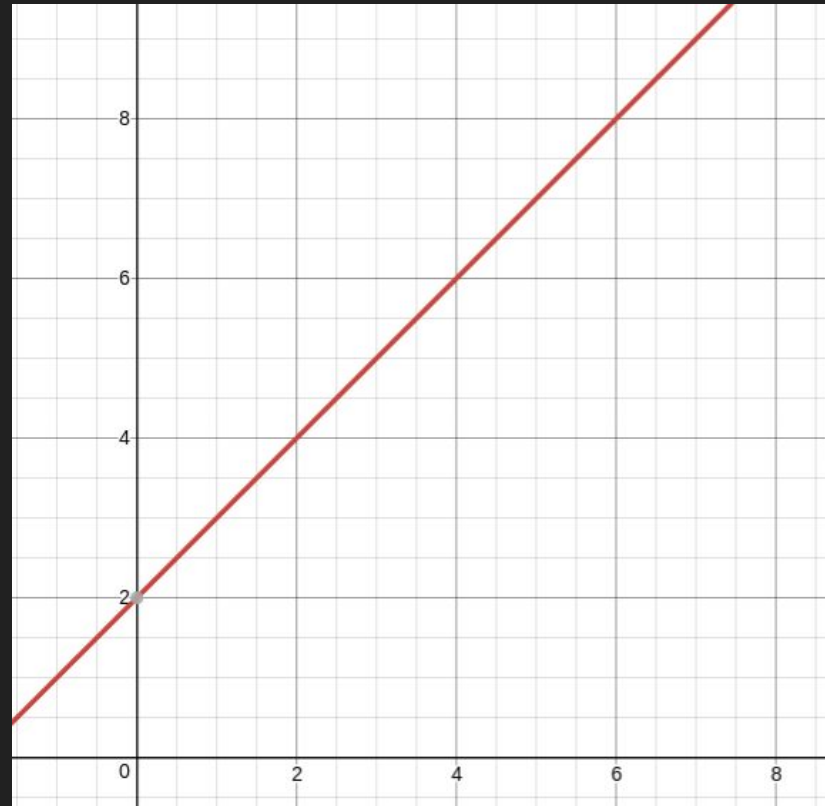


# Session 1: Linear Regression with One Variable

And a Python review

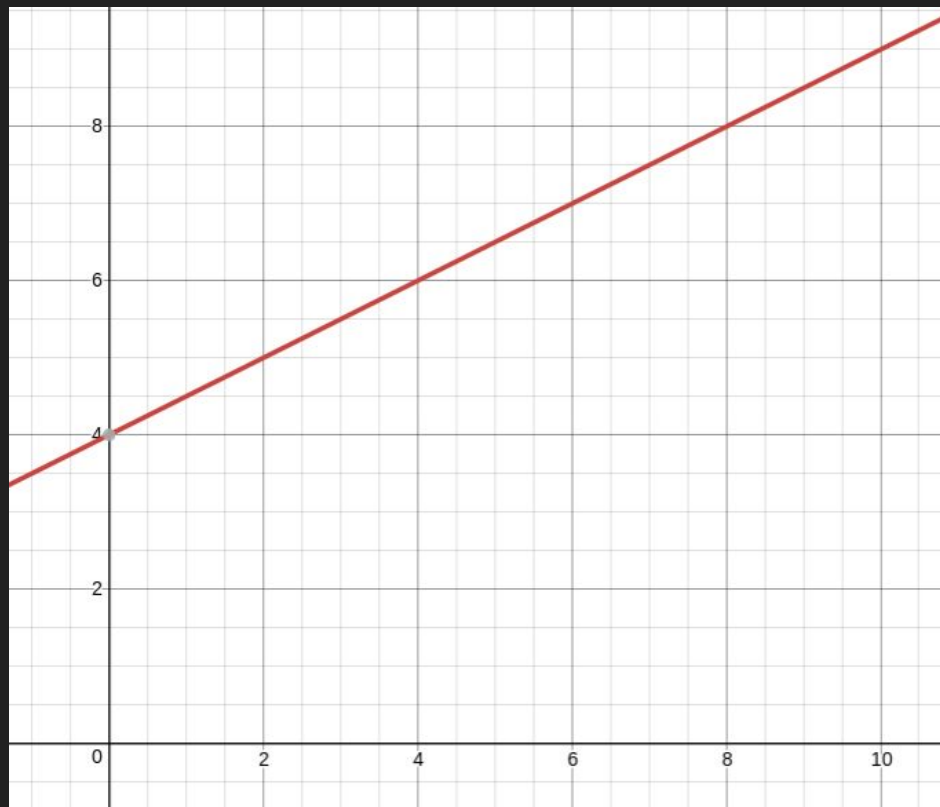
# A linear equation

$$y = 1x + 2$$

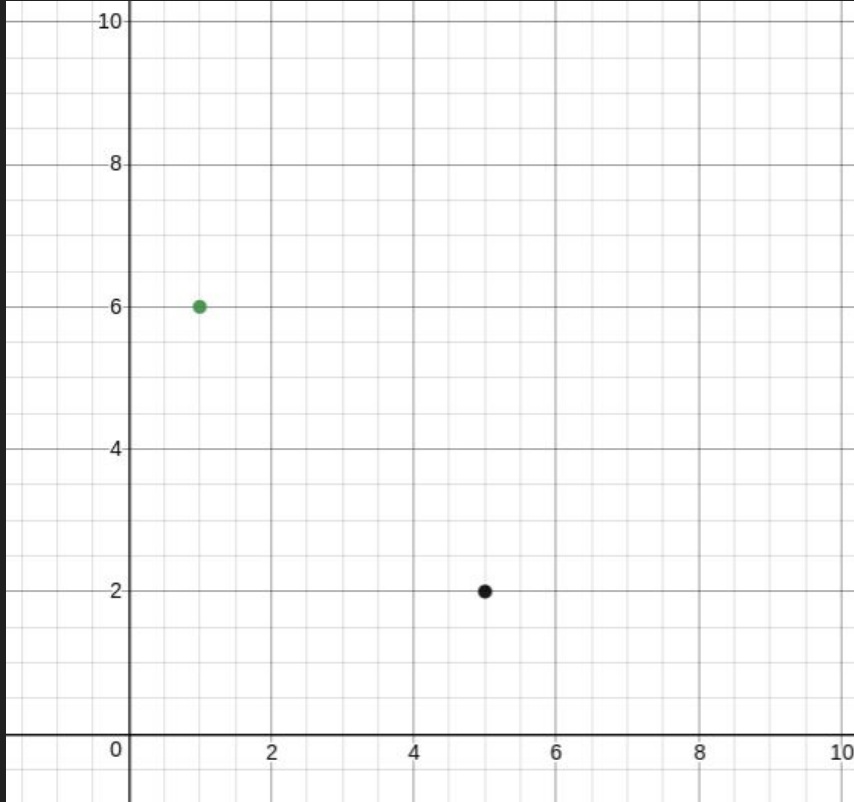


# Another linear equation

$$y = 0.5x + 4$$



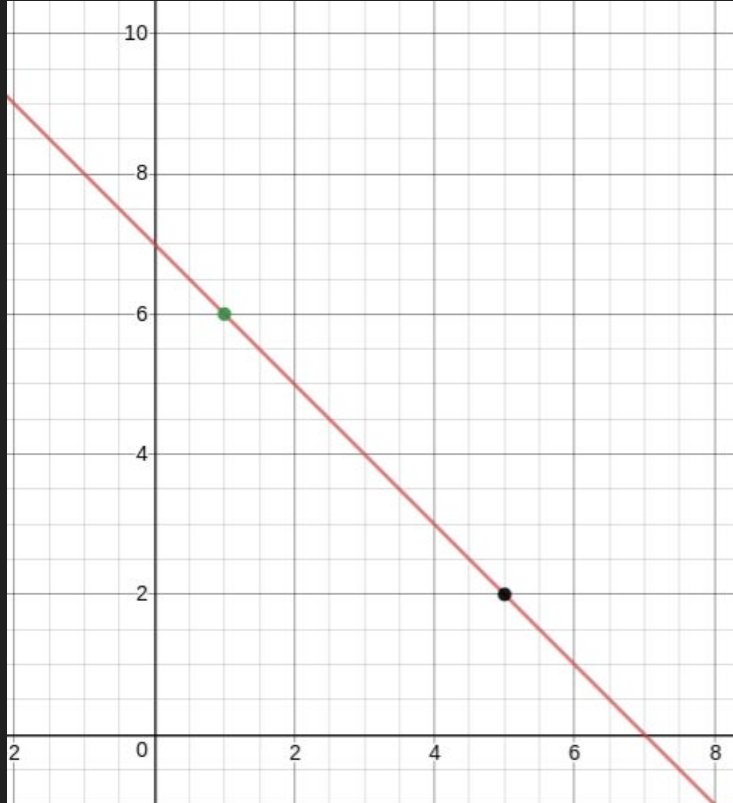
# What if we instead have two values?



$$y = mx + b$$

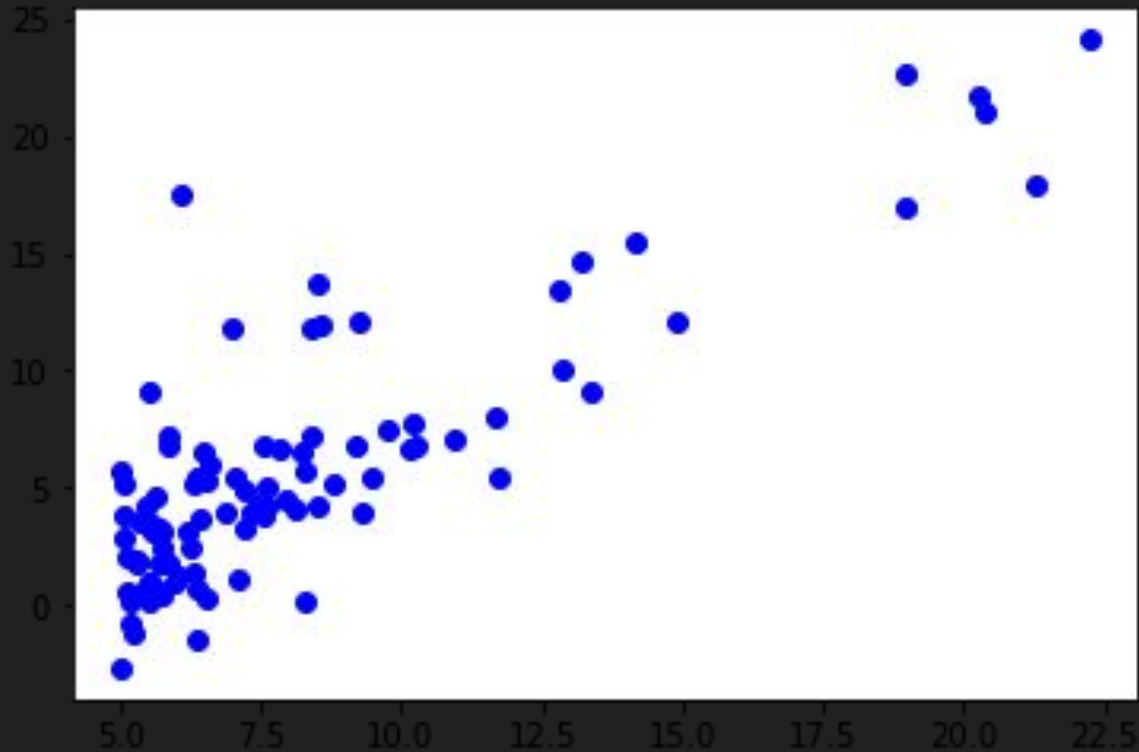
$$slope = \frac{(y_2 - y_1)}{(x_2 - x_1)}$$

We solve the equations and find the line equation



$$y = -x + 7$$

What if we have a bunch of values?



# We need a way to generalize that behaviour

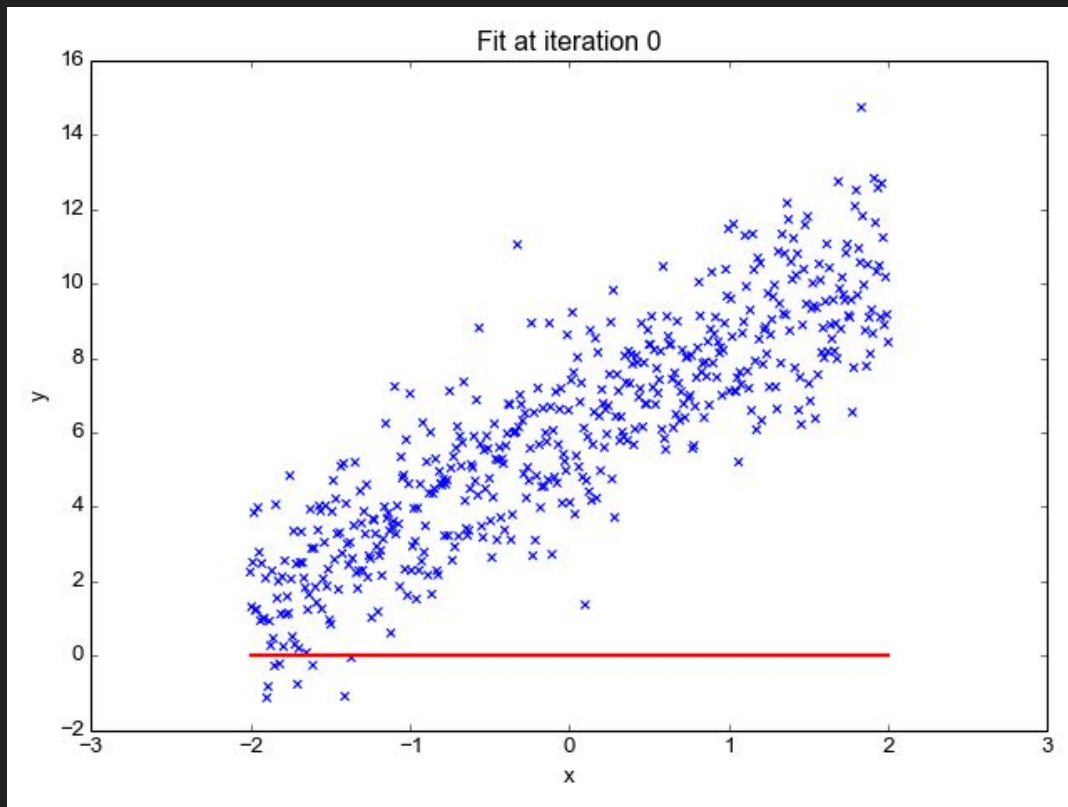
Enter linear regression!

For a set of values  $(x, y)$  we want to find the best model  $y = \theta x + \beta$  such that it minimizes the error.

We will measure the error with the Mean Squared Error (MSE):

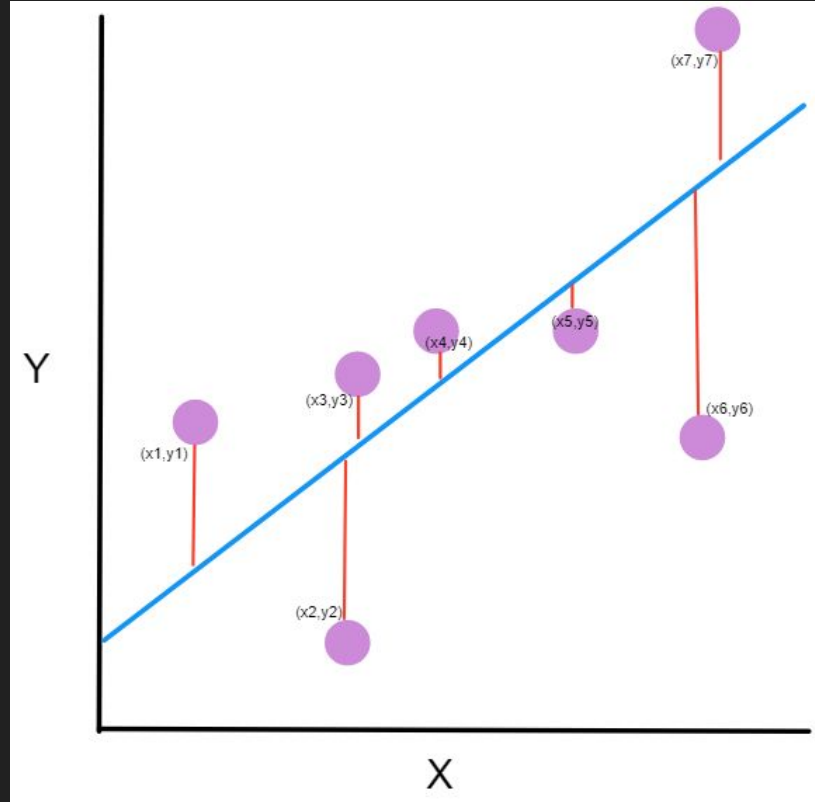
$$\text{MSE} = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2.$$

This is a more visual explanation of that definition

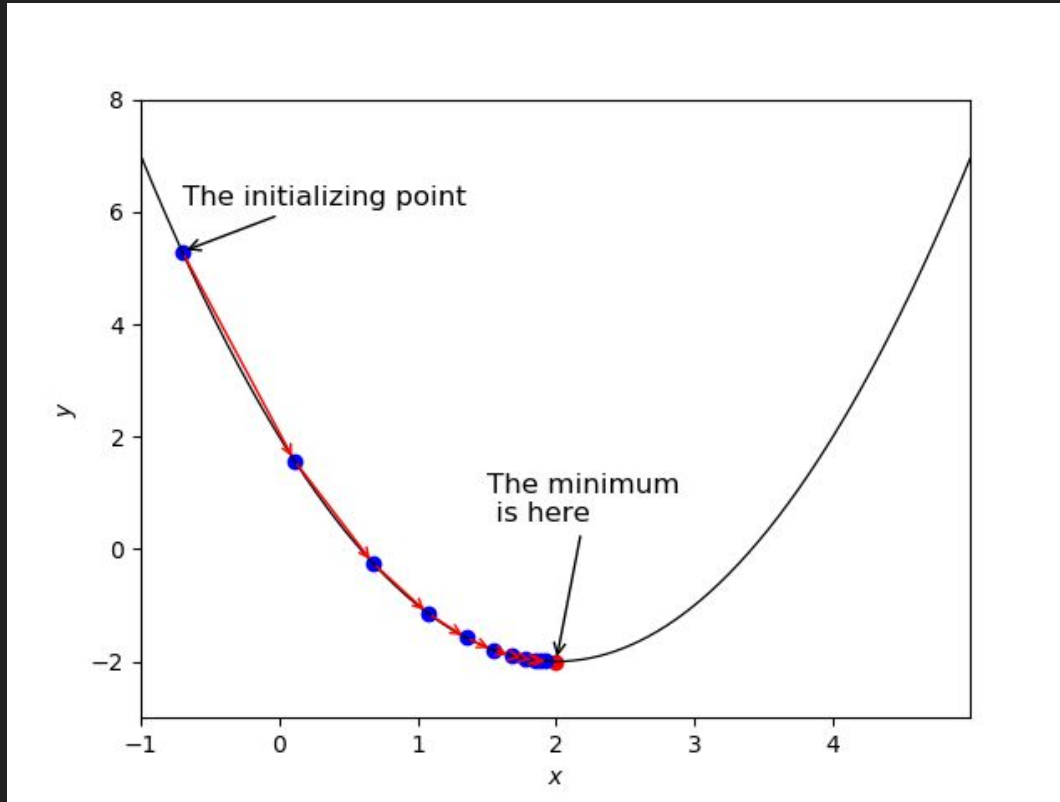




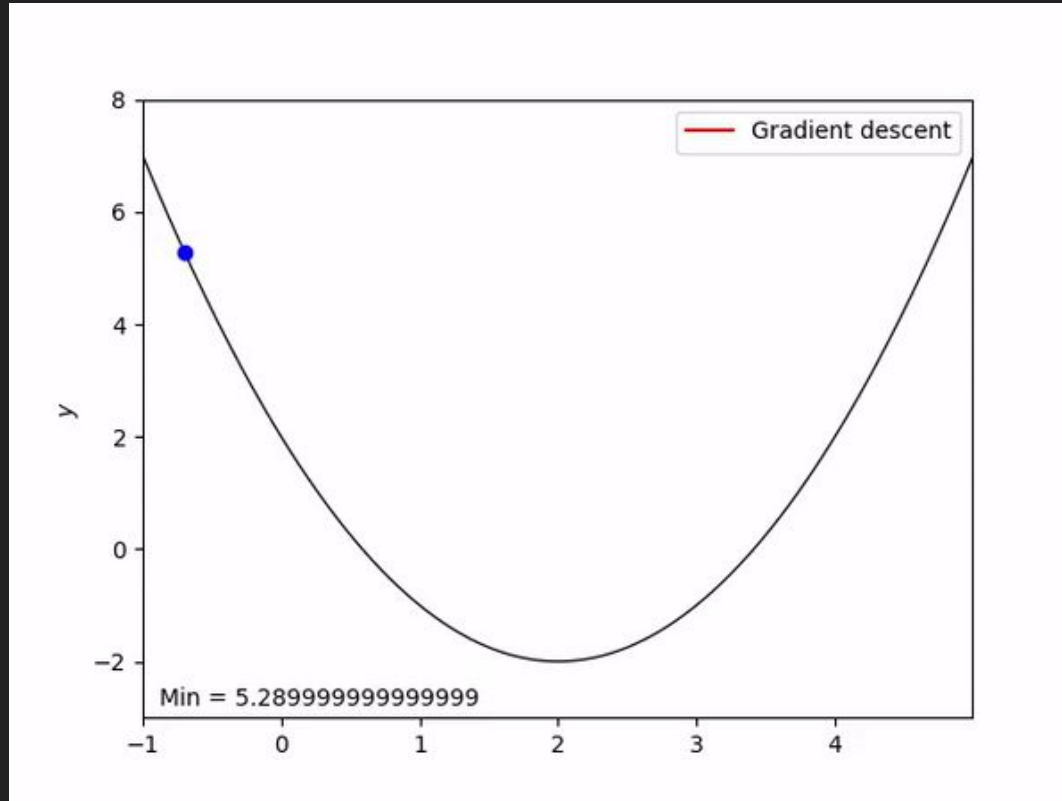
And we try to minimize the length of the red lines



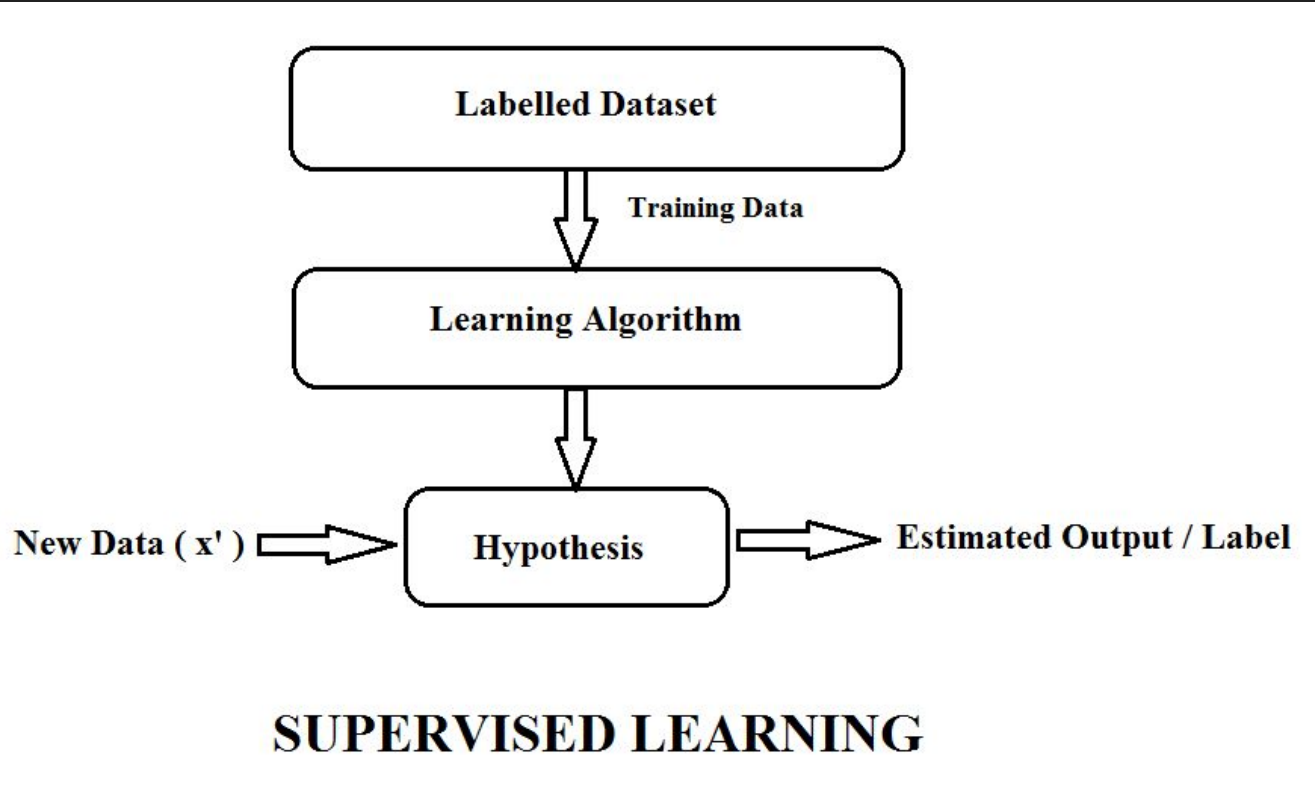
# To do that we use gradient descent



# An animation of gradient descent

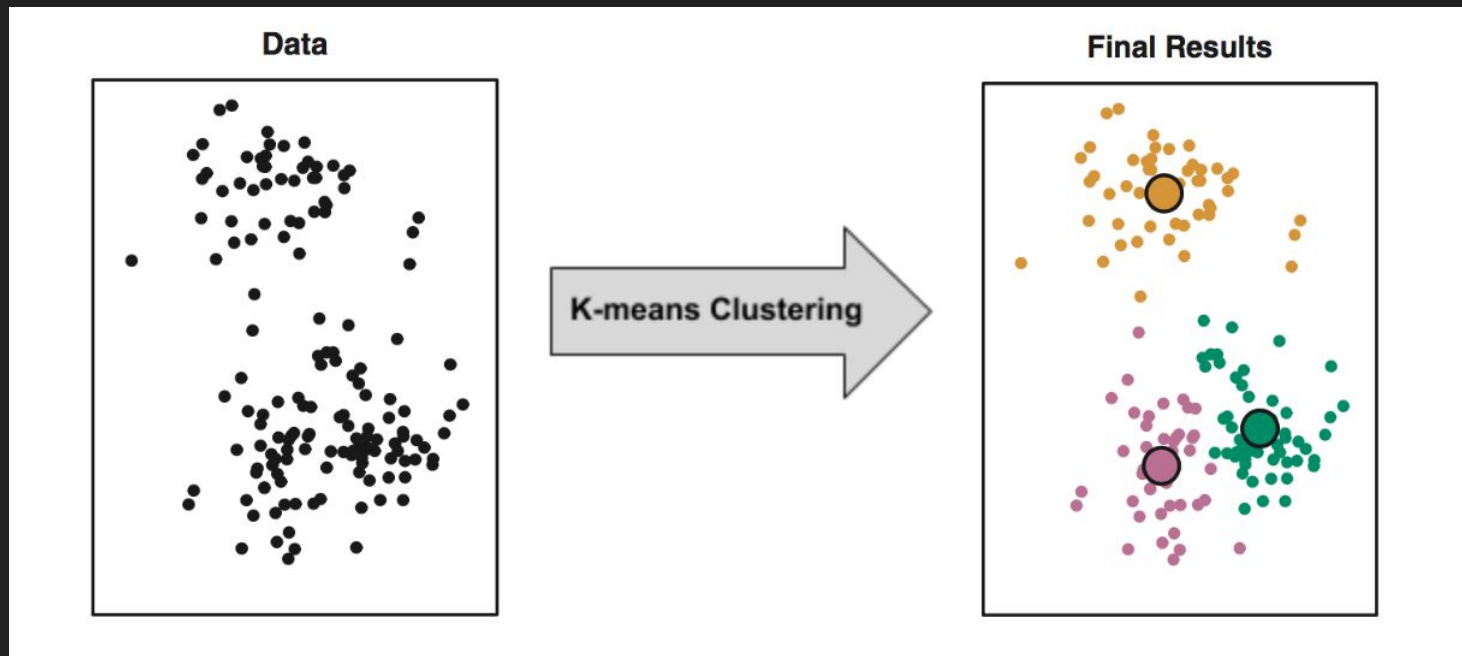


# How does this fit in machine learning?



# What about unsupervised learning?

We throw data to the computer and hope that it gives us something good.



Now let's play with Python! :D