

# MACHINE LEARNING

## WEEK04

اَللّٰهُمَّ ارْزُقْنِيْ عِلْمًا نَّافِعًا وَاسِعًا عَمِيْقًا

اَللّٰهُمَّ ارْزُقْنِيْ رِزْقًا وَّاسِعًا حَلَالًا طَيِّبًا  
مُّبَارَكًا مِنْ عِنْدِكَ

# WEEK 04

# TYPE OF MACHINE LEARNING

- Supervised Learning.
- Unsupervised Learning.
- Reinforcement Learning.

# SUPERVISED LEARNING

Living area (feet <sup>2</sup> )	Price (1000\$s)
2104	400
1600	330
2400	369
1416	232
3000	540
⋮	⋮

# SUPERVISED LEARNING: REGRESSION

- When we try to predict a number from historical data this type of supervised learning problem is called Regression Problem

# FIND THE RELATION

- Can You predict upcoming value based on the following data.

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?



# FIND THE RELATION

- Can you write Generic Formula (function) for following ?

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?

# FIND THE RELATION

- $y=2x+1$  (this is called model)
- We call  $y$  is function of  $x$ .

<b>x</b>	<b>y</b>
2	5
3	7
4	9
5	11
6	13
7	?

# FIND THE RELATION

- We call  $y$  is function of  $x$ .  $f(x) = 2x+1$
- More Formally, we write it  $f: x:\mathbb{R} \rightarrow y:\mathbb{R}$

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?

# LINEAR FUNCTION

## ■ Function on Graph

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?

# LINEAR FUNCTION

- This Function is called Linear Function

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?

# FIND THE RELATION

- Can you predict value of  $y$  when  $x=7$  based on the following data ?

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# FIND THE RELATION

- What could be the general relation ?

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# FIND THE RELATION

- What could be the general relation ?
- $y = x^2 + 2$

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?



# GRAPH THE RELATION

- Draw relation  $y = x^2 + 2$

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# GRAPH THE RELATION

- This kind of relation is called **Quadratic Relation**

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# FIND THE RELATION

- The problem of finding the relation between  $x$  and  $y$  is called Regression

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# FIND THE RELATION

- And if we assume the relation shall be a linear function, we call it linear regression

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# FEATURES: INDEPENDENT VARIABLE

- In this relation  $x$  is independent variable and  $y$  is dependent variable.

<b>x</b>	<b>y</b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?

# UNIVERIATE LINEAR REGRESSION

- If there is only single independent variable or feature that use to predict the  $y$ , then this problem is called Univeriate Linear Regression

# REGRESSION

- As an Human, we identified the function that could be  $f(x) = 2x+1$  or  $f(x) = x^2+2$  where  $f(x)$  is  $y$ .

# FEATURES: INDEPENDENT VARIABLE

- $X$  is also called as features and  $y$  is also called target variable or label variable.

<b><math>x</math></b>	<b><math>y</math></b>
0	2
2	6
3	11
4	18
5	27
6	38
7	?



# LINEAR REGRESSION

given independent variables, we need to find a **linear function** that can **predict** the value of **y** while value of **x** is given.

$$f(x) = \theta_1 x + \theta_0$$

# LINEAR REGRESSION

- In previous example  $\theta_1 = 2$ ,  $\theta_0 = 1$  that make it  $f(x) = 2x + 1$

$$f(x) = \theta_1 x + \theta_0$$

# LINEAR REGRESSION

- This identified function is called **model** or **hypothesis** on data.

$$f(x) = \theta_1 x + \theta_0$$

# LINEAR REGRESSION

The  $\theta_1$  and  $\theta_0$  are called **parameters**, the value of these parameters could be different for different type of data.

$$f(x) = \theta_1 x + \theta_0$$

# MORE FORMALLY

## Linear Regression

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**Given:** Training data:  $(x_1, y_1), \dots, (x_n, y_n)$  /  $x_i \in \mathbb{R}^d$  and  $y_i \in \mathbb{R}$

example $x_1 \rightarrow$	$x_{11}$	$x_{12}$	$\dots$	$x_{1d}$	$y_1 \leftarrow$ label
$\dots$	$\dots$	$\dots$	$\dots$	$\dots$	$\dots$
example $x_i \rightarrow$	$x_{i1}$	$x_{i2}$	$\dots$	$x_{id}$	$y_i \leftarrow$ label
$\dots$	$\dots$	$\dots$	$\dots$	$\dots$	$\dots$
example $x_n \rightarrow$	$x_{n1}$	$x_{n2}$	$\dots$	$x_{nd}$	$y_n \leftarrow$ label

**Task:** Learn a regression function:

$$f : \mathbb{R}^d \longrightarrow \mathbb{R}$$

$$f(x) = y$$

**Linear Regression:** A regression model is said to be linear if it is represented by a linear function.

# LINEAR REGRESSION

<b>x</b>	<b>y</b>
0	1
2	5
3	7
4	9
5	11
6	13
7	?

So finally, we have following problem in hand for any given data we need to find a linear function that map  $x$  to  $y$ .

$$f(x) = \theta_1 x + \theta_0$$

# LINEAR REGRESSION

$$f(x) = \theta_1 x + \theta_0$$

In other terms, we need to find the parameter value of  $\theta_0$  and  $\theta_1$  that best suitable according to the given data.

# LINEAR REGRESSION

$$f(x) = \theta_1 x + \theta_0$$

This linear function also represent as a **line** graphically.



# LINEAR REGRESSION

$$f(x) = \theta_1 x + \theta_0$$

Therefore, sometime the **univariate linear regression** problem also seen as to find equation of line.

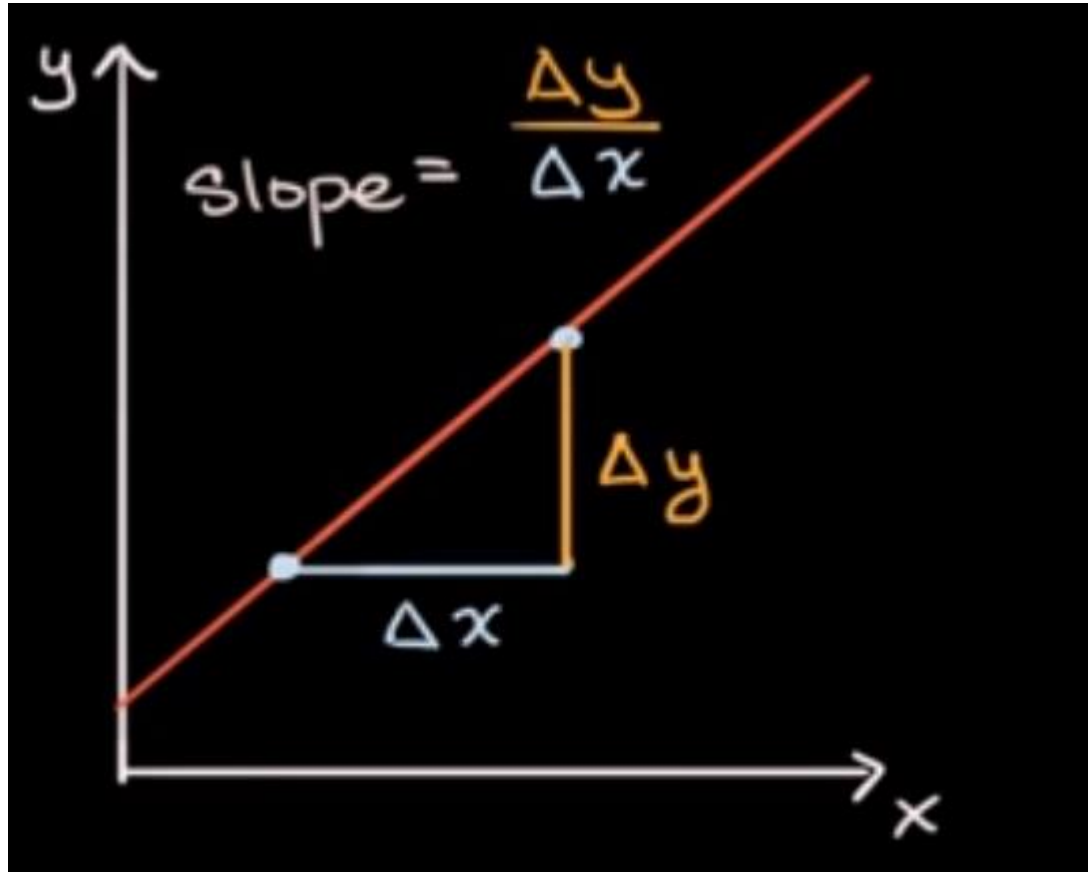
# LINEAR REGRESSION

$$f(x) = \theta_1 x + \theta_0$$

Therefore, sometime the **univariate linear regression** problem also seen as to find equation of line.

# LINES REVIEW

# SLOPE OF LINE

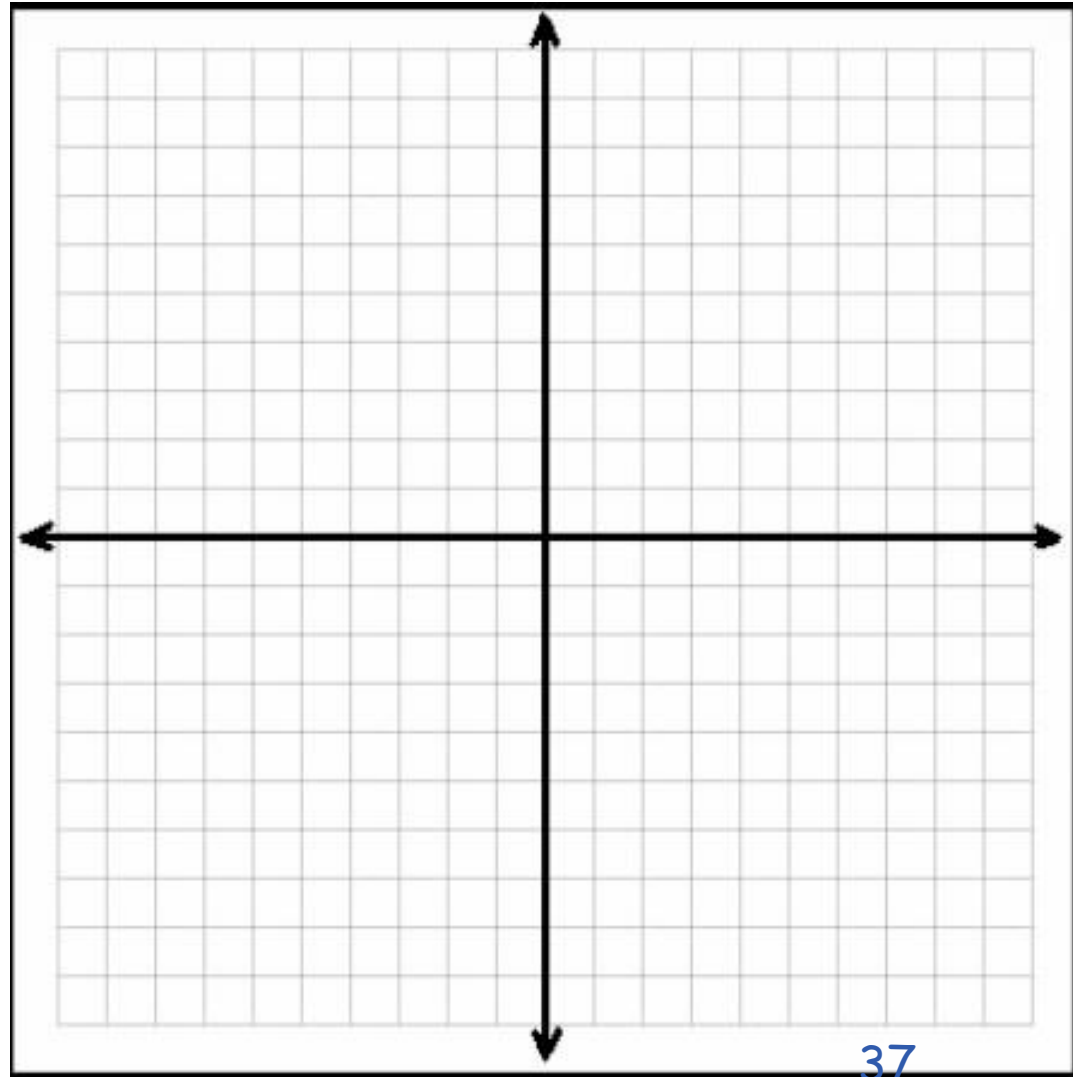


# SLOPE INTERCEPT FORM

$$y = mx + c$$

Draw Line for  
following Equation

$$y = \frac{5}{3}x - 2$$



# SLOPE INTERCEPT FORM

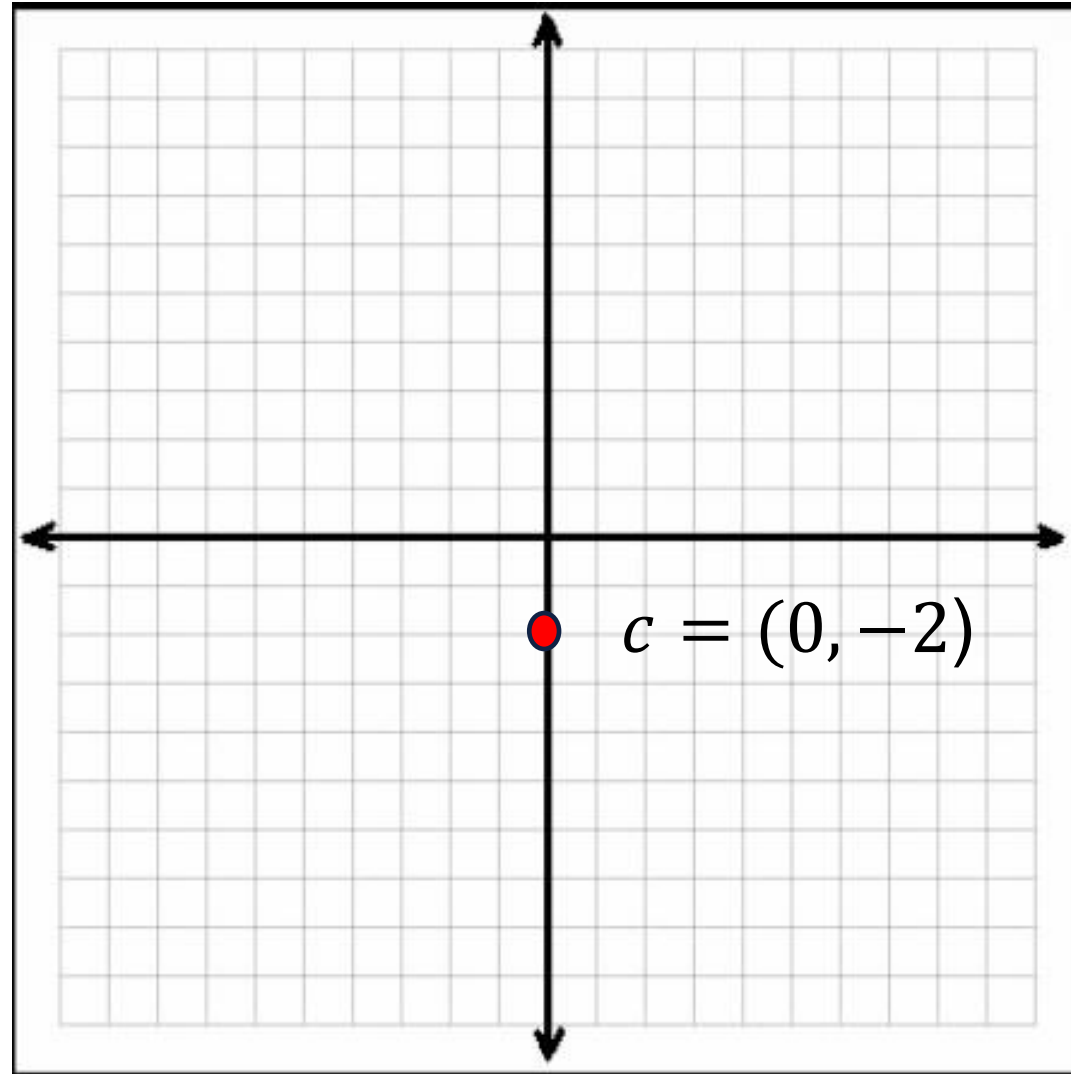
$$y = mx + c$$

$$m = \frac{\Delta Y}{\Delta X}$$

$$y = \frac{5}{3}x - 2$$

$$c = -2$$

$$m = \frac{\Delta Y}{\Delta X} = \frac{5}{3}$$



# SLOPE INTERCEPT FORM

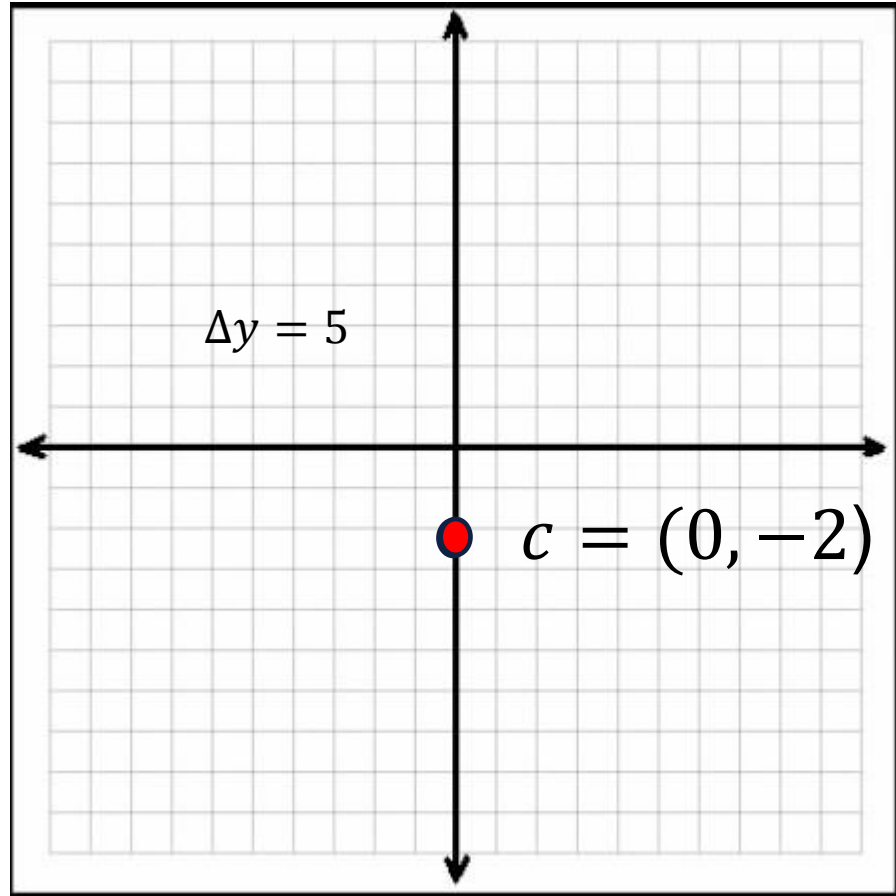
$$y = mx + c$$

$$m = \frac{\Delta Y}{\Delta X}$$

$$y = \frac{5}{3}x - 2$$

$$c = -2$$

$$m = \frac{\Delta Y}{\Delta X} = \frac{5}{3}$$



# SLOPE INTERCEPT FORM

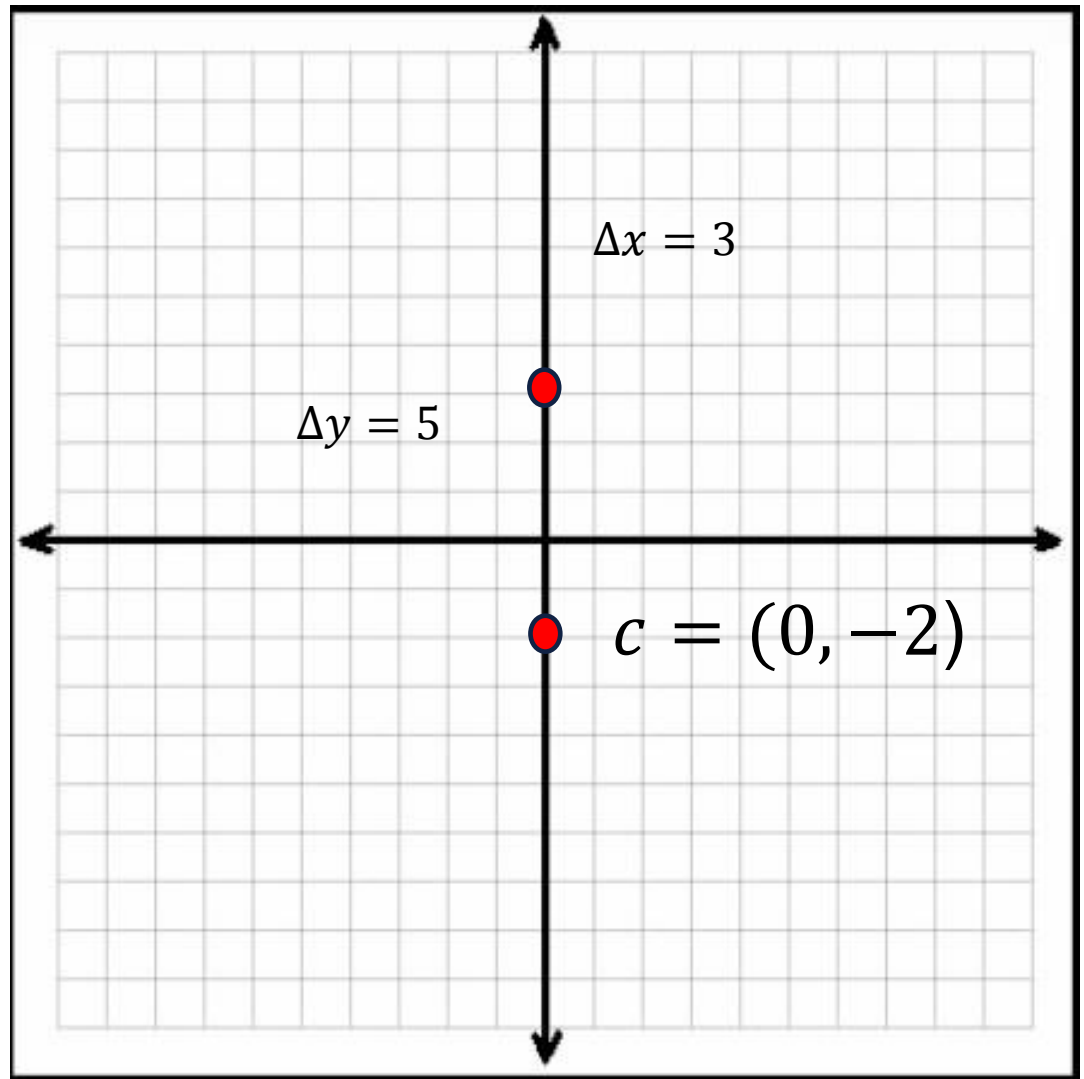
$$y = mx + c$$

$$m = \frac{\Delta Y}{\Delta X}$$

$$y = \frac{5}{3}x - 2$$

$$c = -2$$

$$m = \frac{\Delta Y}{\Delta X} = \frac{5}{3}$$





# SLOPE INTERCEPT FORM

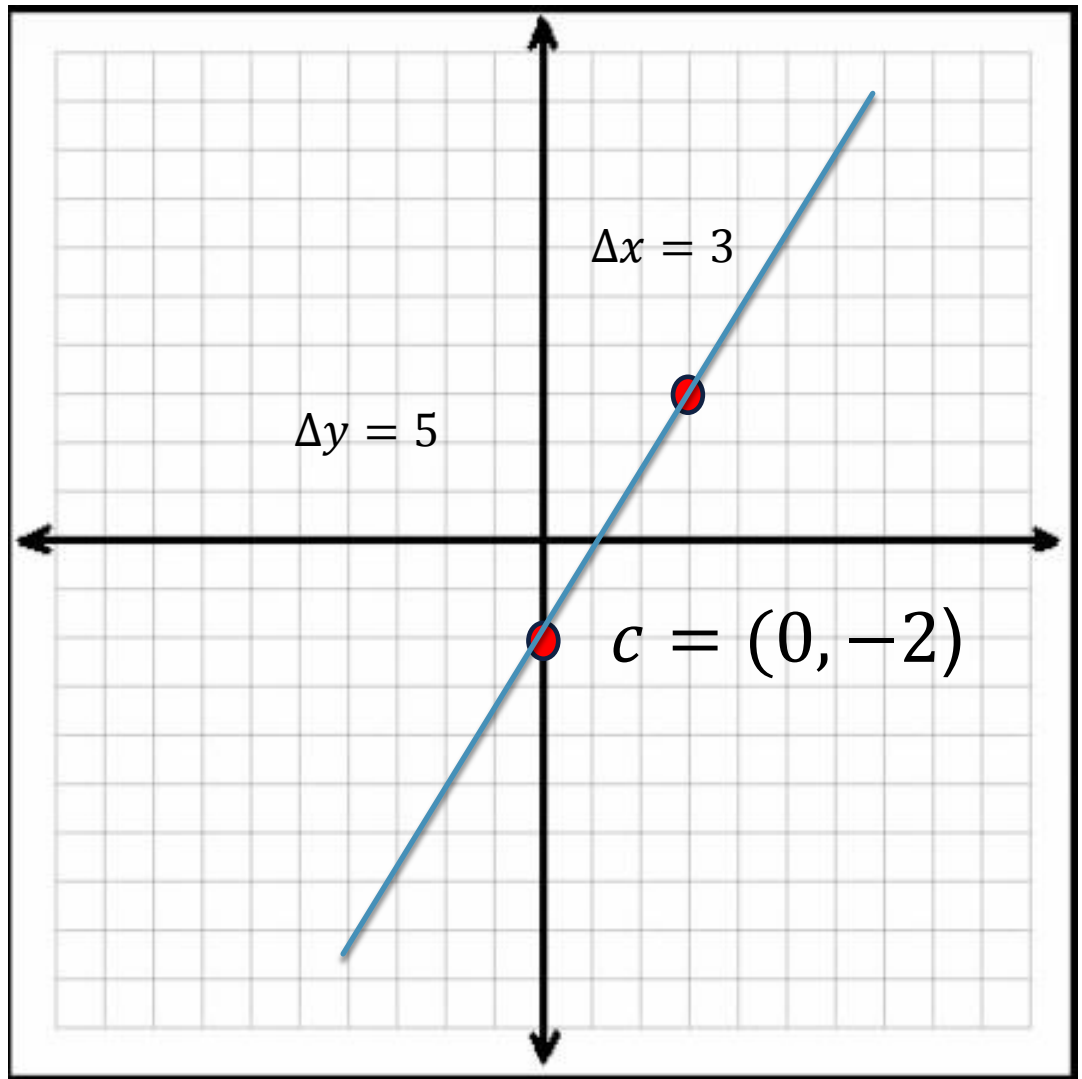
$$y = mx + c$$

$$m = \frac{\Delta Y}{\Delta X}$$

$$y = \frac{5}{3}x - 2$$

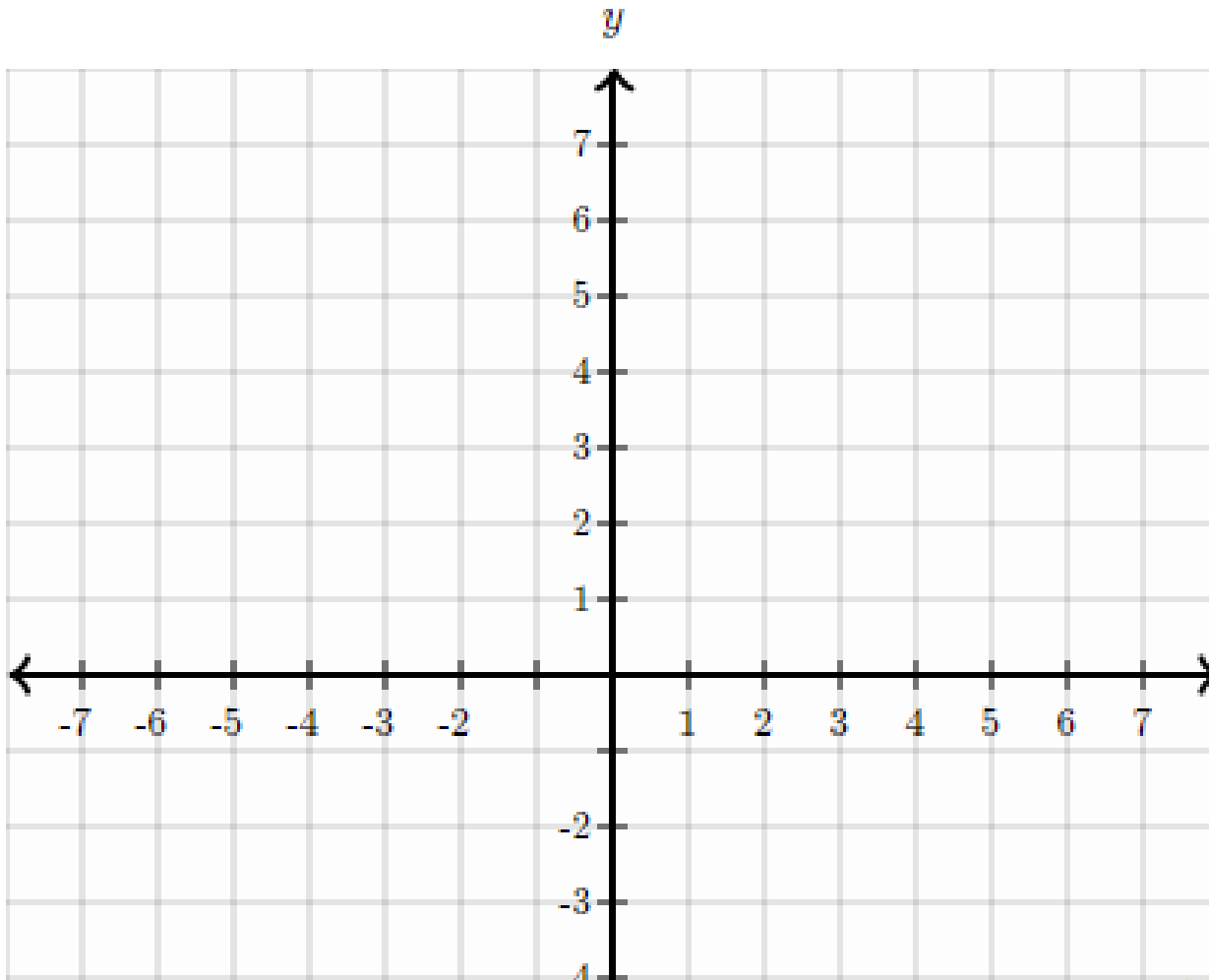
$$c = -2$$

$$m = \frac{\Delta Y}{\Delta X} = \frac{5}{3}$$

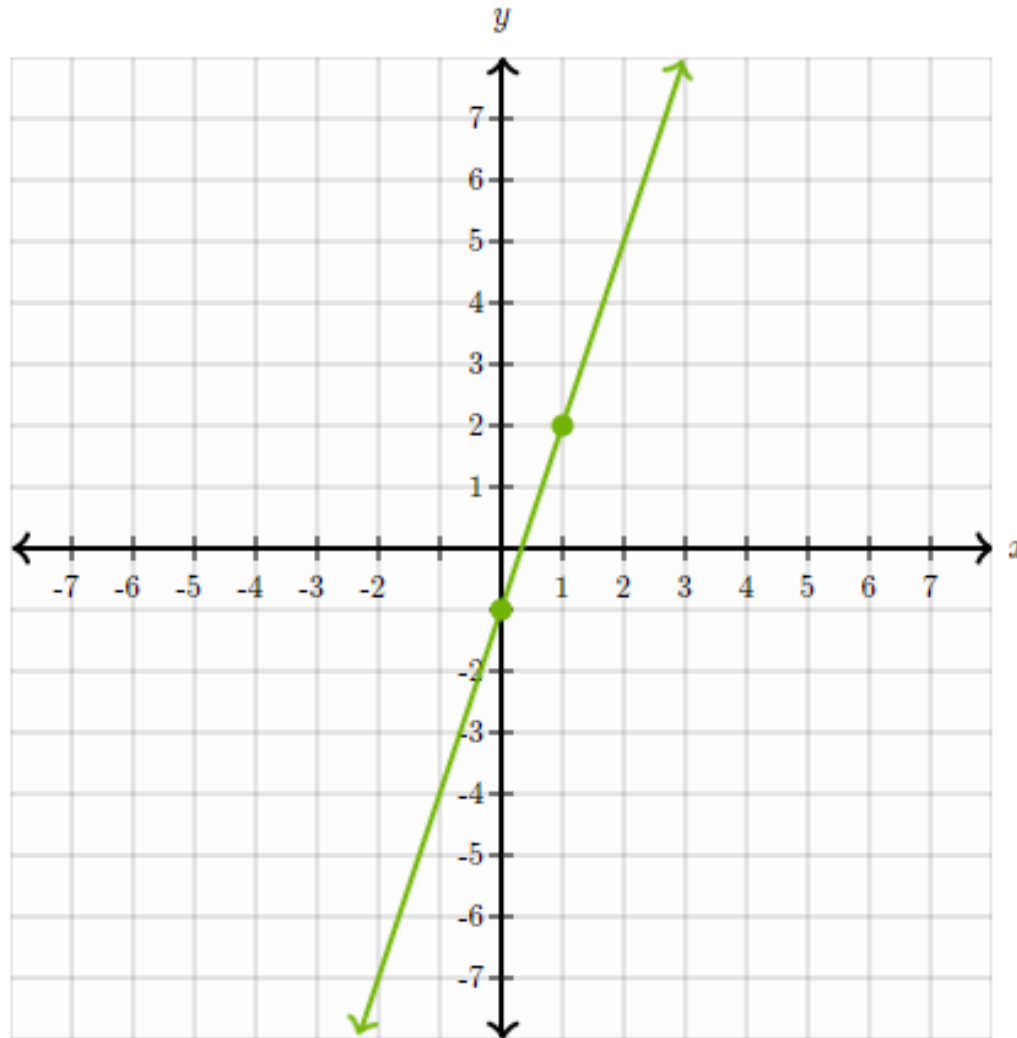


# PRACTICE: DRAW LINE

$$y = 3x - 1$$

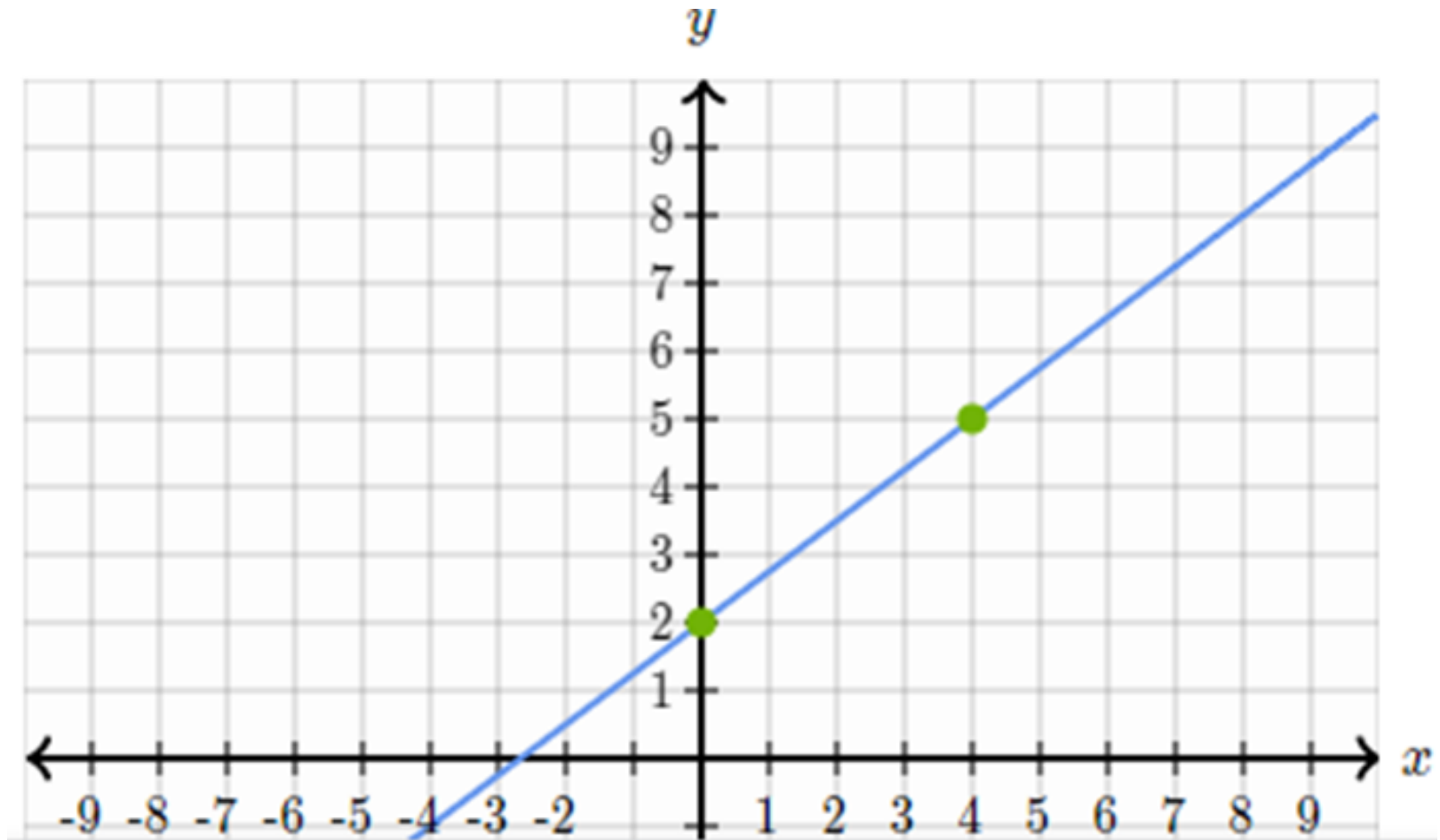


# PRACTICE: DRAW LINE $y=3x-1$



# ONE MORE EXAMPLE

Graph  $y = \frac{3}{4}x + 2$ .



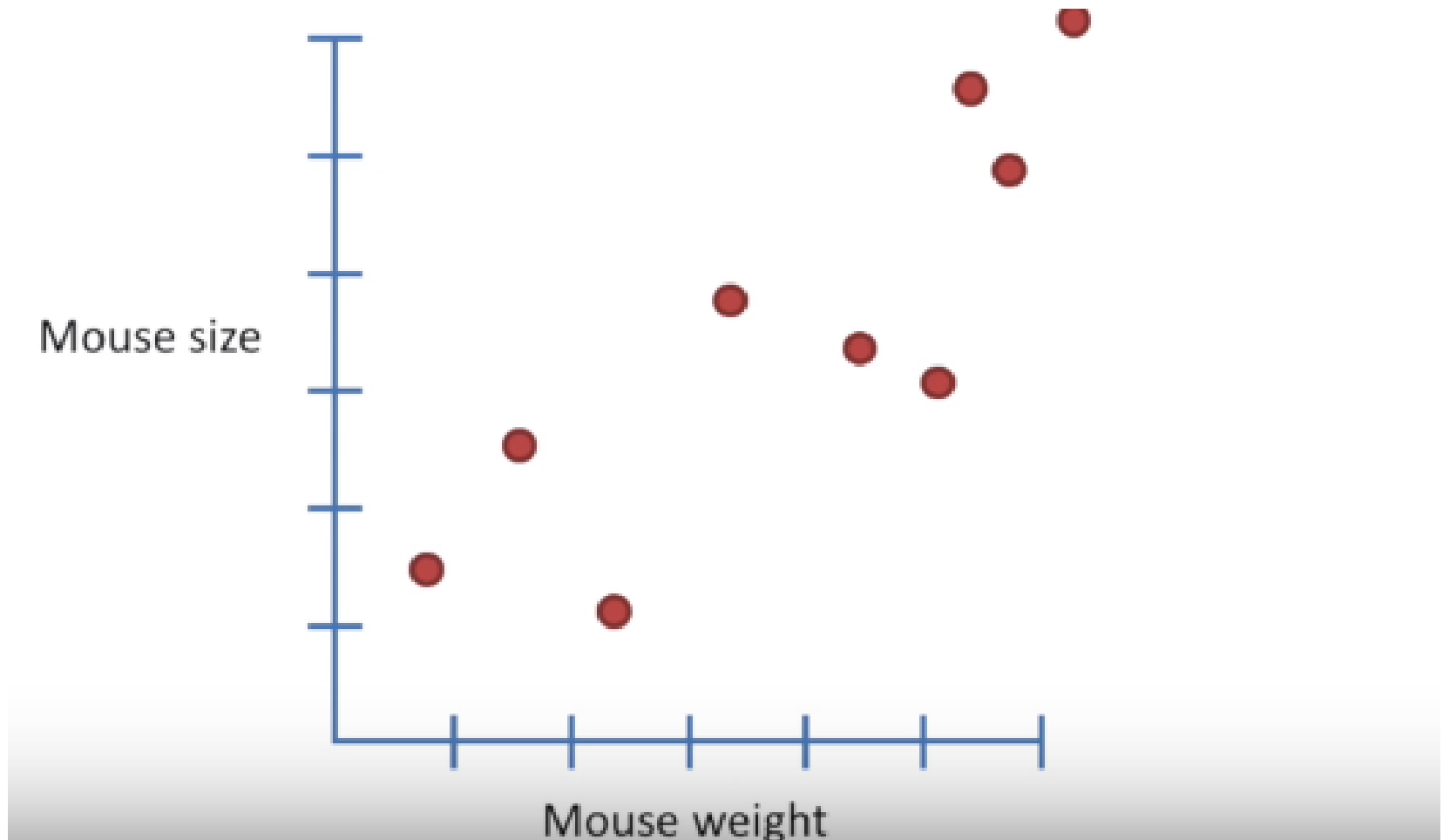
# RATE OF CHANGE

- For Linear Functions, Slope represents the rate of change of function and it is constant.

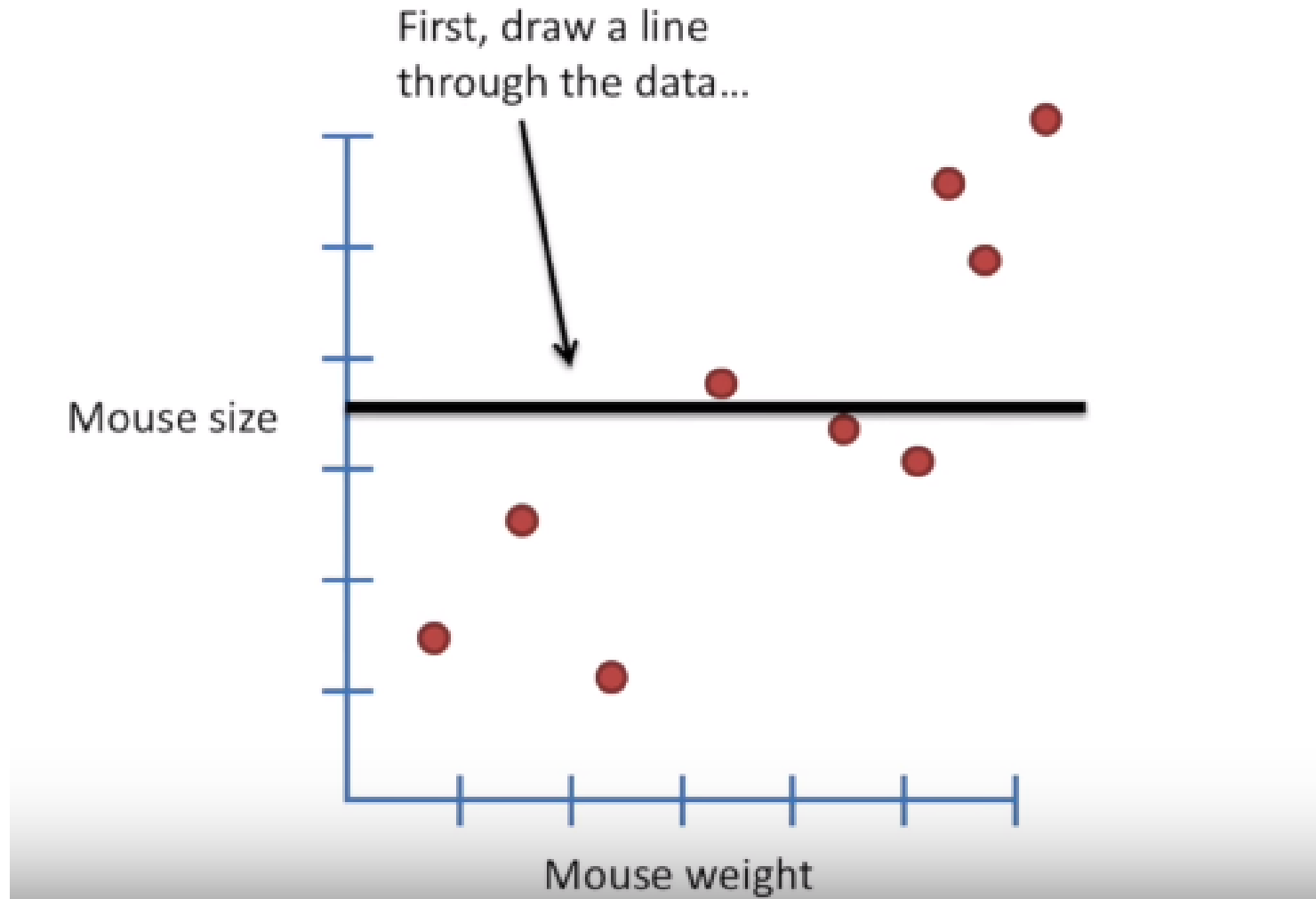
# WHAT IS BEST SUITABLE

- What is best suitable parameters ?  
Let's visualize the problem

# DATA

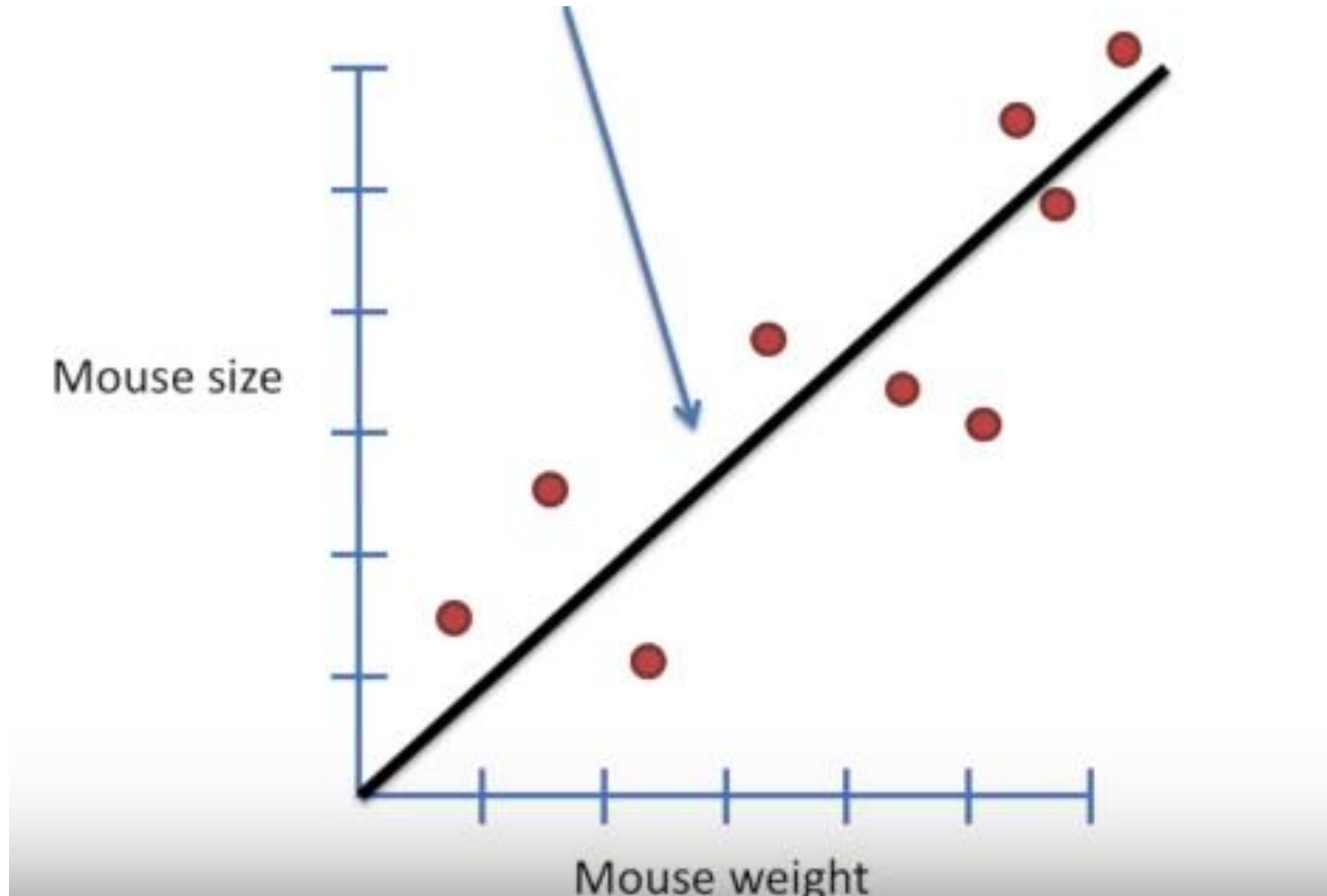


# ONE POSSIBLE FUNCTION

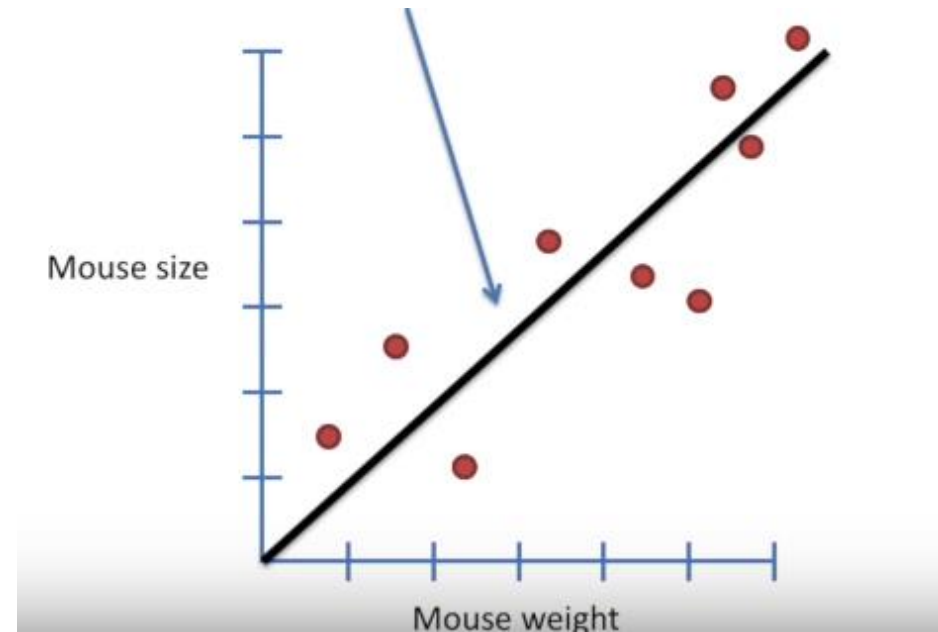
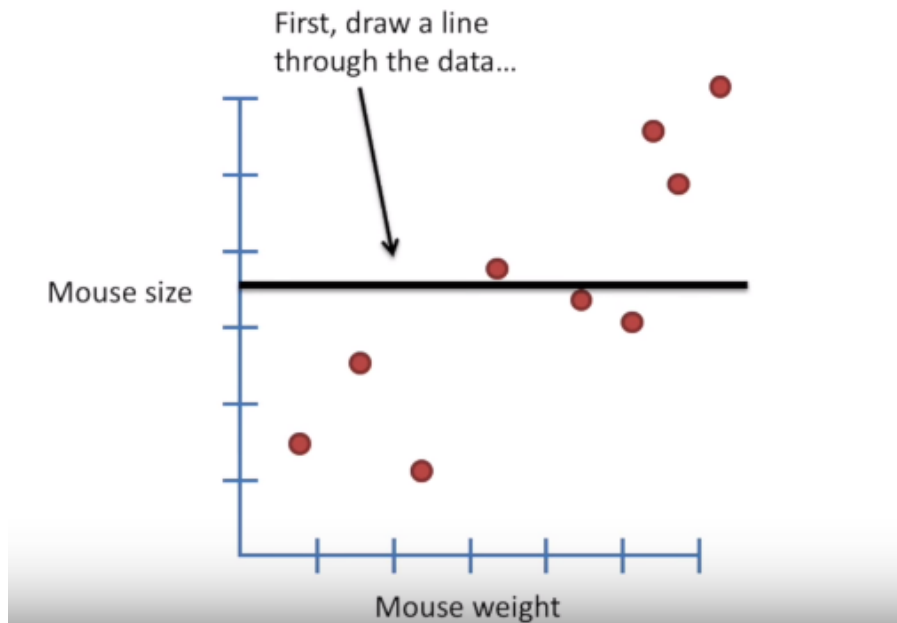




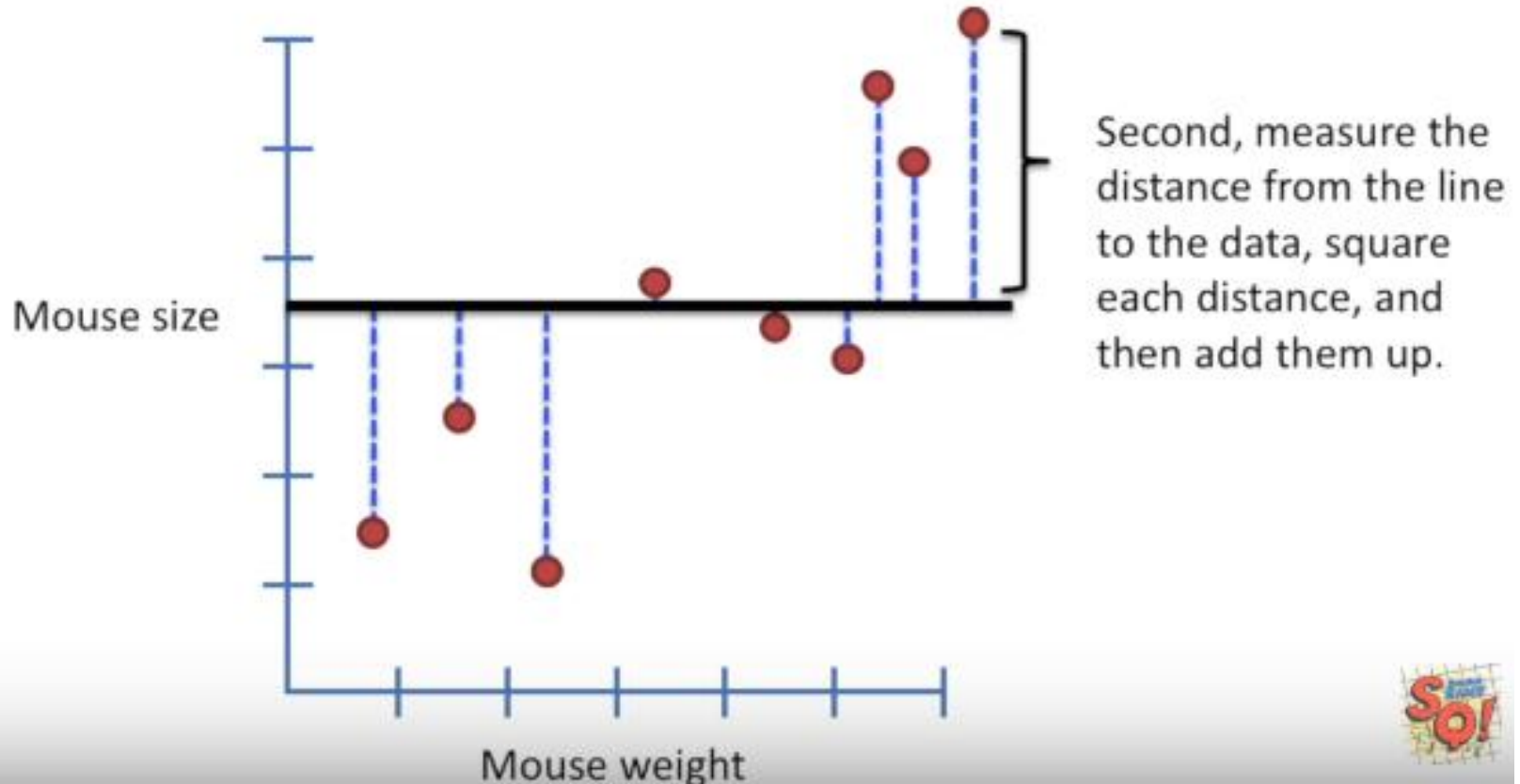
# ANOTHER POSSIBLE FUNCTION



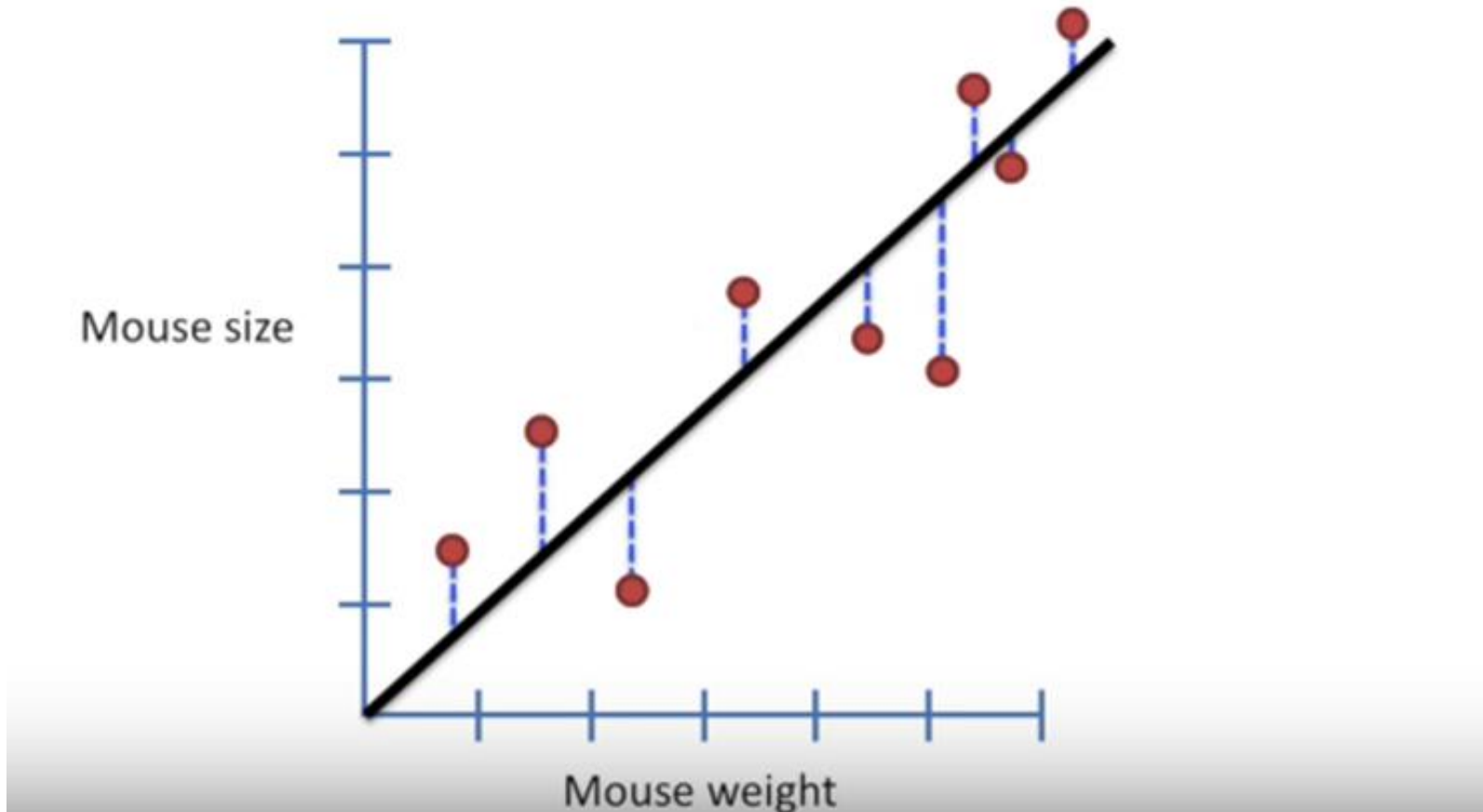
# WHAT IS BEST LINE (LINEAR FUNCTION) ?



# CALCULATE THE ERROR:

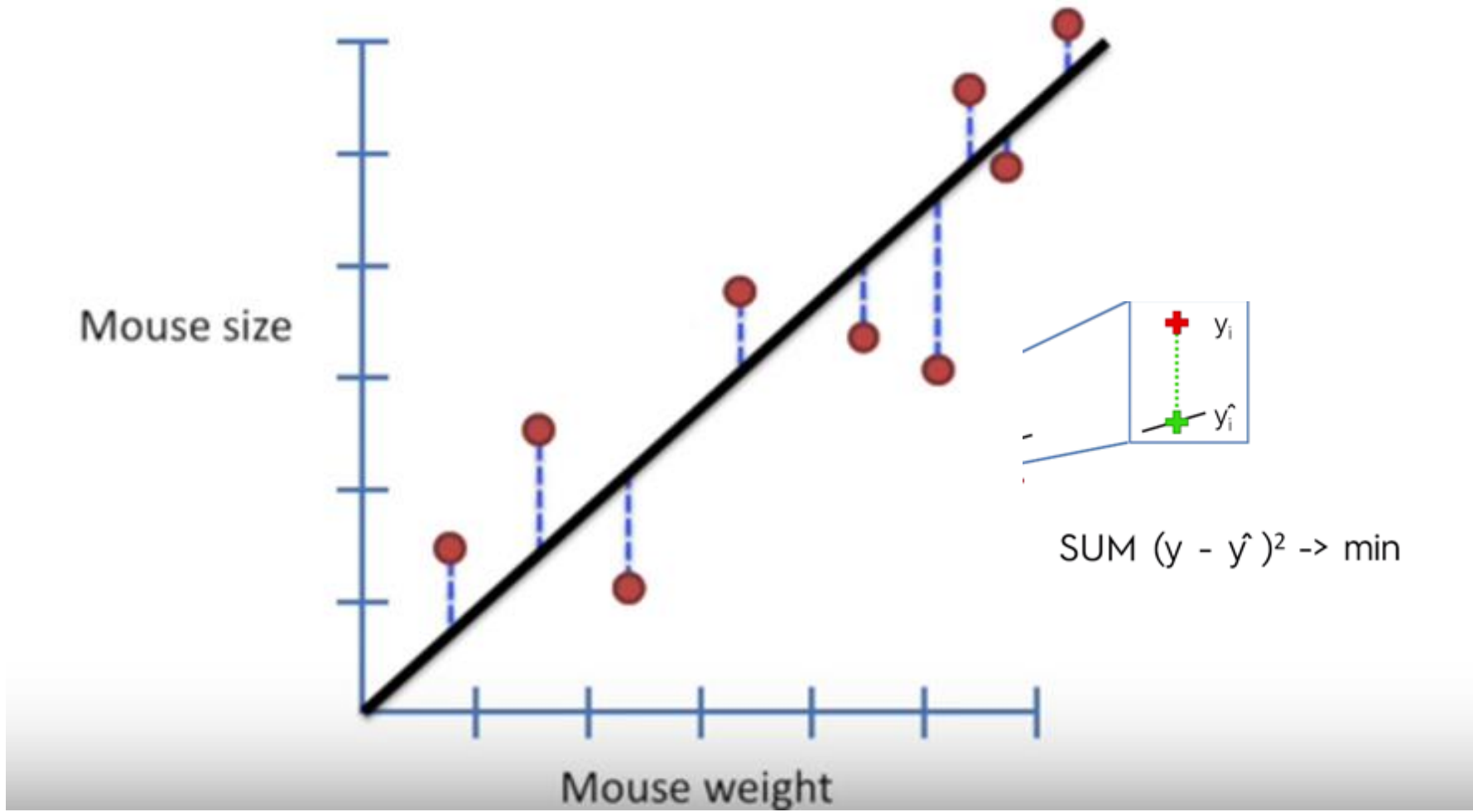


# CALCULATE THE ERROR



## ■ How to calculate Error ?

# LEAST SQUARE: ONE WAY TO FIND ERROR



# LOSS FUNCTION

- The function that find the error is called the LOSS function. This calculate the Loss of prediction for single example.
- $J(\theta) = (y^i - f(x^i))^2$

# COST FUNCTION

- Over all cost for all examples, this function is called cost function.

- $$J(\theta) = \frac{1}{2m} \sum_{i=1}^m (y^i - f(x^i))^2$$



# COST FUNCTION

- Over all cost for all examples, this function is called cost function.

- $$J(\theta) = \frac{1}{2m} \sum_{i=1}^m (y^i - f(x^i))^2$$

Remember here  $f(x^i) = \theta_0 + \theta_1 x^i$

# OUR OBJECTIVE

- So finally, our objective is to find the value of  $\theta_0$  and  $\theta_1$  such that the value of  $J(\theta)$  is minimized.

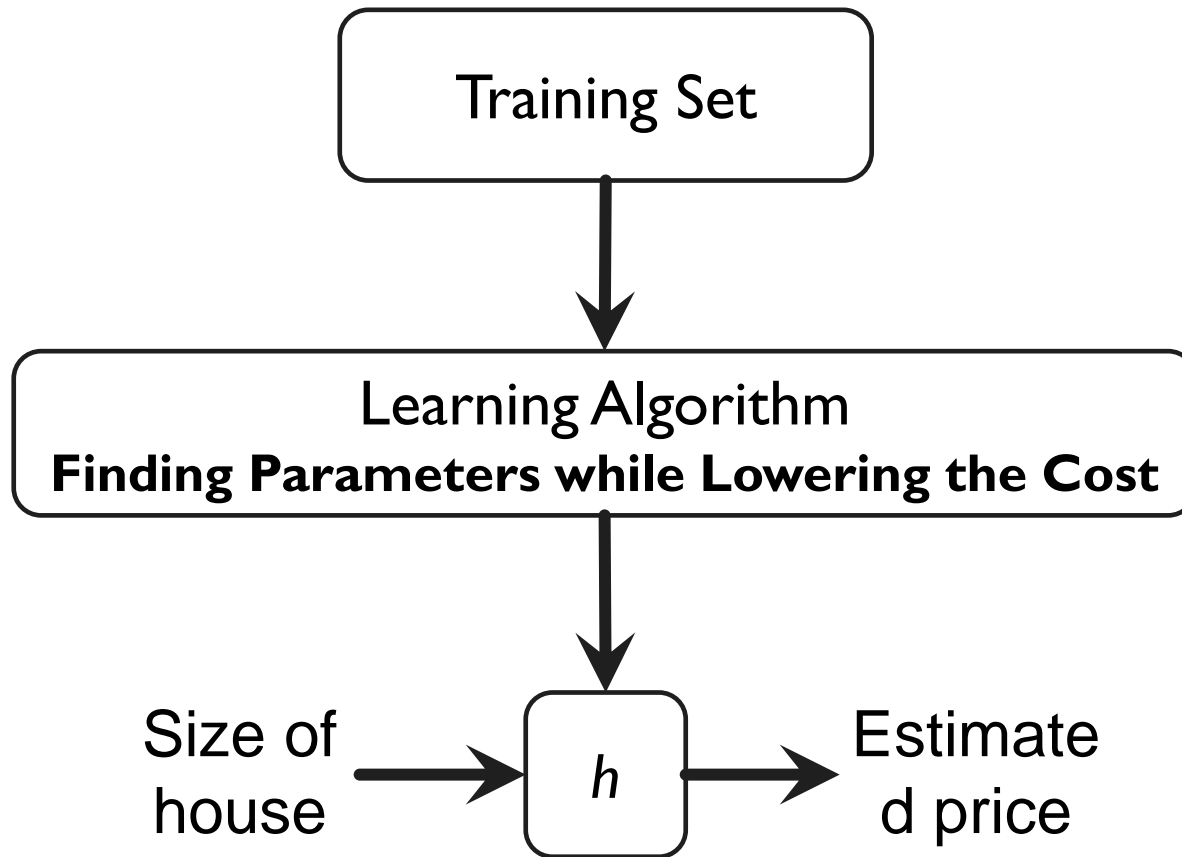
- $$J(\theta) = \frac{1}{2m} \sum_{i=1}^m (y^i - f(x^i))^2$$

Remember here  $f(x^i) = \theta_0 + \theta_1 x^i$

# OUR OBJECTIVE

- So finally, our objective is to find the value of  $\theta_0$  and  $\theta_1$  such that the value of  $J(\theta)$  is minimized.

- $$J(\theta) = \frac{1}{2m} \sum_{i=1}^m (y^i - (\theta_0 + \theta_1 x^i))^2$$



# REVIEW

Hypothesis:  $h_{\theta}(x) = \theta_0 + \theta_1 x$

Parameters:  $\theta_0, \theta_1$

Cost Function:  $J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$

Goal:  $\underset{\theta_0, \theta_1}{\text{minimize}} J(\theta_0, \theta_1)$

# FIND THE MODEL

- With help of cost function find the  $\theta_0$  and  $\theta_1$ .

x	y
0	1
1	4
2	6
3	11
4	14
5	18
6	17

## HINT: FIND THE MODEL

- With help of cost function find the  $\theta_0$  and  $\theta_1$ .
- Assume  $\theta_0$  and  $\theta_1$  values ranges from 1 to 5 (integers) only, now for each combination calculate the  $J(\theta)$
- Select the value of  $\theta_0$  and  $\theta_1$  where  $J(\theta)$  has minimum value.

x	y
0	1
1	4
2	6
3	11
4	14
5	18
6	17

## VIEW THE $J(\theta)$

- Draw graph of  $J(\theta)$  (as dependent variable) and  $\theta_0$  (as independent variable)
- Draw graph of  $J(\theta)$  (as dependent variable) and  $\theta_1$  (as independent variable)



# POSSIBLE EXAM QUESTION

- For the given two model and data set, select best model for Univariate Linear Regression

# PROGRAMMING ASSIGNMENT 01

- Add comma separated Dataset in data.csv file.
- Python program take input two parameter  $\theta_0$  and  $\theta_1$  of a hypothesis function and display total cost based on the given data.
- (write a python function that take two input and return the cost value)

# PROGRAMMING ASSIGNMENT 02

- Extend the previous assignment 01, now it should take parameters of two models and return the parameters of best model.

# PROGRAMMING ASSIGNMENT 03

- Extend the previous assignments, write a program that read the data and proposed the linear regression parameters.
- (You can add range constrain on parameters).

# PROGRAMMING ASSIGNMENT 03

- Extend the previous assignments, write a program that read the data and proposed the linear regression parameters.
- (You can add range constrain on parameters between 1 to 10).
- Hint you need to change the first parameter value and keep the second parameter constant

# PROGRAMMING ASSIGNMENT 04

- Calculate the runtime of algorithm

# PROGRAMMING ASSIGNMENT 05

- Repeat the Experiment for following ranges of parameters and fill the table.

Range	Number of Iterations	Time in Seconds
1 to 10		
1 to 100		
1 to 1000		
1 to 10000		