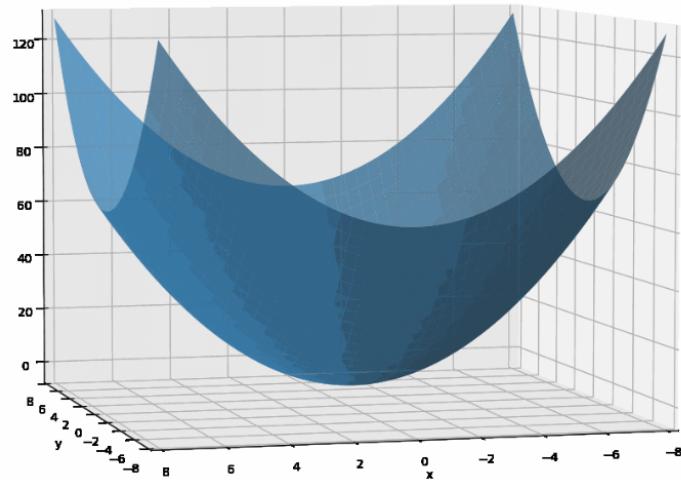
# Slicing the Space



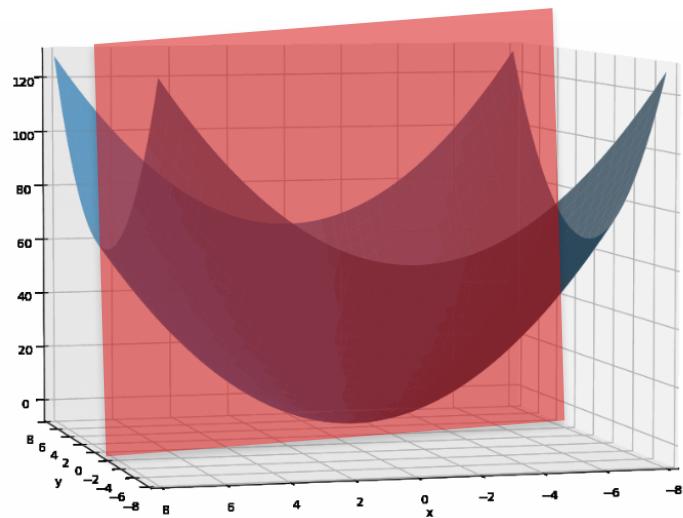
# Slicing the Space



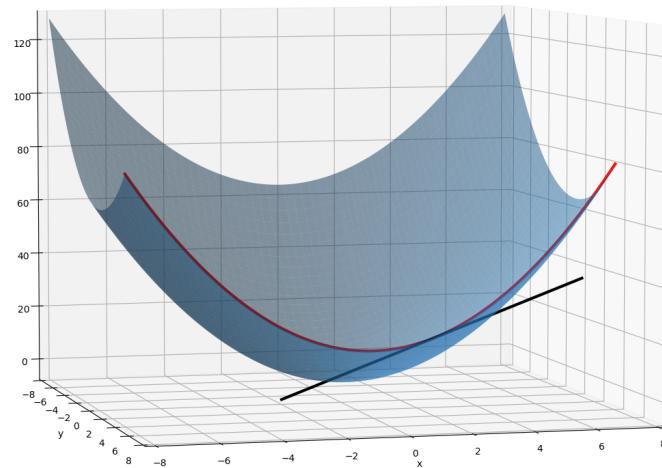
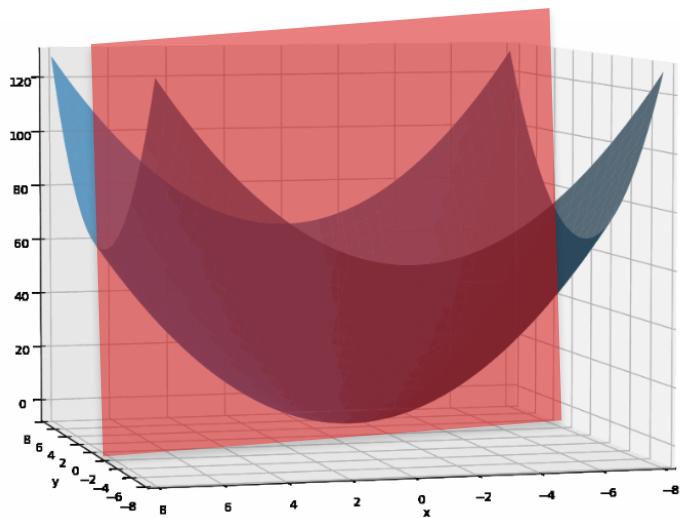
# Slicing the Space



# Slicing the Space

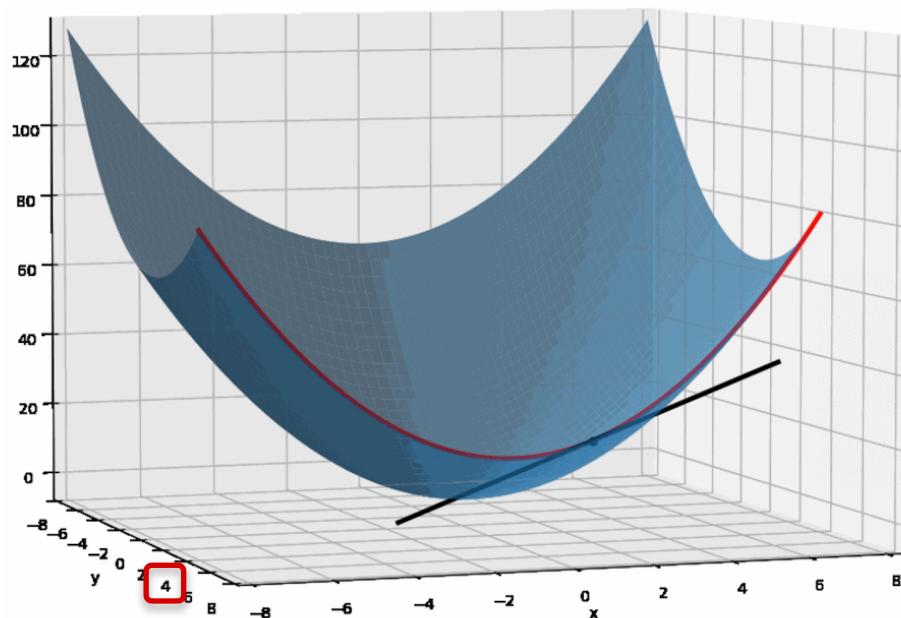


# Slicing the Space



# Partial Derivatives

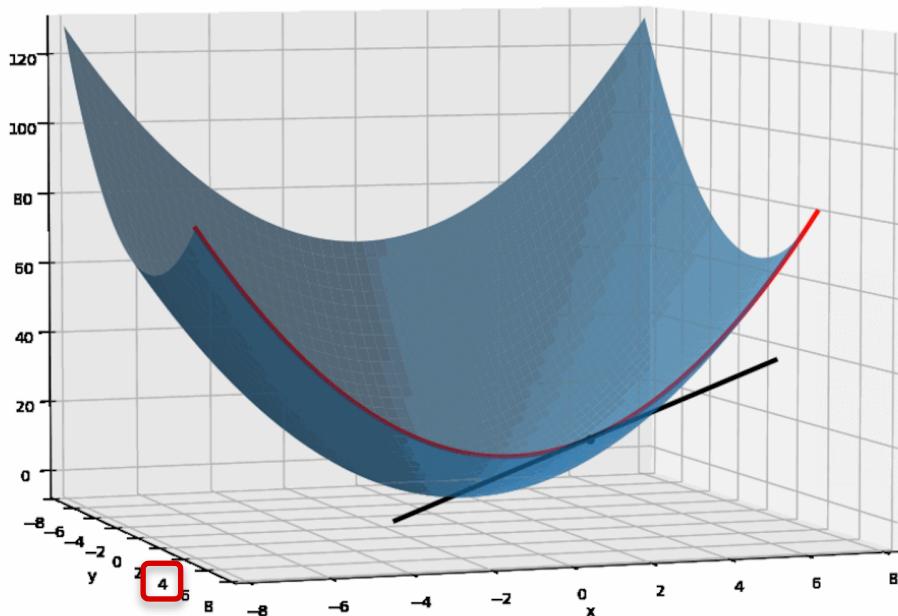
$$f(x, y) = x^2 + y^2$$



# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

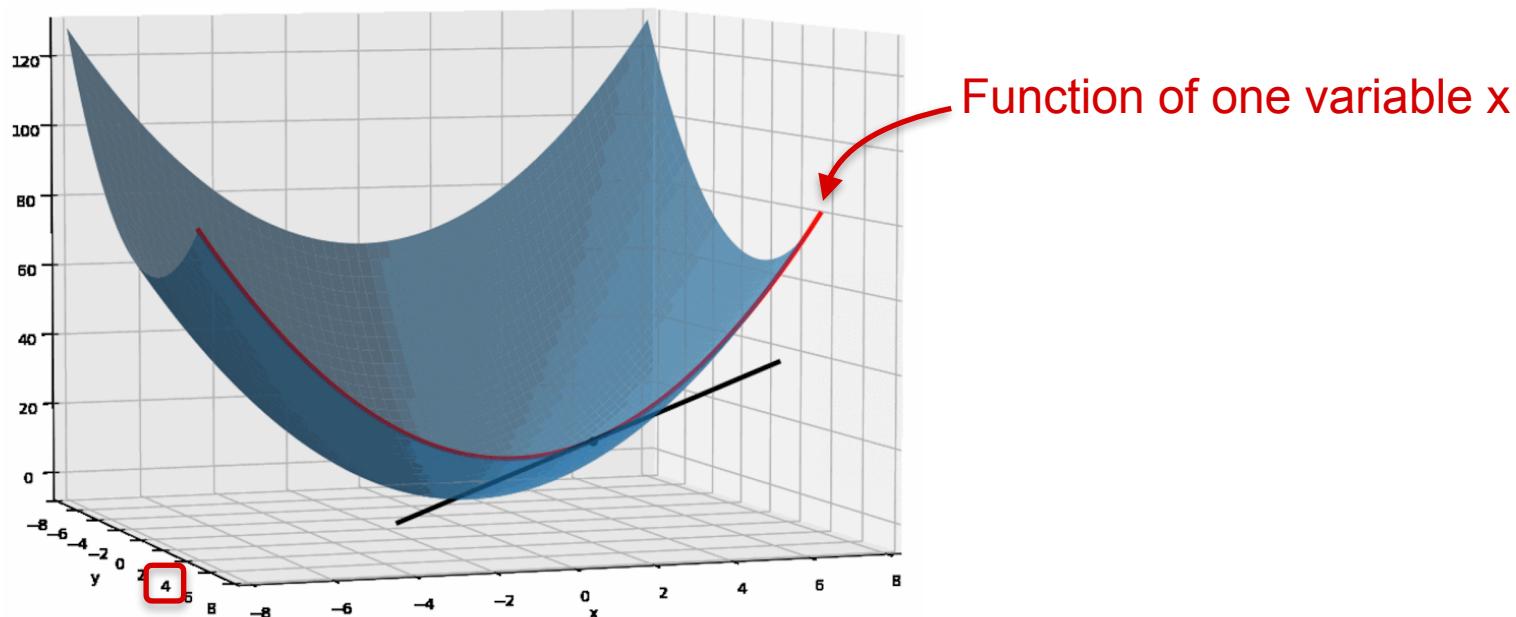
Treat y as a constant



# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

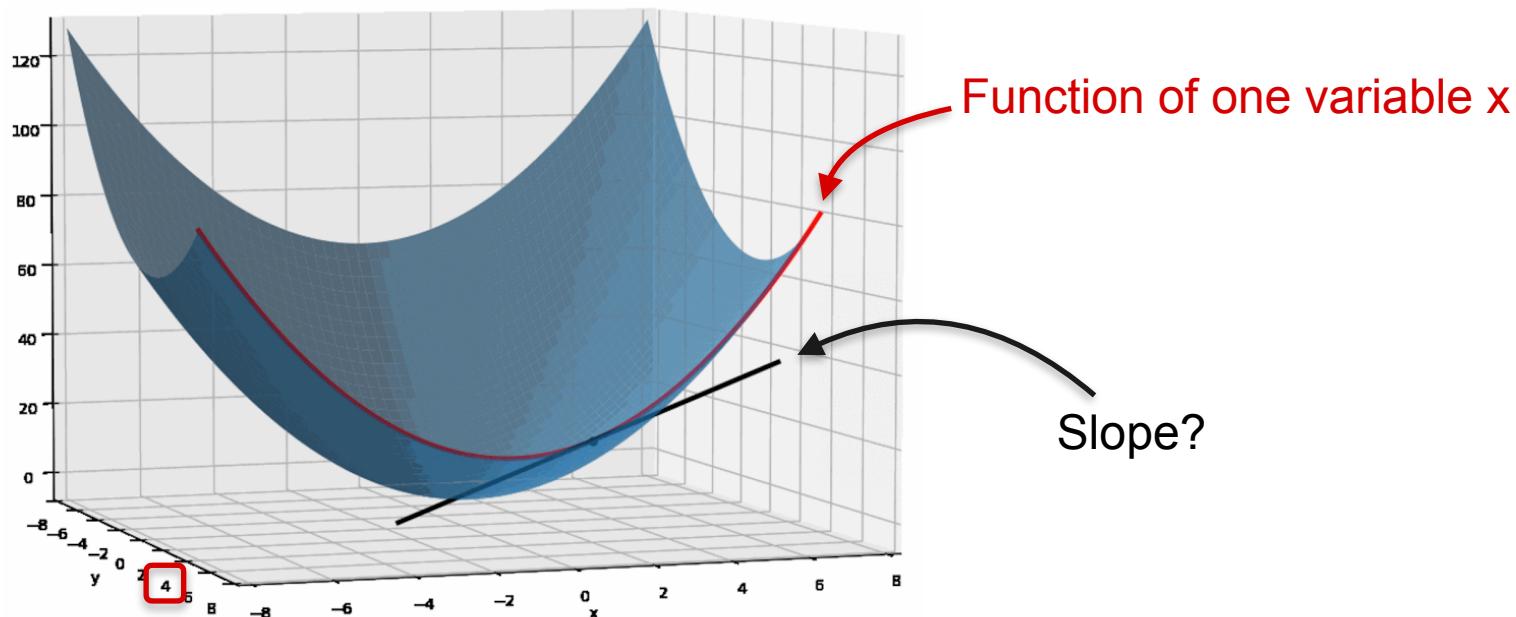
Treat y as a constant



# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

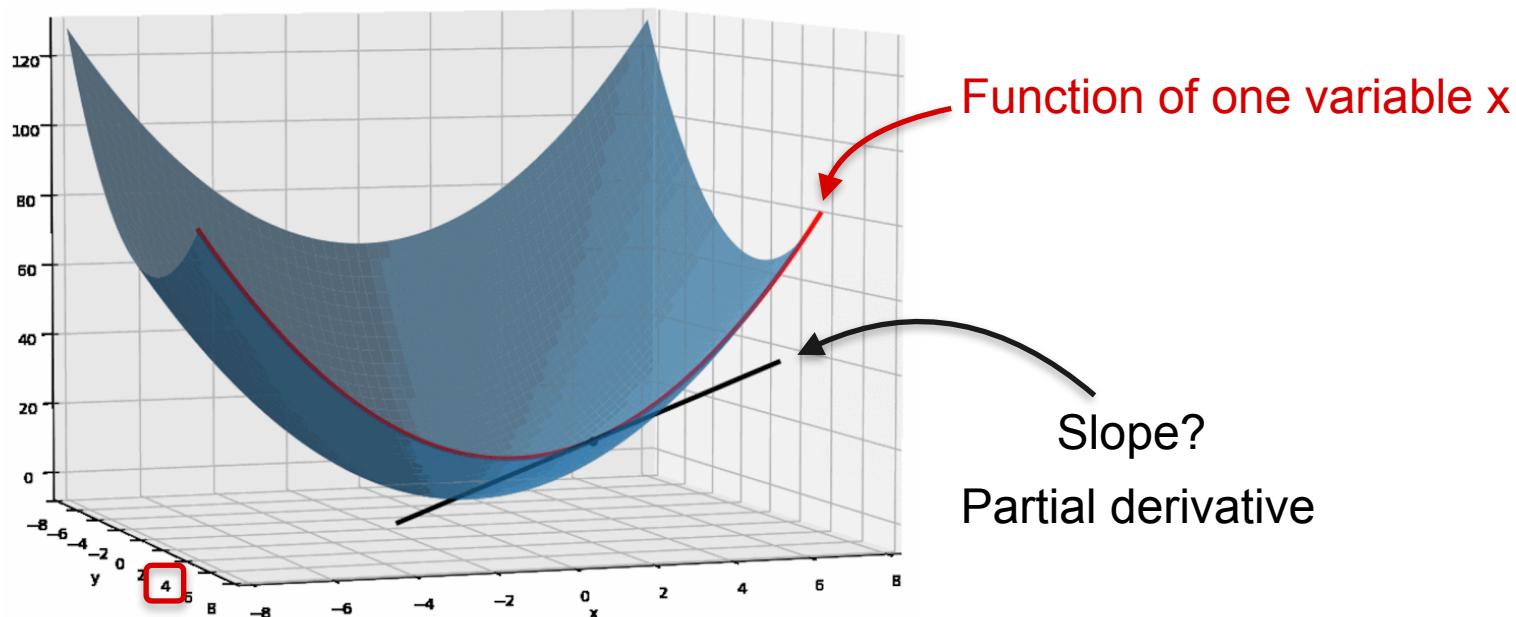
Treat y as a constant



# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

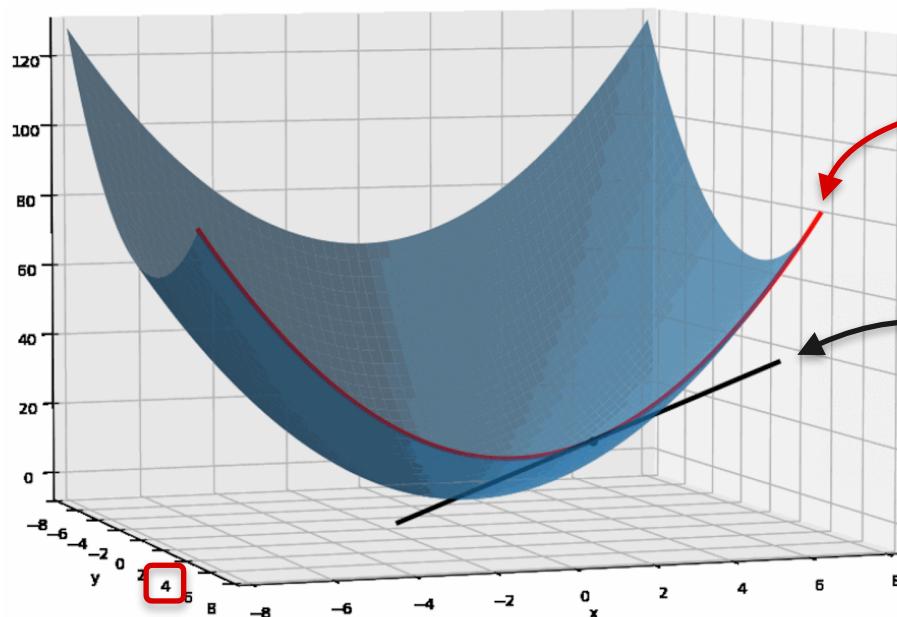
Treat y as a constant



# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

Treat y as a constant



Function of one variable x

$$f(x, y) = x^2 + y^2$$

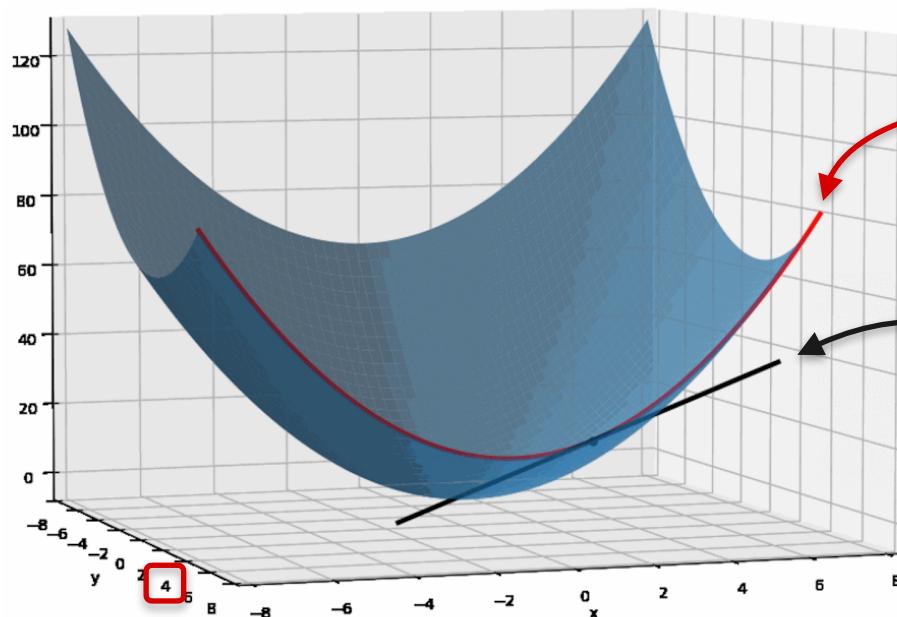
Slope?

Partial derivative

# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

Treat y as a constant



Function of one variable x

Constant

$$f(x, y) = x^2 + \boxed{y^2}$$

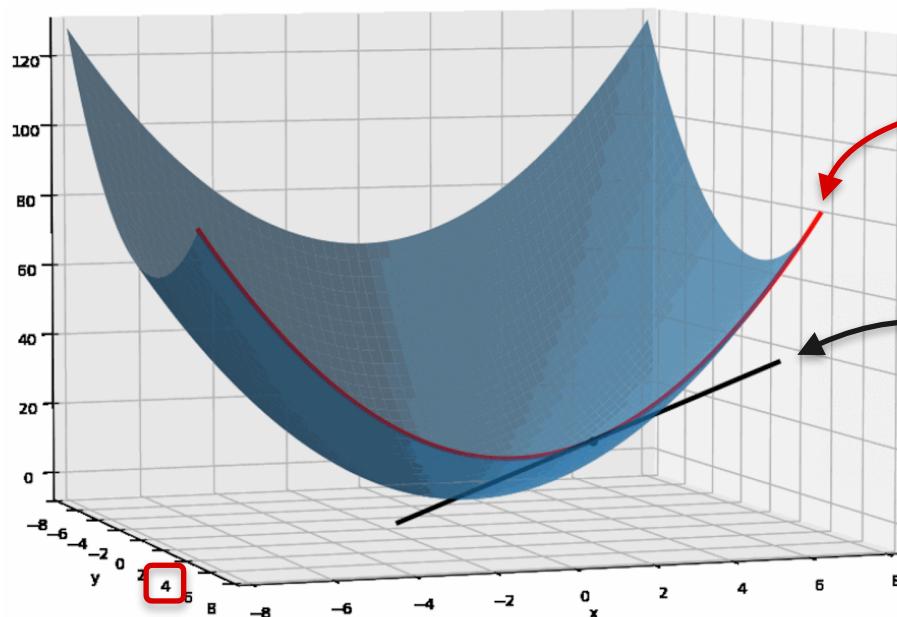
Slope?

Partial derivative

# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

Treat y as a constant



Function of one variable x

Slope?

Partial derivative

Constant

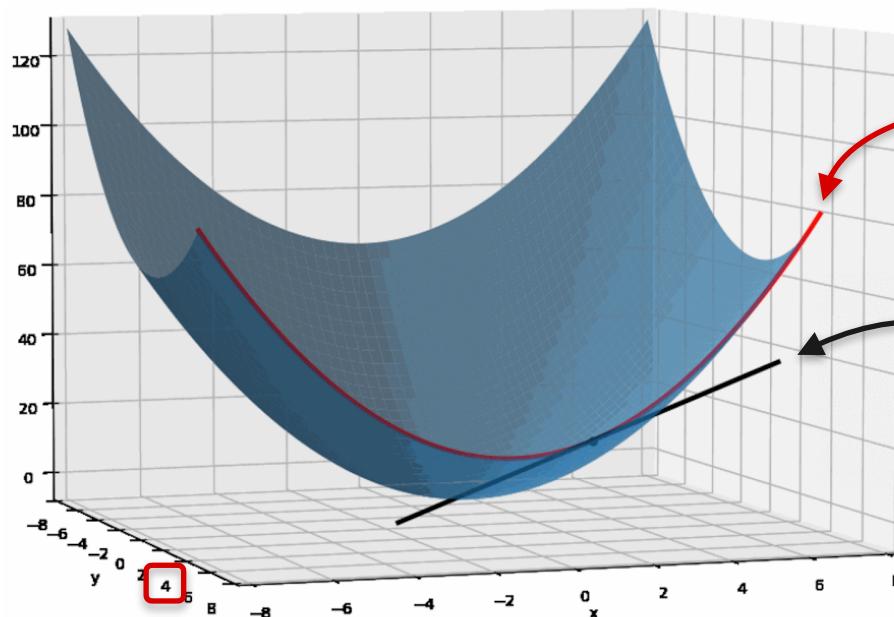
$$f(x, y) = x^2 + \boxed{y^2}$$

$$\frac{f}{x} = 2x + 0$$

# Partial Derivatives

$$f(x, y) = x^2 + y^2$$

Treat y as a constant



Function of one variable x

Slope?

Partial derivative

Constant

$$f(x, y) = x^2 + \boxed{y^2}$$

$$\frac{f}{x} = 2x + \boxed{0}$$

Derivative = 0

# Partial Derivatives

# Partial Derivatives

$$x^2 + y^2$$

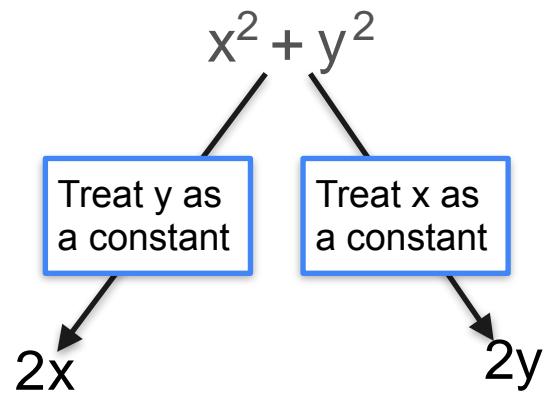
# Partial Derivatives

$$x^2 + y^2$$

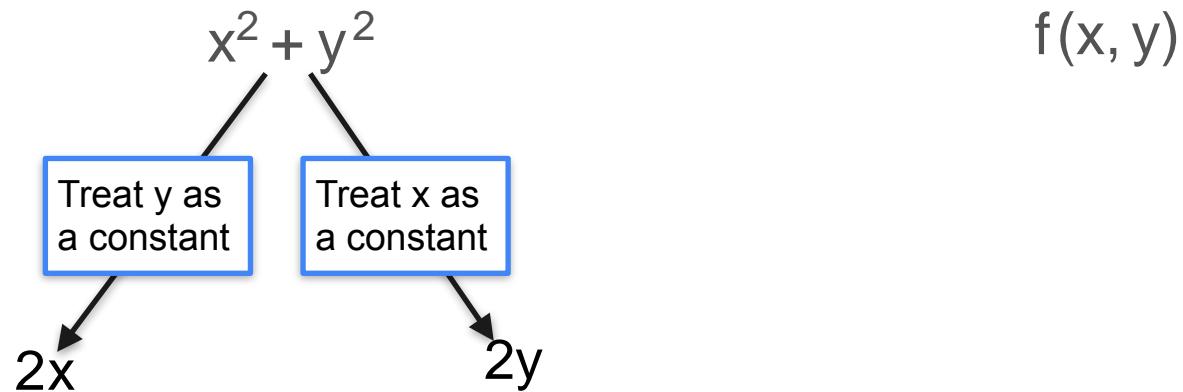
Treat y as a constant

$$2x$$

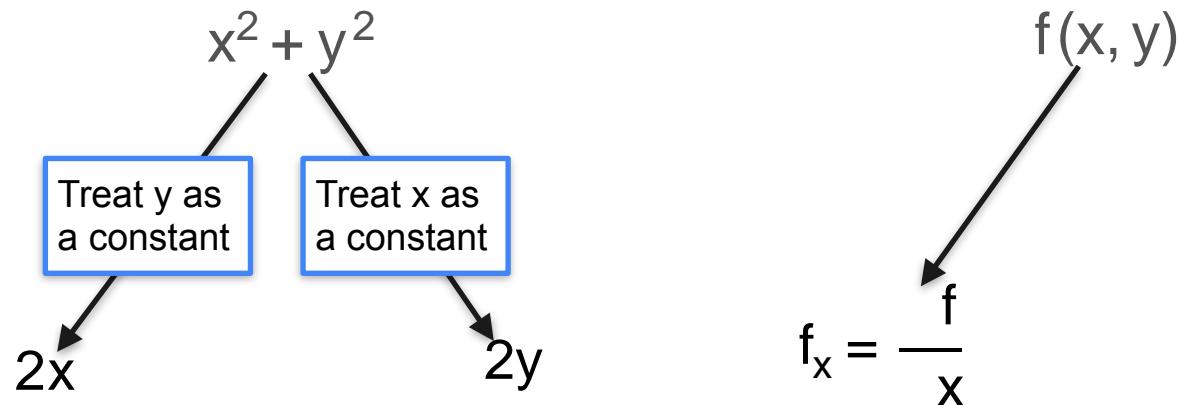
# Partial Derivatives



# Partial Derivatives



# Partial Derivatives



The diagram illustrates the calculation of the first-order partial derivative  $f_x$  for the function  $f(x, y)$ . At the top, the function is written as  $f(x, y)$ . An arrow points downwards to the formula  $f_x = \frac{f}{x}$ , which represents the derivative of the function with respect to  $x$ , treating  $y$  as a constant.

$f(x, y)$

$f_x = \frac{f}{x}$

# Partial Derivatives

$$x^2 + y^2$$

Treat y as a constant

Treat x as a constant

$$2x$$
$$2y$$

```
graph TD; A[x^2 + y^2] --> B["Treat y as a constant"]; A --> C["Treat x as a constant"]; B --> D[2x]; C --> E[2y]
```

$$f(x, y)$$
$$f_x = \frac{f}{x}$$
$$f_y = \frac{f}{y}$$

```
graph TD; A[f(x, y)] --> B[f_x]; A --> C[f_y]; B --> D["f / x"]; C --> E["f / y"]
```

# Partial Derivatives

$$x^2 + y^2$$

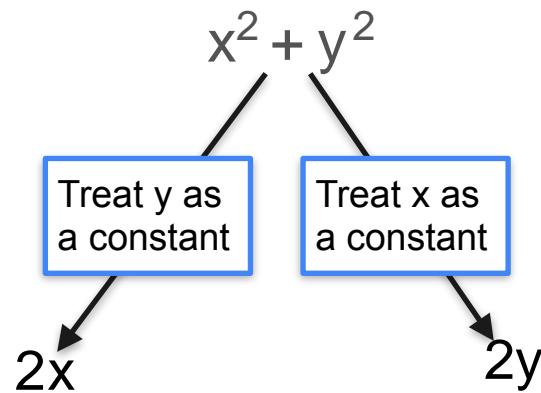
Treat y as a constant      Treat x as a constant

$$2x \quad 2y$$

$$f(x, y)$$
$$f_x = \frac{f}{x} \quad f_y = \frac{f}{y}$$

Partial derivative of  
f with respect to x

# Partial Derivatives



A diagram illustrating the general form of partial derivatives. At the top center is the expression  $f(x, y)$ . Two arrows point downwards from this expression to two equations. The left equation is  $f_x = \frac{f}{x}$  and the right equation is  $f_y = \frac{f}{y}$ .

Partial derivative of  
f with respect to x

Partial derivative of  
f with respect to x

# Intro To Partial Derivatives

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

**TASK**

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivatives of  $f$  with respect to  
 $x$  and  $y$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

$$\frac{f}{x} =$$

Find partial derivatives of  $f$  with respect to  
 $x$  and  $y$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

$$\frac{f}{x} =$$

Find partial derivatives of  $f$  with respect to  $x$  and  $y$

$$\frac{f}{y} =$$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivatives of  $f$  with respect to  $x$  and  $y$

# Intro To Partial Derivatives

Partial derivative notation

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivatives of  $f$  with respect to  $x$  and  $y$

# Intro To Partial Derivatives

Partial derivative notation

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivatives of  $f$  with respect to  $x$  and  $y$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

**TASK**

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

Step 1:

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} =$$

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:**

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = ?$$
  
$$\frac{f}{y} = ?$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$


## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

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# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + 1$$

$$f(x, y) = x^2 + y^2$$


## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + 1$$

$$f(x, y) = x^2 + y^2$$


## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

$$\frac{f}{x} = 2x$$

$$\frac{f}{y} =$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Intro To Partial Derivatives

$$f(x, y) = x^2 + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

$$\frac{f}{x} = 2x$$

**Step 2:** Differentiate the function using the normal rules of differentiation.

$$\frac{f}{y} =$$

# Intro To Partial Derivatives

$$f(x, y) = \boxed{x^2} + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

$$\frac{f}{x} = 2x$$

**Step 2:** Differentiate the function using the normal rules of differentiation.

$$\frac{f}{y} =$$

# Intro To Partial Derivatives

$$f(x, y) = 1 + y^2$$

$$f(x, y) = \boxed{x^2} + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

$$\frac{f}{x} = 2x$$

**Step 2:** Differentiate the function using the normal rules of differentiation.

$$\frac{f}{y} =$$

# Intro To Partial Derivatives

$$f(x, y) = 1 + y^2$$

$$f(x, y) = \boxed{x^2} + y^2$$

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

$$\frac{f}{x} = 2x$$

$$\frac{f}{y} = 2y$$

**Step 2:** Differentiate the function using the normal rules of differentiation.



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# Gradients and Gradient Descent

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## Partial derivatives -Part 2

# Partial Derivatives (More Examples)

# Partial Derivatives (More Examples)

$$f(x,y) = 3x^2y^3$$

# Partial Derivatives (More Examples)

$$f(x,y) = 3x^2y^3$$

## TASK

What is the partial derivate of  $f$  with respect to  $x$ ?

# Partial Derivatives (More Examples)

$$f(x,y) = 3x^2y^3$$

$$\frac{f}{x} = ?$$

## TASK

What is the partial derivate of  $f$  with respect to  $x$ ?

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{x} =$$

**TASK**

Find partial derivate of  $f$  with respect to  $x$

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{x} =$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

$$\frac{f}{x} =$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

$$\frac{f}{x} =$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$



$$\frac{f}{x} =$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

Constant coefficient

$$\frac{f}{x} =$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

Constant coefficient

$$\frac{f}{x} = 3$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

$$\frac{f}{x} = 3$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

Differentiate with respect to  $x$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$


Differentiate with respect to  $x$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

$$\frac{f}{x} = 3(2x)$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2 +$$

treat as constant coefficient

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$\frac{f}{x} = 3(2x)$$

$$f(x, y) = 3x^2$$



treat as constant coefficient

## TASK

Find partial derivative of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{x} = 3(2x)y^3$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\begin{aligned}\frac{f}{x} &= 3(2x)y^3 \\ &= 6xy^3\end{aligned}$$

## TASK

Find partial derivate of  $f$  with respect to  $x$

**Step 1:** Treat all other variables as a constant. In our case  $y$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{y} =$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{y} = ?$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\frac{f}{y} =$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

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# Partial Derivatives (More Examples)

$$f(x, y) = 3x^3y^3$$

$$\frac{f}{y} =$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

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# Partial Derivatives (More Examples)

$$f(x, y) = 3x^3y^3$$



$$\frac{f}{y} = 3$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

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# Partial Derivatives (More Examples)

$$\frac{f}{y} = 3$$

$$f(x, y) = 3x^3y^3$$



## TASK

What is the partial derivate of f with respect to y?

**Step 1:** Treat all other variables as a constant. In our case x.

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$\frac{f}{y} = 3 \cdot (3y^2)$$

**TASK**

What is the partial derivate of  $f$  with respect to  $y$ ?

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.

# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$



$$\begin{aligned}\frac{f}{y} &= 3(x^2)(3y^2) \\ &= 9x^2y^2\end{aligned}$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

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# Partial Derivatives (More Examples)

$$f(x, y) = 3x^2y^3$$

$$\begin{aligned}\frac{f}{y} &= 3(x^2)(3y^2) \\ &= 9x^2y^2\end{aligned}$$

## TASK

What is the partial derivate of  $f$  with respect to  $y$ ?

**Step 1:** Treat all other variables as a constant. In our case  $x$ .

**Step 2:** Differentiate the function using the normal rules of differentiation.



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# Gradients and Gradient Descent

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## Gradients

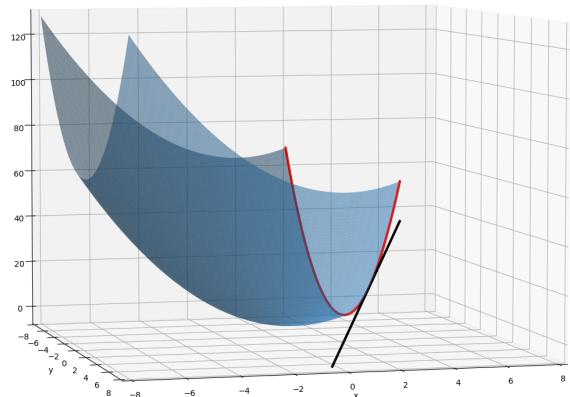
# Gradient

$$f(x, y) = x^2 + y^2$$

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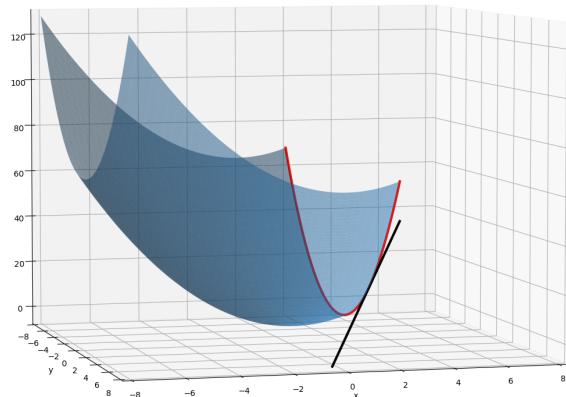
Treat  $y$  as a constant



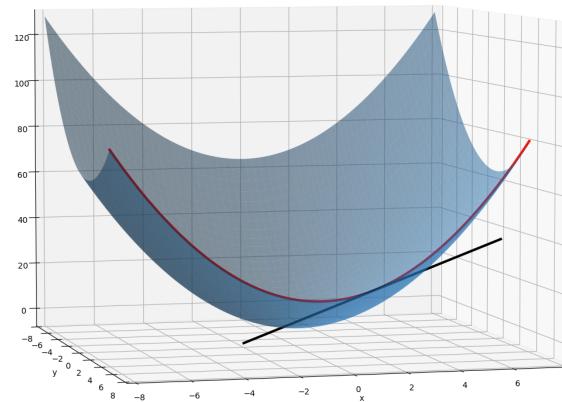
# Gradient

$$f(x, y) = x^2 + y^2$$

Treat  $y$  as a constant



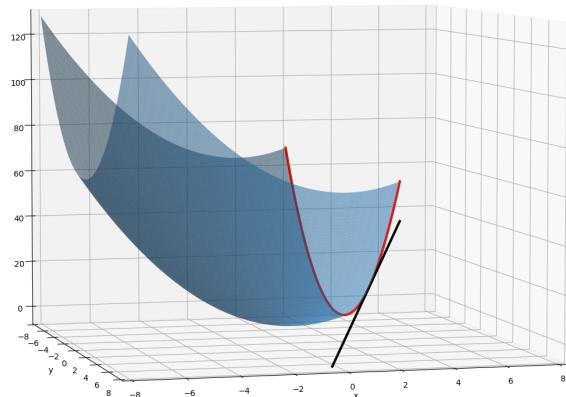
Treat  $x$  as a constant



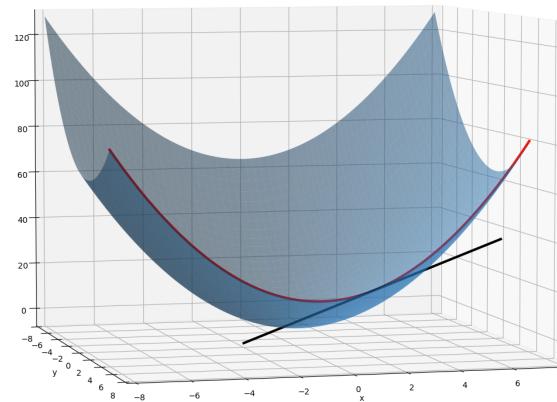
# Gradient

$$f(x, y) = x^2 + y^2$$

Treat y as a constant



Treat x as a constant

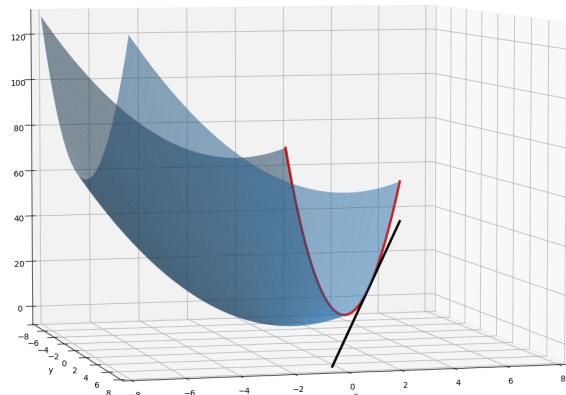


$$\frac{f}{x} = 2x$$

# Gradient

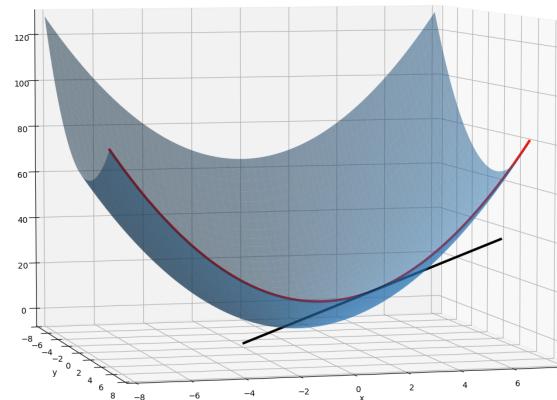
$$f(x, y) = x^2 + y^2$$

Treat y as a constant



$$\frac{f}{x} = 2x$$

Treat x as a constant

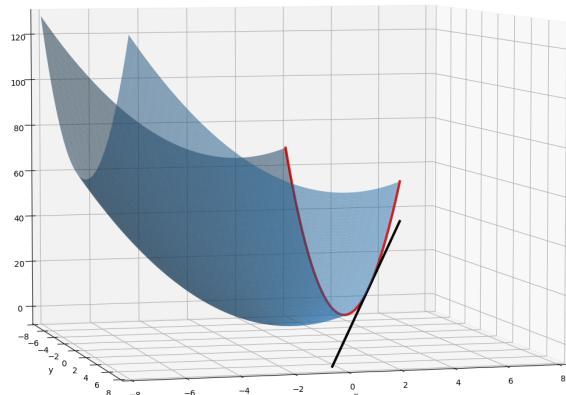


$$\frac{f}{y} = 2y$$

# Gradient

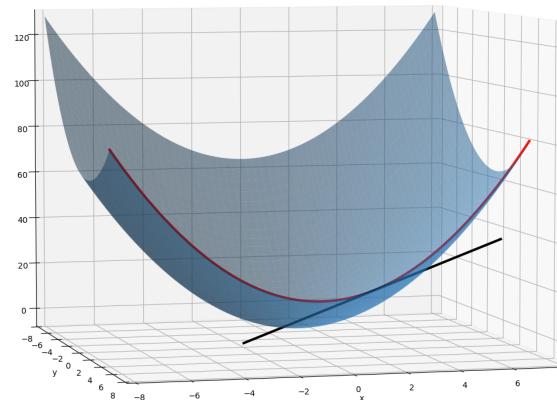
$$f(x, y) = x^2 + y^2$$

Treat y as a constant



$$\frac{f}{x} = 2x$$

Treat x as a constant



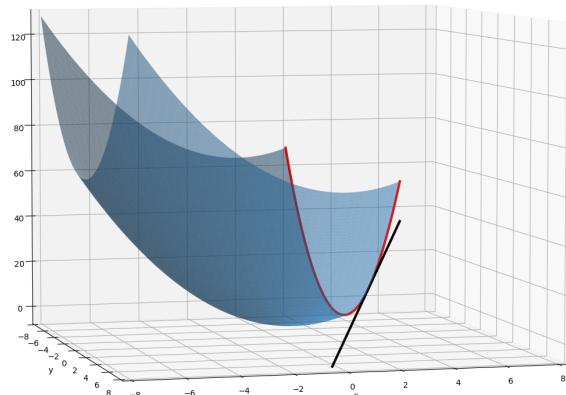
$$\frac{f}{y} = 2y$$

Gradient

# Gradient

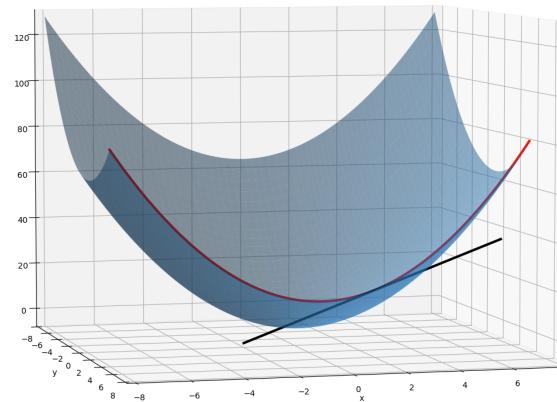
$$f(x, y) = x^2 + y^2$$

Treat y as a constant



$$\frac{\partial f}{\partial x} = 2x$$

Treat x as a constant



$$\frac{\partial f}{\partial y} = 2y$$

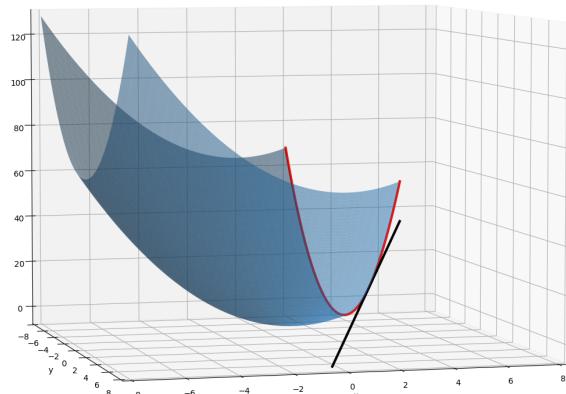
Gradient

$$\begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

# Gradient

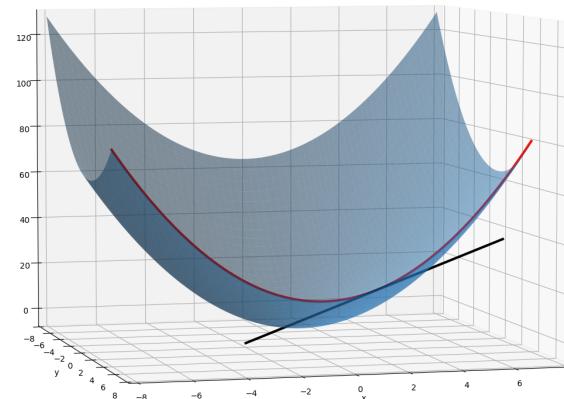
$$f(x, y) = x^2 + y^2$$

Treat  $y$  as a constant



$$\frac{\partial f}{\partial x} = 2x$$

Treat  $x$  as a constant



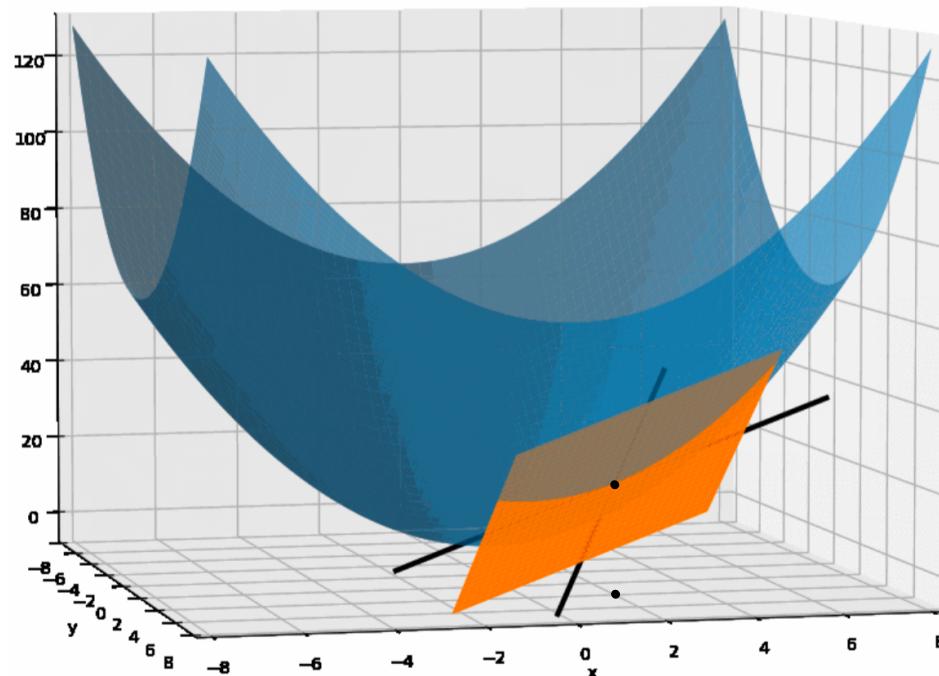
$$\frac{\partial f}{\partial y} = 2y$$

Gradient

$$\begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

$$f = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

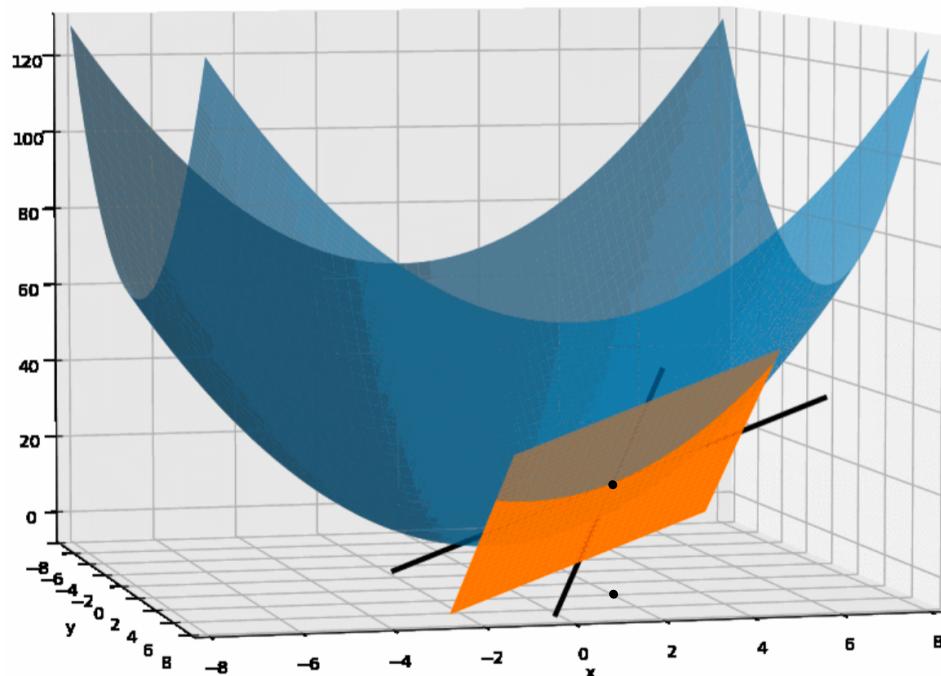
# Gradient



$$f(x, y) = x^2 + y^2$$

The gradient of  $f(x, y)$  is:  $\nabla f = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$

# Gradient



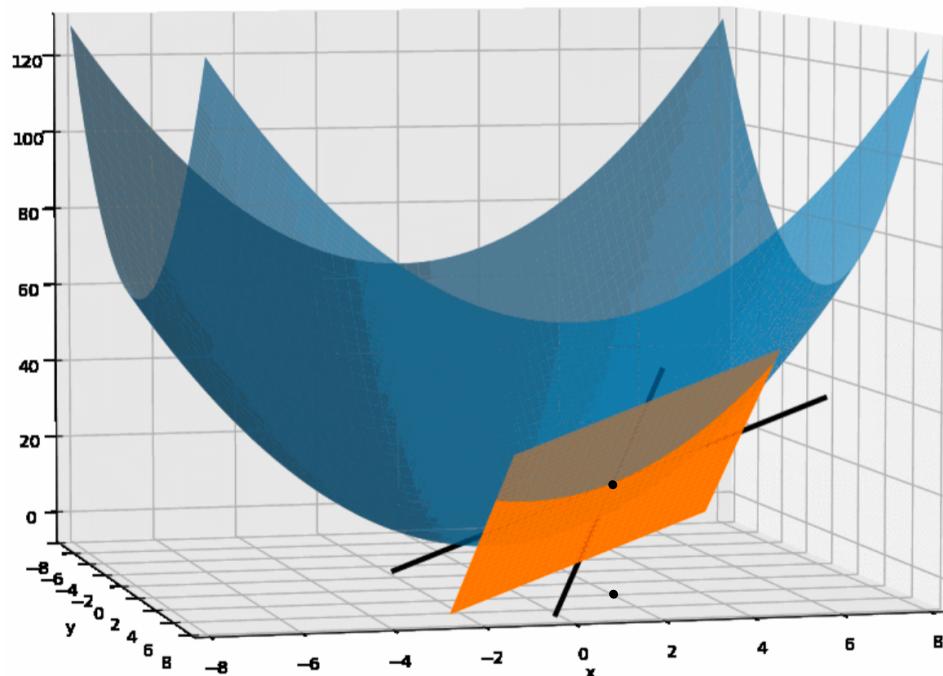
$$f(x, y) = x^2 + y^2$$

The gradient of  $f(x, y)$  is:  $\nabla f = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$

## TASK

Find the gradient of  $f(x, y)$  at  $(2, 3)$

# Gradient



$$f(x, y) = x^2 + y^2$$

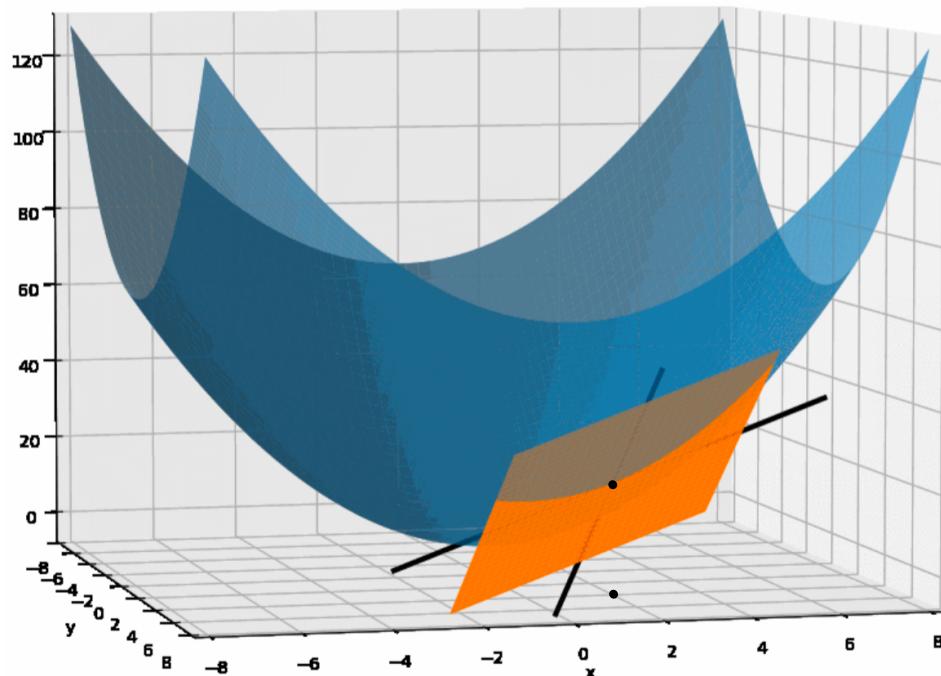
The gradient of  $f(x, y)$  is:  $\begin{bmatrix} 2x \\ 2y \end{bmatrix}$

## TASK

Find the gradient of  $f(x, y)$  at  $(2, 3)$

The gradient of  $f(x, y)$  is given as:

# Gradient



$$f(x, y) = x^2 + y^2$$

The gradient of  $f(x, y)$  is:  $f = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$

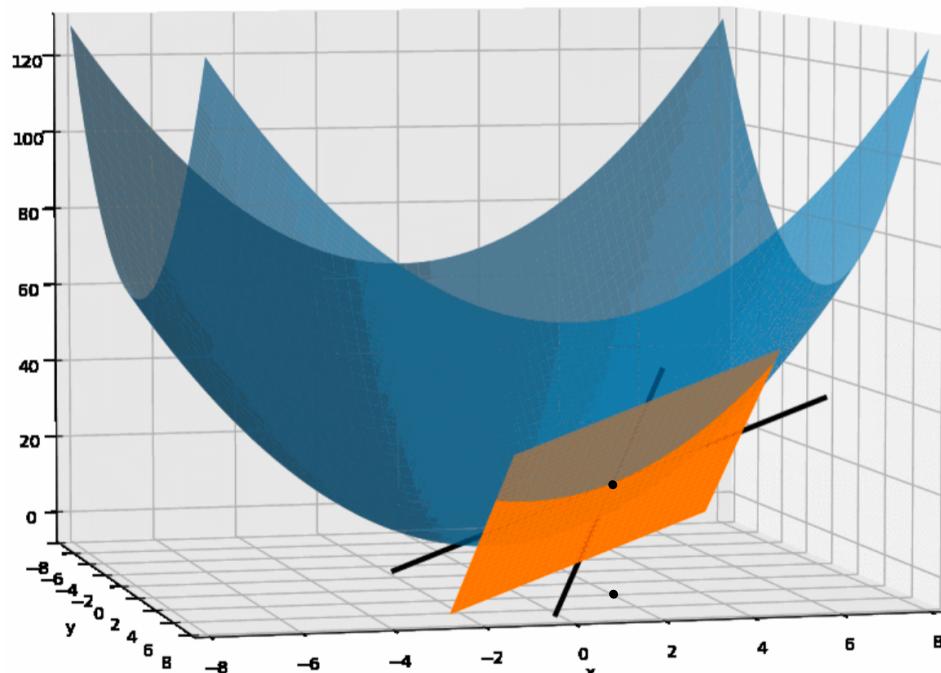
## TASK

Find the gradient of  $f(x, y)$  at  $(2, 3)$

The gradient of  $f(x, y)$  is given as:

$$f = \begin{bmatrix} 2 \\ 2 \\ 3 \end{bmatrix}$$

# Gradient



$$f(x, y) = x^2 + y^2$$

The gradient of  $f(x, y)$  is:  $f = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$

## TASK

Find the gradient of  $f(x, y)$  at  $(2, 3)$

The gradient of  $f(x, y)$  is given as:

$$f = \begin{bmatrix} 2 \\ 2 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \end{bmatrix}$$



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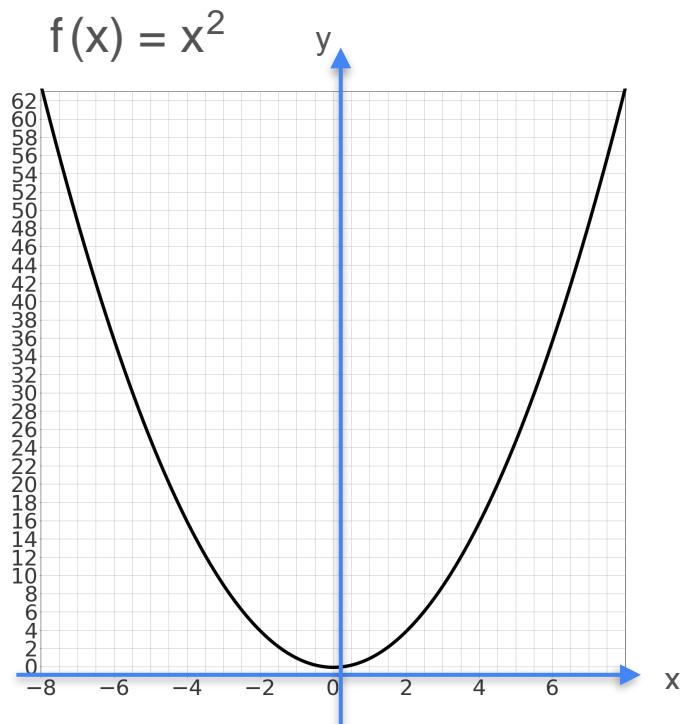
## Gradients and Gradient Descent

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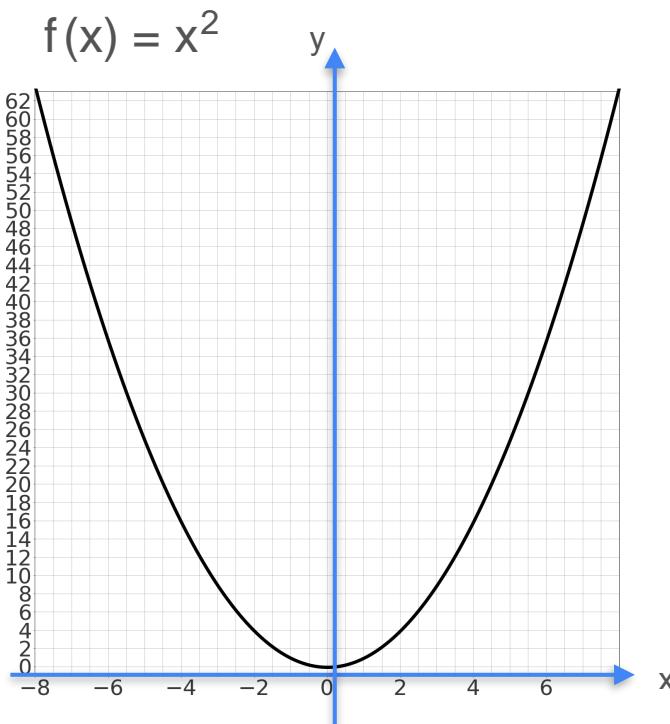
**Gradients and maxima/  
minima**

# Functions of Two Variables

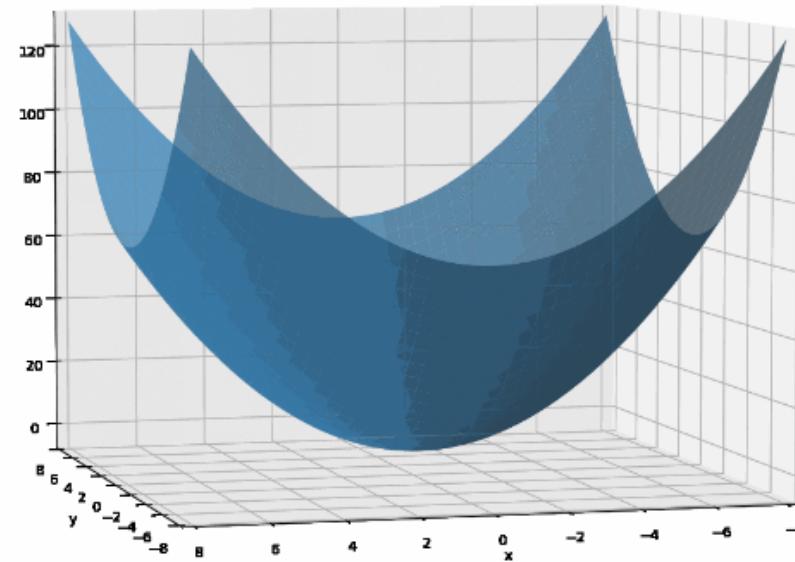
# Functions of Two Variables



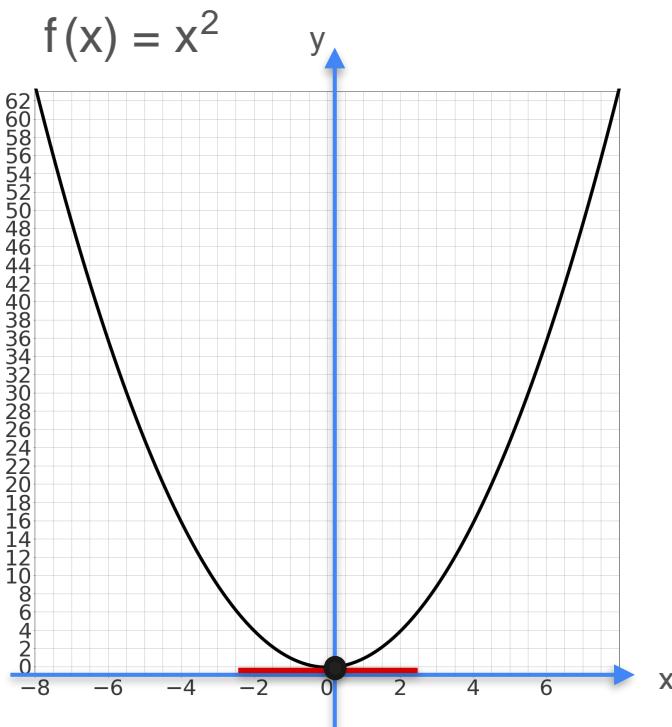
# Functions of Two Variables



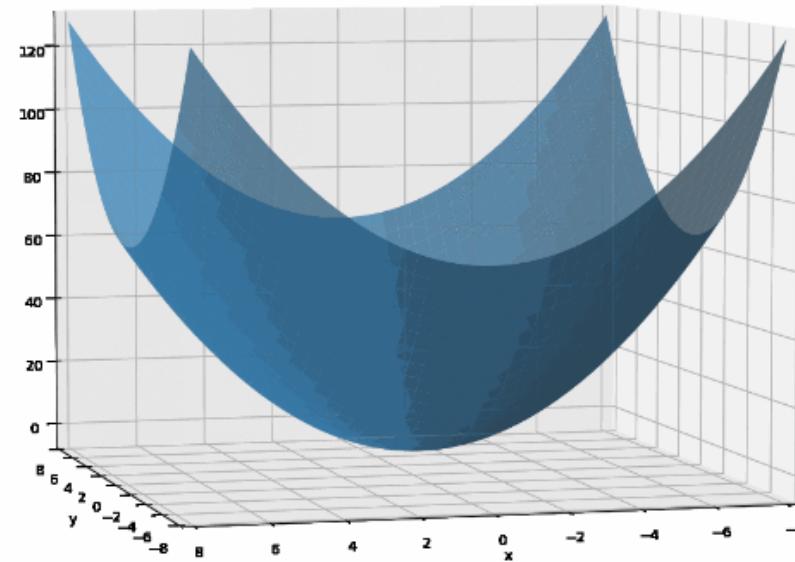
$$f(x, y) = x^2 + y^2$$



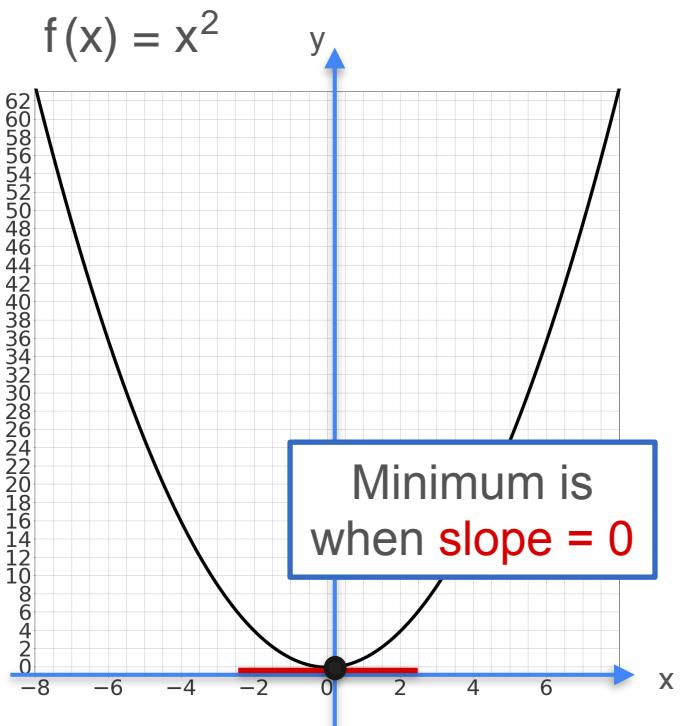
# Functions of Two Variables



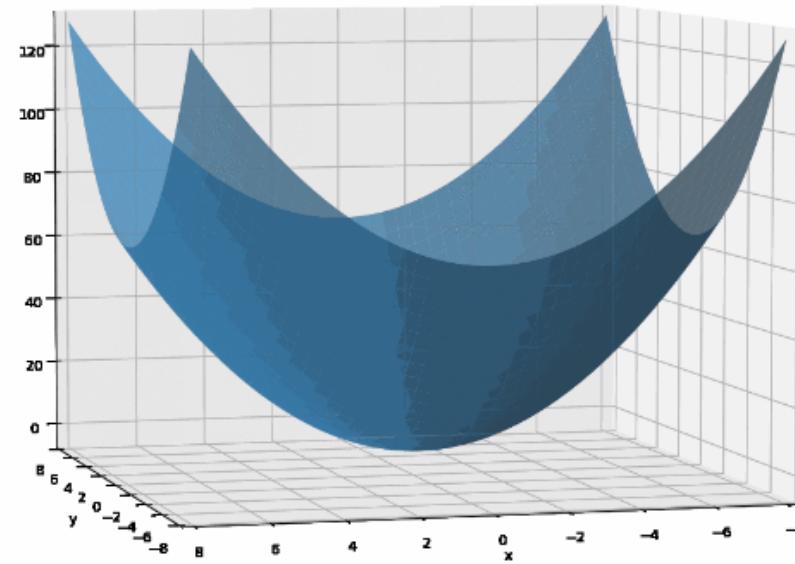
$$f(x, y) = x^2 + y^2$$



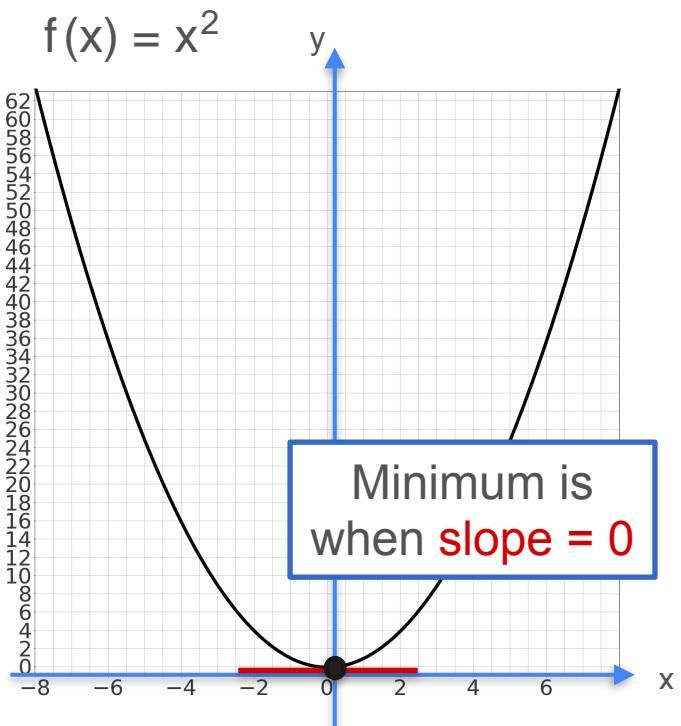
# Functions of Two Variables



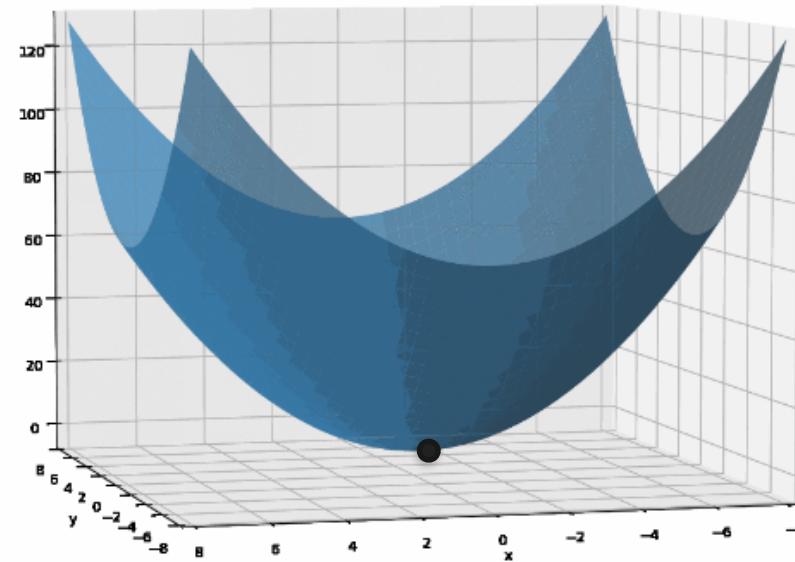
$$f(x, y) = x^2 + y^2$$



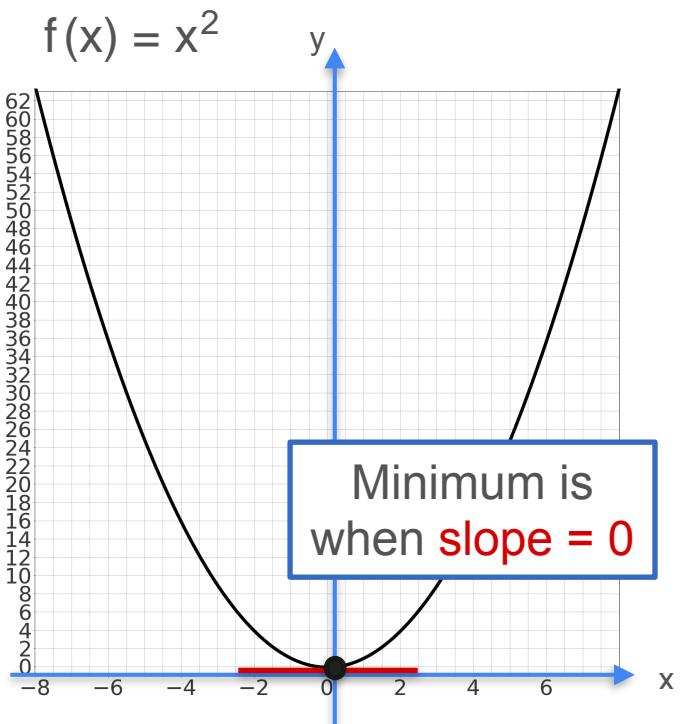
# Functions of Two Variables



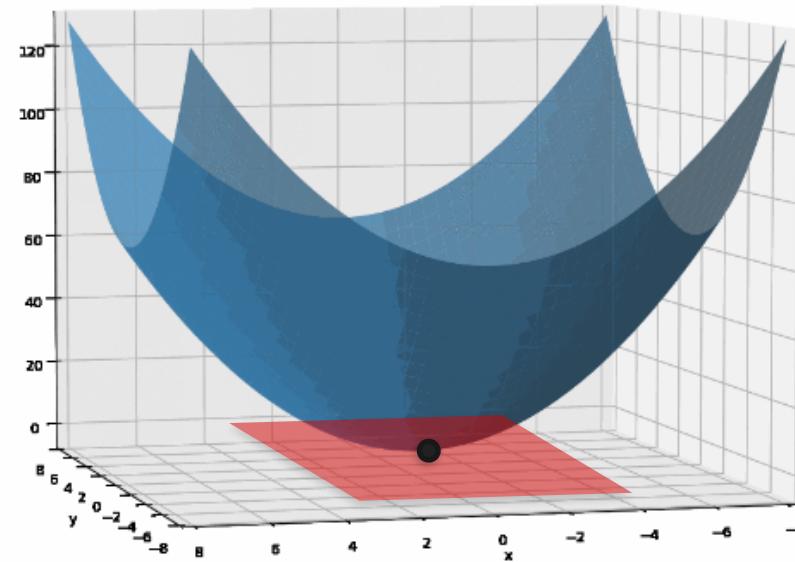
$$f(x, y) = x^2 + y^2$$



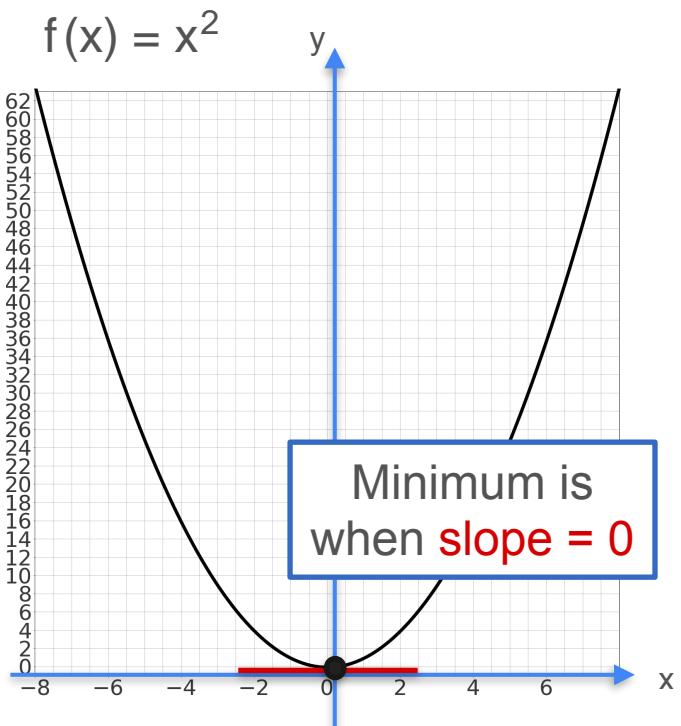
# Functions of Two Variables



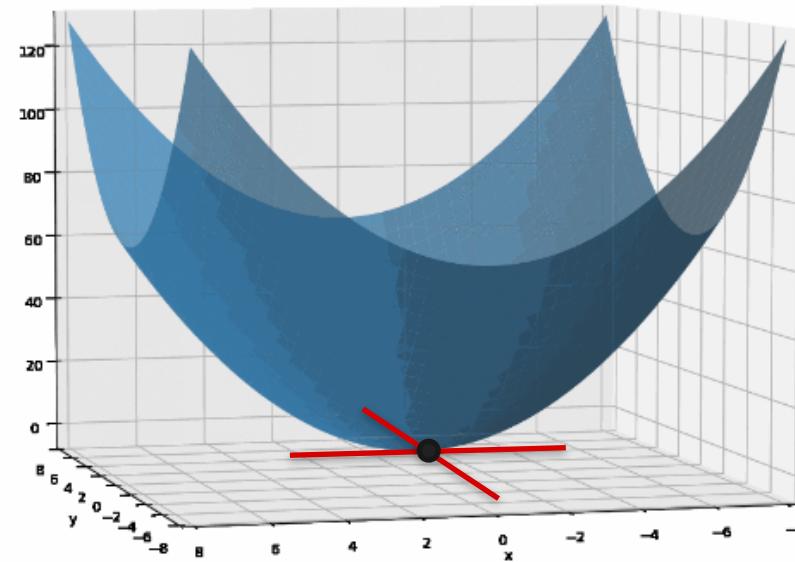
$$f(x, y) = x^2 + y^2$$



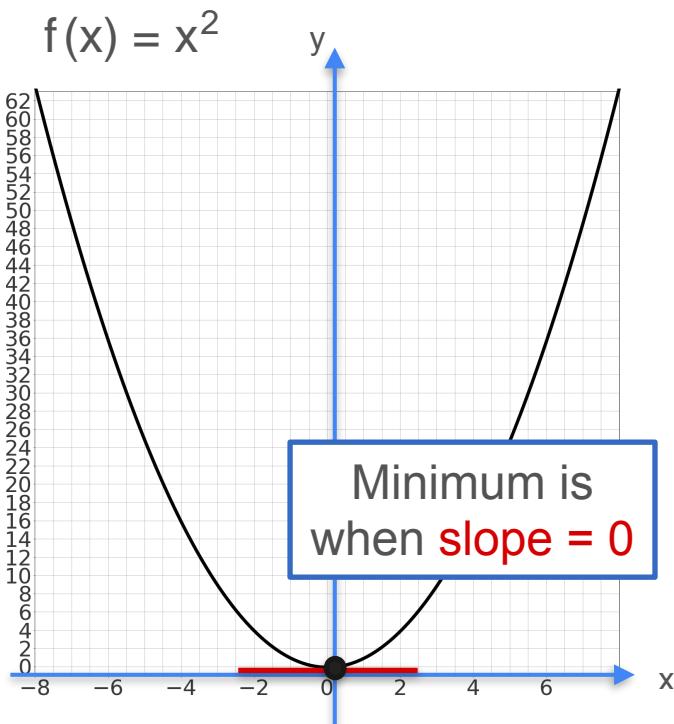
# Functions of Two Variables



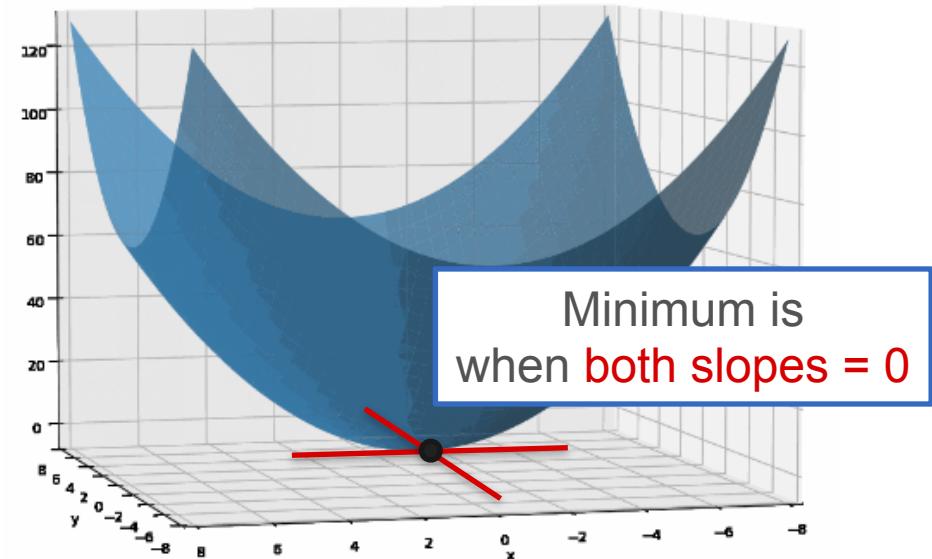
$$f(x, y) = x^2 + y^2$$



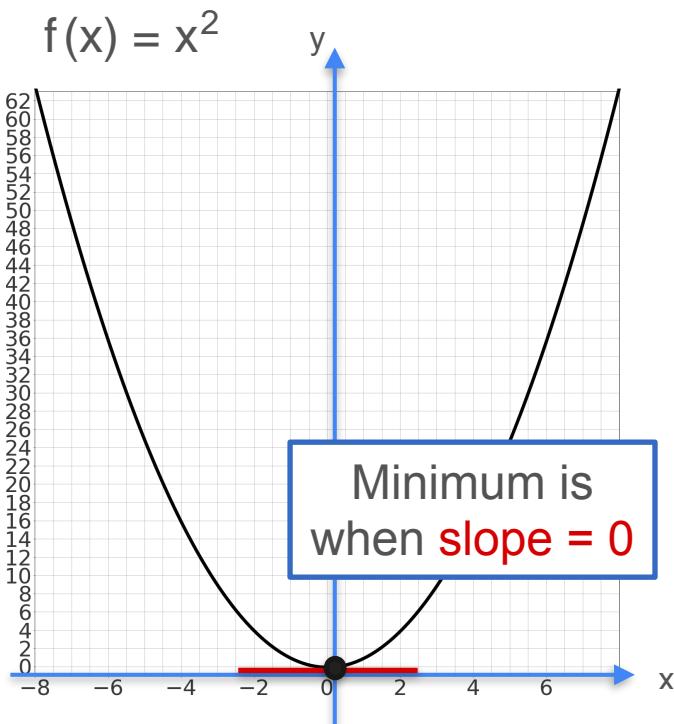
# Functions of Two Variables



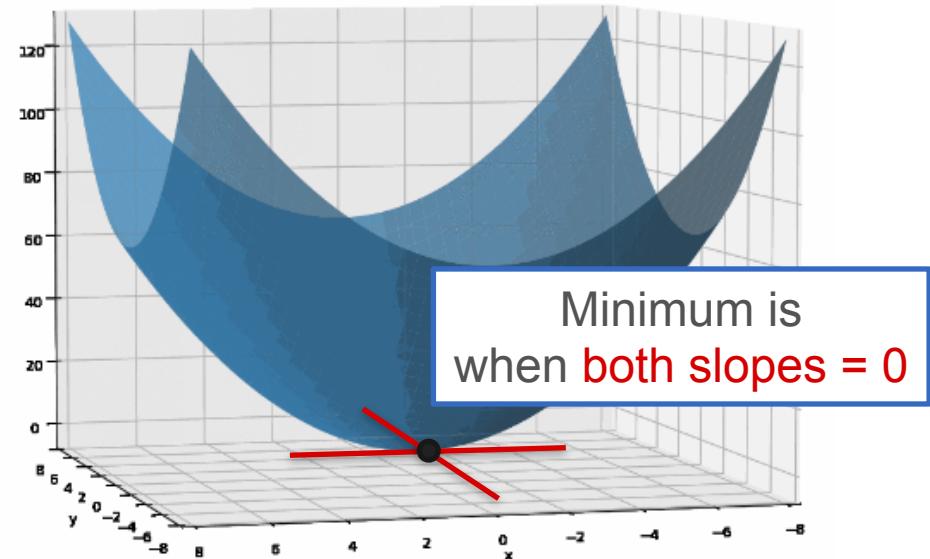
$$f(x, y) = x^2 + y^2$$



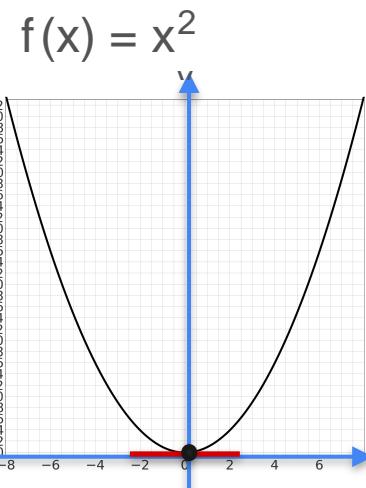
# Functions of Two Variables



$$f(x, y) = x^2 + y^2$$

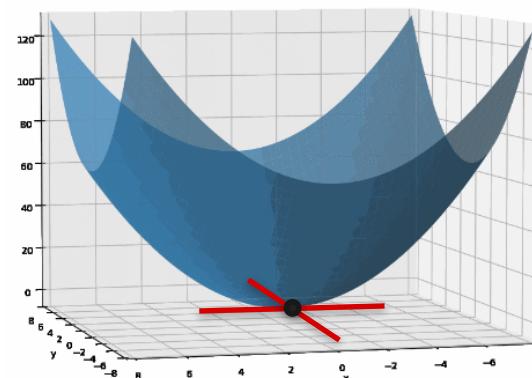


# Functions of Two Variables



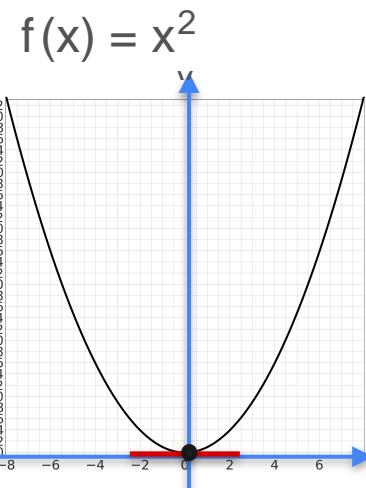
Minimum is  
when **slope = 0**

$$f(x, y) = x^2 + y^2$$



Minimum is  
when **both slopes = 0**

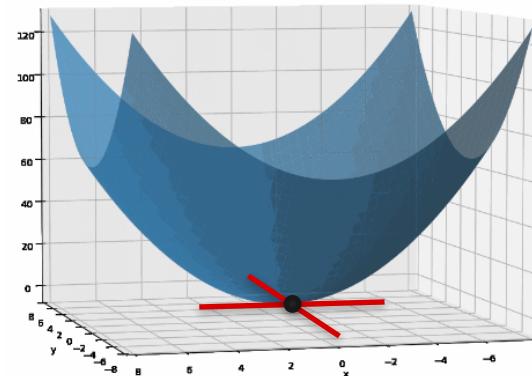
# Functions of Two Variables



Minimum is  
when **slope = 0**

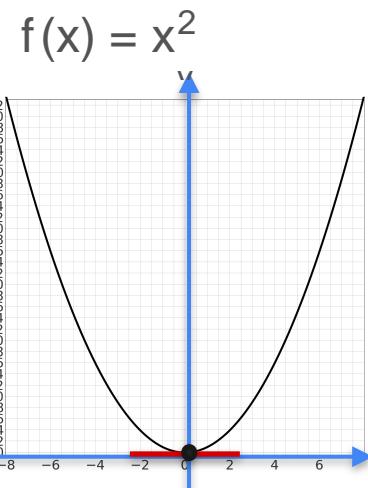
$$f(x) = 0$$

$$f(x, y) = x^2 + y^2$$



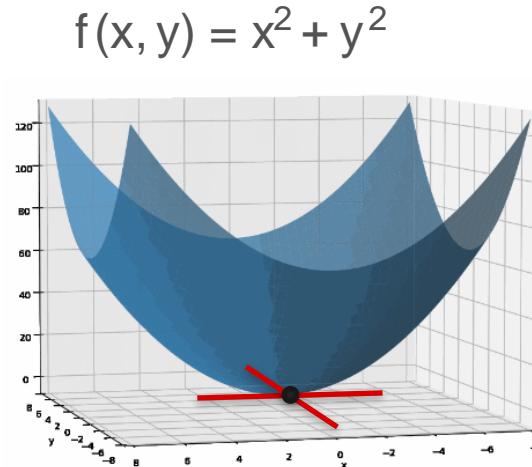
Minimum is  
when **both slopes = 0**

# Functions of Two Variables



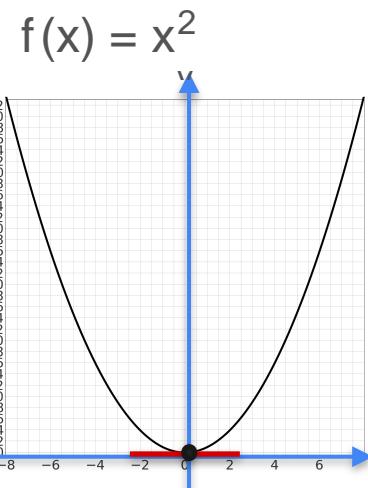
Minimum is  
when slope = 0

$$\begin{aligned}f'(x) &= 0 \\2x &= 0\end{aligned}$$



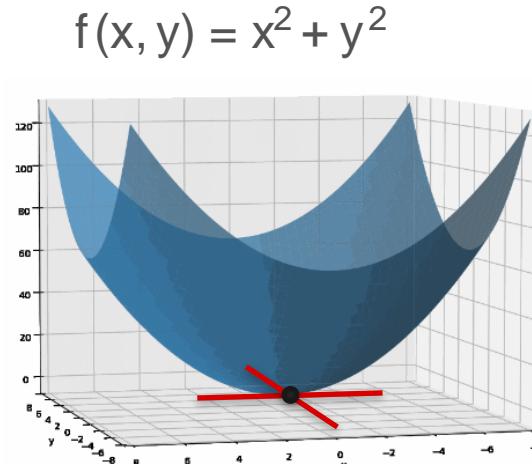
Minimum is  
when both slopes = 0

# Functions of Two Variables



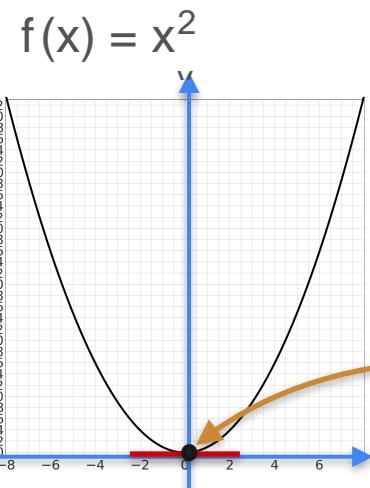
Minimum is  
when **slope = 0**

$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



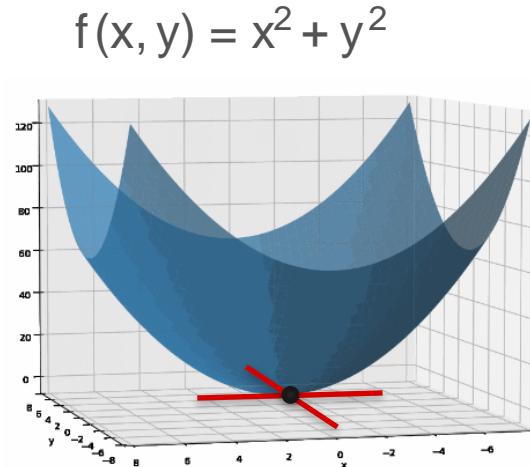
Minimum is  
when **both slopes = 0**

# Functions of Two Variables



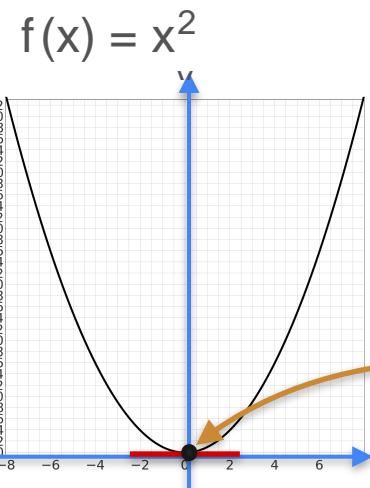
Minimum is  
when slope = 0

$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



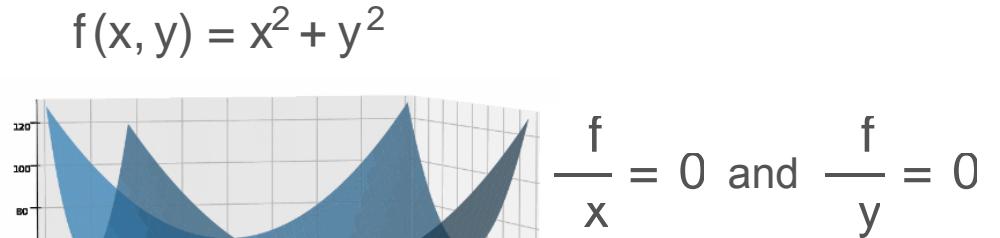
Minimum is  
when both slopes = 0

# Functions of Two Variables



Minimum is  
when slope = 0

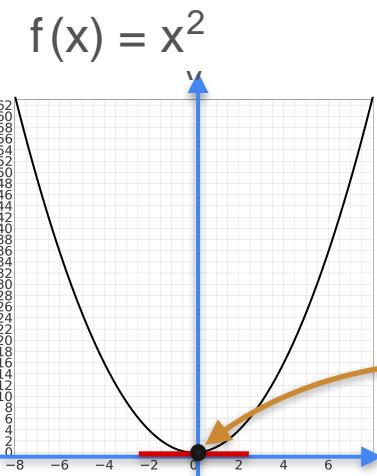
$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



$$\frac{\partial f}{\partial x} = 0 \text{ and } \frac{\partial f}{\partial y} = 0$$

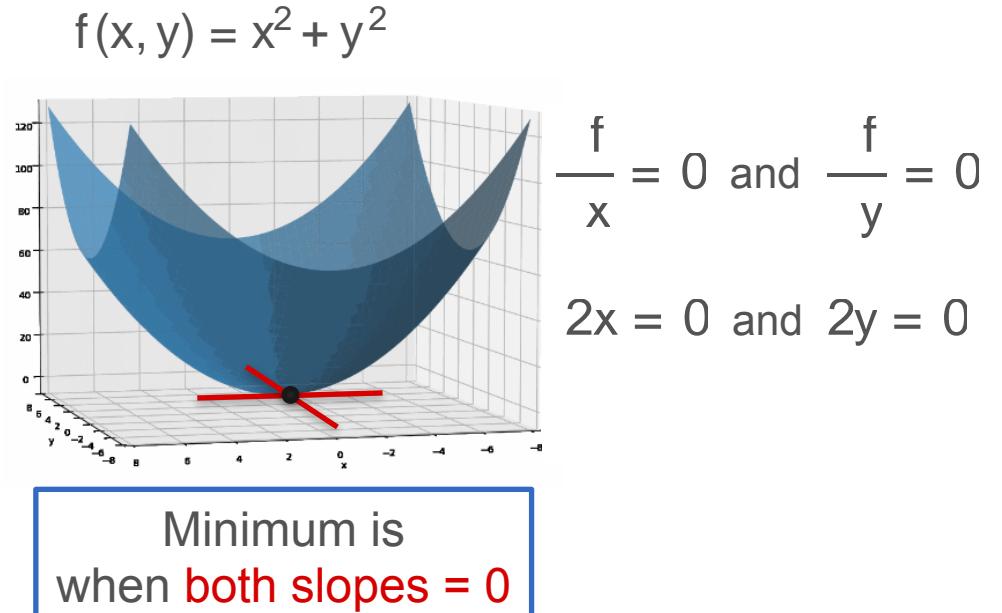
Minimum is  
when both slopes = 0

# Functions of Two Variables



Minimum is  
when slope = 0

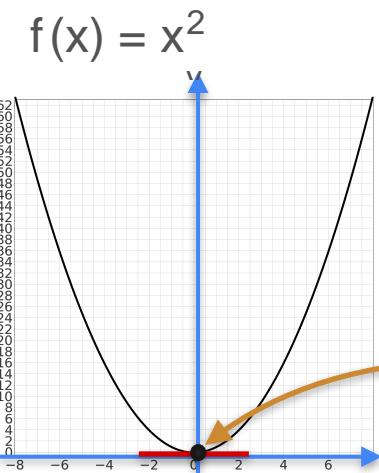
$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



$$\begin{aligned}\frac{f}{x} &= 0 \text{ and } \frac{f}{y} = 0 \\2x &= 0 \text{ and } 2y = 0\end{aligned}$$

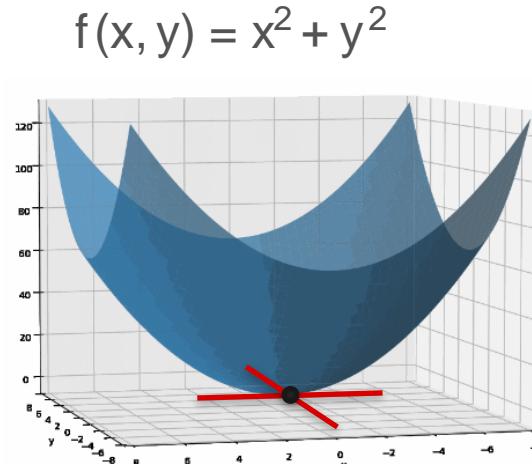
Minimum is  
when both slopes = 0

# Functions of Two Variables



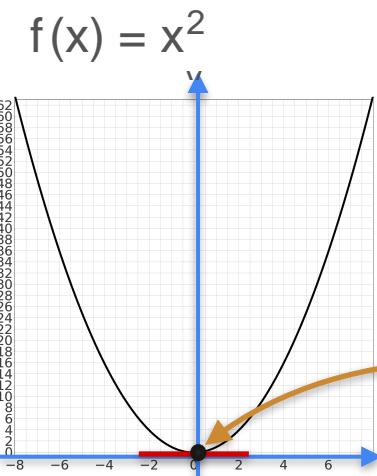
Minimum is  
when slope = 0

$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



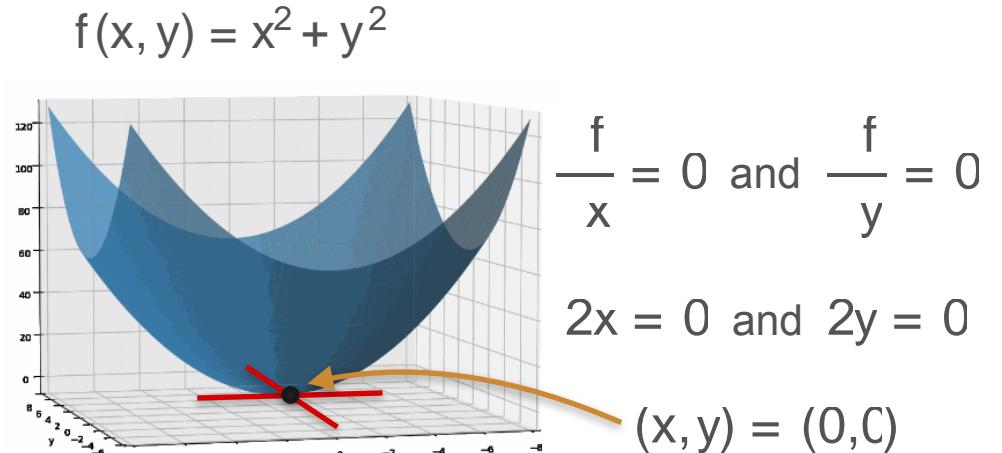
Minimum is  
when both slopes = 0

# Functions of Two Variables



Minimum is  
when slope = 0

$$\begin{aligned}f(x) &= 0 \\2x &= 0 \\x &= 0\end{aligned}$$



Minimum is  
when both slopes = 0

$$\begin{aligned}\frac{f}{x} &= 0 \text{ and } \frac{f}{y} = 0 \\2x &= 0 \text{ and } 2y = 0 \\(x, y) &= (0, 0)\end{aligned}$$



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## Gradients and Gradient Descent

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**Optimization with gradients:  
An example**

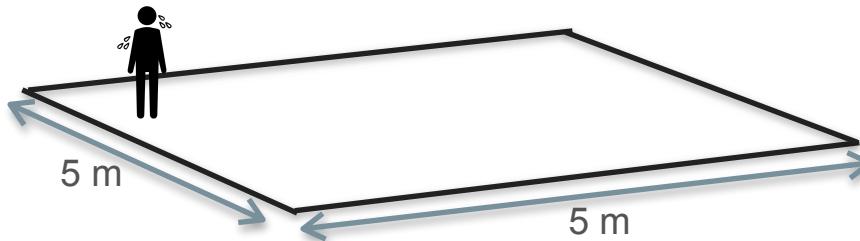
# Motivation for Optimization in Two Variables



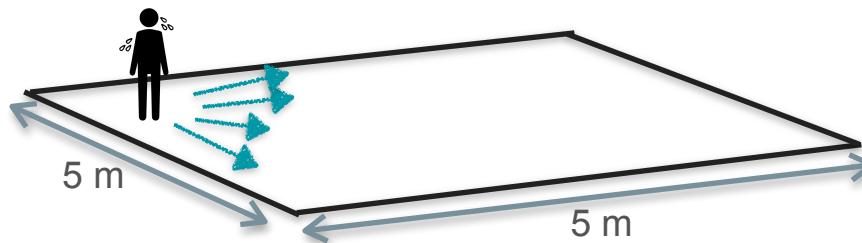
# Motivation for Optimization in Two Variables



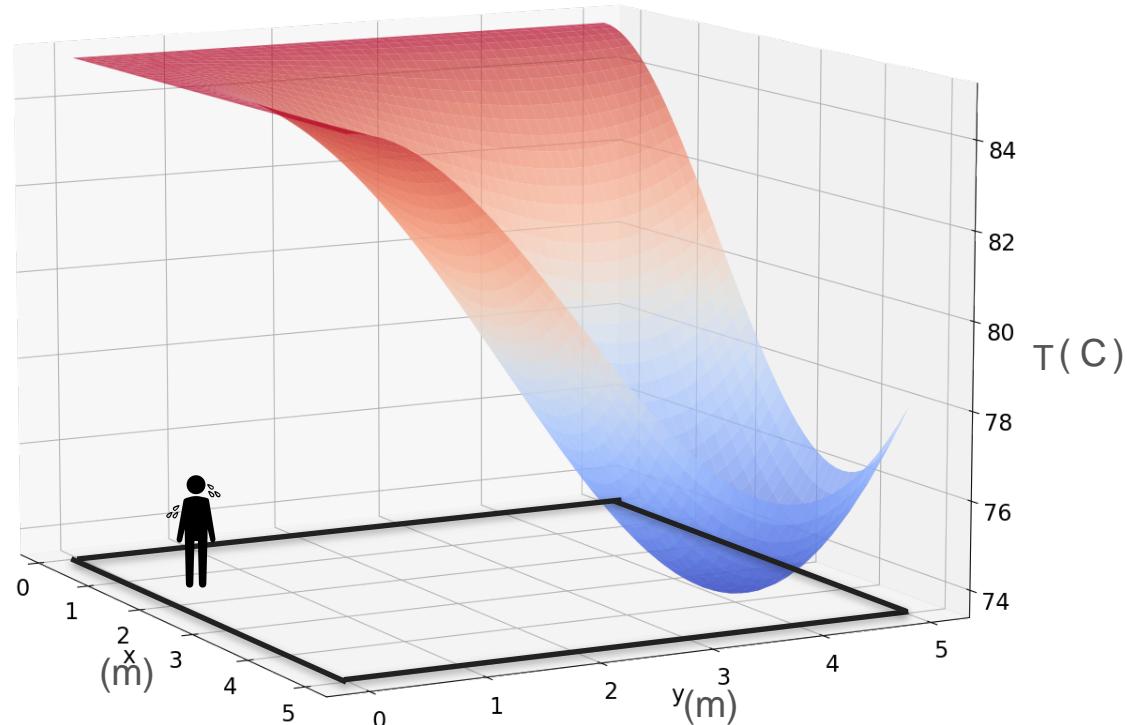
# Motivation for Optimization in Two Variables



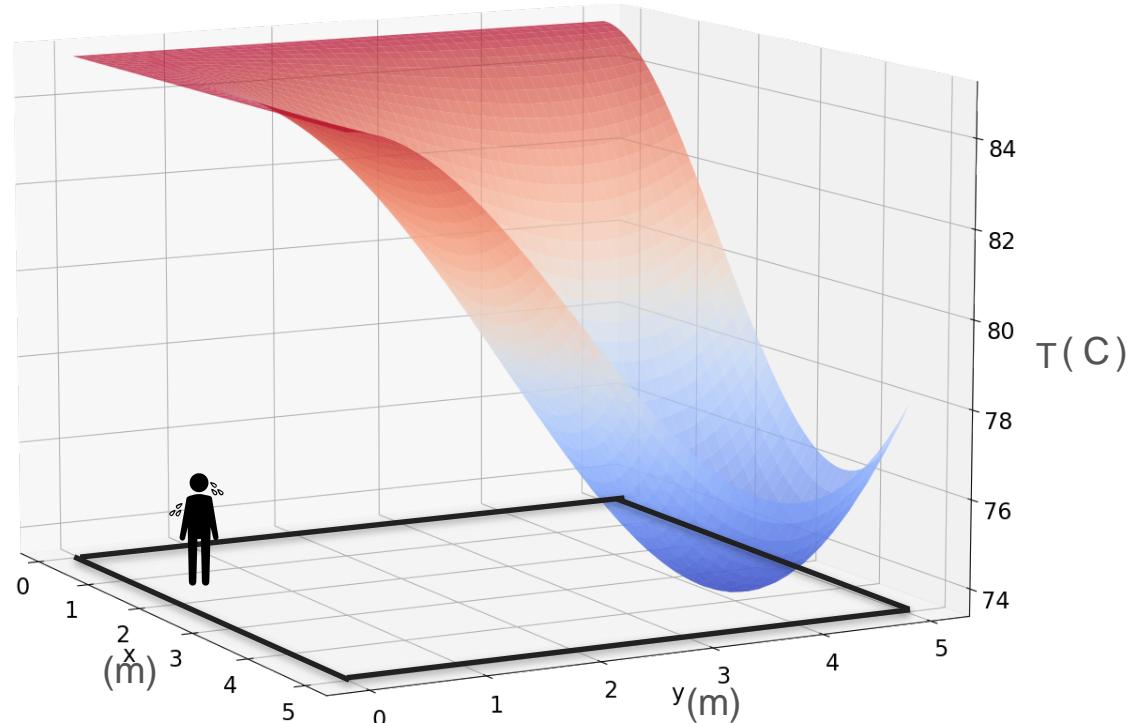
# Motivation for Optimization in Two Variables



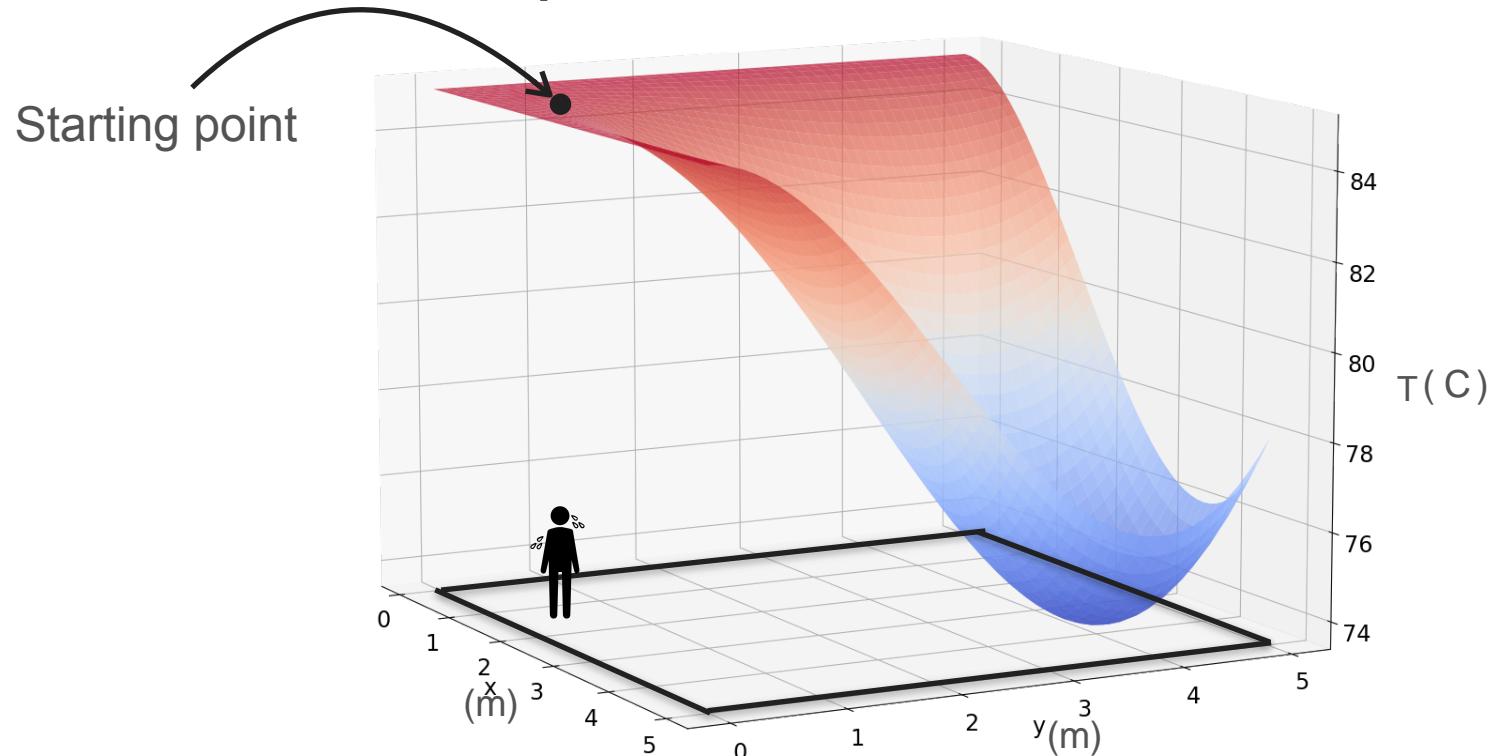
# Motivation for Optimization in Two Variables



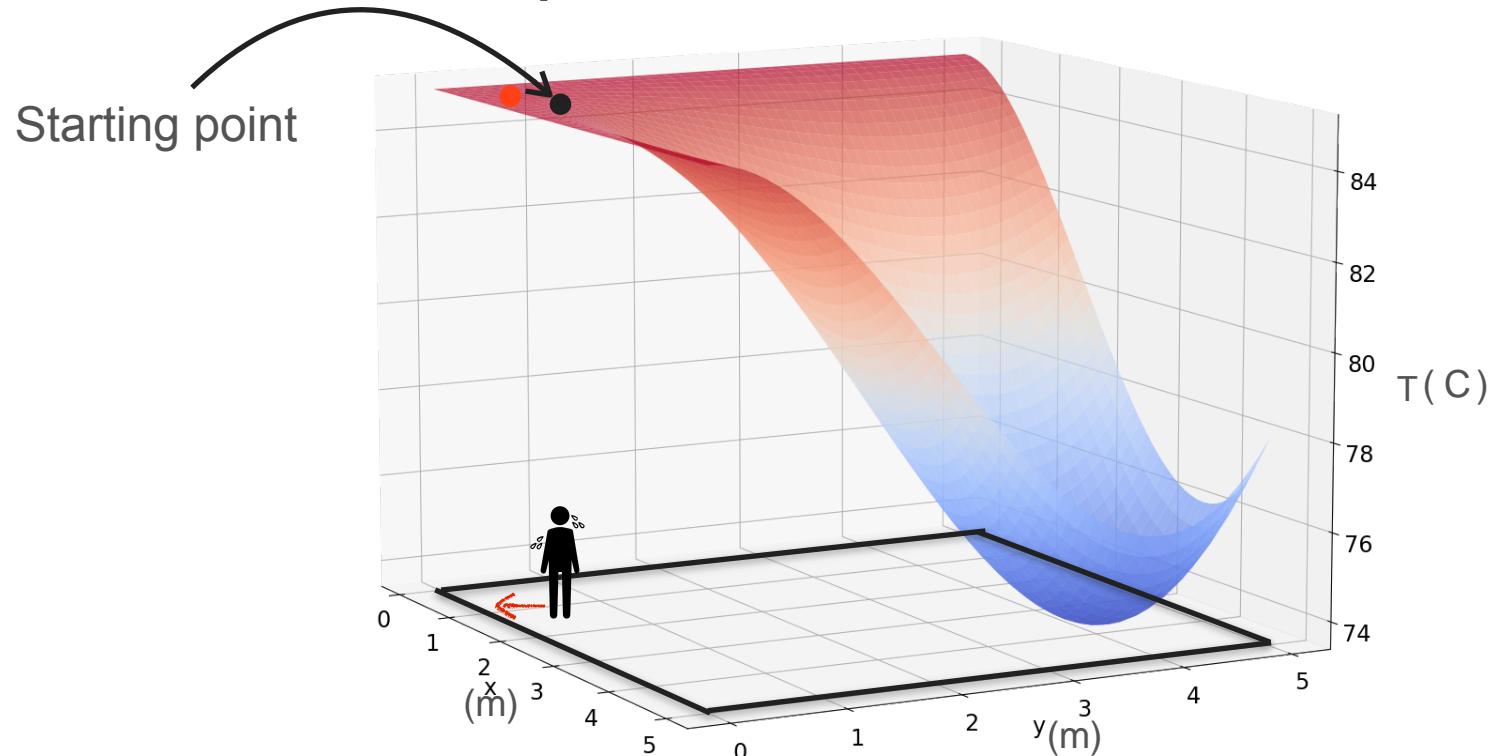
# Motivation for Optimization in Two Variables



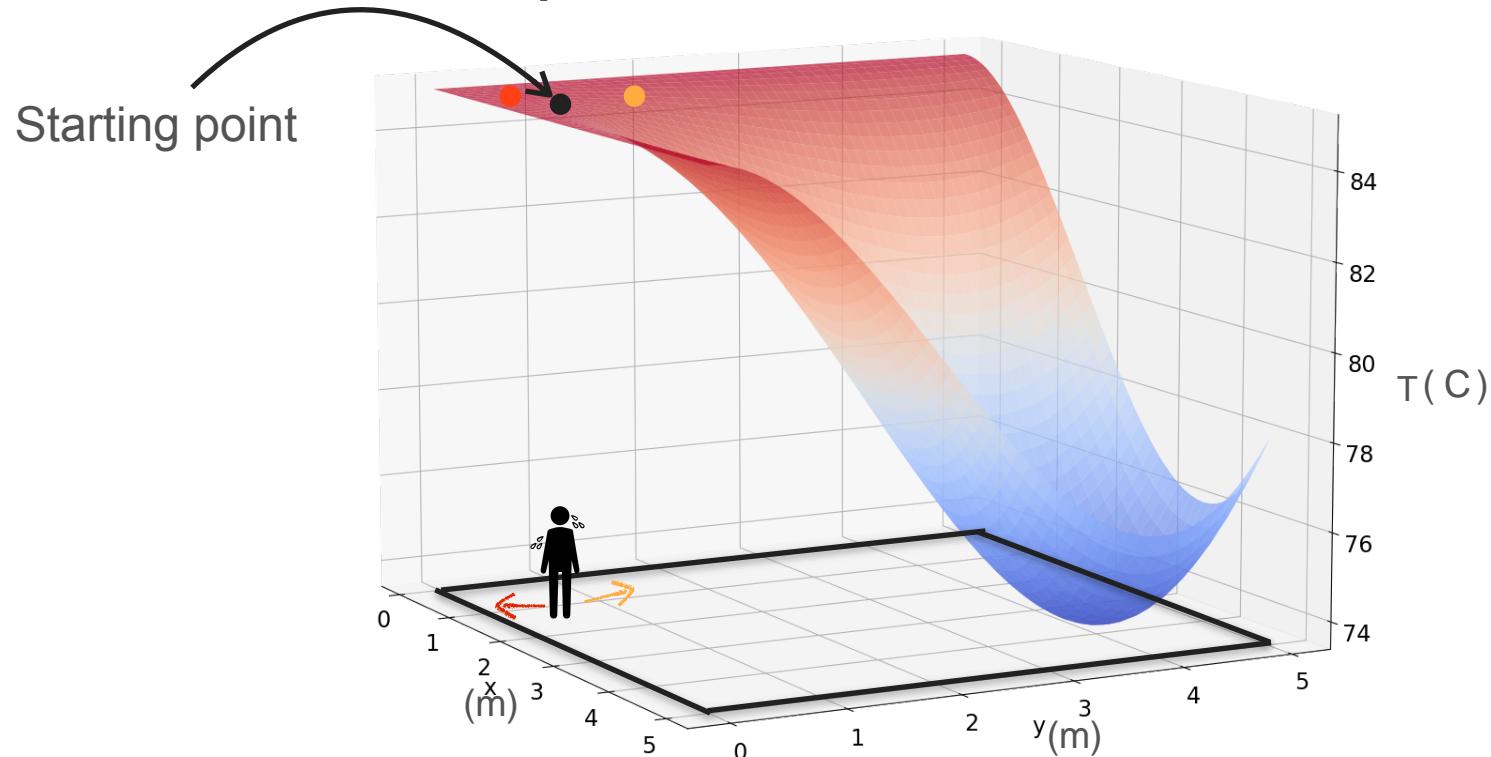
# Motivation for Optimization in Two Variables



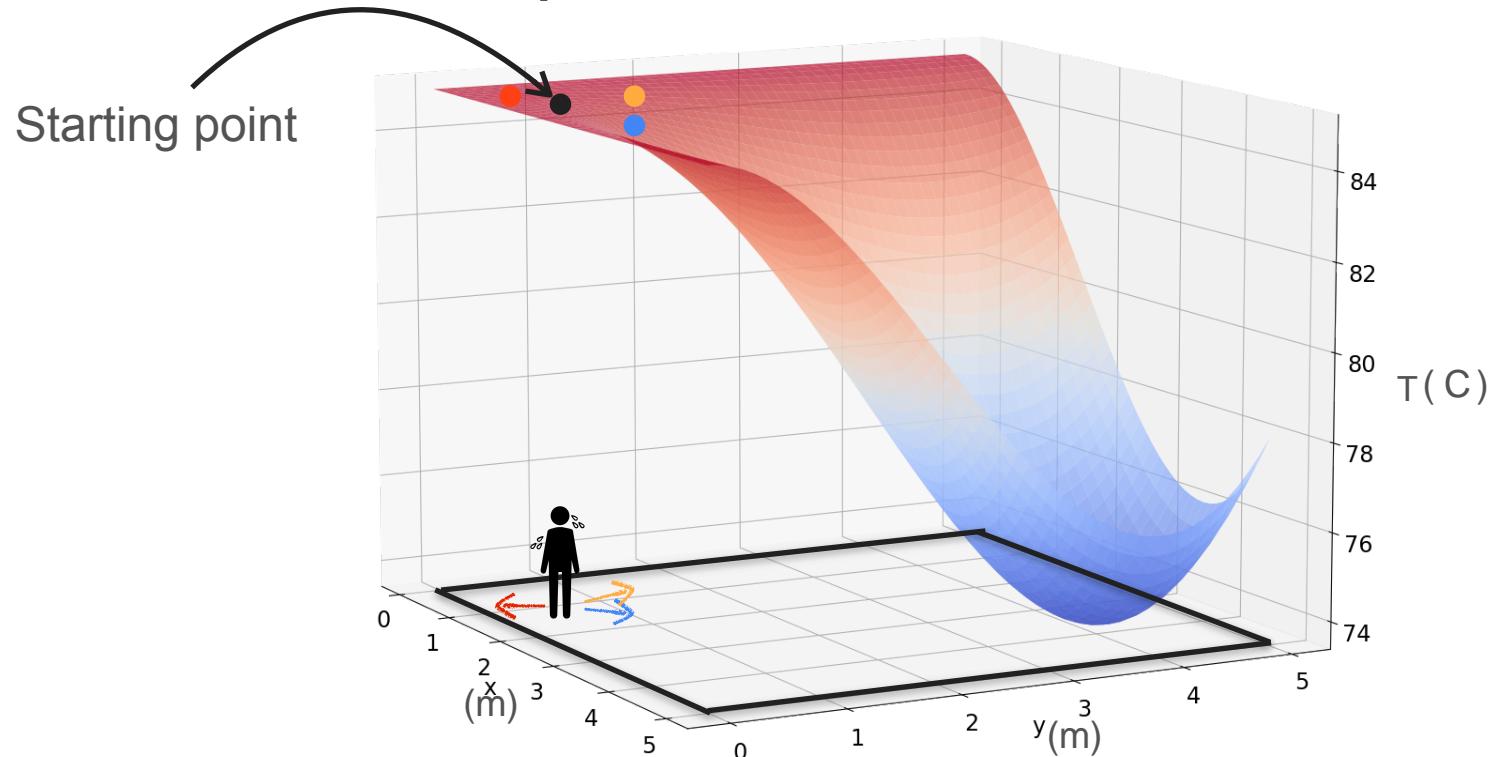
# Motivation for Optimization in Two Variables



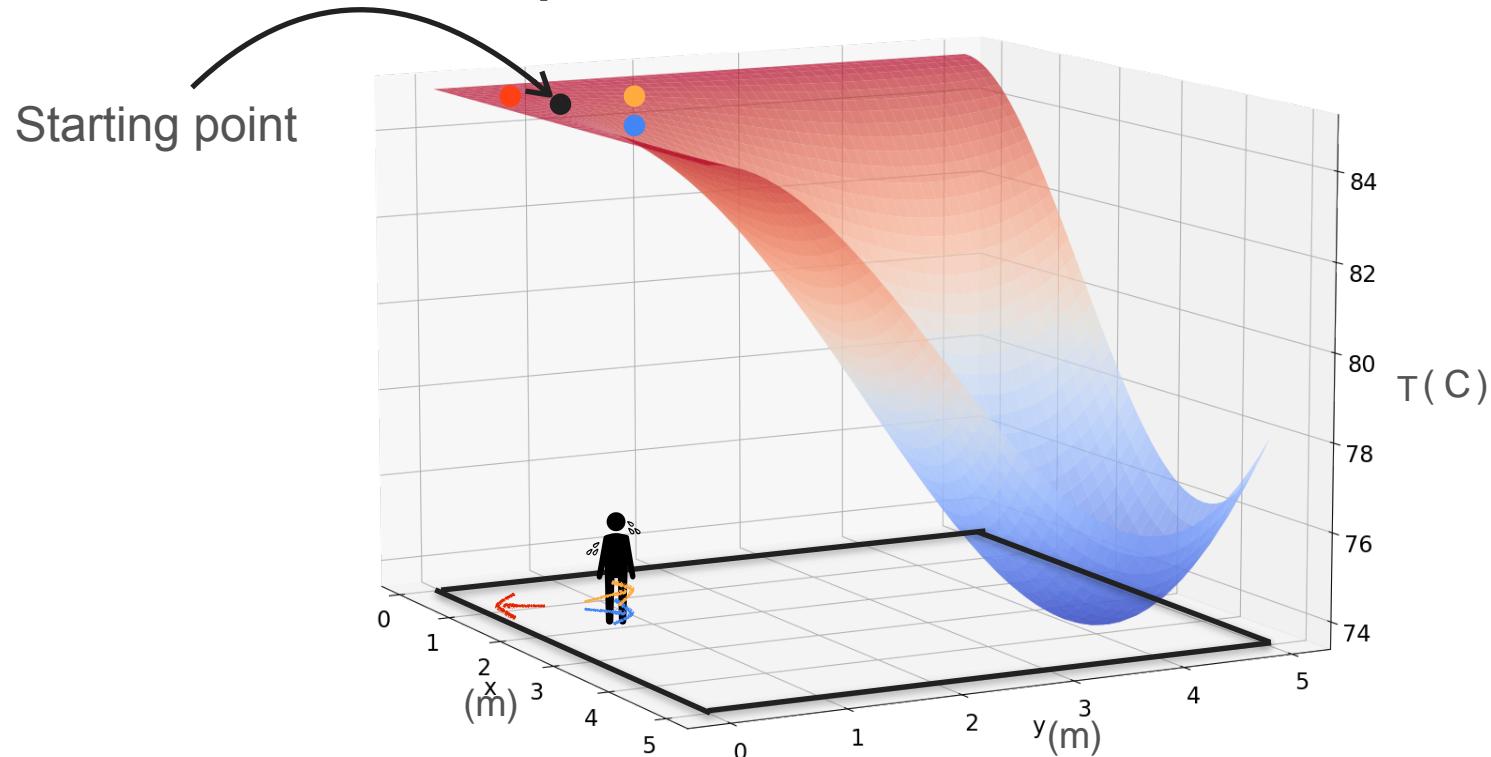
# Motivation for Optimization in Two Variables



# Motivation for Optimization in Two Variables

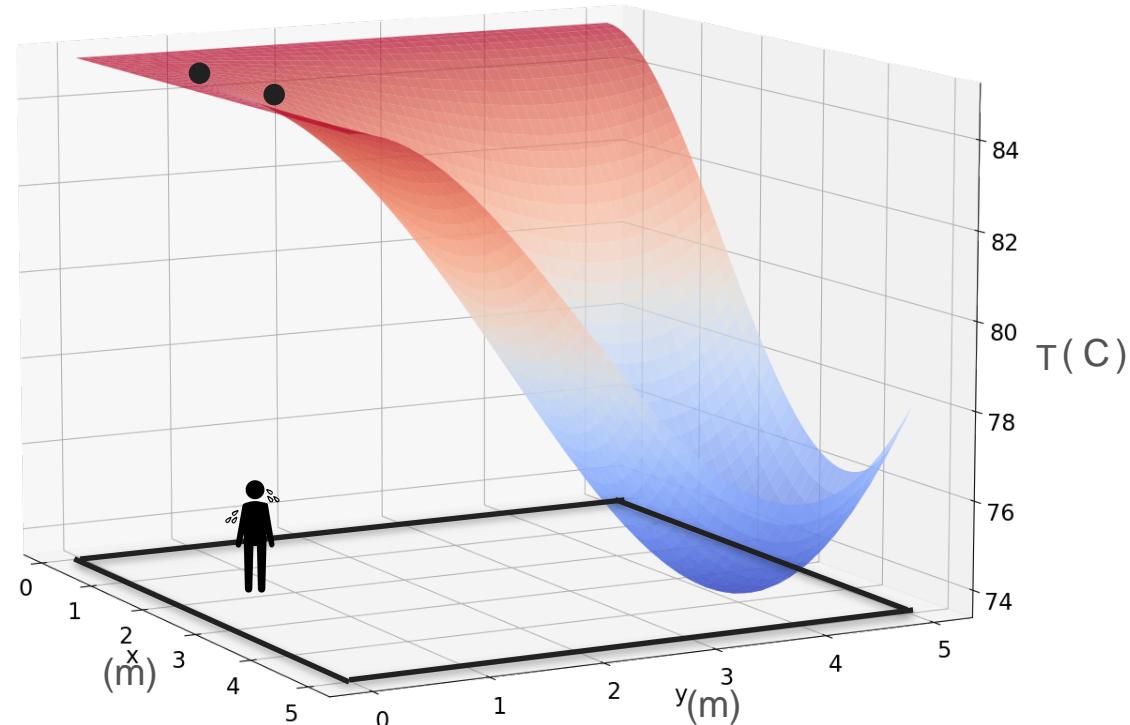


# Motivation for Optimization in Two Variables

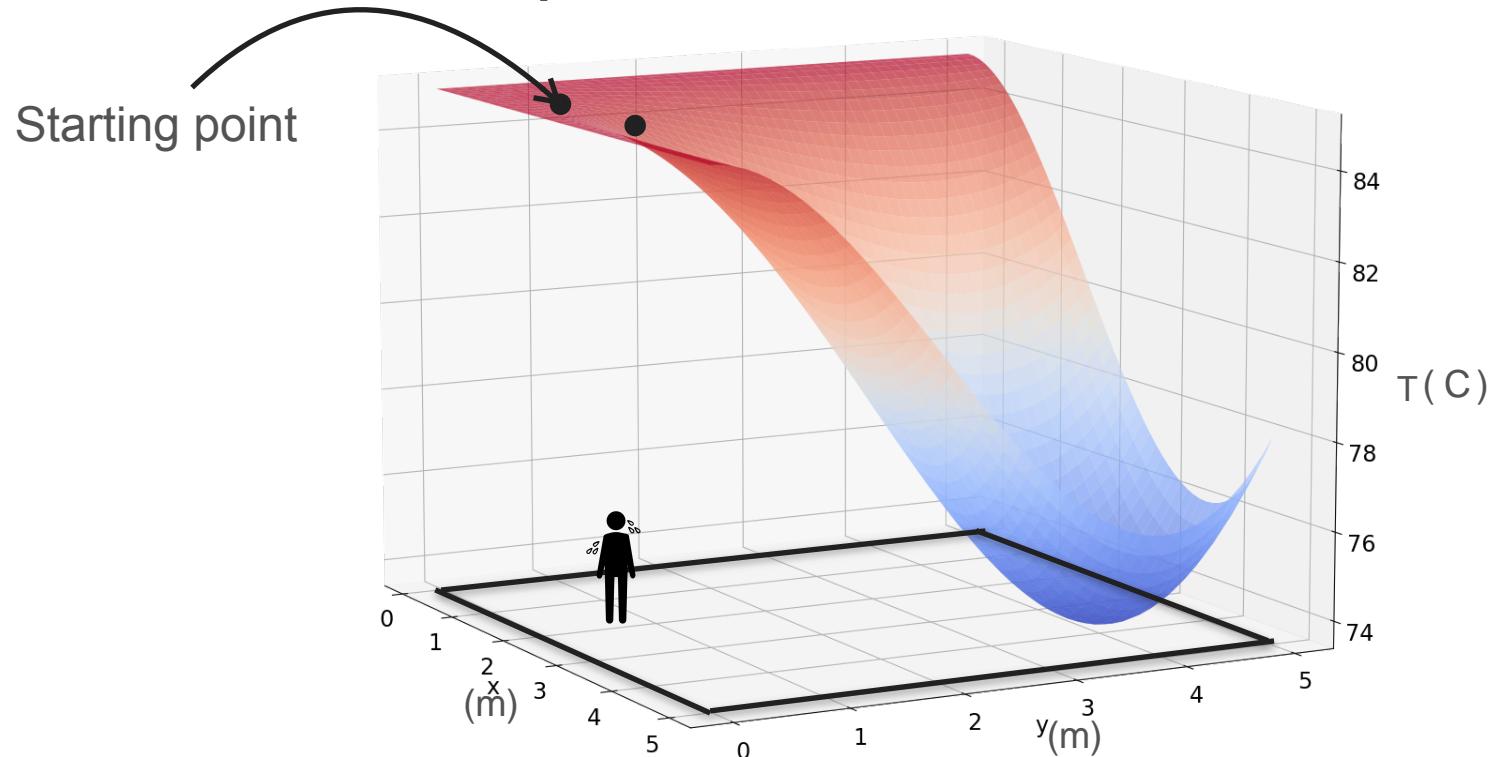


# Motivation for Optimization in Two Variables

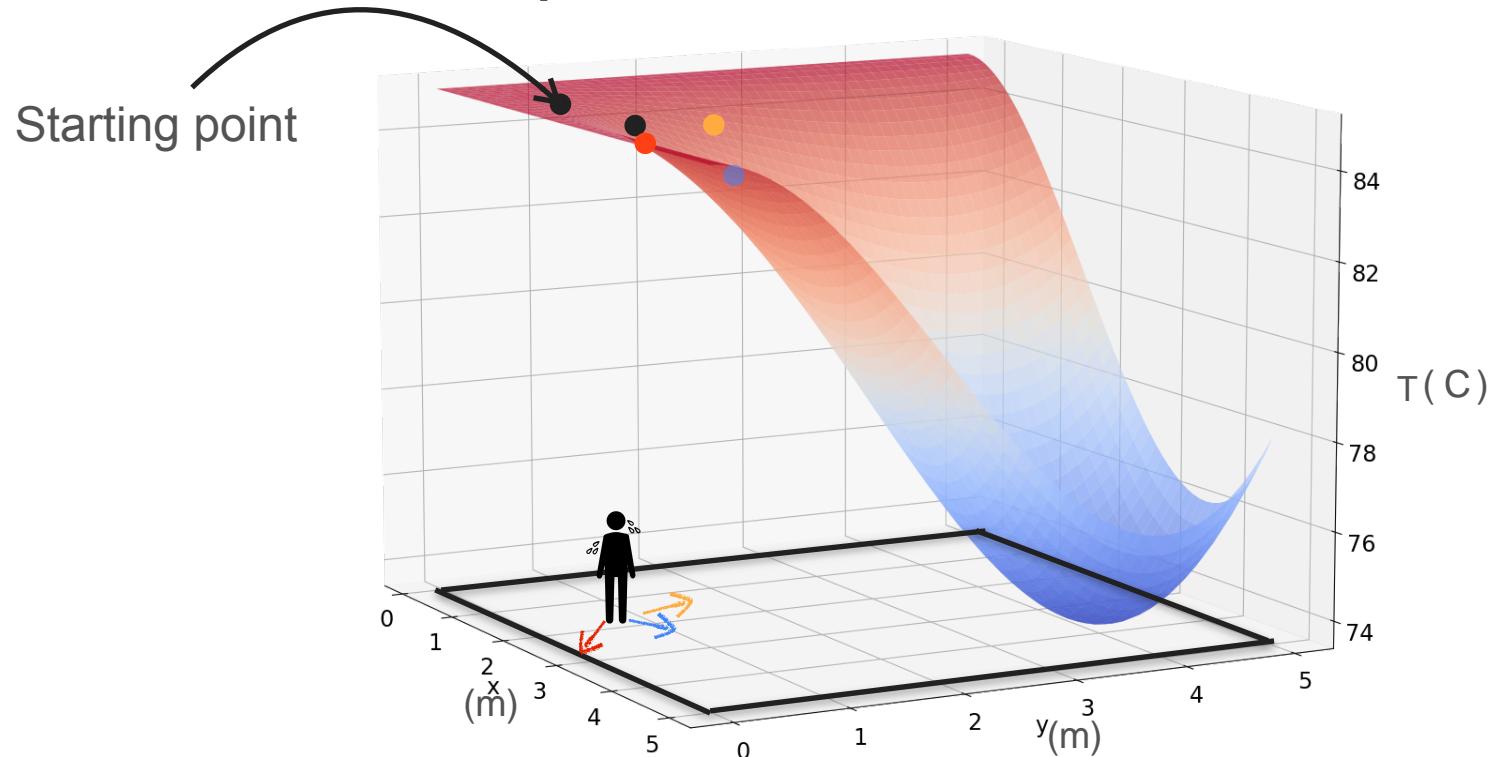
Starting point



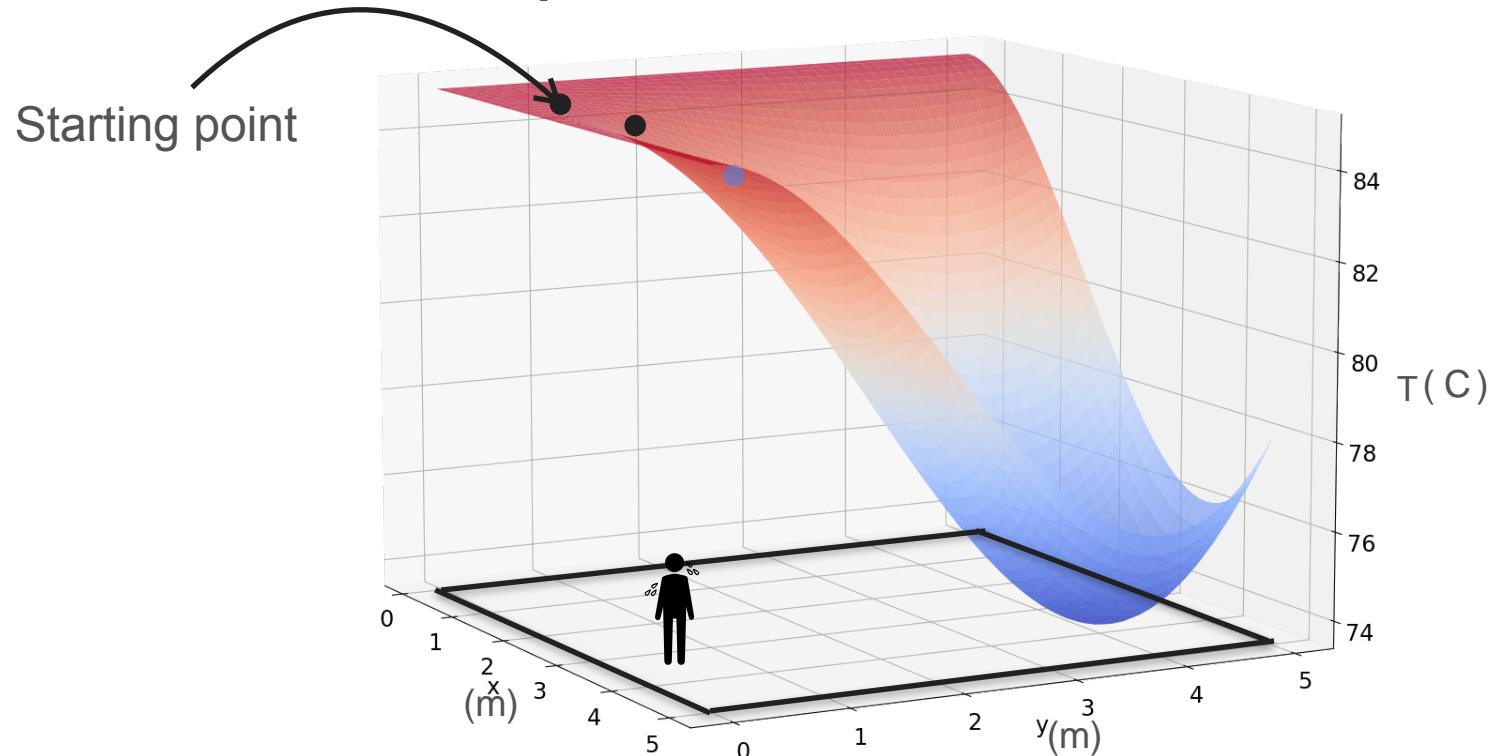
# Motivation for Optimization in Two Variables



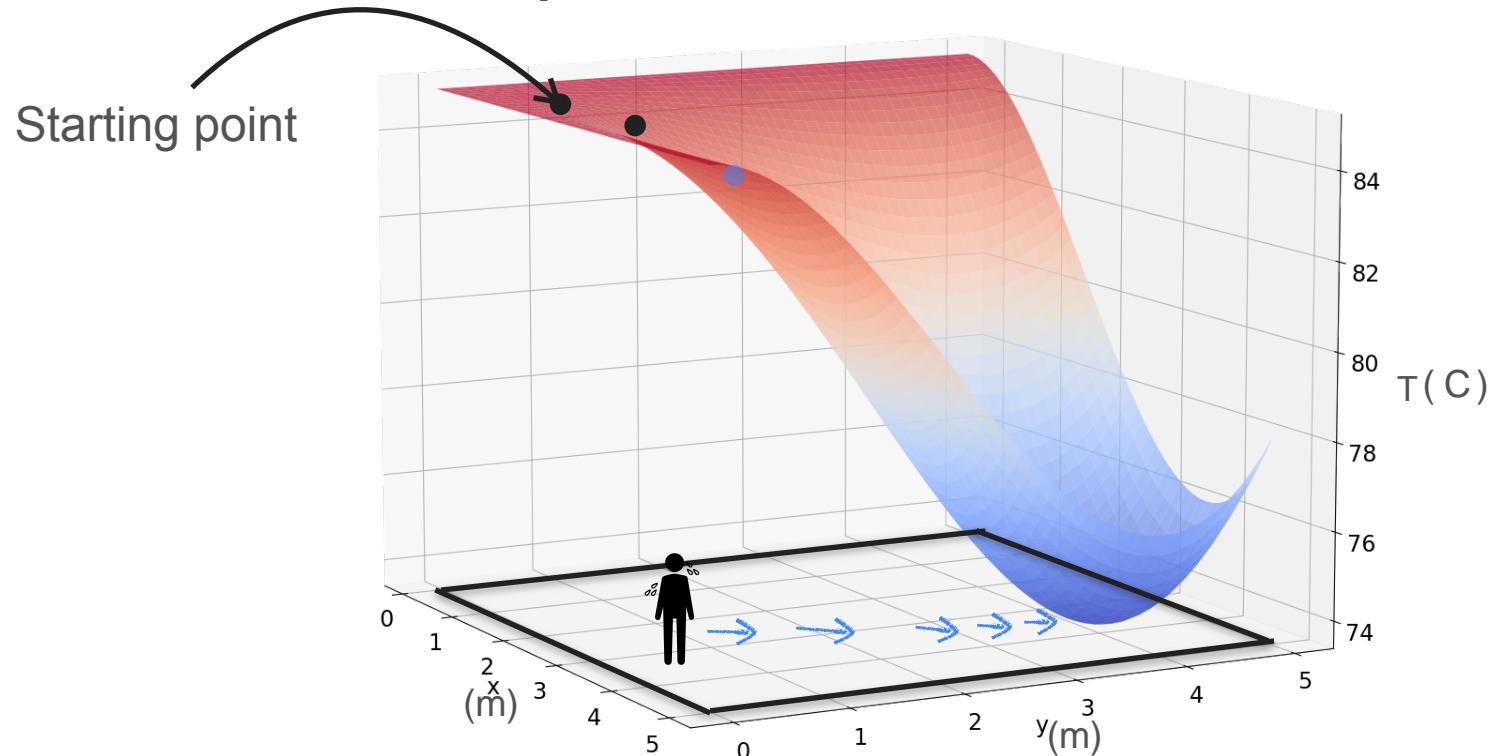
# Motivation for Optimization in Two Variables



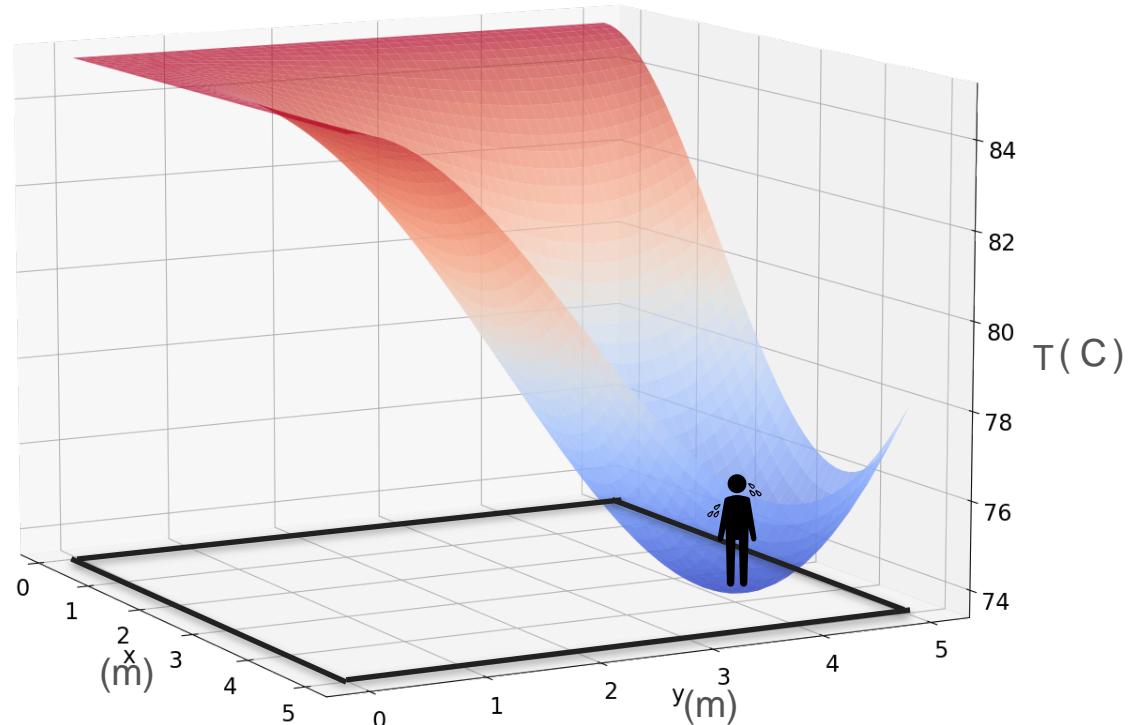
# Motivation for Optimization in Two Variables



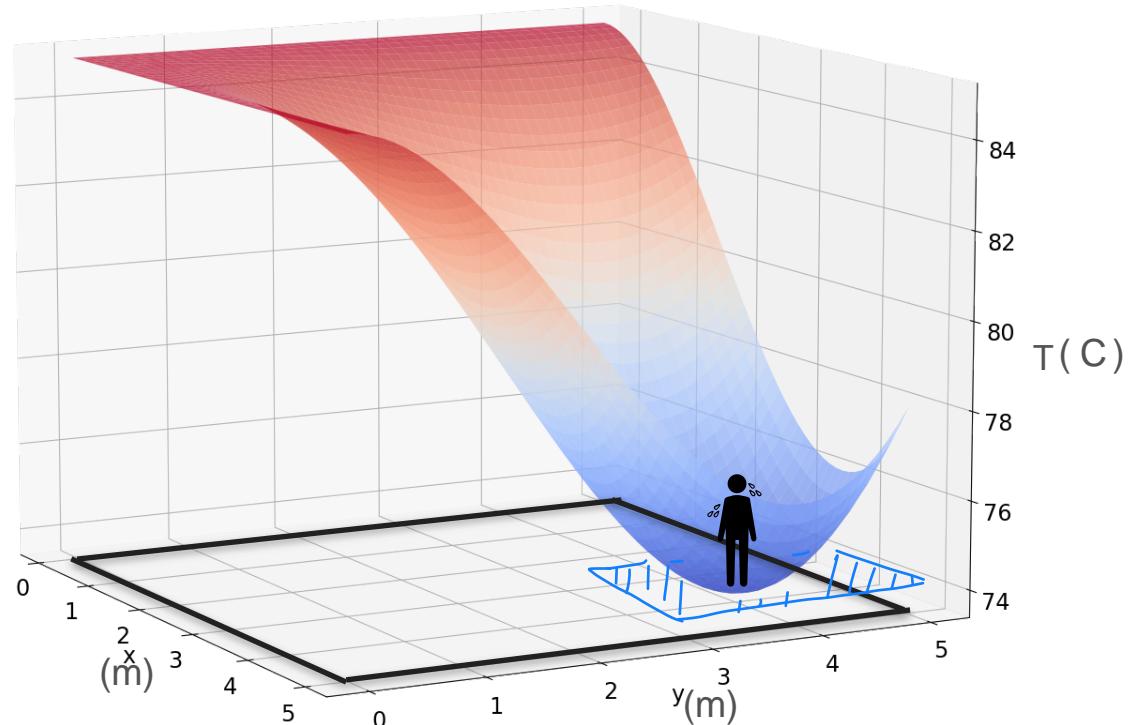
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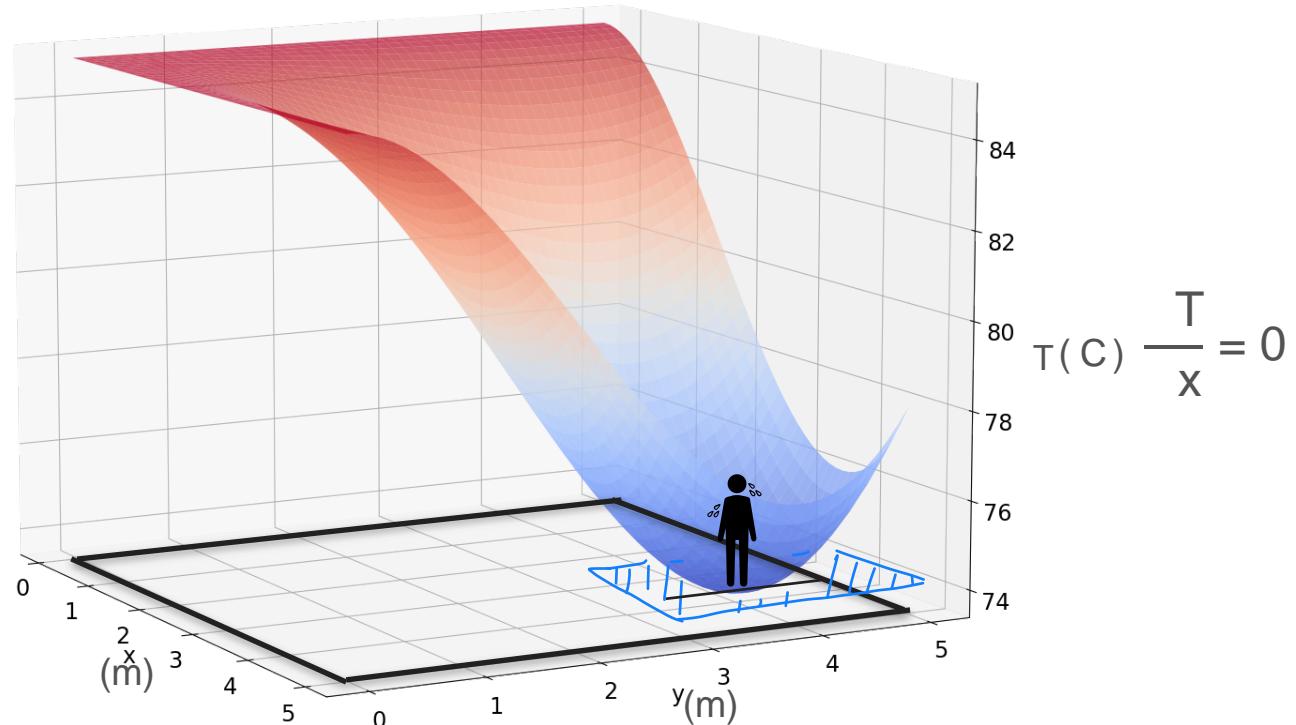
# Motivation for Optimization in Two Variables



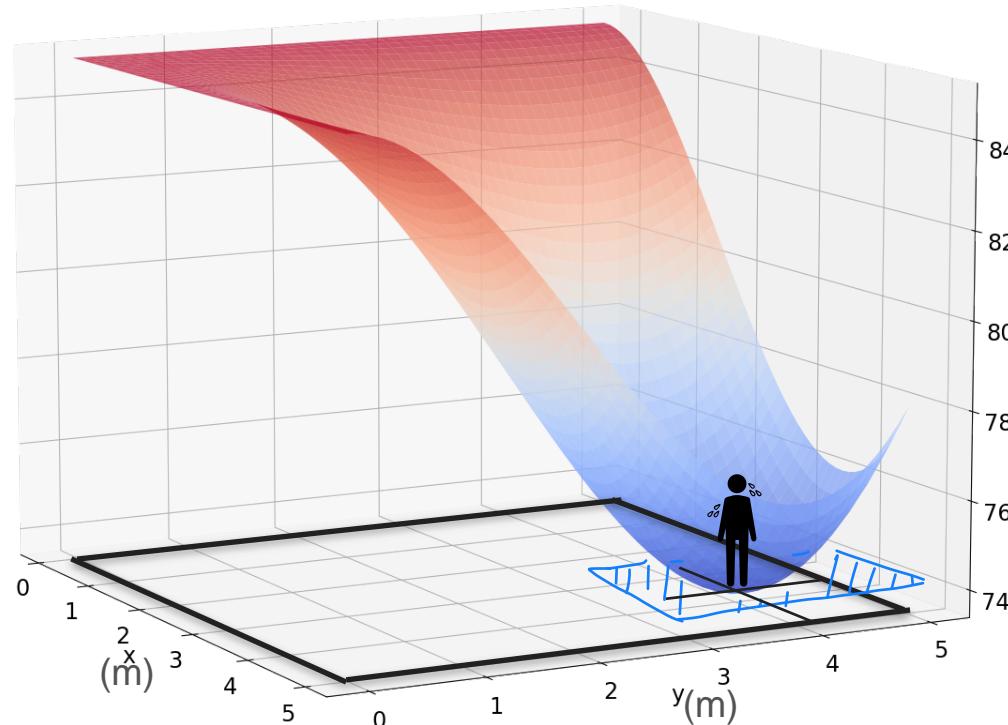
# Motivation for Optimization in Two Variables



# Motivation for Optimization in Two Variables



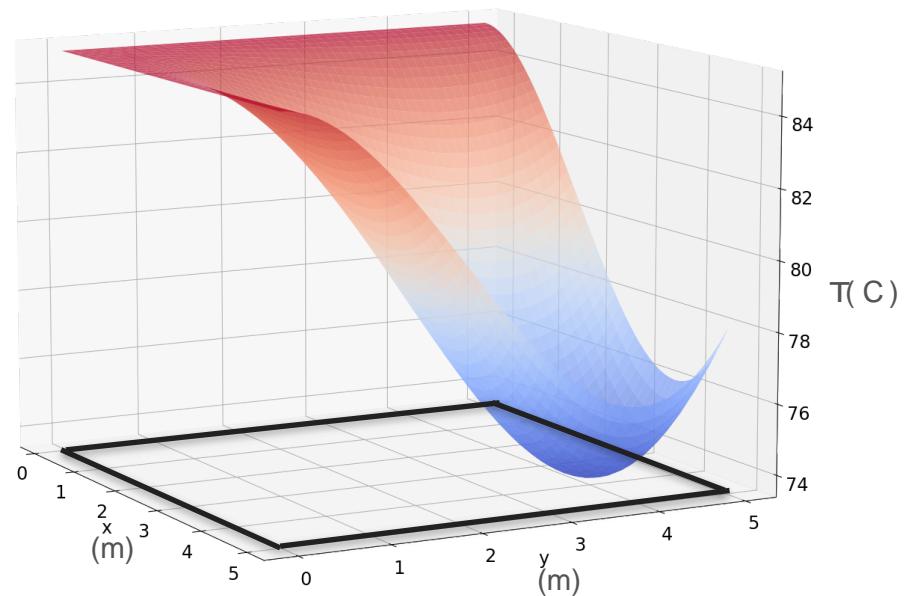
# Motivation for Optimization in Two Variables



$$T(C) \frac{\partial}{\partial x} = 0$$

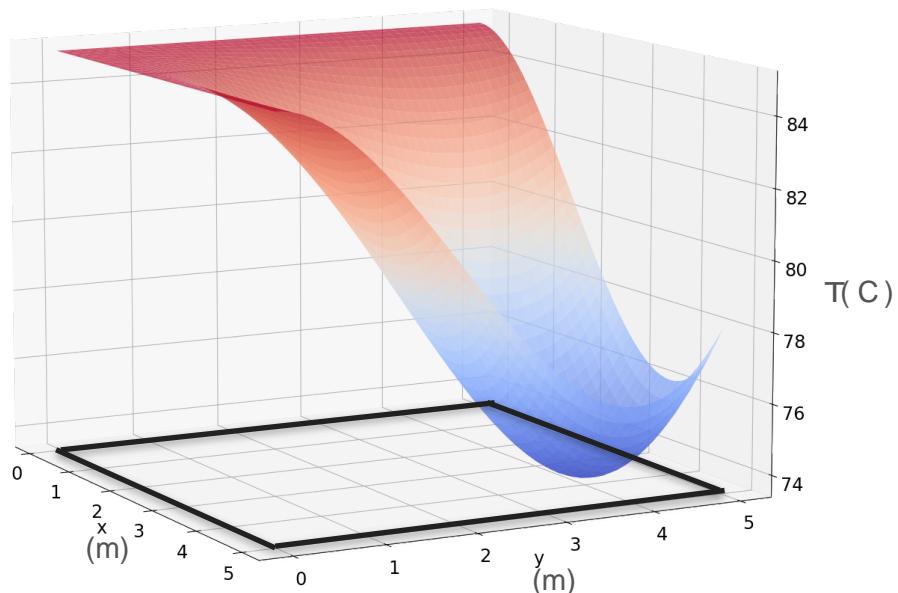
$$\frac{\partial}{\partial y} = 0$$

# Exercise



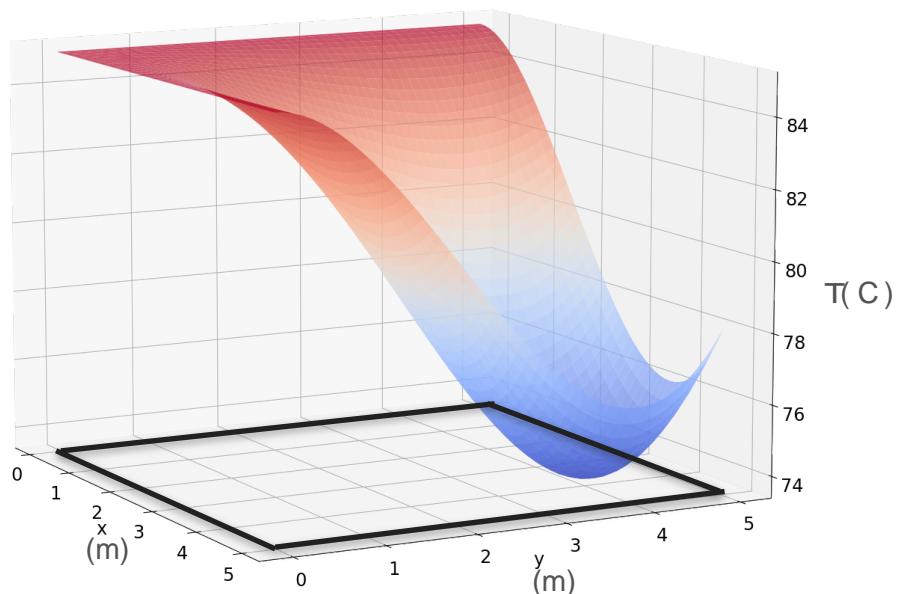
# Exercise

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$



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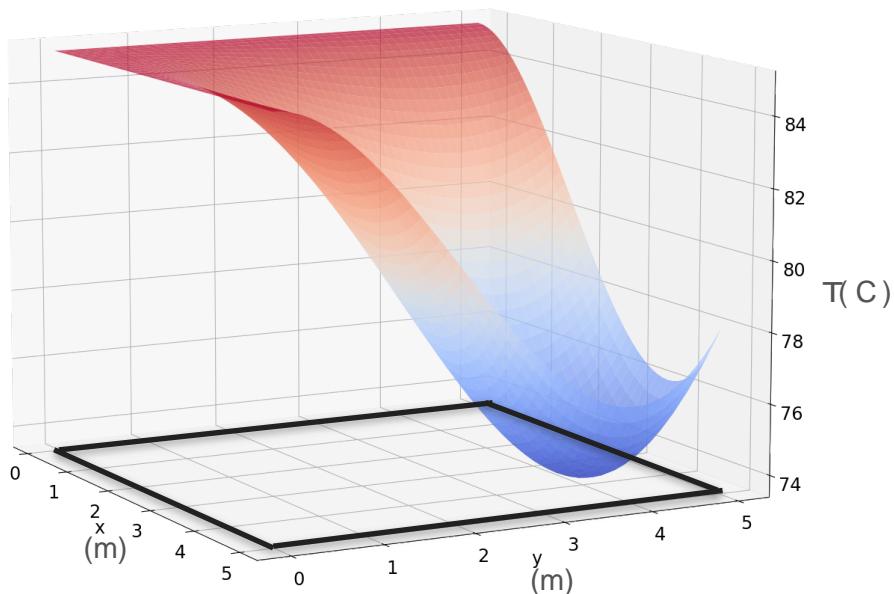


Try and calculate

$$\frac{\partial f}{\partial x}$$

# Exercise

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$



Try and calculate

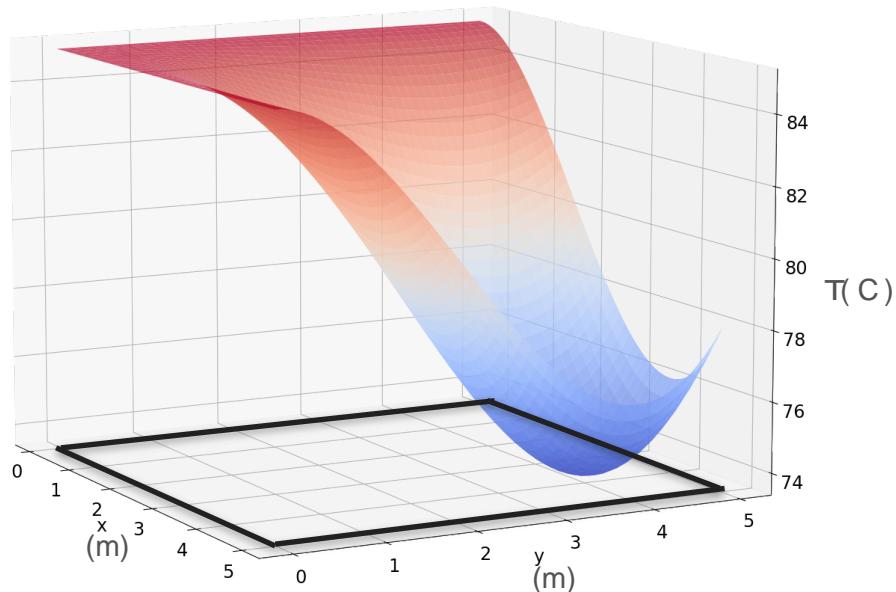
$$\frac{\partial f}{\partial x}$$

and

$$\frac{\partial f}{\partial y}$$

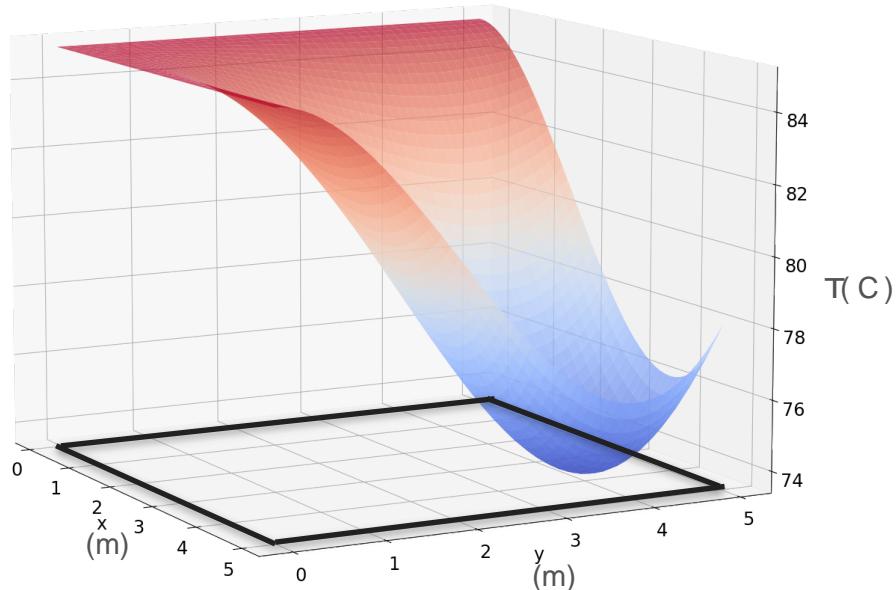
# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$



# Motivation for Optimization in Two Variables

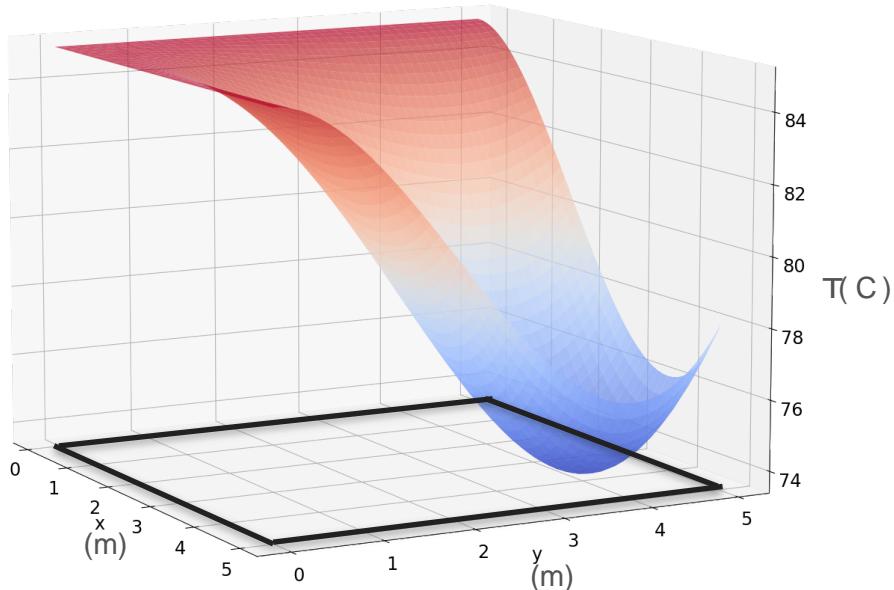
$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$
$$\frac{\partial f}{\partial x} = \frac{1}{90}x(3x-12)y^2(y-6)$$



# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$

$$\frac{\partial f}{\partial x} = -\frac{1}{90}x(3x - 12)y^2(y - 6)$$

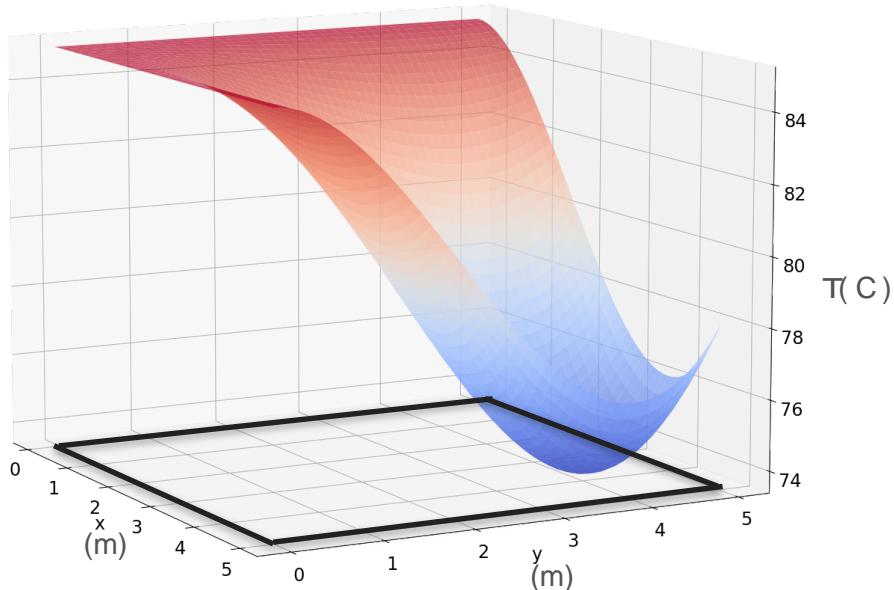


$$\frac{\partial f}{\partial y} = -\frac{1}{90}x^2(x - 6)y(3y - 12)$$

# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$

$$\frac{\partial f}{\partial x} = \frac{1}{90}x(3x - 12)y^2(y - 6) = 0$$



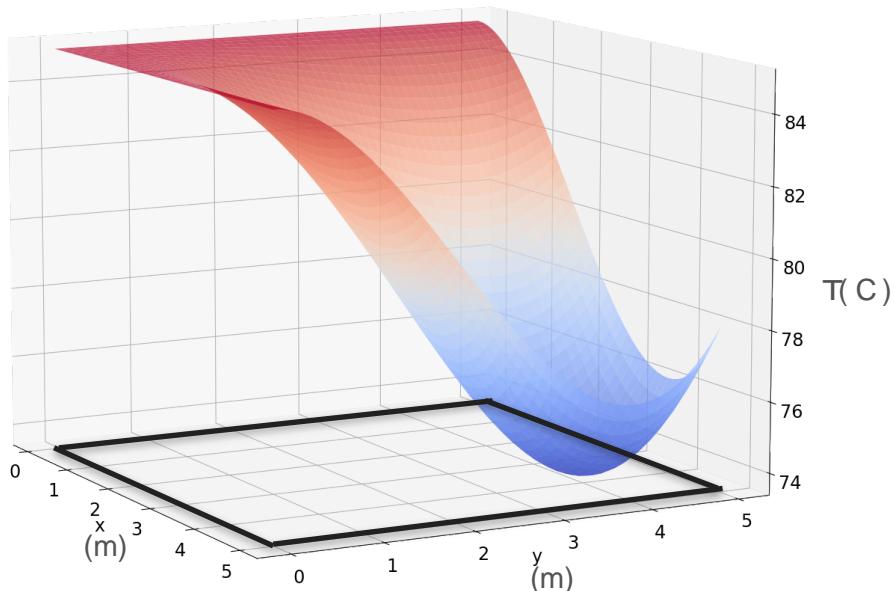
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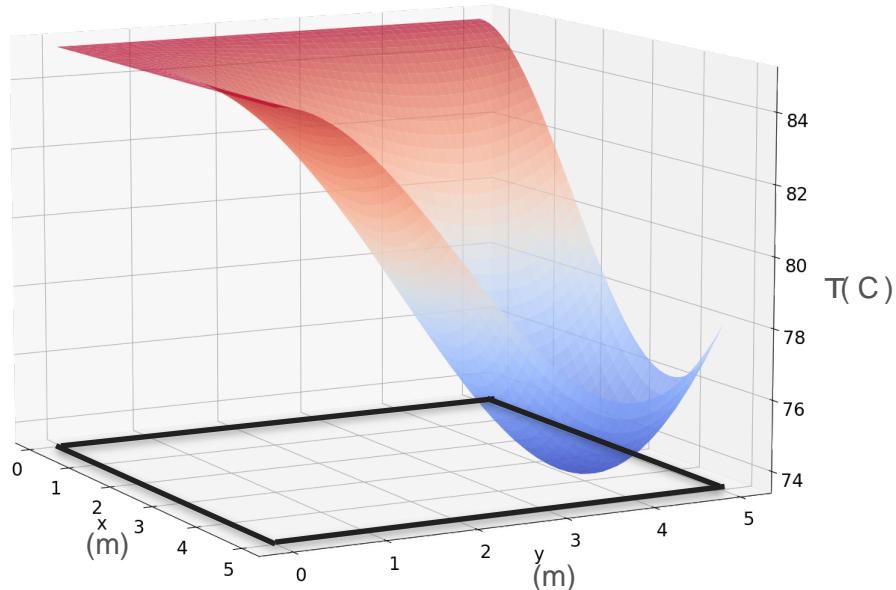
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# Motivation for Optimization in Two Variables

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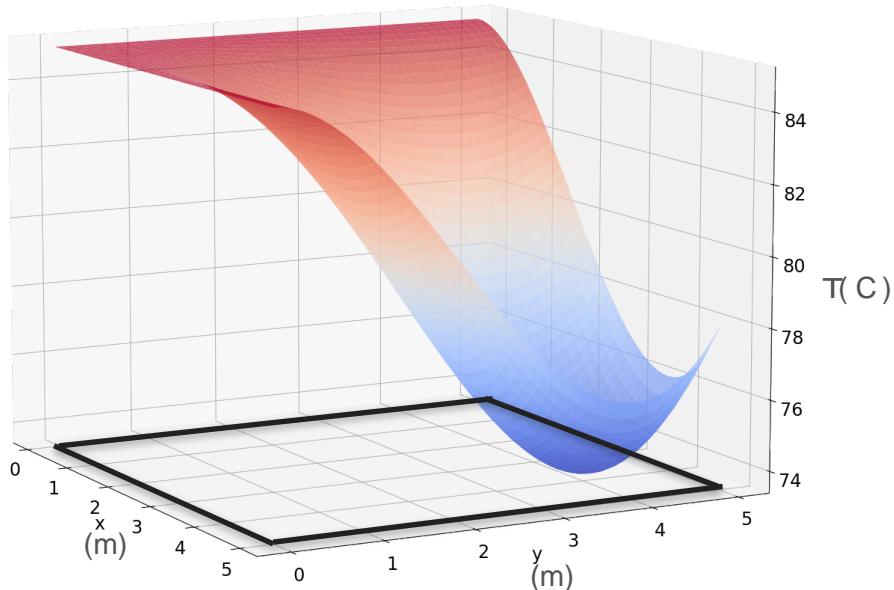


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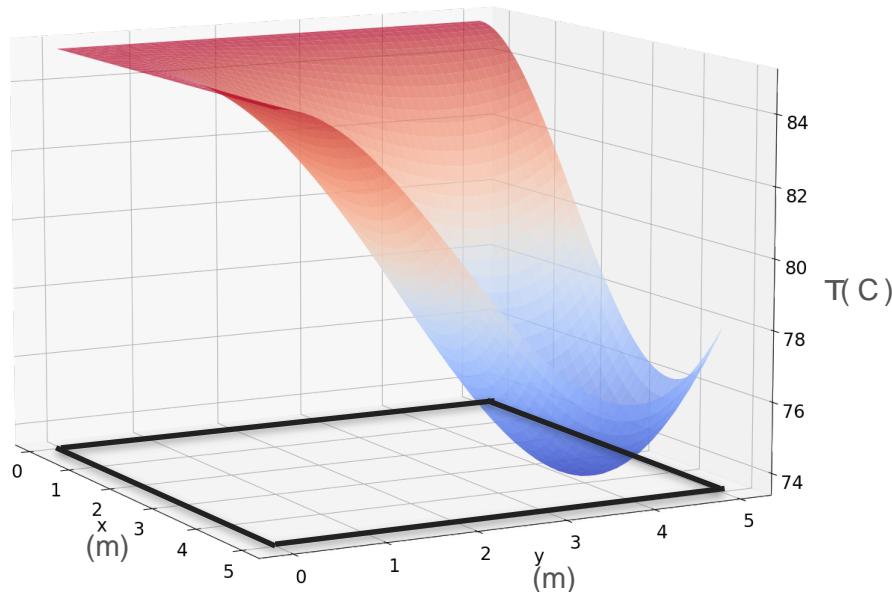


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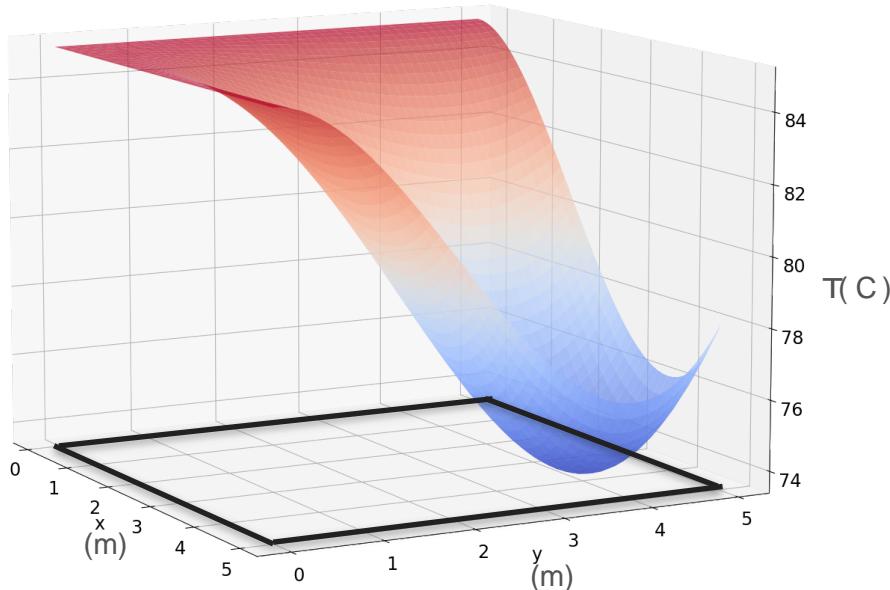
$$\frac{\partial f}{\partial x} = \frac{1}{90}x(3x-12)y^2(y-6) = 0$$

$x = 0 \quad x = 4 \quad y = 0 \quad y = 6$

$$\frac{\partial f}{\partial y} = \frac{1}{90}x^2(x-6)y(3y-12) = 0$$

# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$



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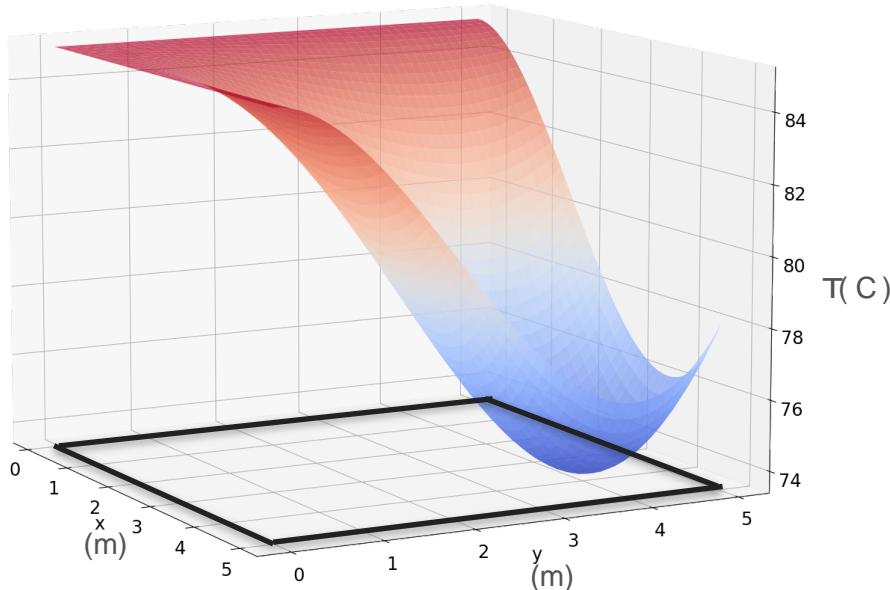
$x = 0$     $x = 4$     $y = 0$     $y = 6$

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$x = 0$

# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$



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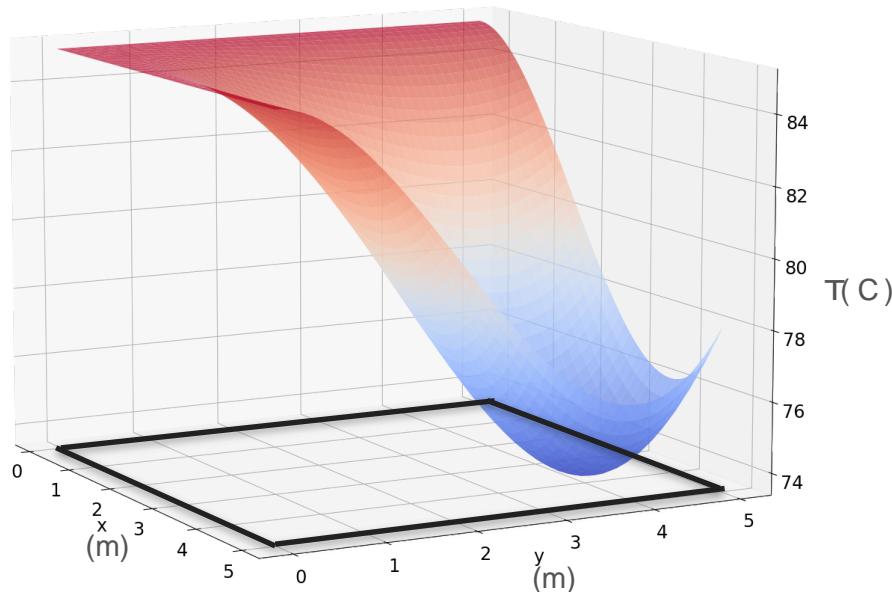
$x = 0$     $x = 4$     $y = 0$     $y = 6$

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# Motivation for Optimization in Two Variables

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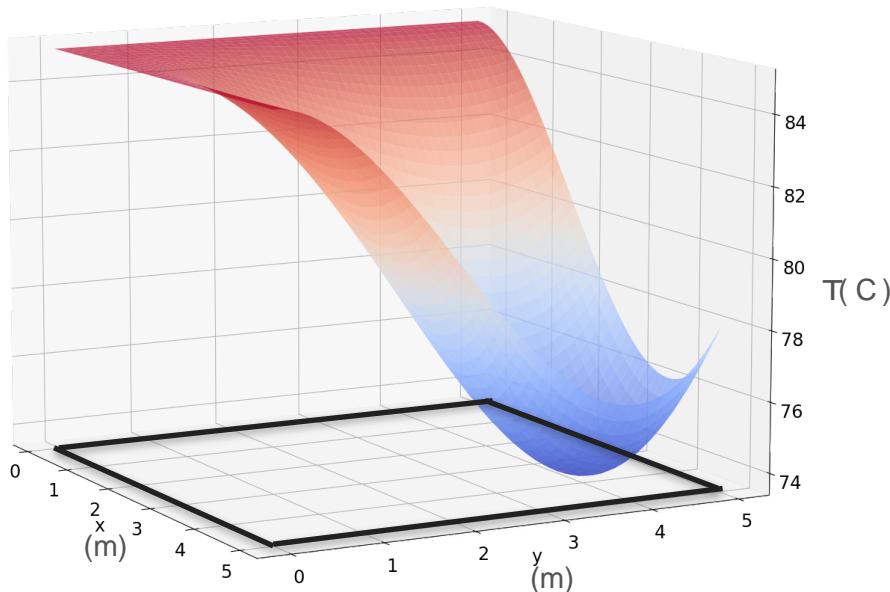


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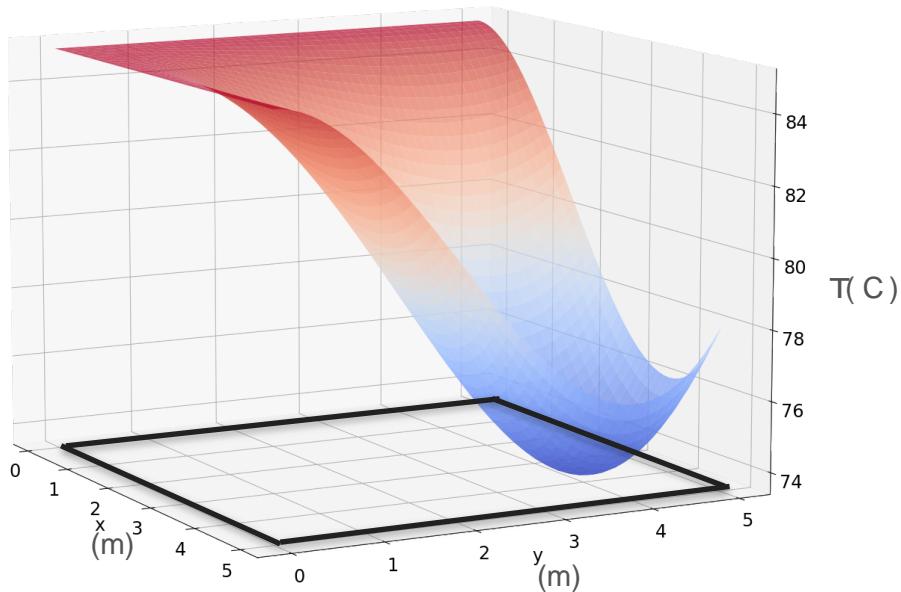


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# Motivation for Optimization in Two Variables

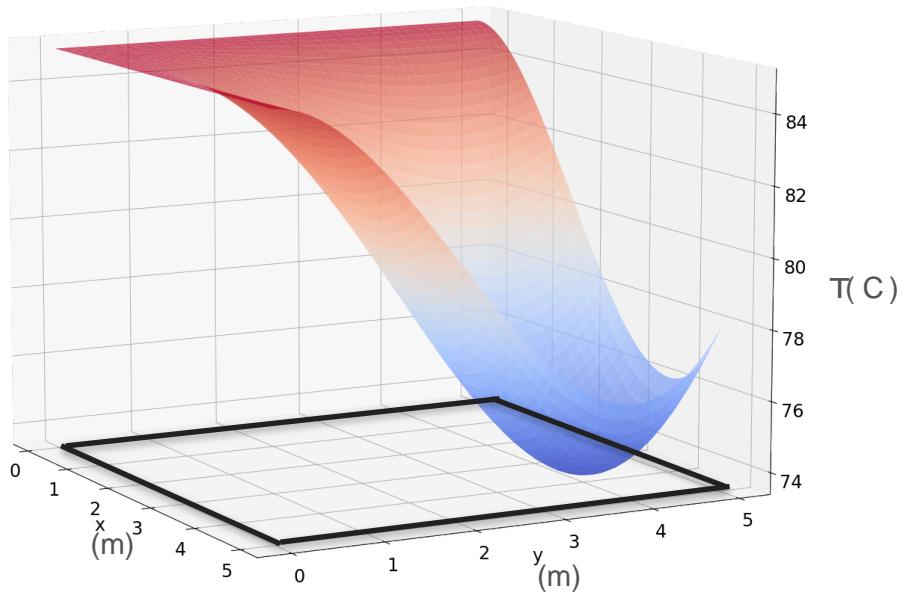
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# Motivation for Optimization in Two Variables

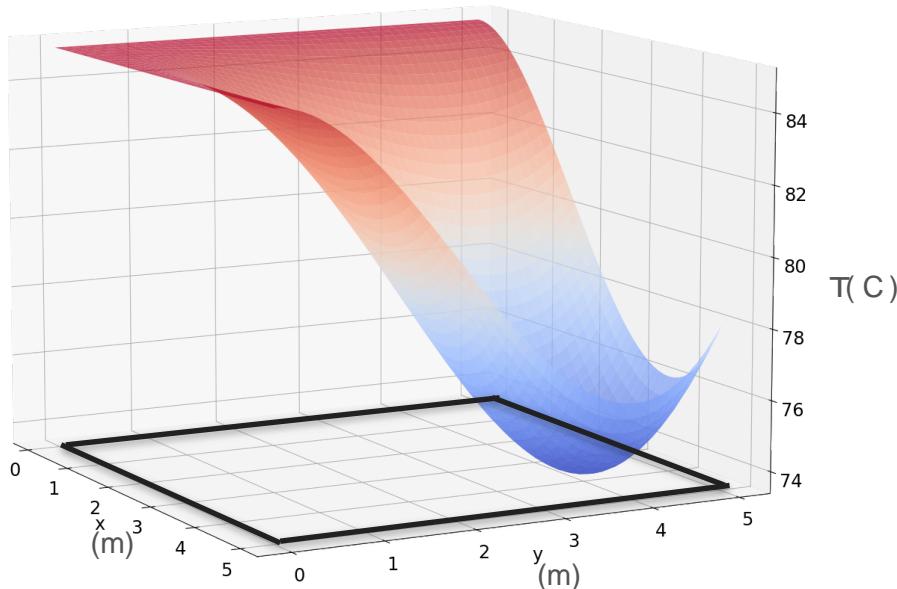
$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$

Candidate points for the minima



# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$



Candidate points for the minima

$$x = 0$$

$$y = 0$$

$$x = 0, y = 0$$

$$x = 0, y = 4$$

$$x = 0, y = 6$$

$$x = 4, y = 0$$

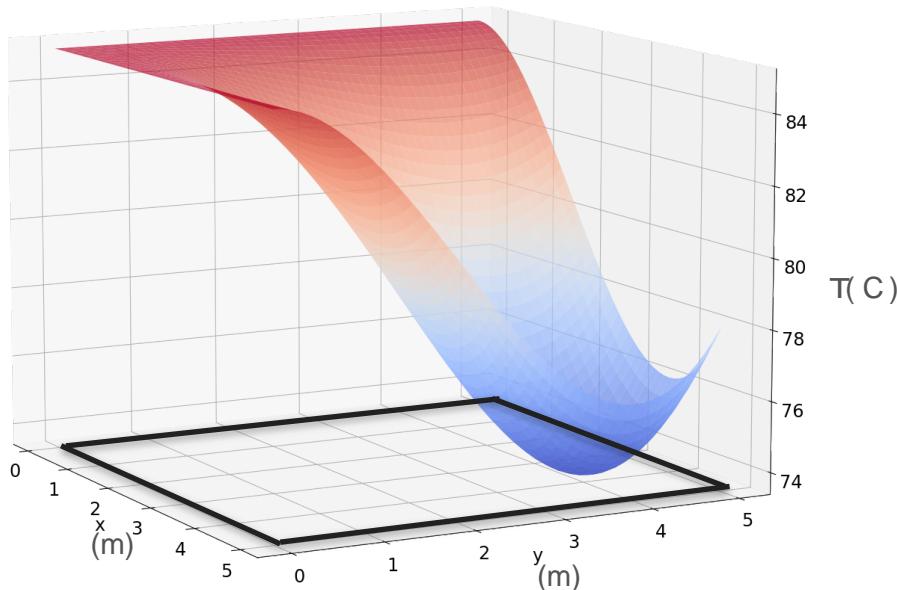
$$x = 4, y = 4$$

$$x = 6, y = 0$$

$$x = 6, y = 6$$

# Motivation for Optimization in Two Variables

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x - 6)y^2(y - 6)$$



Candidate points for the minima

$$x = 0$$

$$y = 0$$

$$x = 0, y = 0$$

$$x = 0, y = 4$$

$$x = 0, y = 6$$

Outside

$$x = 4, y = 0$$

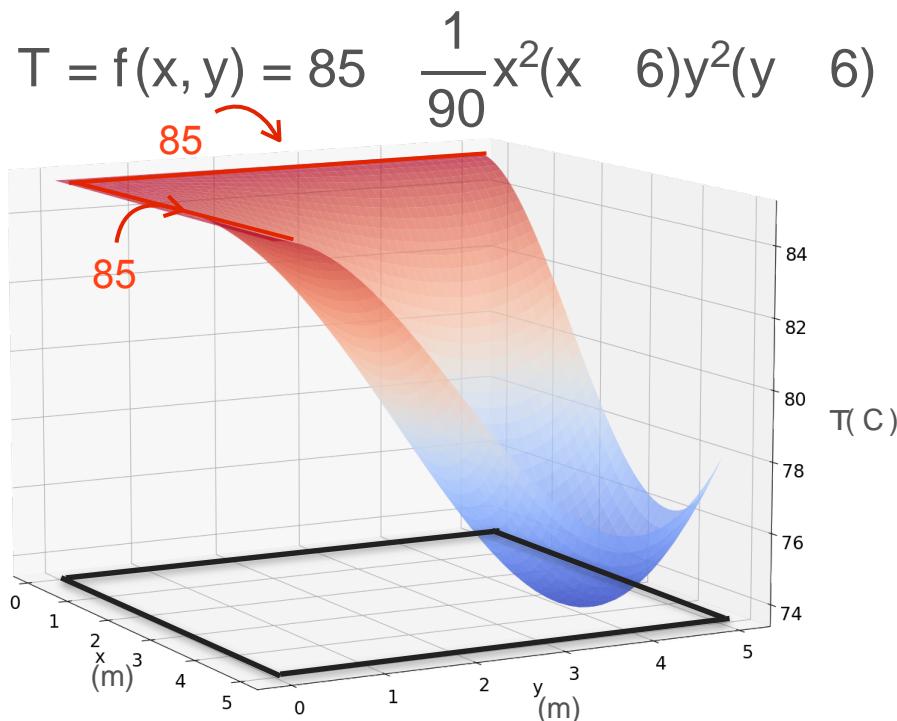
$$x = 4, y = 4$$

$$x = 6, y = 0$$

Outside

$$x = 6, y = 6$$

# Motivation for Optimization in Two Variables



Candidate points for the minima

$x = 0$   
 $y = 0$

Maxima

$x = 0, y = 0$

$x = 0, y = 4$

$x = 0, y = 6$

Outside

$x = 4, y = 0$

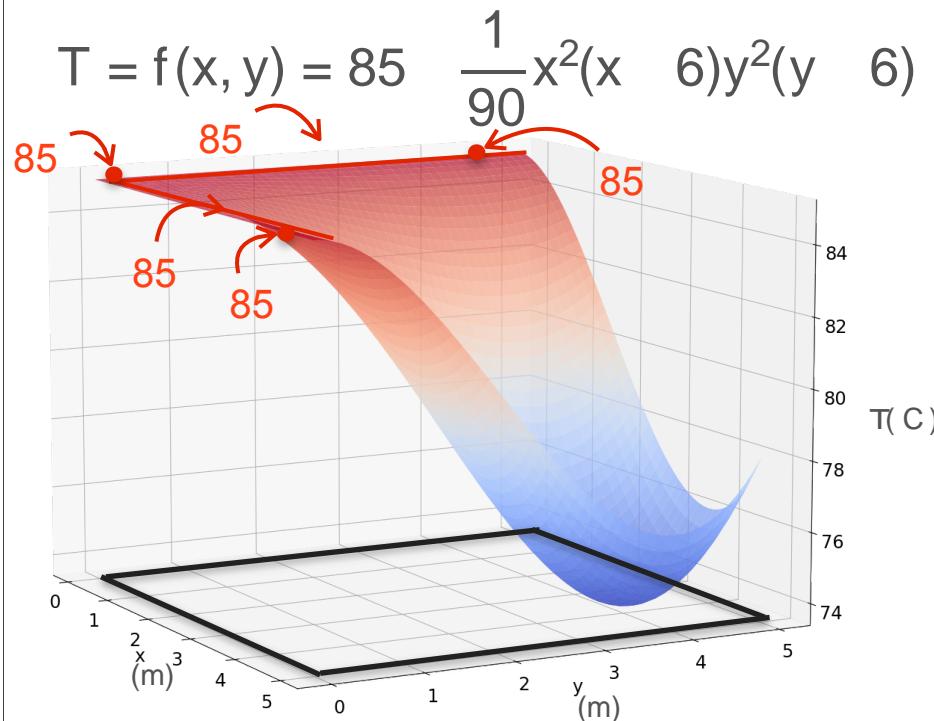
$x = 4, y = 4$

$x = 6, y = 0$

Outside

$x = 6, y = 6$

# Motivation for Optimization in Two Variables



Candidate points for the minima

$x = 0$   
 $y = 0$

Maxima

$x = 0, y = 0$   
 $x = 0, y = 4$

Maxima

$x = 0, y = 6$

Outside

$x = 4, y = 0$

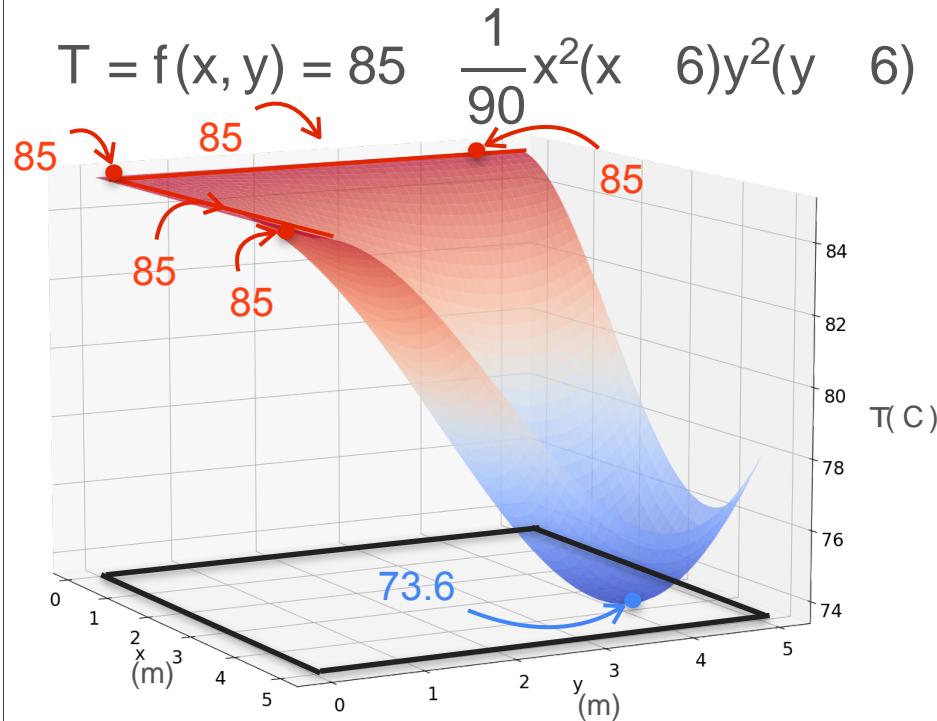
Maxima

$x = 4, y = 4$

$x = 6, y = 0$   
 $x = 6, y = 6$

Outside

# Motivation for Optimization in Two Variables



Candidate points for the minima

x = 0  
y = 0

Maxima

x = 0, y = 0  
x = 0, y = 4

Maxima

x = 0, y = 6

Outside

x = 4, y = 0

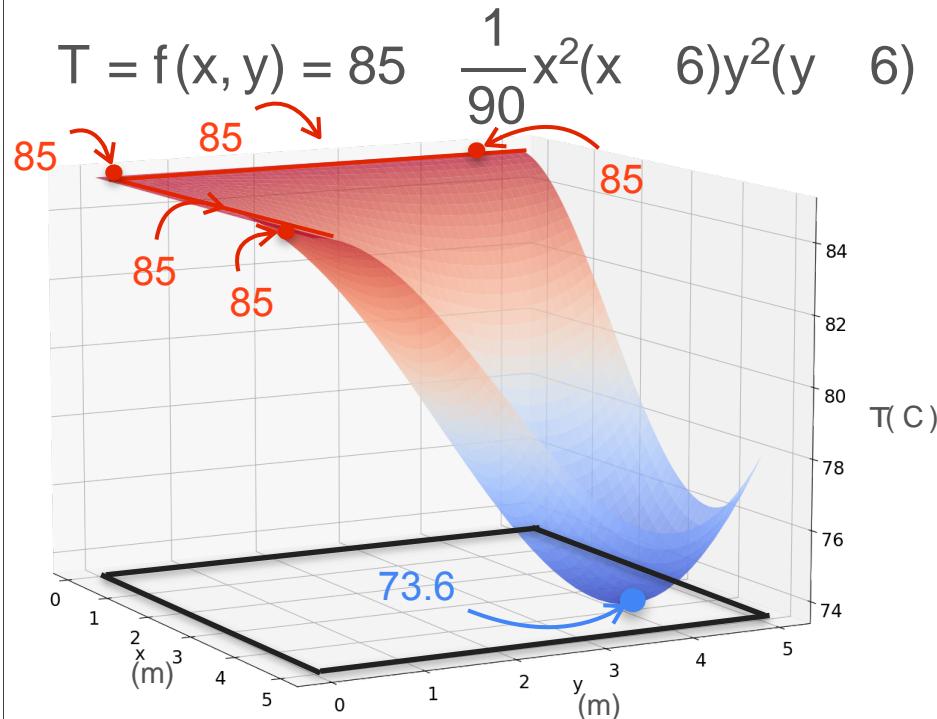
Maxima

x = 4, y = 4

Outside

x = 6, y = 0  
x = 6, y = 6

# Motivation for Optimization in Two Variables



Candidate points for the minima

x = 0  
y = 0

Maxima

x = 0, y = 0  
x = 0, y = 4

Maxima

x = 0, y = 6

Outside

x = 4, y = 0

Maxima

x = 4, y = 4

Minimum

x = 6, y = 0  
x = 6, y = 6

Outside



DeepLearning.AI

## Gradients and Gradient Descent

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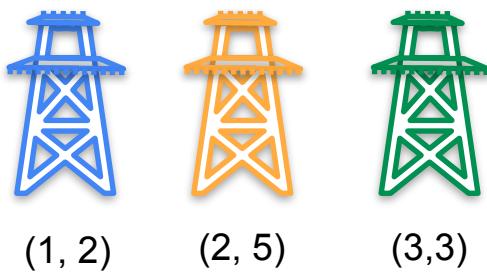
**Optimization using gradients**  
**- Analytical method**

# Linear Regression: Analytical Approach

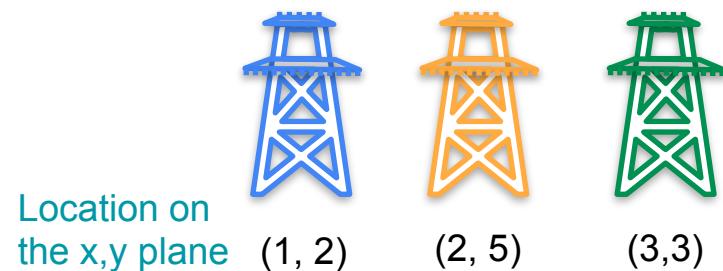
# Linear Regression: Analytical Approach



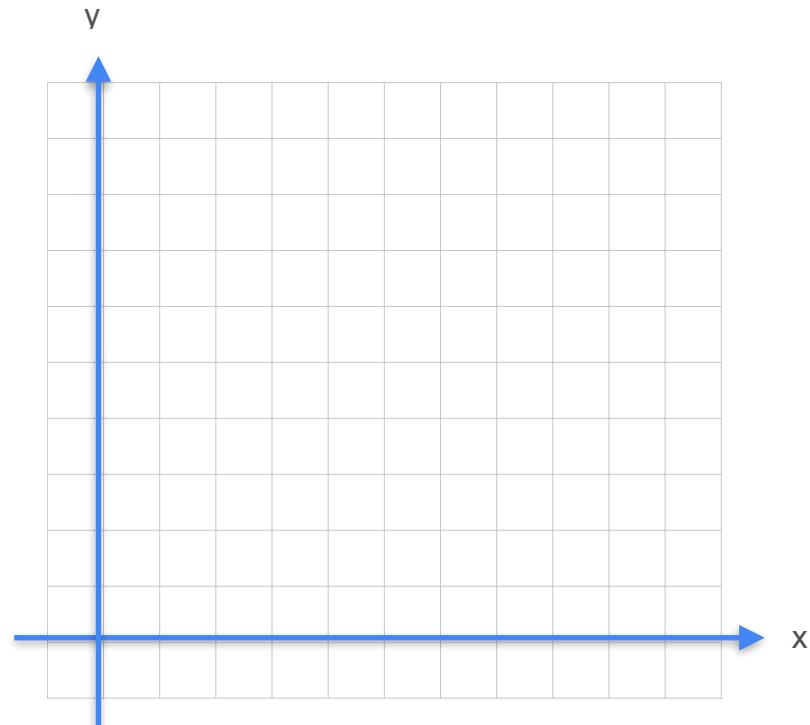
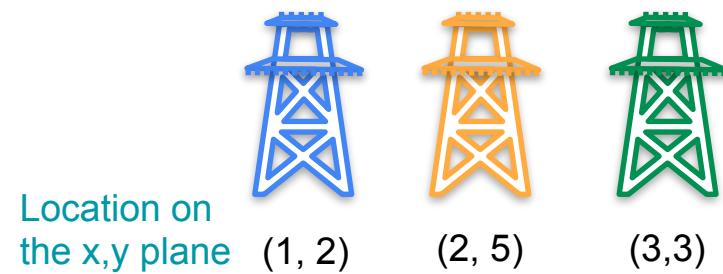
# Linear Regression: Analytical Approach



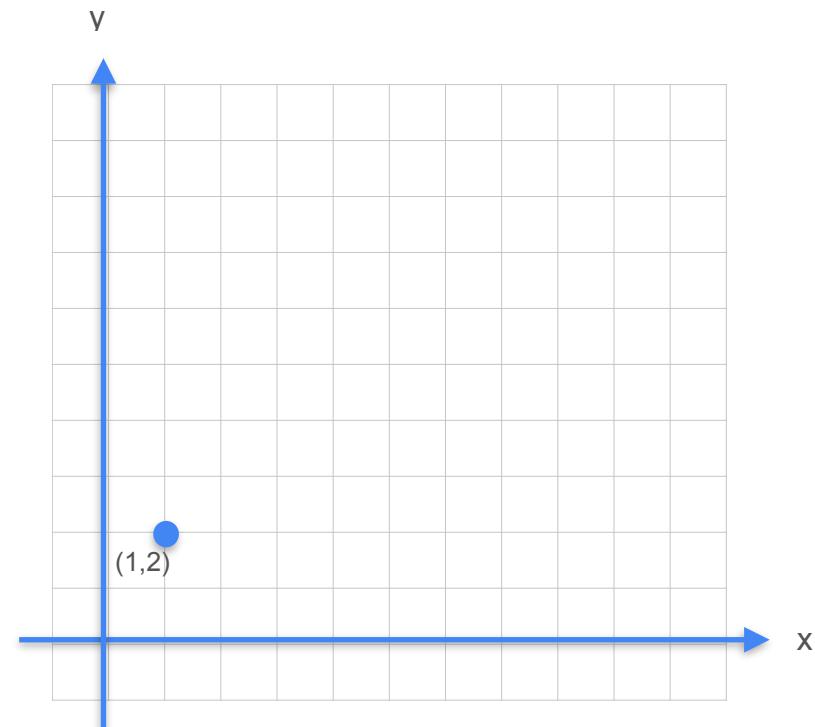
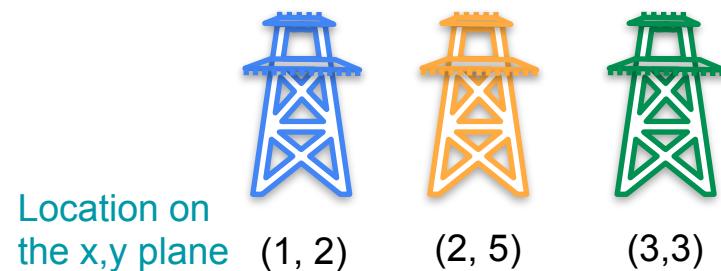
# Linear Regression: Analytical Approach



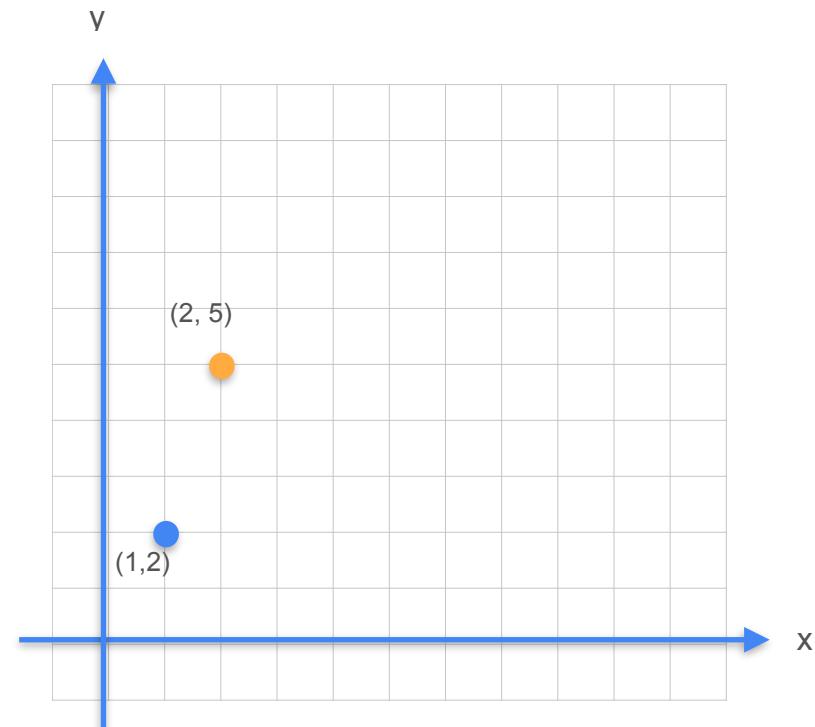
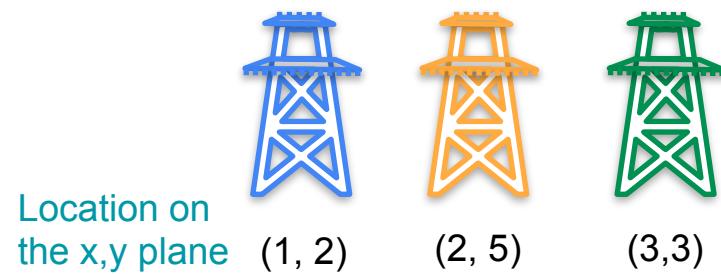
# Linear Regression: Analytical Approach



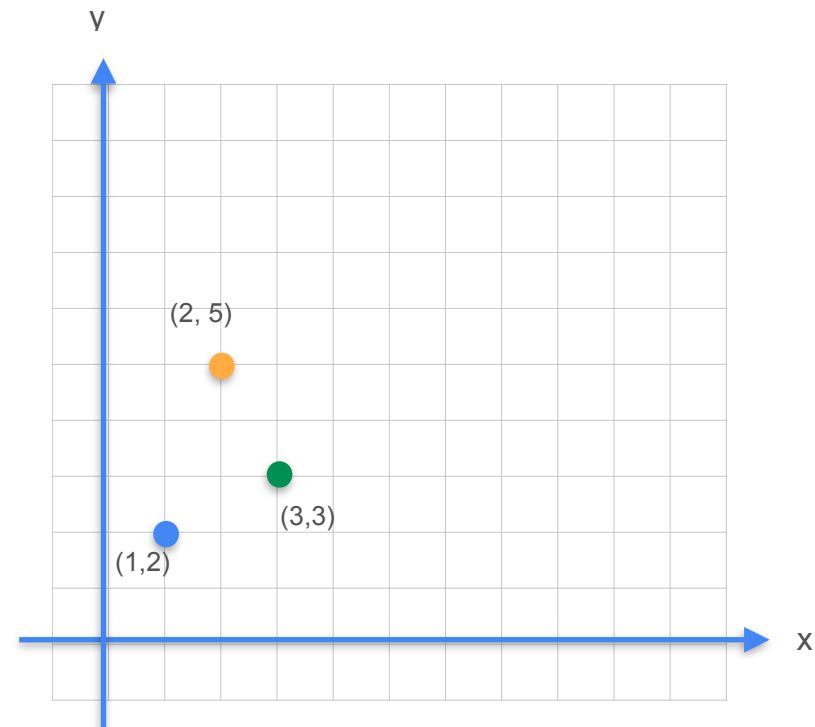
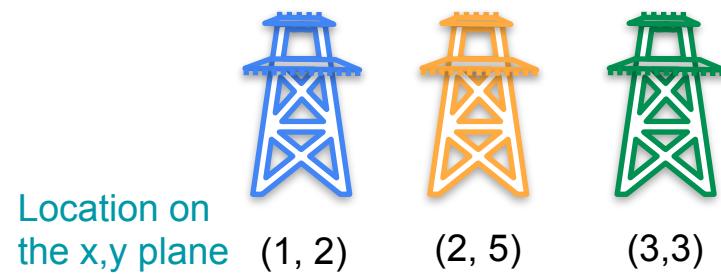
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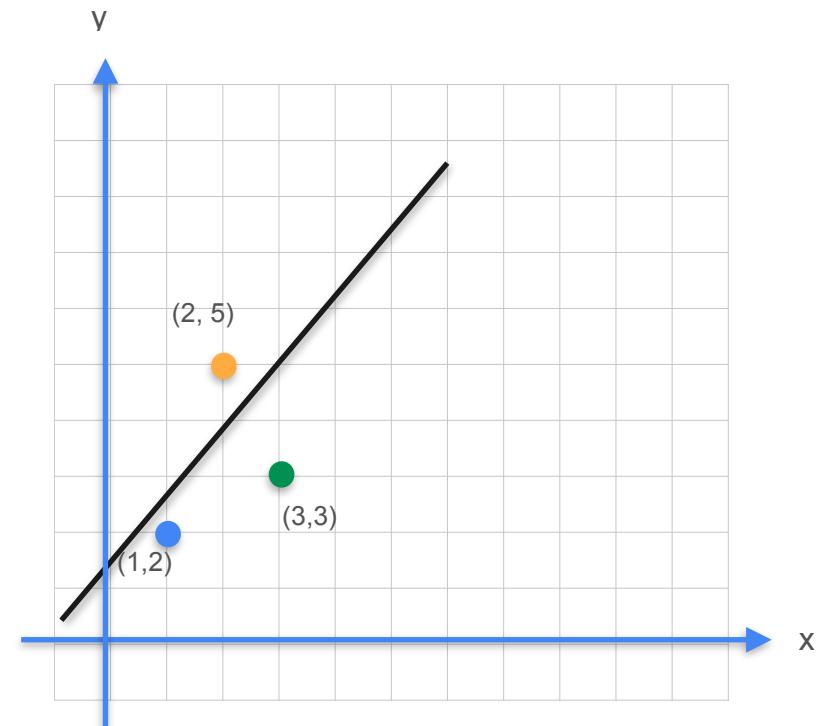
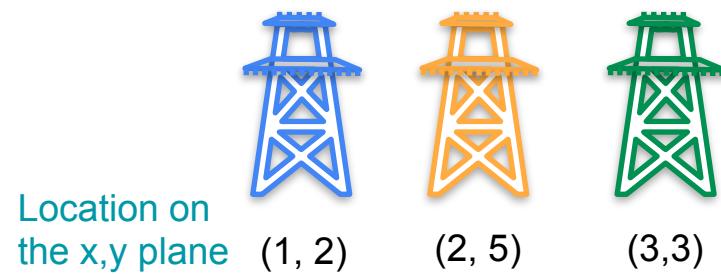
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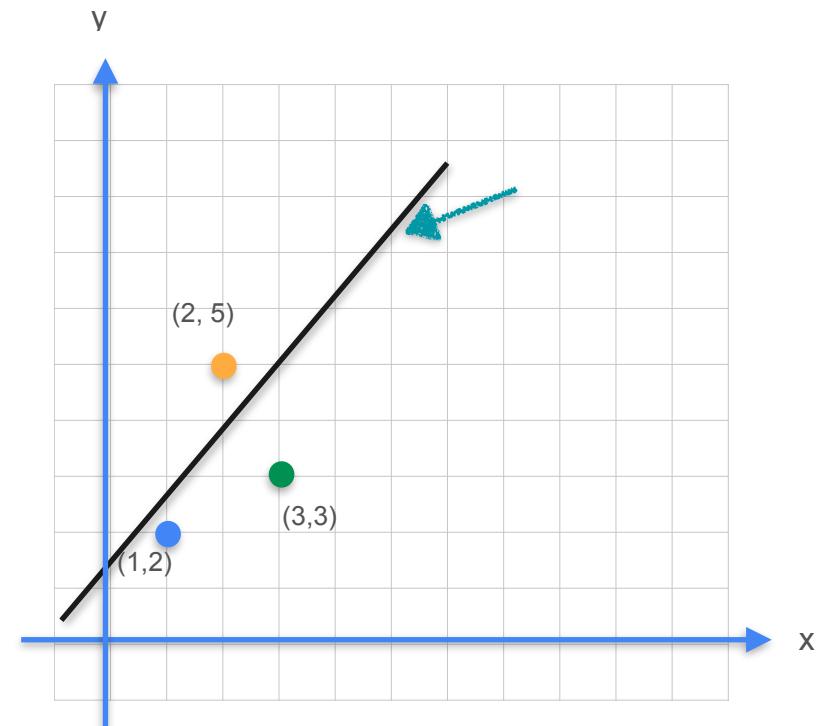
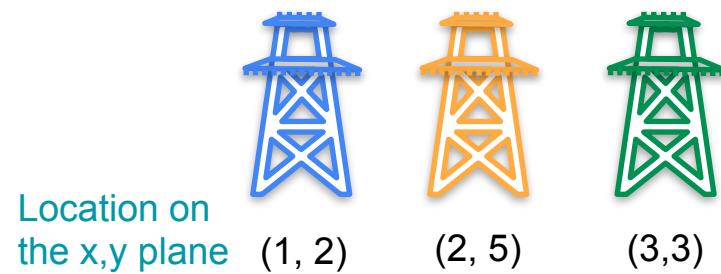
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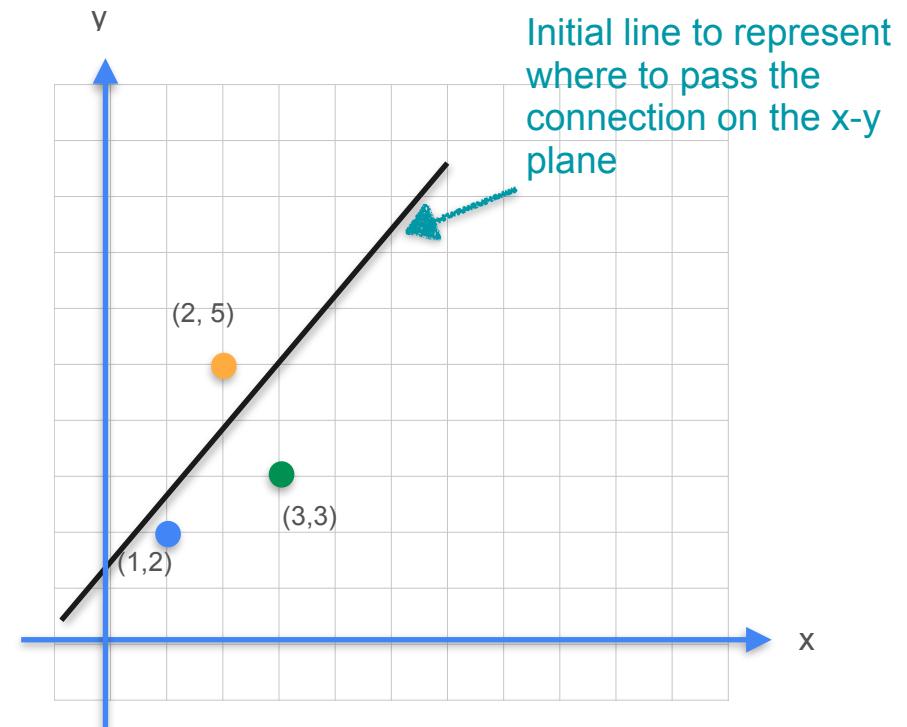
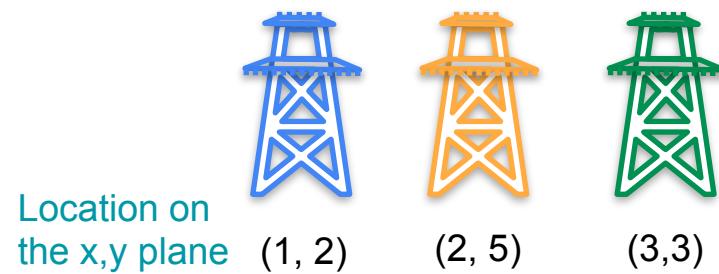
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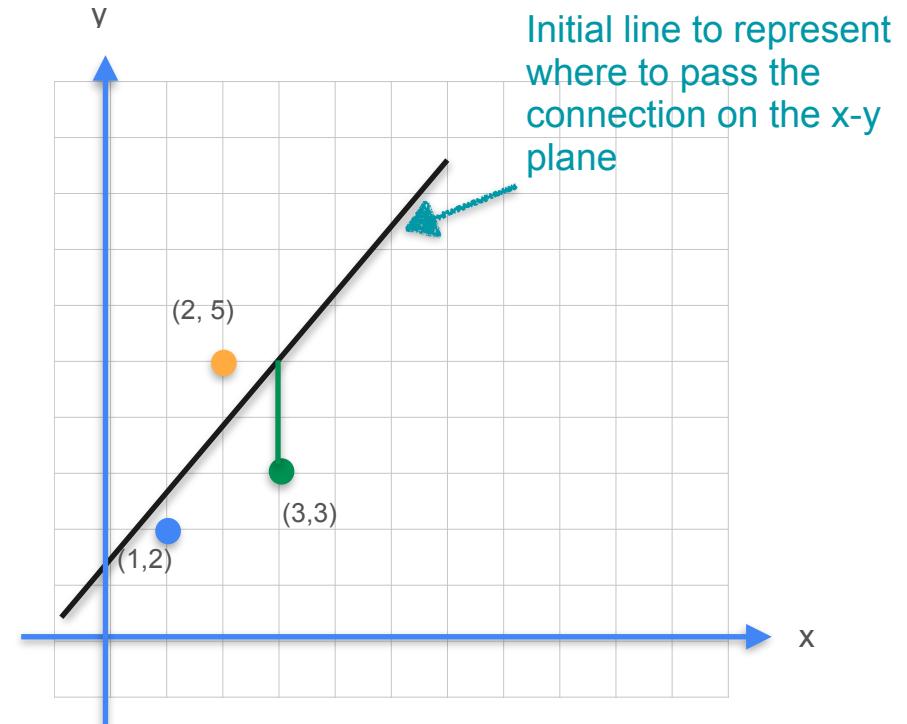
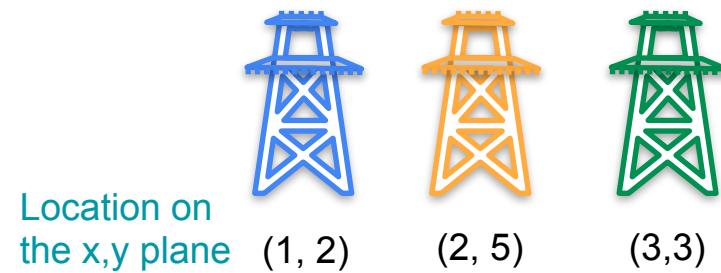
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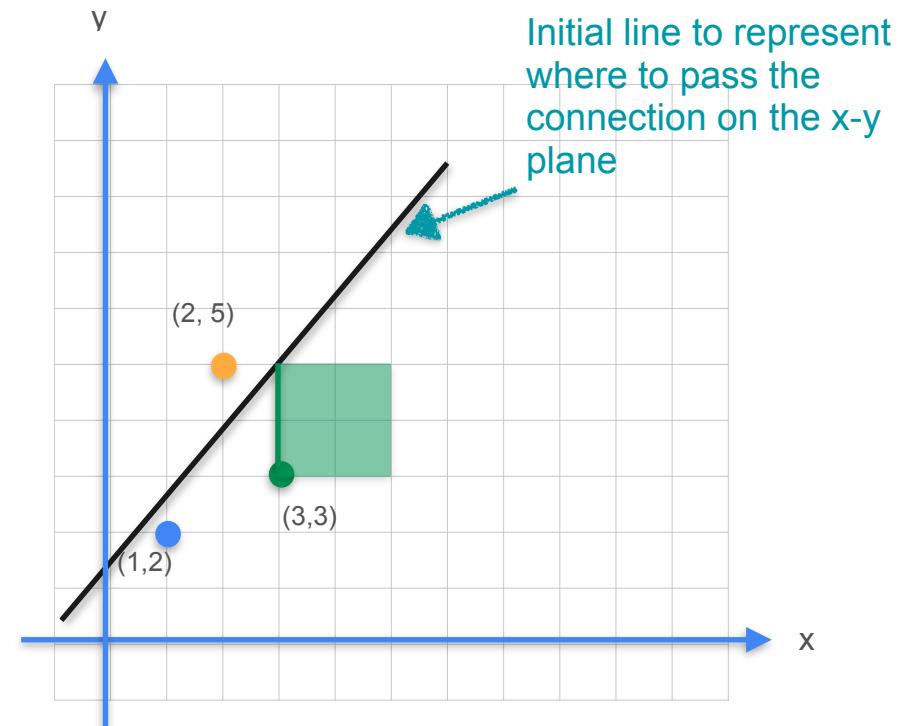
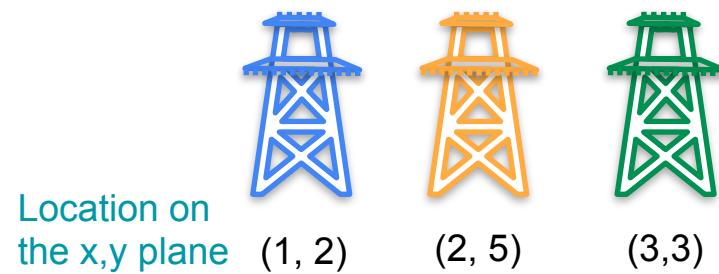
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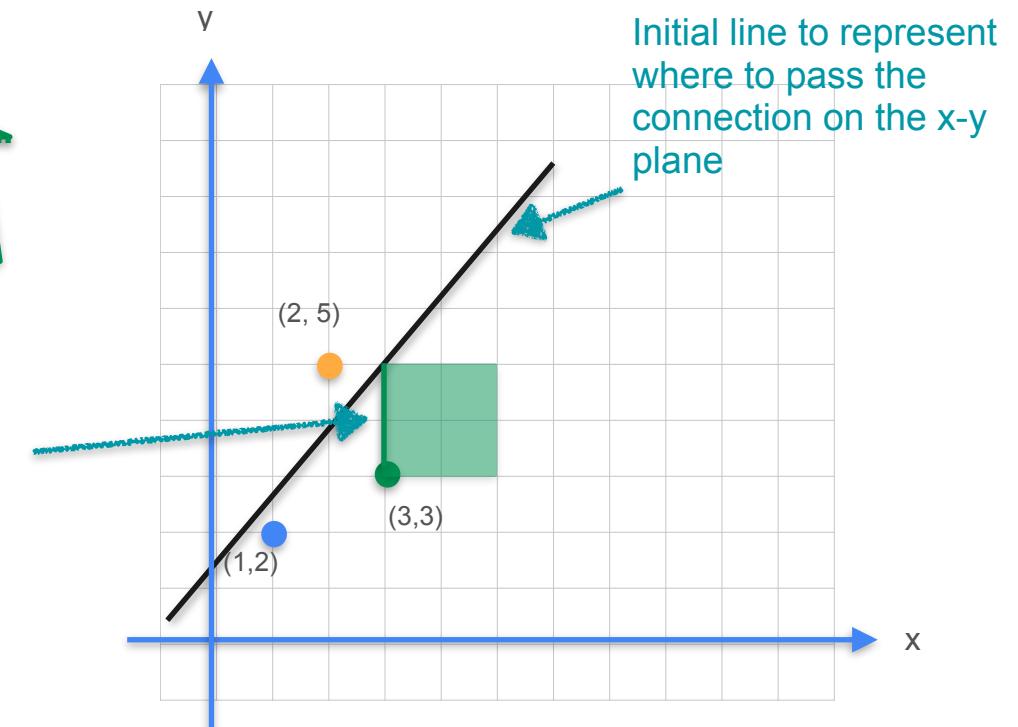
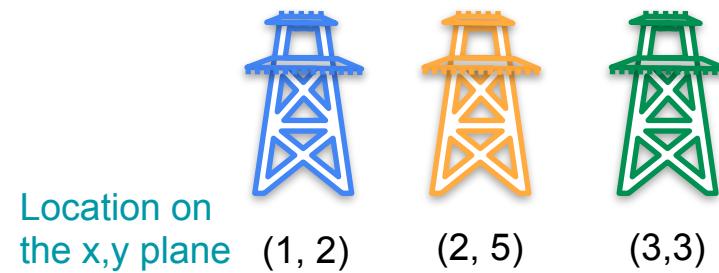
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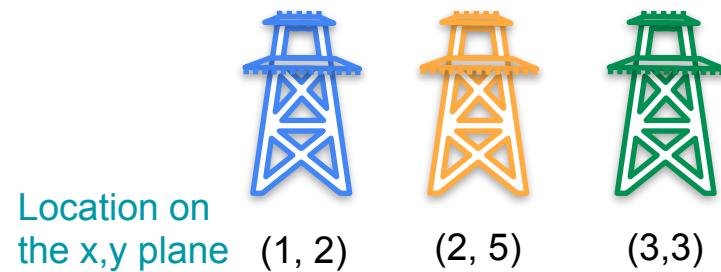
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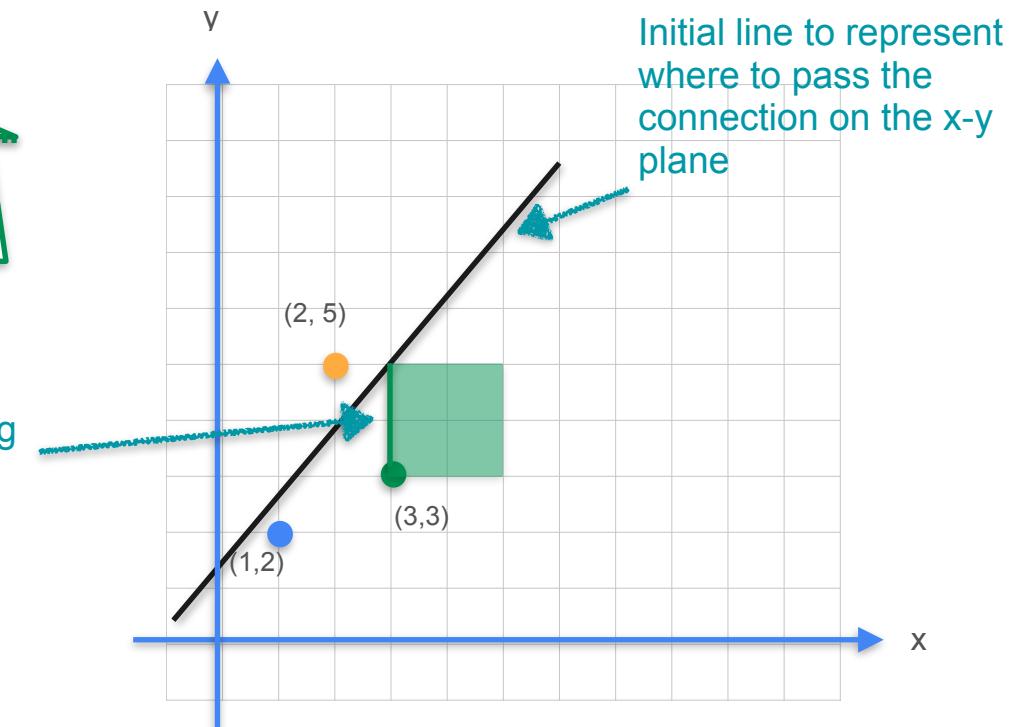
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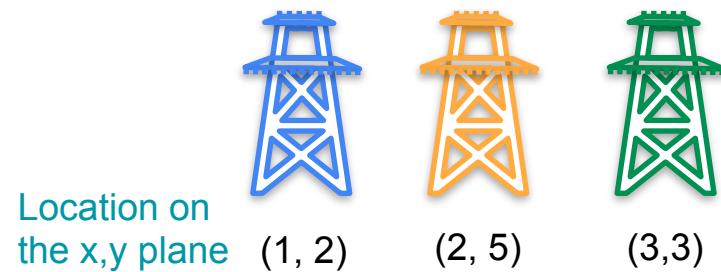
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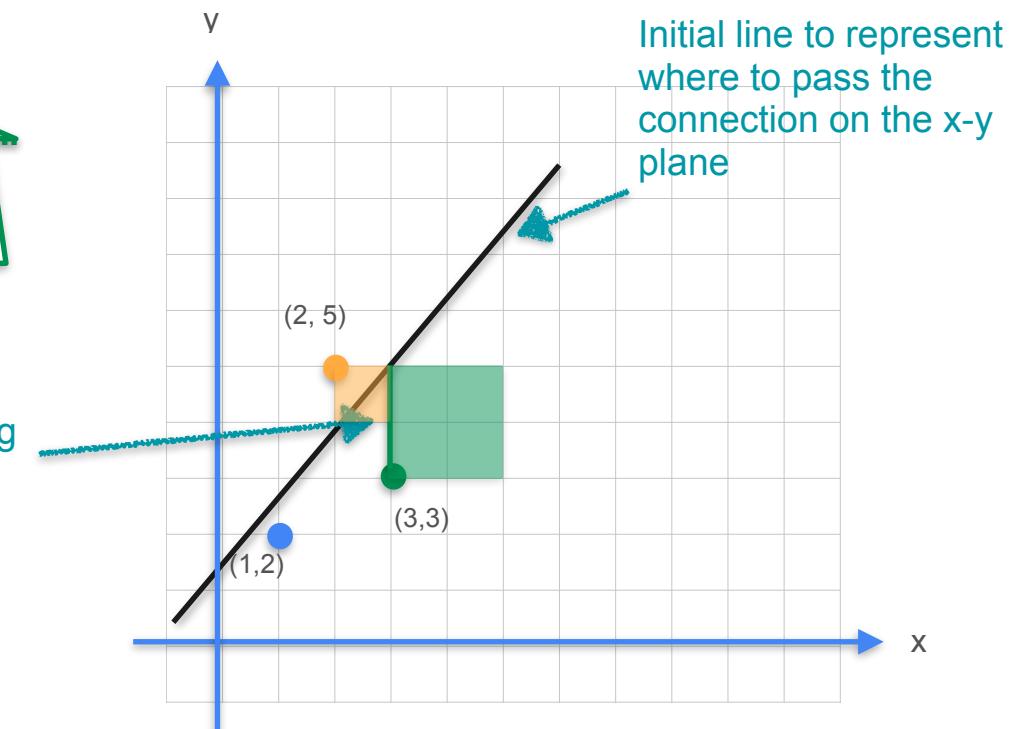
The cost of connecting connection to the powerline



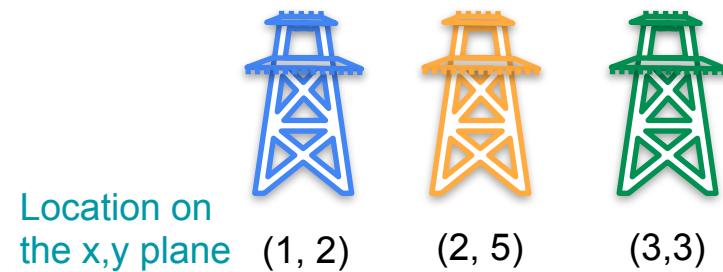
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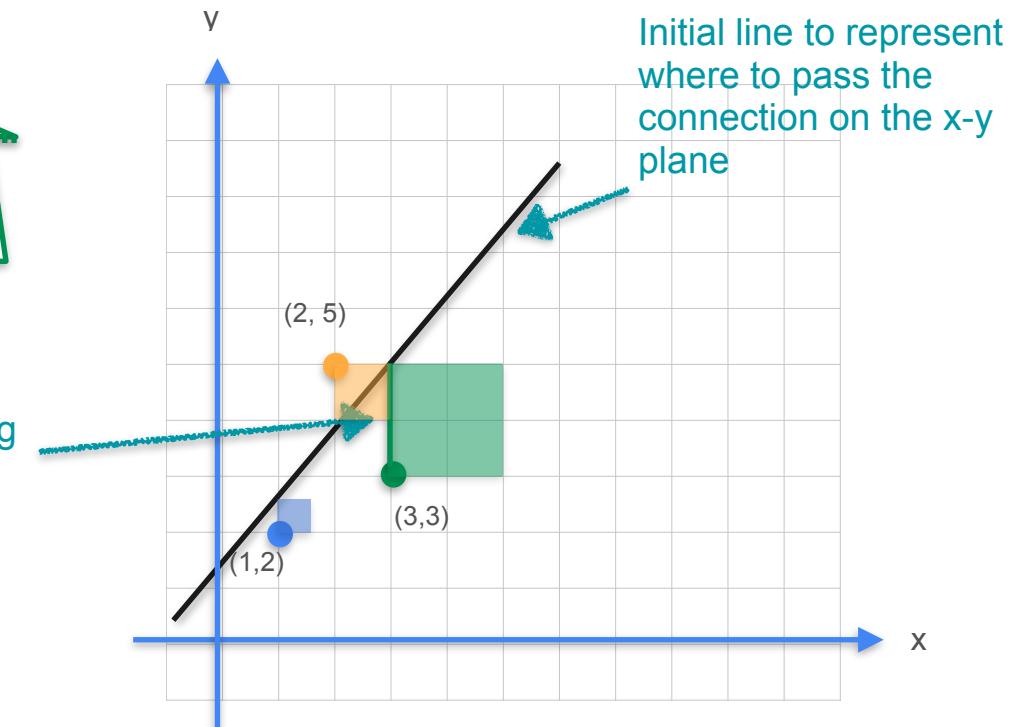
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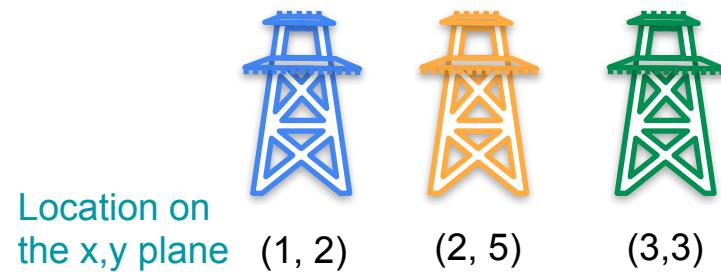
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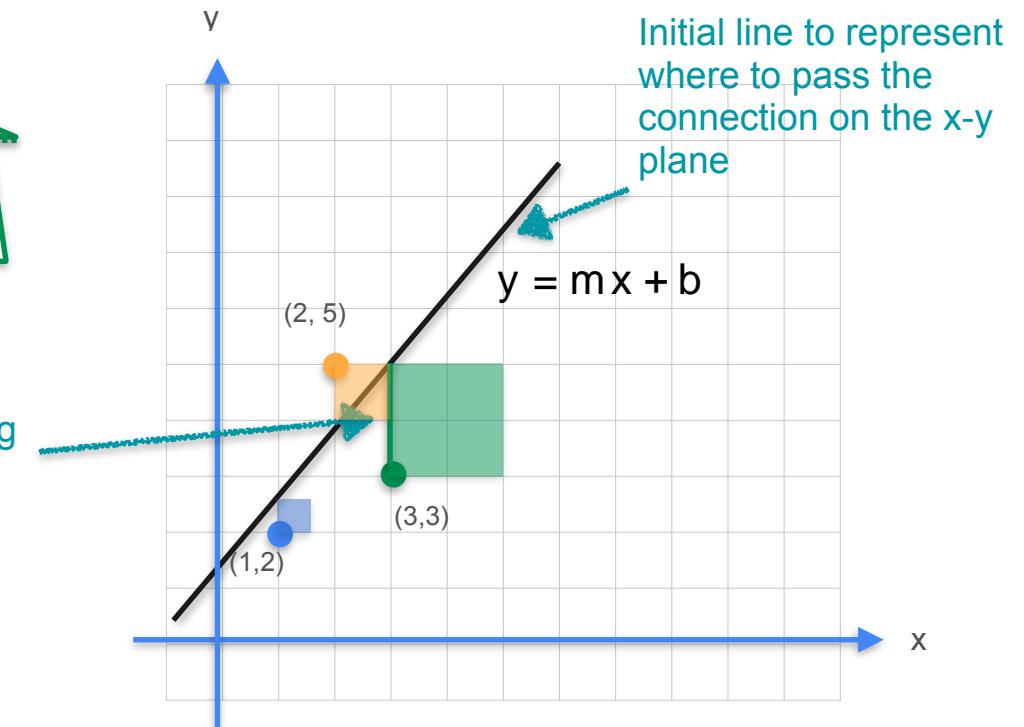
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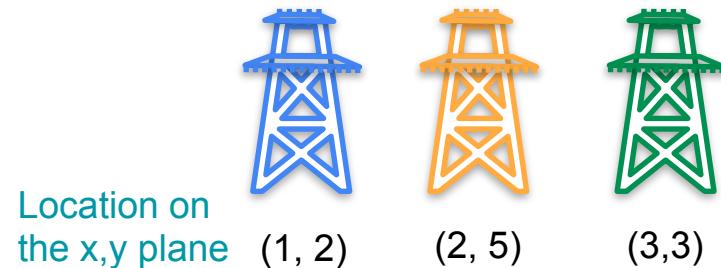
# Linear Regression: Analytical Approach



The cost of connecting connection to the powerline

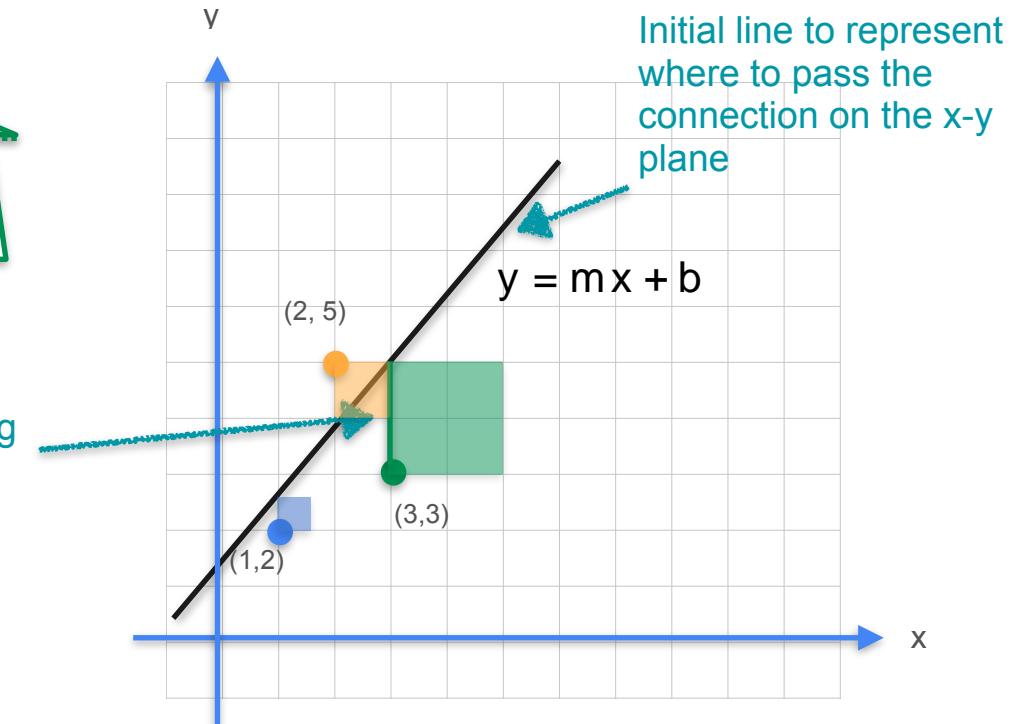


# Linear Regression: Analytical Approach

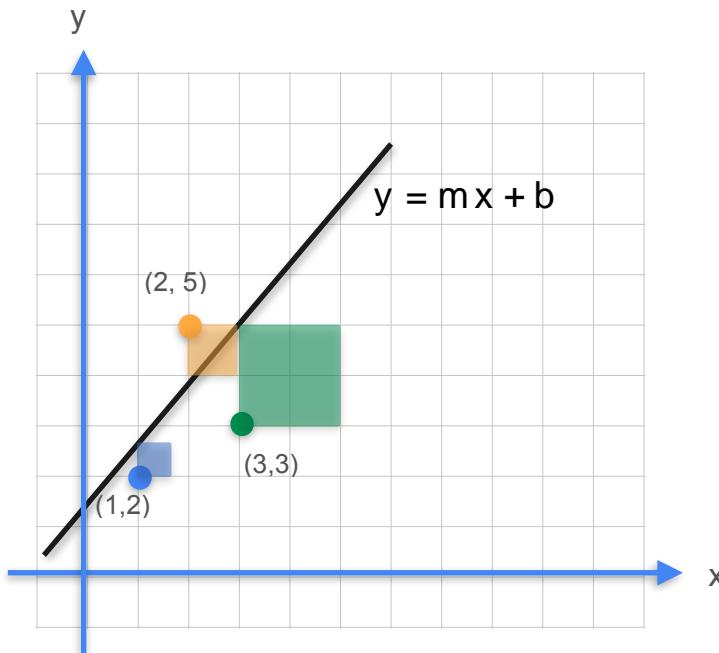


The cost of connecting connection to the powerline

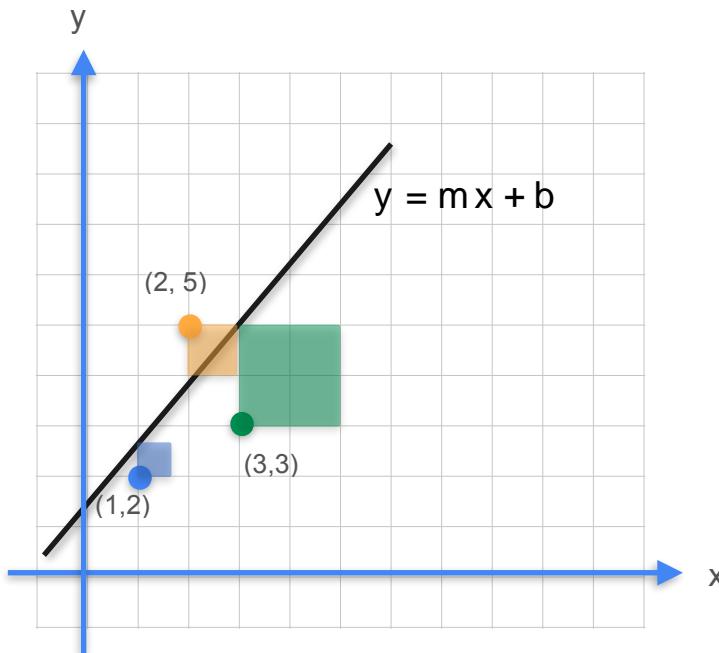
**Goal:** Find  $m, b$  such that you minimize sum of squares cost



# Linear Regression: Analytical Approach

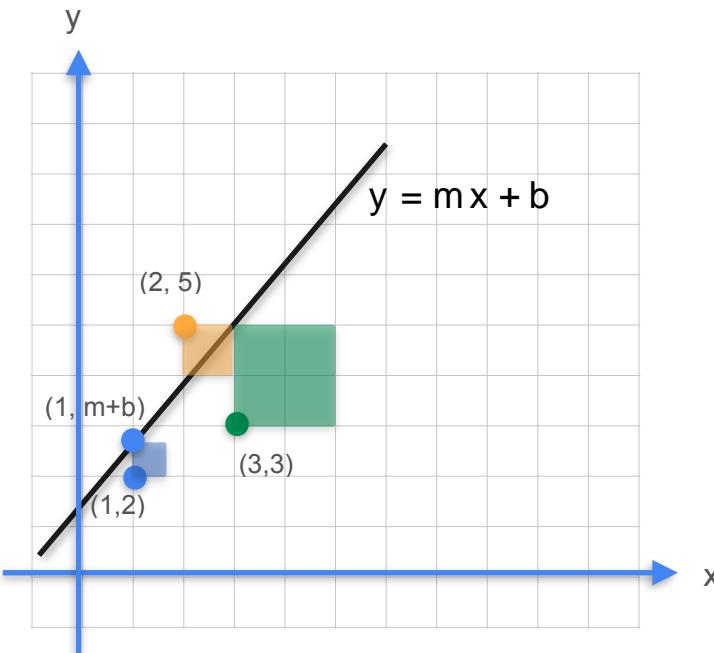


# Linear Regression: Analytical Approach



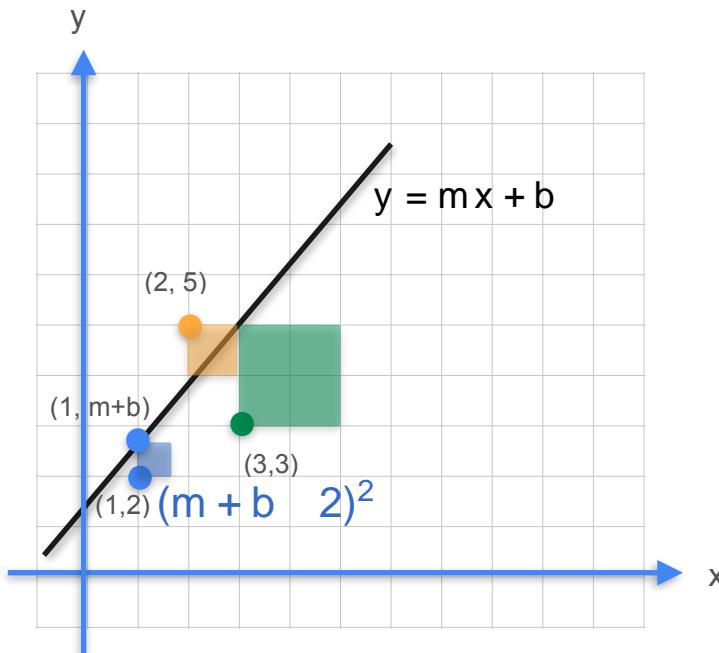
**Goal: Minimize sum of squares cost**

# Linear Regression: Analytical Approach



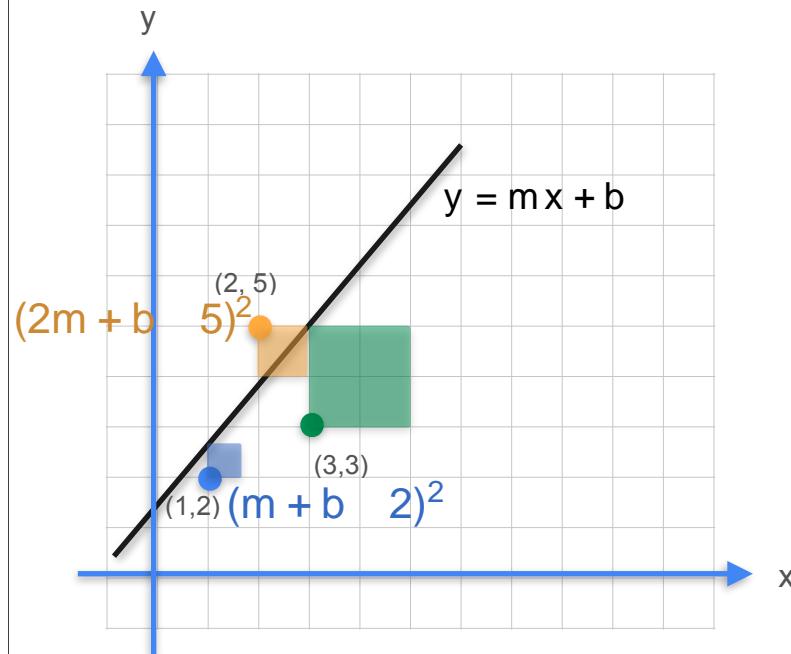
**Goal: Minimize sum of squares cost**

# Linear Regression: Analytical Approach



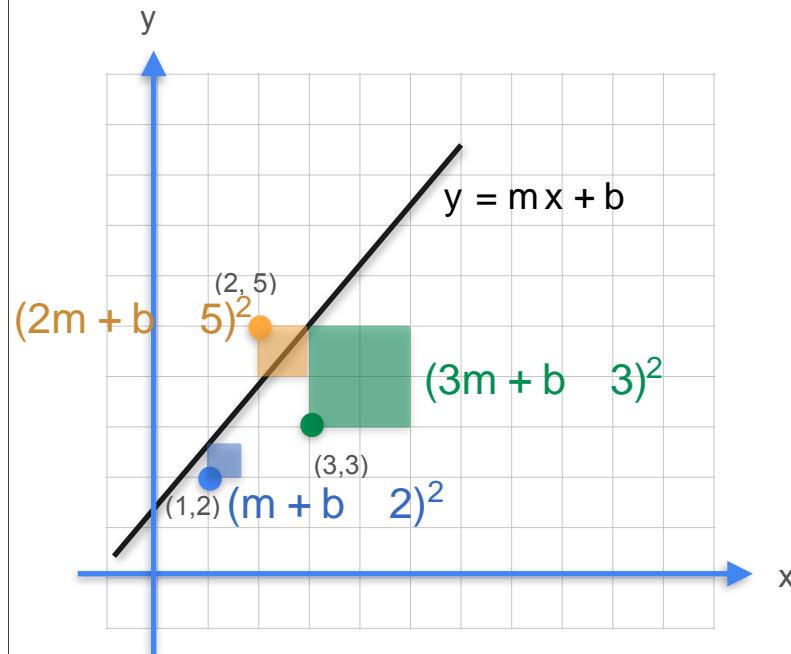
**Goal: Minimize sum of squares cost**

# Linear Regression: Analytical Approach



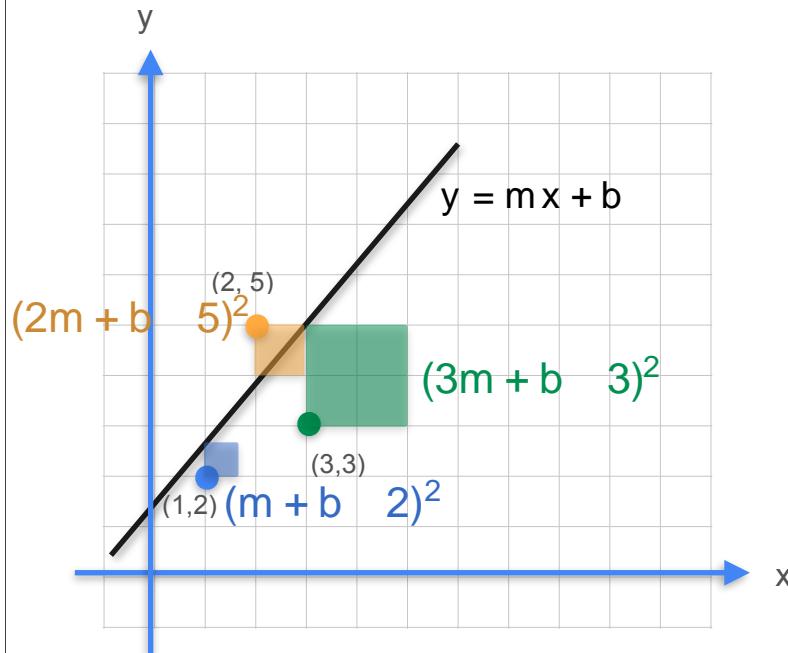
**Goal: Minimize sum of squares cost**

# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

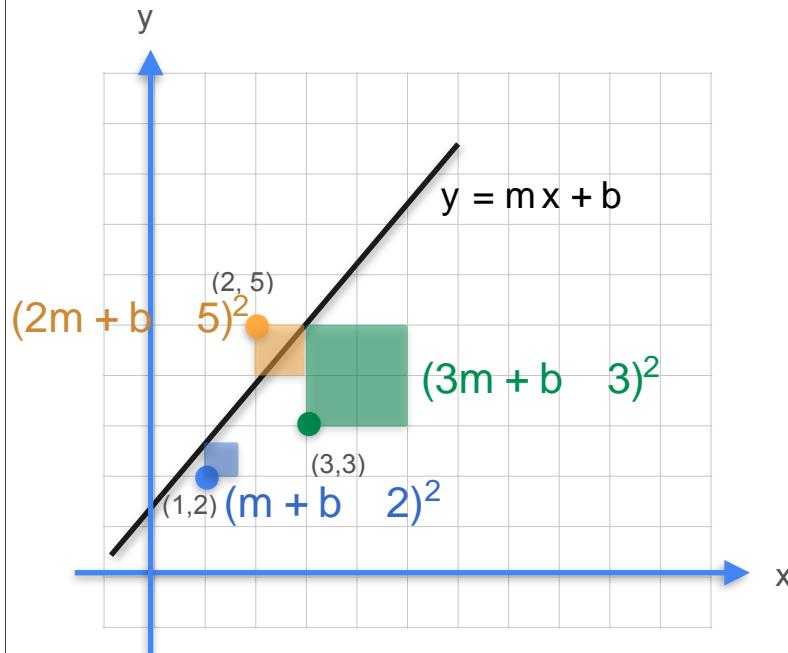
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$(m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2$$

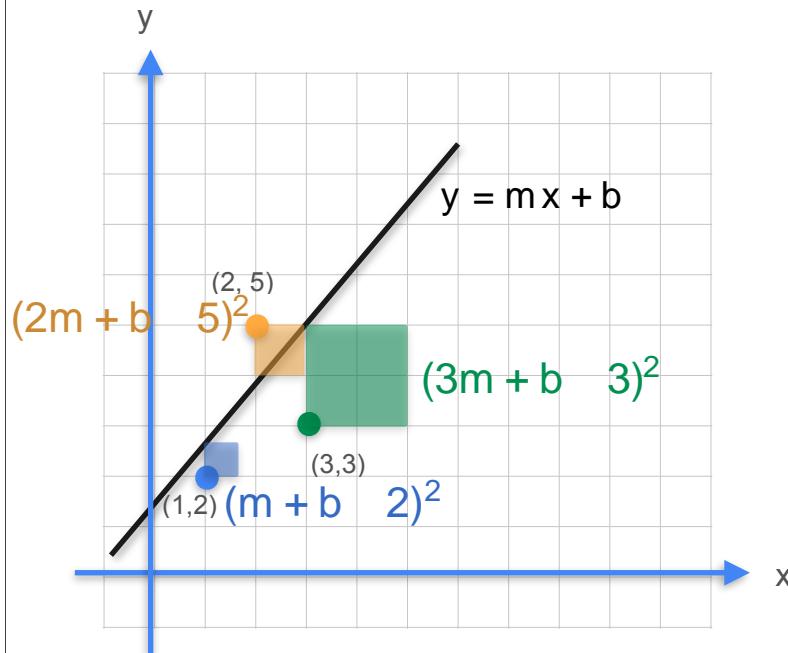
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$(m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2$$
$$m^2 + b^2 + 4 + 2mb - 4m - 4b$$

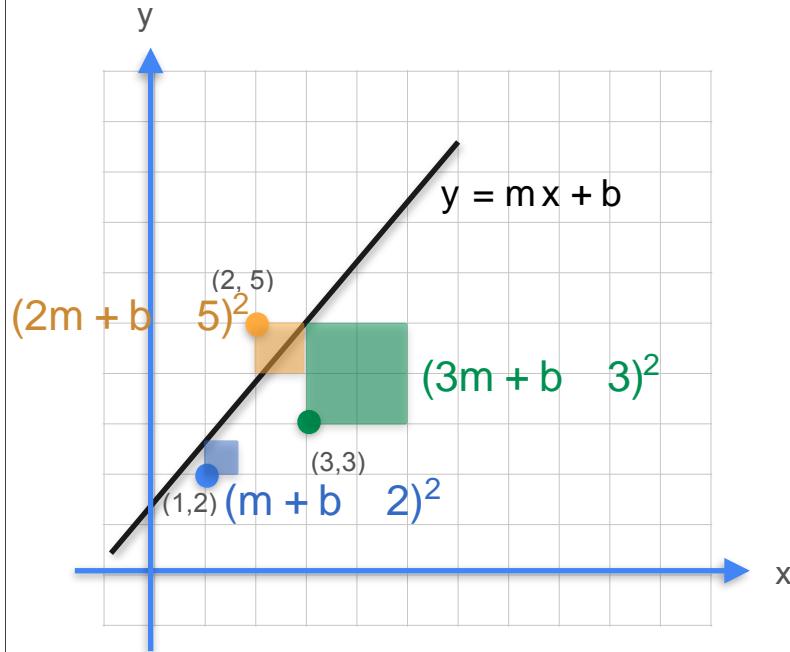
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$\begin{aligned}(m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2 \\ m^2 + b^2 + 4 + 2mb - 4m - 4b \\ + 4m^2 + b^2 + 25 + 4mb - 20m - 10b\end{aligned}$$

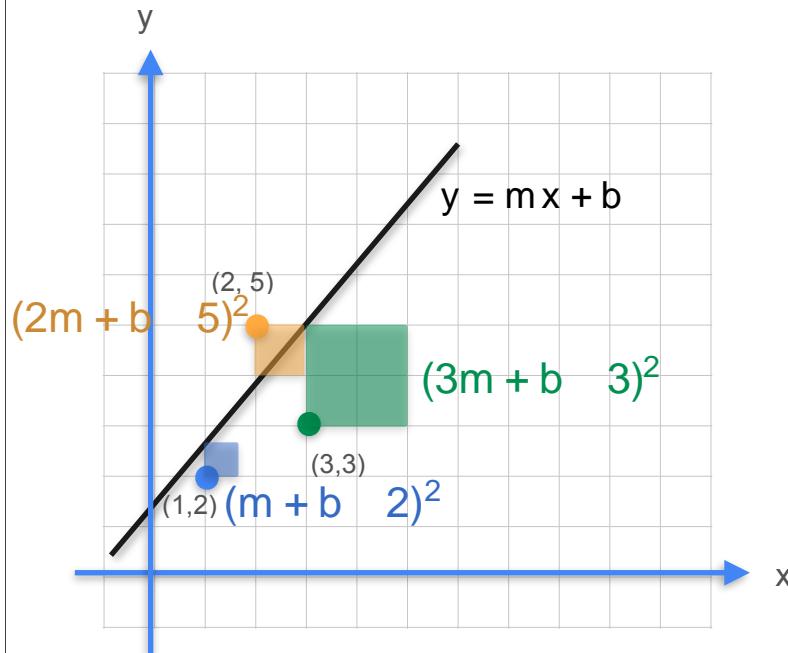
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$\begin{aligned} & (m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2 \\ & m^2 + b^2 + 4 + 2mb - 4m - 4b \\ & + 4m^2 + b^2 + 25 + 4mb - 20m - 10b \\ & + 9m^2 + b^2 + 9 + 6mb - 18m - 6b \end{aligned}$$

# Linear Regression: Analytical Approach

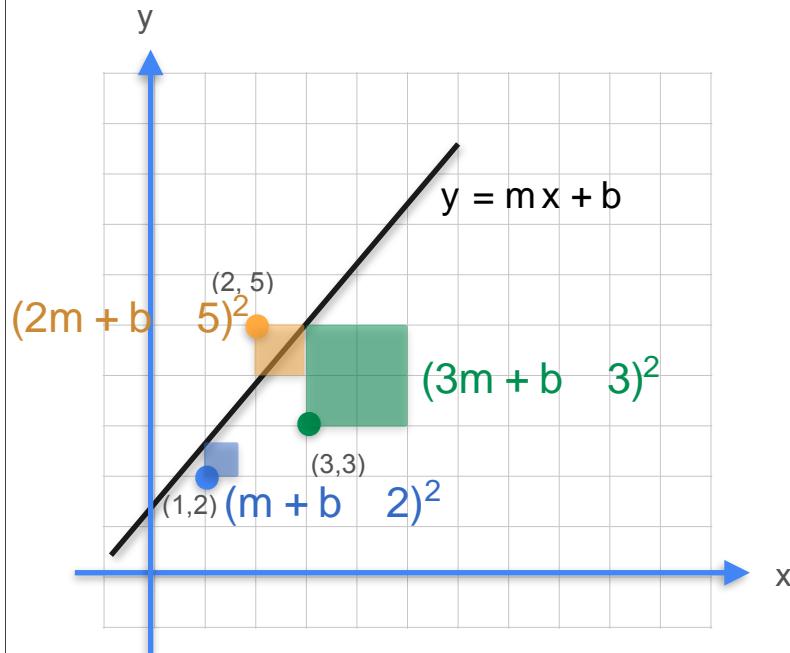


**Goal: Minimize sum of squares cost**

$$\begin{aligned} & (m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2 \\ & m^2 + b^2 + 4 + 2mb - 4m - 4b \\ & + 4m^2 + b^2 + 25 + 4mb - 20m - 10b \\ & + 9m^2 + b^2 + 9 + 6mb - 18m - 6b \end{aligned}$$

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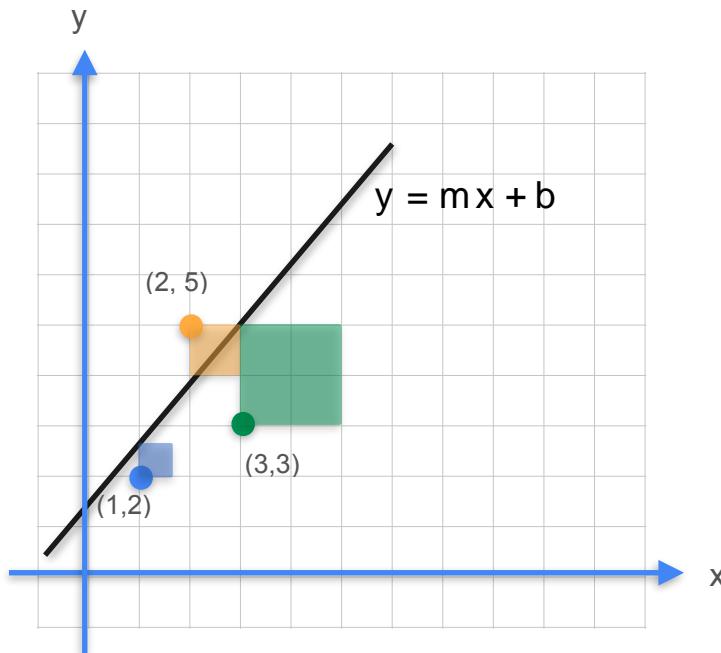
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$\begin{aligned} & (m + b - 2)^2 + (2m + b - 5)^2 + (3m + b - 3)^2 \\ & m^2 + b^2 + 4 + 2mb - 4m - 4b \\ & + 4m^2 + b^2 + 25 + 4mb - 20m - 10b \\ & + 9m^2 + b^2 + 9 + 6mb - 18m - 6b \\ \hline E(m, b) & = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b \end{aligned}$$

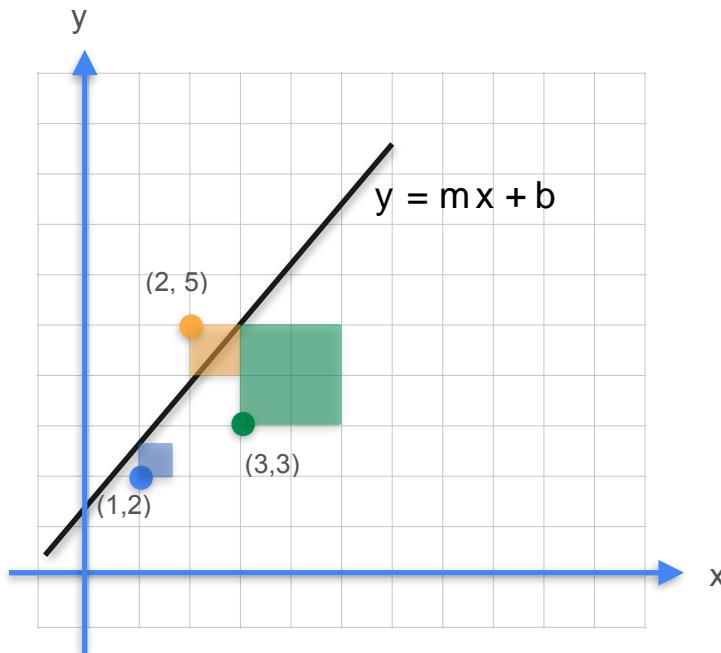
# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

# Linear Regression: Analytical Approach

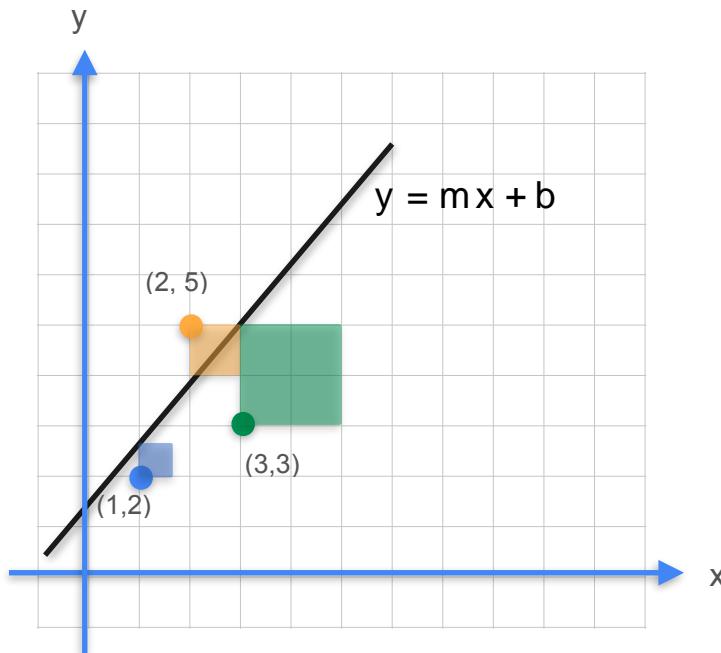


**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 0$$

# Linear Regression: Analytical Approach



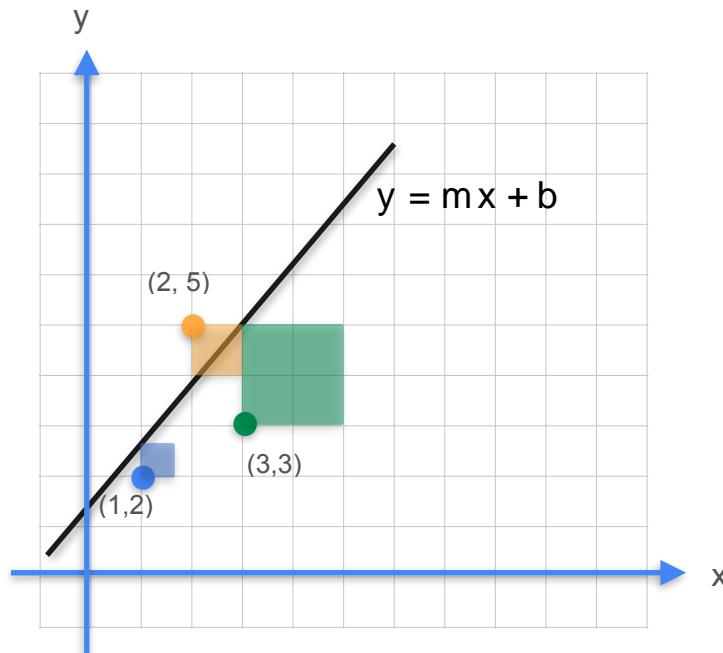
**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 0$$

$$\frac{\partial E}{\partial b} = 0$$

# Linear Regression: Analytical Approach



Goal: Minimize sum of squares cost

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

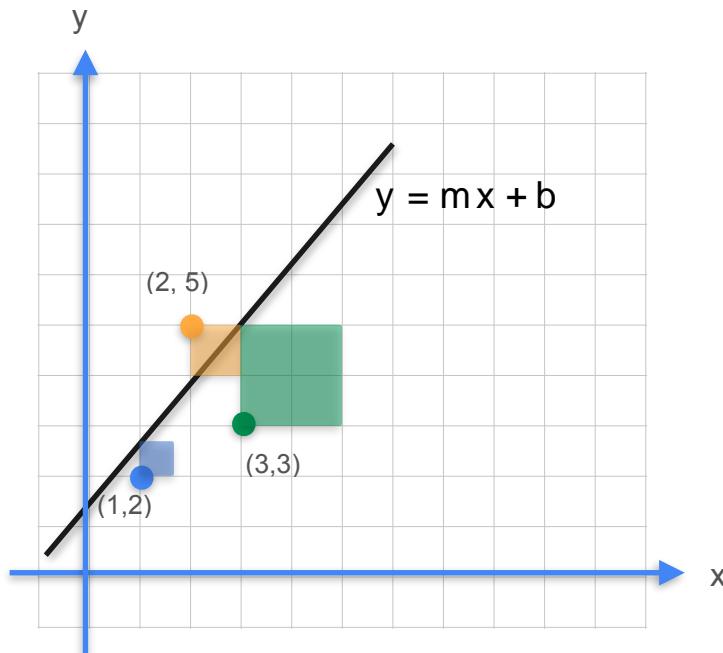
$$\frac{\partial E}{\partial m} = 0$$

Quiz:

$$\frac{\partial E}{\partial b} = 0$$

Find the partial derivative of  $E$  with respect to  $m$

# Linear Regression: Analytical Approach



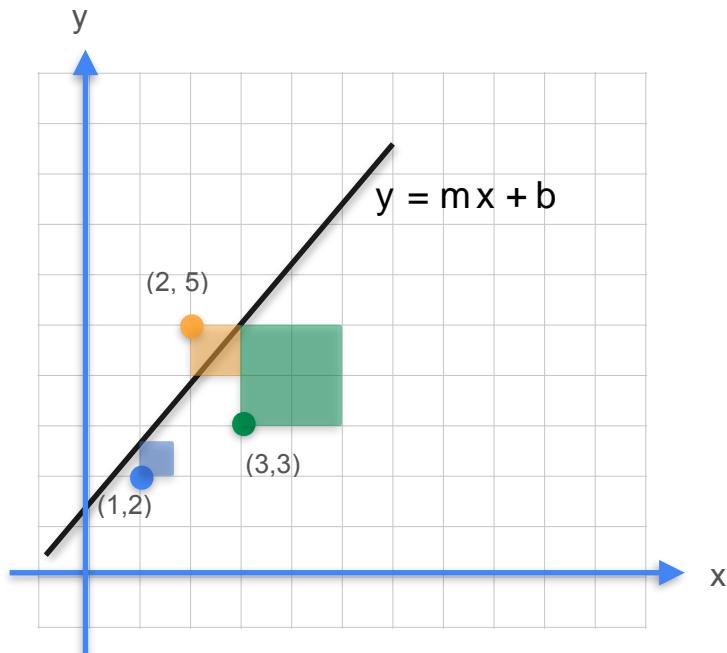
**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42$$

$$\frac{\partial E}{\partial b} =$$

# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

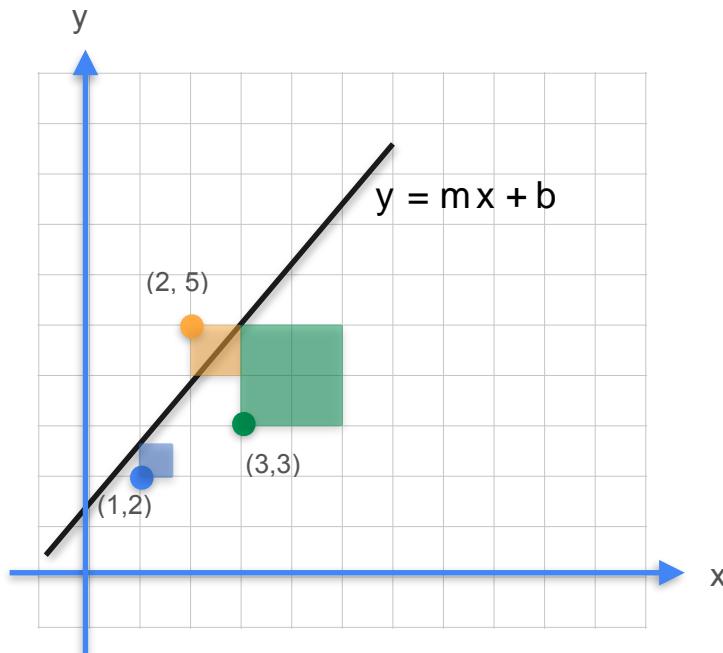
$$\frac{\partial E}{\partial m} = 28m + 12b - 42$$

$$\frac{\partial E}{\partial b} =$$

**Quiz:**

**Find the partial derivative of  $E$  with respect to  $b$**

# Linear Regression: Analytical Approach



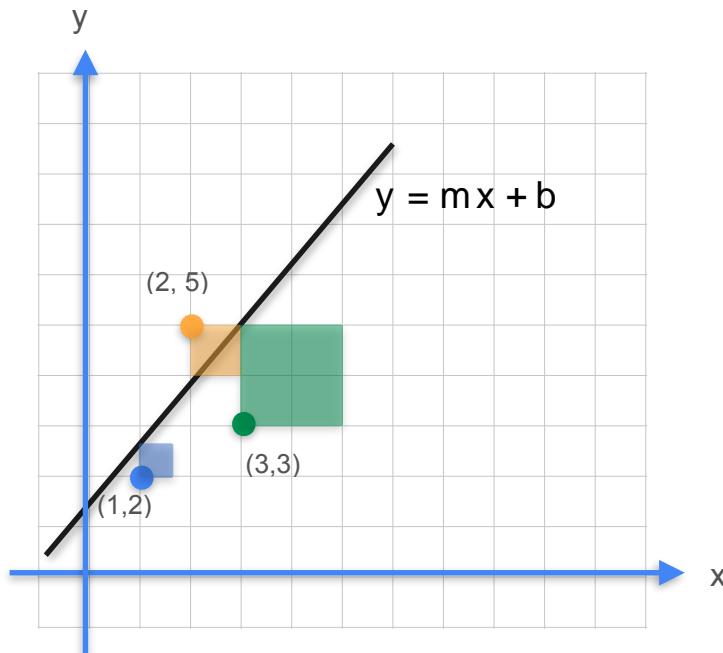
**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20$$

# Linear Regression: Analytical Approach



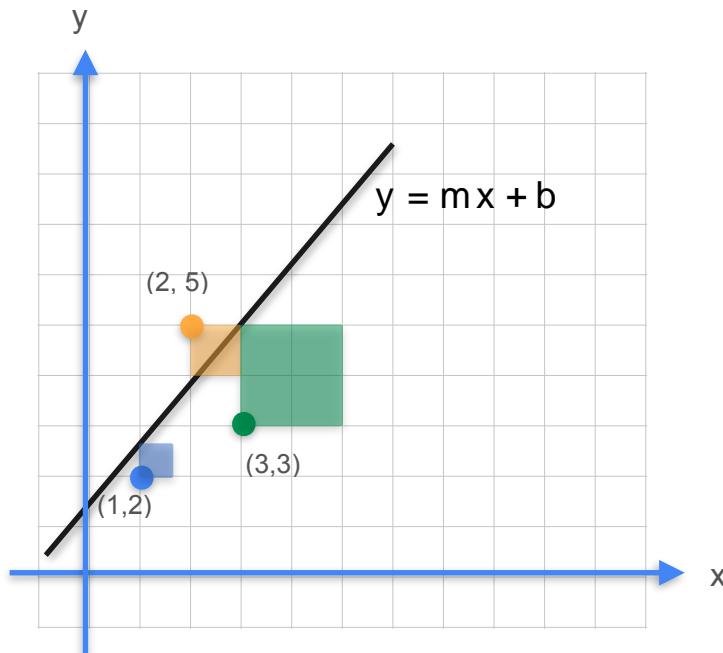
**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{E}{m} = 28m + 12b - 42$$

$$\frac{E}{b} = 6b + 12m - 20$$

# Linear Regression: Analytical Approach



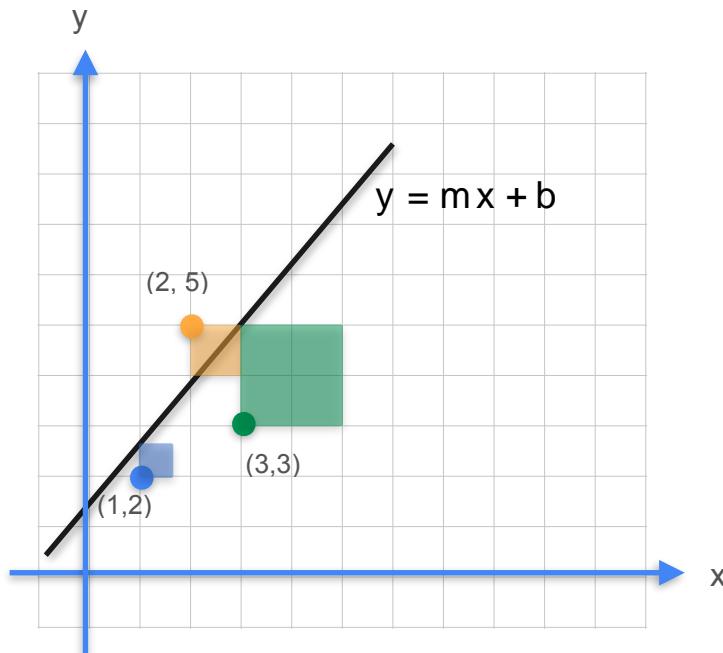
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$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

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# Linear Regression: Analytical Approach



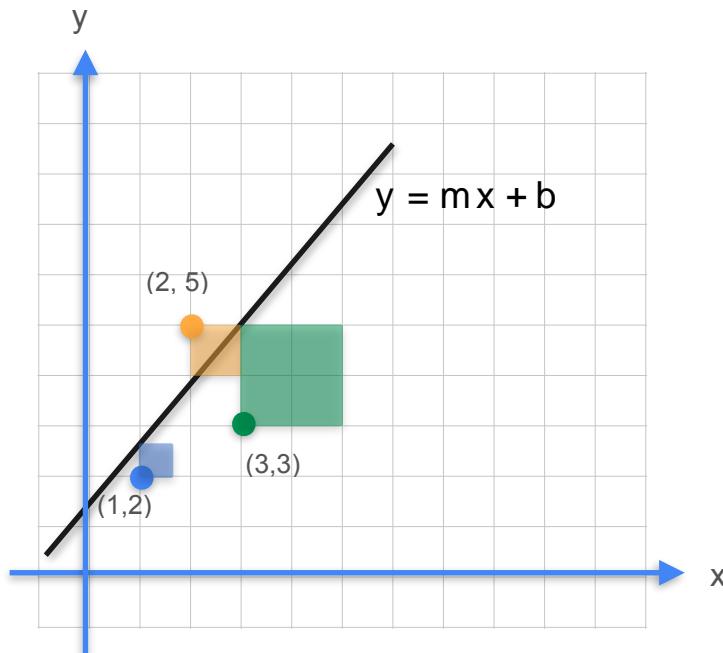
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# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

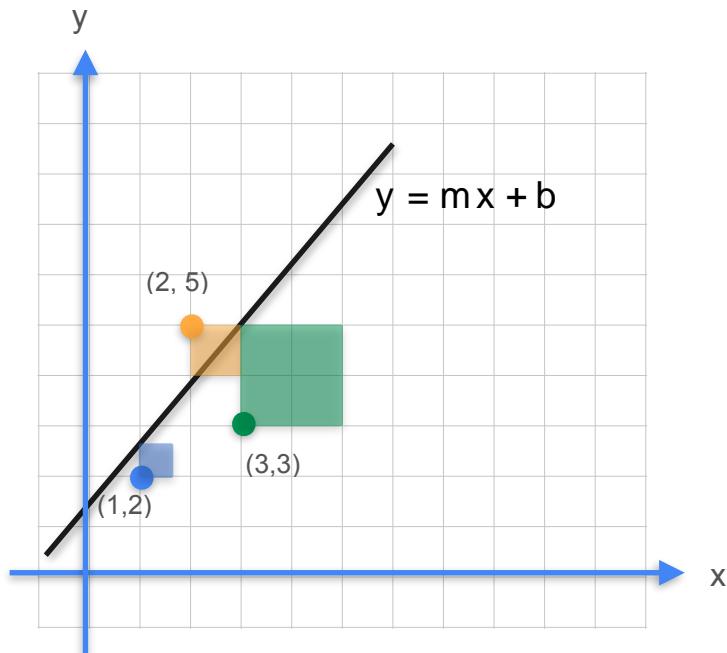
$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$m =$$

# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

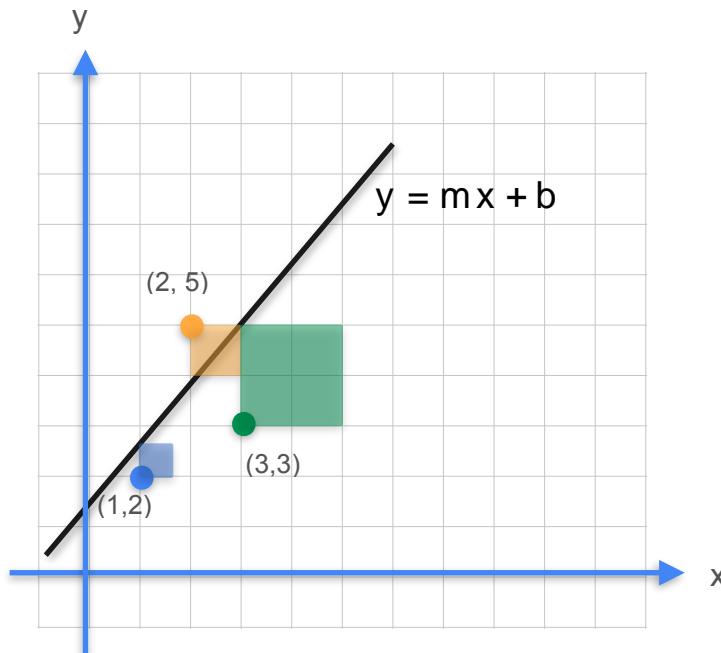
$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$m =$$

$$b =$$

# Linear Regression: Analytical Approach



**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$

# Linear Regression: Analytical Approach

Goal: Minimize sum of squares cost

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$\begin{matrix} m = ? \\ b = ? \end{matrix}$$

# Linear Regression: Analytical Approach

Goal: Minimize sum of squares cost

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

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$$\begin{matrix} m = ? \\ b = ? \end{matrix}$$



# Linear Regression: Analytical Approach

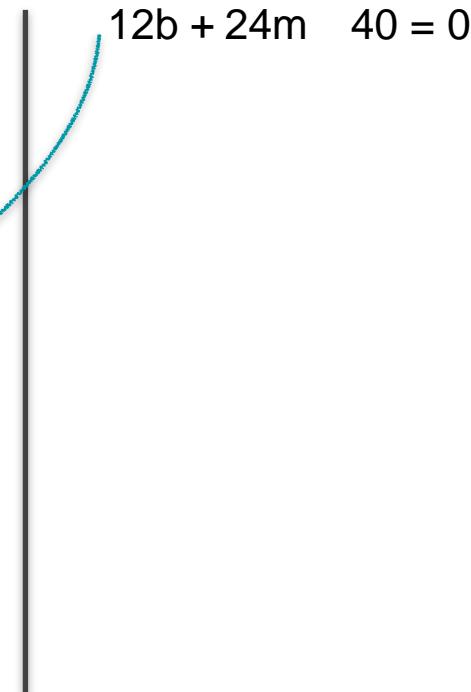
Goal: Minimize sum of squares cost

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# Linear Regression: Analytical Approach

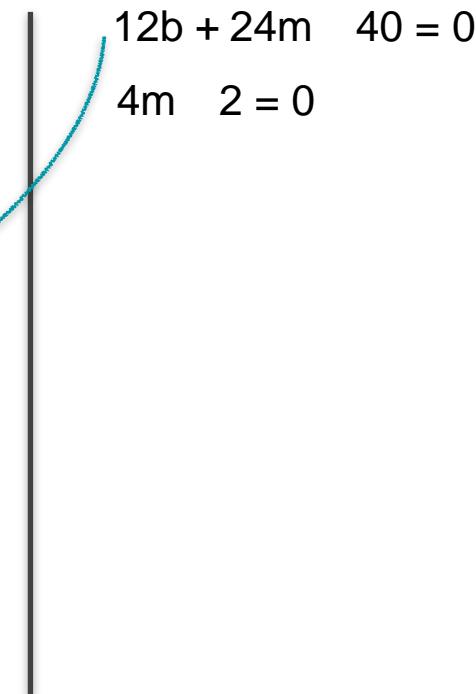
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# Linear Regression: Analytical Approach

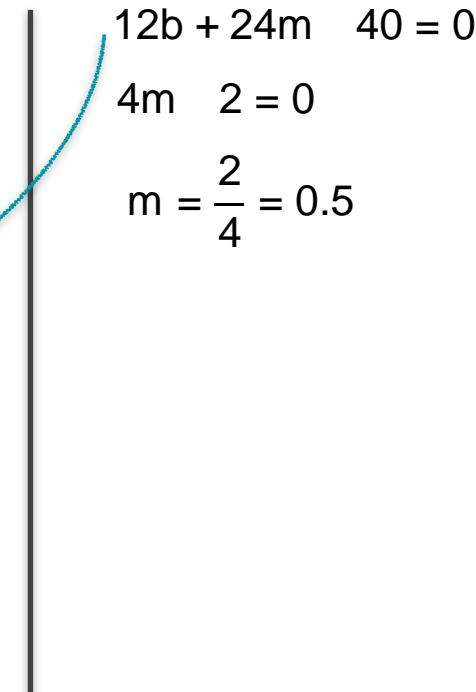
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$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

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$$\begin{matrix} m = \\ b = \end{matrix} ?$$



# Linear Regression: Analytical Approach

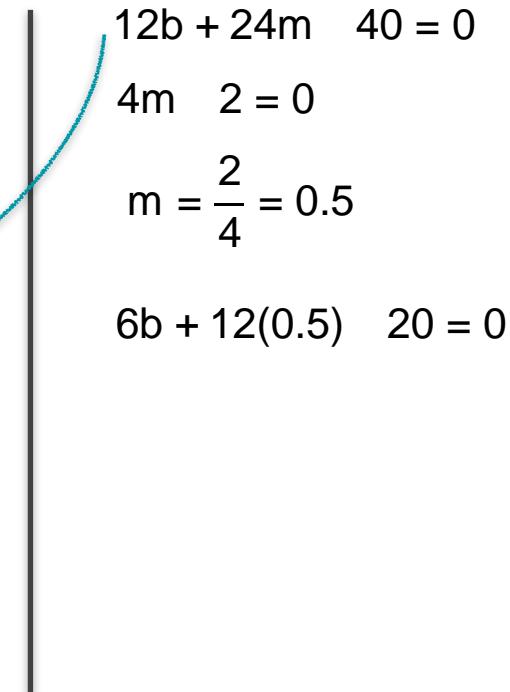
Goal: Minimize sum of squares cost

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$$\begin{matrix} m = \\ b = \end{matrix} ?$$



# Linear Regression: Analytical Approach

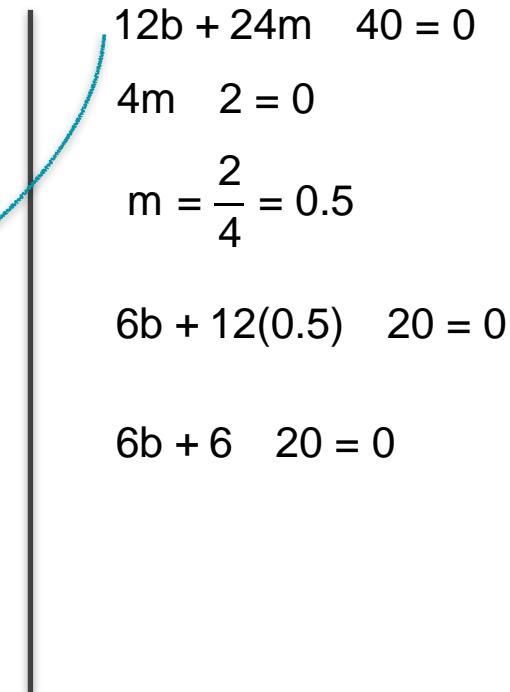
Goal: Minimize sum of squares cost

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

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$$\begin{matrix} m = \\ b = \end{matrix} ?$$



# Linear Regression: Analytical Approach

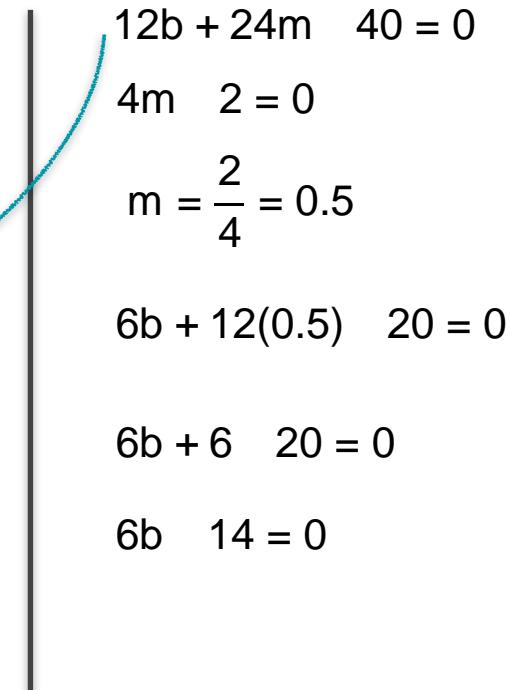
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$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

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$$\begin{matrix} m = \\ b = \end{matrix} ?$$



# Linear Regression: Analytical Approach

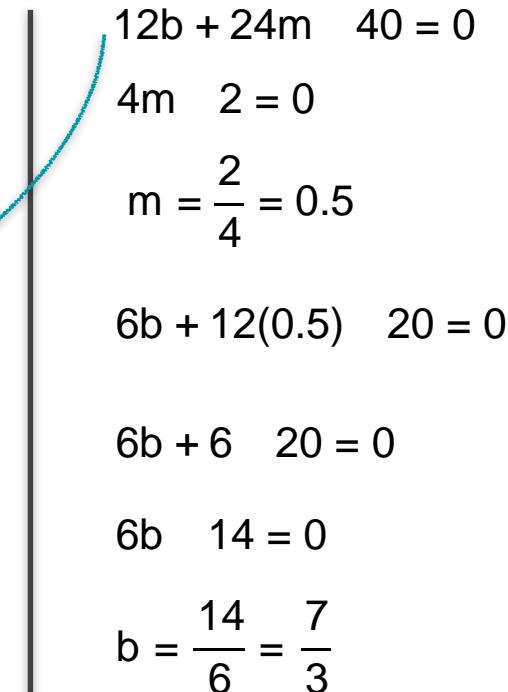
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$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$


$$\begin{aligned} 12b + 24m - 40 &= 0 \\ 4m - 2 &= 0 \\ m &= \frac{2}{4} = 0.5 \\ 6b + 12(0.5) - 20 &= 0 \\ 6b + 6 - 20 &= 0 \\ 6b - 14 &= 0 \\ b &= \frac{14}{6} = \frac{7}{3} \end{aligned}$$

# Linear Regression: Analytical Approach

Goal: Minimize sum of squares cost

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

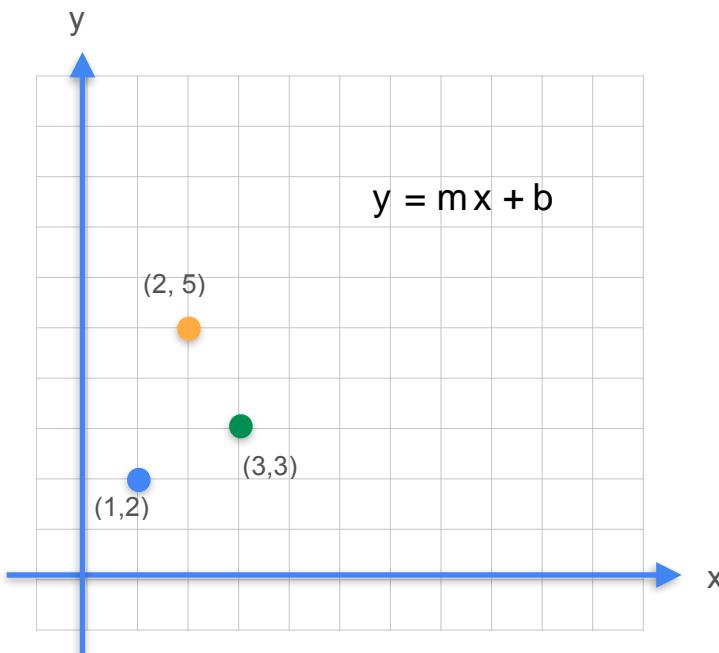
$$\begin{matrix} m = ? \\ b = ? \end{matrix}$$

$$m = \frac{1}{2}$$

$$b = \frac{7}{3}$$

$$E(m = \frac{1}{2}, b = \frac{7}{3}) = 4.167$$

# Linear Regression: Optimal Solution

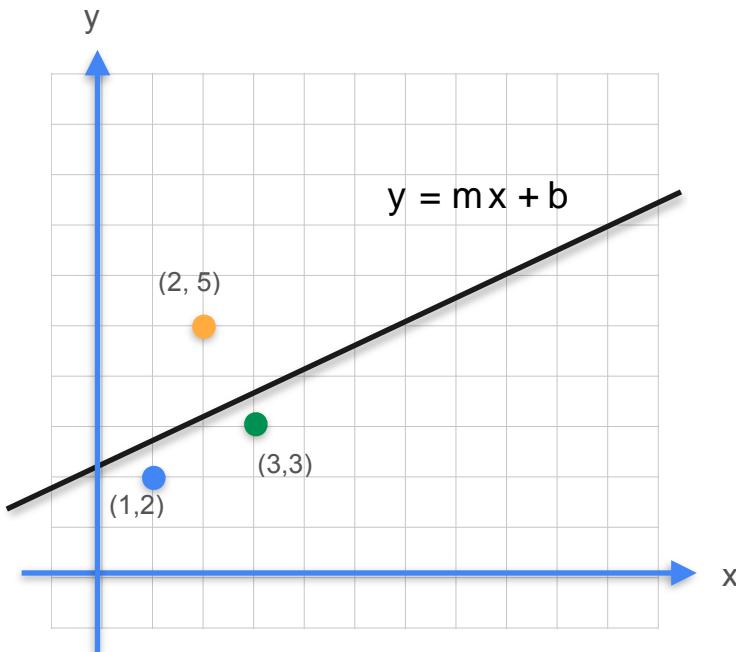


$$m = \frac{1}{2}$$

$$b = \frac{7}{3}$$

$$E(m = \frac{1}{2}, b = \frac{7}{3}) = 4.167$$

# Linear Regression: Optimal Solution

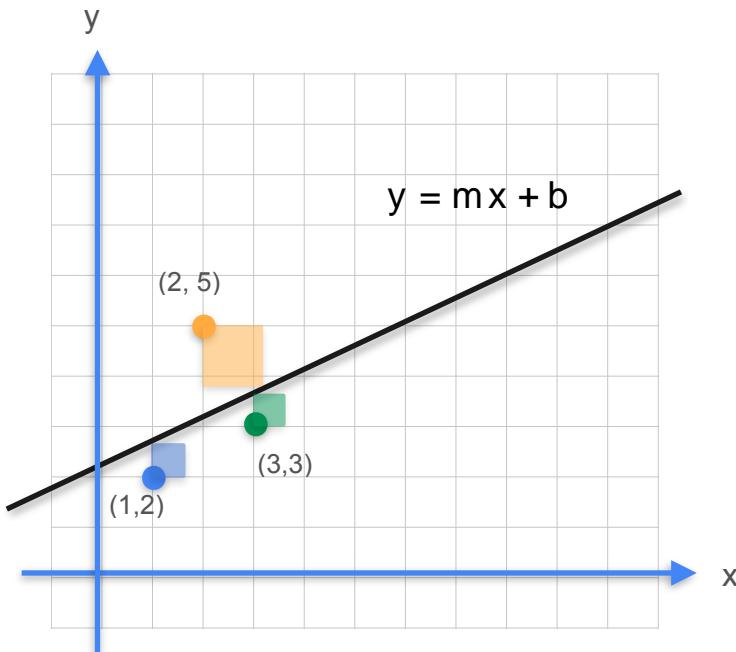


$$m = \frac{1}{2}$$

$$b = \frac{7}{3}$$

$$E(m = \frac{1}{2}, b = \frac{7}{3}) = 4.167$$

# Linear Regression: Optimal Solution



$$m = \frac{1}{2}$$

$$b = \frac{7}{3}$$

$$E(m = \frac{1}{2}, b = \frac{7}{3}) = 4.167$$

# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

$$\frac{\partial E}{\partial m} = 28m + 12b - 42 = 0$$

$$\frac{\partial E}{\partial b} = 6b + 12m - 20 = 0$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$

**Gradient Descent to the rescue**



DeepLearning.AI

## Gradients and Gradient Descent

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# Optimization using Gradient Descent in one variable - Part 1

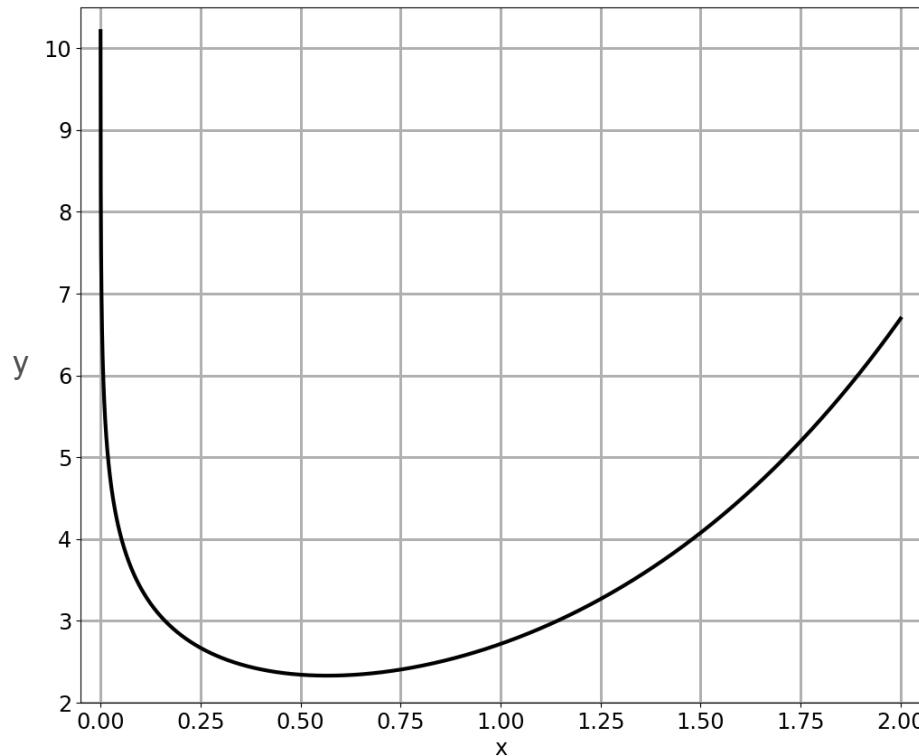
# Hard To Optimize Functions

# Hard To Optimize Functions

$$f(x) = e^x - \log(x)$$

# Hard To Optimize Functions

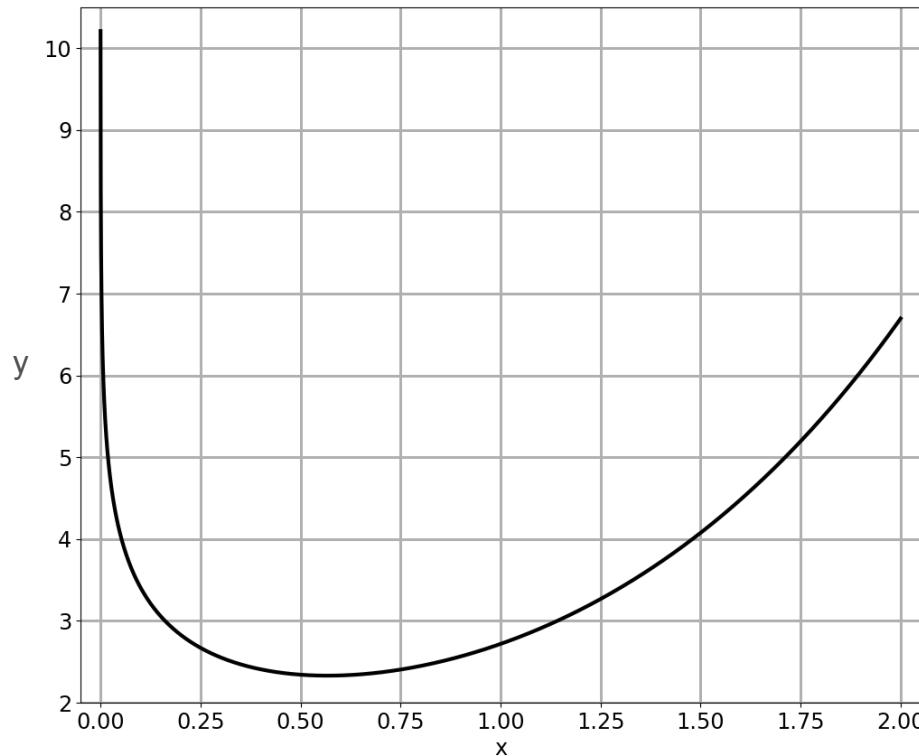
$$f(x) = e^x - \log(x)$$



# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

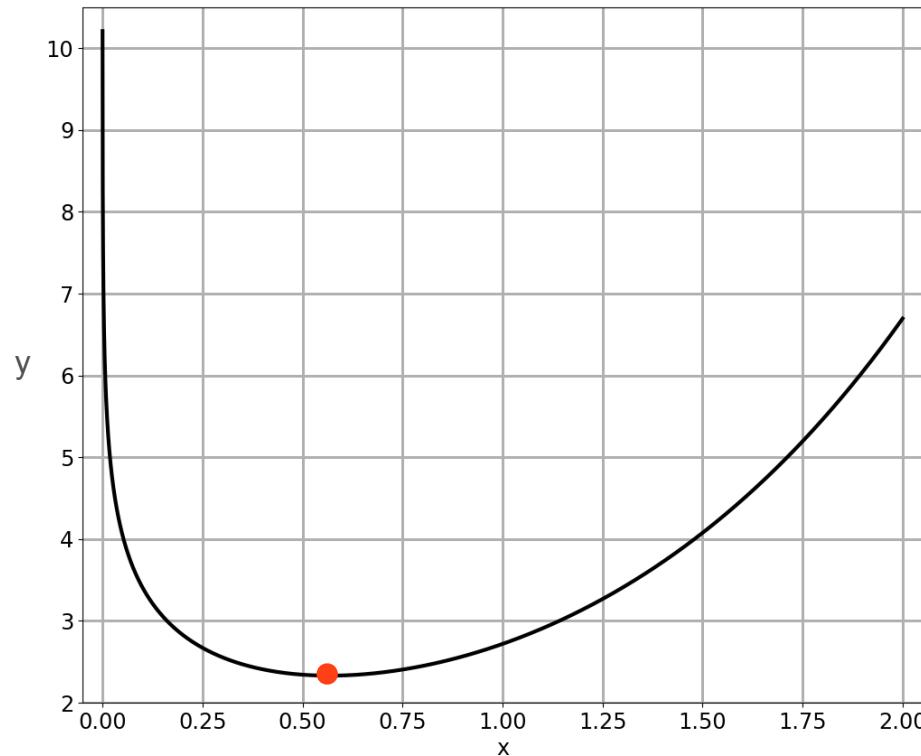
Minimum?



# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

Minimum?

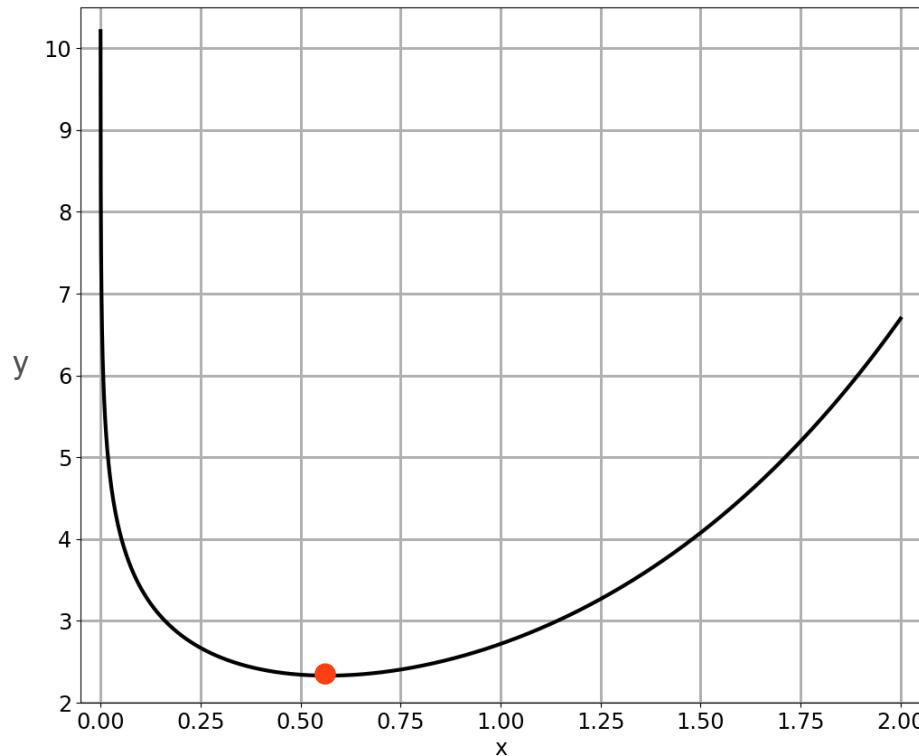


# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

Minimum?

$$f'(x)=0$$

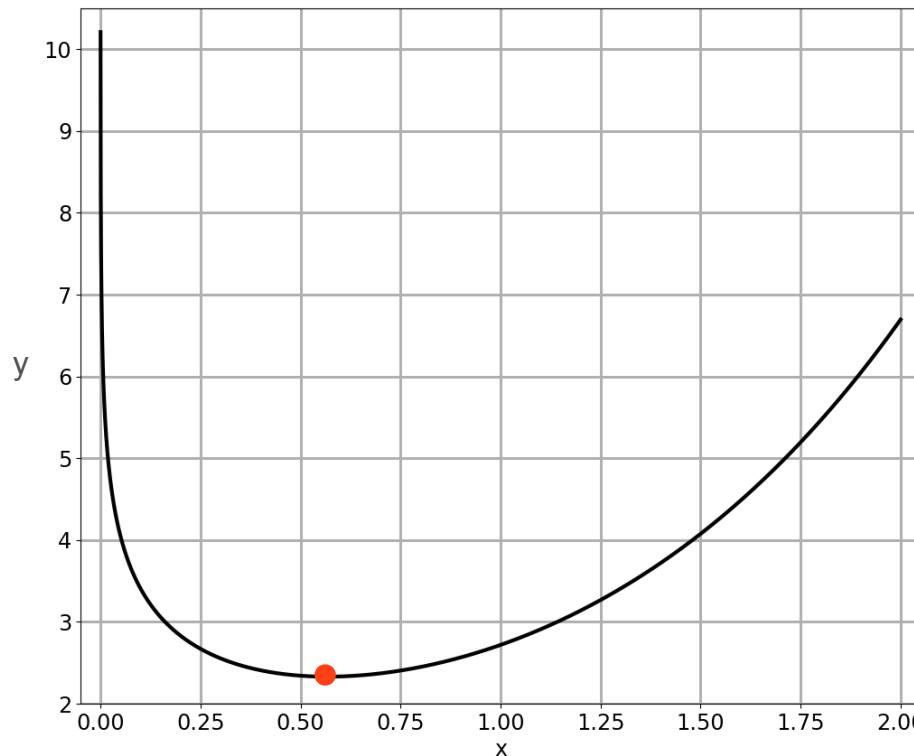


# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

Minimum?

$$f'(x) = e^x - \frac{1}{x} = 0$$

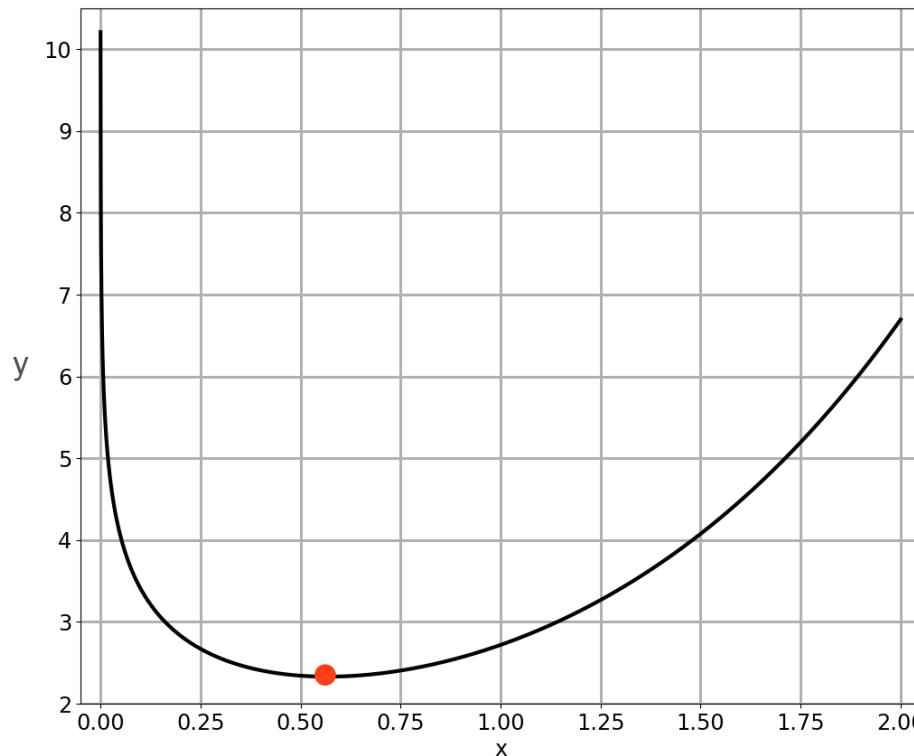


# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

Minimum?

$$f'(x) = e^x - \frac{1}{x} = 0$$

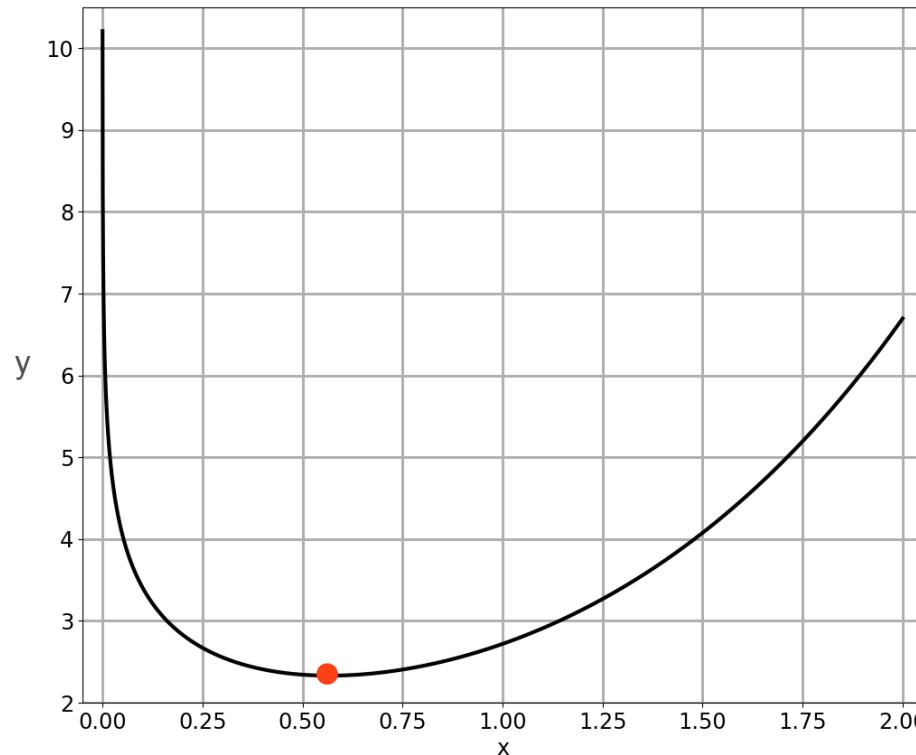


# Hard To Optimize Functions

$$f(x) = e^x \quad \log(x)$$

Minimum?

$$f'(x) = e^x \quad \frac{1}{x} = 0$$

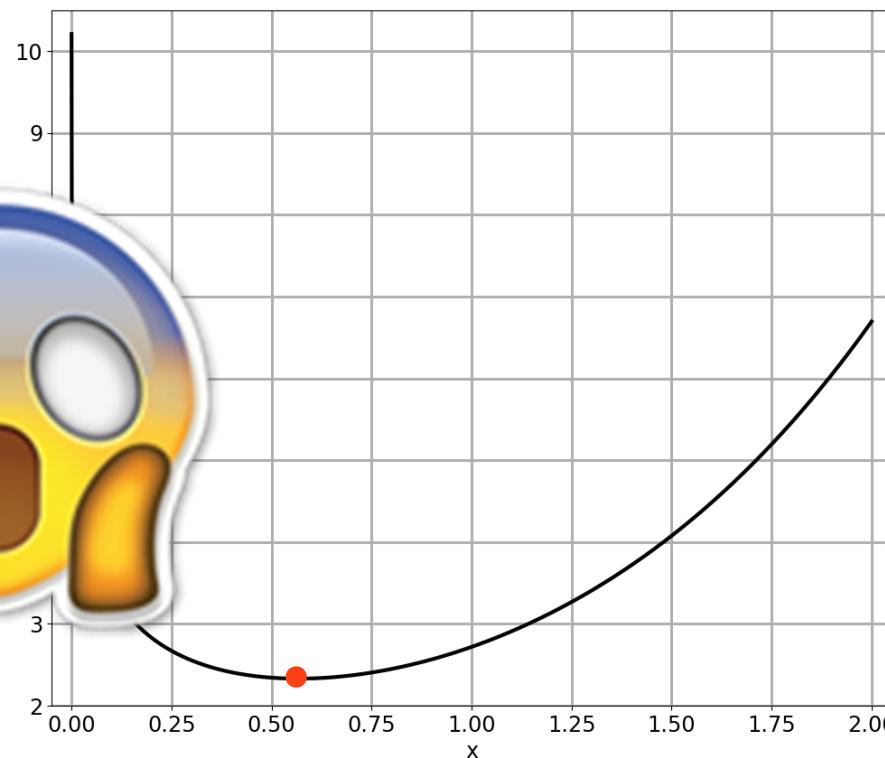


# Hard To Optimize Functions

$$f(x) = e^x \quad \log(x)$$

Minimum?

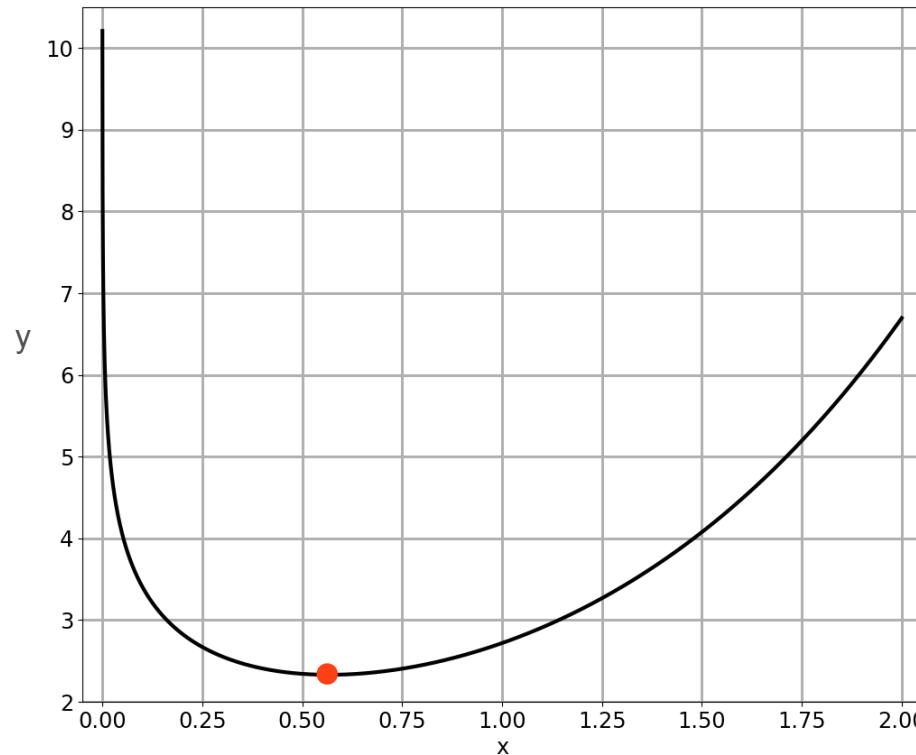
$$f(x) = e^x \quad \frac{1}{x} = 0$$



# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

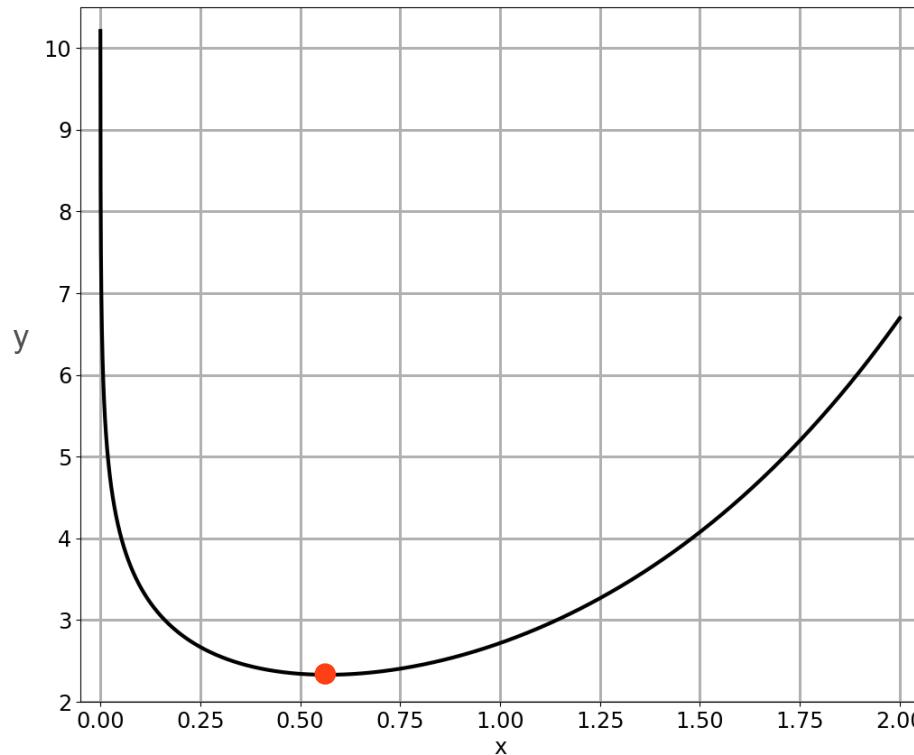
$$f(x) = e^x - \frac{1}{x}$$



# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x} = 0$$



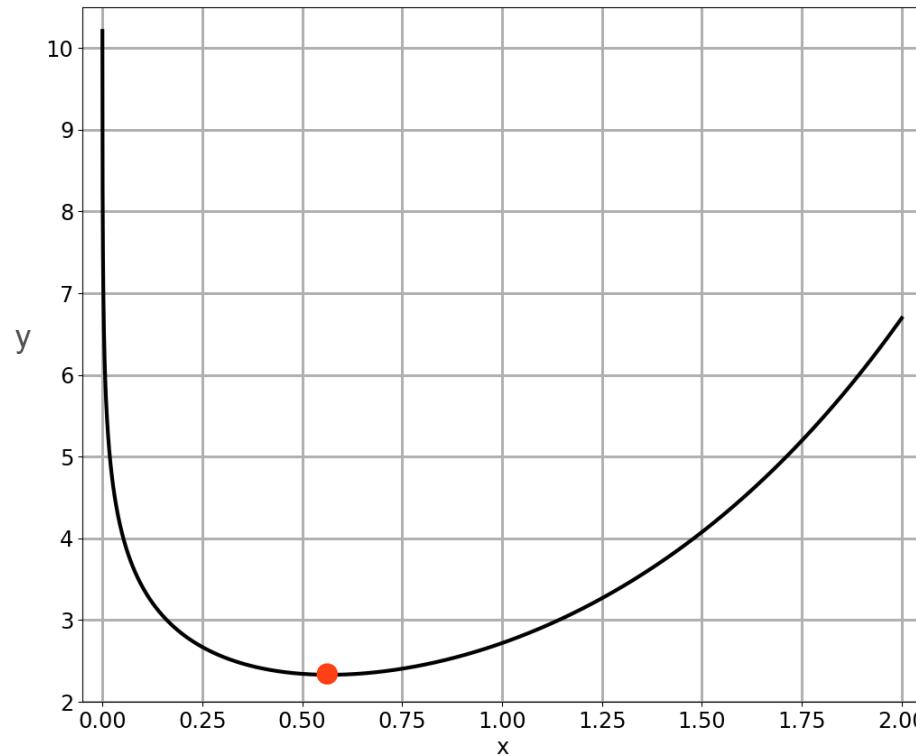
# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x} = 0$$



$$e^x = \frac{1}{x}$$



# Hard To Optimize Functions

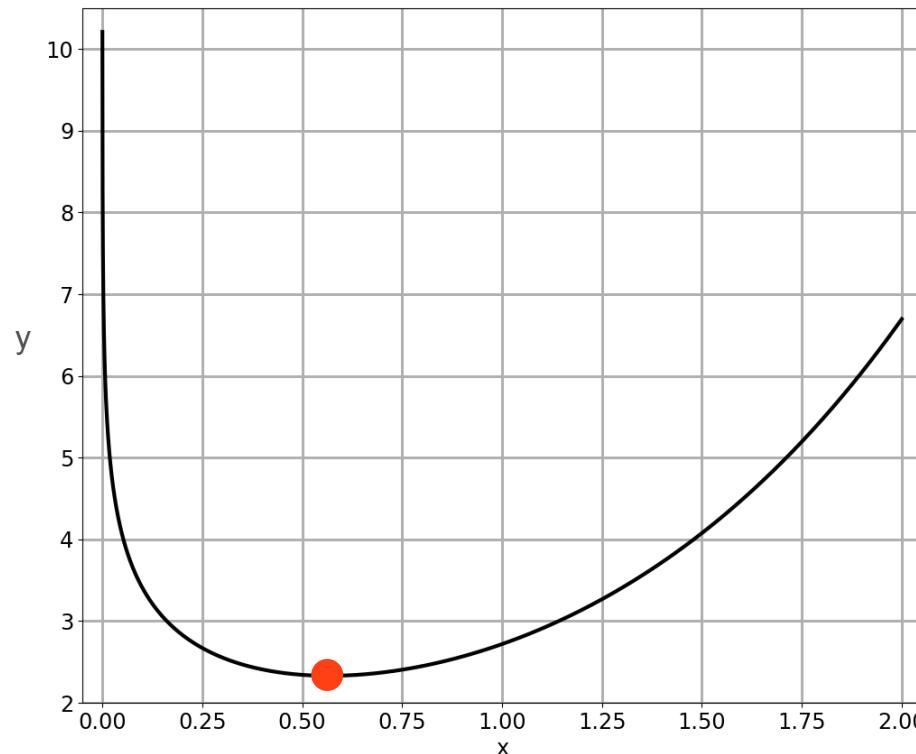
$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x} = 0$$



$$e^x = \frac{1}{x}$$

Solution:  $x = 0.5671\ldots$



# Hard To Optimize Functions

$$f(x) = e^x \log(x)$$

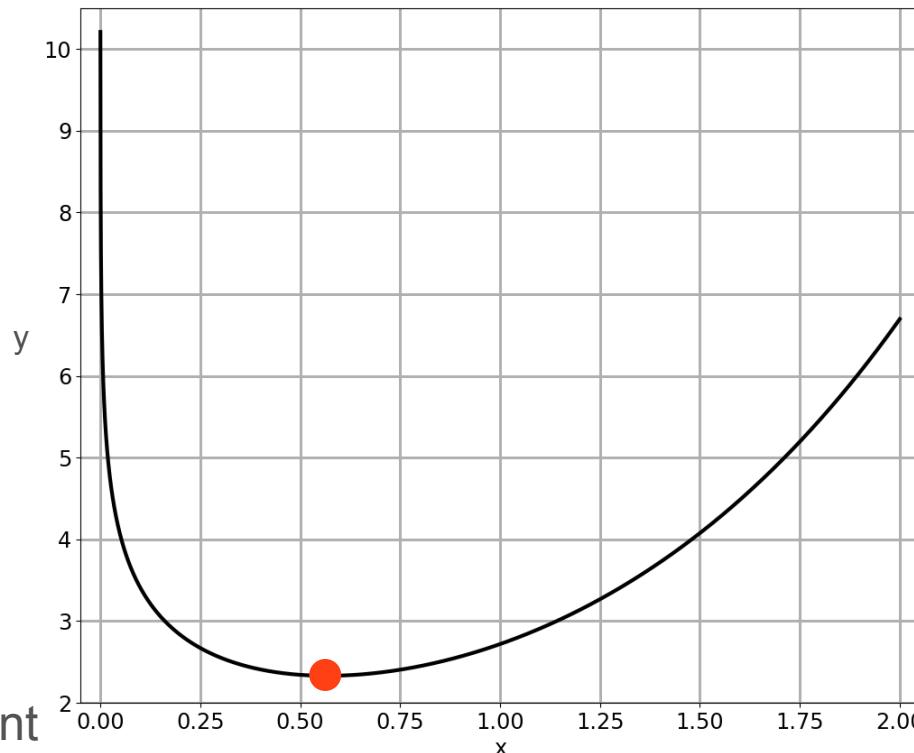
$$f'(x) = e^x - \frac{1}{x} = 0$$



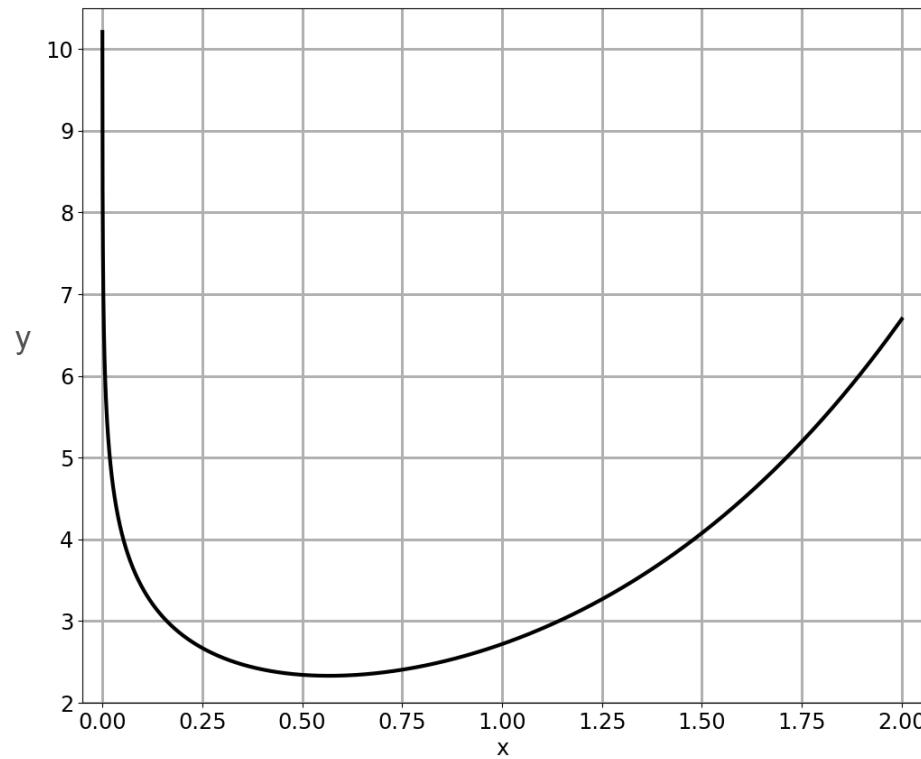
$$e^x = \frac{1}{x}$$

Solution:  $x = 0.5671\ldots$

Also known as the Omega constant

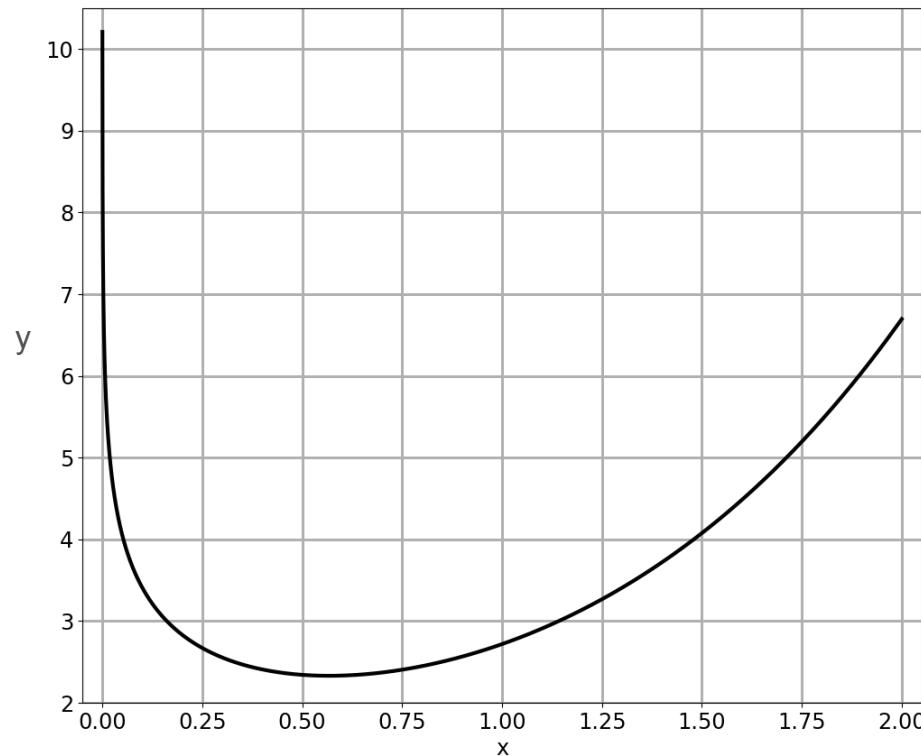


# Method 1: Try Both Directions



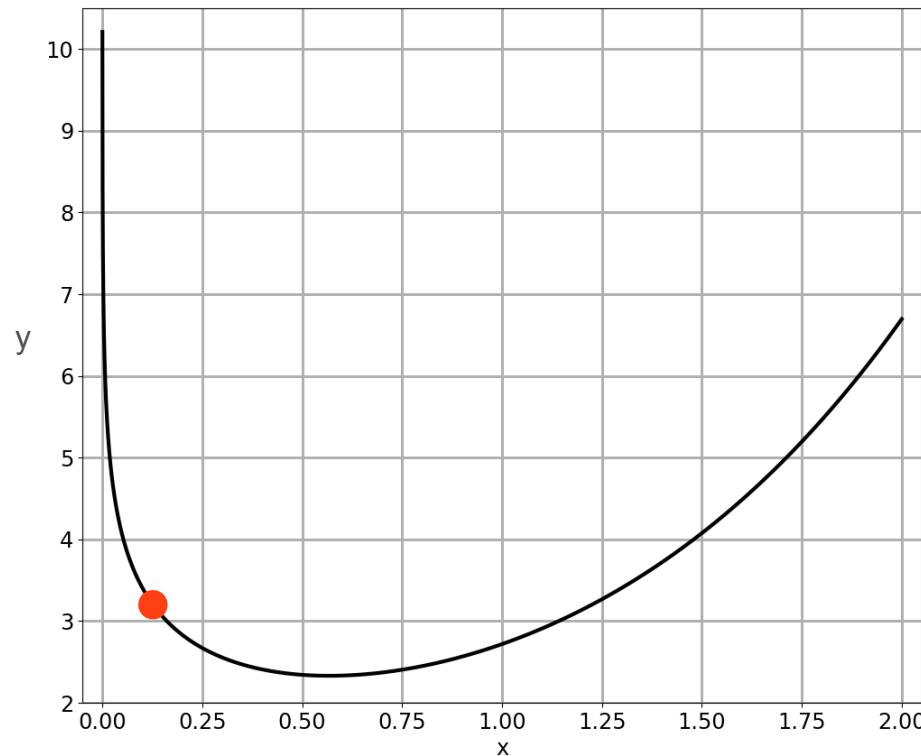
# Method 1: Try Both Directions

Is there any  
other way?



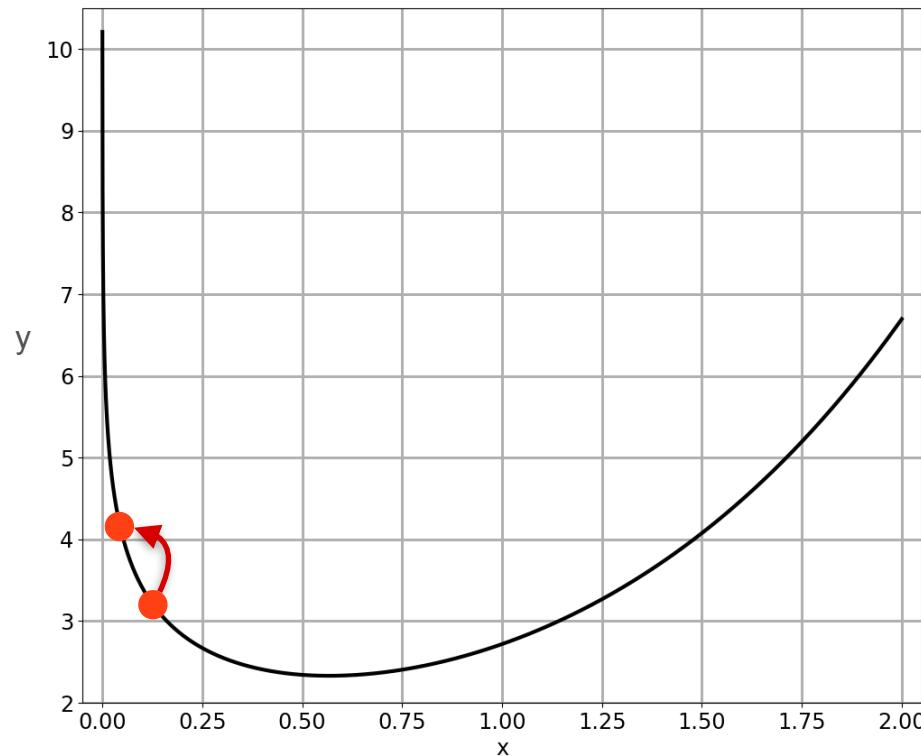
# Method 1: Try Both Directions

Is there any  
other way?



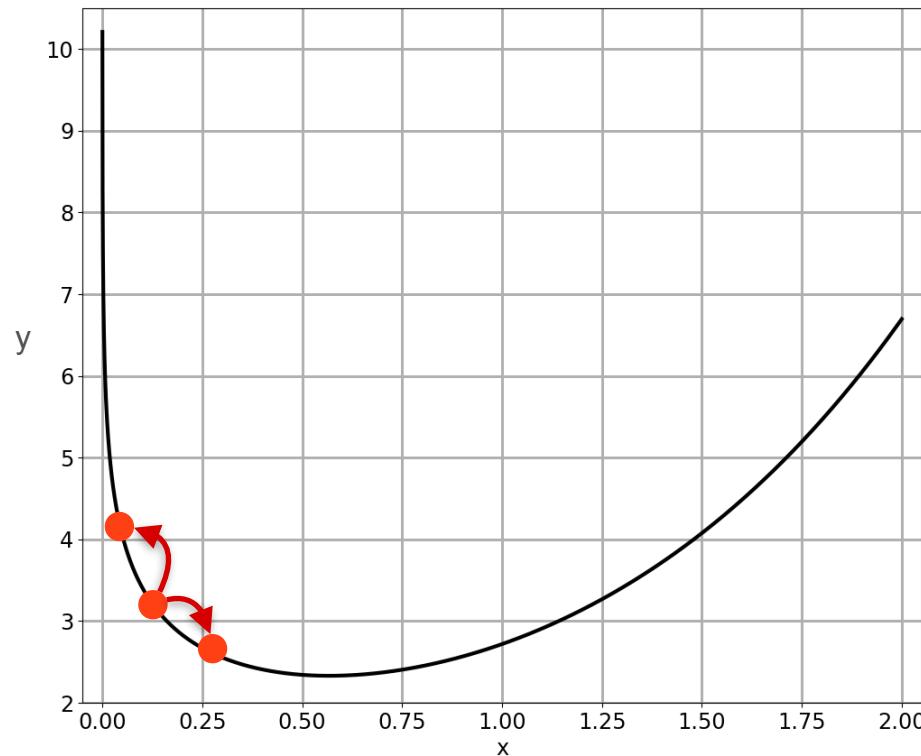
# Method 1: Try Both Directions

Is there any  
other way?



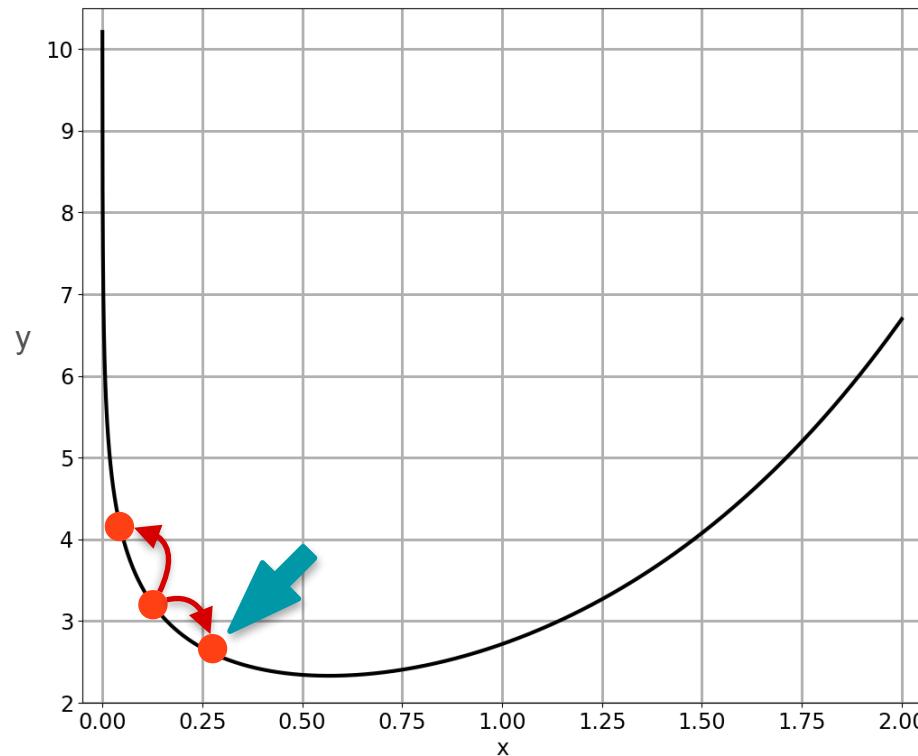
# Method 1: Try Both Directions

Is there any  
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# Method 1: Try Both Directions

Is there any  
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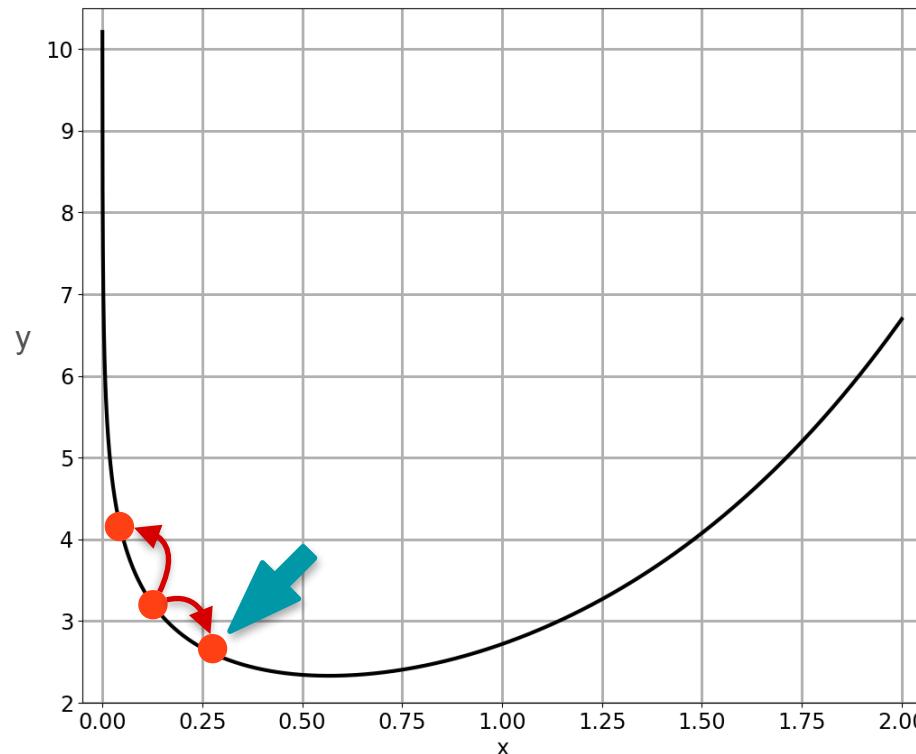


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

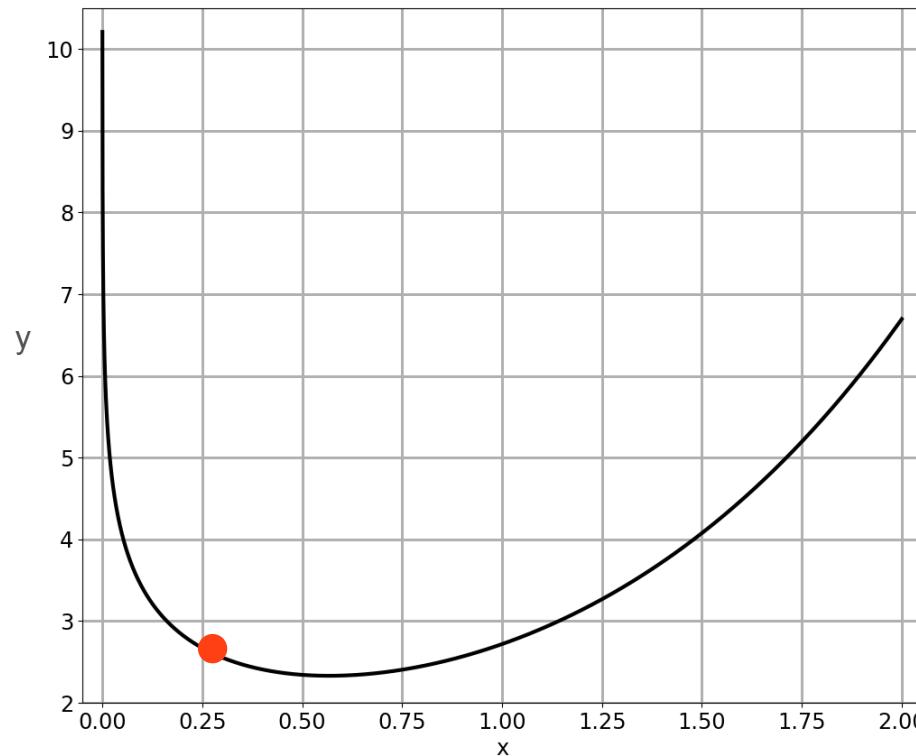


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

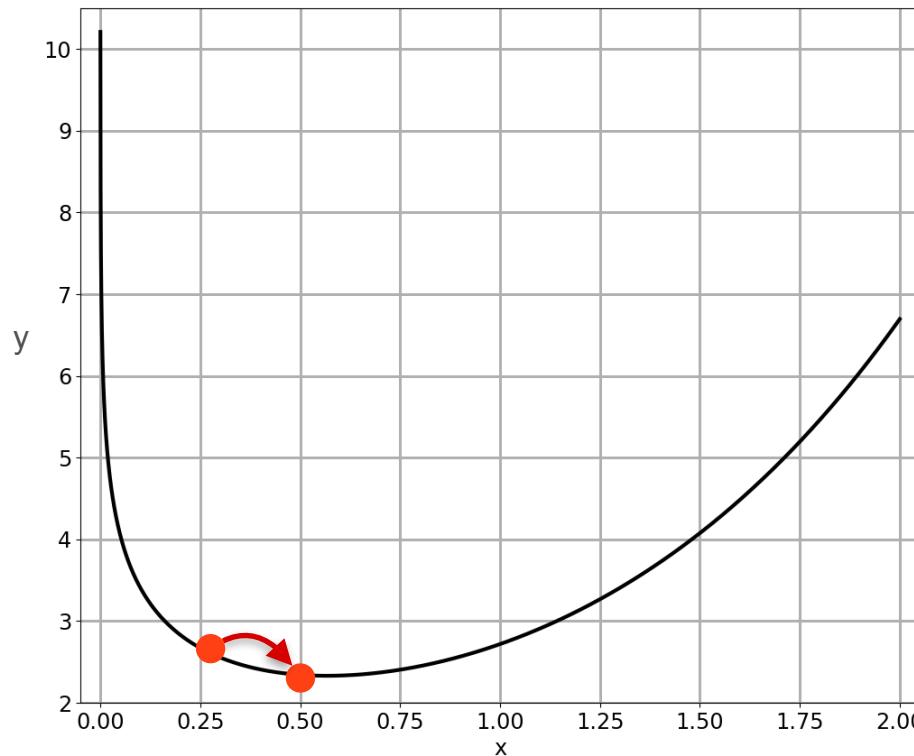


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

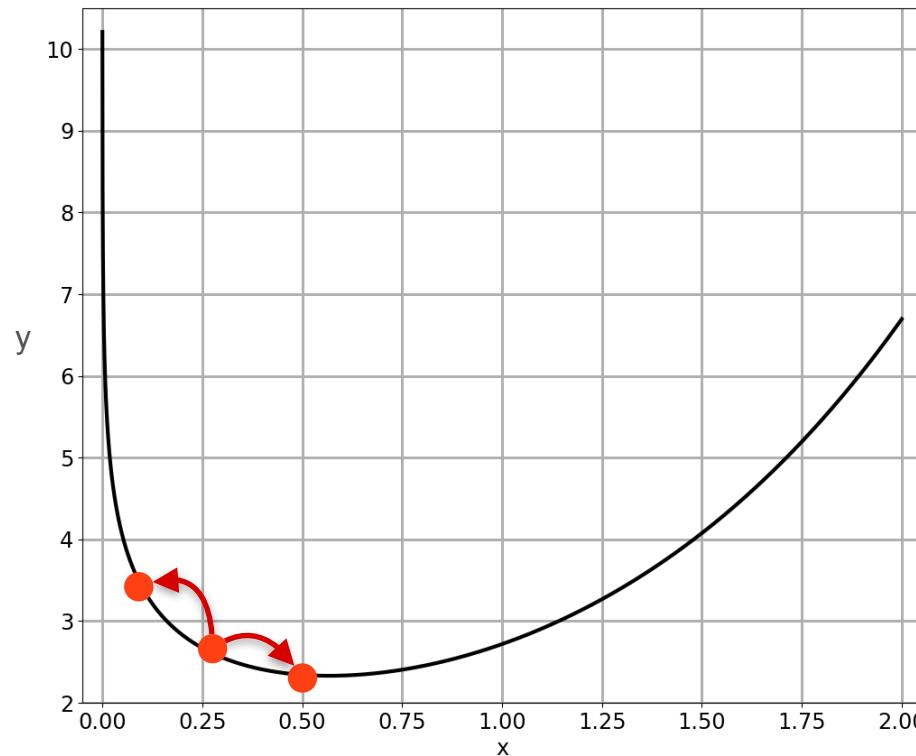


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

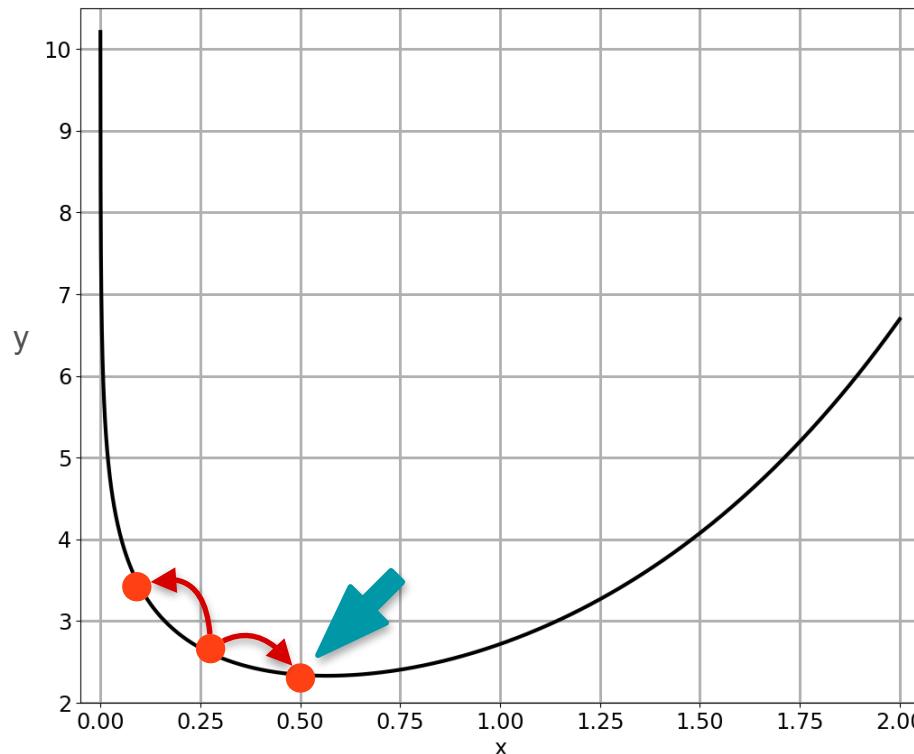


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

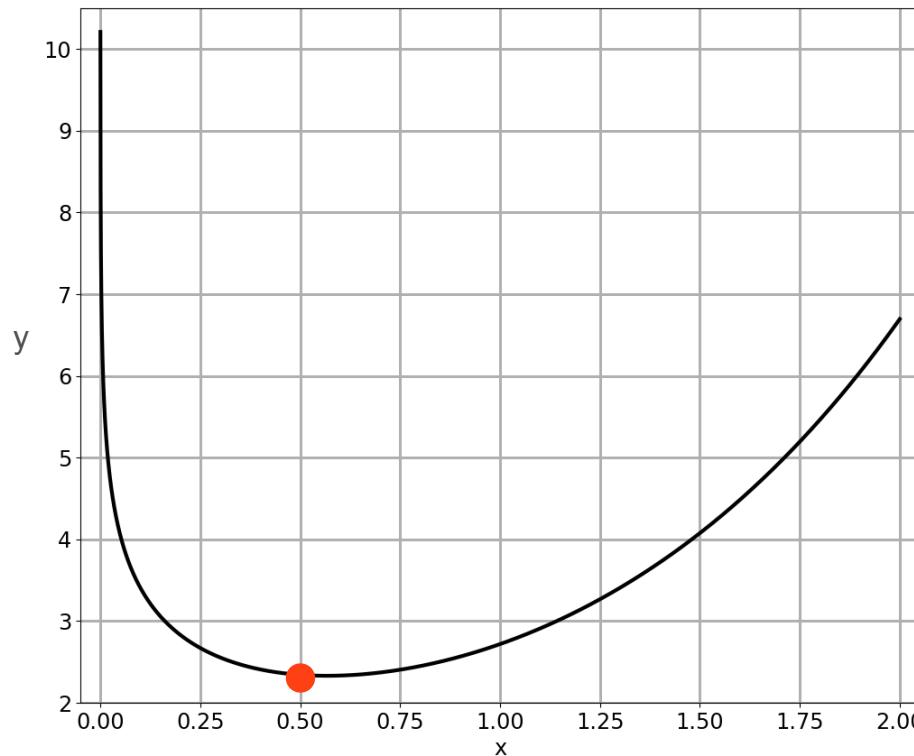


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

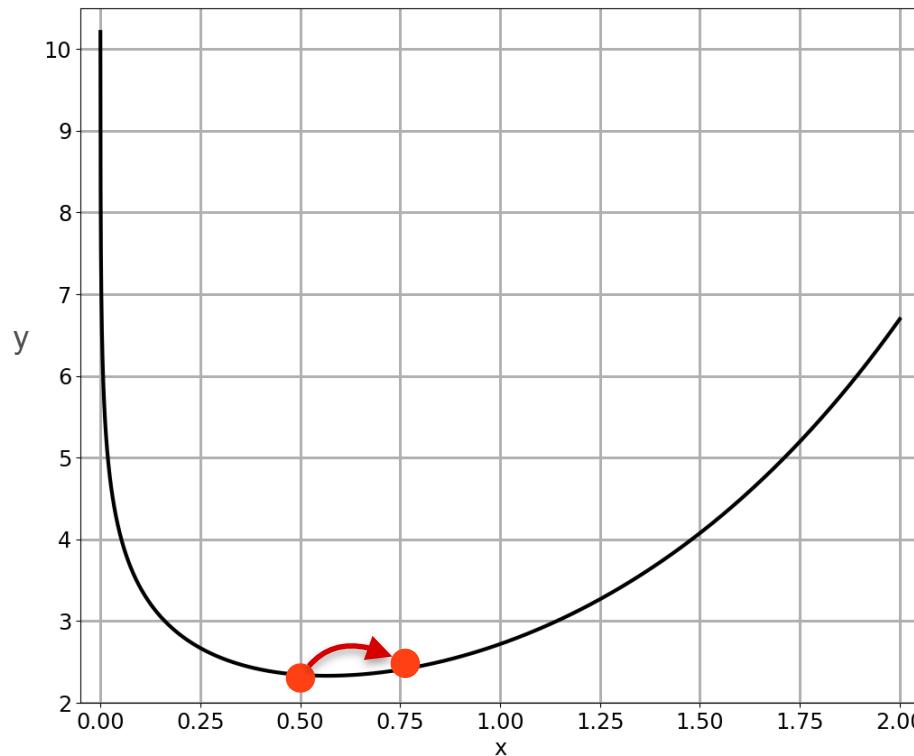


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

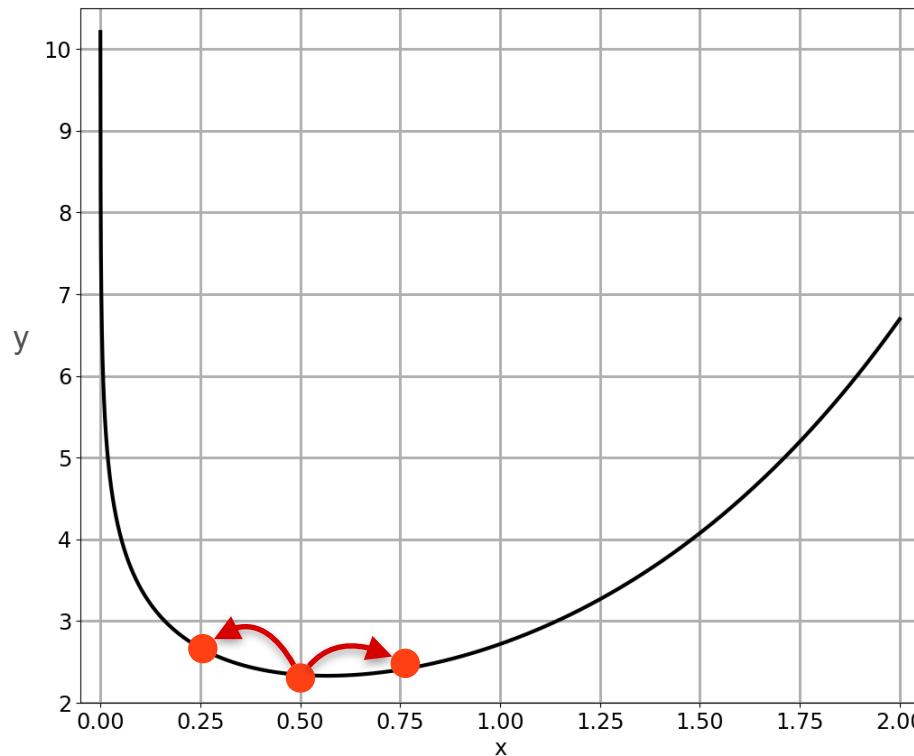


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

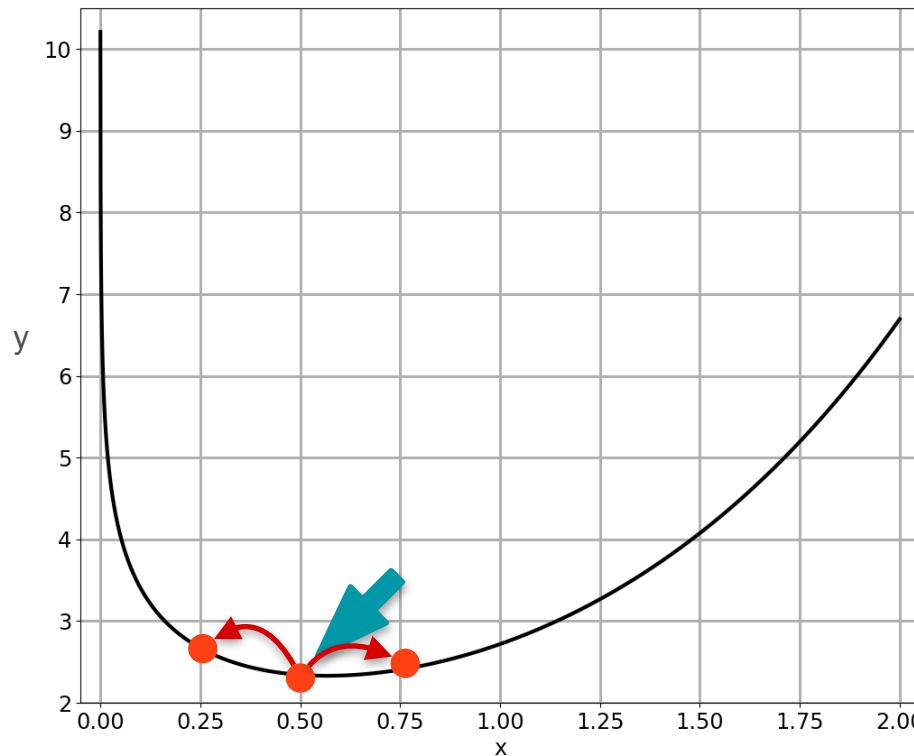


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!

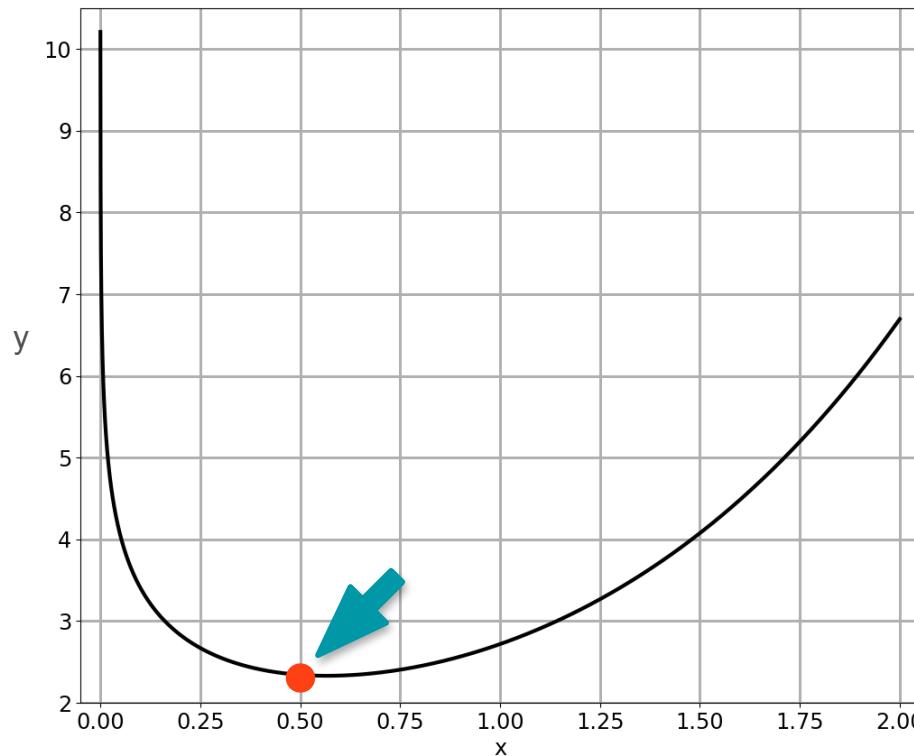


# Method 1: Try Both Directions

Is there any  
other way?



Repeat!





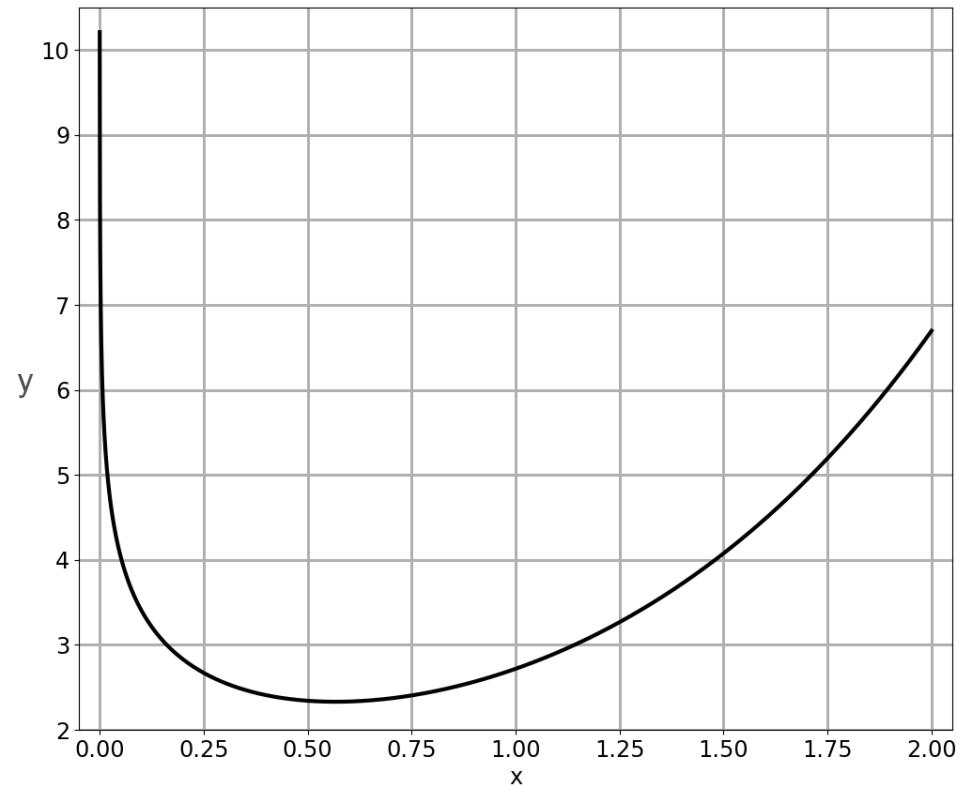
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## Gradients and Gradient Descent

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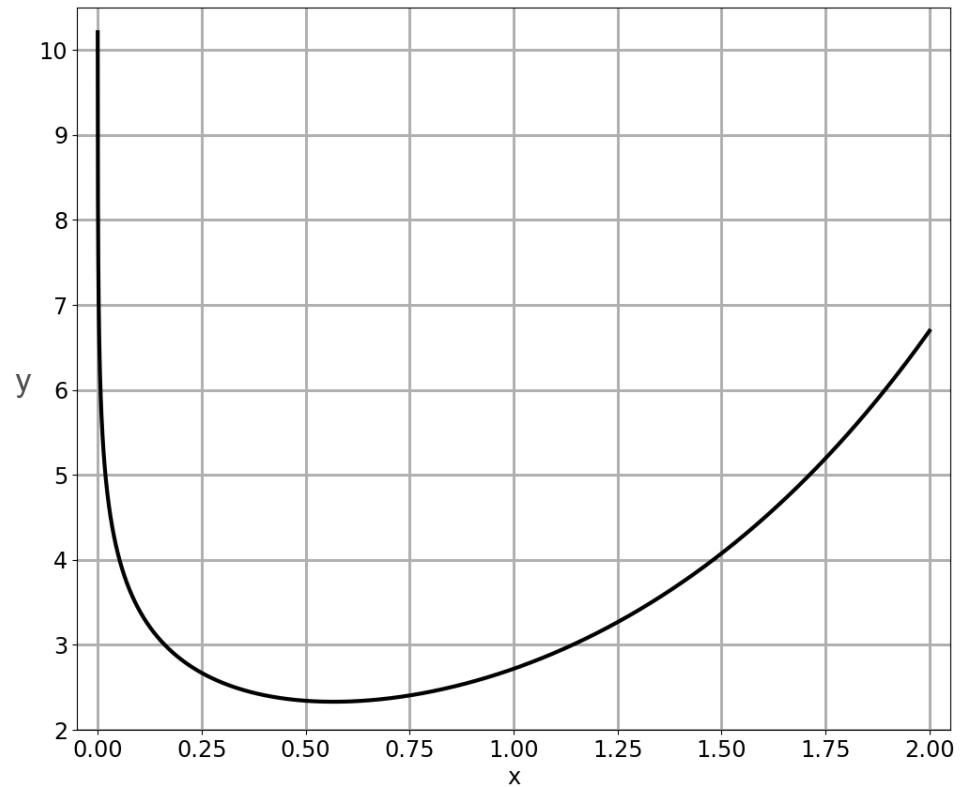
**Optimization using Gradient  
Descent in one variable -  
Part 2**

## Method 2: Be Clever



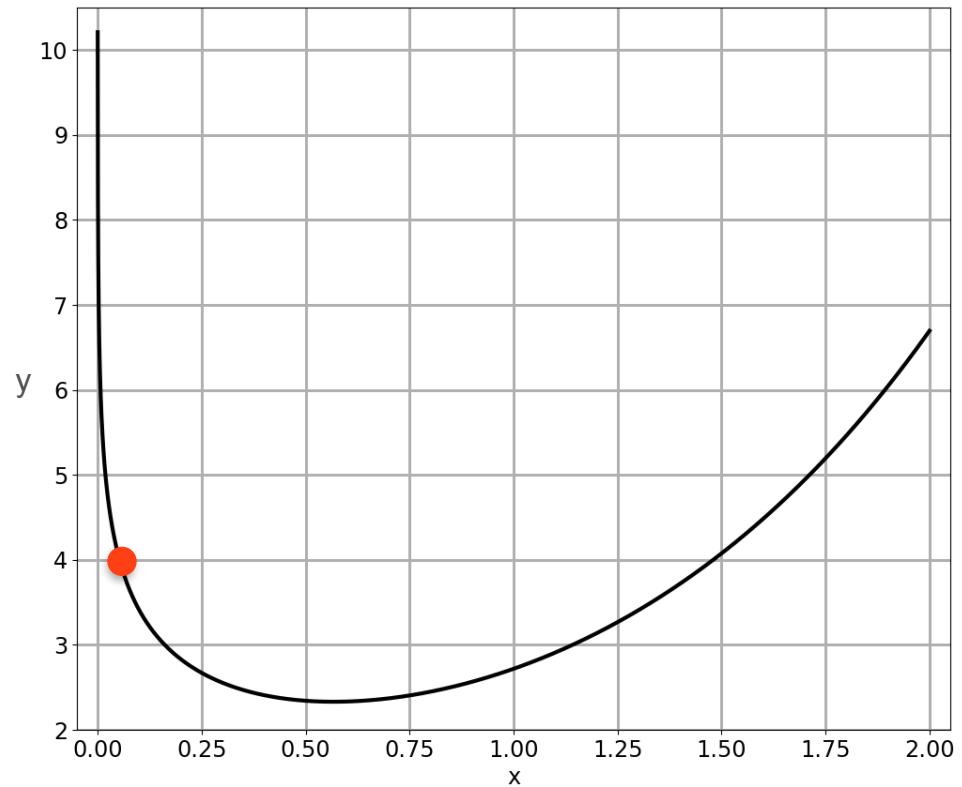
## Method 2: Be Clever

Try something  
smarter...



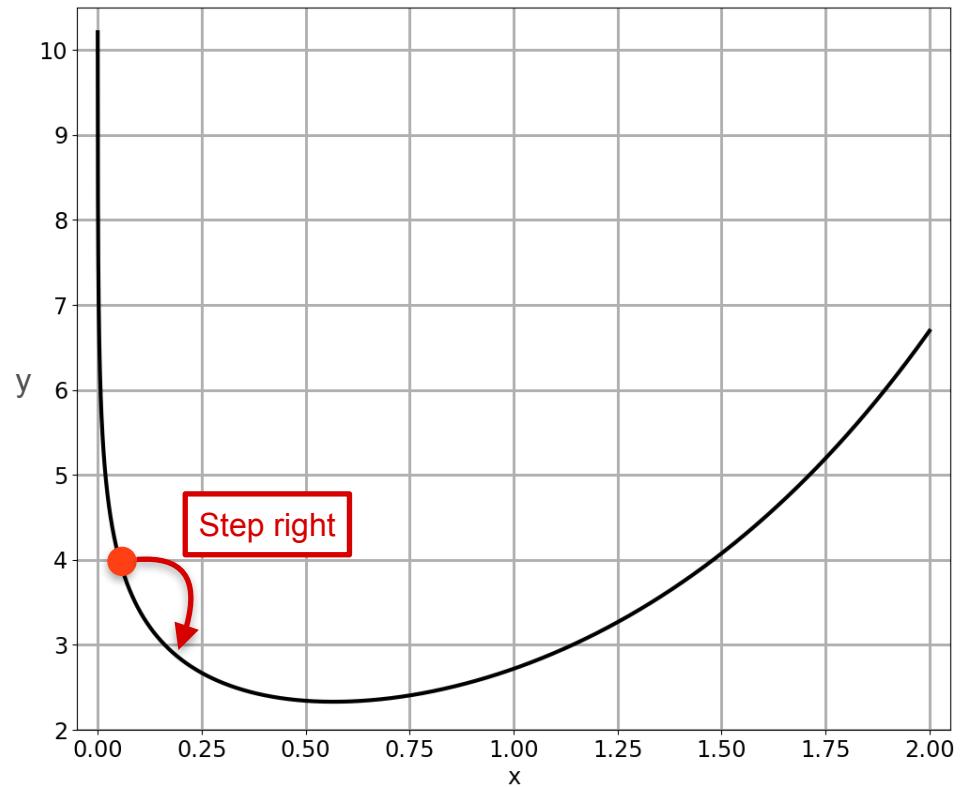
## Method 2: Be Clever

Try something  
smarter...



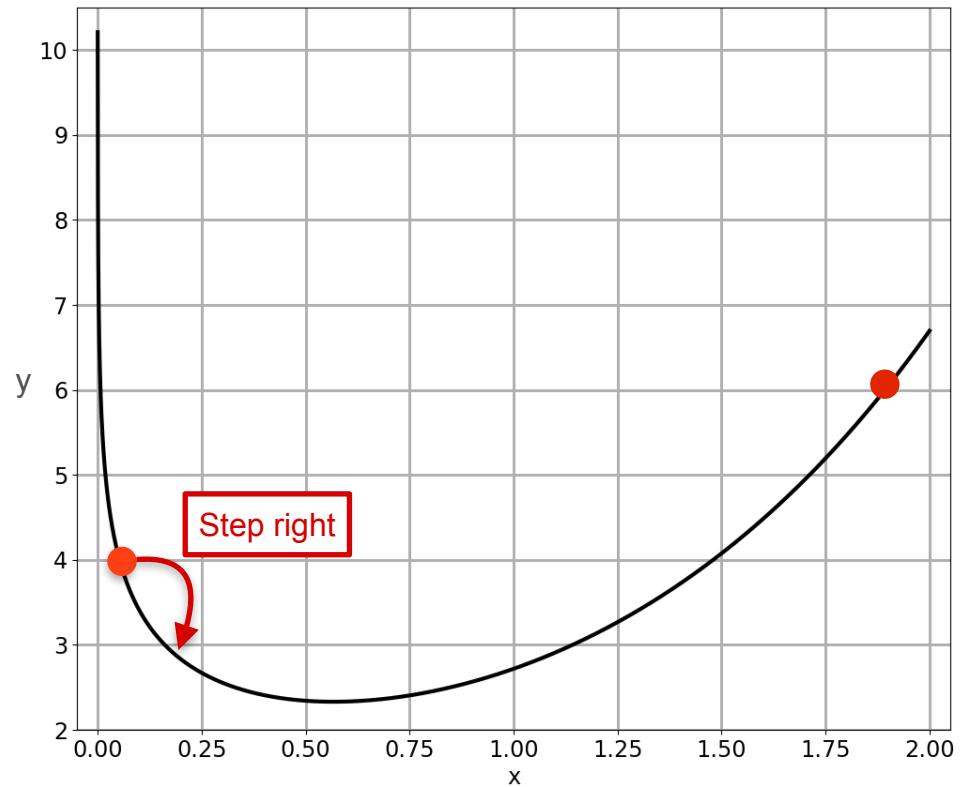
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Try something  
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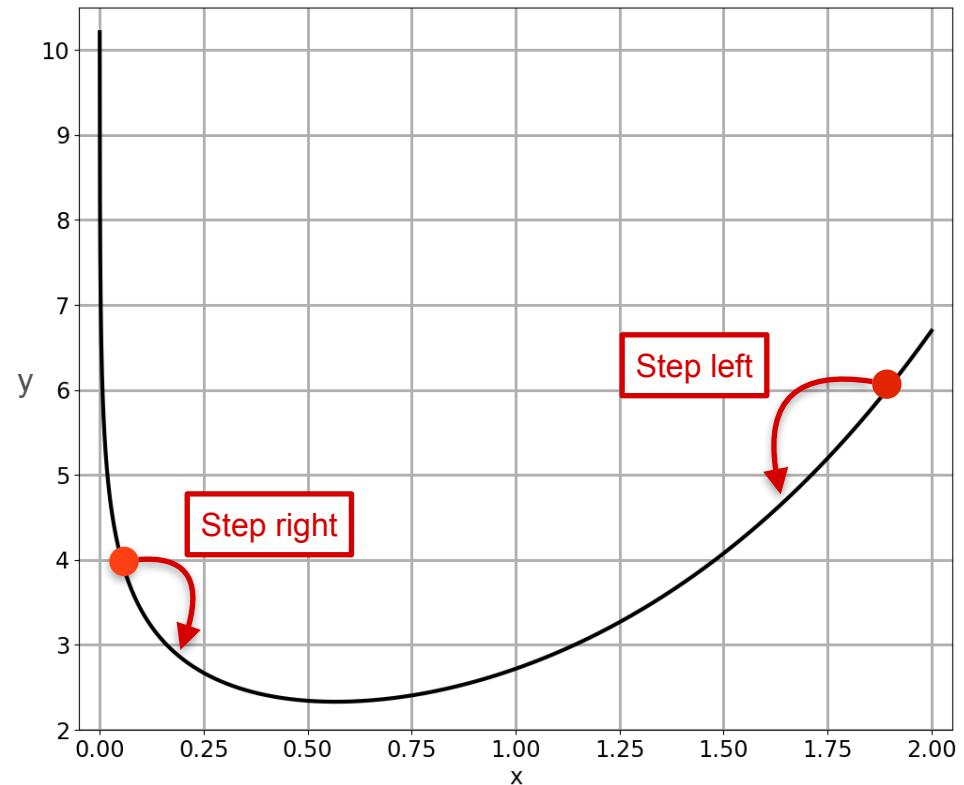
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Try something  
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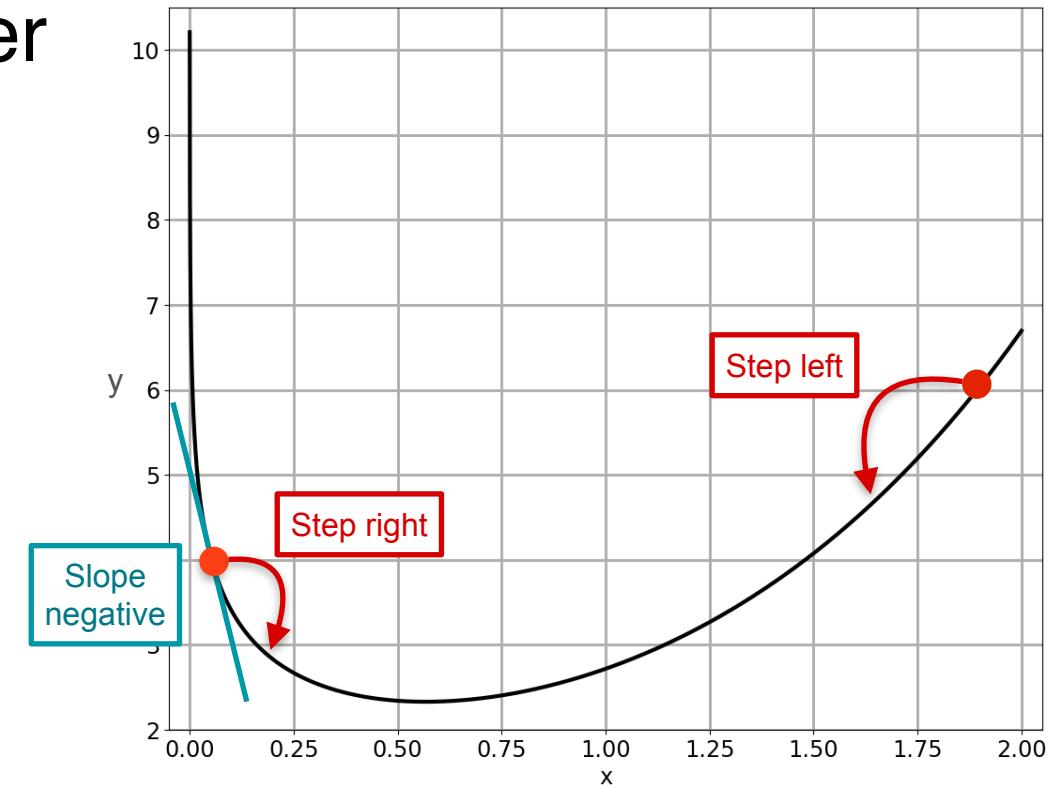
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Try something  
smarter...



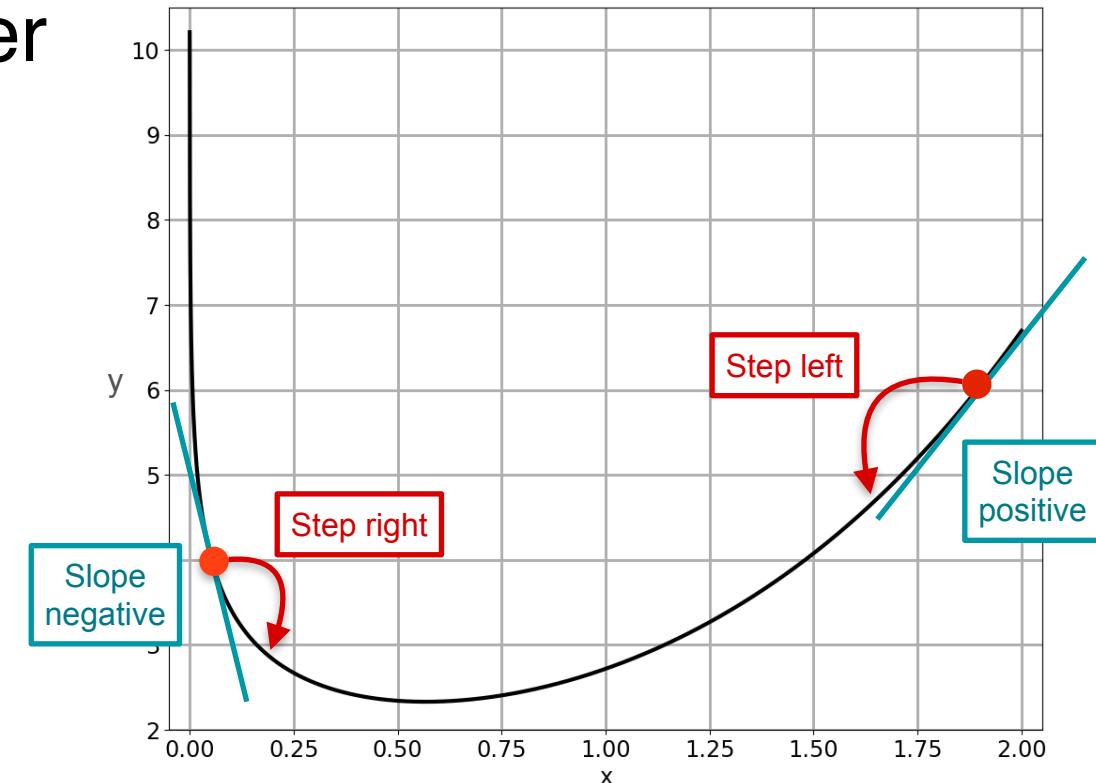
## Method 2: Be Clever

Try something  
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## Method 2: Be Clever

Try something  
smarter...

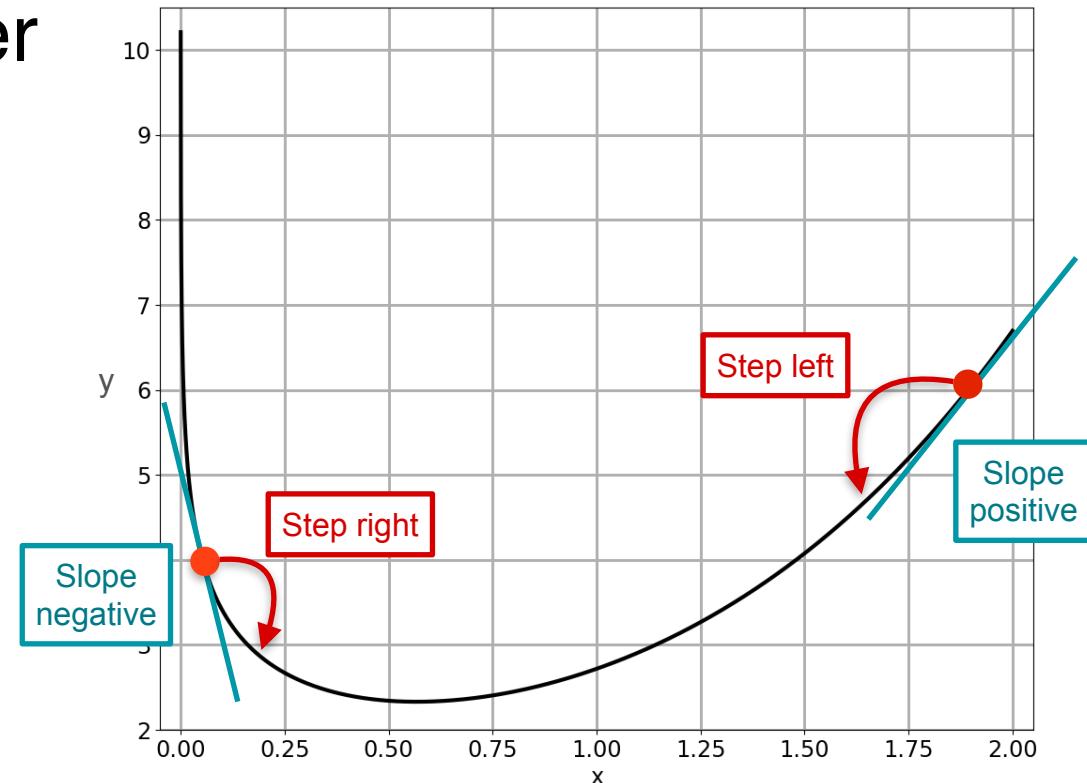


## Method 2: Be Clever

Try something  
smarter...



new point

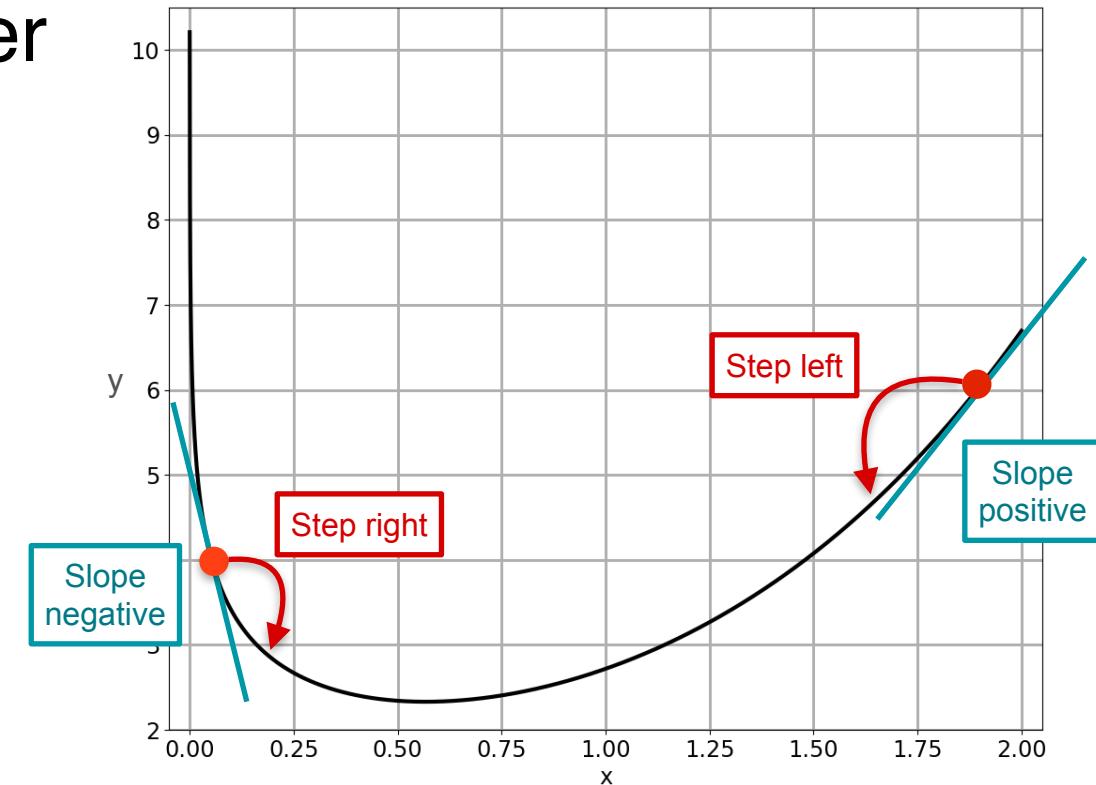


## Method 2: Be Clever

Try something  
smarter...



new point = old point

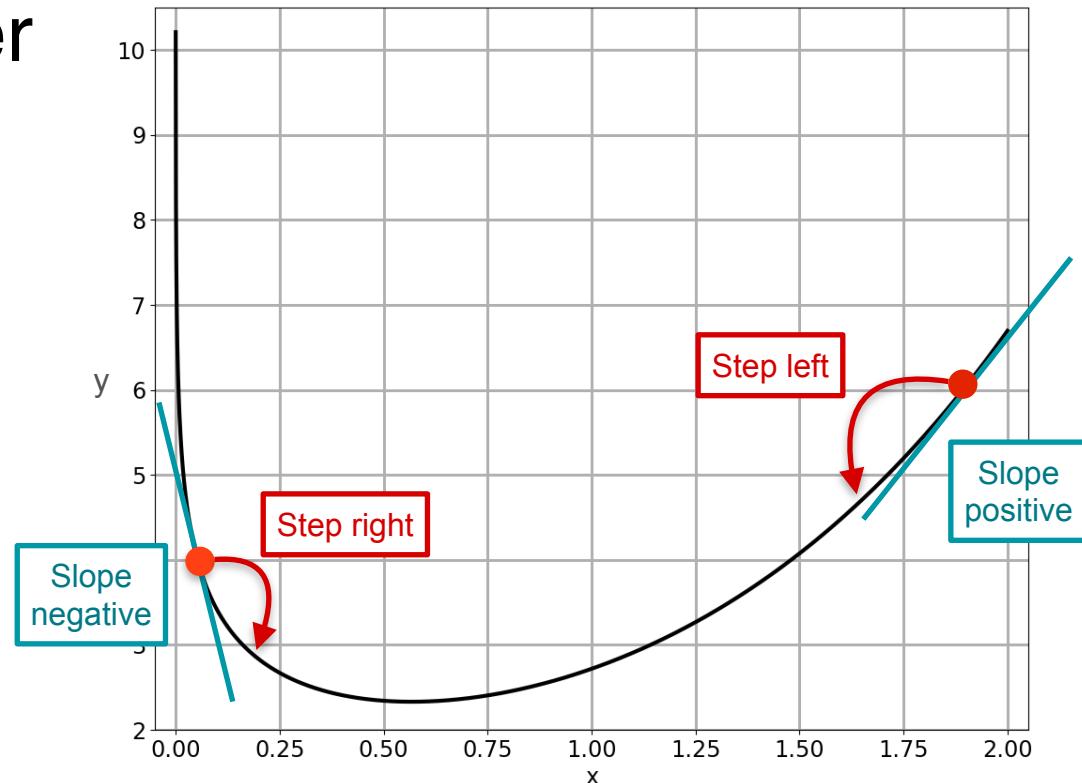


## Method 2: Be Clever

Try something  
smarter...



$\text{new point} = \text{old point} - \text{slope}$



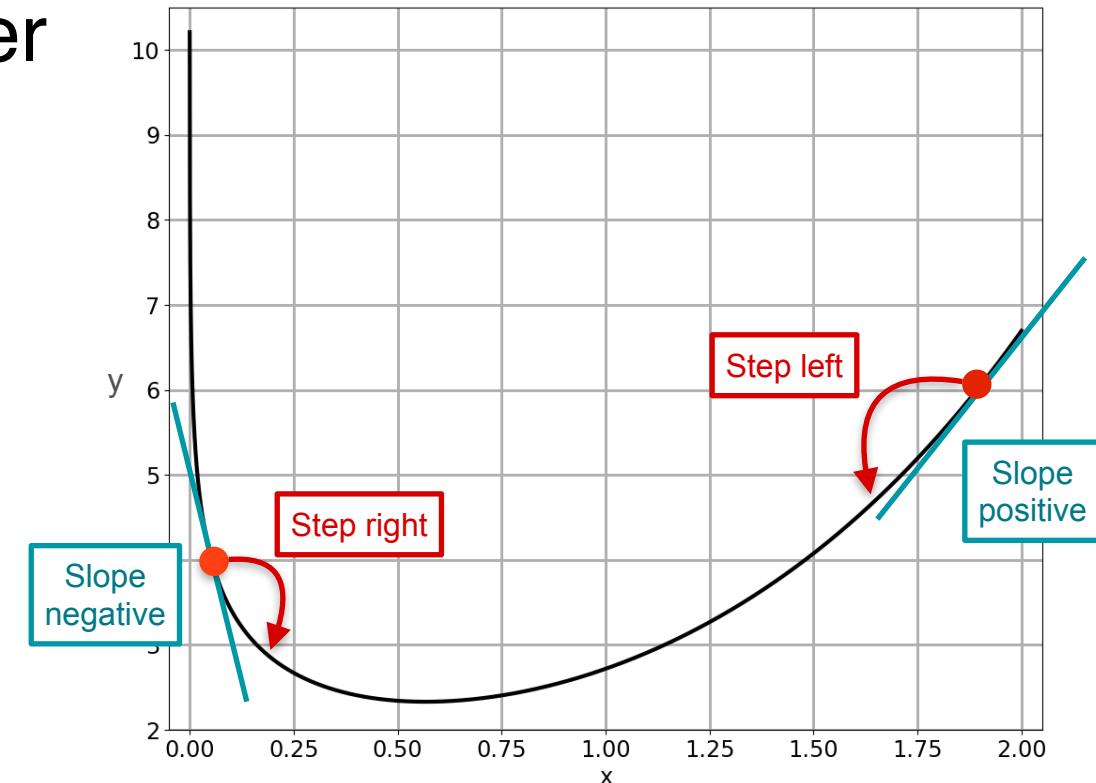
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$x_1$



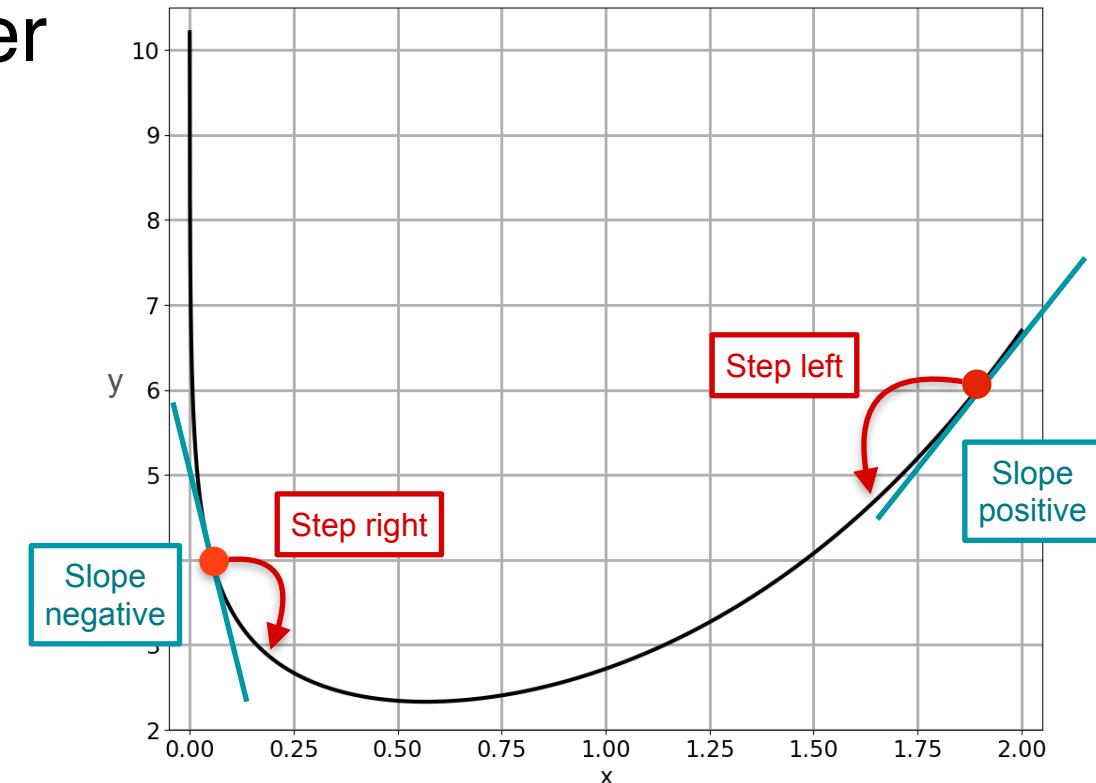
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0$$



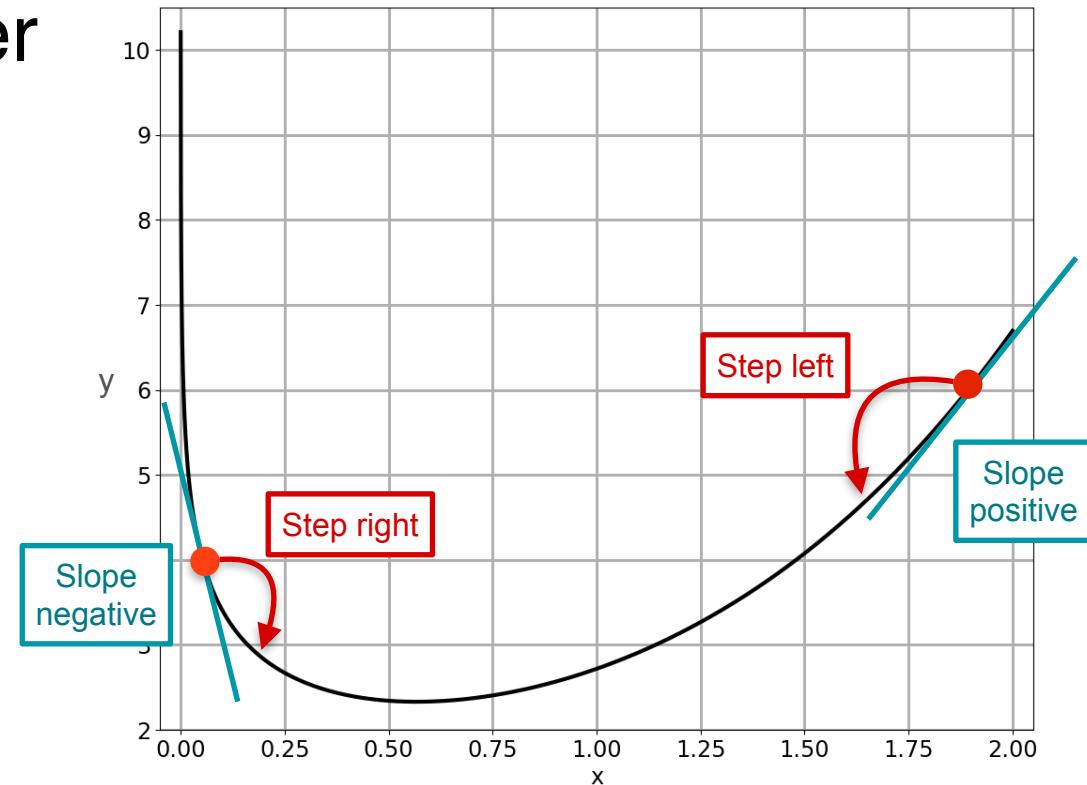
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f'(x_0)$$



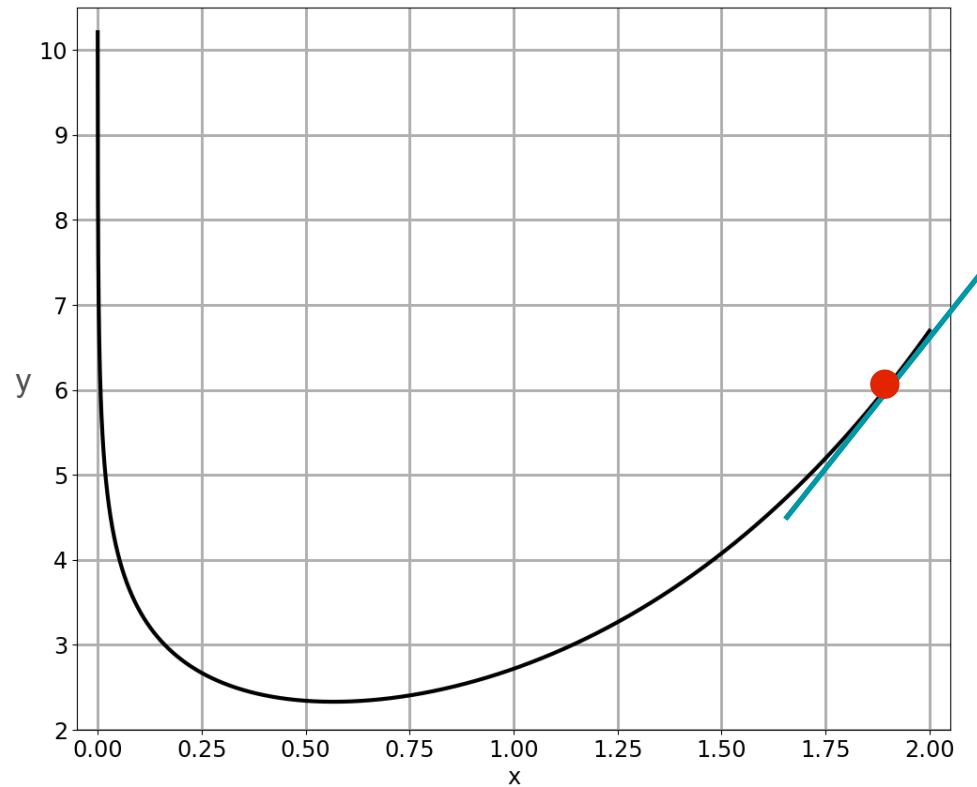
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$



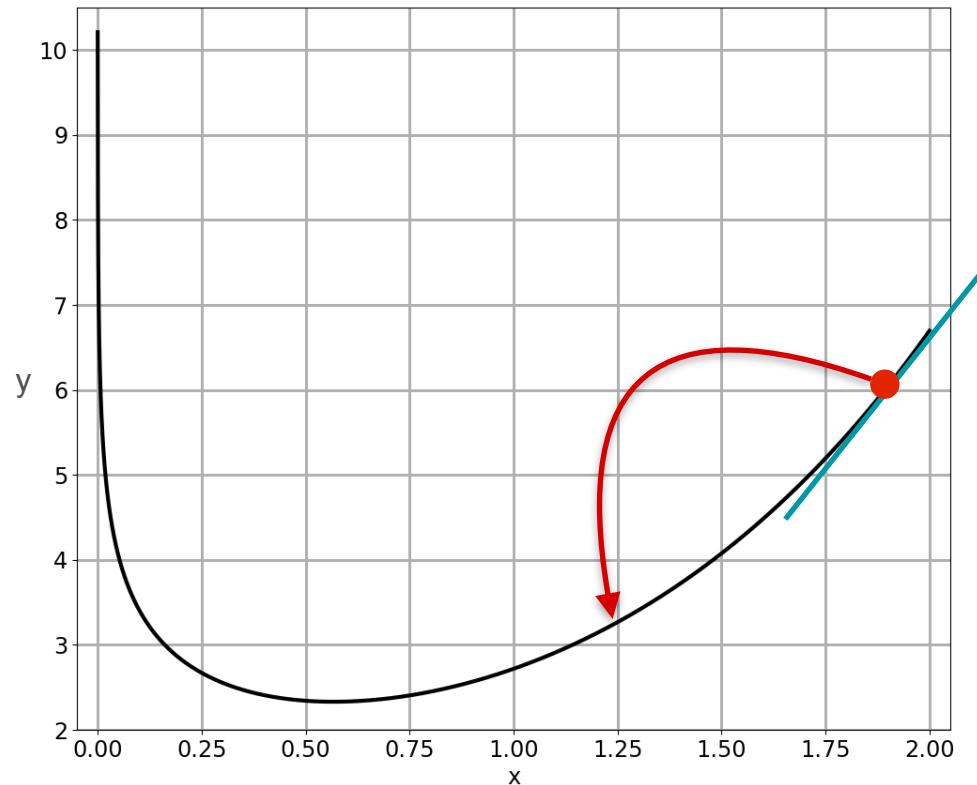
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$



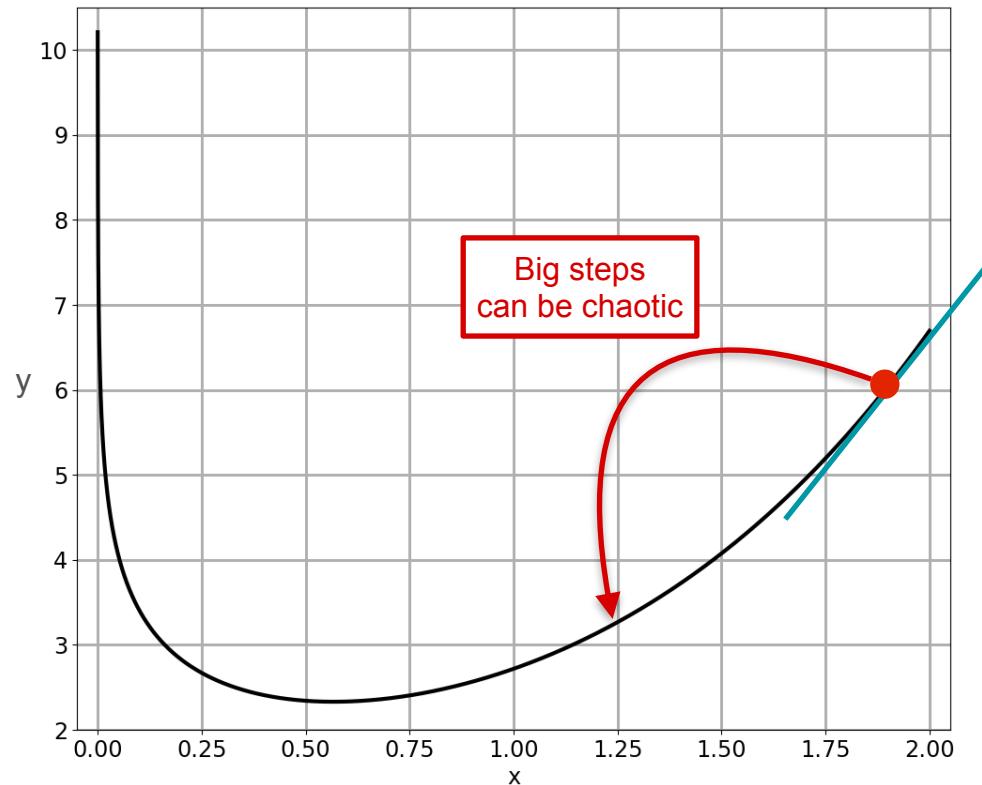
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$



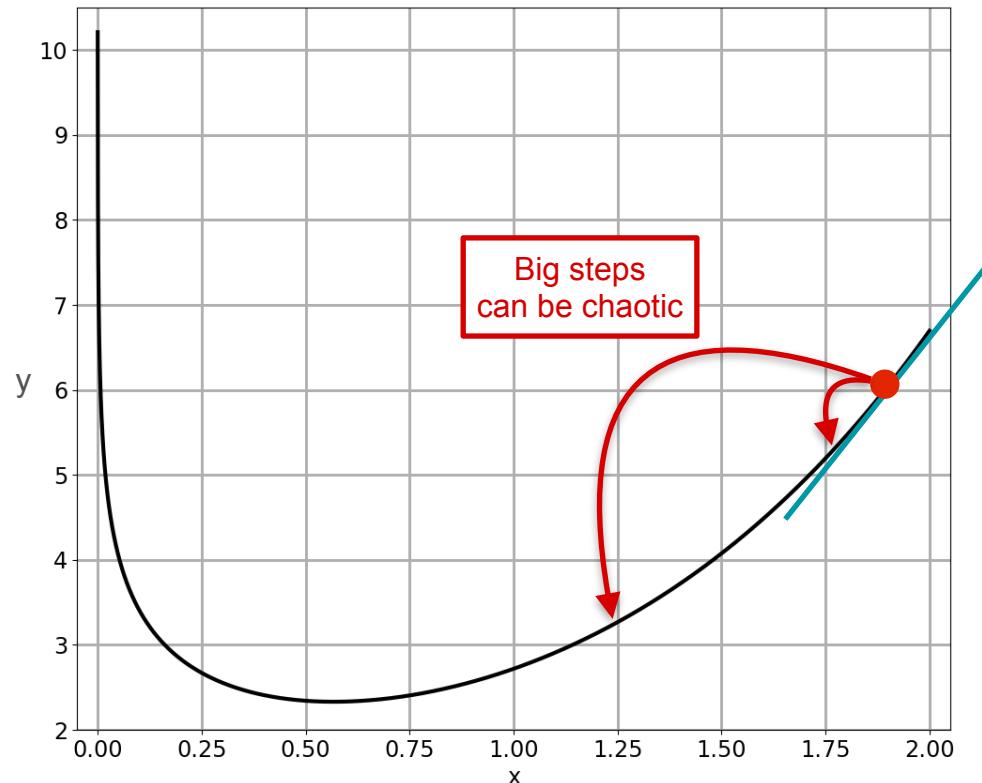
## Method 2: Be Clever

Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$



## Method 2: Be Clever

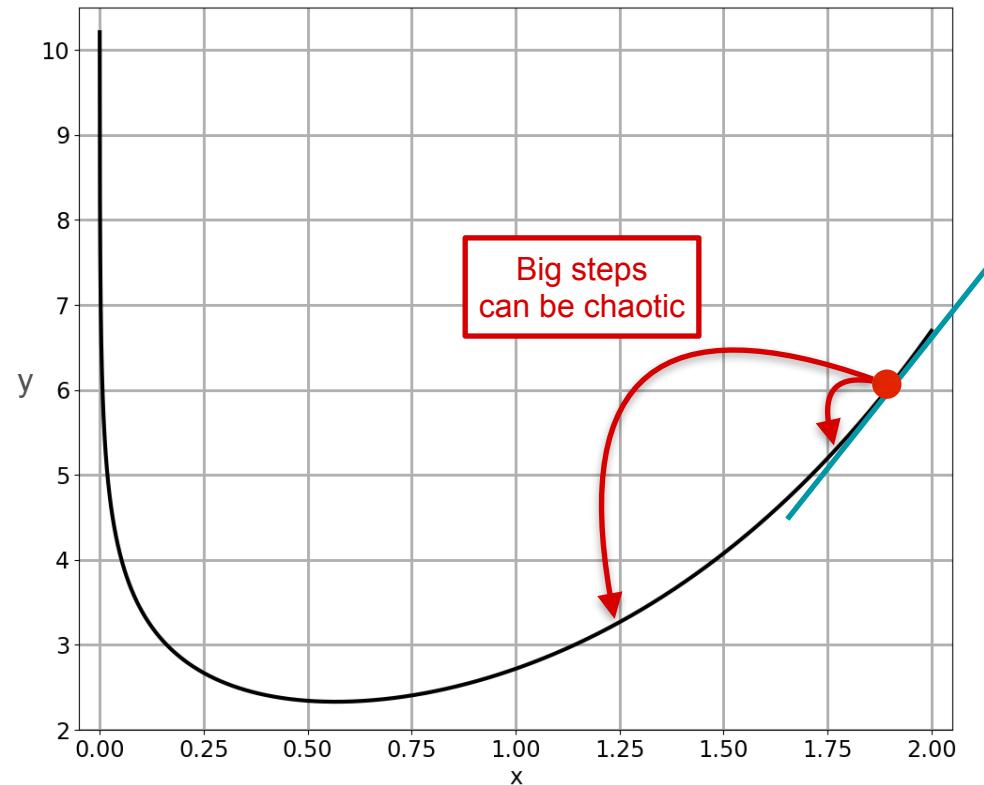
Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$

$$x_1 = x_0 - 0.01 f(x_0)$$



## Method 2: Be Clever

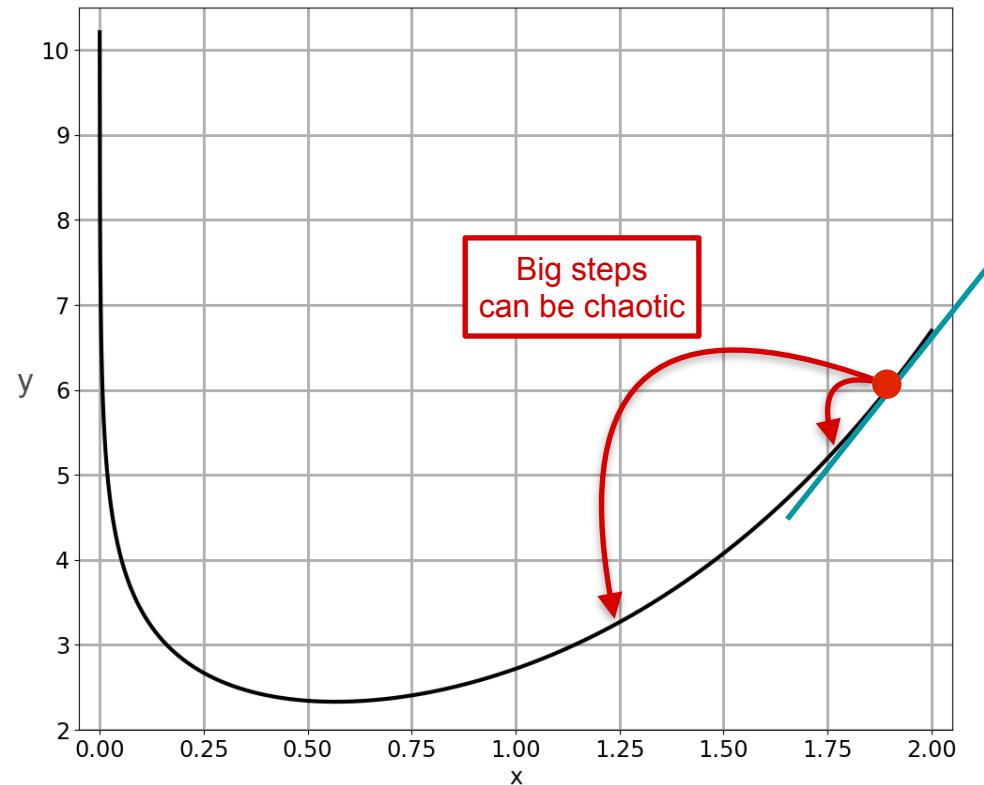
Try something  
smarter...



new point = old point - slope

$$x_1 = x_0 - f(x_0)$$

$$x_1 = x_0 - f(x_0)$$



## Method 2: Be Clever

Try something  
smarter...

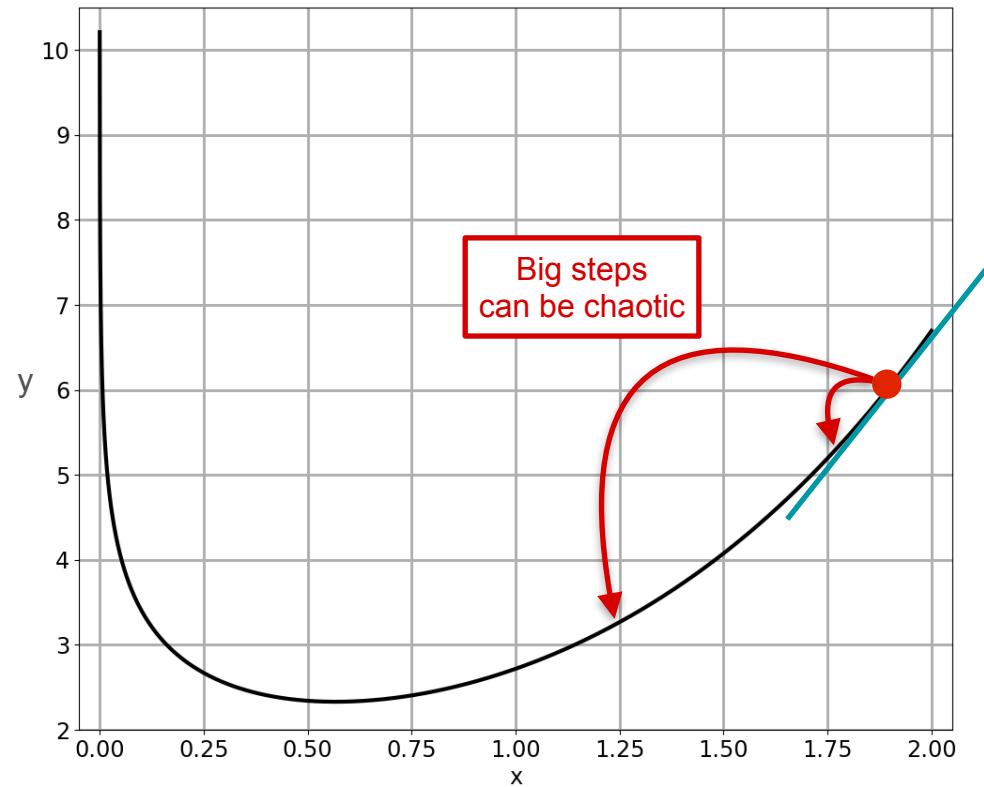


new point = old point - slope

$$x_1 = x_0 - f(x_0)$$

$$x_1 = x_0 - f(x_0)$$

↑  
Learning rate



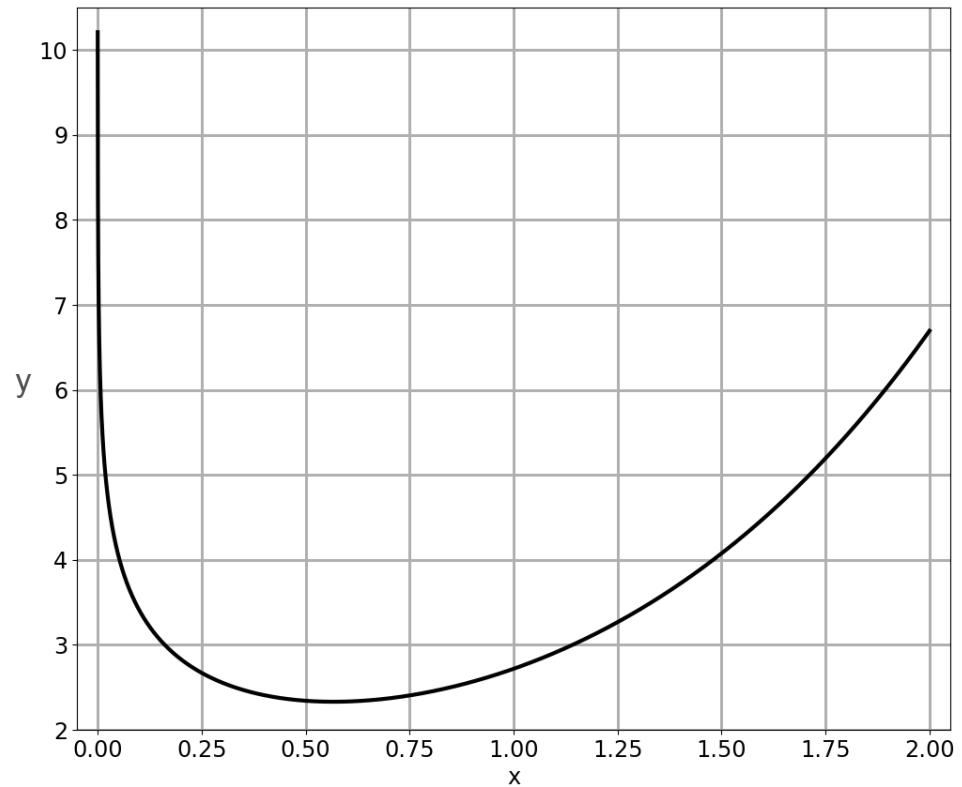
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



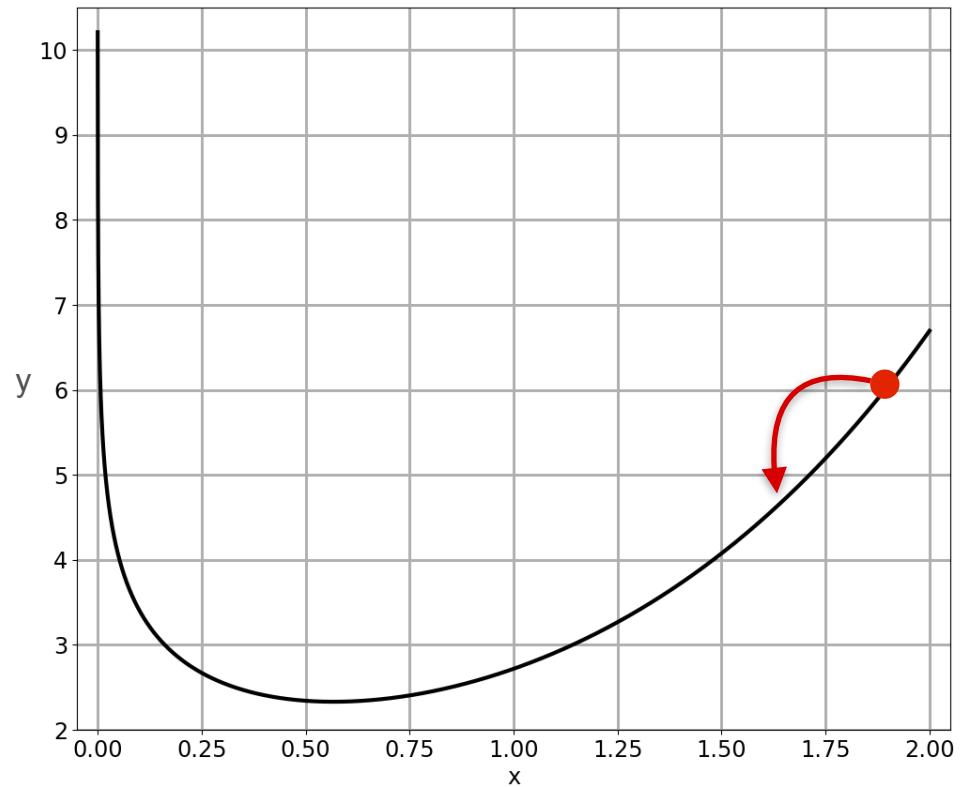
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



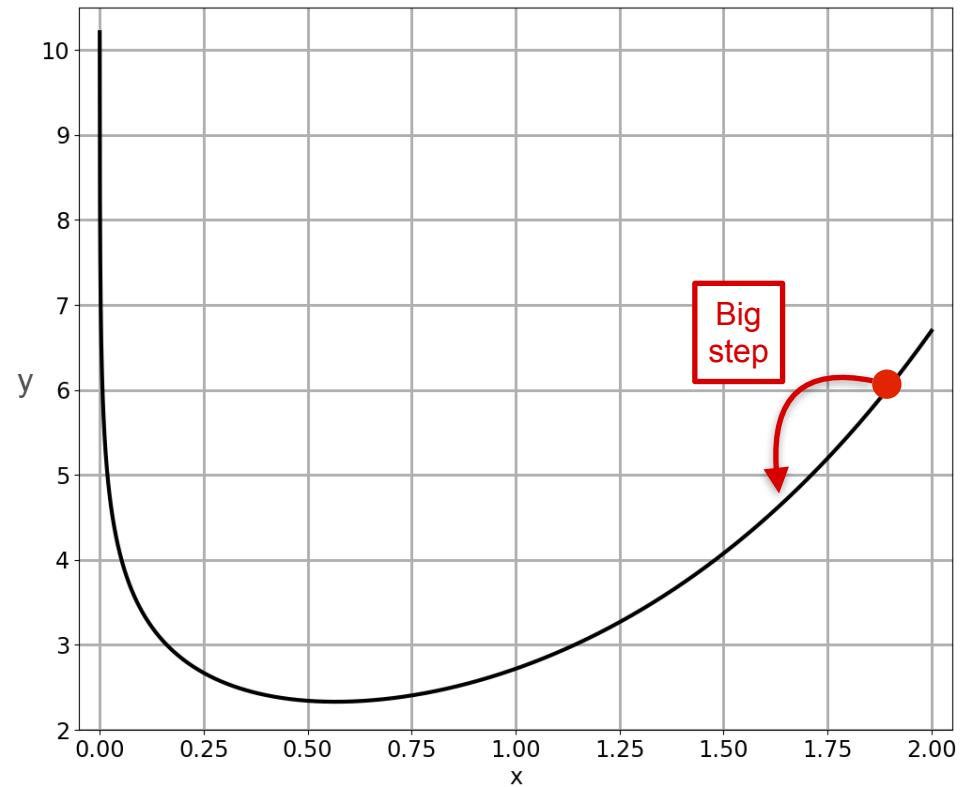
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



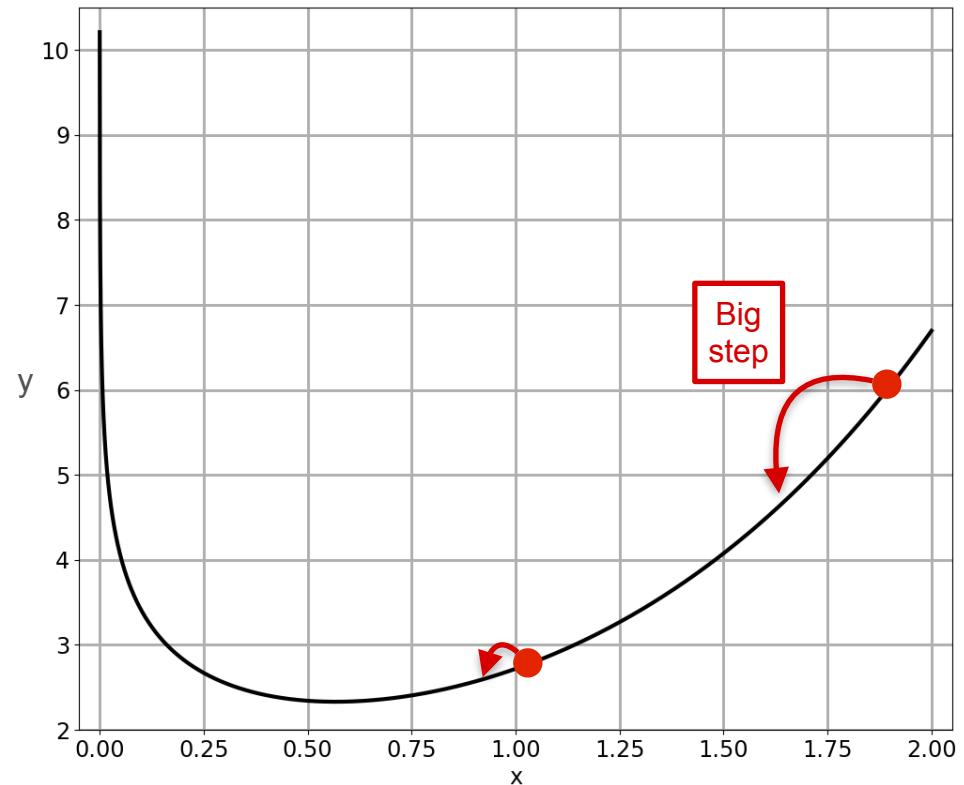
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



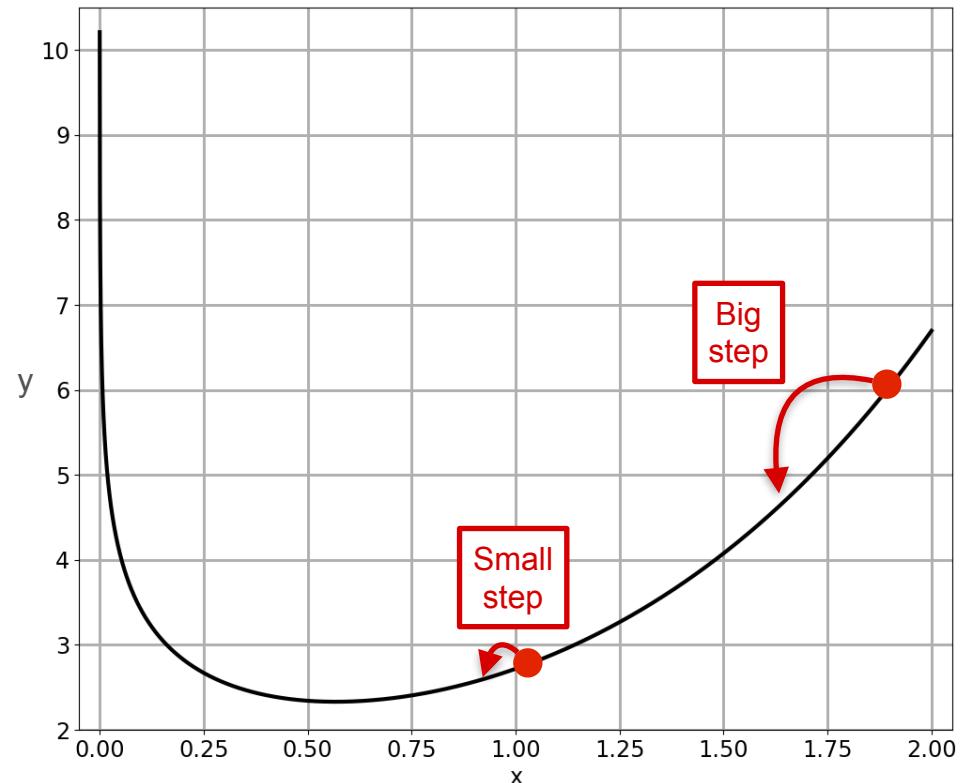
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



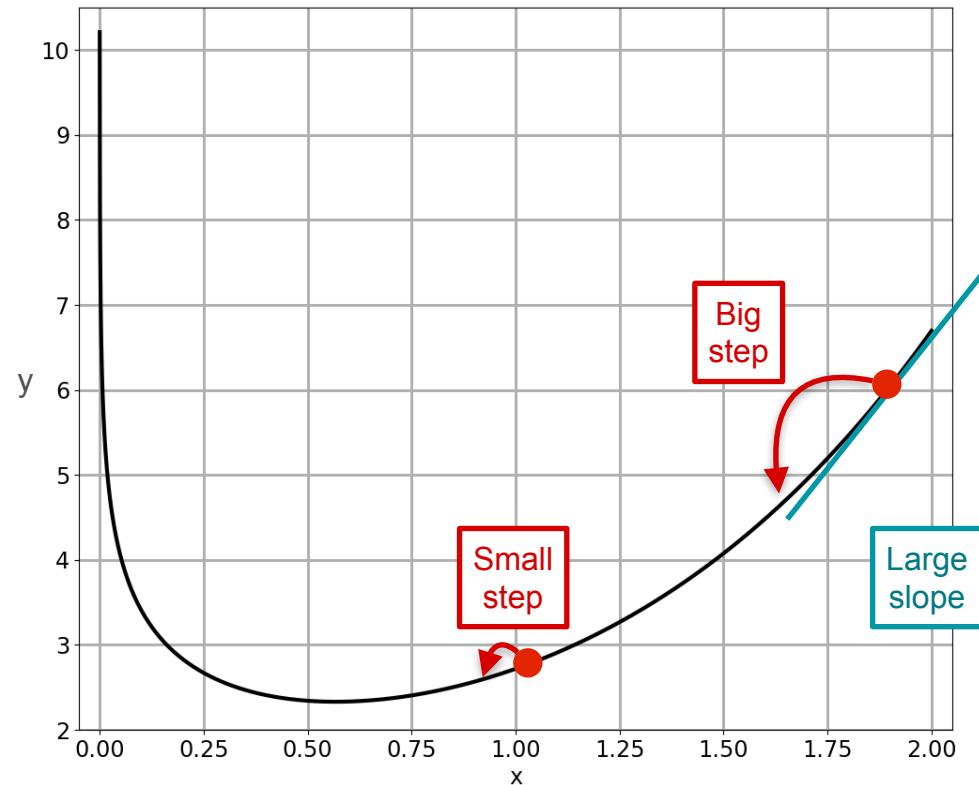
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



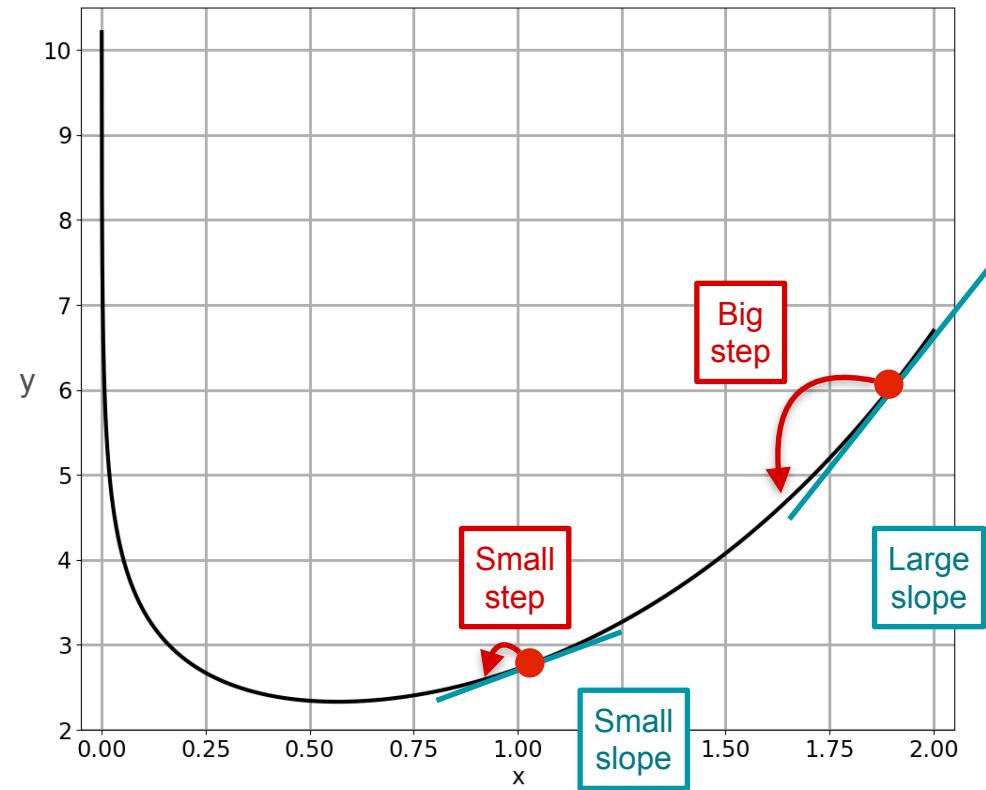
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0$$

$$f(x_0)$$



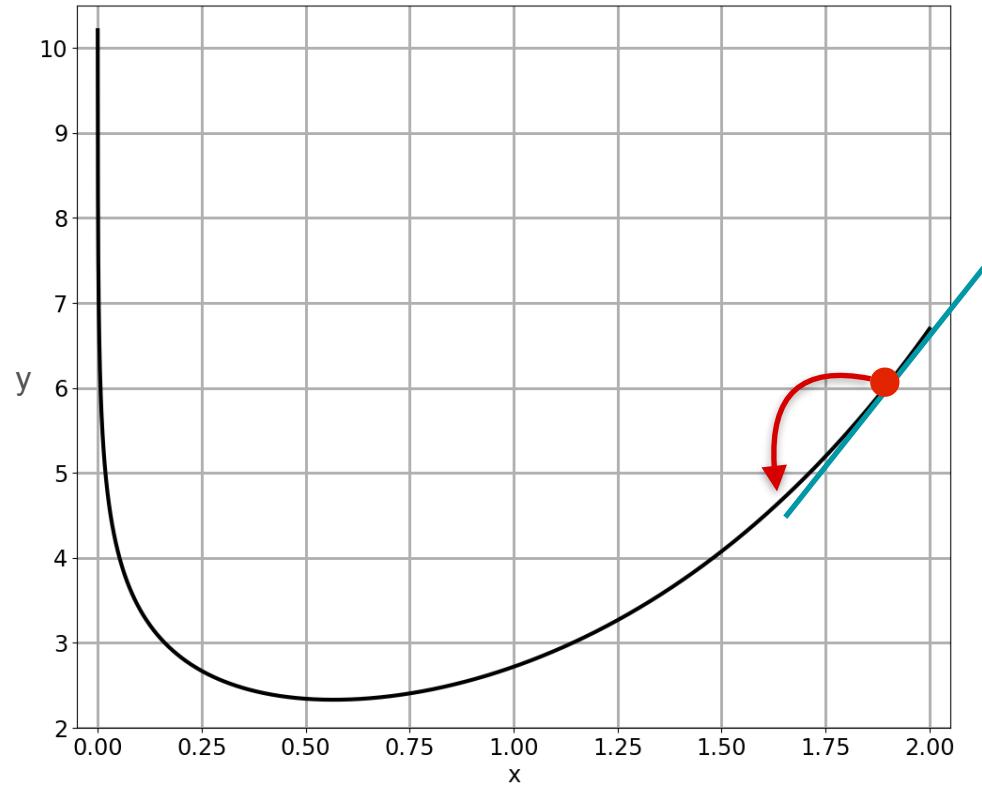
## Method 2: Be Clever

Try something  
smarter...



$$x_1 = x_0 - f(x_0)$$

Gradient descent



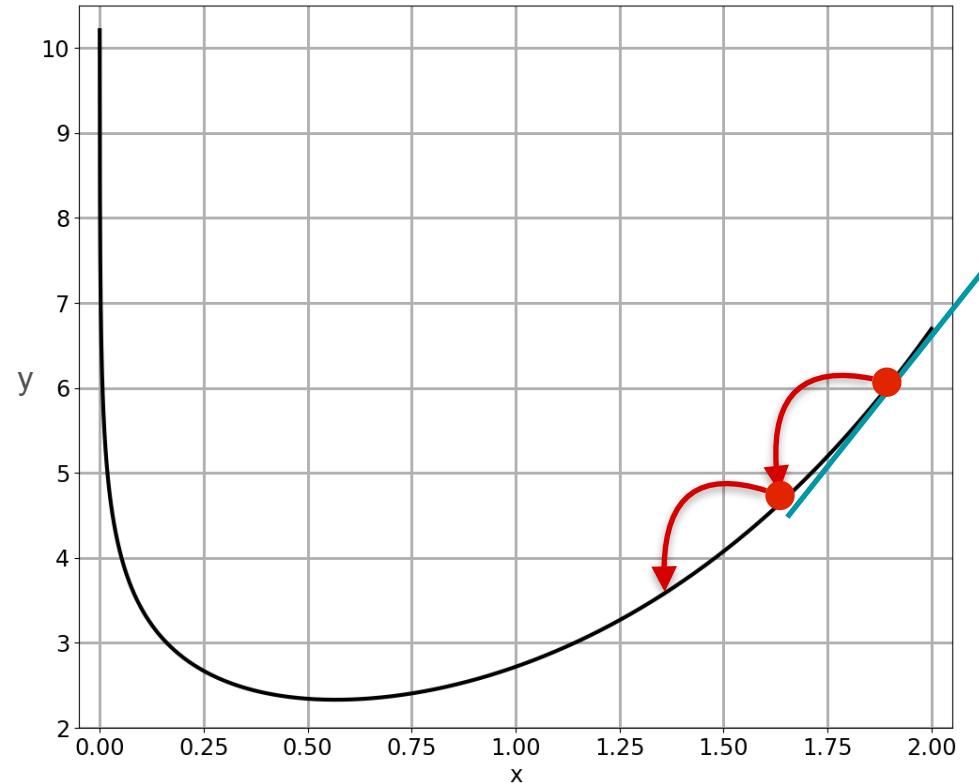
## Method 2: Be Clever

Try something  
smarter...



$$x_2 = x_1 \quad f(x_1)$$

Gradient descent



## Method 2: Be Clever

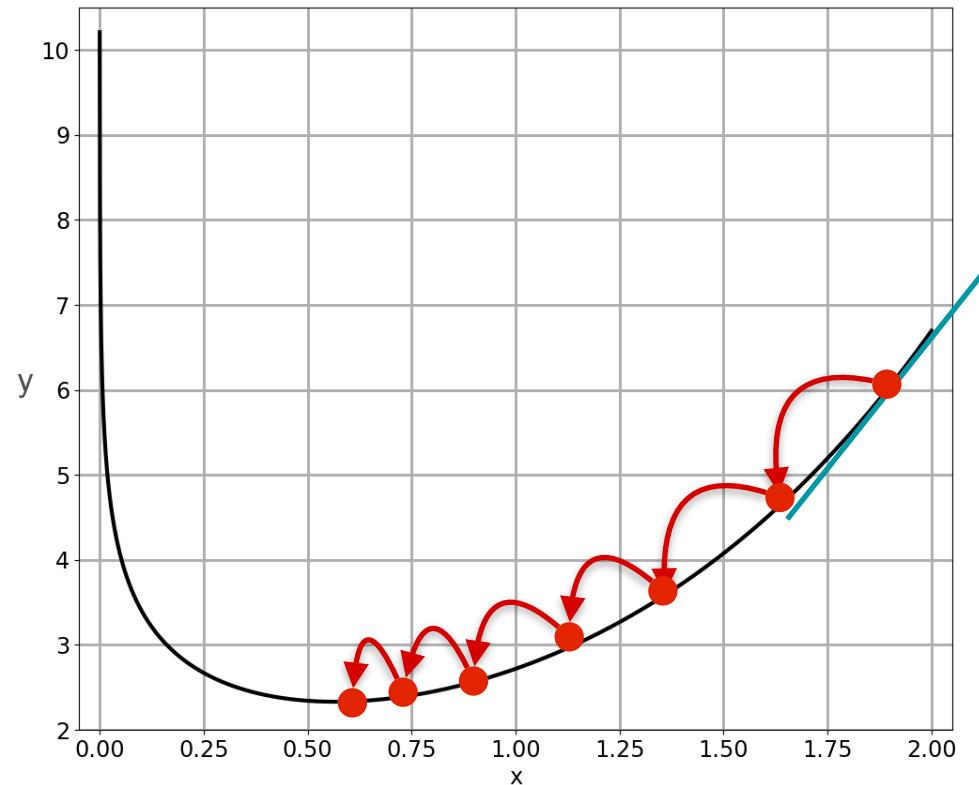
Try something  
smarter...



$$x_{20} = x_{19}$$

$$f(x_{19})$$

Gradient descent



# Gradient Descent

Function:  $f(x)$

Goal: find minimum of  $f(x)$

Step 1:

Define a learning rate

Choose a starting point  $x_0$

Step 2:

Update:  $x_k = x_{k-1} - \alpha f'(x_{k-1})$

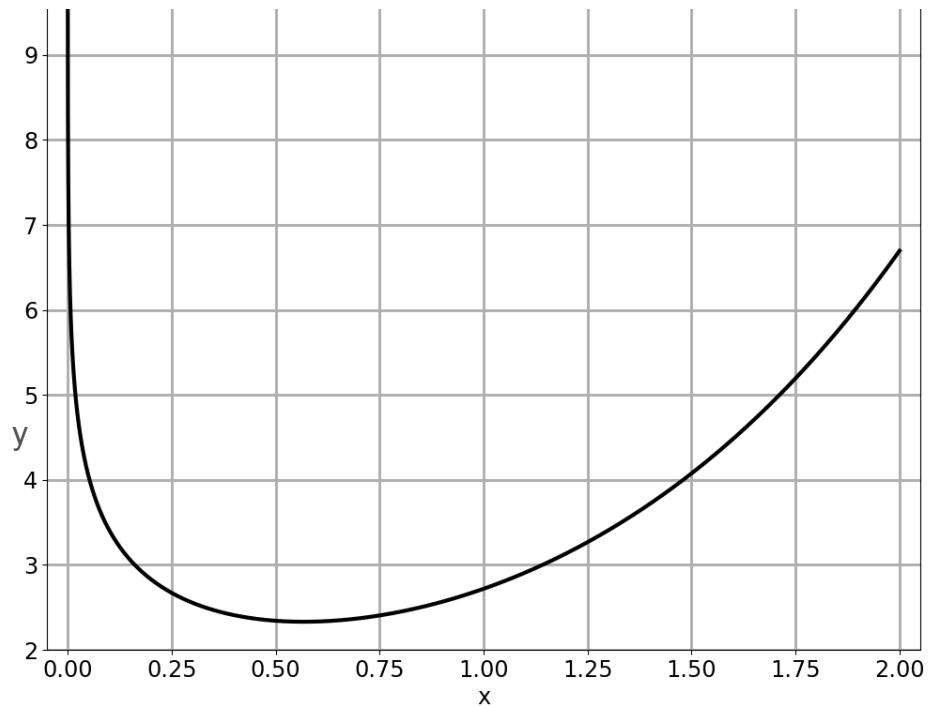
Step 3:

Repeat Step 2 until you are close enough to  
the true minimum  $x^*$

# Gradient Descent

$$f(x) = e^x - \log(x)$$

$$f'(x) = e^x - \frac{1}{x}$$



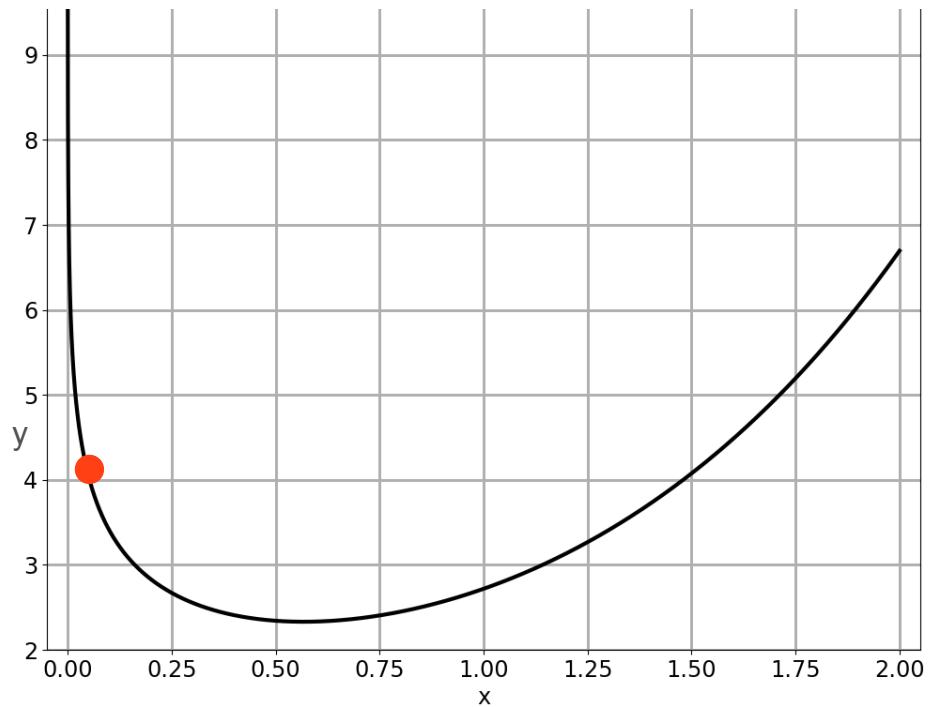
# Gradient Descent

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x}$$

Start:  $x = 0.05$

Rate:  $= 0.005$



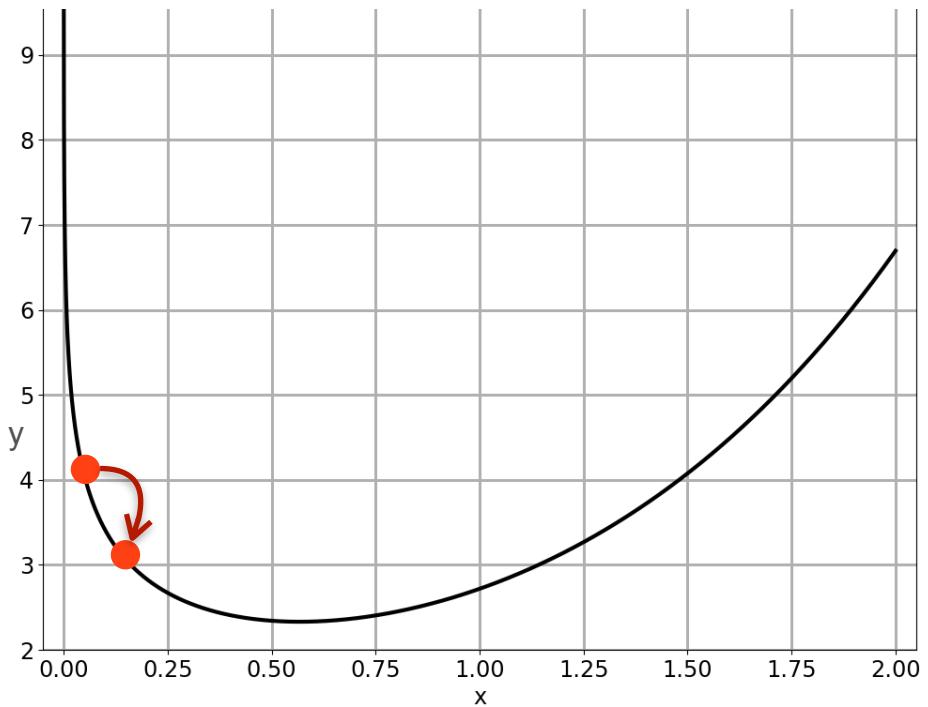
# Gradient Descent

$$f(x) = e^x \log(x) \quad f'(x) = e^x \frac{1}{x}$$

Start:  $x = 0.05$       Rate:  $= 0.005$

Find:  
 $f(0.05) = 18.9$

Move by  $0.005f(0.05)$        $x = 0.1447$



# Gradient Descent

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x}$$

Start:  $x = 0.05$       Rate:  $= 0.005$

Find:

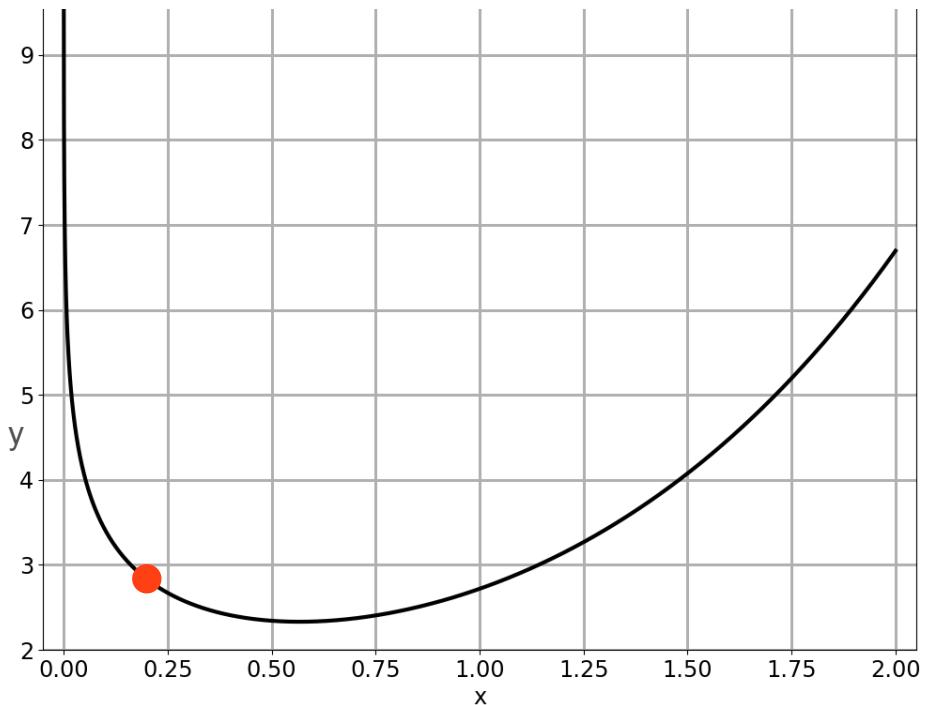
$$f(0.05) = 18.9$$

Move by  $0.005f(0.05)$        $x = 0.1447$

Find:

$$f(0.1447) = 5.7552$$

Move by  $0.005f(0.05)$        $x = 0.1735$



# Gradient Descent

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x}$$

Start:  $x = 0.05$       Rate:  $= 0.005$

Find:

$$f(0.05) = 18.9$$

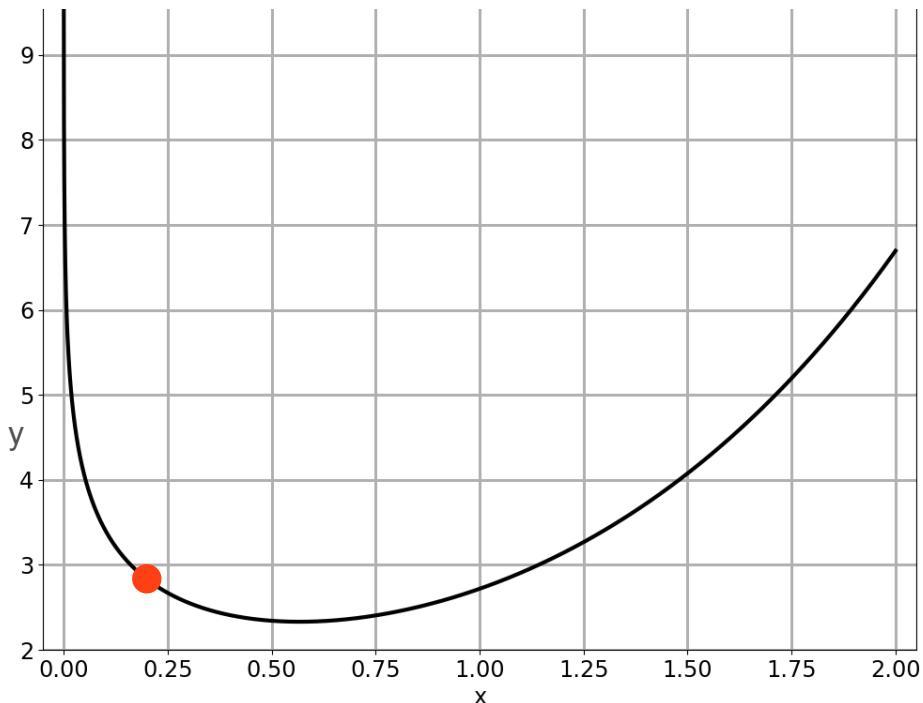
Move by  $0.005f(0.05)$        $x = 0.1447$

Find:

$$f(0.1447) = 5.7552$$

Move by  $0.005f(0.05)$        $x = 0.1735$

Repeat!



# Gradient Descent

$$f(x) = e^x \log(x)$$

$$f'(x) = e^x - \frac{1}{x}$$

Start:  $x = 0.05$       Rate:  $= 0.005$

Find:

$$f(0.05) = 18.9$$

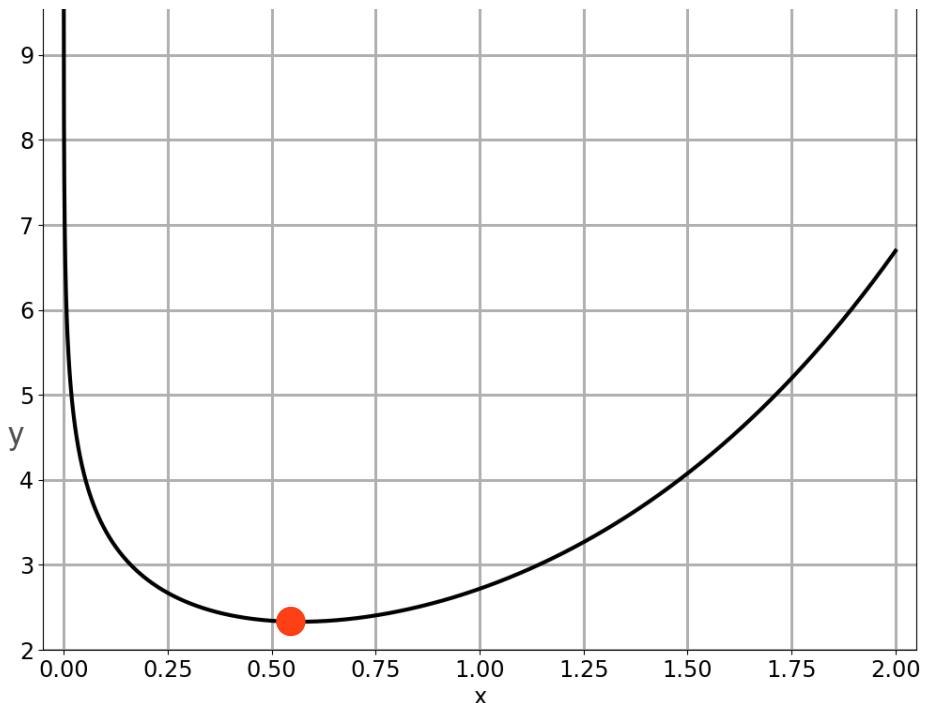
Move by  $0.005f(0.05)$        $x = 0.1447$

Find:

$$f(0.1447) = 5.7552$$

Move by  $0.005f(0.05)$        $x = 0.1735$

Repeat!





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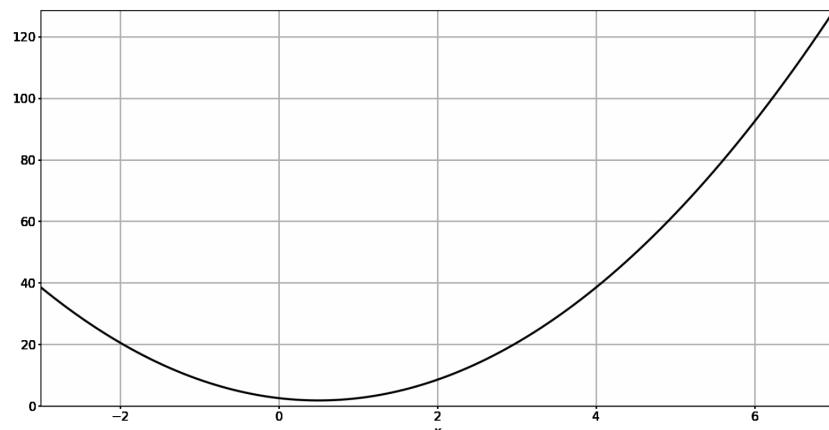
## Gradients and Gradient Descent

---

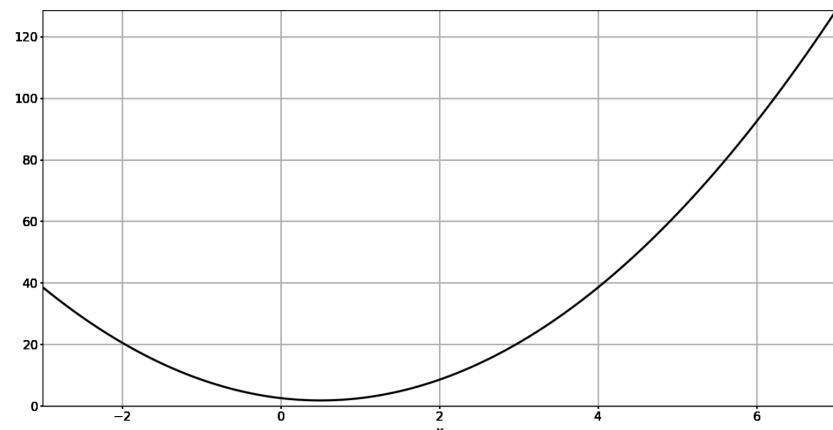
# Optimization using Gradient Descent in one variable - Part 3

# What Is a Good Learning Rate?

Too large

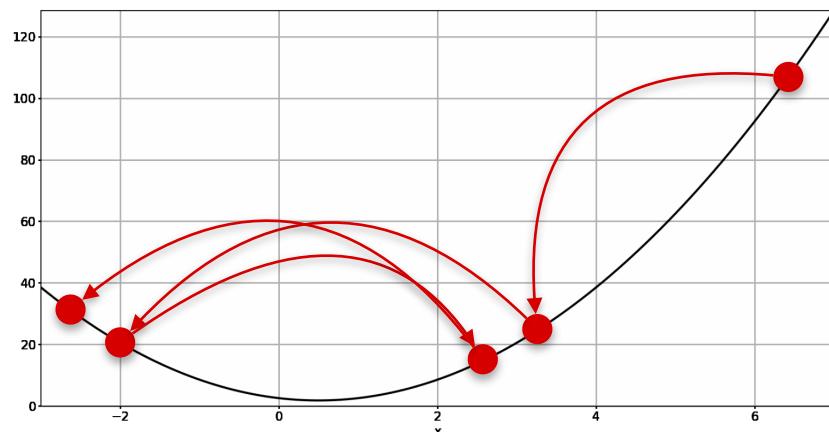


Too small

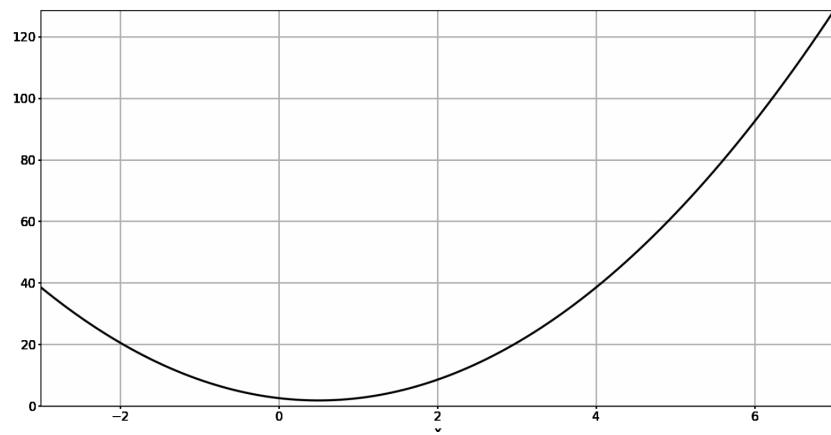


# What Is a Good Learning Rate?

Too large

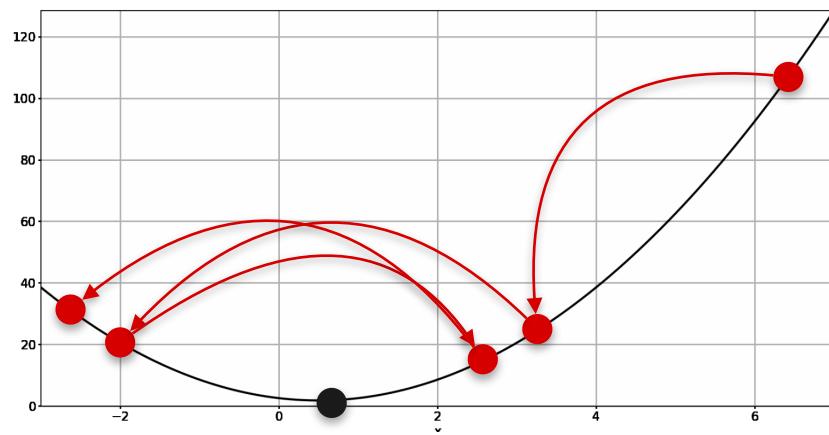


Too small

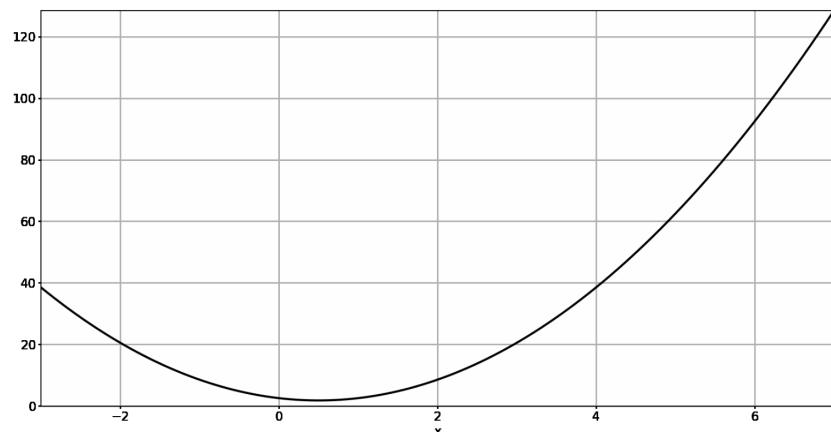


# What Is a Good Learning Rate?

Too large

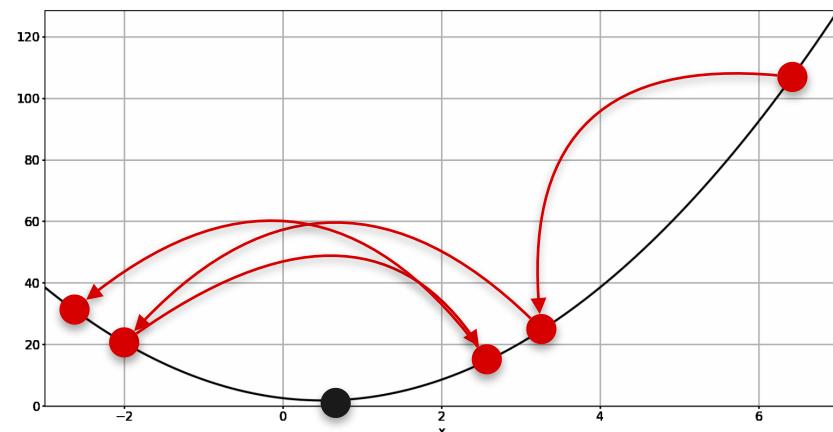


Too small

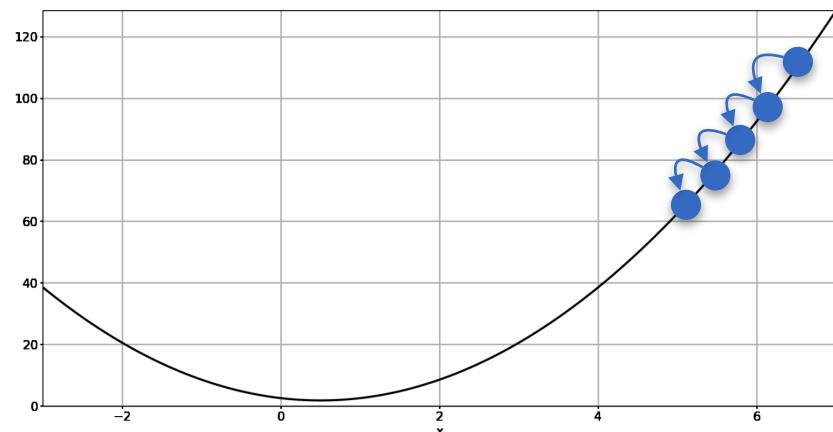


# What Is a Good Learning Rate?

Too large

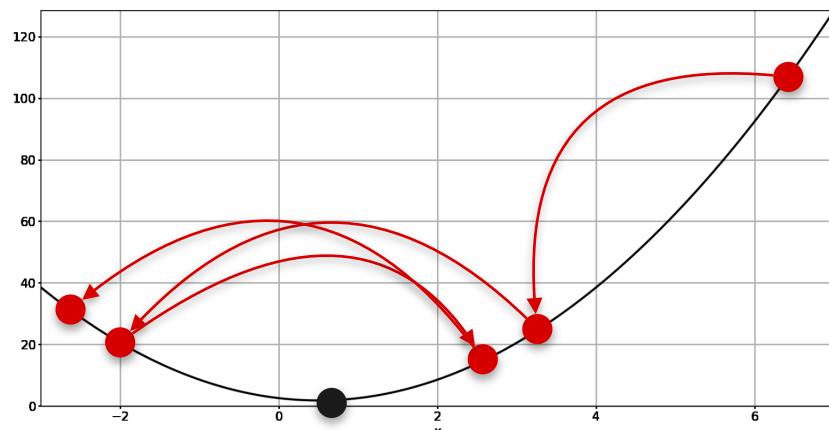


Too small

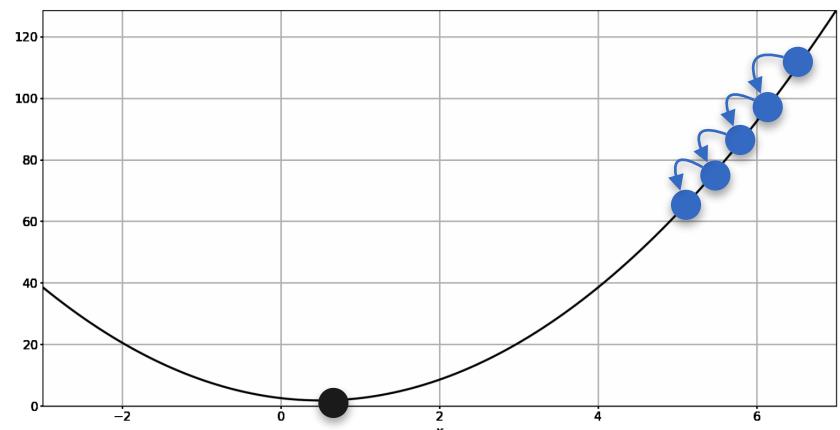


# What Is a Good Learning Rate?

Too large

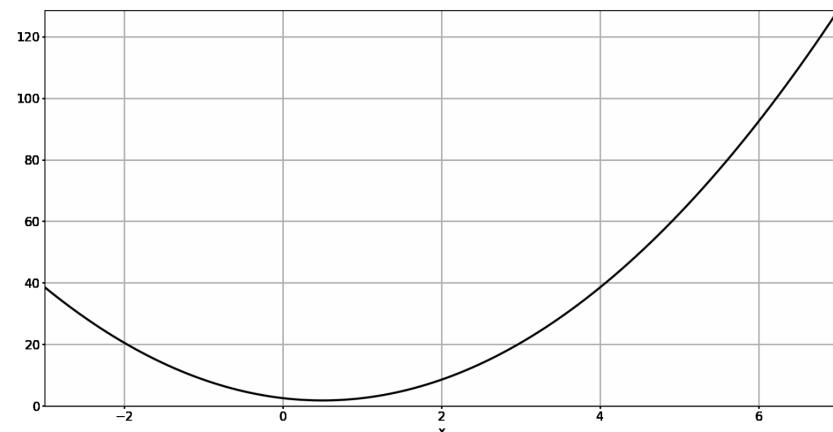


Too small



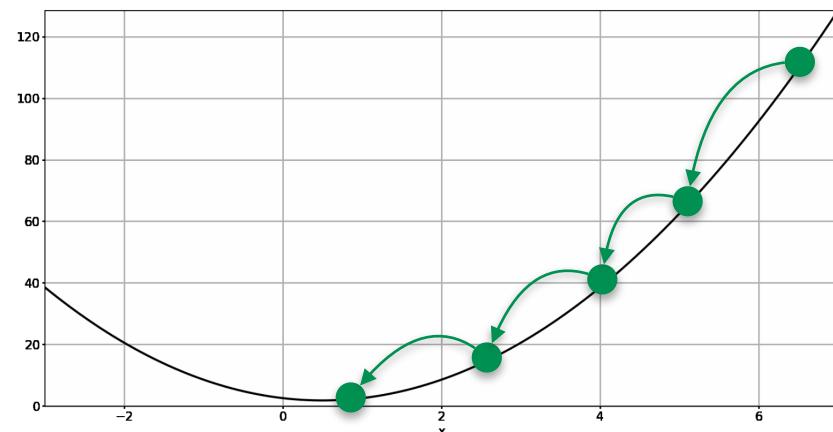
# What Is a Good Learning Rate?

Just right



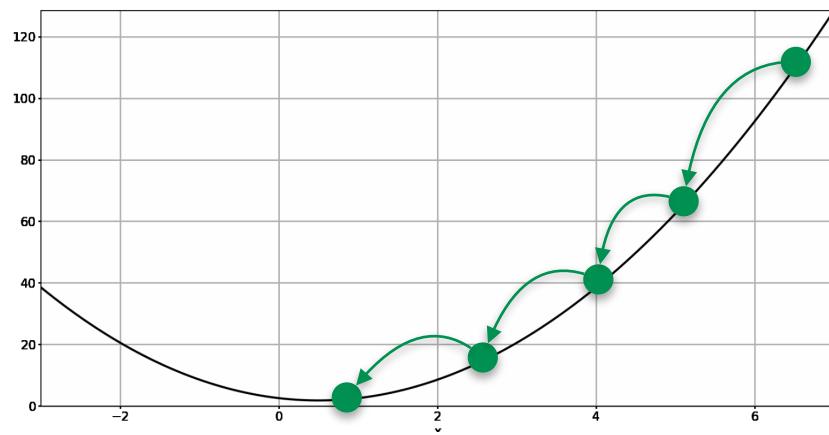
# What Is a Good Learning Rate?

Just right

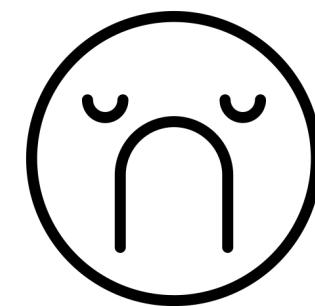


# What Is a Good Learning Rate?

Just right

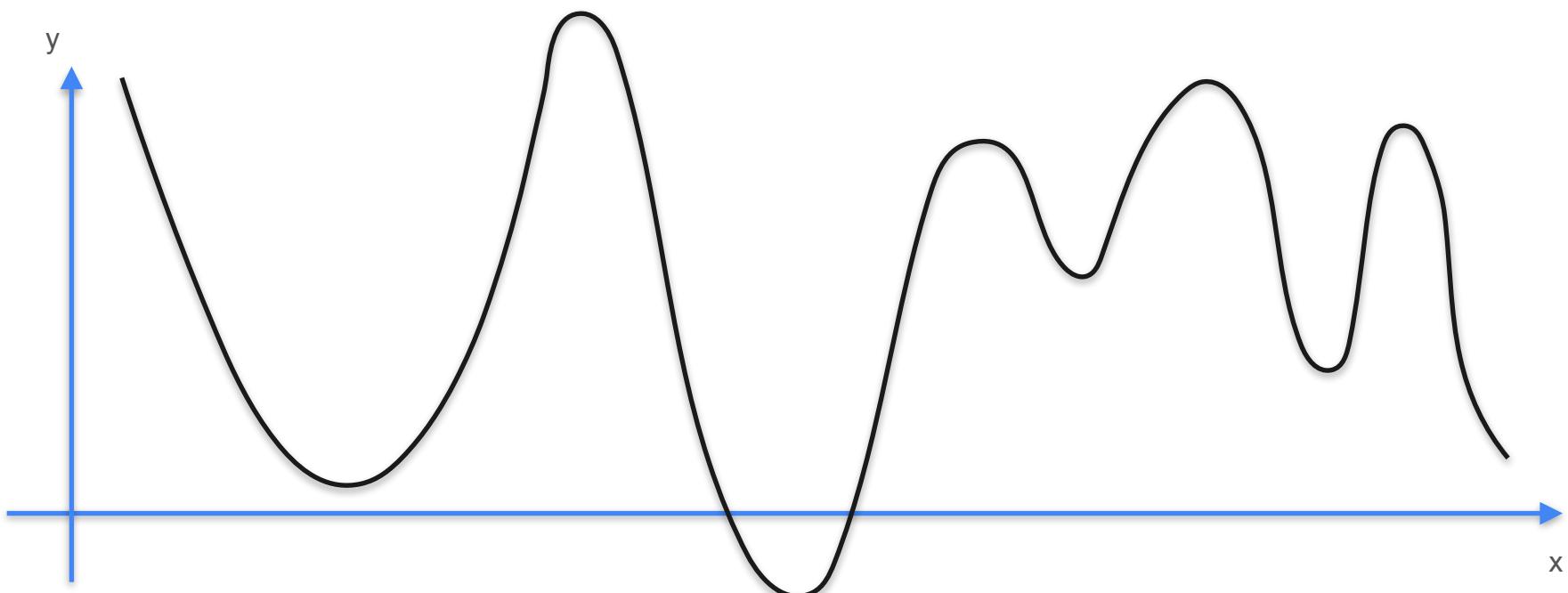


Unfortunately, there is no rule to give the best learning rate

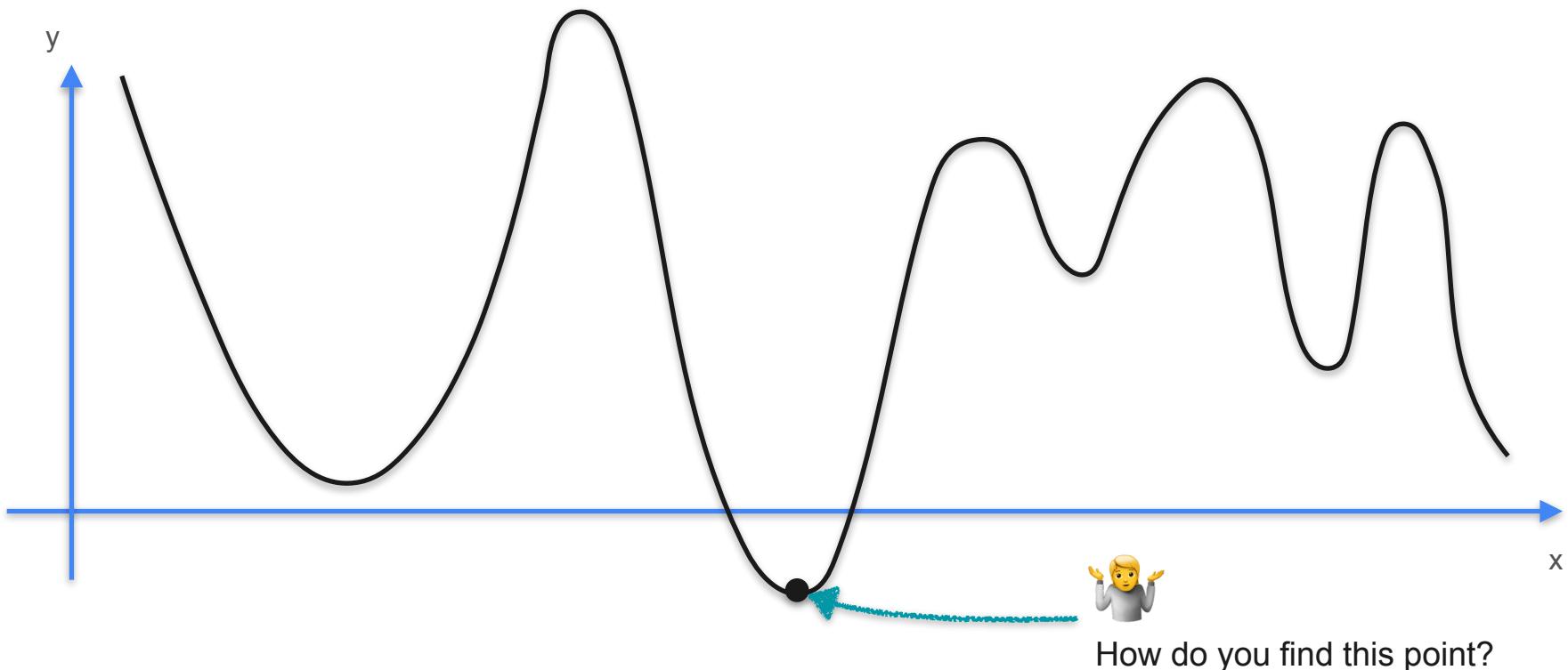


# Drawbacks of Gradient Descent

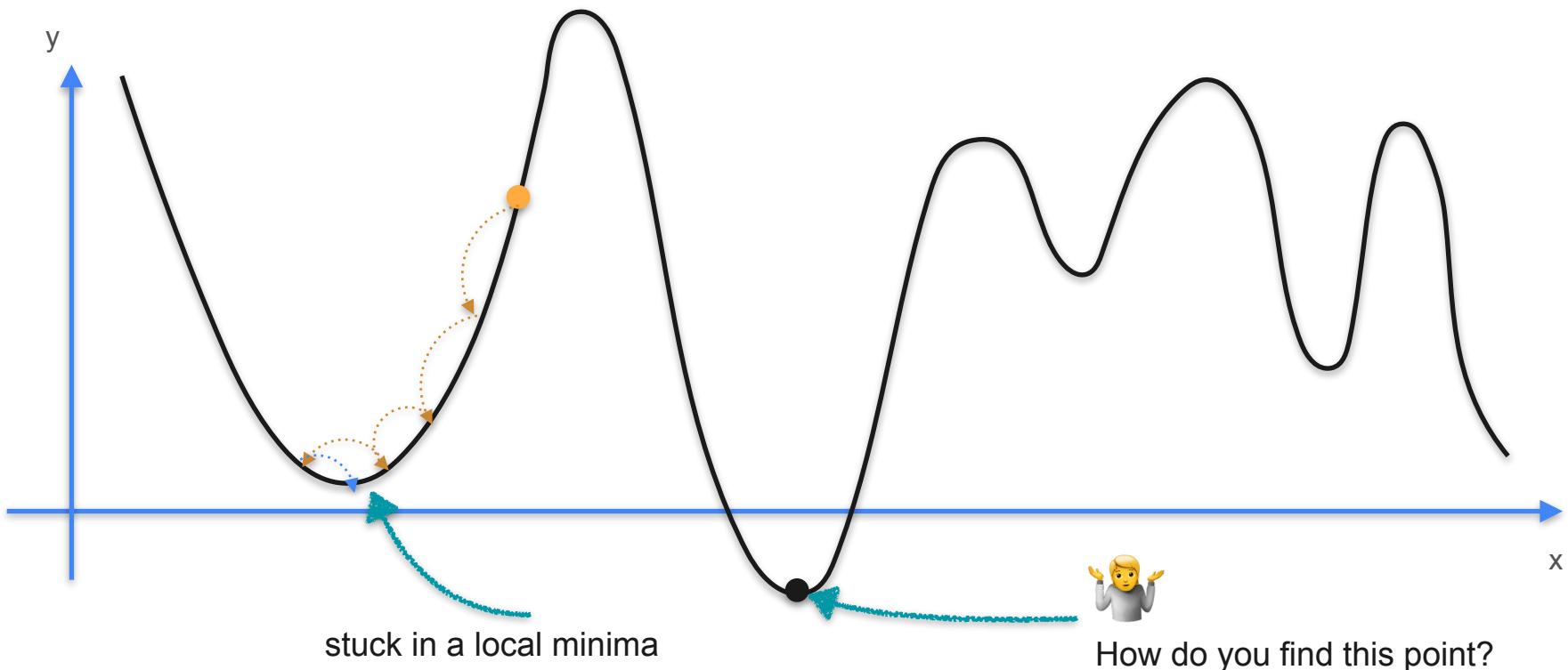
# Drawbacks of Gradient Descent



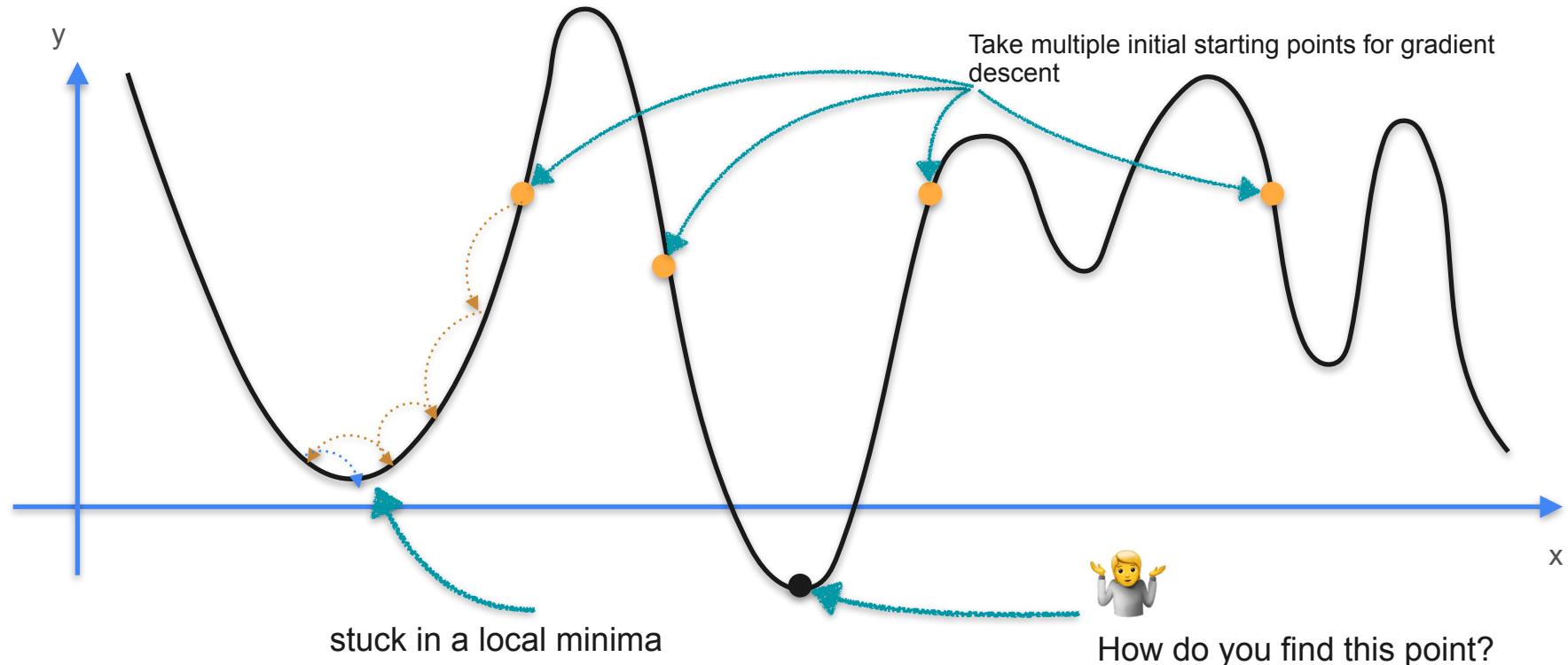
# Drawbacks of Gradient Descent



# Drawbacks of Gradient Descent



# Drawbacks of Gradient Descent





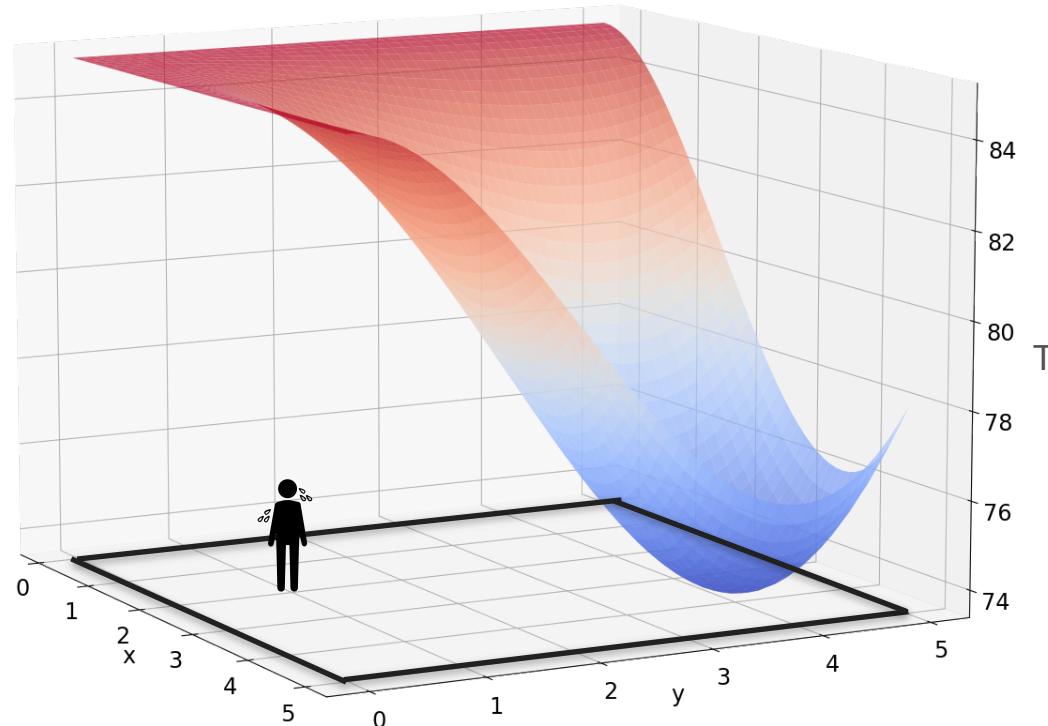
DeepLearning.AI

## Gradients and Gradient Descent

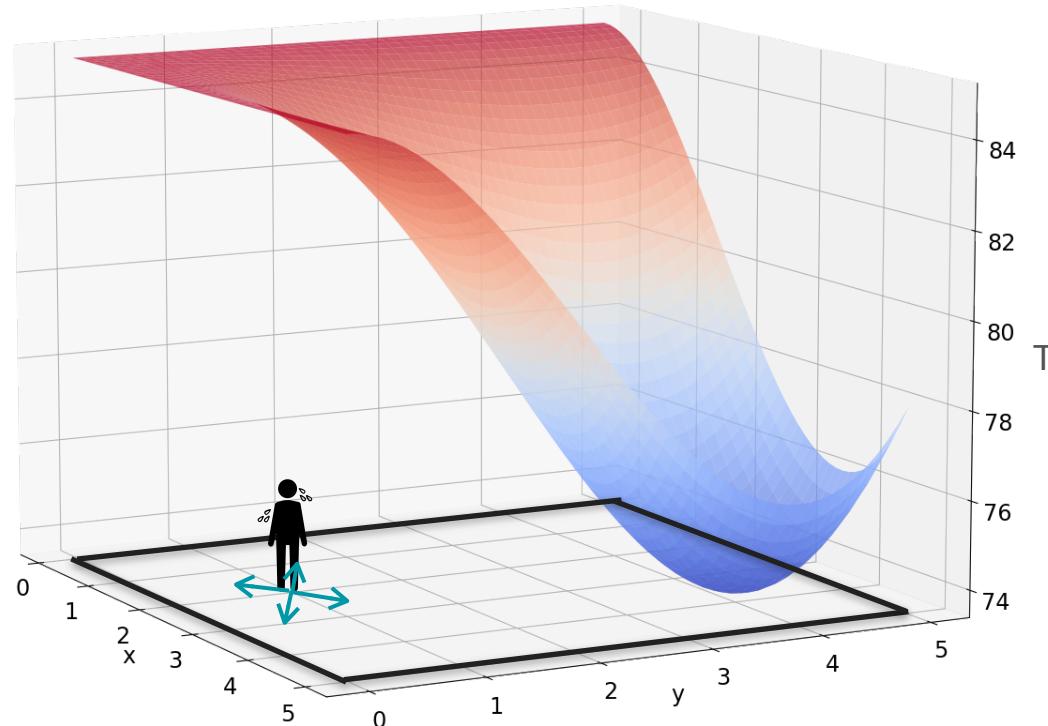
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**Optimization using Gradient  
Descent in two variables -  
Part 1**

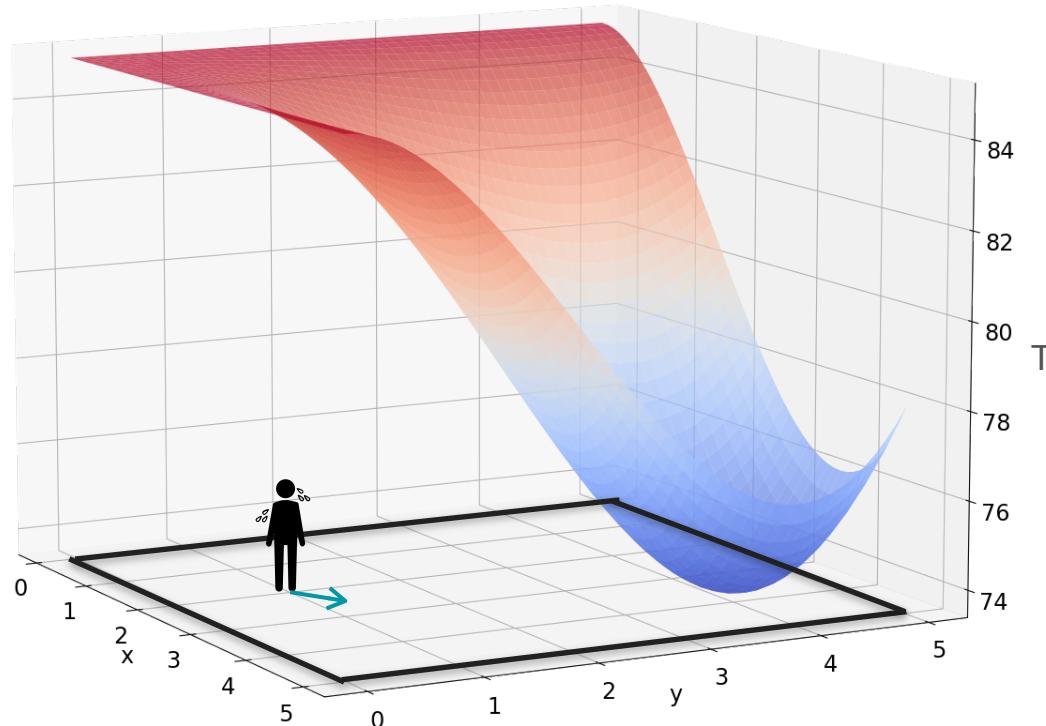
# Gradient Descent With Heat Example



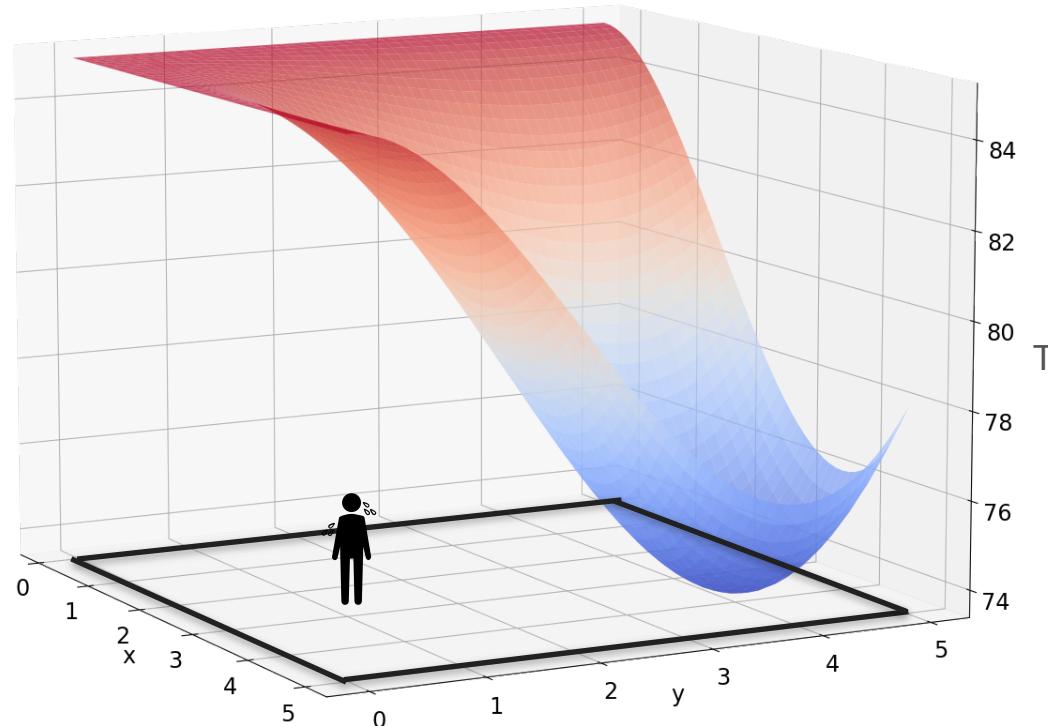
# Gradient Descent With Heat Example



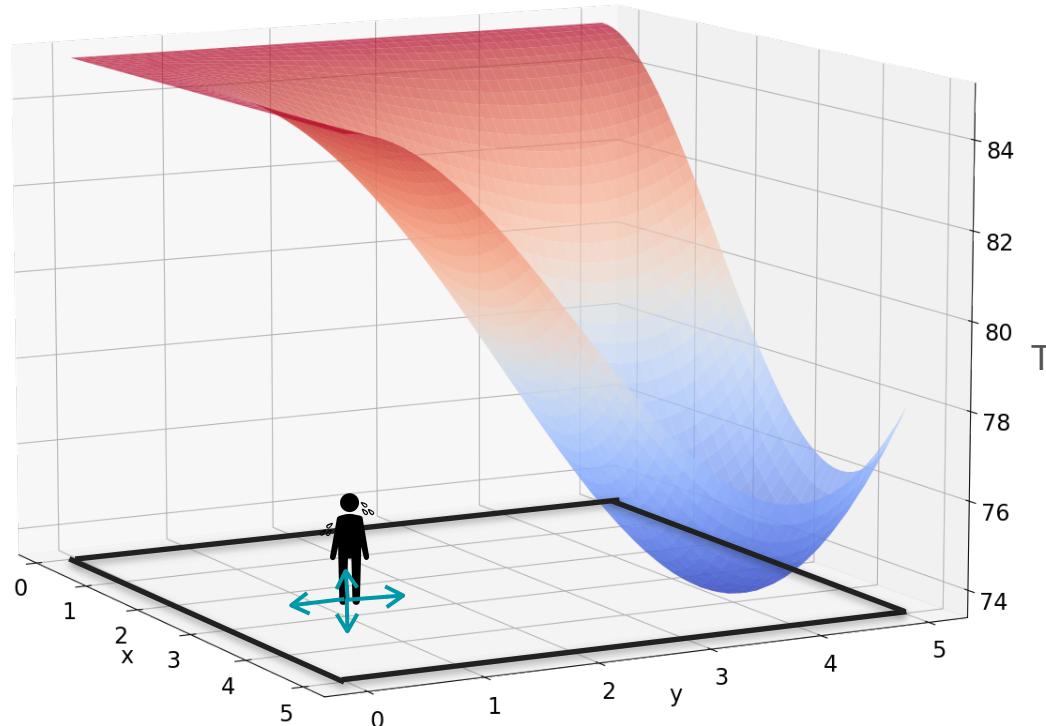
# Gradient Descent With Heat Example



# Gradient Descent With Heat Example

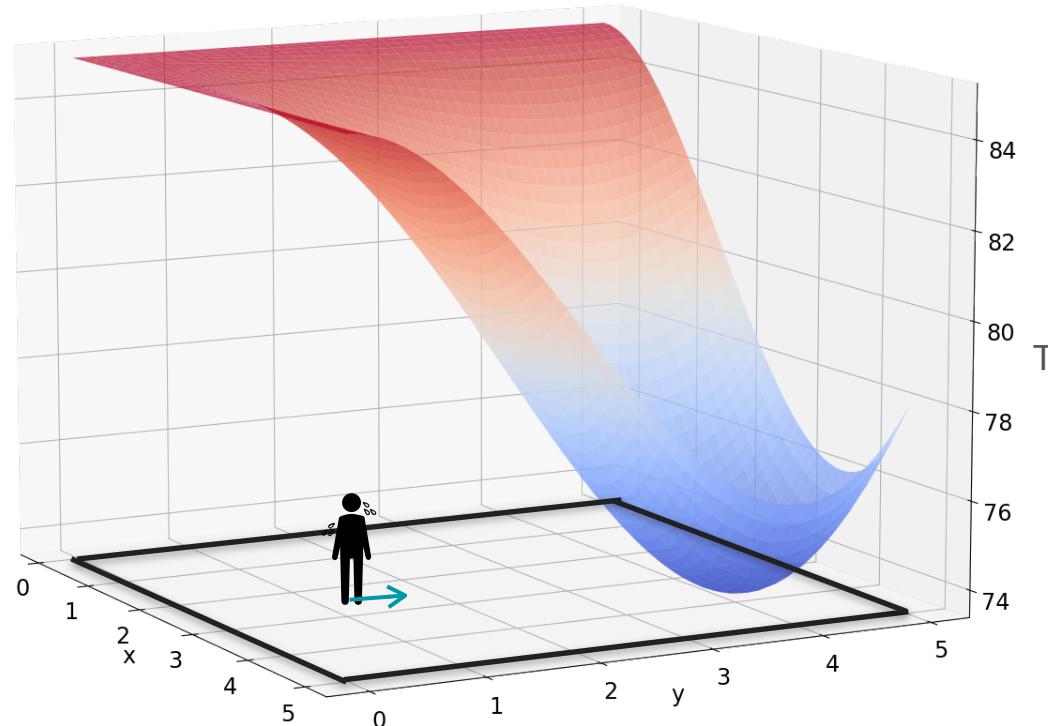


# Gradient Descent With Heat Example

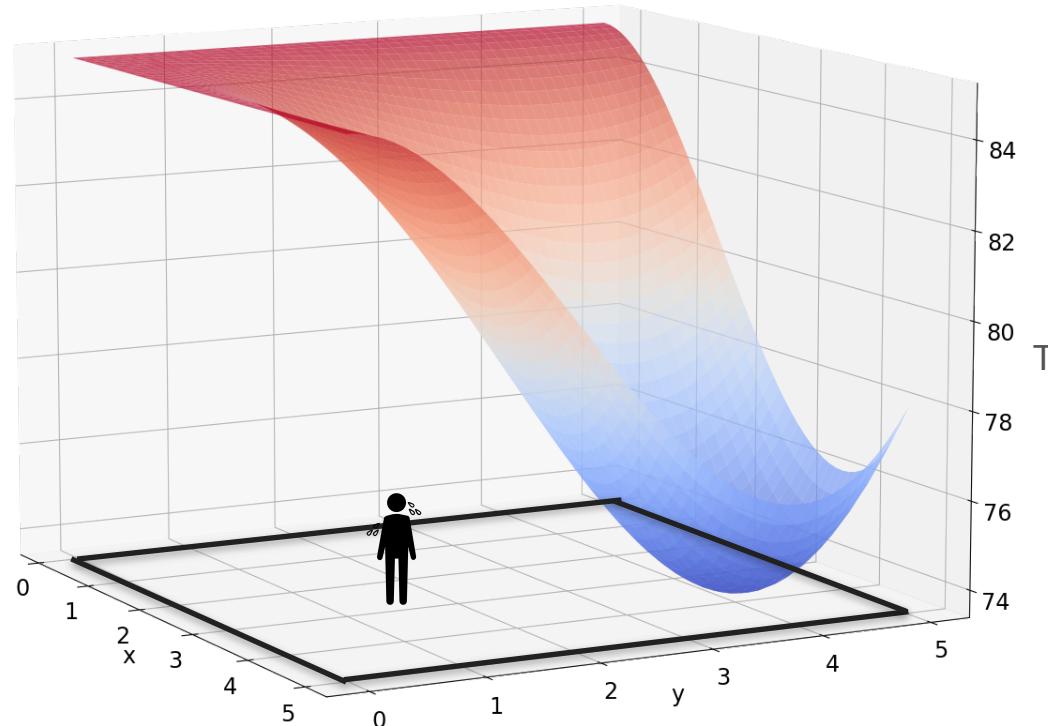


# Gradient Descent With Heat Example

Repeat!

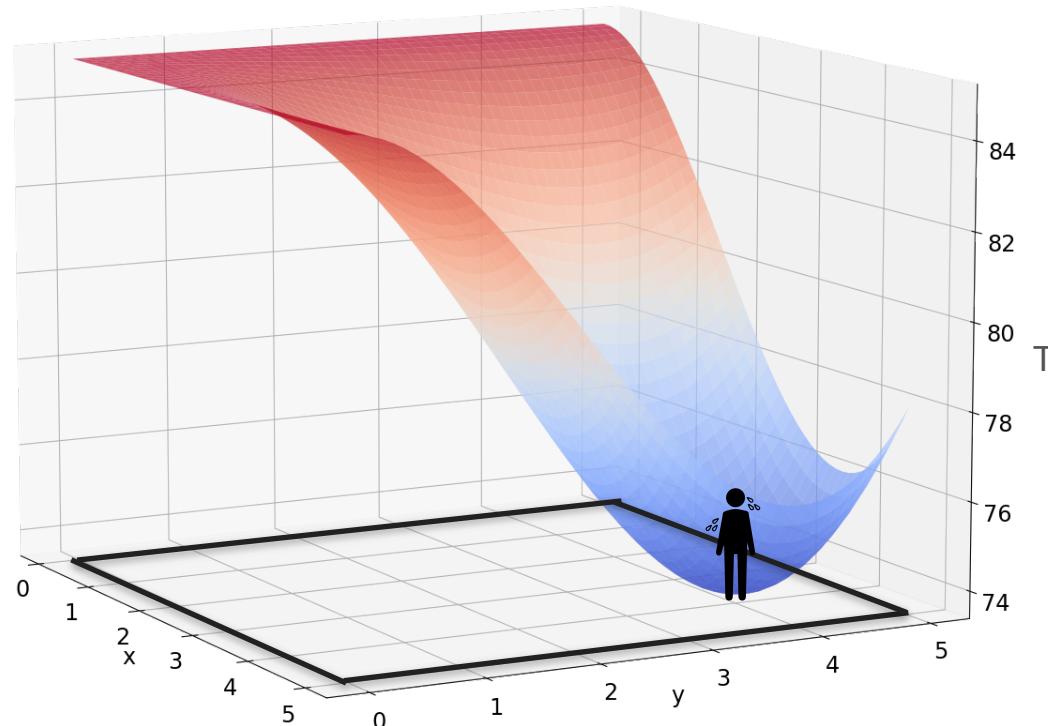


# Gradient Descent With Heat Example



# Gradient Descent With Heat Example

Repeat!





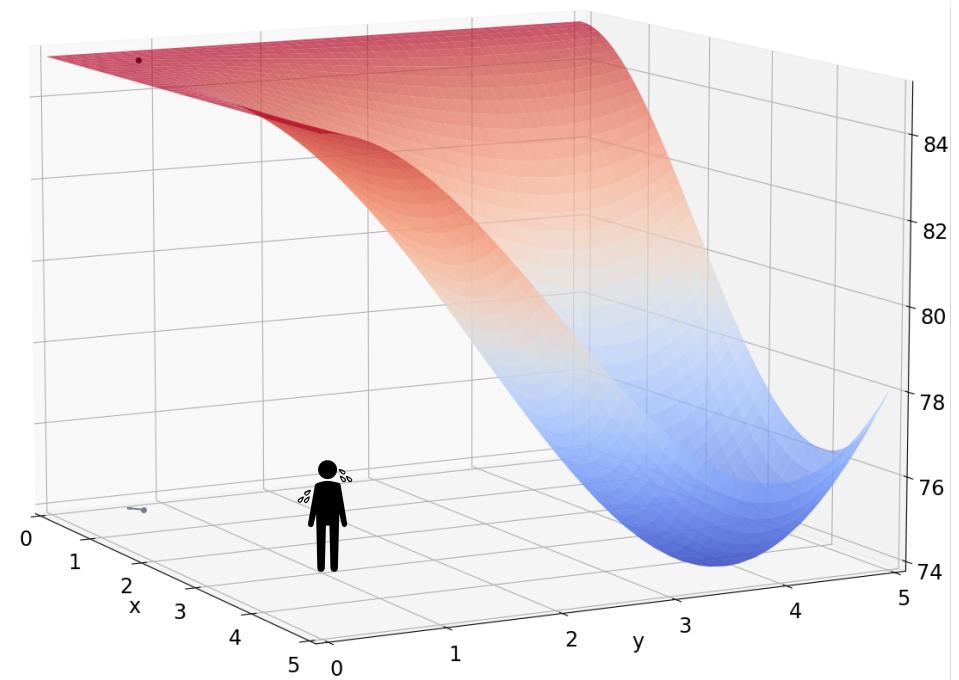
DeepLearning.AI

## Gradients and Gradient Descent

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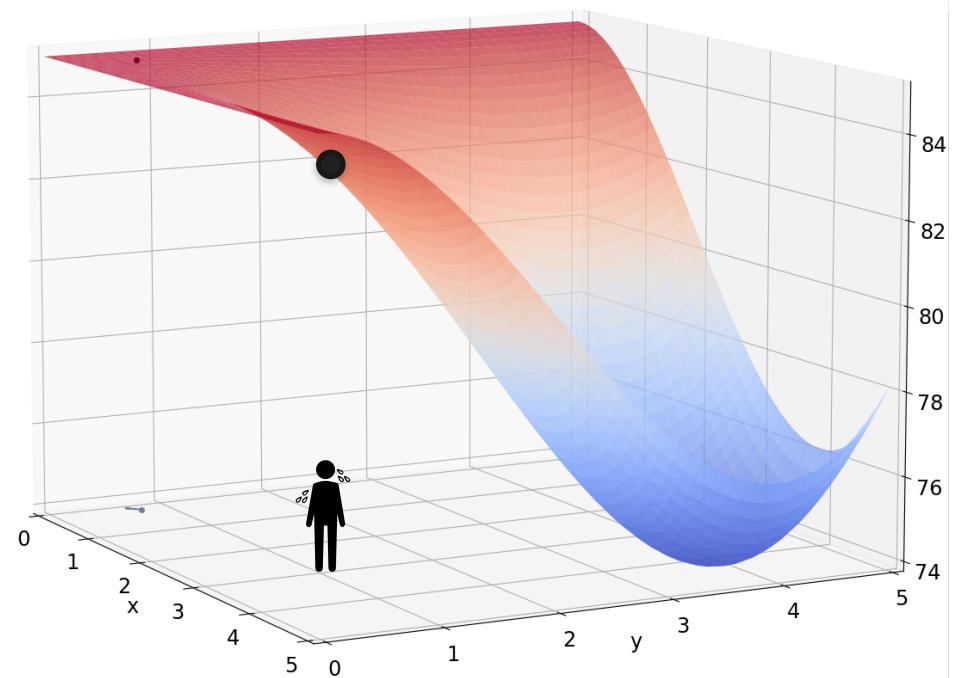
**Optimization using Gradient  
Descent in two variables -  
Part 2**

# Idea for Gradient Descent



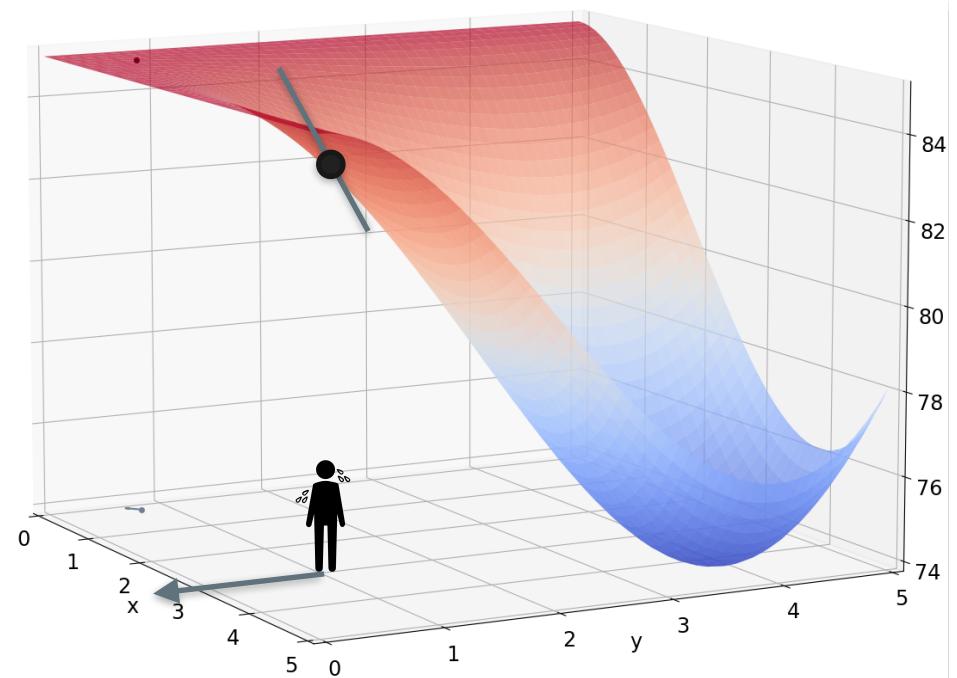
# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$



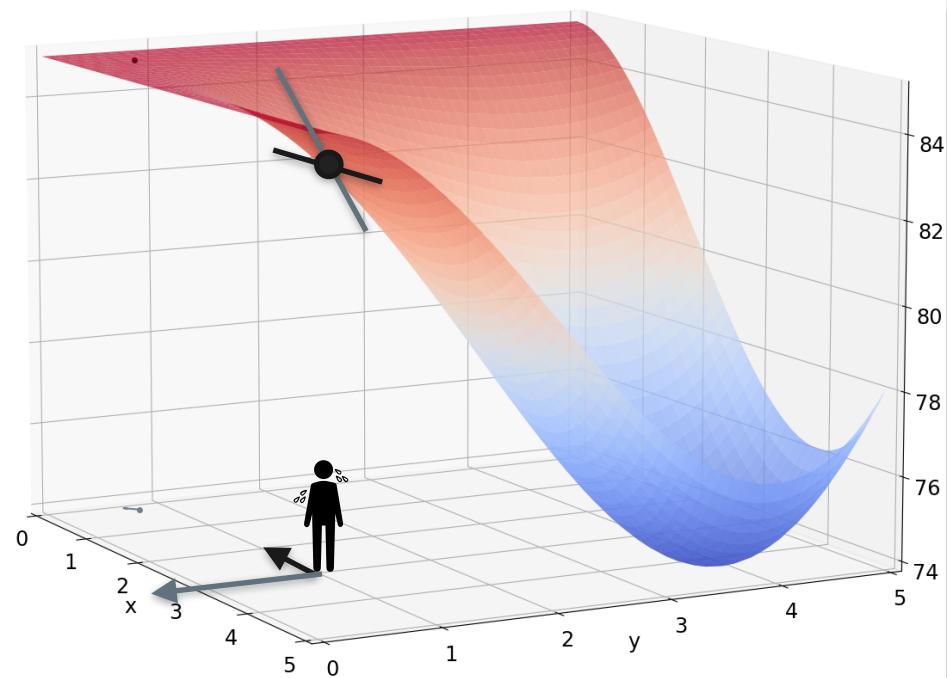
# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$



# Idea for Gradient Descent

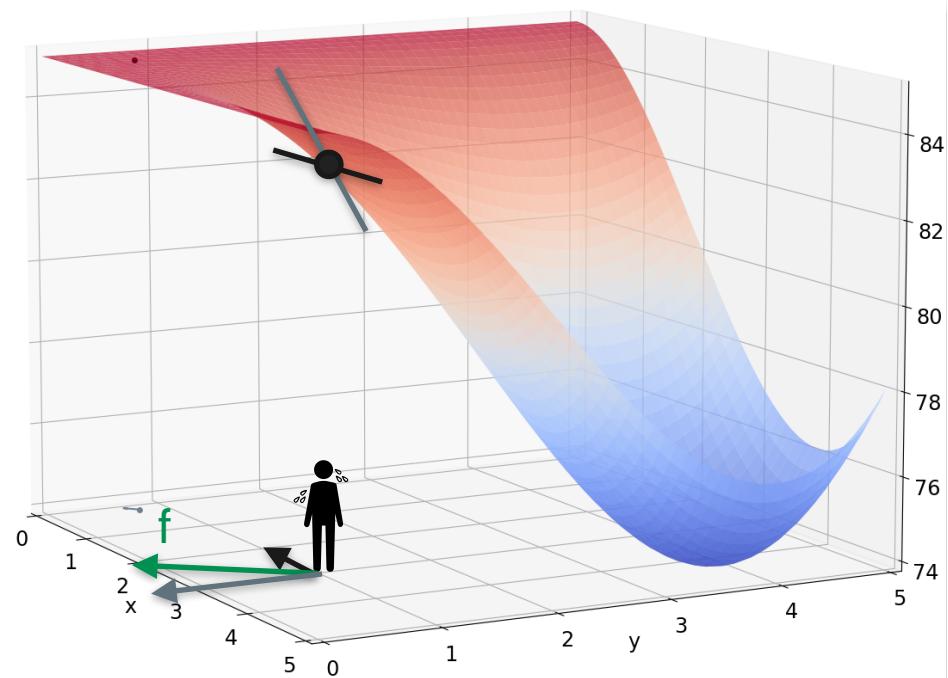
Initial position:  $(x_0, y_0)$



# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

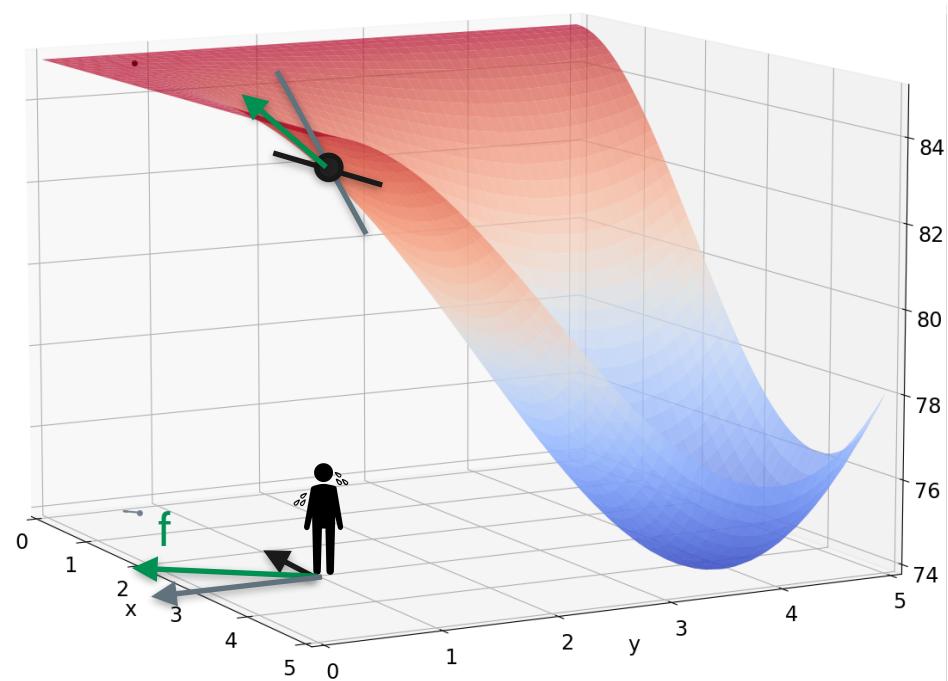
Direction of greatest ascent:  $\nabla f$



# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

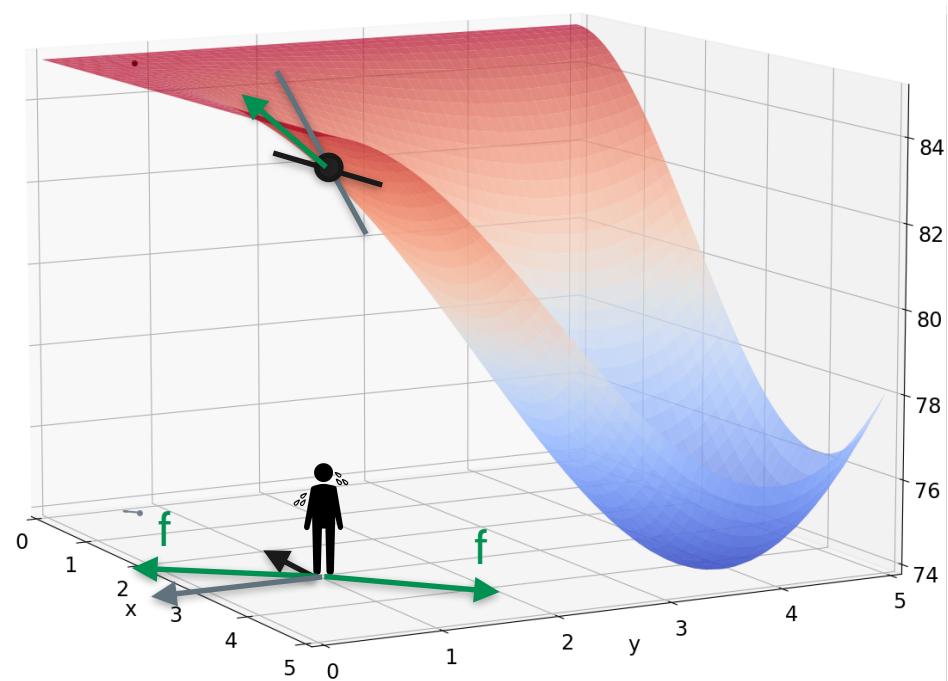


# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

Direction of greatest descent:  $-\nabla f$

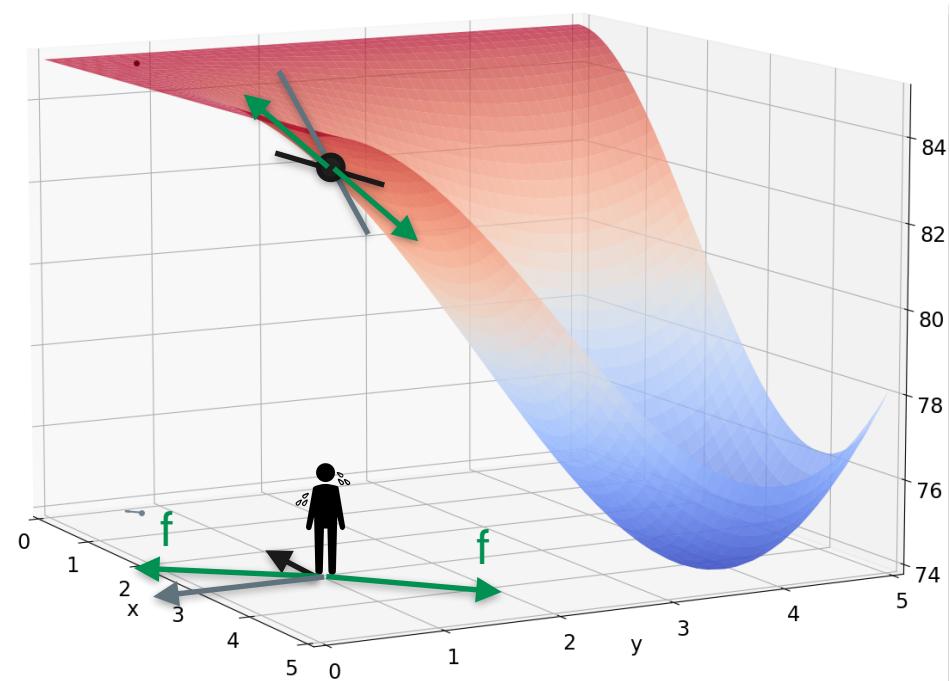


# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

Direction of greatest descent:  $-\nabla f$

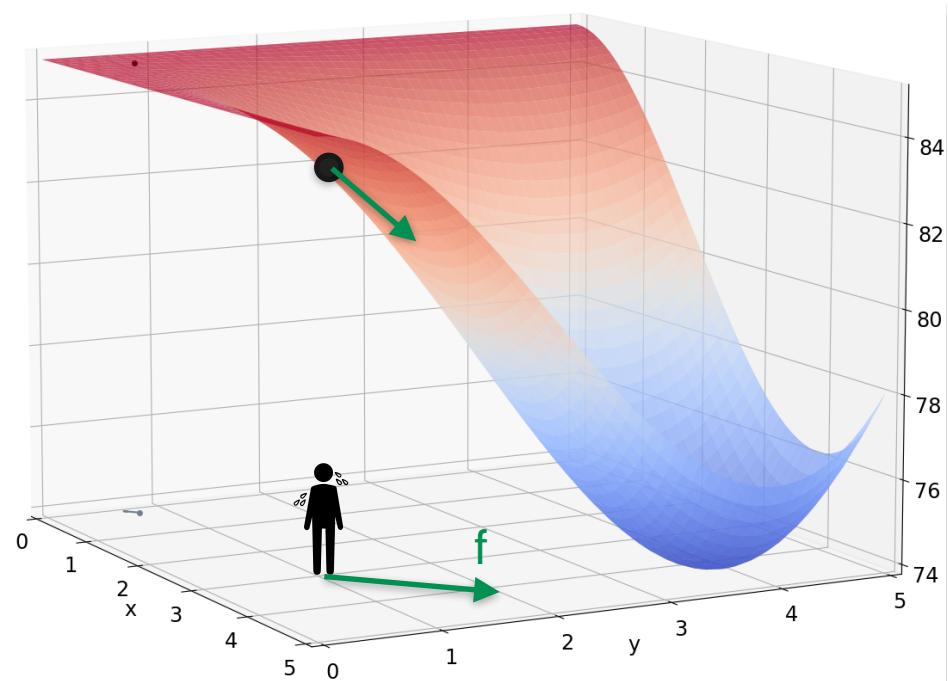


# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

Direction of greatest descent:  $-\nabla f$



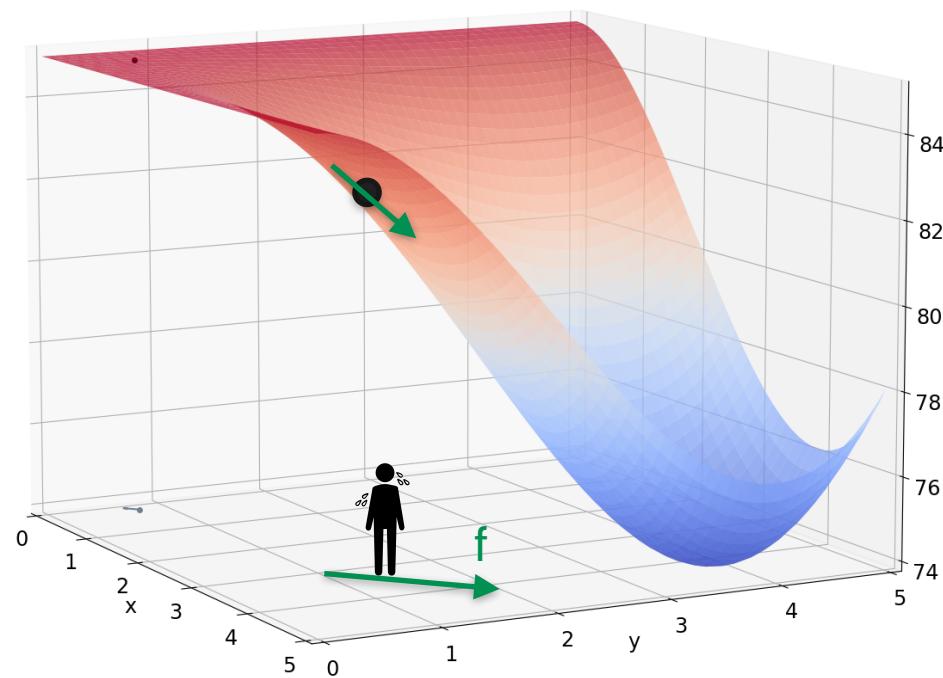
# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

Direction of greatest descent:  $-\nabla f$

Updated position:  $(x_0 + \delta, y_0)$   $\nabla f$



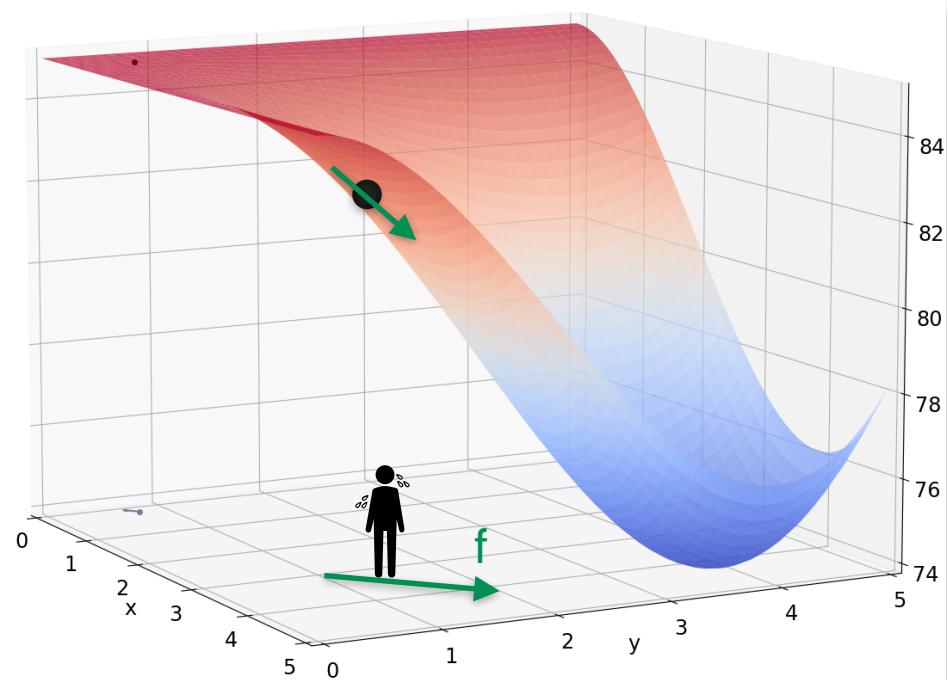
# Idea for Gradient Descent

Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $\nabla f$

Direction of greatest descent:  $-\nabla f$

Updated position:  $(x_0, y_0) \xrightarrow{\nabla f} (x_1, y_1)$



# Idea for Gradient Descent

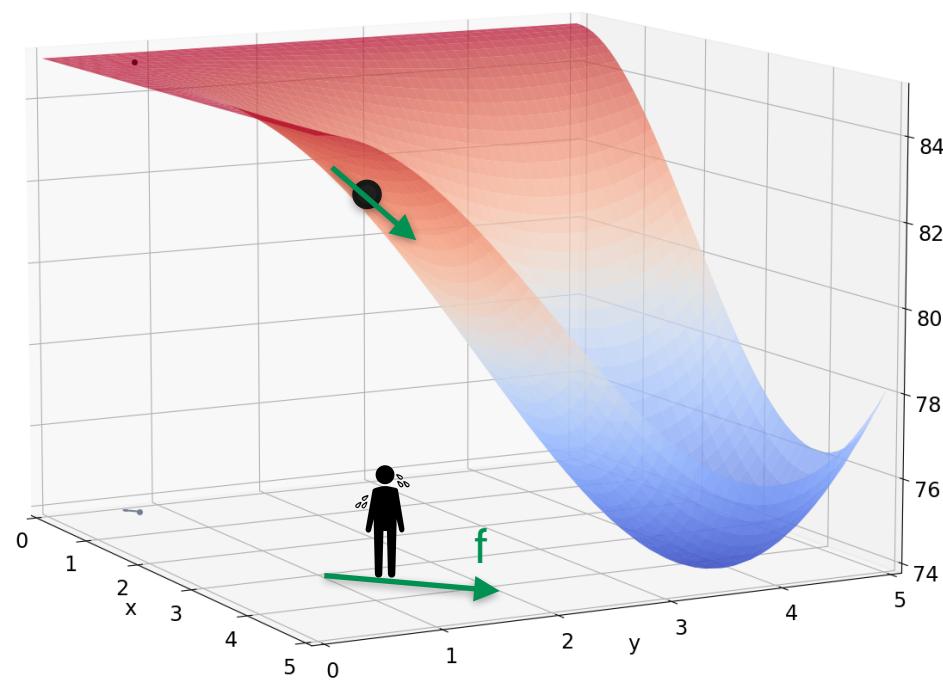
Initial position:  $(x_0, y_0)$

Direction of greatest ascent:  $f$

Direction of greatest descent:  $-f$

Updated position:  $(x_0, y_0) \xrightarrow{f} (x_1, y_1)$

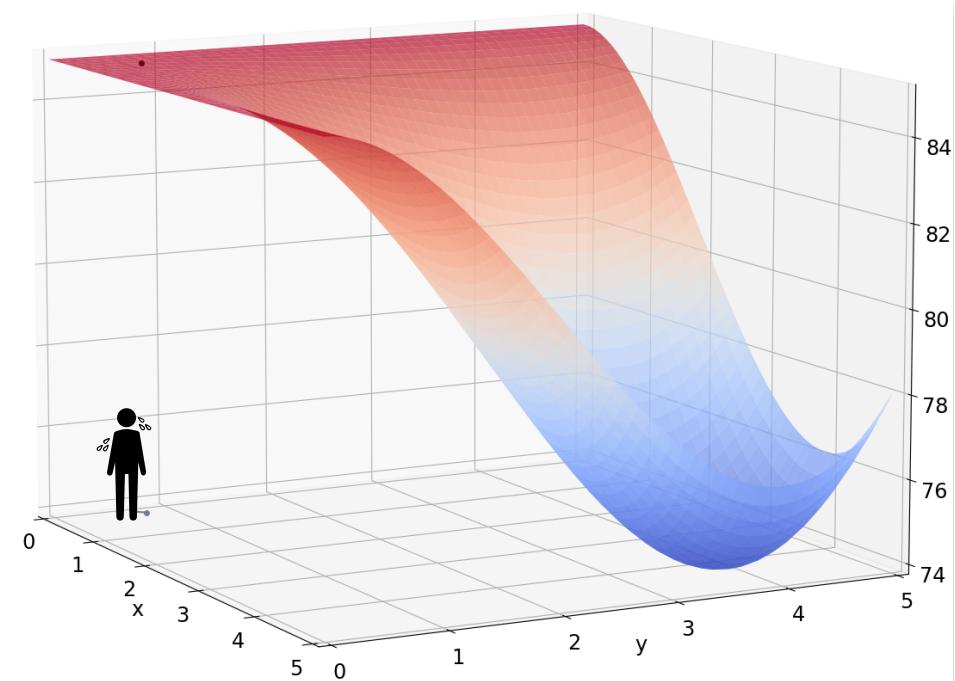
Better point!



## Method 2: Gradient Descent

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$

Start:  $x = 0.5, y = 0.6$

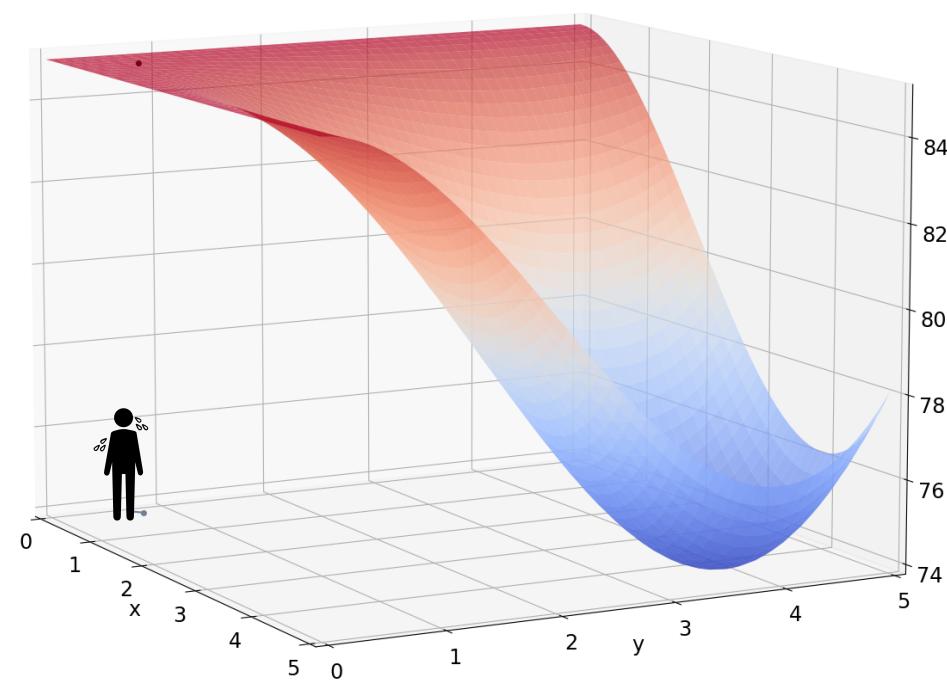


## Method 2: Gradient Descent

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$

Start:  $x = 0.5, y = 0.6$

$$f = \begin{bmatrix} f \\ \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$



## Method 2: Gradient Descent

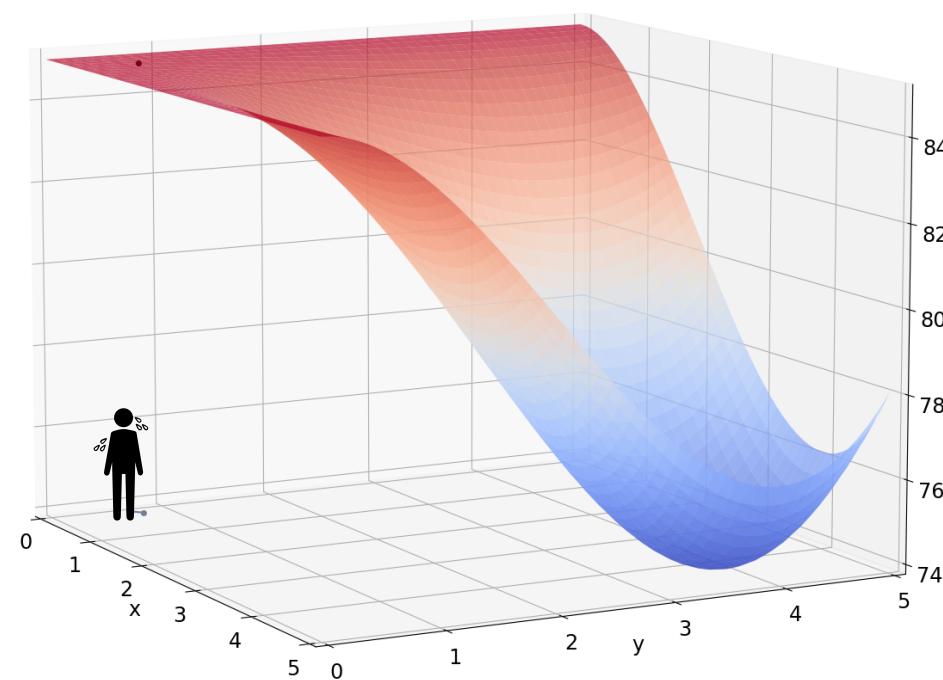
$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$

Start:  $x = 0.5, y = 0.6$

$$\begin{bmatrix} f \\ \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

$$\frac{\partial f}{\partial x} = \frac{1}{90}x(3x-12)y^2(y-6)$$

$$\frac{\partial f}{\partial y} = \frac{1}{90}x^2(x-6)y(3y-12)$$



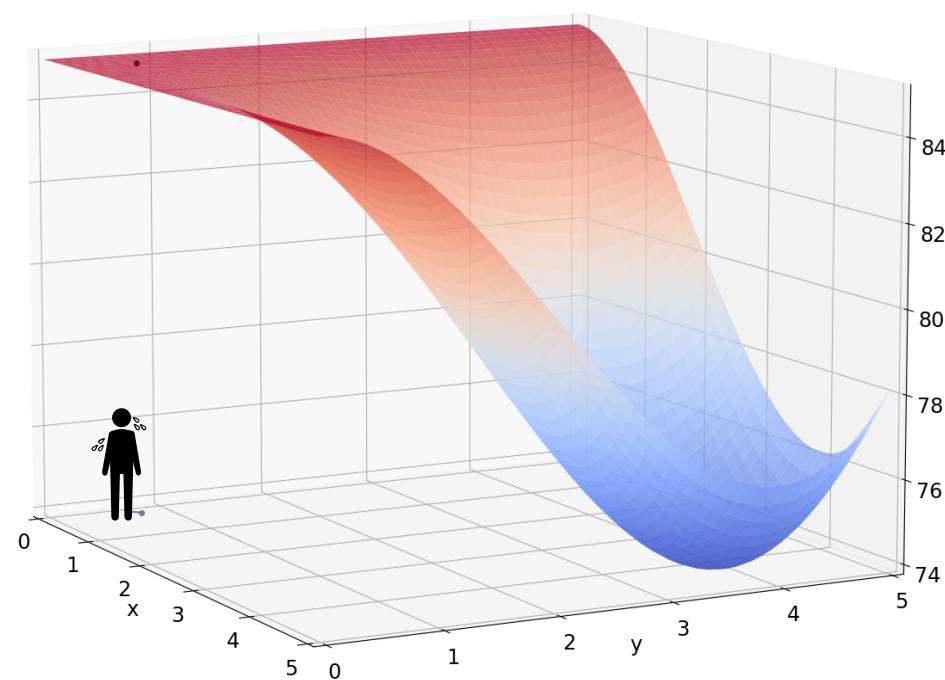
## Method 2: Gradient Descent

$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$

Start:  $x = 0.5, y = 0.6$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

$$\nabla f = \begin{bmatrix} \frac{1}{90}x(3x-12)y^2(y-6) \\ \frac{1}{90}x^2(x-6)y(3y-12) \end{bmatrix}$$



## Method 2: Gradient Descent

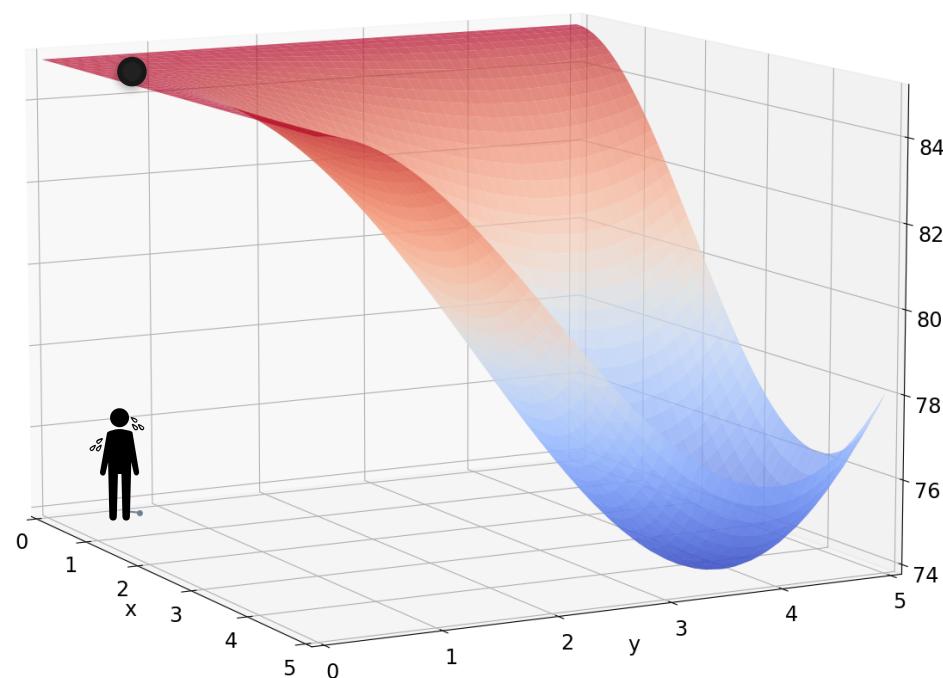
$$T = f(x, y) = 85 - \frac{1}{90}x^2(x-6)y^2(y-6)$$

Start:  $x = 0.5, y = 0.6$

$$f = \begin{bmatrix} f \\ \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

$$f = \begin{bmatrix} \frac{1}{90}x(3x-12)y^2(y-6) \\ \frac{1}{90}x^2(x-6)y(3y-12) \end{bmatrix}$$

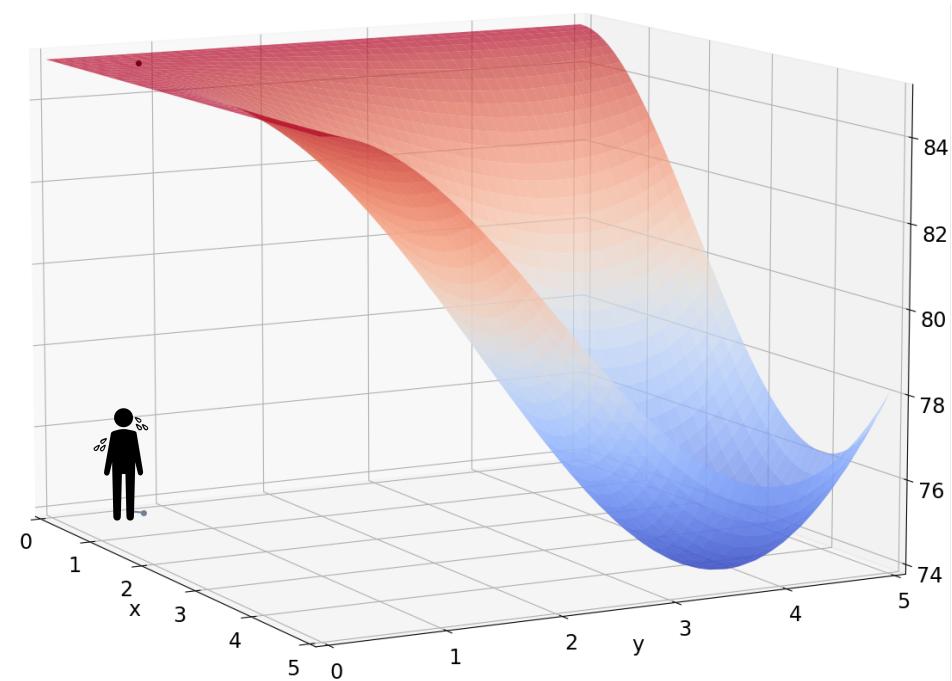
$$f(0.5, 0.6) = \begin{bmatrix} 0.1134 \\ 0.0935 \end{bmatrix}$$



# Method 2: Gradient Descent

Start:  $x = 0.5, y = 0.6$

$$f(0.5, 0.6) = \begin{bmatrix} 0.1134 \\ 0.0935 \end{bmatrix}$$

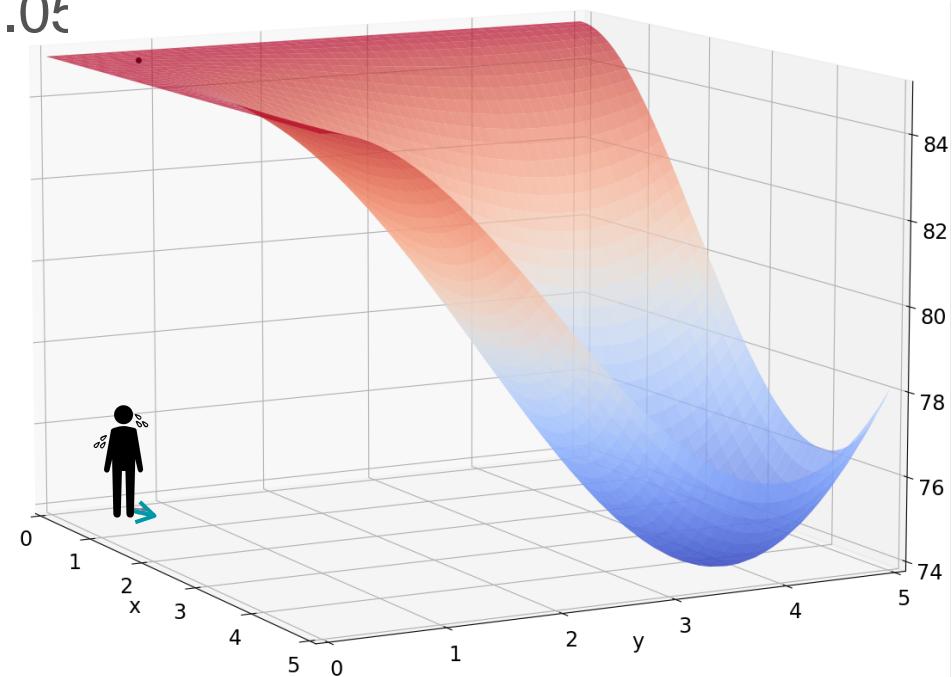


## Method 2: Gradient Descent

Start:  $x = 0.5, y = 0.6$    Rate:  $= 0.05$

$$f(0.5, 0.6) = \begin{bmatrix} 0.1134 \\ 0.0935 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5, 0.6)$



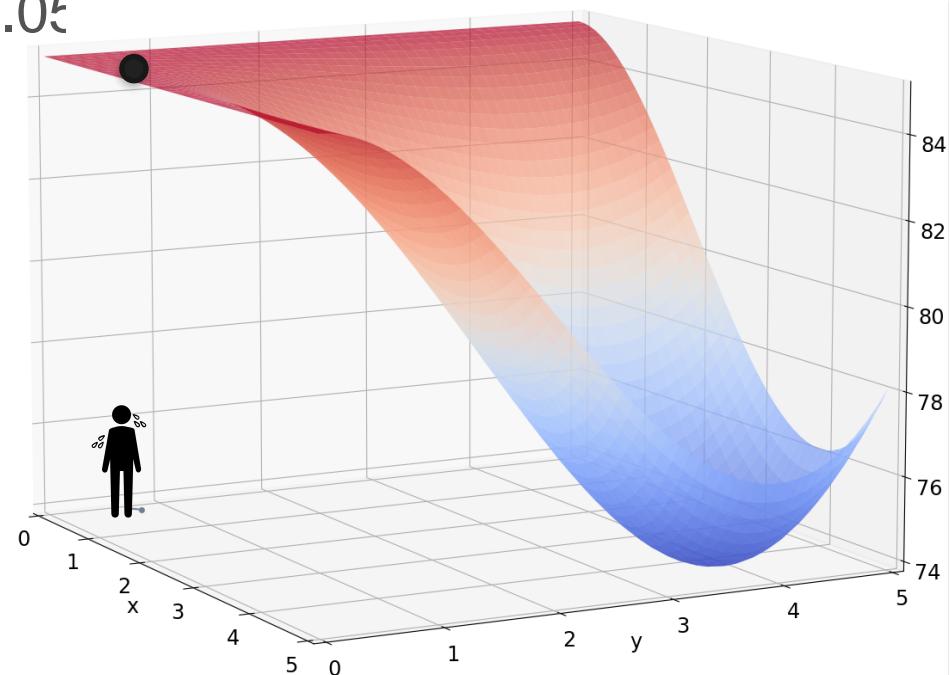
## Method 2: Gradient Descent

Start:  $x = 0.5$ ,  $y = 0.6$     Rate:  $= 0.05$

$$f(0.5, 0.6) = \begin{bmatrix} 0.1134 \\ 0.0935 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5, 0.6)$

$$\begin{aligned} x &= 0.5057 \\ y &= 0.6047 \end{aligned}$$



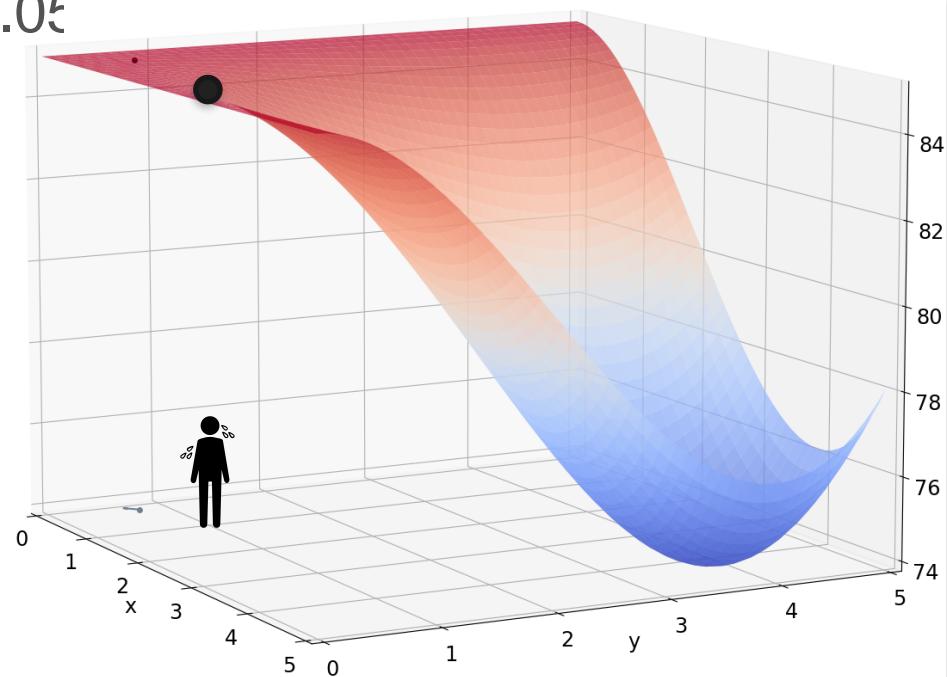
## Method 2: Gradient Descent

Start:  $x = 0.5$ ,  $y = 0.6$    Rate:  $= 0.05$

$$f(0.5, 0.6) = \begin{bmatrix} 0.1134 \\ 0.0935 \end{bmatrix}$$

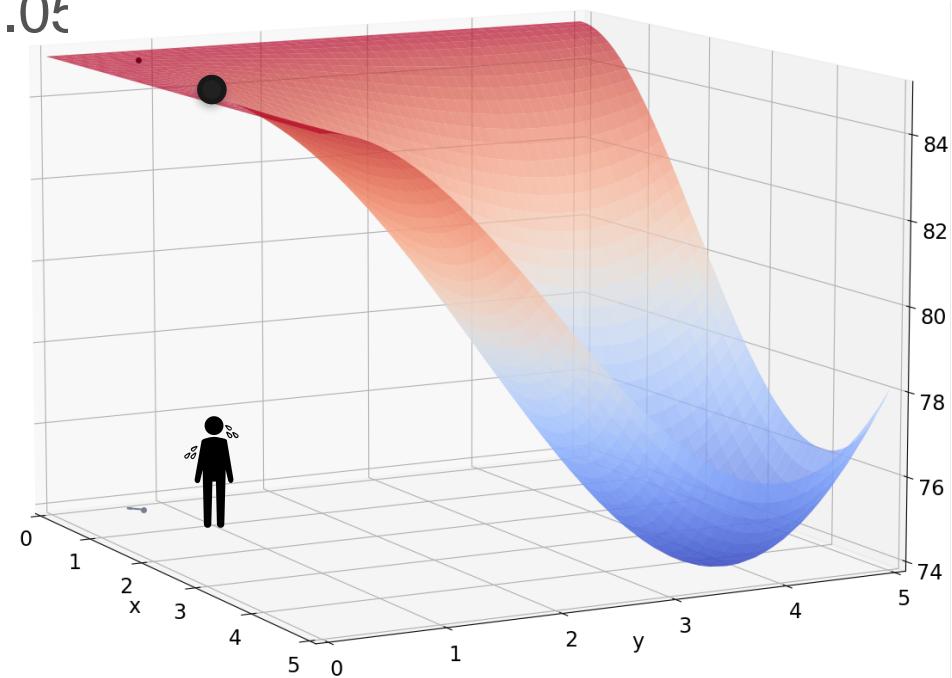
Move by  
 $0.05 \cdot f(0.5, 0.6)$

$$\begin{array}{ll} x & 0.5057 \\ y & 0.6047 \end{array}$$



# Method 2

Start:  $x = 0.5, y = 0.6$     Rate:  $= 0.05$



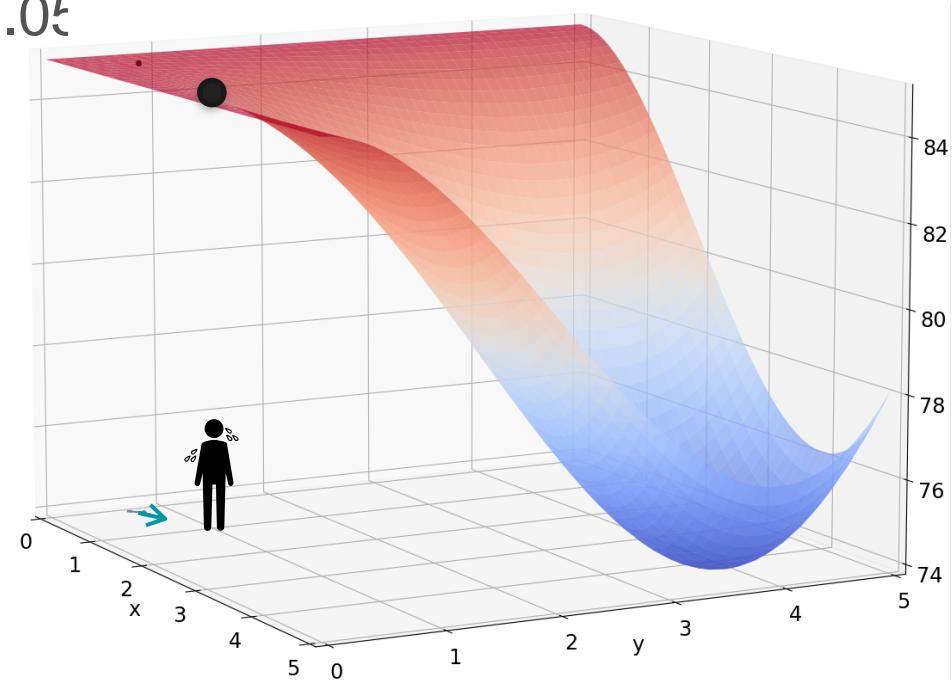
## Method 2

Start:  $x = 0.5$ ,  $y = 0.6$     Rate:  $= 0.05$

Find:

$$f(0.5057, 0.6047) = \begin{bmatrix} 0.1162 \\ 0.0961 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5057, 0.604)$



## Method 2

Start:  $x = 0.5$ ,  $y = 0.6$     Rate:  $= 0.05$

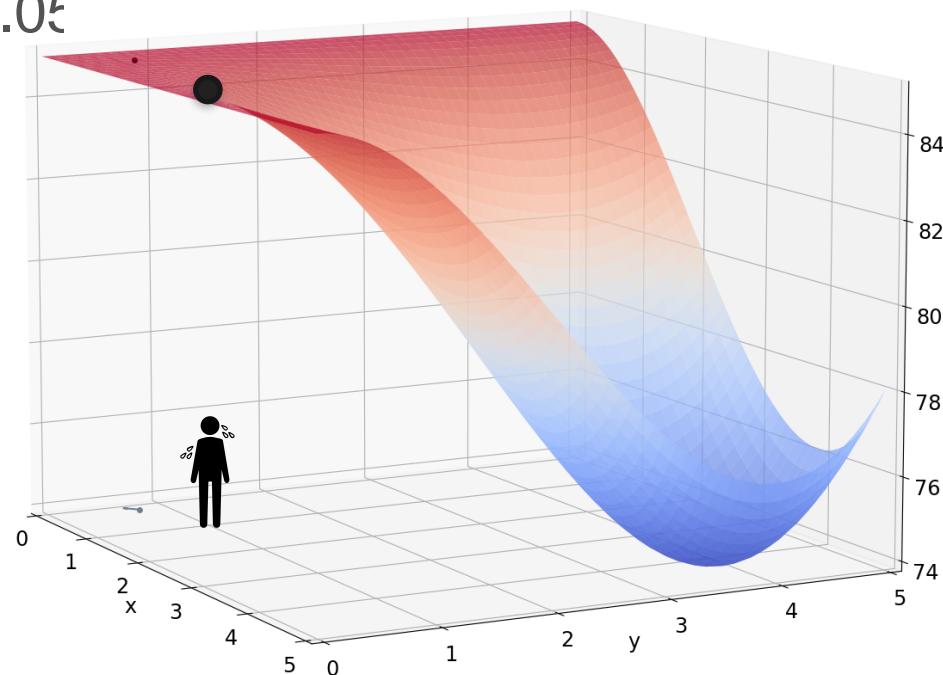
Find:

$$f(0.5057, 0.6047) = \begin{bmatrix} 0.1162 \\ 0.0961 \end{bmatrix}$$

Move by  
0.05  $f(0.5057, 0.604)$

$$\begin{array}{ll} x & 0.5115 \\ y & 0.6095 \end{array}$$

Repeat!



# Gradient Descent

Start:  $x = 0.5$ ,  $y = 0.6$    Rate:  $= 0.05$

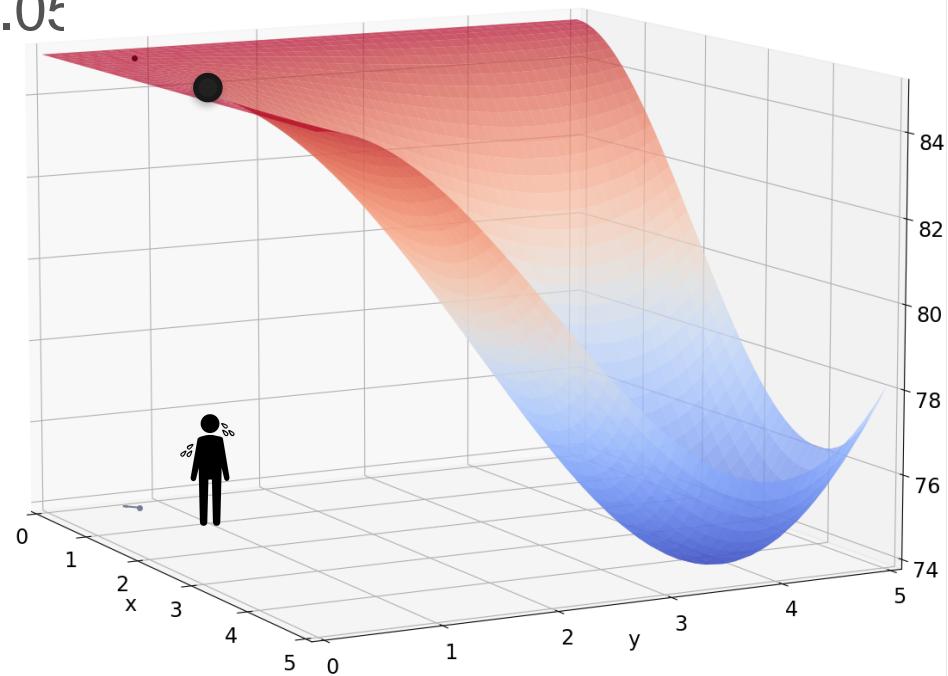
Find:

$$f(0.5057, 0.6047) = \begin{bmatrix} 0.1162 \\ 0.0961 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5057, 0.604)$

$$\begin{array}{ll} x & 0.5115 \\ y & 0.6095 \end{array}$$

Repeat!



# Gradient Descent

Start:  $x = 0.5$ ,  $y = 0.6$    Rate:  $= 0.05$

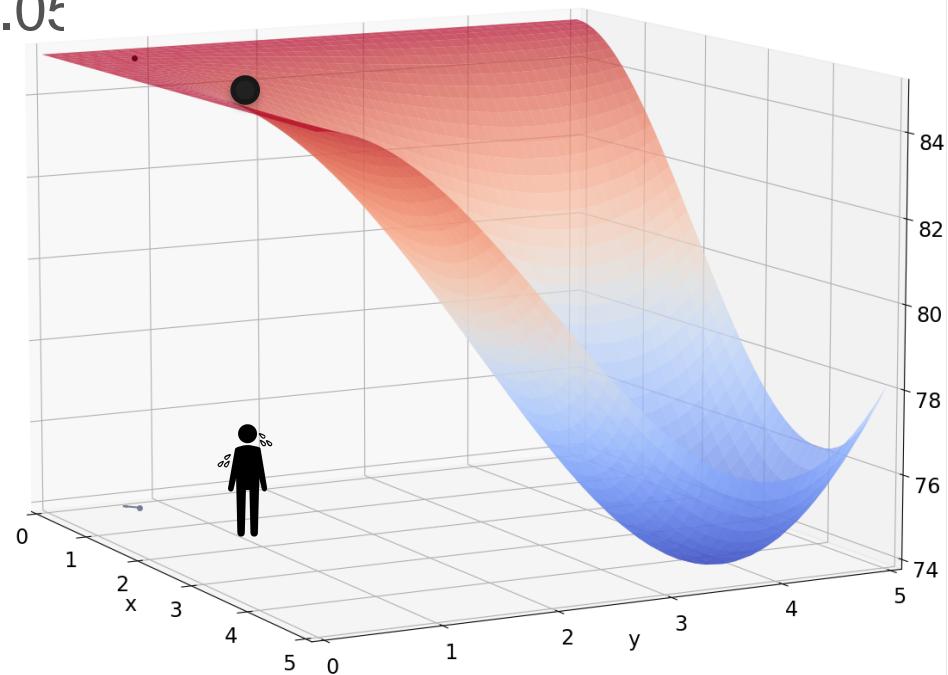
Find:

$$f(0.5057, 0.6047) = \begin{bmatrix} 0.1162 \\ 0.0961 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5057, 0.604)$

$$\begin{array}{ll} x & 0.5115 \\ y & 0.6095 \end{array}$$

Repeat!



# Gradient Descent

Start:  $x = 0.5$ ,  $y = 0.6$    Rate:  $= 0.05$

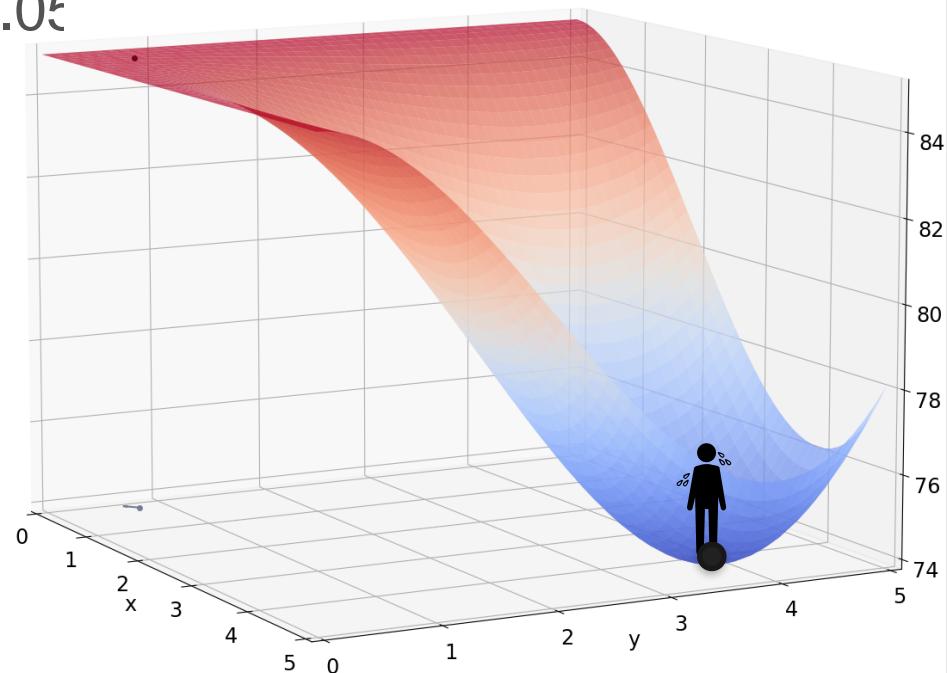
Find:

$$f(0.5057, 0.6047) = \begin{bmatrix} 0.1162 \\ 0.0961 \end{bmatrix}$$

Move by  
 $0.05 \cdot f(0.5057, 0.604)$

$$\begin{array}{ll} x & 0.5115 \\ y & 0.6095 \end{array}$$

Repeat!



# Gradient Descent

# Gradient Descent

Function:  $f(x, y)$

# Gradient Descent

Function:  $f(x, y)$

Goal: find minimum of  $f(x, y)$

# Gradient Descent

Function:  $f(x, y)$

Goal: find minimum of  $f(x, y)$

Step 1:

Define a learning rate

Choose a starting point  $(x_0, y_0)$

# Gradient Descent

Function:  $f(x, y)$

Goal: find minimum of  $f(x, y)$

Step 1:

Define a learning rate

Choose a starting point  $(x_0, y_0)$

Step 2:

Update:  $\begin{bmatrix} x_k \\ y_k \end{bmatrix} = \begin{bmatrix} x_{k-1} \\ y_{k-1} \end{bmatrix} - f(x_{k-1}, y_{k-1})$

# Gradient Descent

Function:  $f(x, y)$

Goal: find minimum of  $f(x, y)$

Step 1:

Define a learning rate

Choose a starting point  $(x_0, y_0)$

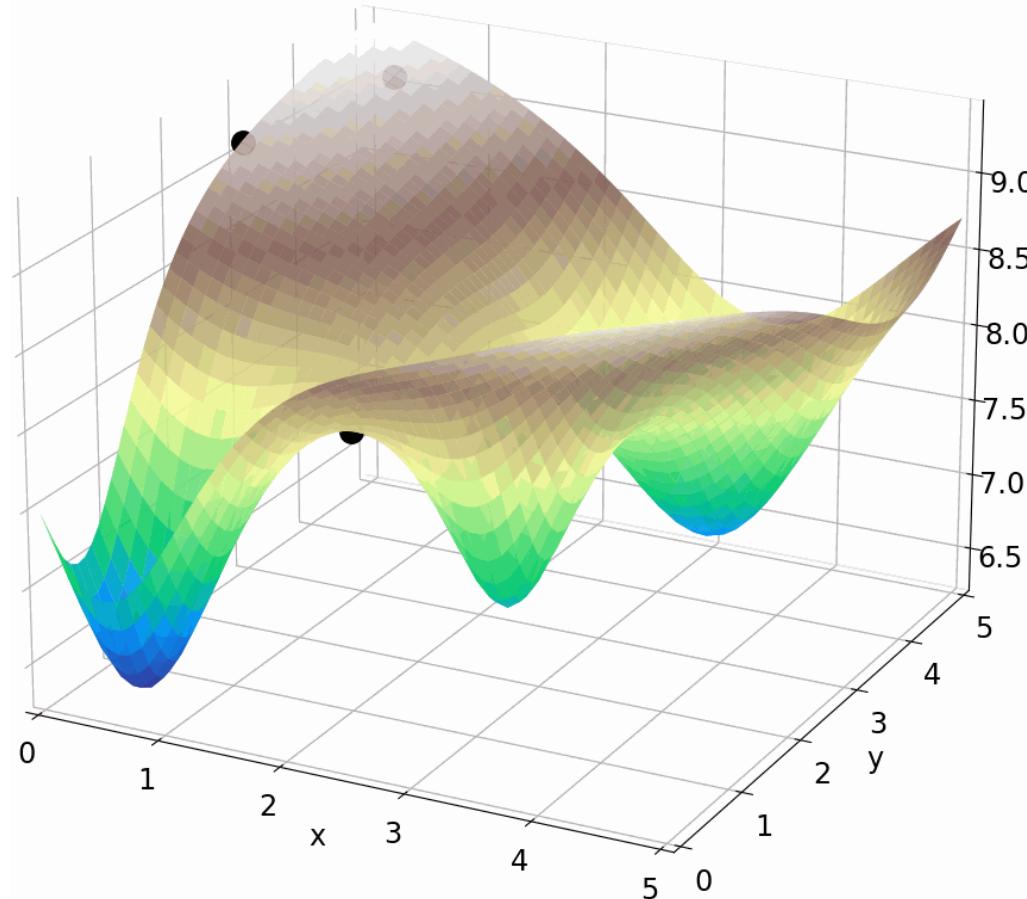
Step 2:

Update:  $\begin{bmatrix} x_k \\ y_k \end{bmatrix} = \begin{bmatrix} x_{k-1} \\ y_{k-1} \end{bmatrix} - f(x_{k-1}, y_{k-1})$

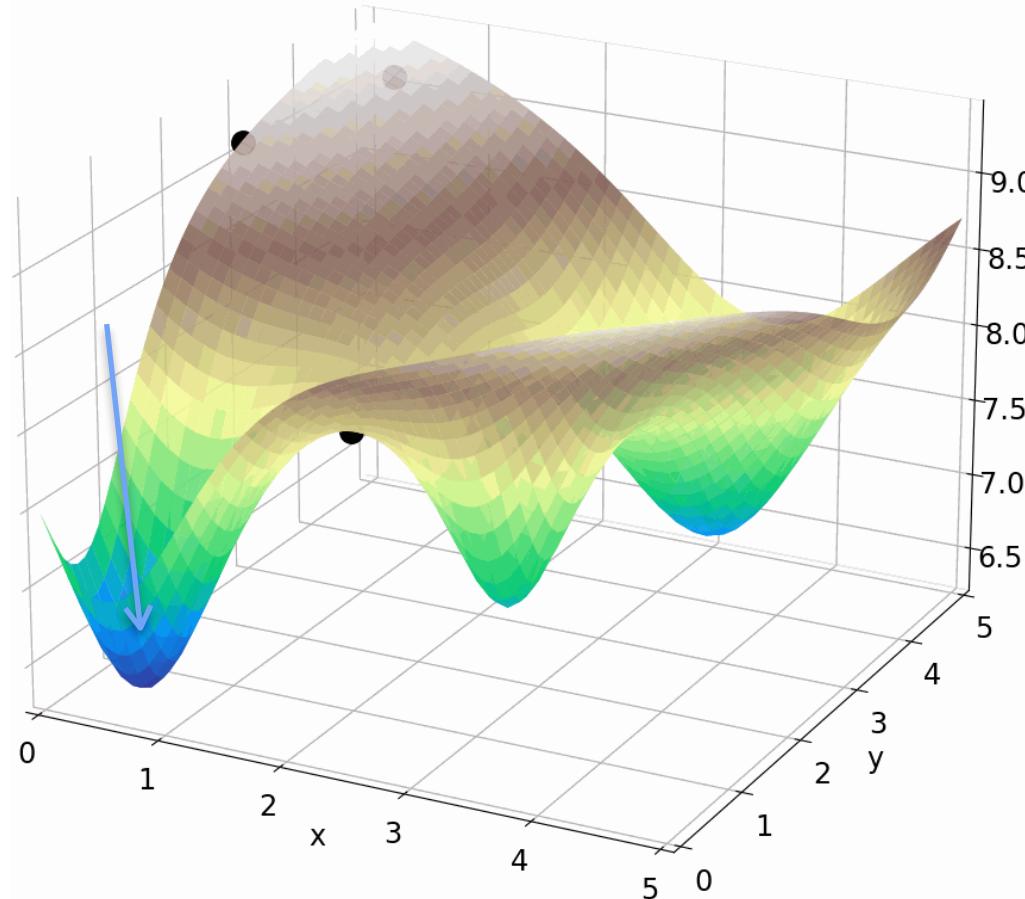
Step 3:

Repeat Step 2 until you are close enough to  
the true minimum  $(x^*, y^*)$

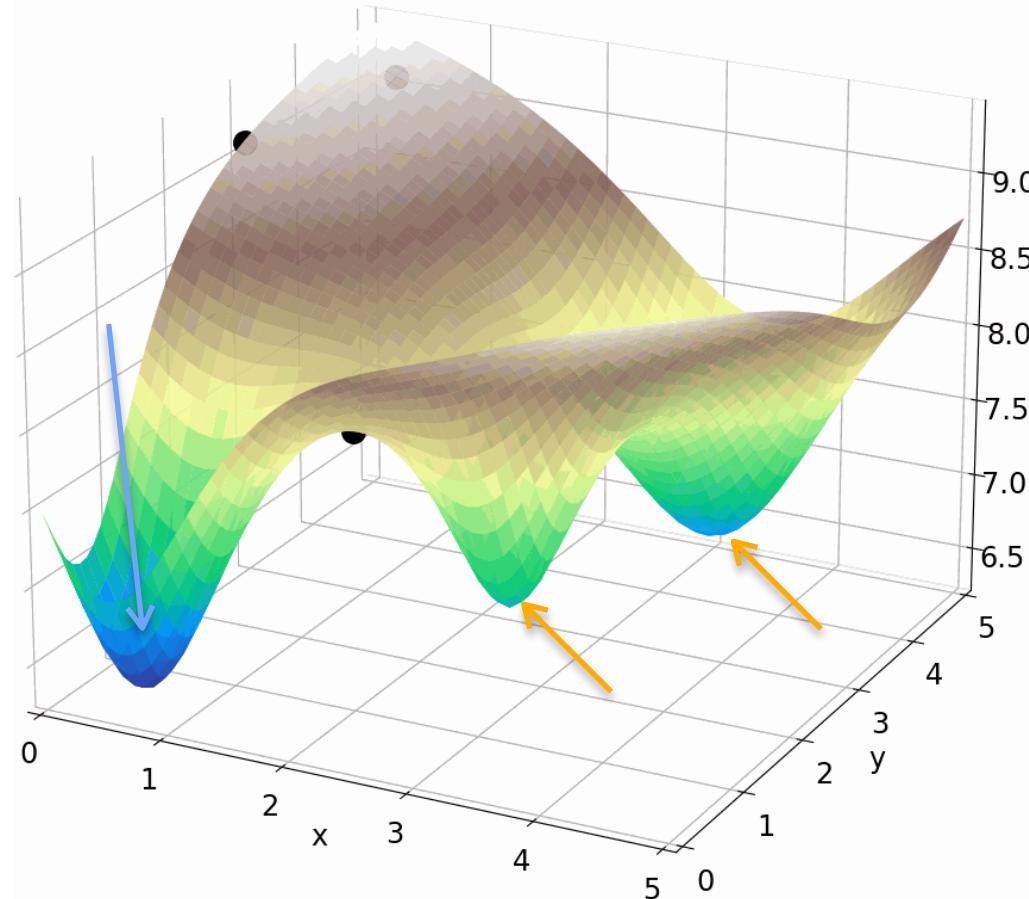
# Gradient Descent With Local Minima



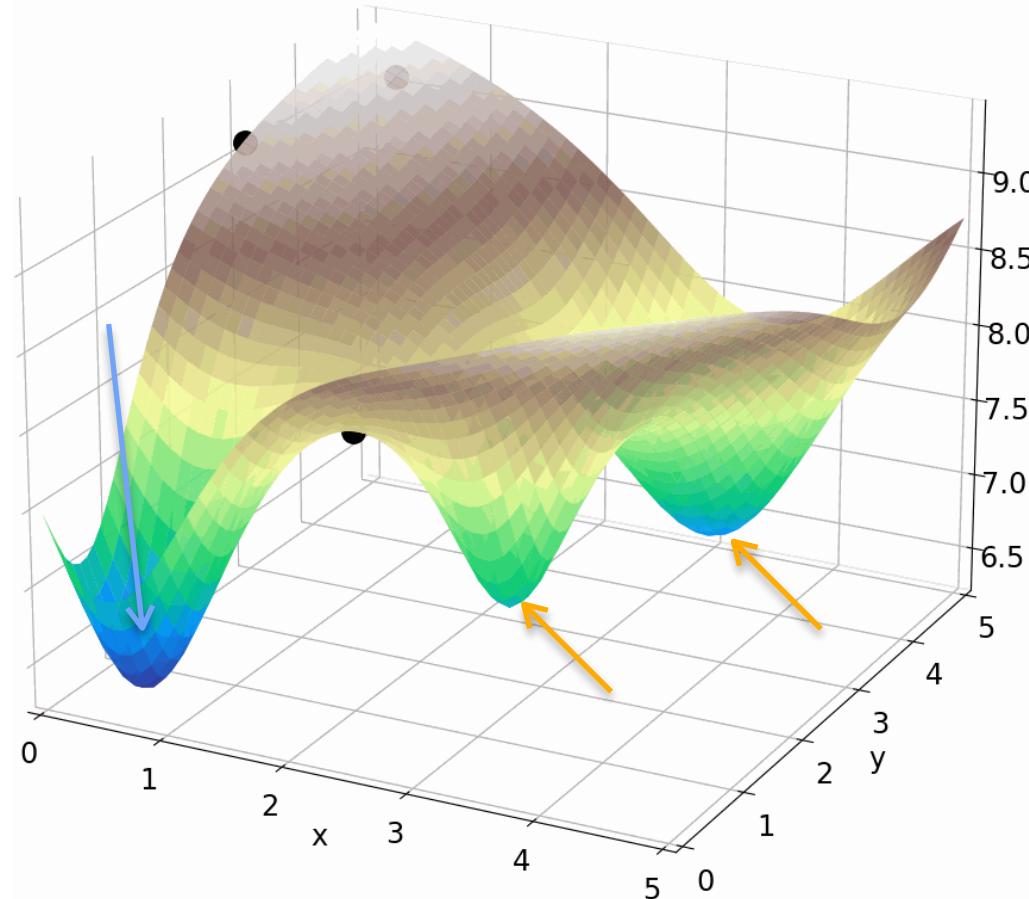
# Gradient Descent With Local Minima



# Gradient Descent With Local Minima



# Gradient Descent With Local Minima





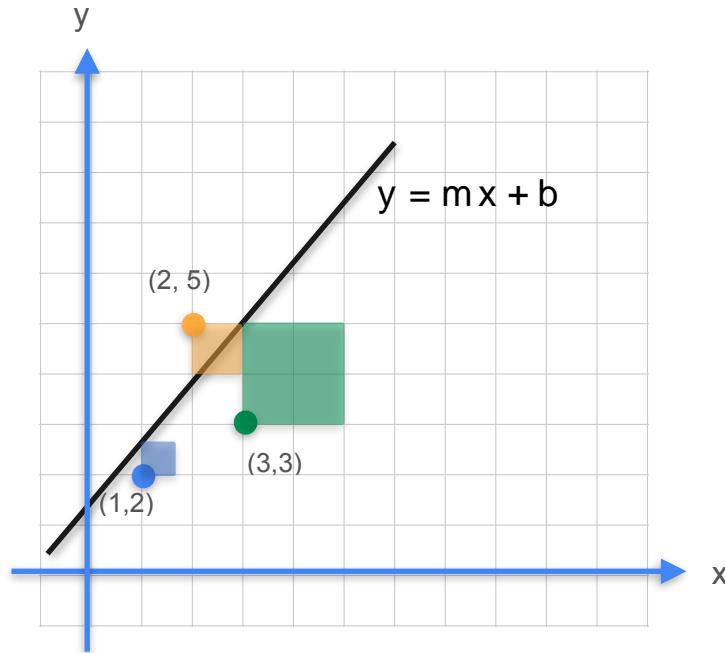
DeepLearning.AI

## Gradients and Gradient Descent

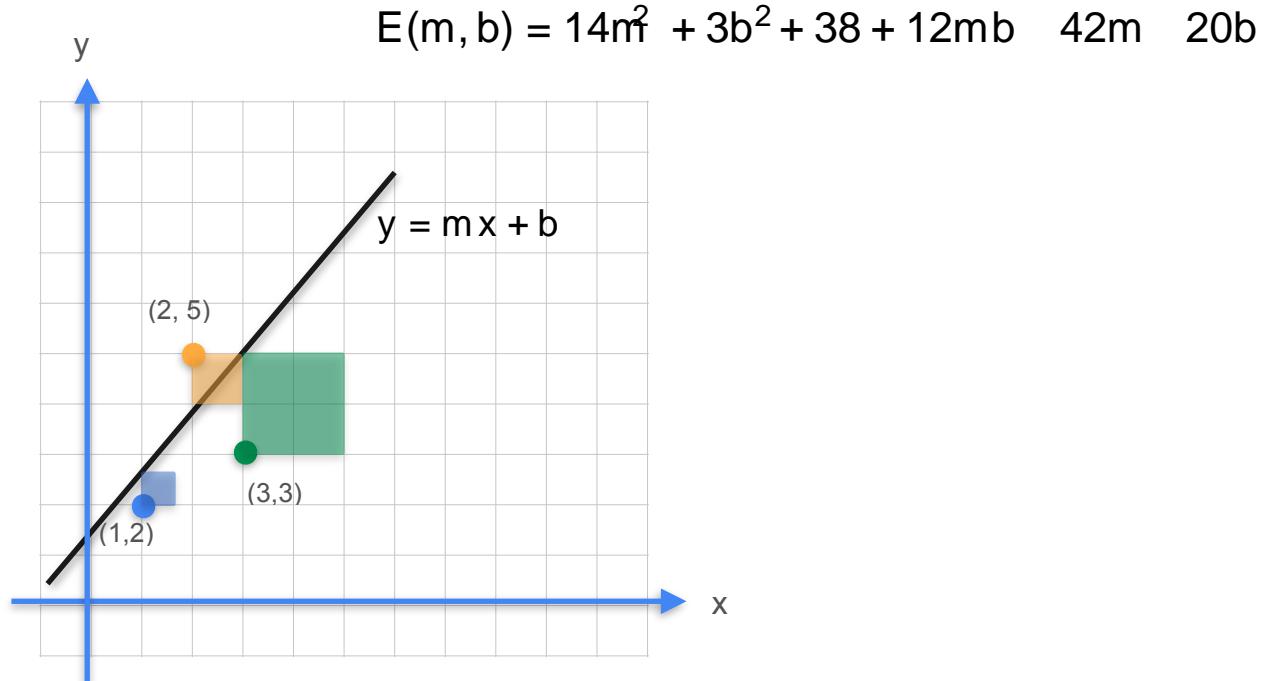
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**Optimization using Gradient  
Descent - Least squares**

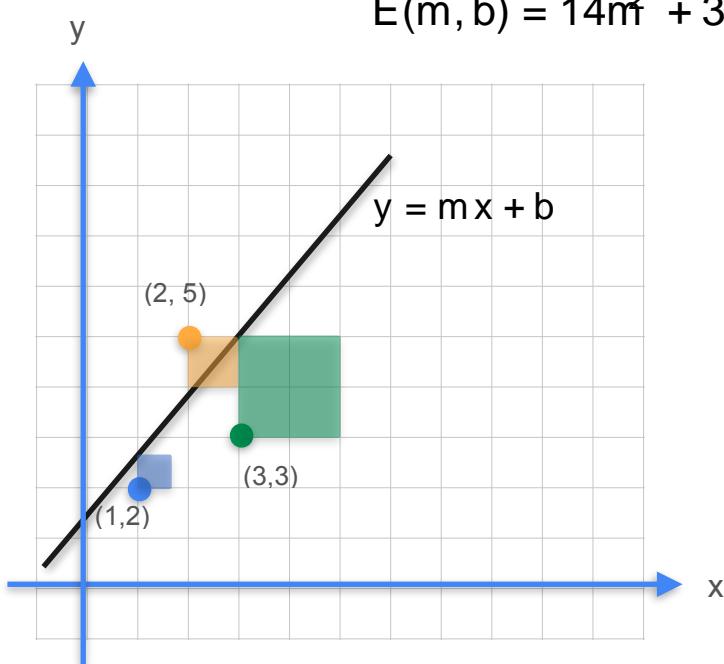
# Gradient Descent With Power Lines Example



# Gradient Descent With Power Lines Example



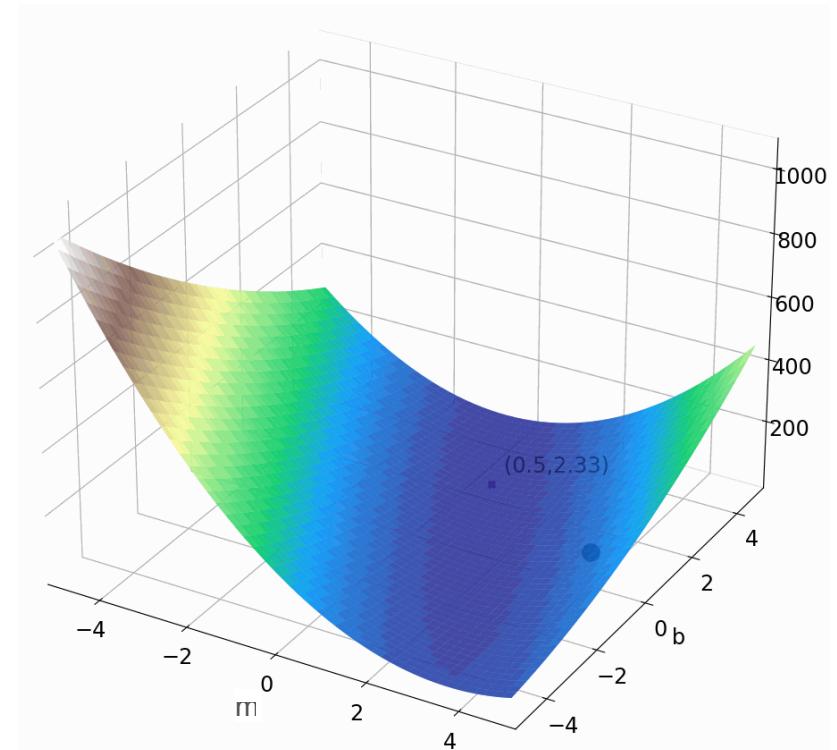
# Gradient Descent With Power Lines Example



$$E(m, b) = 14m^2 + 3b^2 + 38 + 12mb - 42m - 20b$$

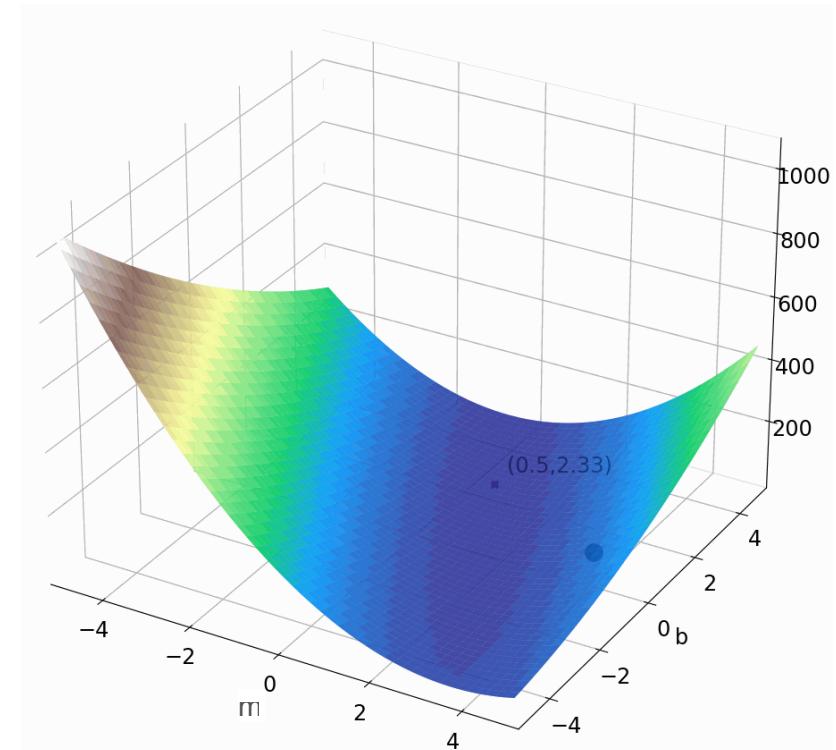
$$E(m = \frac{1}{2}, b = \frac{7}{3}) = 4.167$$

# Linear Regression: Gradient Descent



# Linear Regression: Gradient Descent

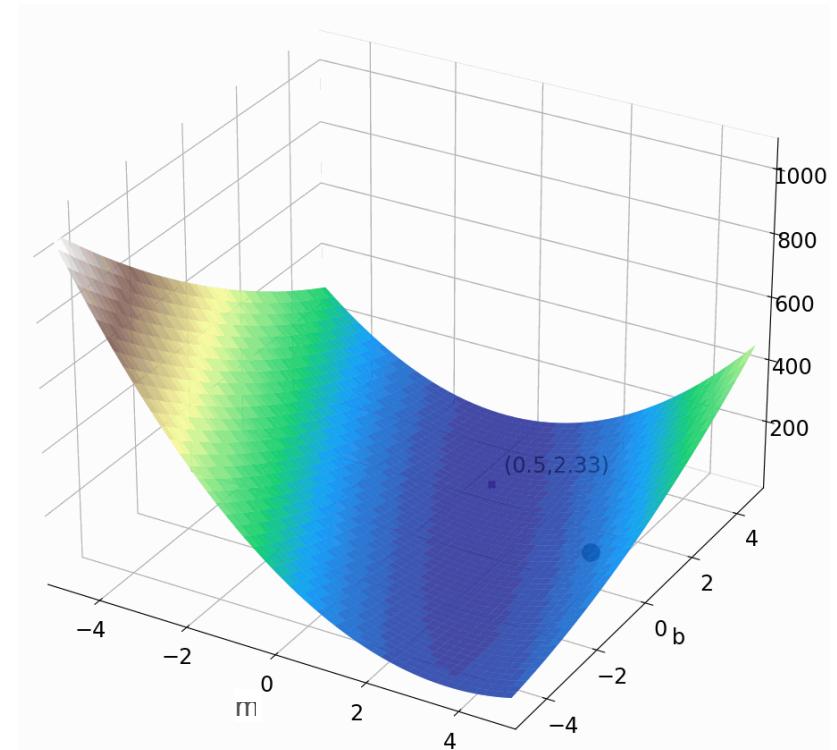
**Goal: Minimize sum of squares cost**



# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E = [28m + 12b - 42, 6b + 12m - 20]$$



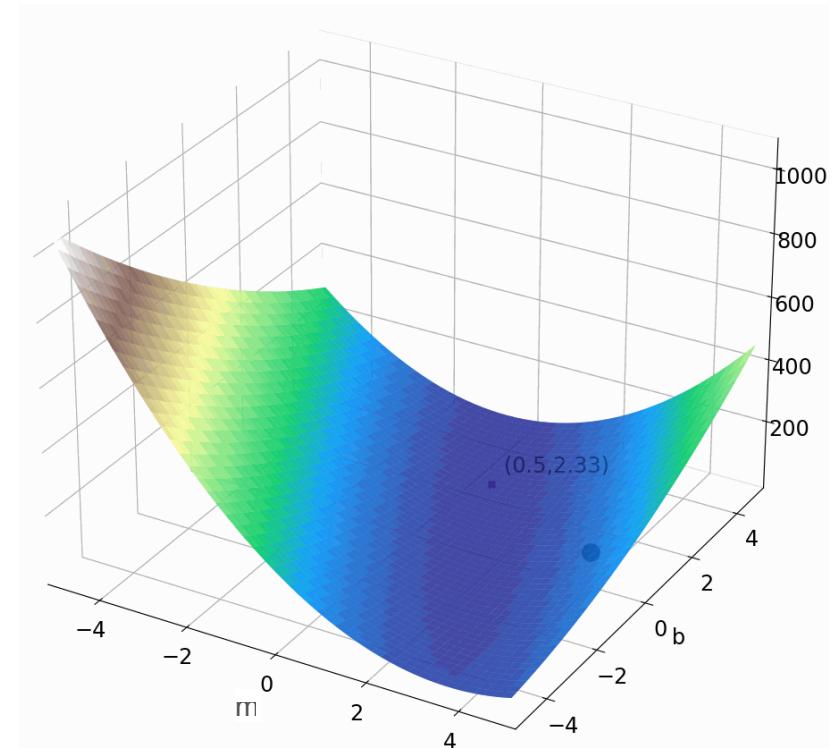
# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E = [28m + 12b - 42, 6b + 12m - 20]$$

$$m =$$

$$b =$$

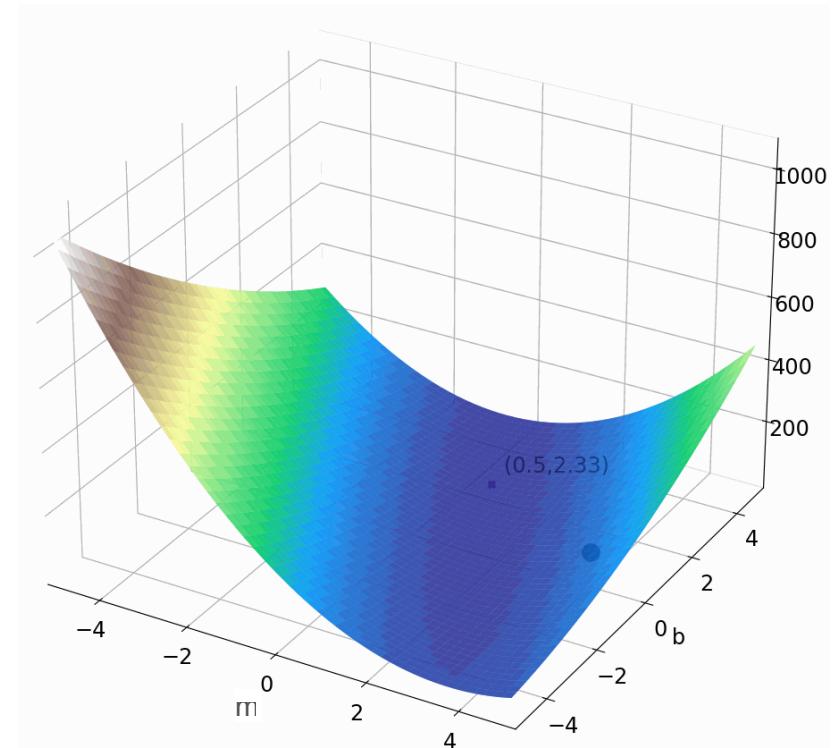


# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E = [28m + 12b - 42, 6b + 12m - 20]$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$

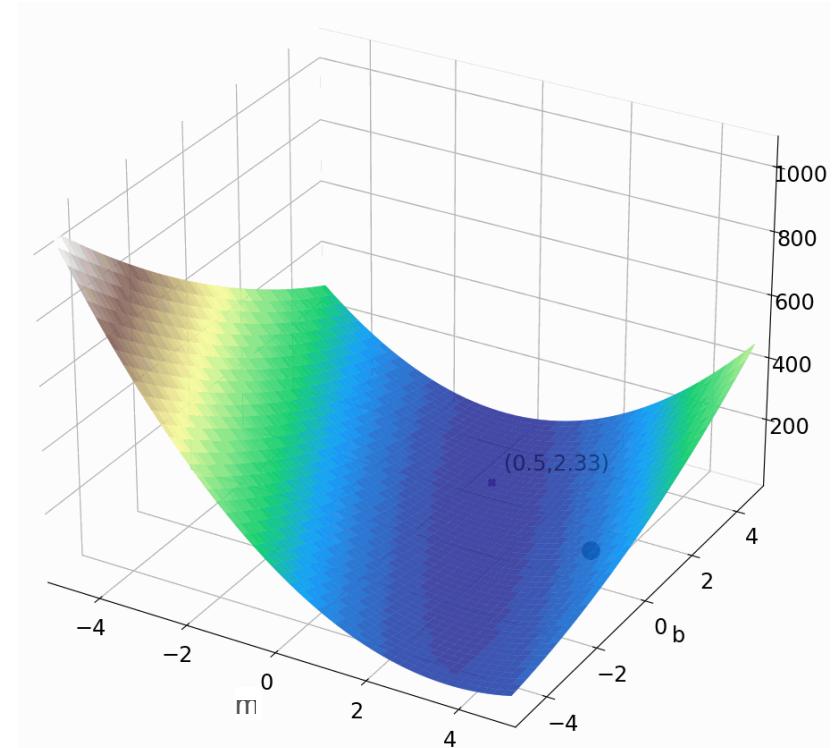


# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E = [28m + 12b - 42, 6b + 12m - 20]$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$

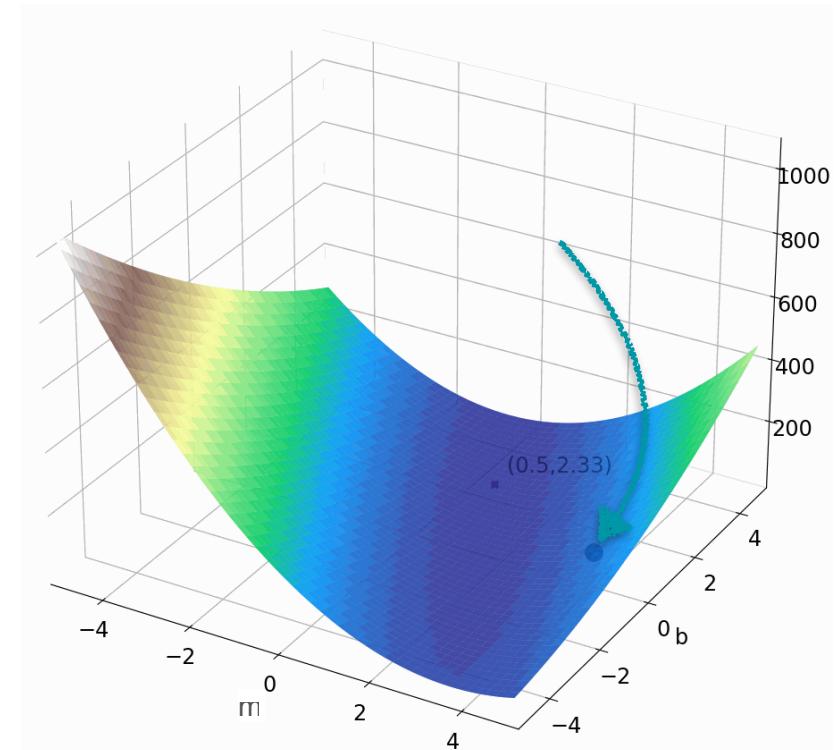


# Linear Regression: Gradient Descent

**Goal: Minimize sum of squares cost**

$$E = [28m + 12b - 42, 6b + 12m - 20]$$

$$\begin{aligned} m &= ? \\ b &= ? \end{aligned}$$

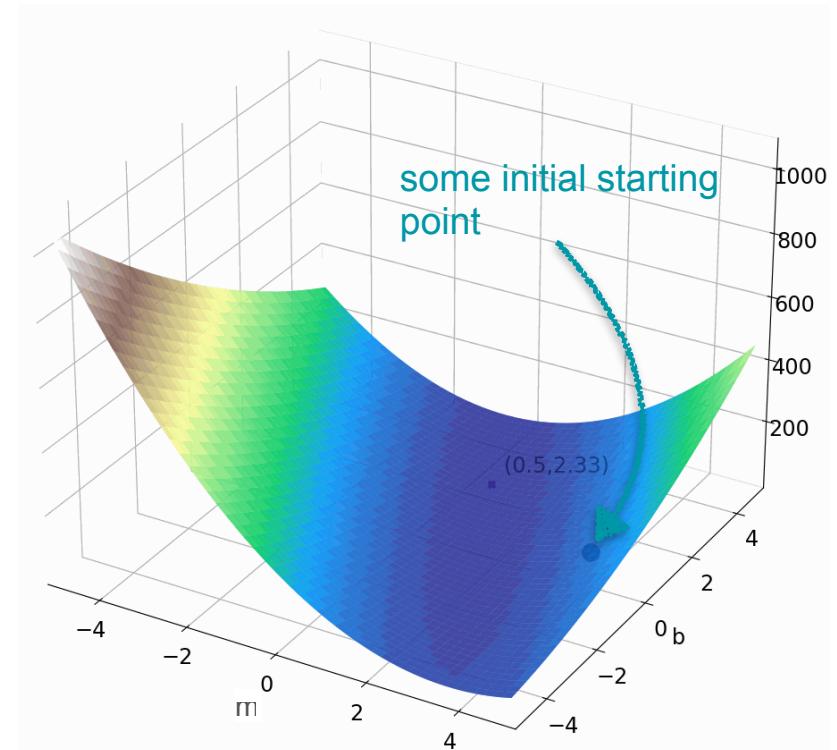


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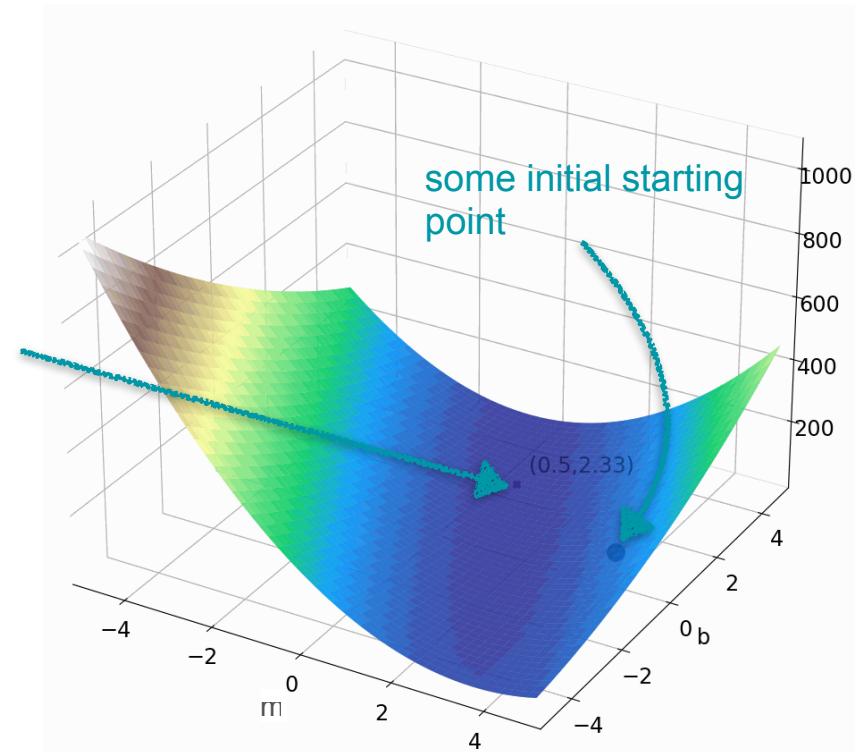


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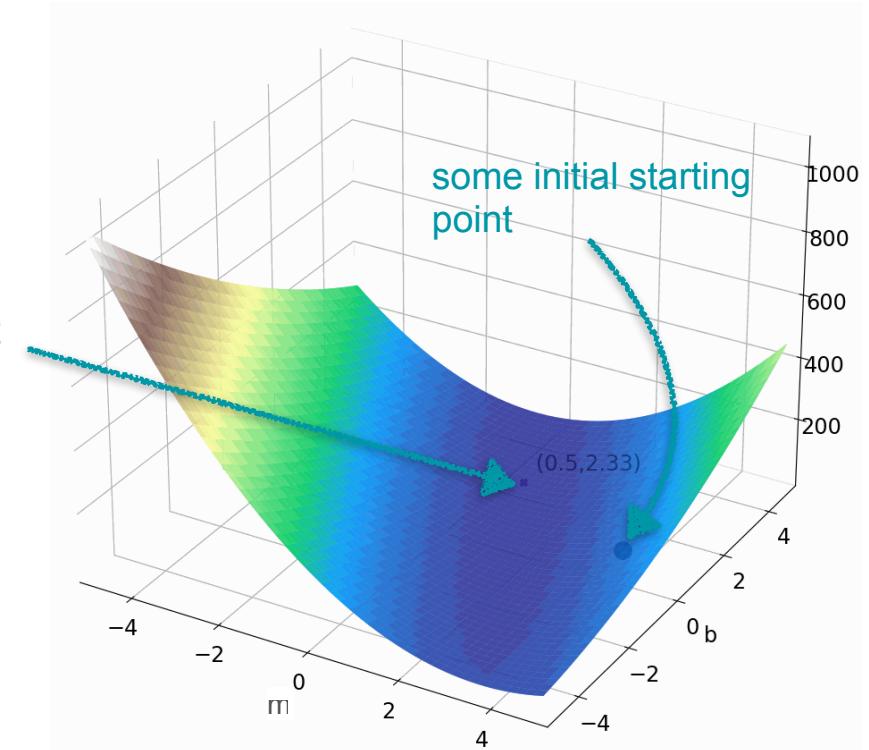
# Linear Regression: Gradient Descent

Goal: Minimize sum of squares cost

$$E = [28m + 12b - 42, 6b + 12m - 20]$$

$$\begin{matrix} m = \\ b = \end{matrix} ?$$

The points  $m, b$  such that  
the cost is minimum



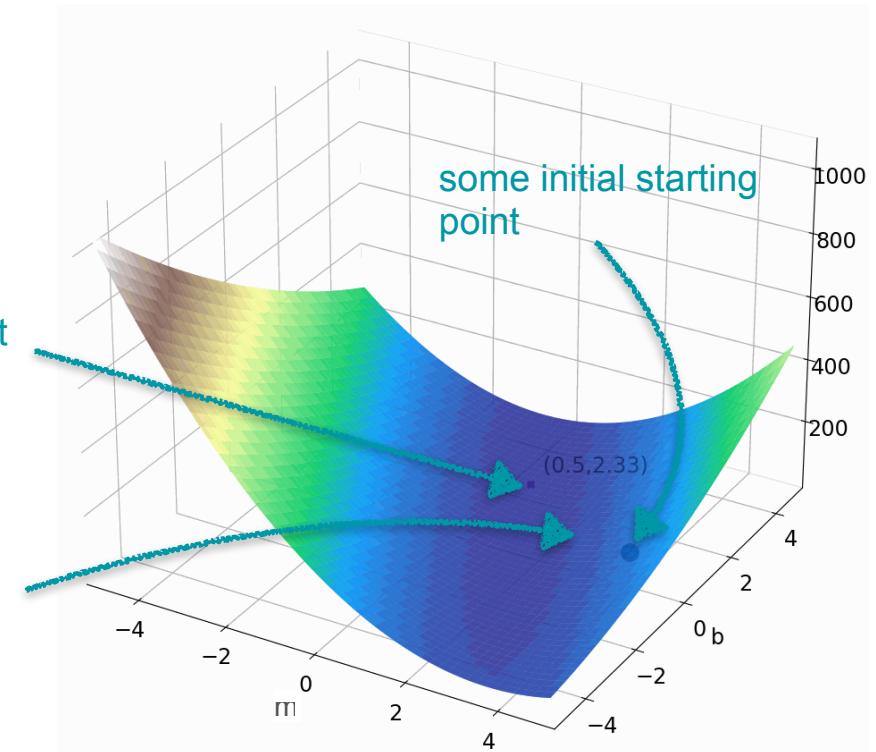
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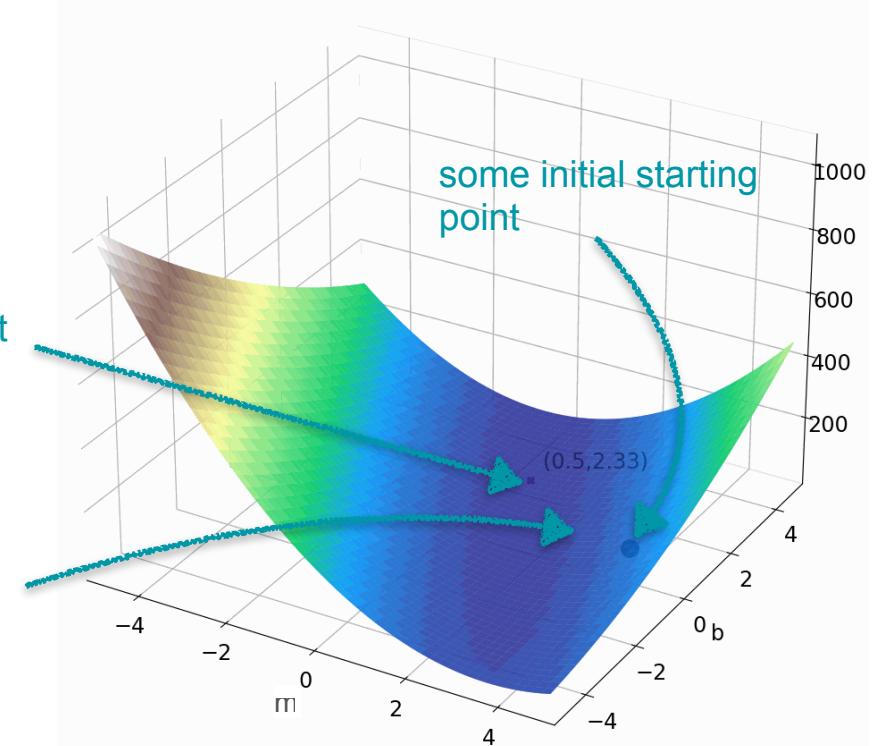
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descend until you  
find the minimum



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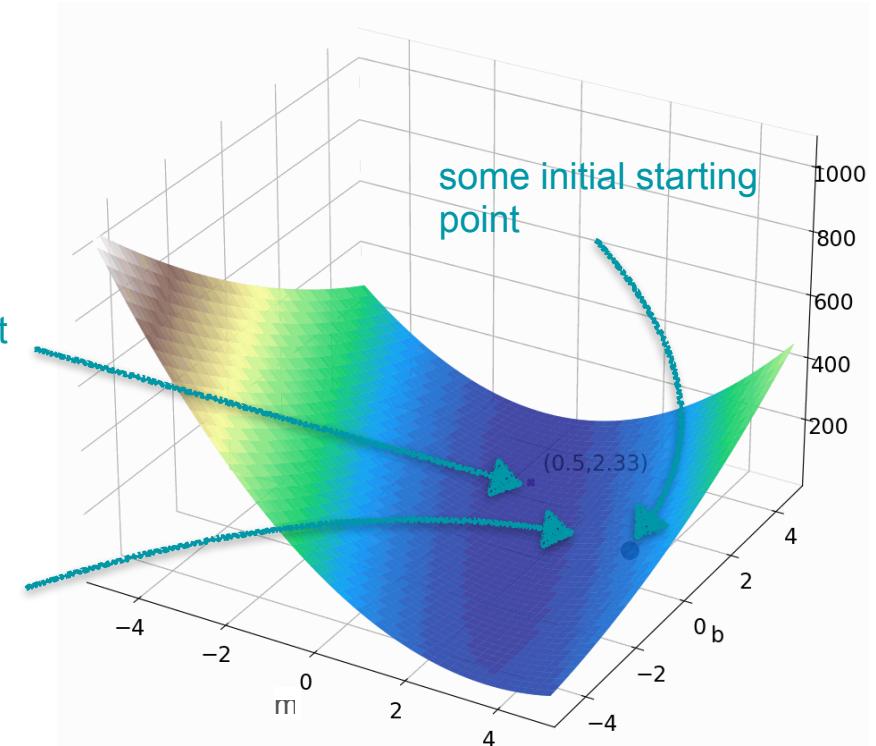
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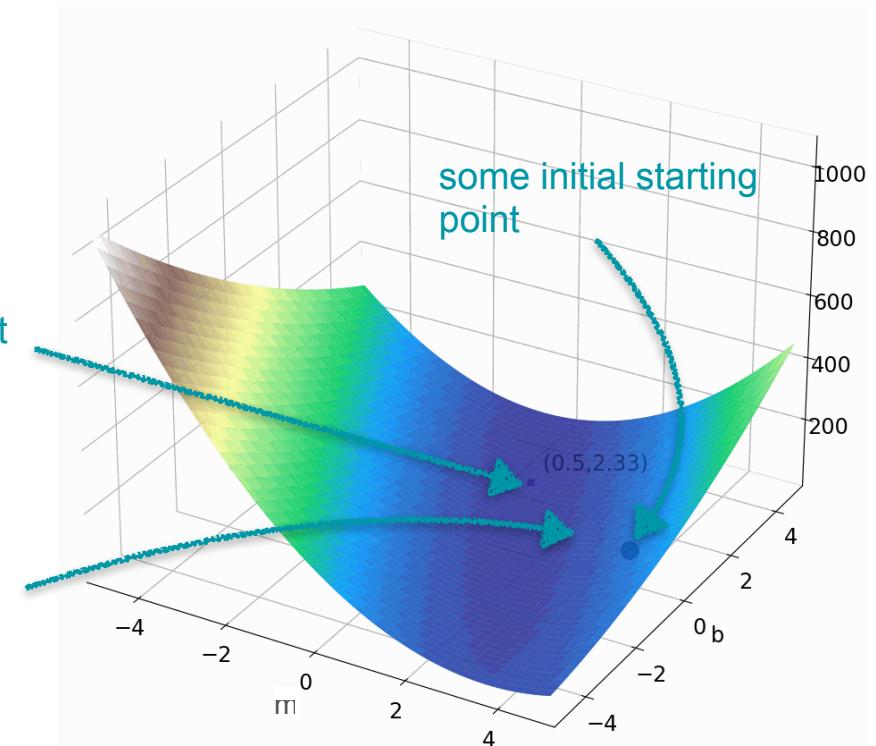
$$\begin{matrix} m = \\ b = \end{matrix} ?$$

Steps:

Start with  $(m_0, b_0)$

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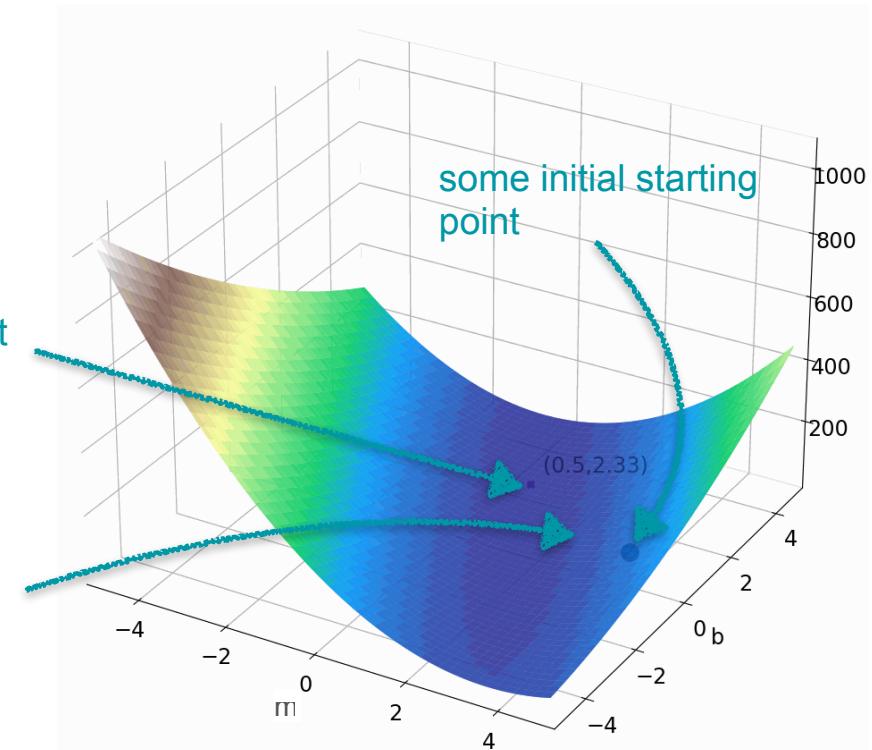
Steps:

Start with  $(m_0, b_0)$

descend until you  
find the minimum

Iterate  
 $(m_{k+1}, b_{k+1}) = (m_k, b_k)$

$E(m_k, b_k)$





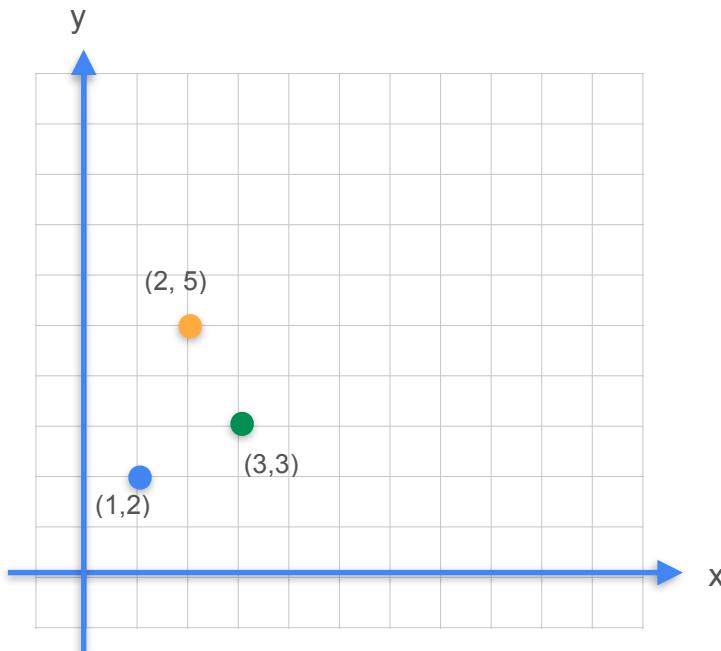
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## Gradients and Gradient Descent

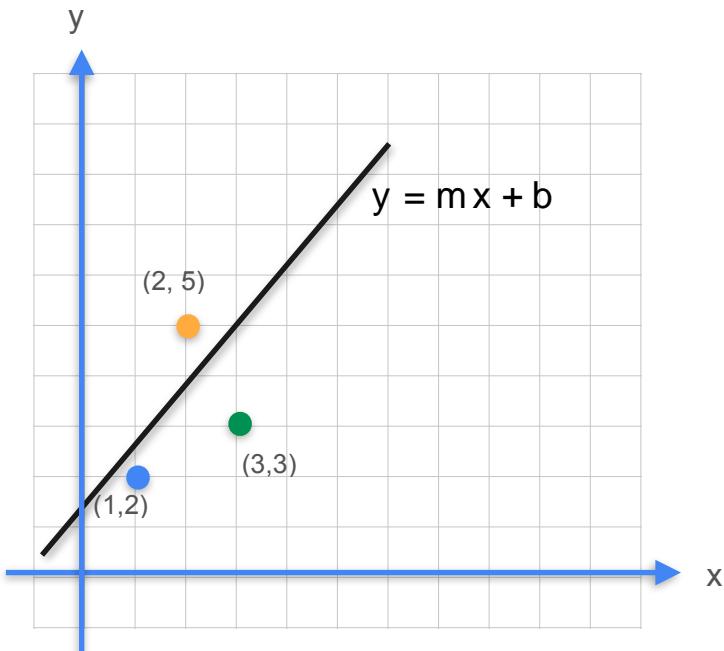
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**Optimization using Gradient  
Descent - Least squares  
with multiple observations**

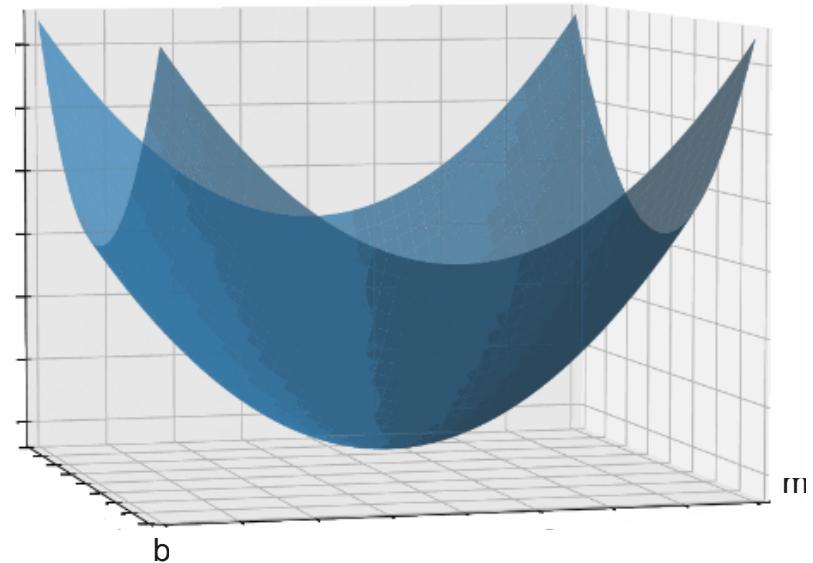
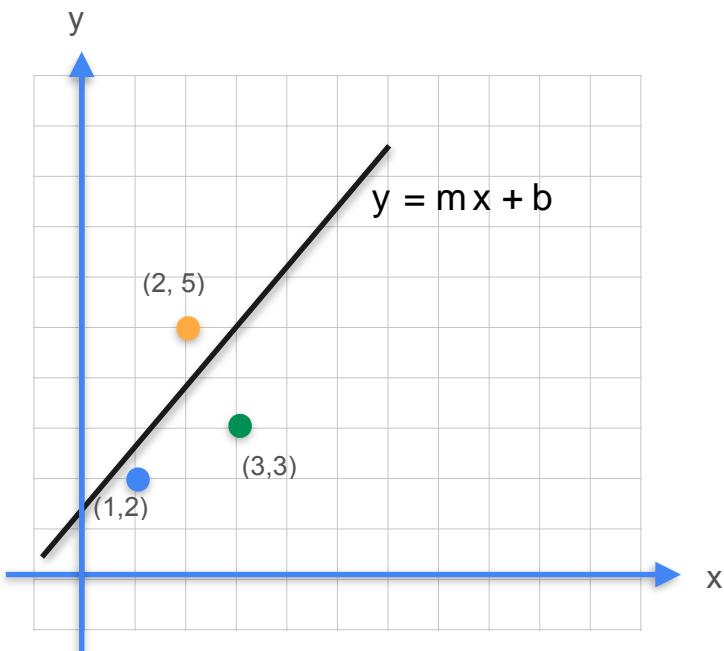
# Gradient Descent



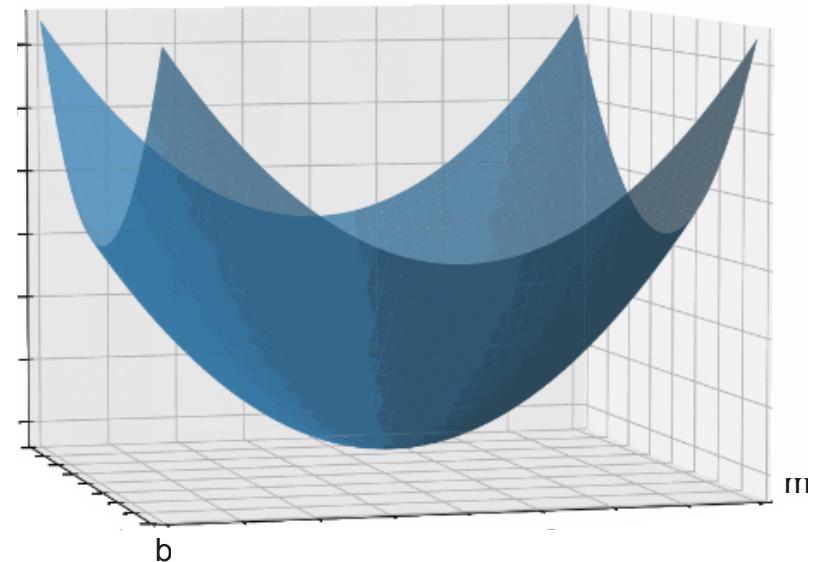
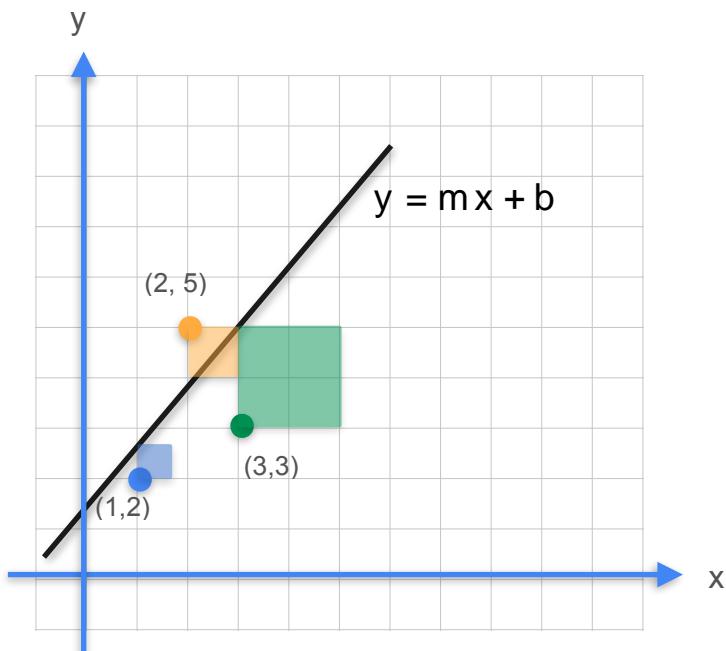
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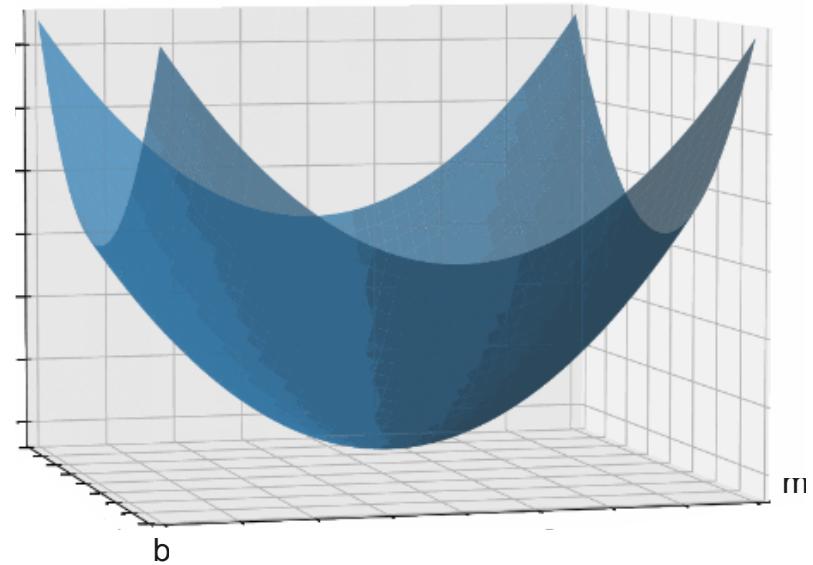
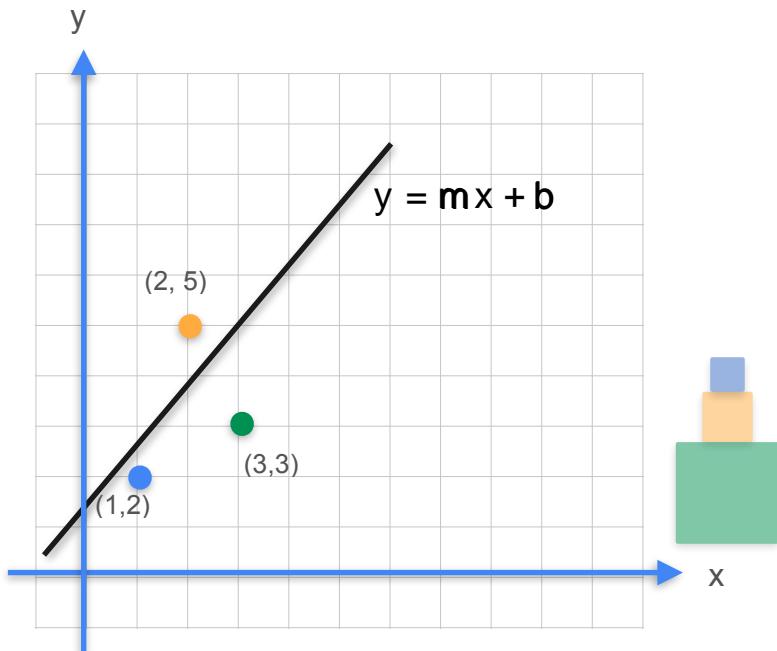
# Gradient Descent



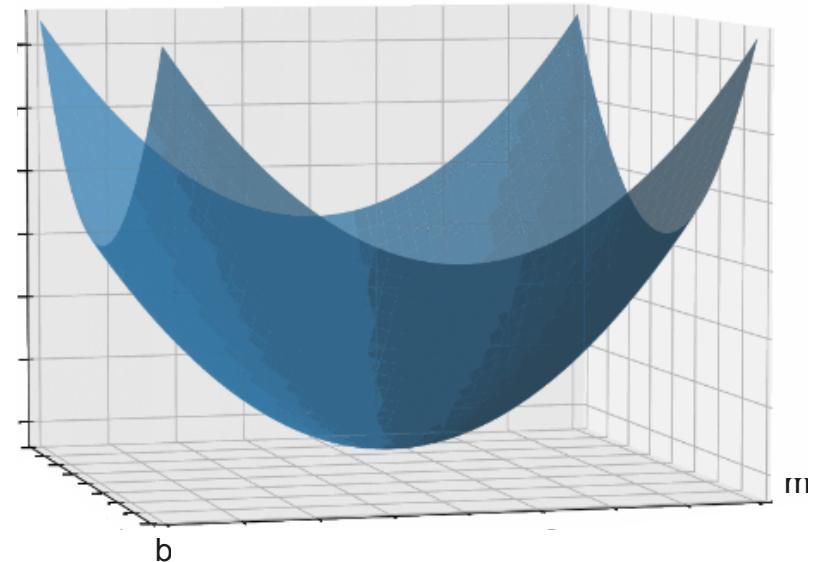
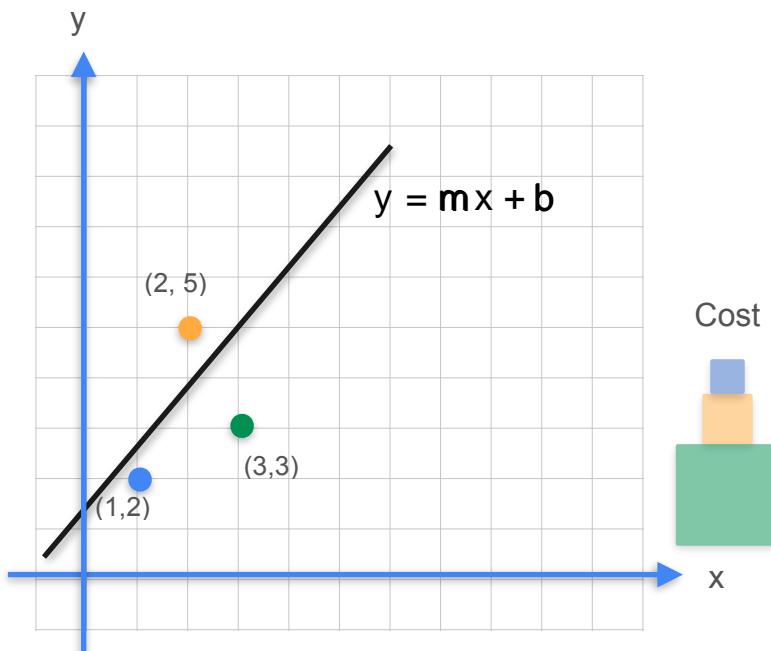
# Gradient Descent



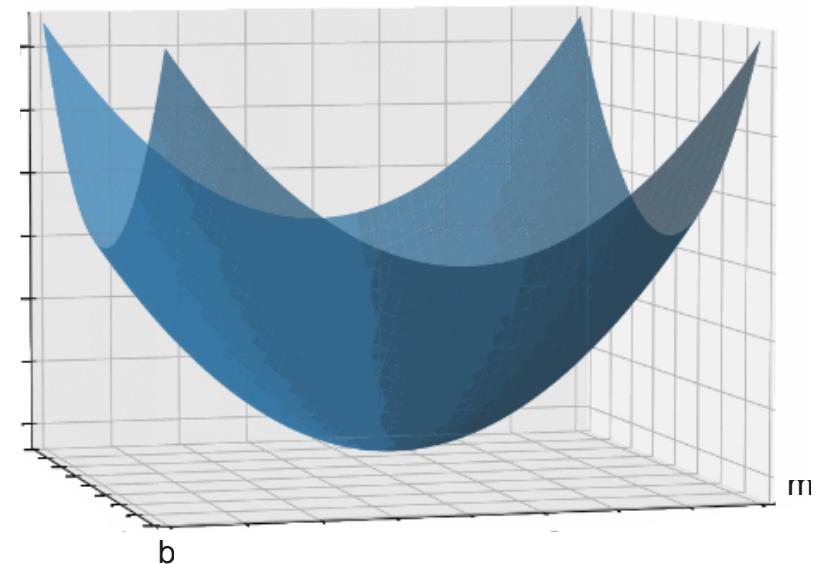
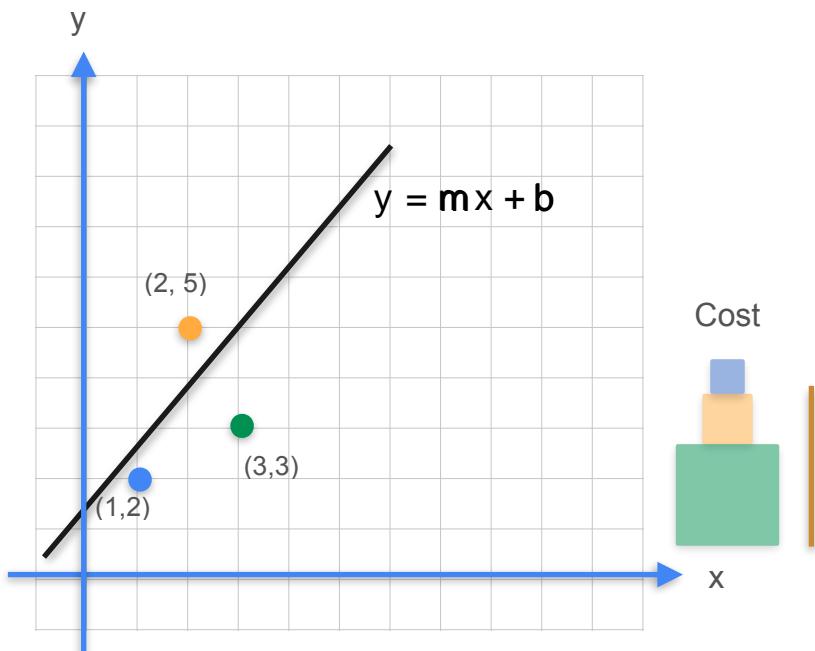
# Gradient Descent



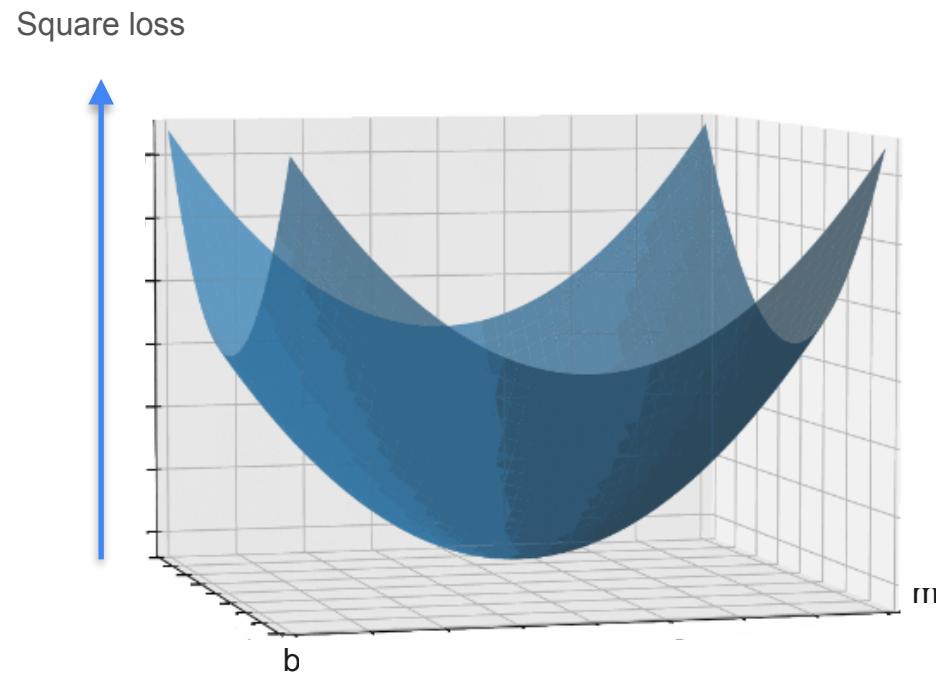
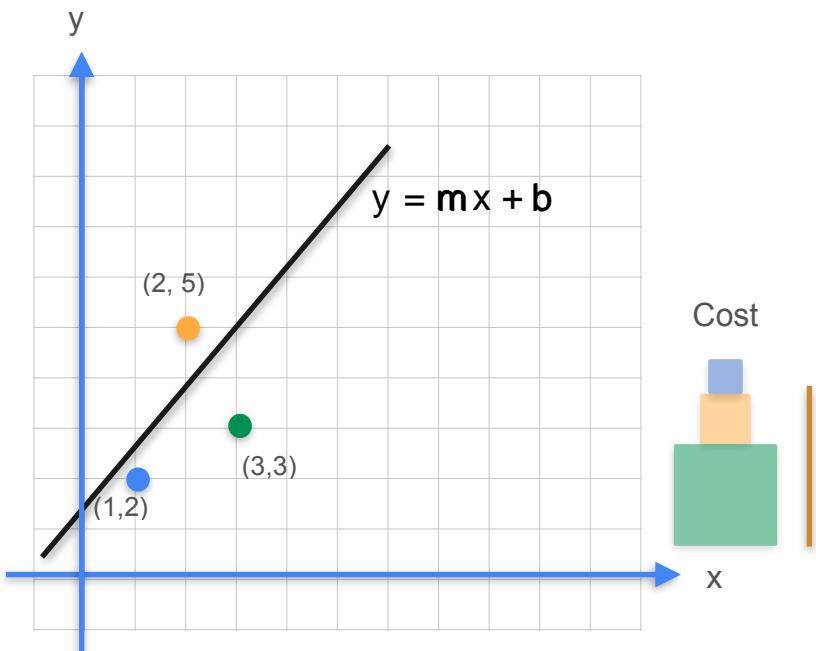
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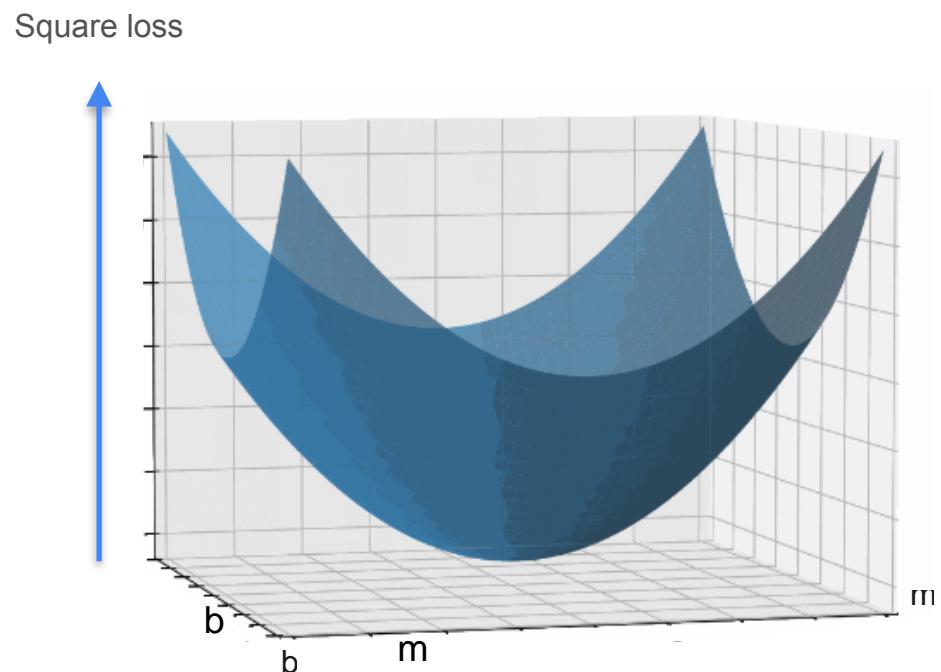
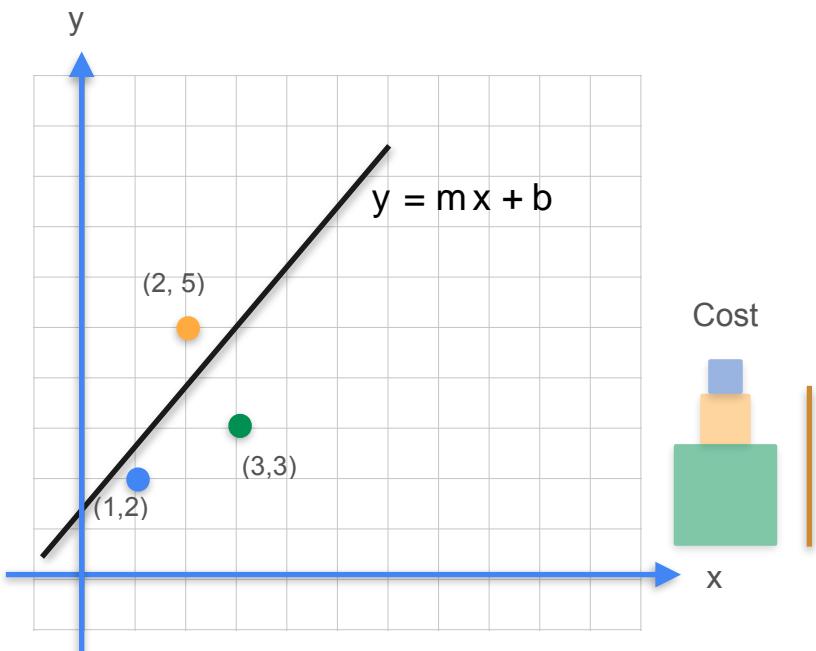
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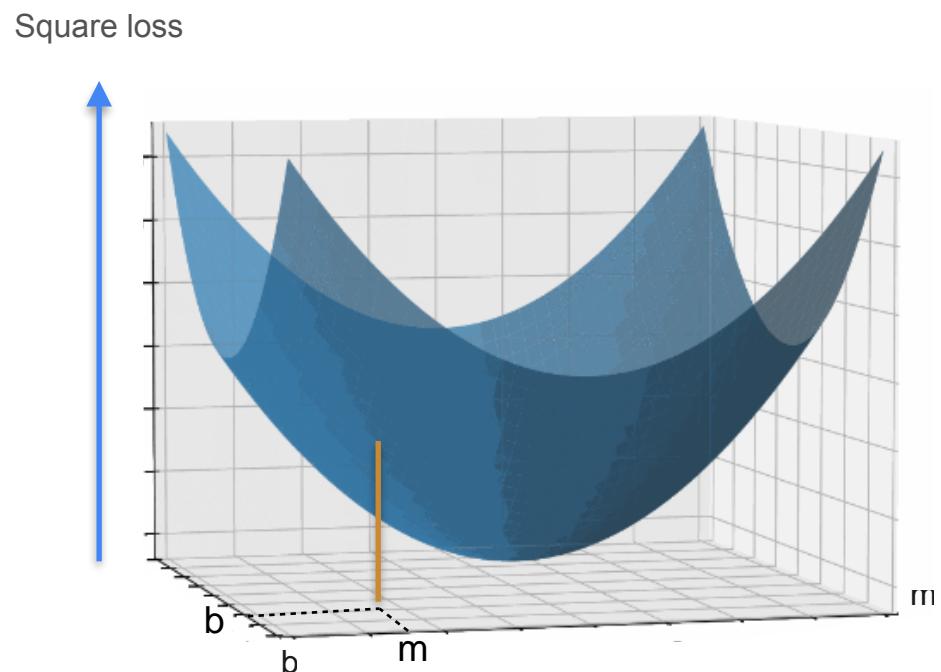
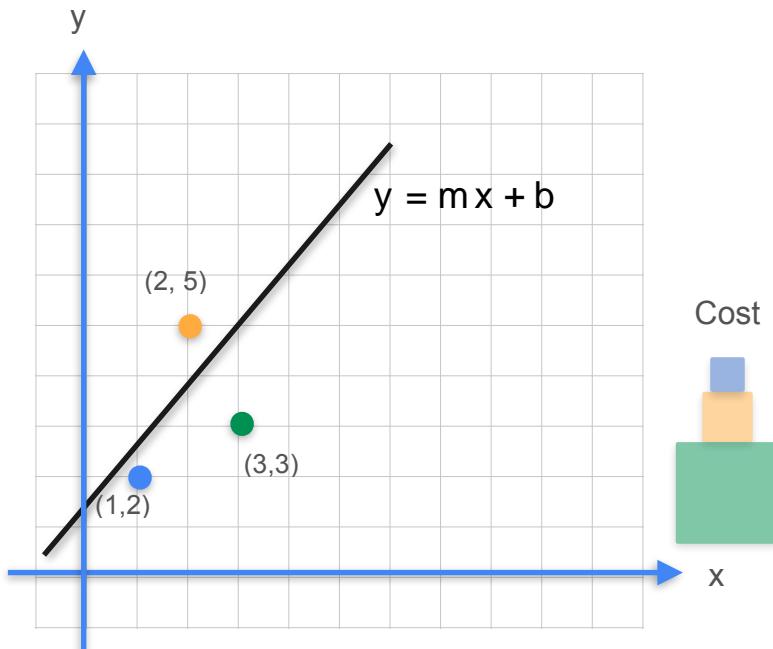
# Gradient Descent



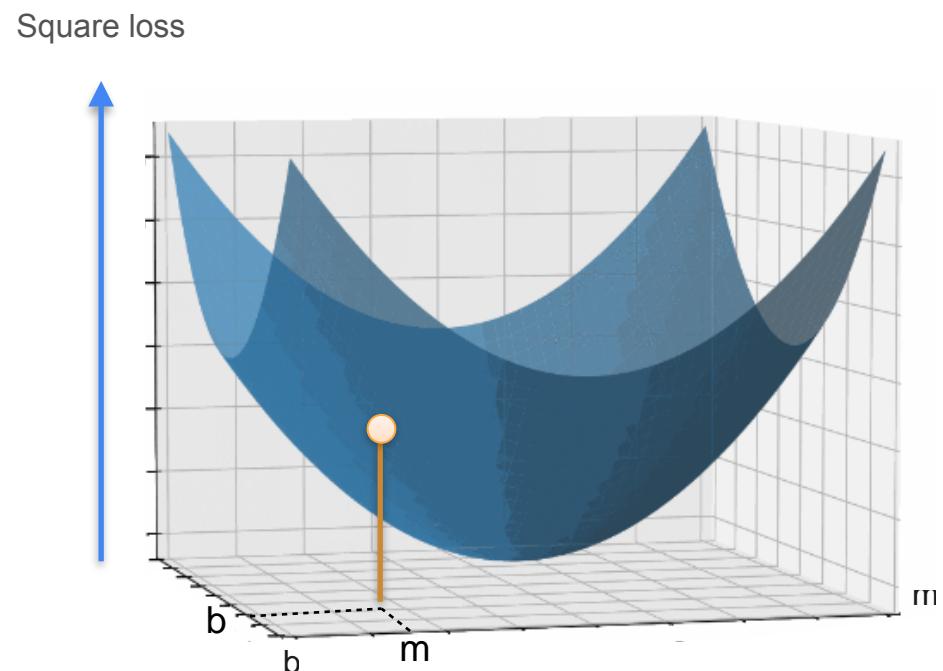
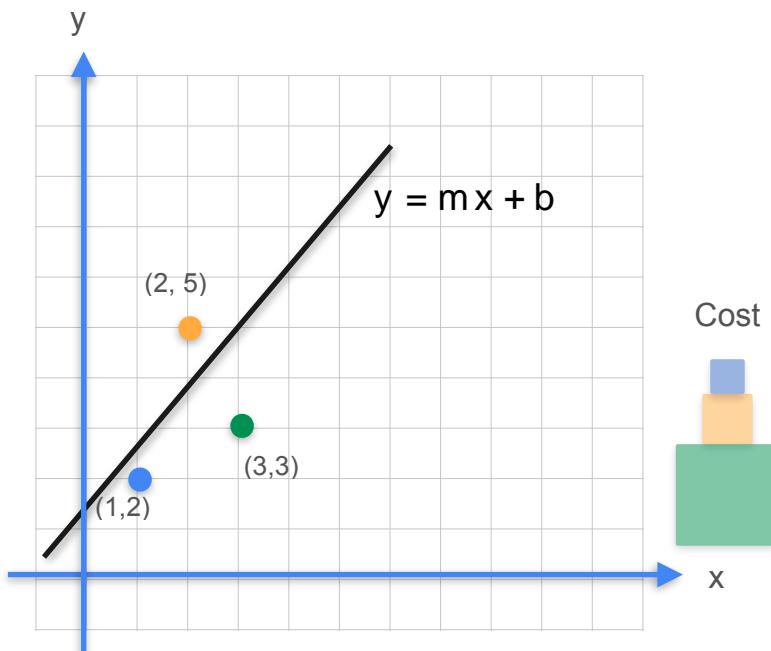
# Gradient Descent



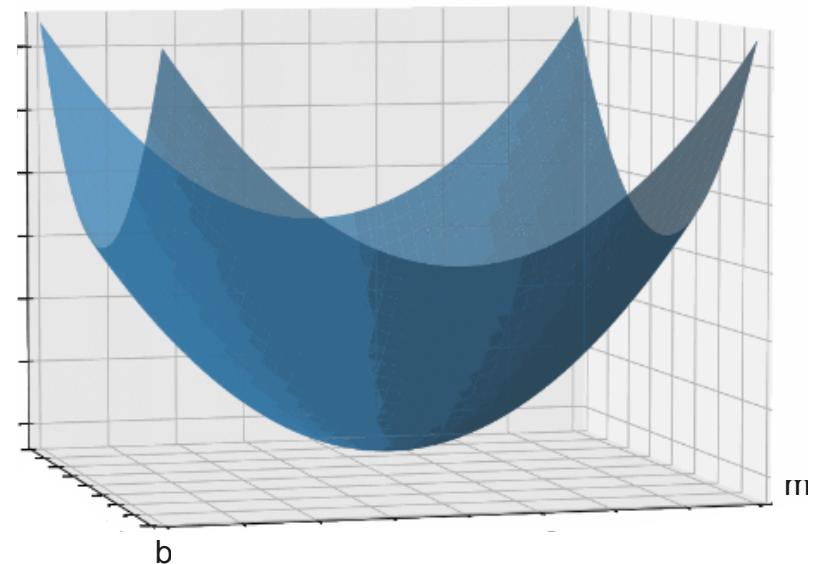
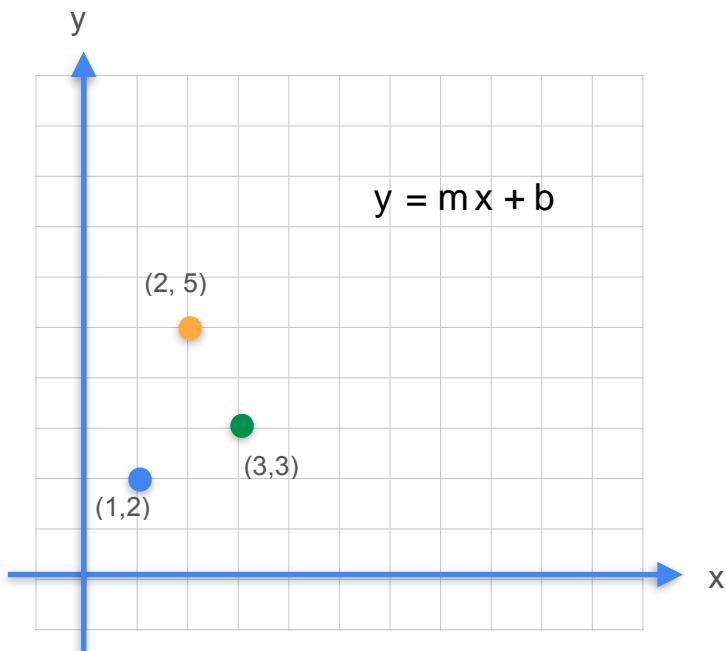
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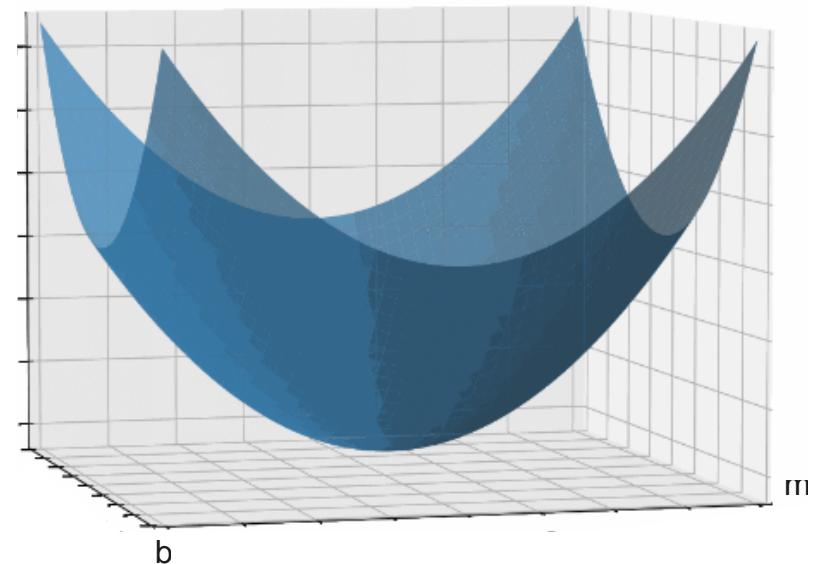
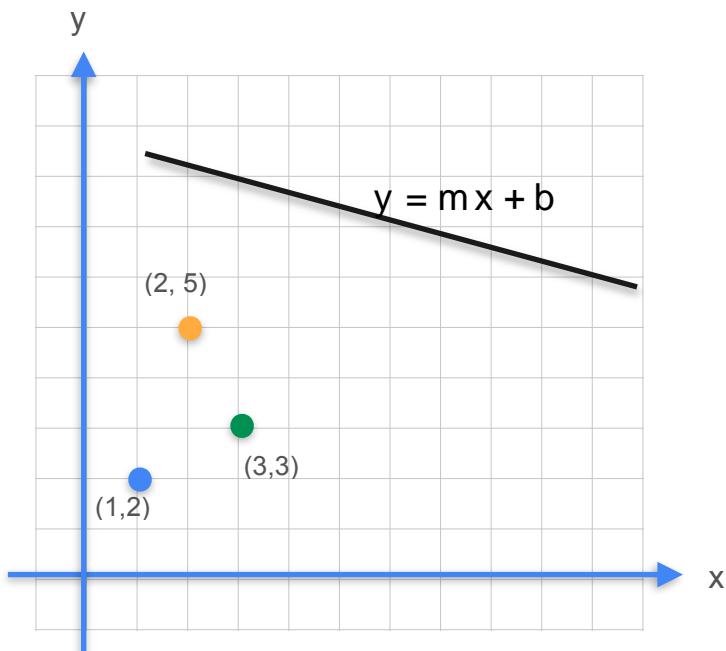
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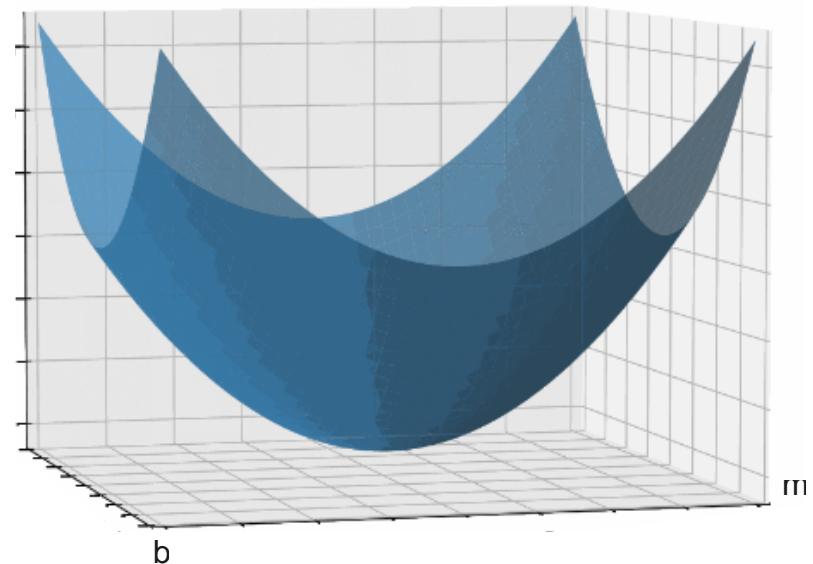
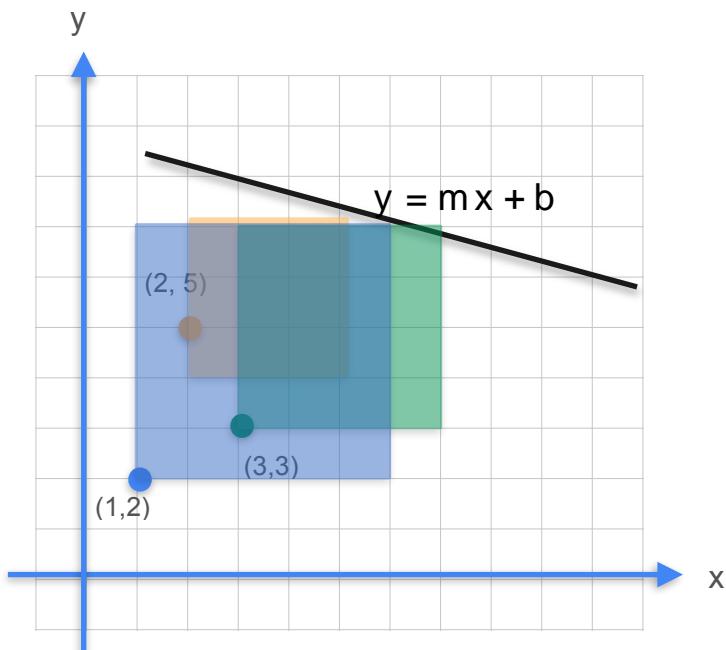
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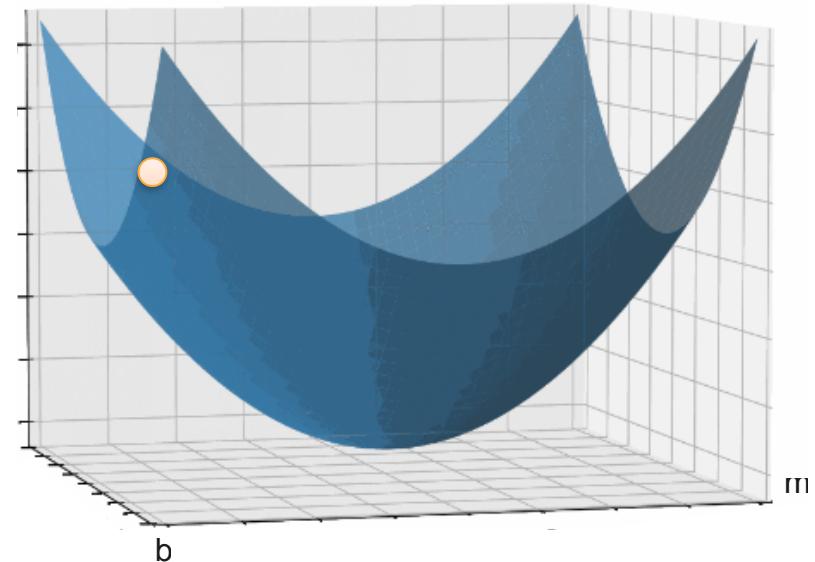
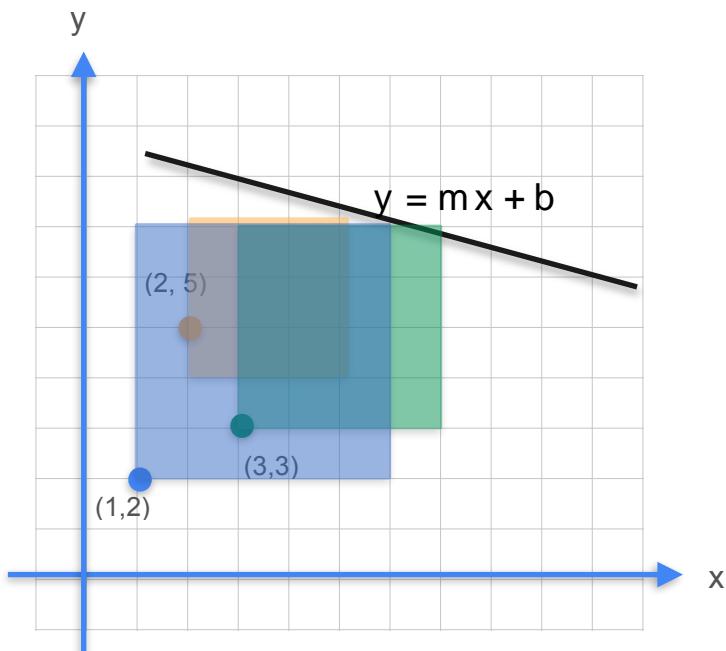
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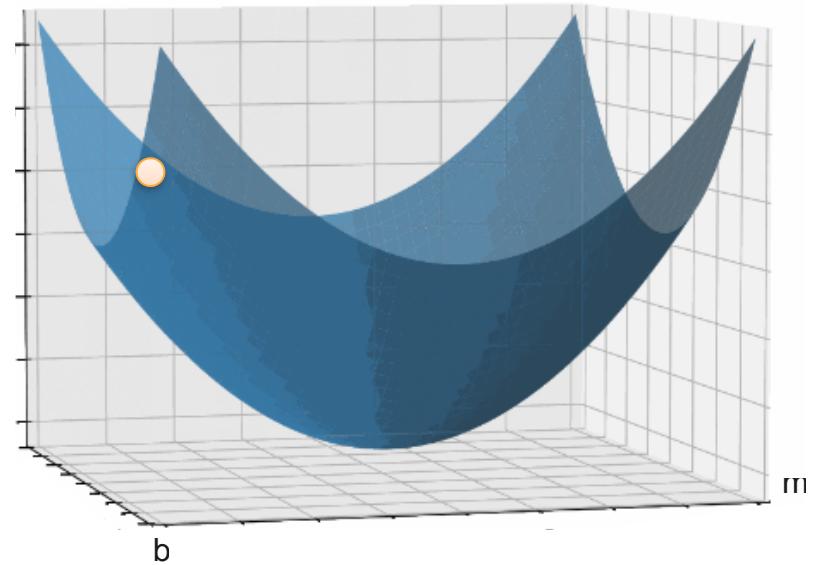
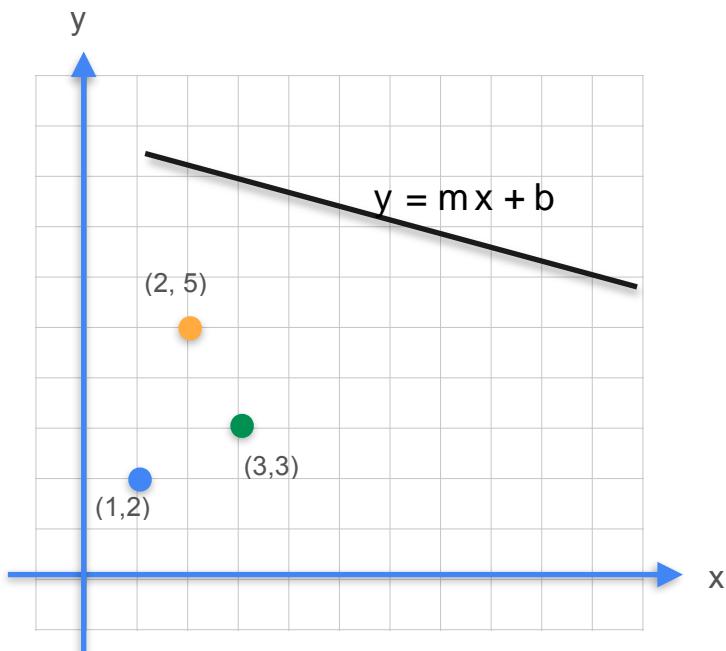
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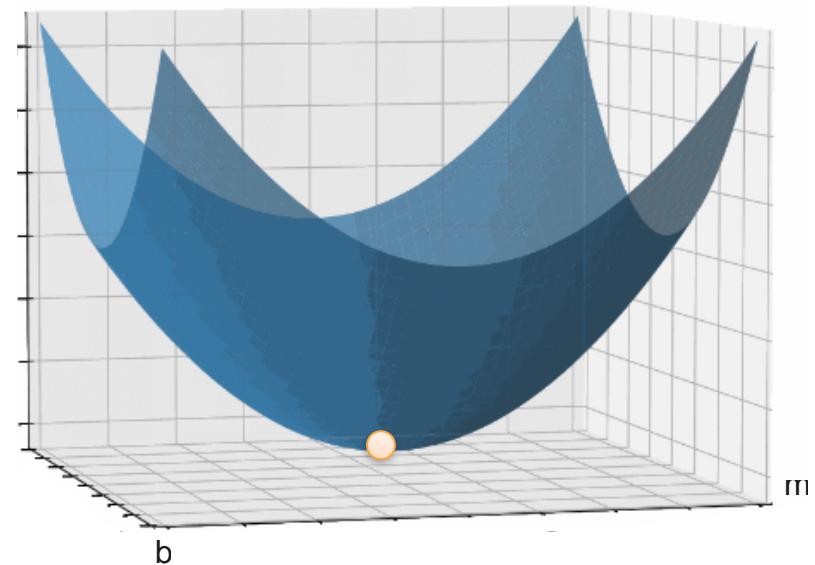
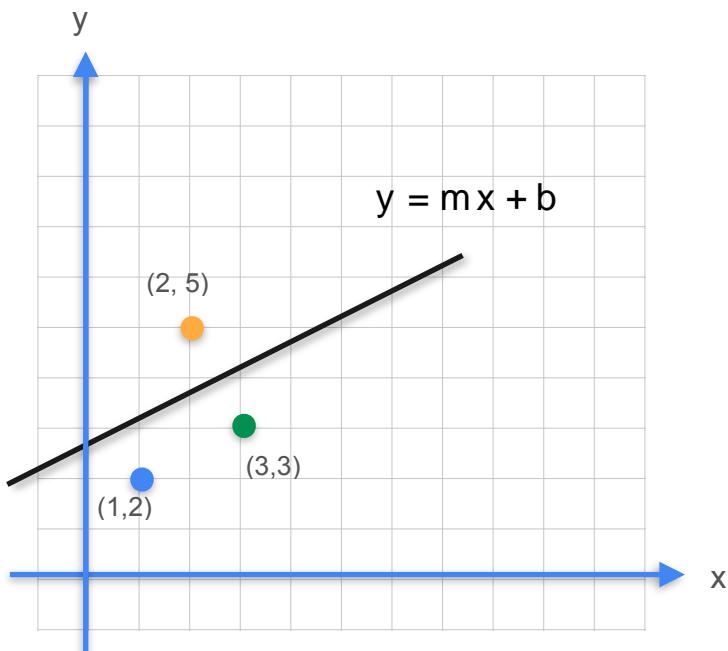
# Gradient Descent



# Gradient Descent



# Gradient Descent



# Another Example

# Another Example



# Another Example



TV advertisement  
budget

## Another Example



TV advertisement  
budget



## Another Example



TV advertisement  
budget



Number of sales

# Another Example

# Another Example

TV budget

Sales

# Another Example

TV budget	Sales
230.1	22.1

# Another Example

TV budget	Sales
230.1	22.1
44.5	10.4

# Another Example

TV budget	Sales
230.1	22.1
44.5	10.4
17.2	9.3

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Goal: Predict sales in terms of TV budget

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Tool: Linear regression

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$$y = mx + b$$

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TV budget	Sales
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Goal: Predict sales in terms of TV budget

Tool: Linear regression

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# Another Example

TV budget	Sales
230.1	22.1
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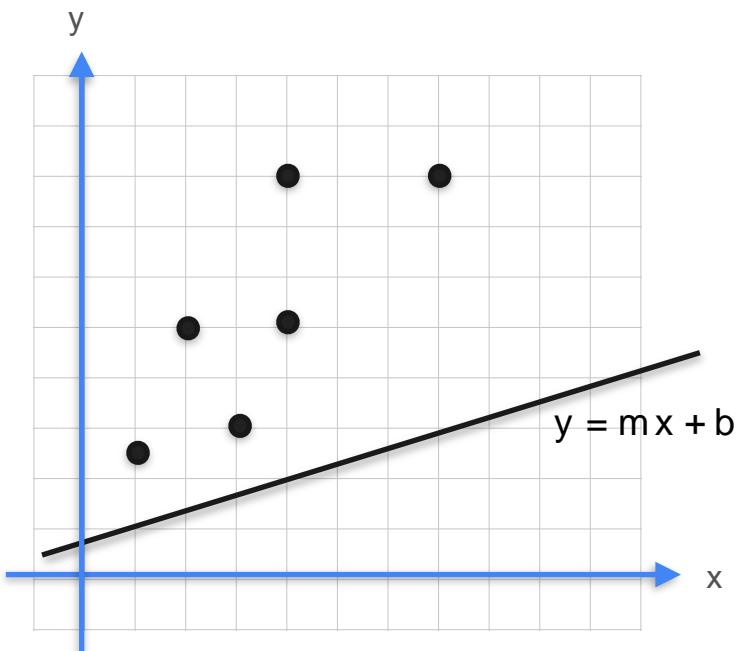
Multiple observations

Goal: Predict sales in terms of TV budget

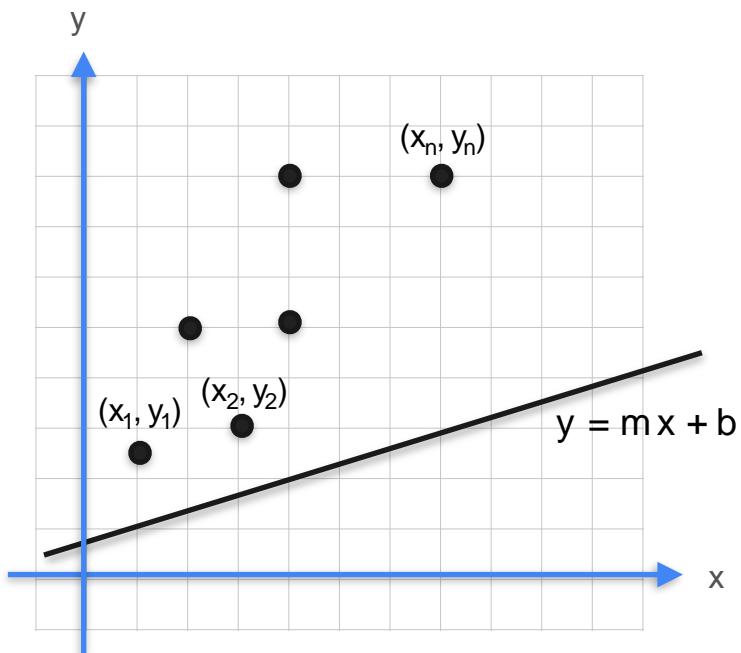
Tool: Linear regression

$$y = mx + b$$

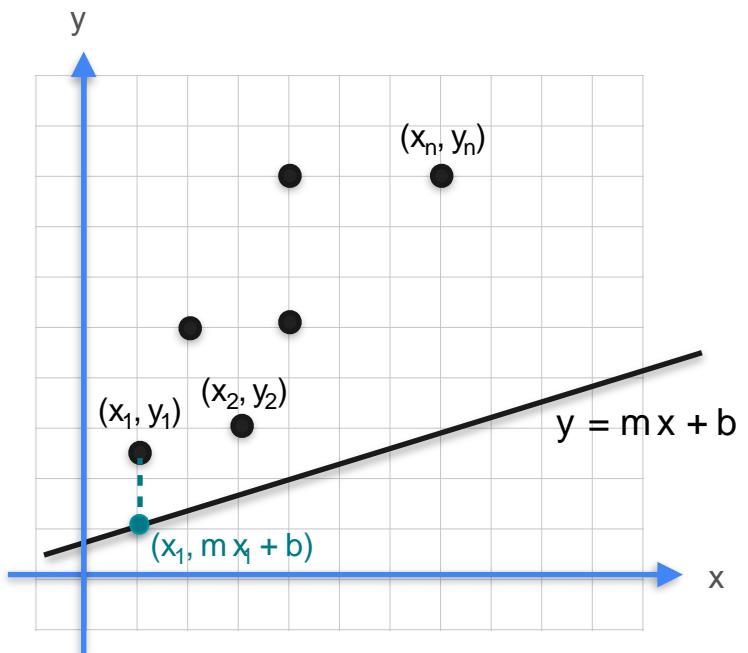
# Gradient Descent



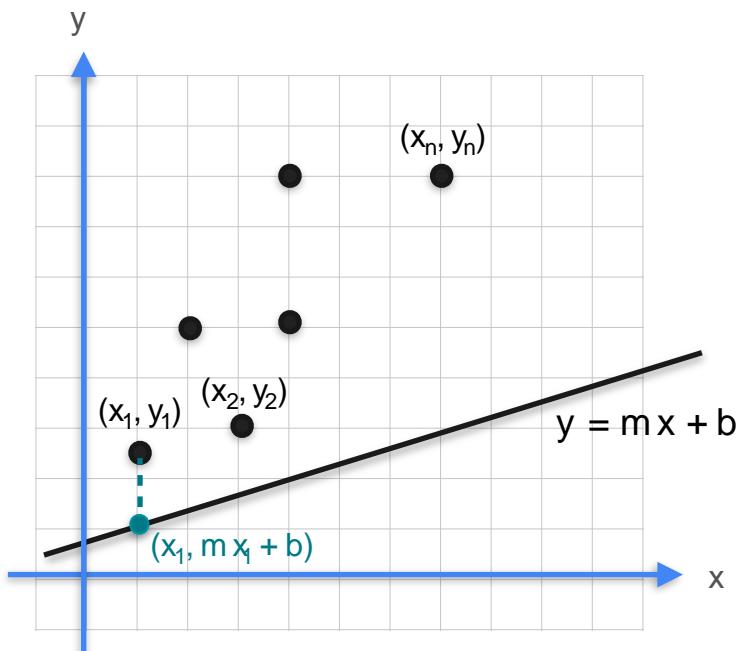
# Gradient Descent



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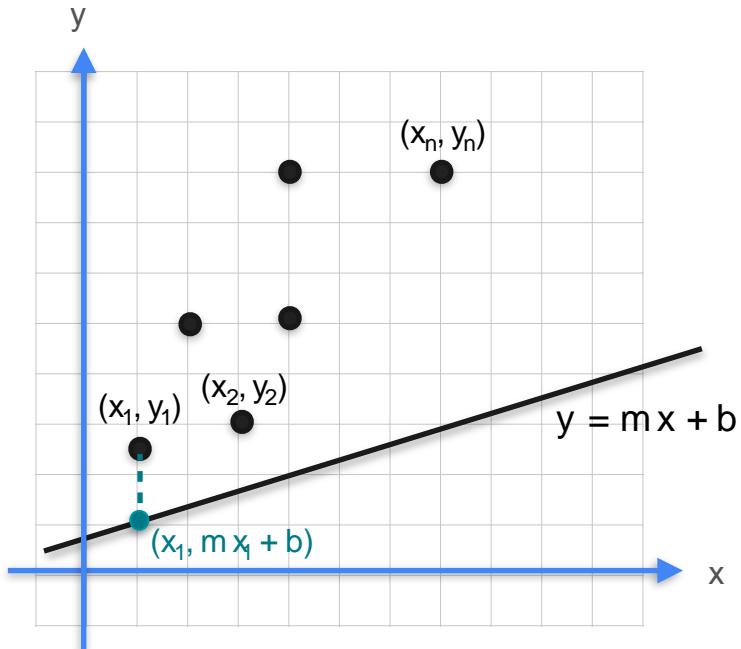


# Gradient Descent



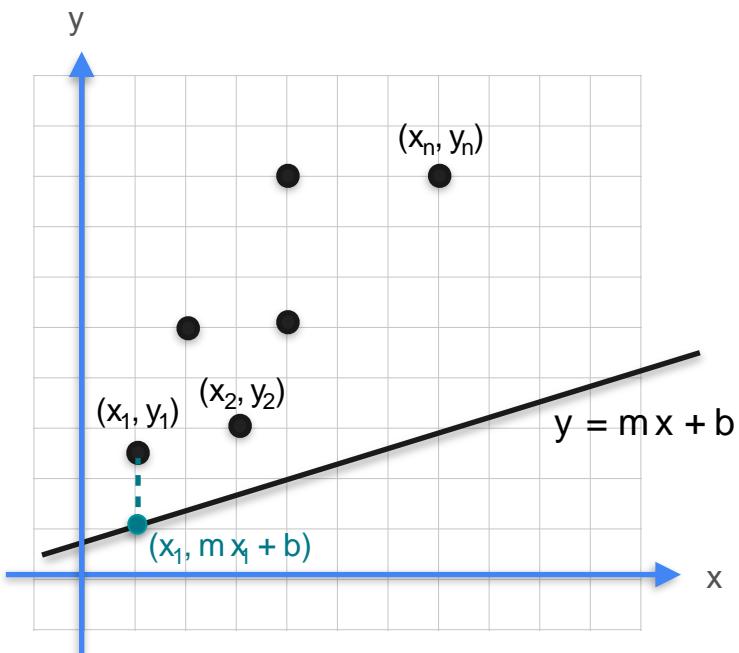
Loss  
↓  
 $mx_1 + b - y_1$

# Gradient Descent



Loss  
↓  
 $(mx_1 + b - y_1)^2$

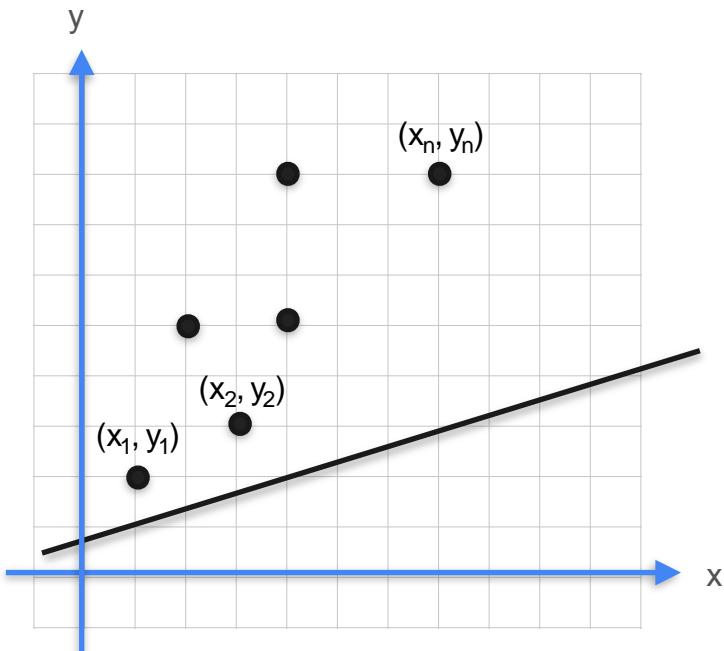
# Gradient Descent



Cost ↓      Loss ↓

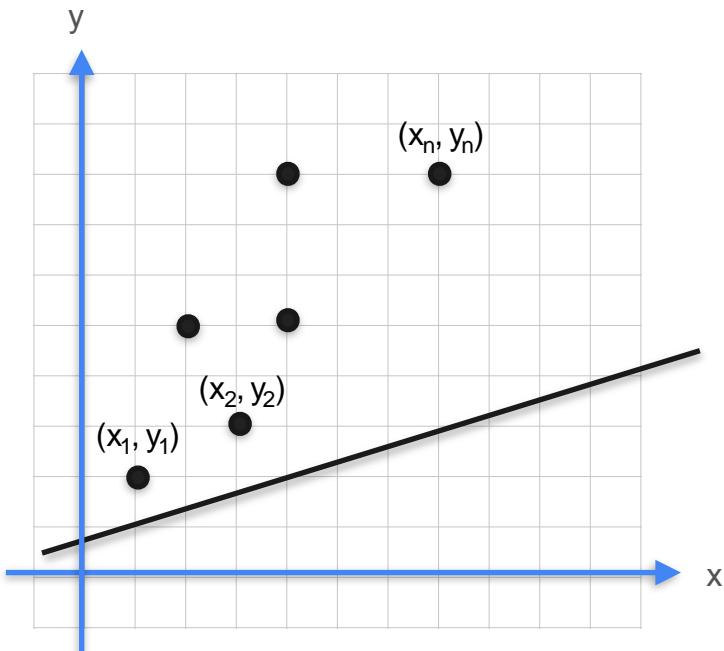
$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

# Gradient Descent



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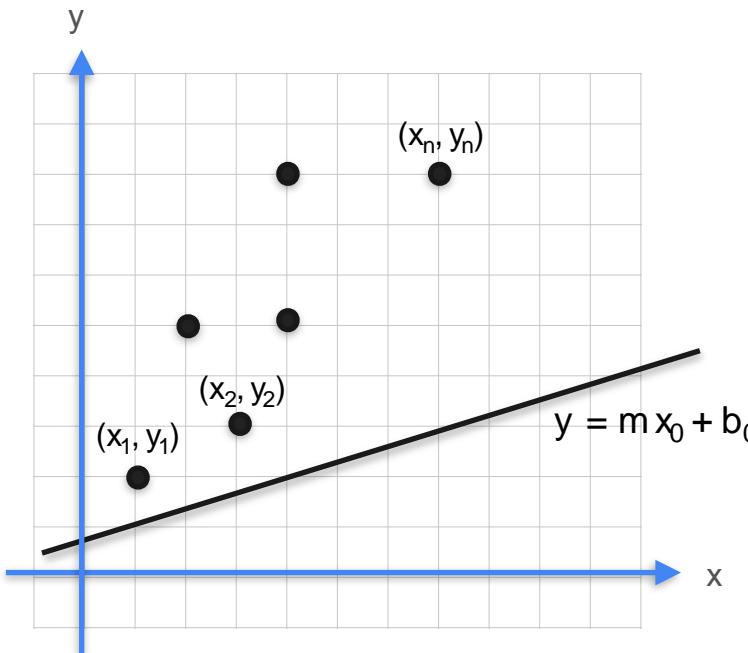
# Gradient Descent



$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$m_0$   
[  $b_0$  ]

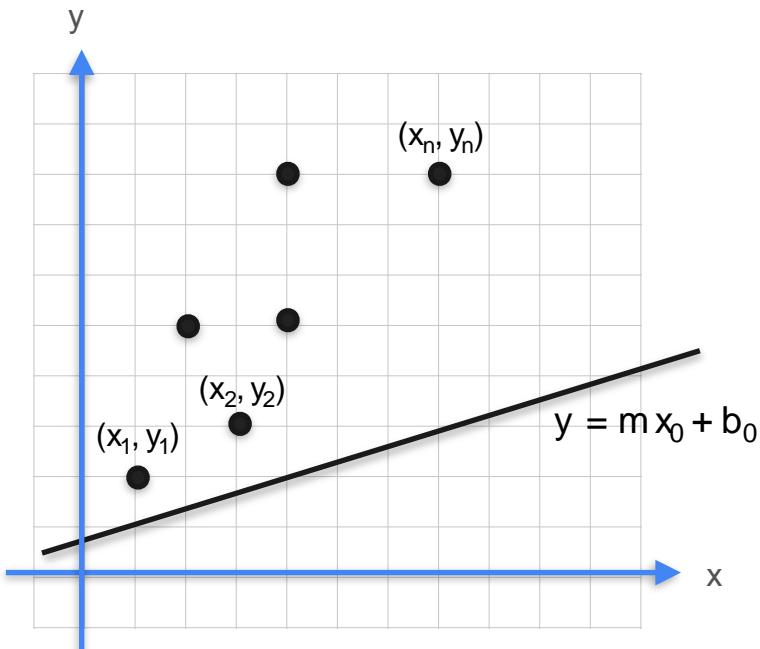
# Gradient Descent



$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_0 \\ b_0 \end{bmatrix}$$

# Gradient Descent

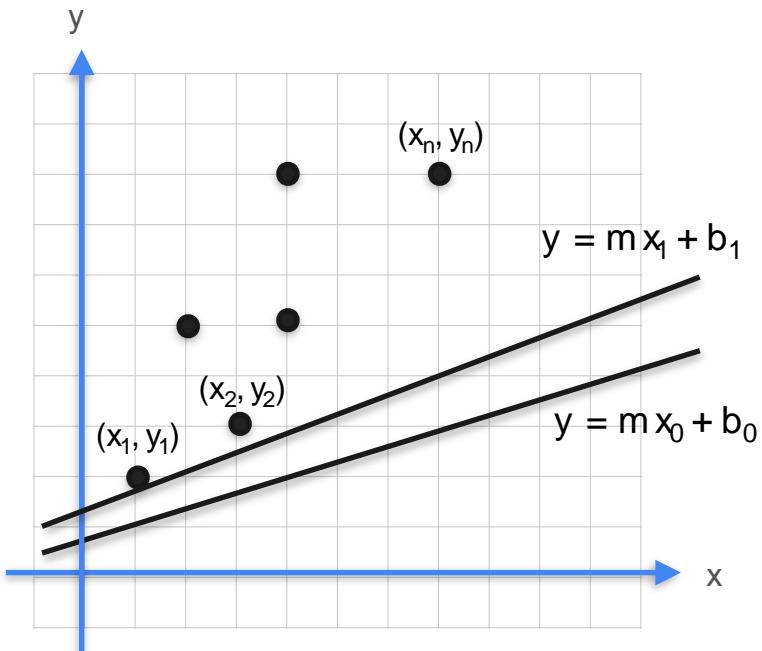


$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_0 \\ b_0 \end{bmatrix} \rightarrow \begin{bmatrix} m_1 \\ b_1 \end{bmatrix} = \begin{bmatrix} m_0 \\ b_0 \end{bmatrix}$$

$\downarrow(m_0, b_0)$

# Gradient Descent

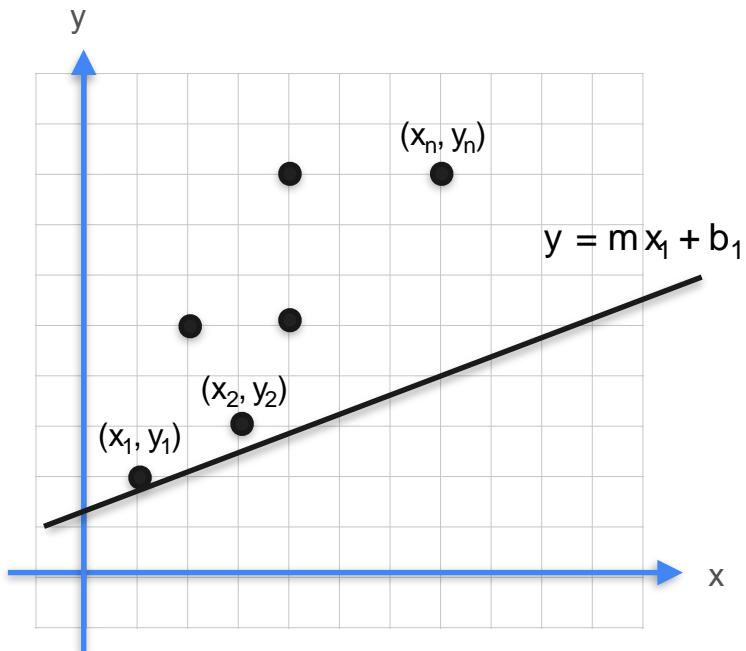


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# Gradient Descent

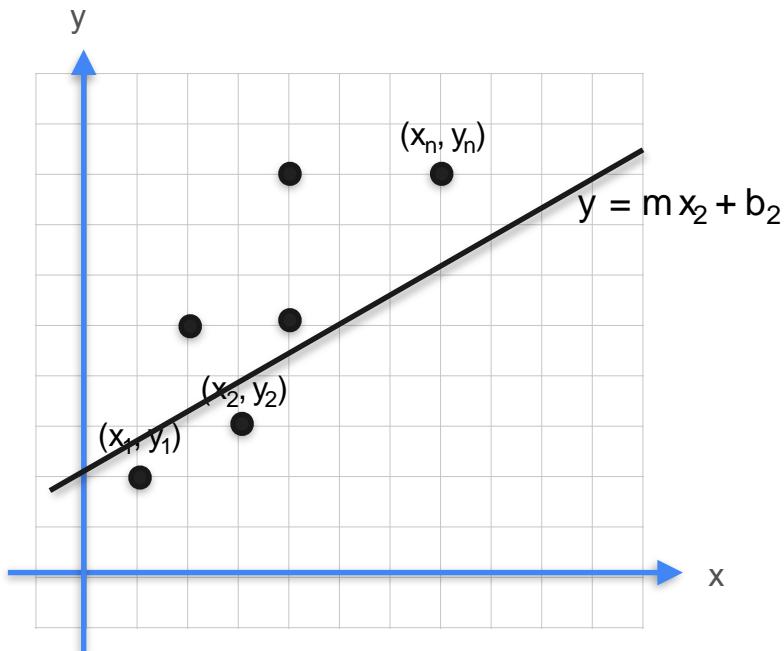


$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_0 \\ b_0 \end{bmatrix} \xrightarrow{\quad} \begin{bmatrix} m_1 \\ b_1 \end{bmatrix} = \begin{bmatrix} m_0 \\ b_0 \end{bmatrix}$$

$\downarrow(m_0, b_0)$

# Gradient Descent

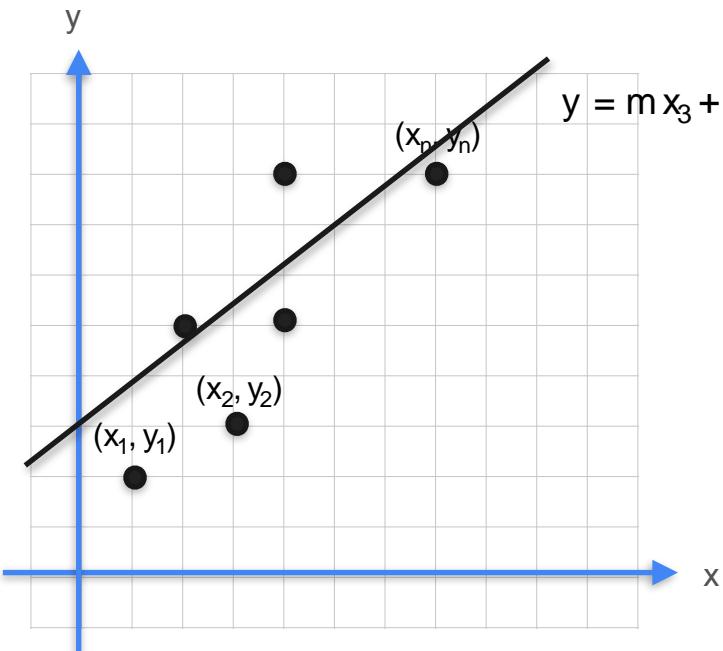


$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_1 \\ b_1 \end{bmatrix} \rightarrow \begin{bmatrix} m_2 \\ b_2 \end{bmatrix} = \begin{bmatrix} m_1 \\ b_1 \end{bmatrix}$$

$\downarrow(m_1, b_1)$

# Gradient Descent

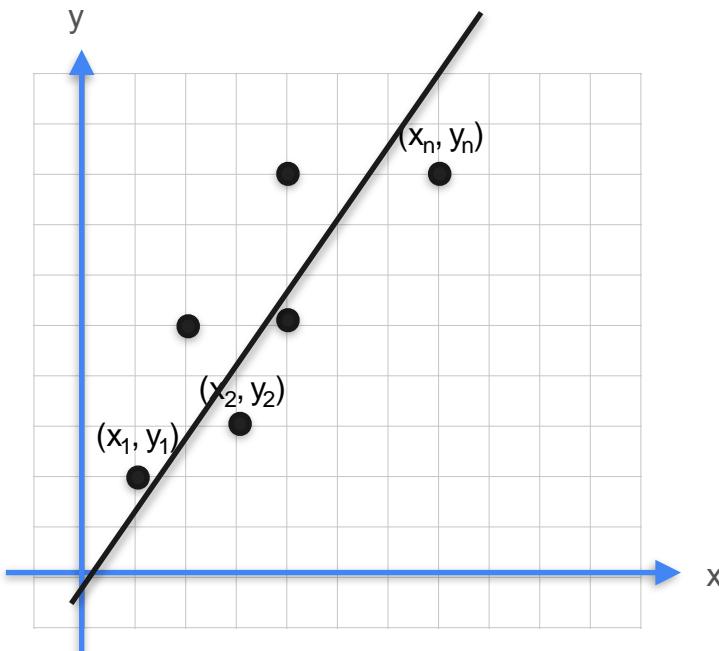


$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_2 \\ b_2 \end{bmatrix} \rightarrow \begin{bmatrix} m_3 \\ b_3 \end{bmatrix} = \begin{bmatrix} m_2 \\ b_2 \end{bmatrix}$$

$\downarrow(m_2, b_2)$

# Gradient Descent

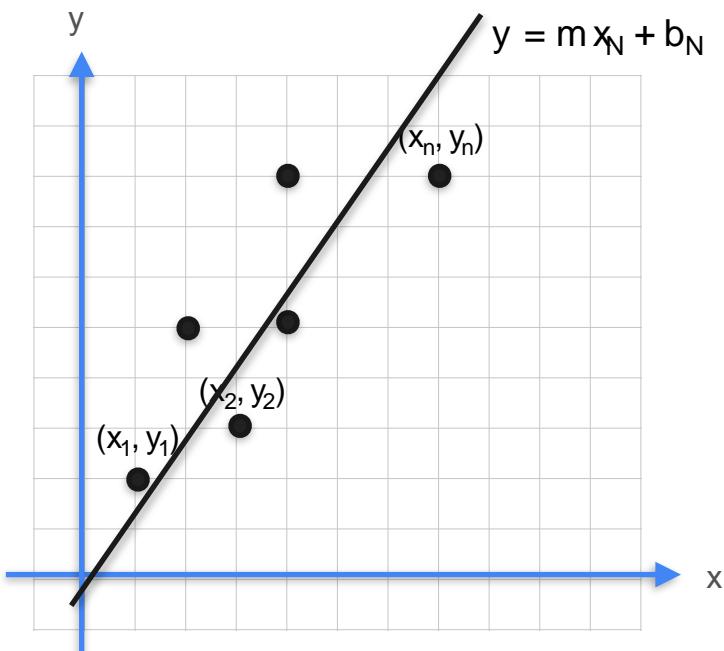


$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_N \\ b_N \end{bmatrix} \rightarrow \begin{bmatrix} m_N \\ b_N \end{bmatrix} = \begin{bmatrix} m_{N-1} \\ b_{N-1} \end{bmatrix}$$

 $_1(m_{N-1}, b_{N-1})$

# Gradient Descent



$$(m, b) = \frac{1}{2m} [(mx_1 + b - y_1)^2 + \dots + (mx_n + b - y_n)^2]$$

$$\begin{bmatrix} m_N \\ b_N \end{bmatrix} \rightarrow \begin{bmatrix} m_N \\ b_N \end{bmatrix} = \begin{bmatrix} m_{N-1} \\ b_{N-1} \end{bmatrix}$$

 $_1(m_{N-1}, b_{N-1})$



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# Gradients and Gradient Descent

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## Conclusion