

[320] Optimization and Gradient Descent

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Optimization Problems

minimize or maximize something

1

find the x value that **minimizes** the y , when $y=f(x)$

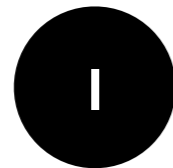
2

find the fit line **coefficients** (slope and intercept) that **minimize** the **average squared differences** between the data and the line

3

find the **weights on edges between neurons** to **minimize** the **mistakes** made by the neural network

Techniques



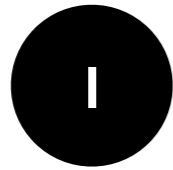
find the x value that **minimizes** the y , when $y=f(x)$

Calculus: find derivative of continuous function f , set to zero, evaluate x solutions

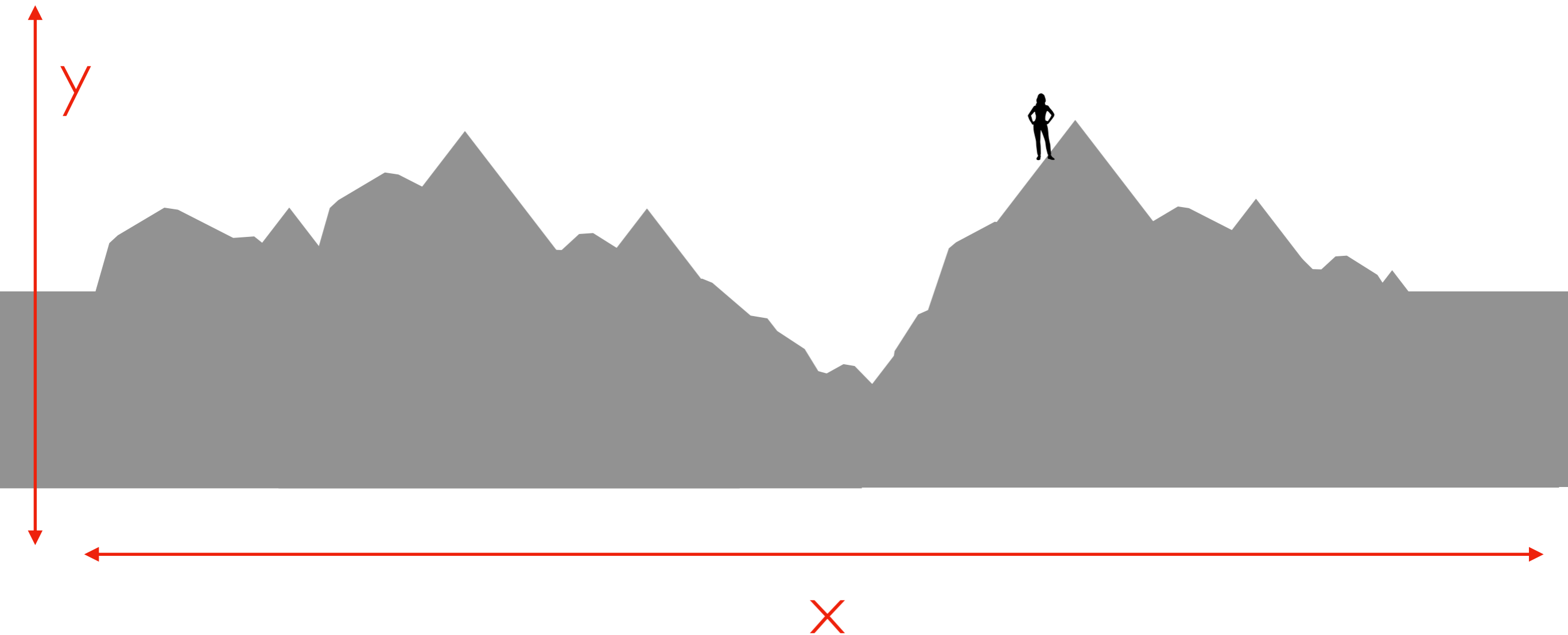
Compute: loop over lots of x values $(-5, -4.9, -4.8, \dots, 4.8, 4.9, 5)$

Compute: gradient descent (keep tweaking x based on gradient, searching for best)

Gradient Descent

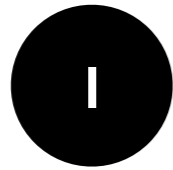


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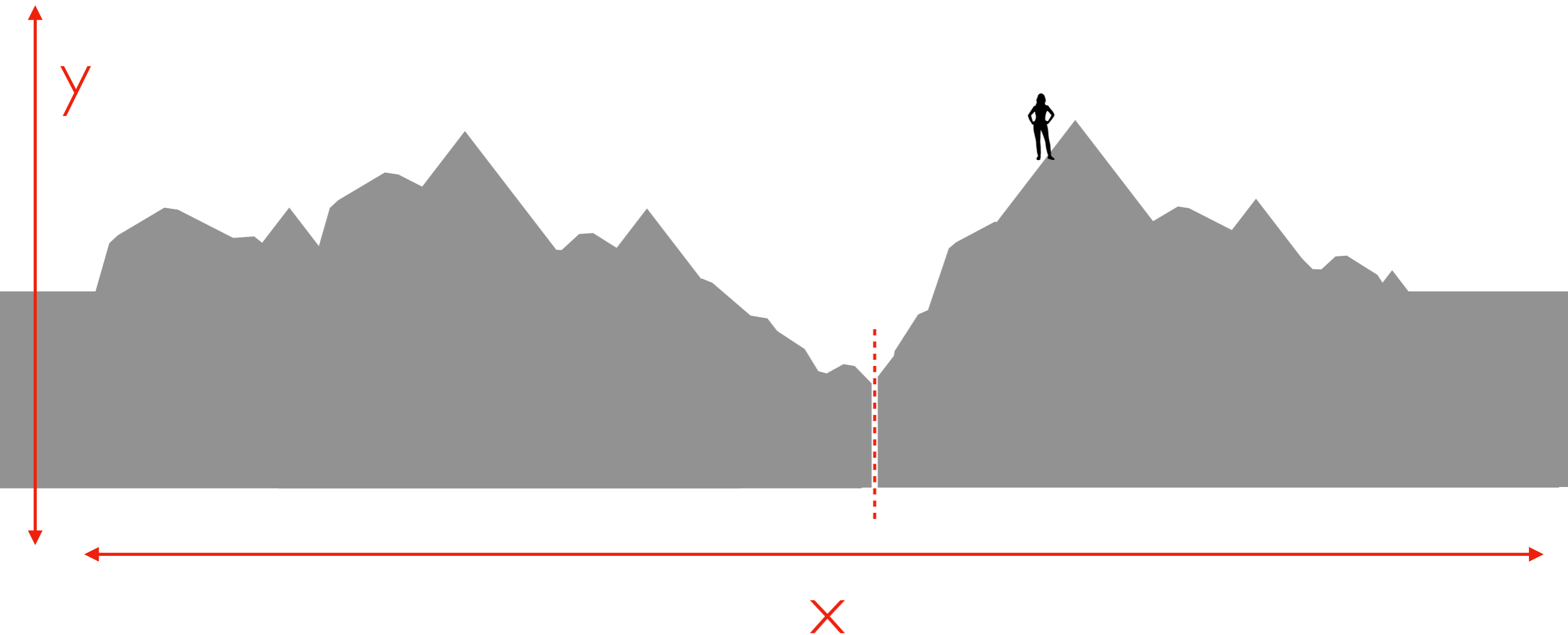


imagine you're in the mountains...

Gradient Descent

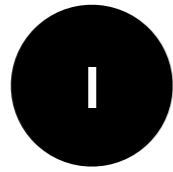


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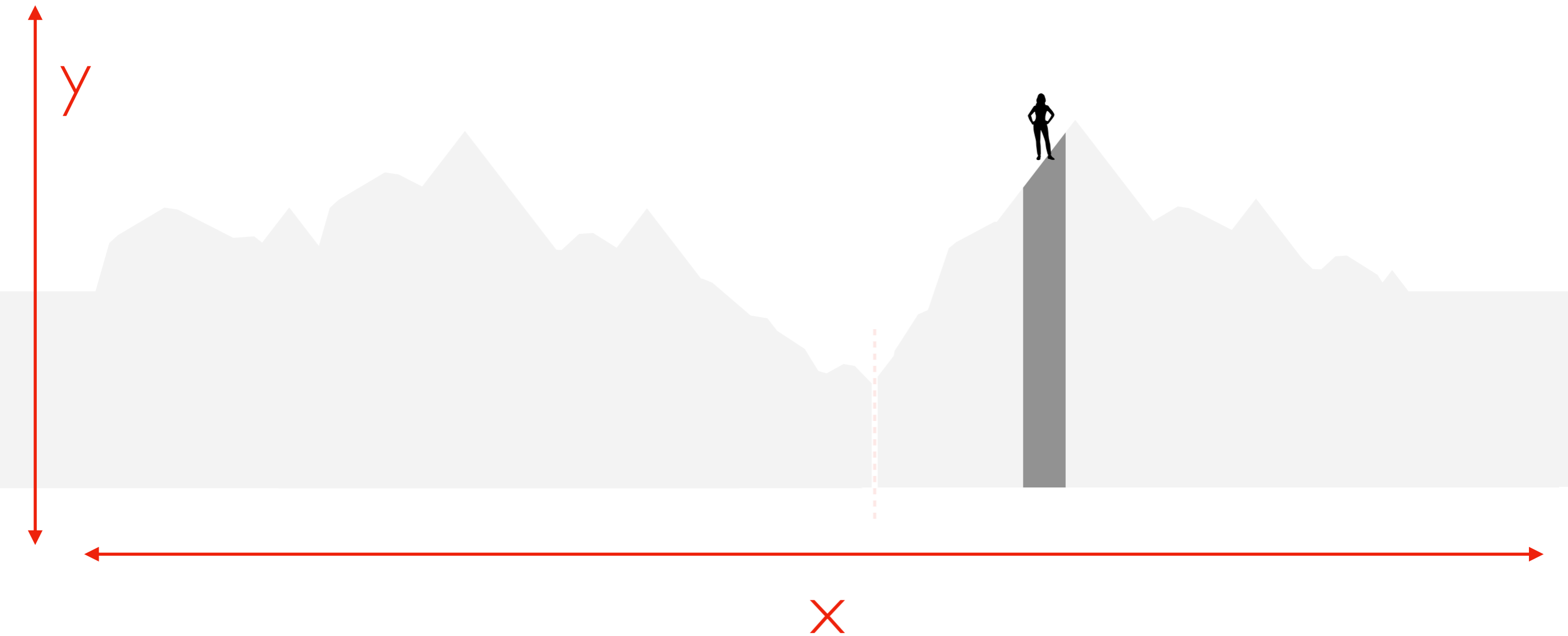


...trying to find the lowest point...

Gradient Descent

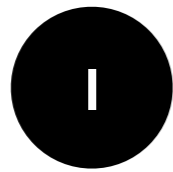


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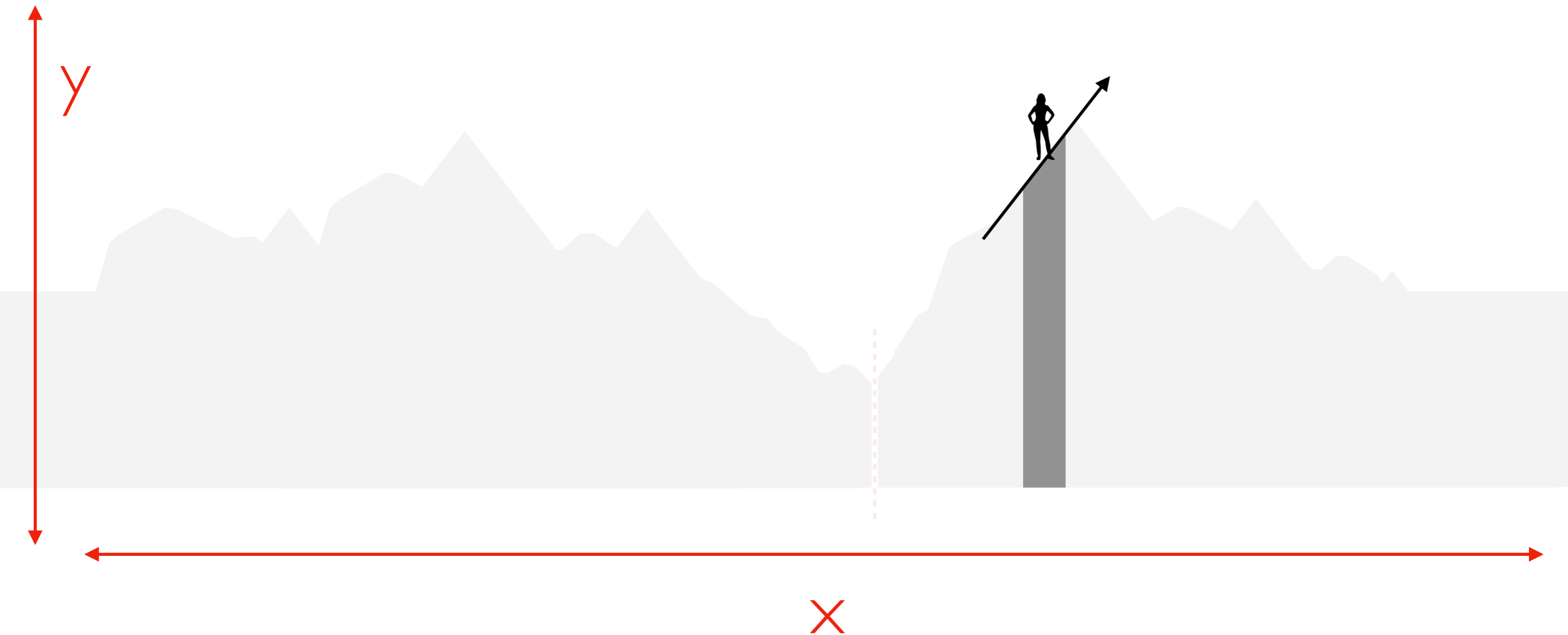


...in a heavy fog

Gradient Descent

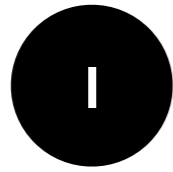


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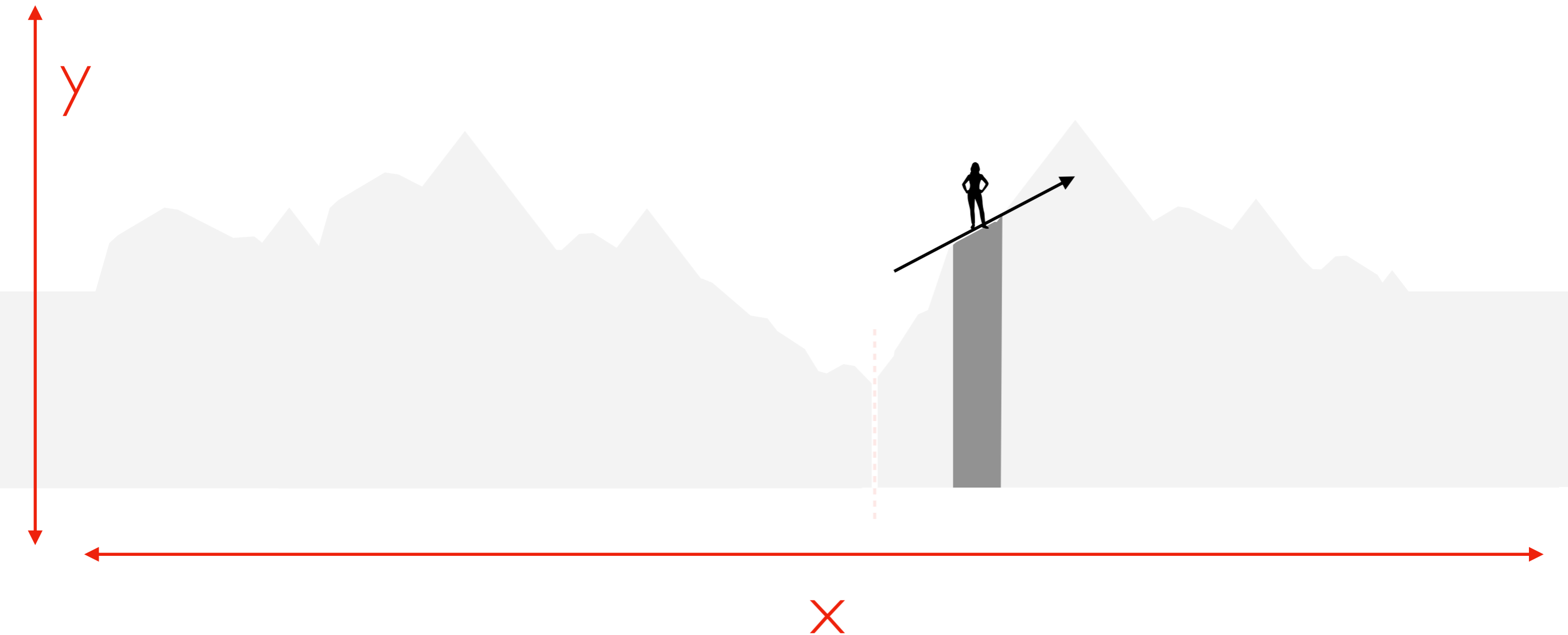


Move to bigger or smaller x ? Smaller because the gradient is positive!

Gradient Descent

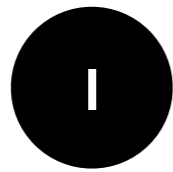


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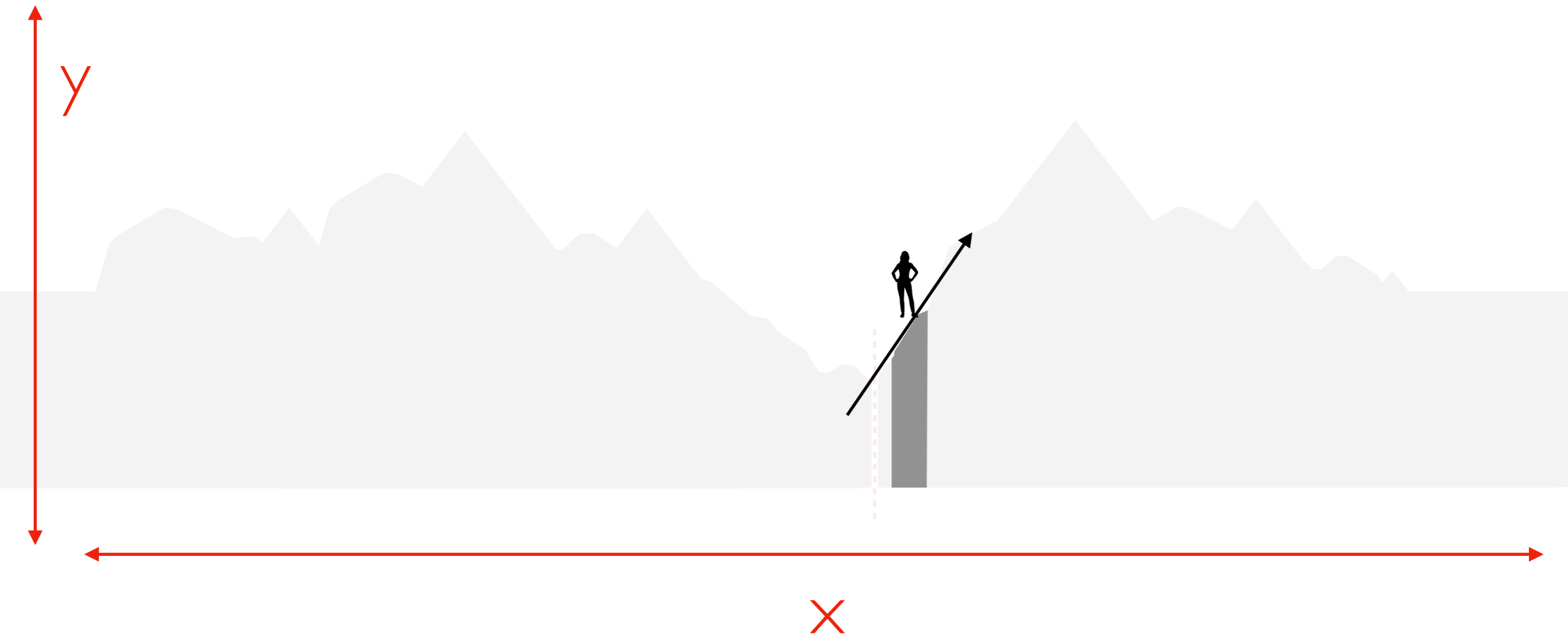


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Gradient Descent

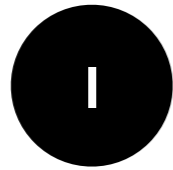


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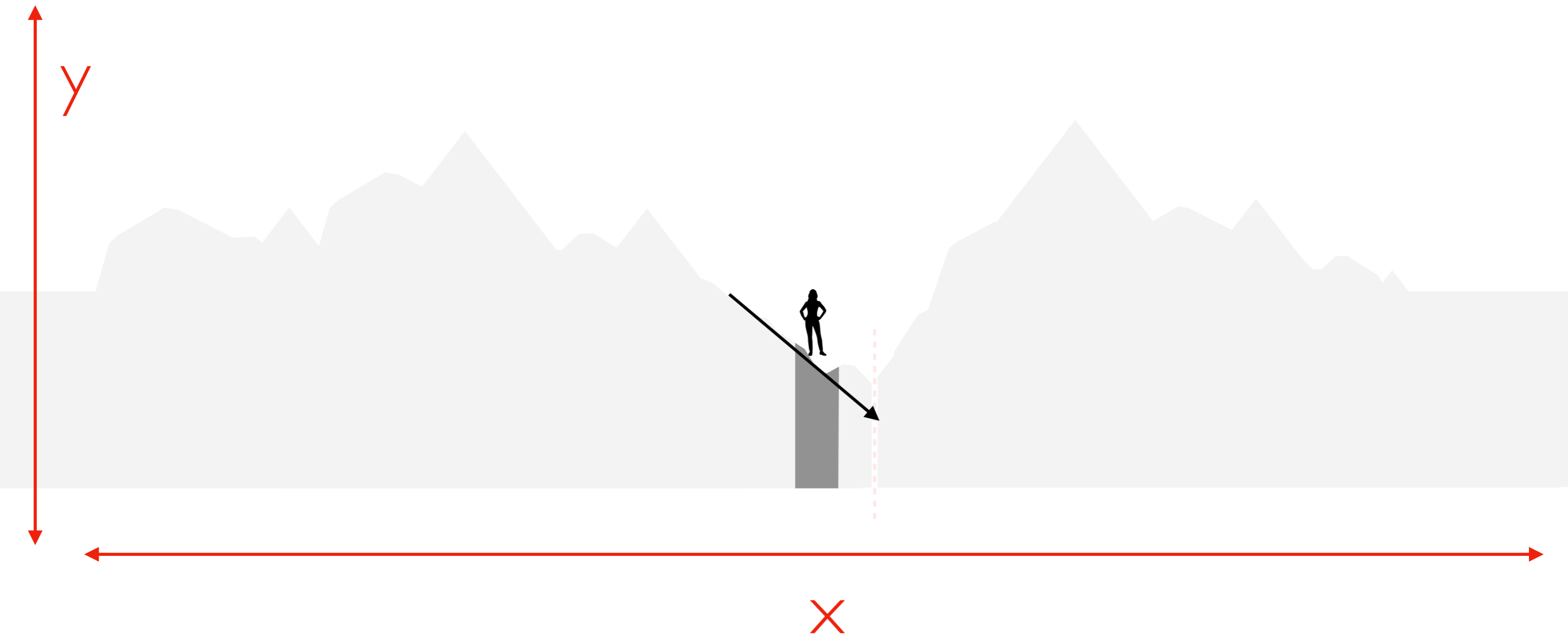


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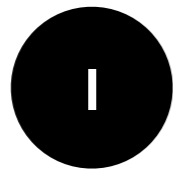


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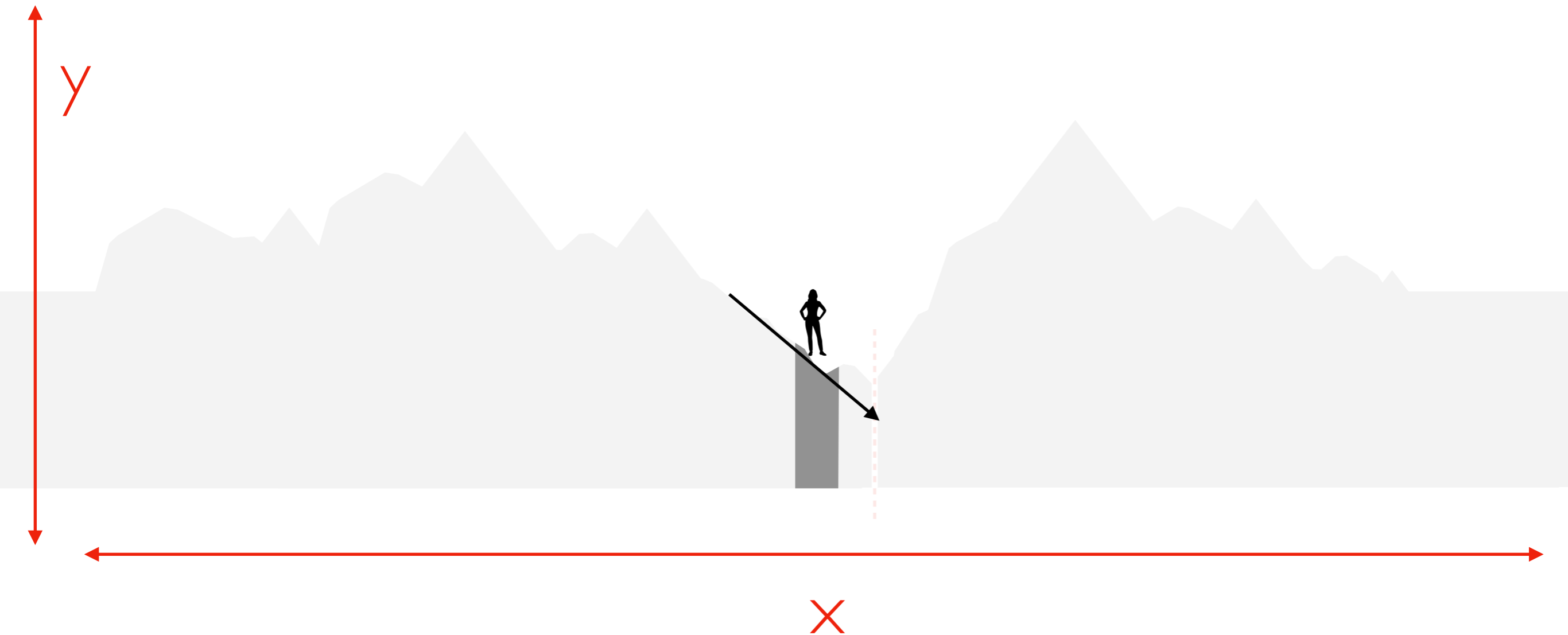


Hiking Analogy Breaks Down: you "jump" without crossing area between

Gradient Descent

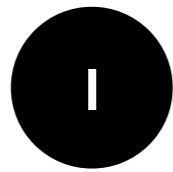


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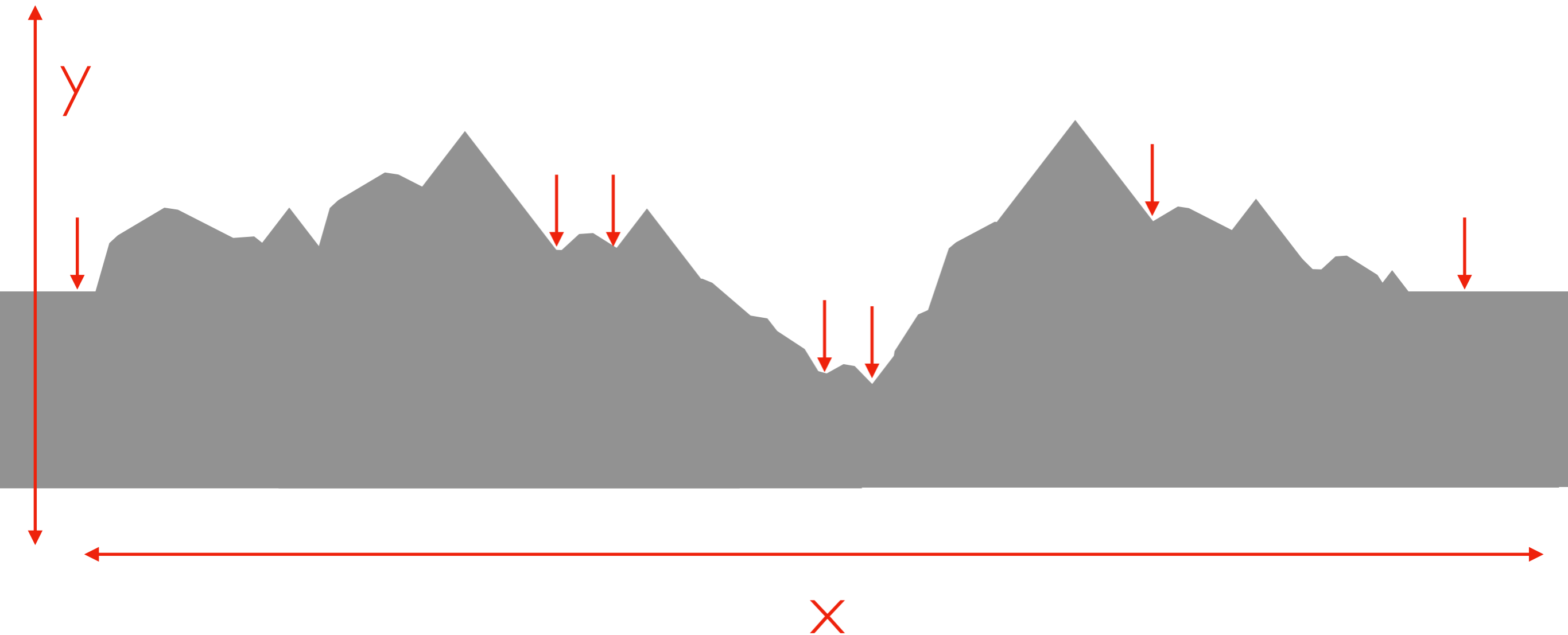


Problem 1: jumping past the optimum without realizing it (how far should we jump each time?)

Gradient Descent

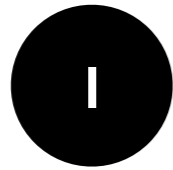


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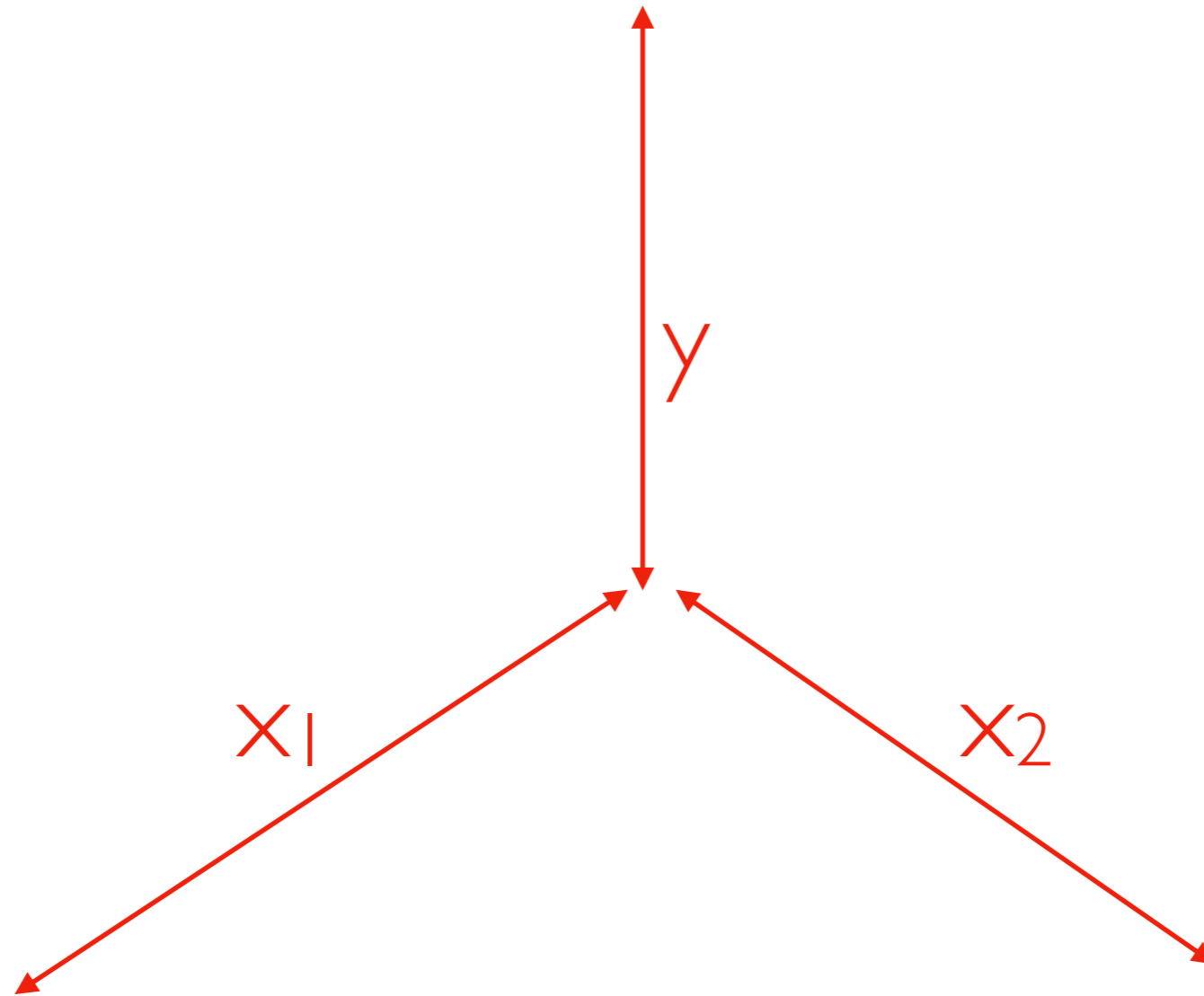


Problem 2: lots of local minima (for certain problems)

Gradient Descent

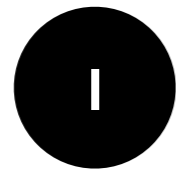


find the x value(s) that **minimize** the y , when $y=f(x_1, x_2)$



Hiking Analogy Breaks Down: there may be MANY dimensions

Gradient Descent



find the x value that **minimizes** the y , when $y=f(x_1, x_2, x_3, x_4, \dots, x_N)$



Hiking Analogy Breaks Down: there may be MANY dimensions

Least Squares, with Gradient Descent

- 1 find the x value that **minimizes** the y , when $y=f(x_1, x_2, x_3, x_4, \dots, x_N)$
- 2 find the fit line **coefficients** (slope and intercept) that **minimize** the **average squared differences** between the data and the line

$$y = f(x) \quad \text{where } f(x) = \text{slope} * x + \text{intercept}$$

$$\text{error} = \text{mean_squared_error}(\text{slope}, \text{intercept})$$

use gradient descent to find best slope, intercept!

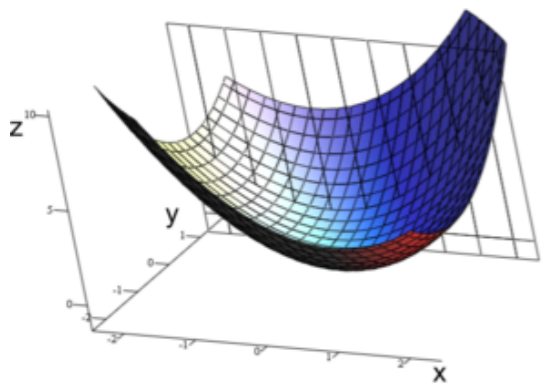
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mean_squared_error is a convex function: https://en.wikipedia.org/wiki/Convex_function