

# **Data Science**

- Start of the Journey

# What is the Relationship ?

$Y = \text{????????}$

<u>X</u>	<u>Y</u>
2	8
6	20
4	14
3	11
7	23
4	14
2	8
5	17

# Relationship

$$Y = 2 + 3(X)$$

<u>X</u>	<u>Y</u>
2	8
6	20
4	14
3	11
7	23
4	14
2	8
5	17

Find the Y in ?

$$Y = 2 + 3(X)$$

<u>X</u>	<u>Y</u>
2	8
6	20
4	14
3	11
7	23
4	14
2	8
5	17
10	?
1	?

## Value for Y with given X

$$Y = 2 + 3(X)$$

<u>X</u>	<u>Y</u>
2	8
6	20
4	14
3	11
7	23
4	14
2	8
5	17
10	32
1	5

# Terminology

$$Y = 2 + 3(X)$$

**Y = Model**

**2 = Intercept**

**3 = Slope**

**X = input**

<u><b>X</b></u>	<u><b>Y</b></u>
2	8
6	20
4	14
3	11
7	23
4	14
2	8
5	17
10	32
1	5

Predict the price of House ?

# Price of a House



\$70,000



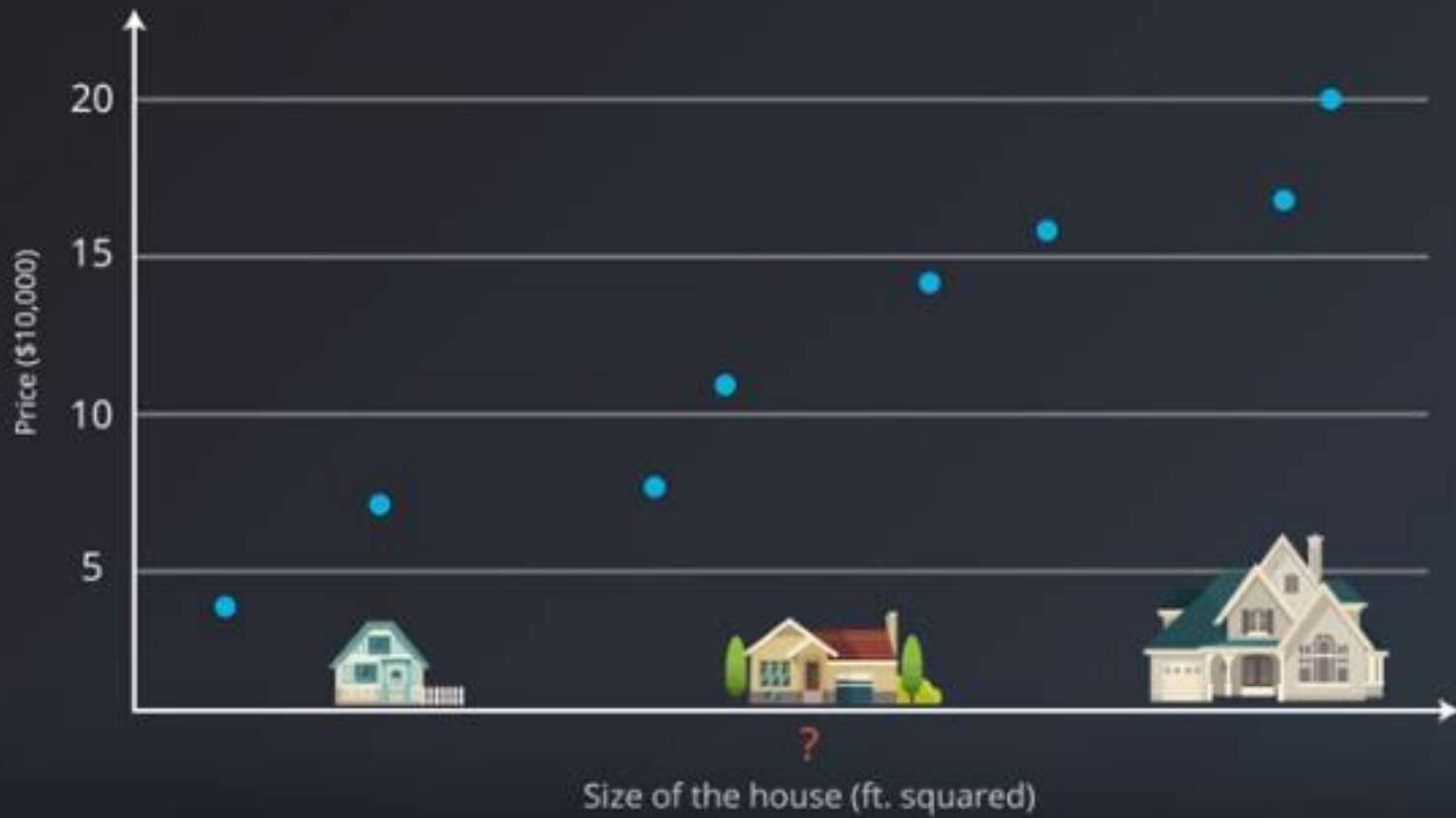
?

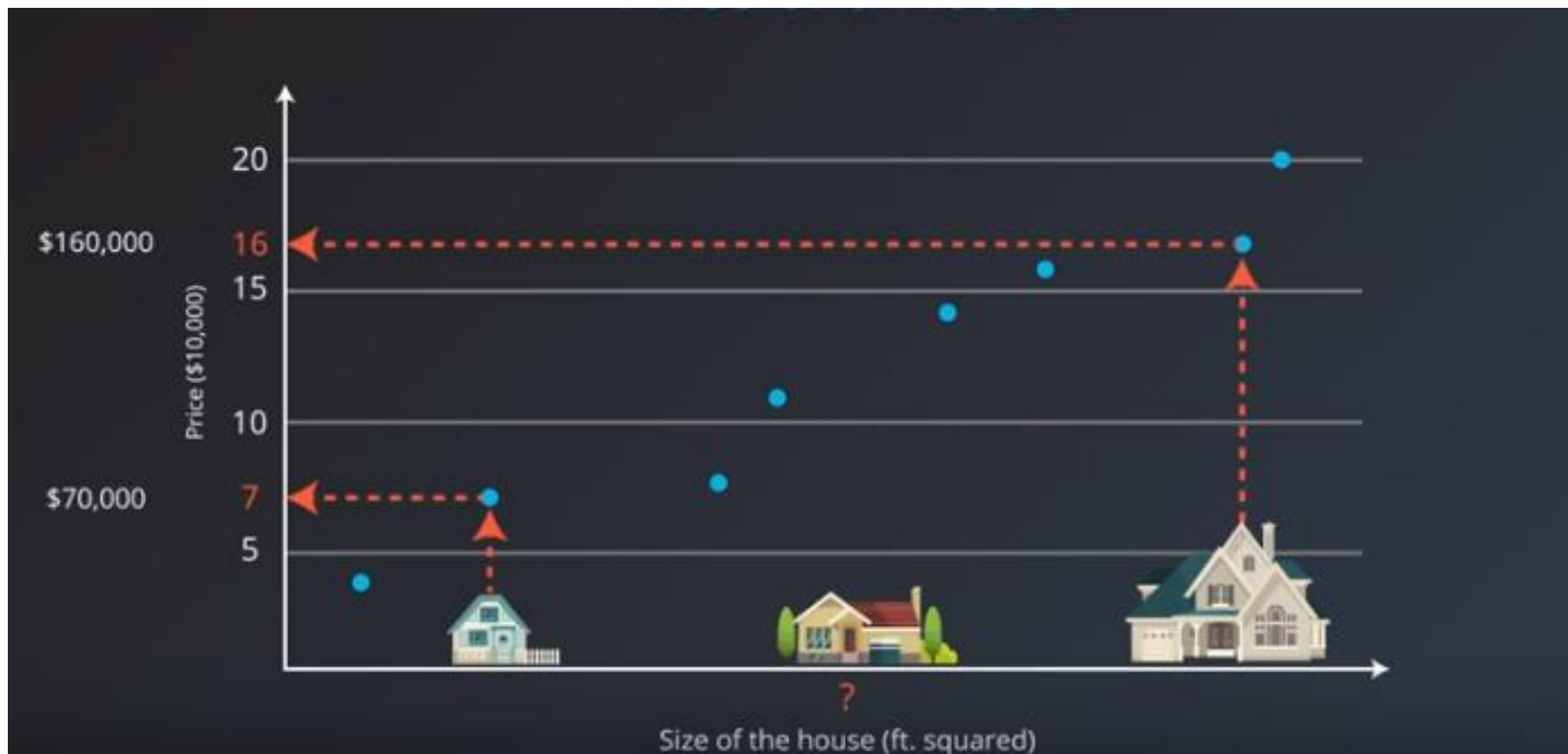


\$160,000









# It's all about

- Finding the “best-fit” line is the **goal** of simple linear regression.

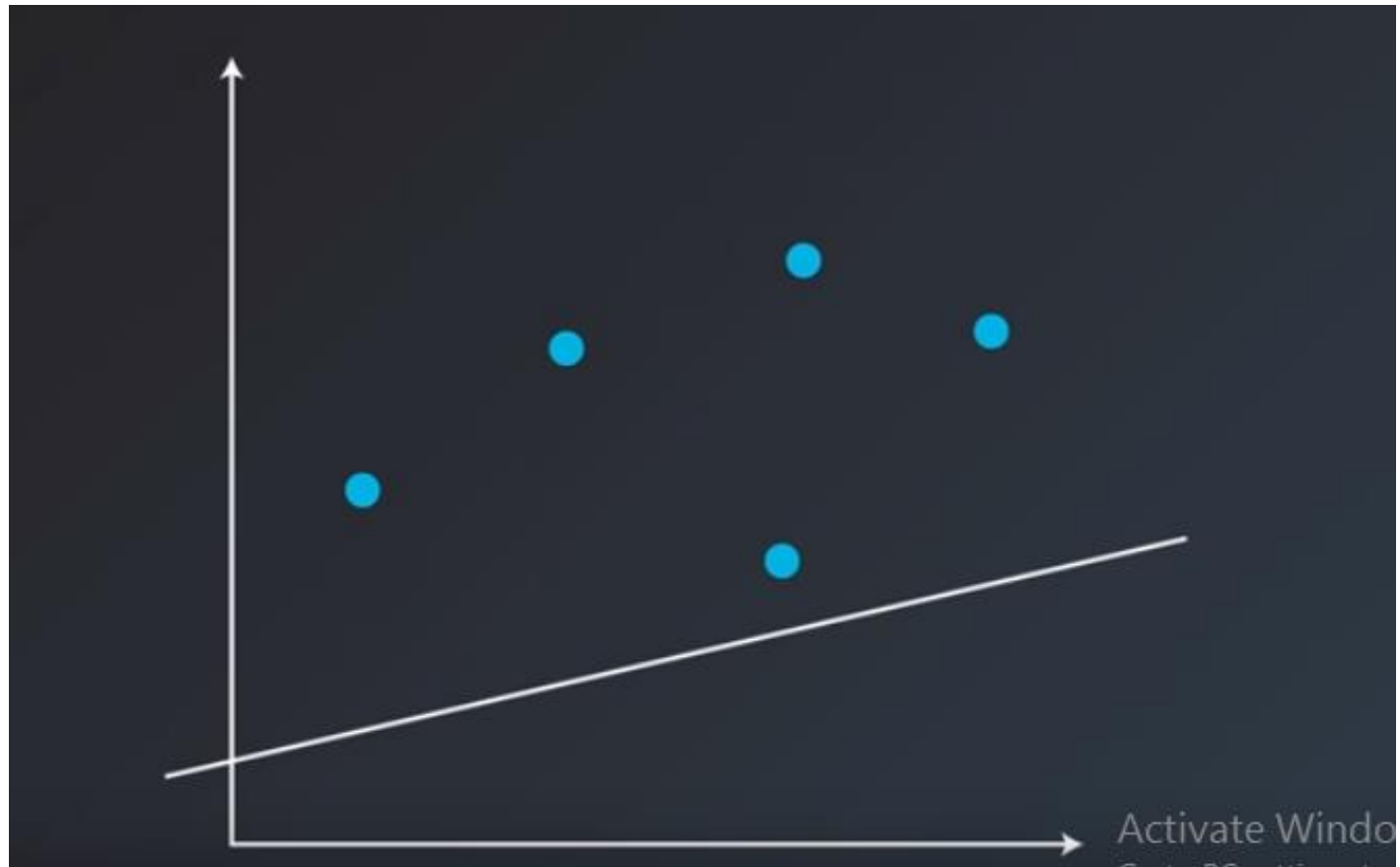
# Linear Regression

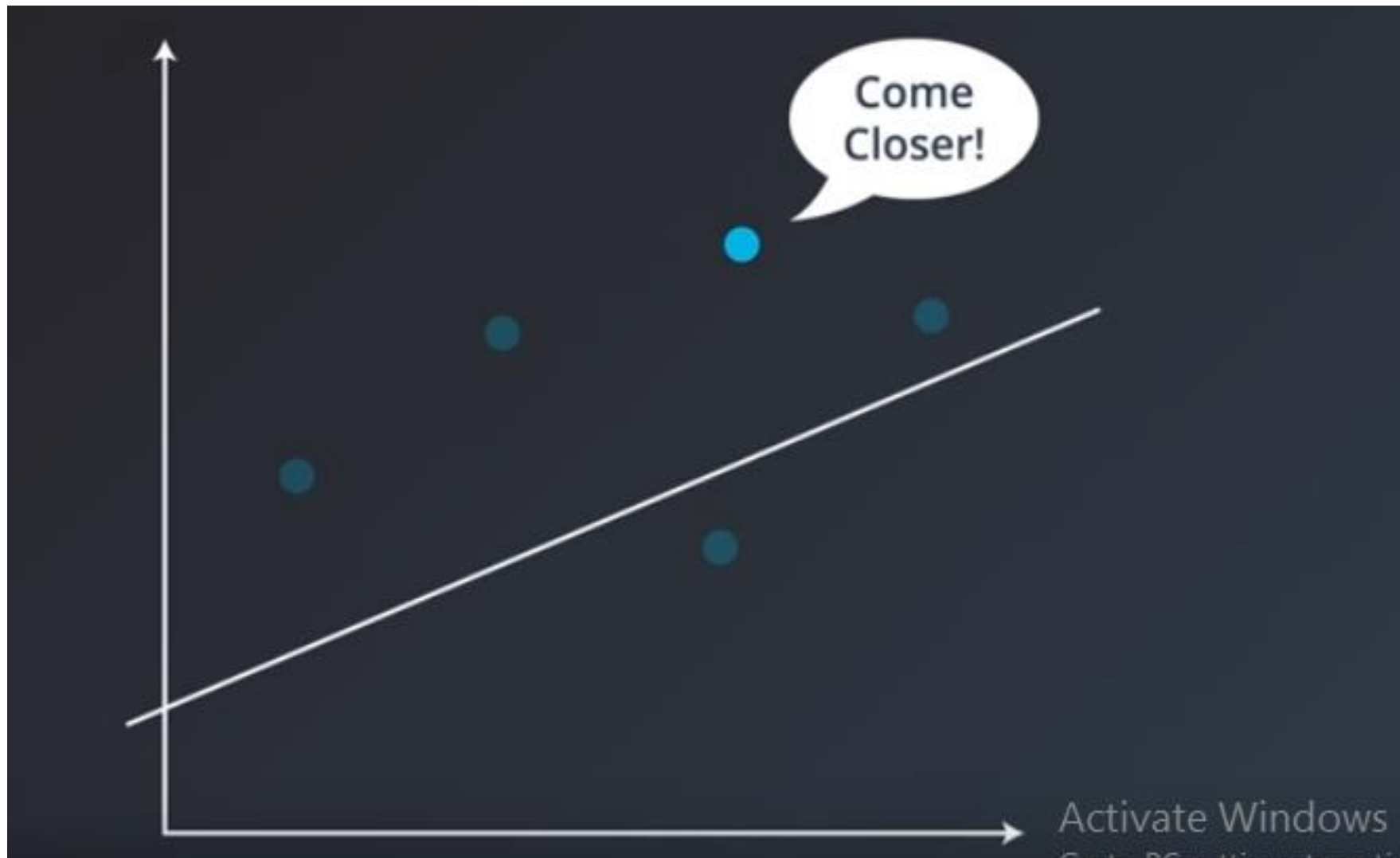
- *Welcome to the world of data science*

# What is Simple Linear Regression?

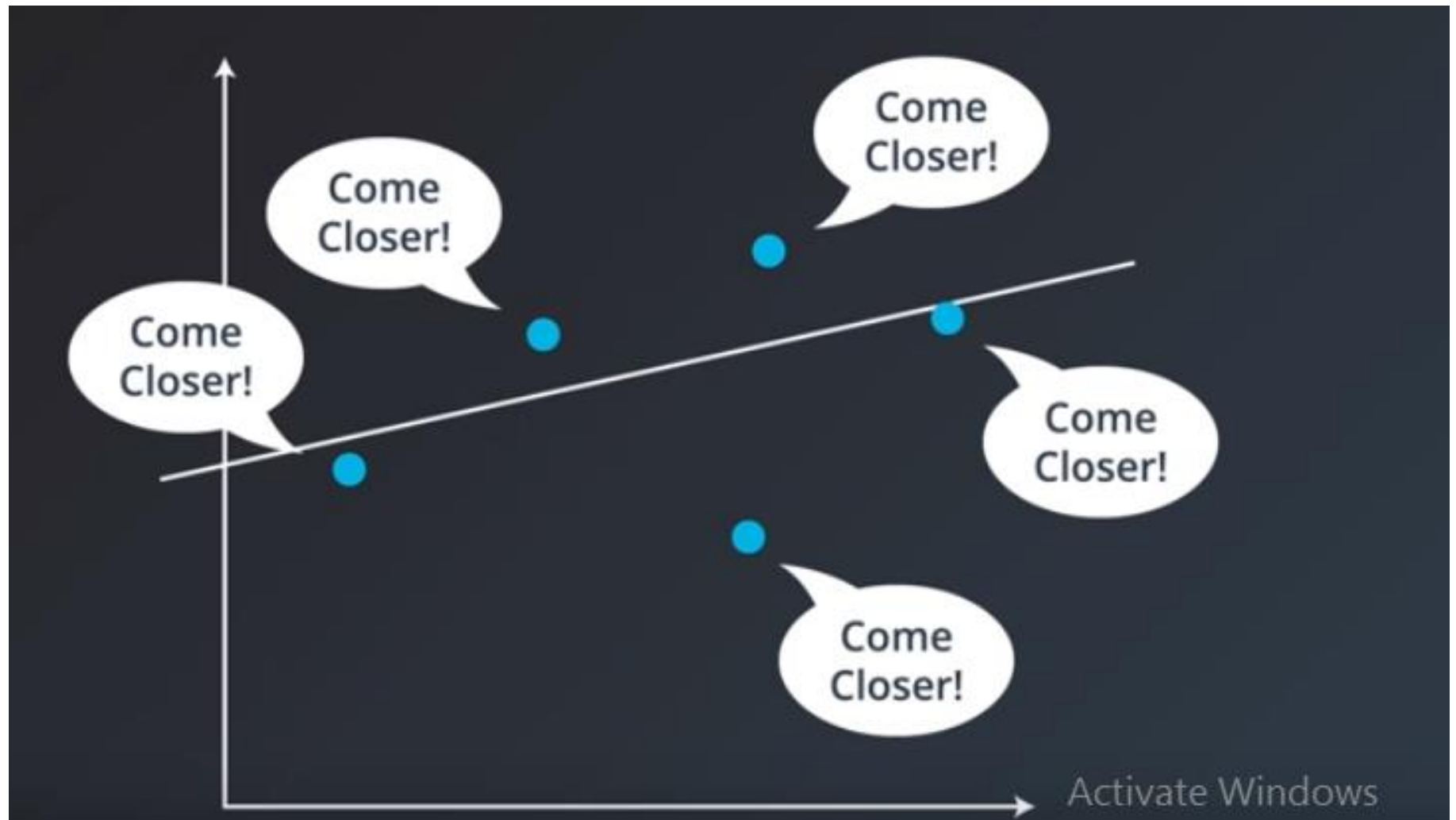
- Simple Linear Regression is a method used to fit the **best straight line** between a set of datapoints.
- After a graph is properly scaled, the data points must “look” like they would fit a straight line, not a parabola, or any other shape.
- The line is used as a model in order to predict a variable  $y$  from another variable  $x$ .
- A regression line must involve 2 variables, the dependent and the independent variable.
- Finding the “best-fit” line is the **goal** of simple linear regression.

# Fitting A Line





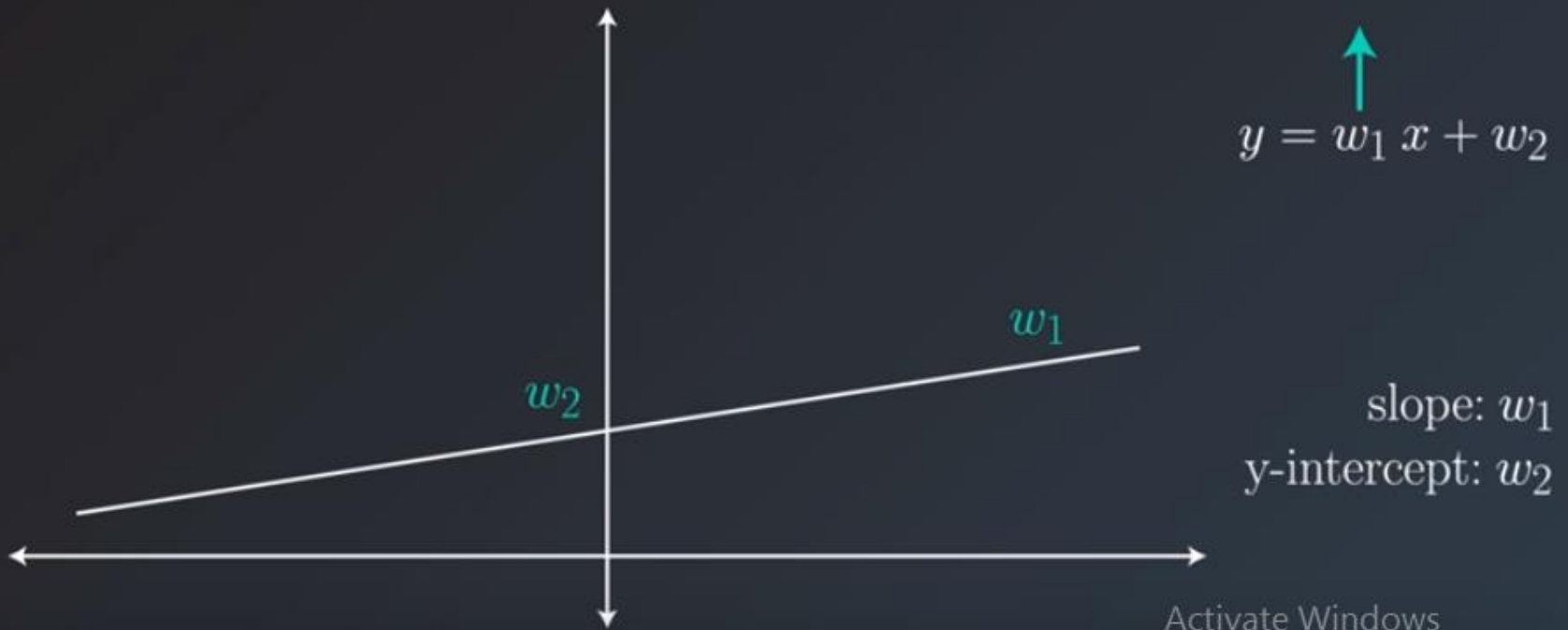


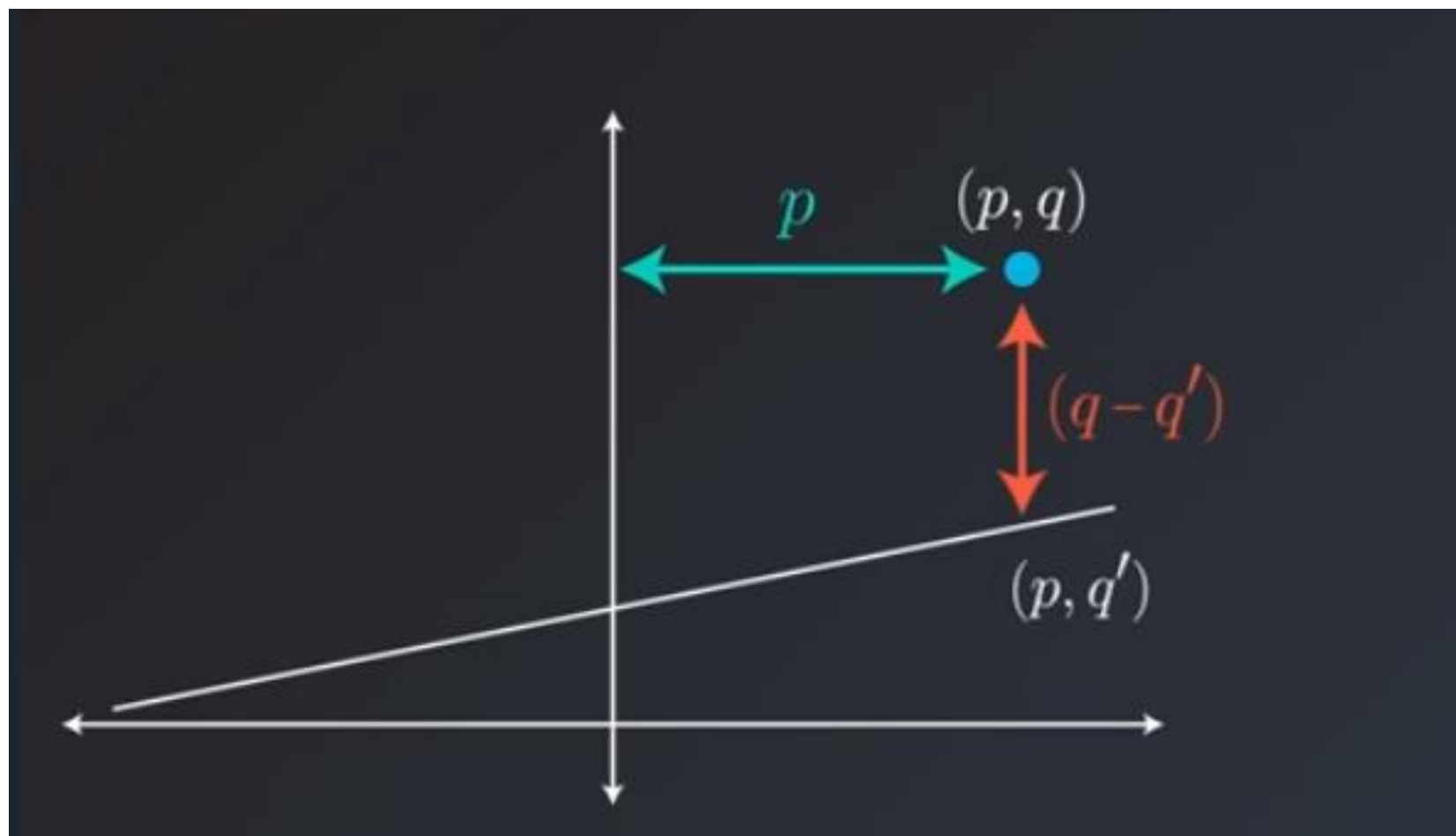


# Equation of a Straight Line

$$y = w_1 x + w_2$$

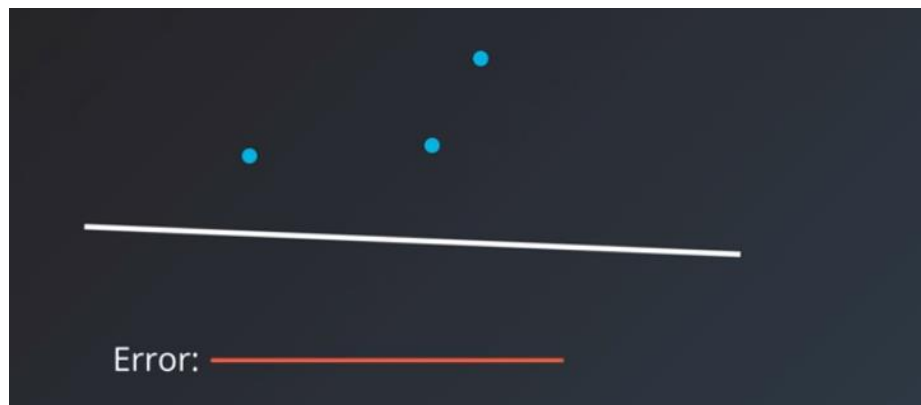
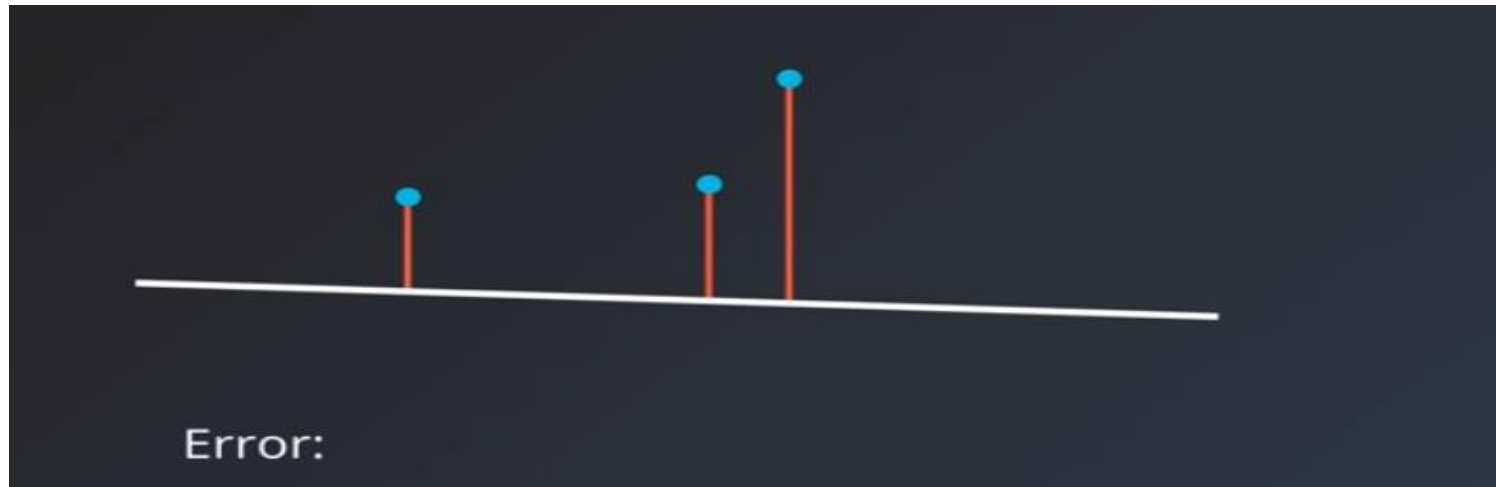
# Moving A Line

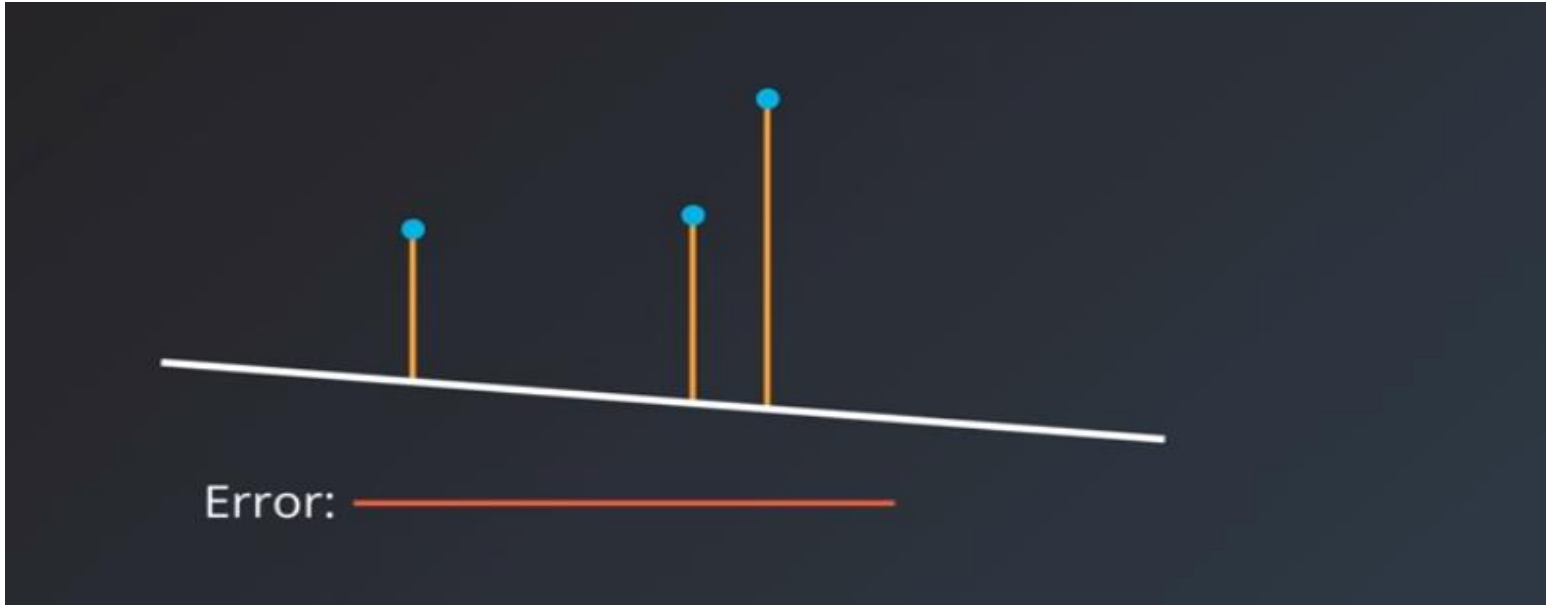


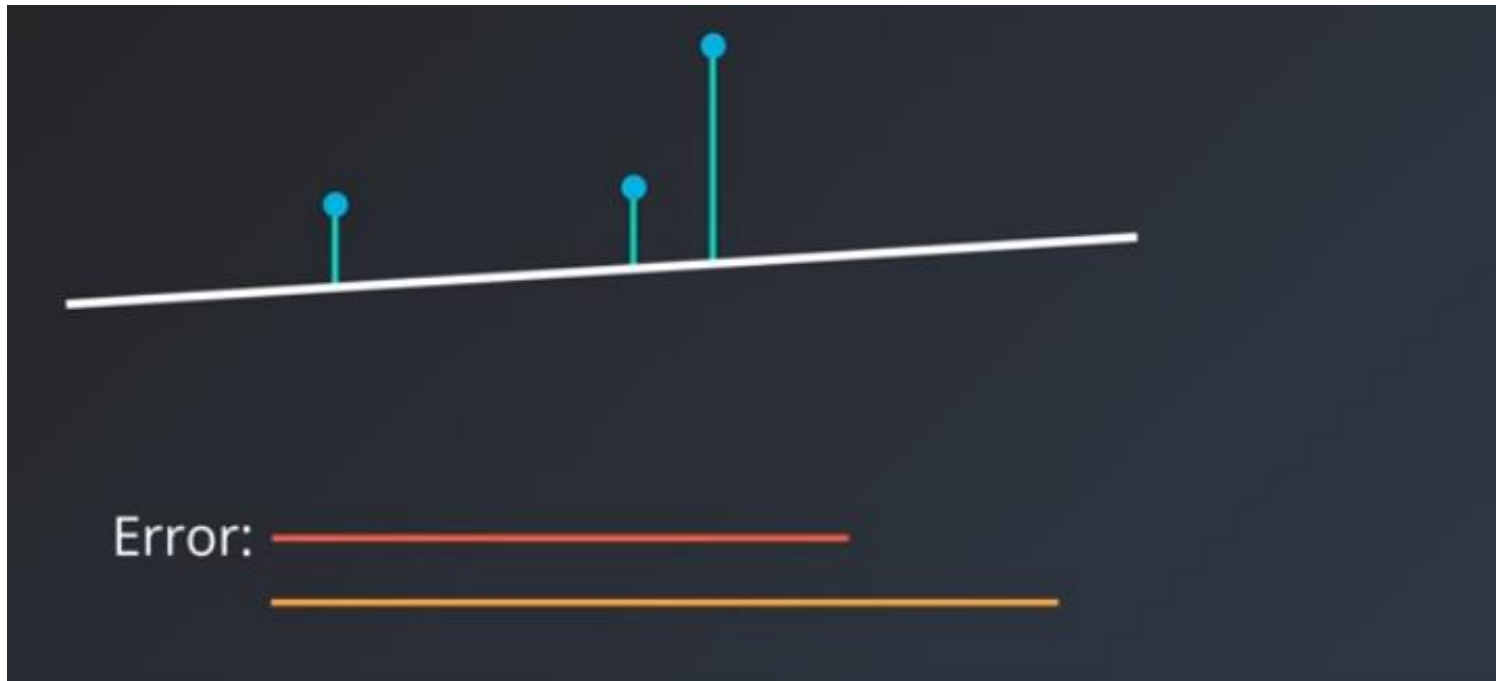


$$y = w_1 x + w_2$$

# Line vs Error







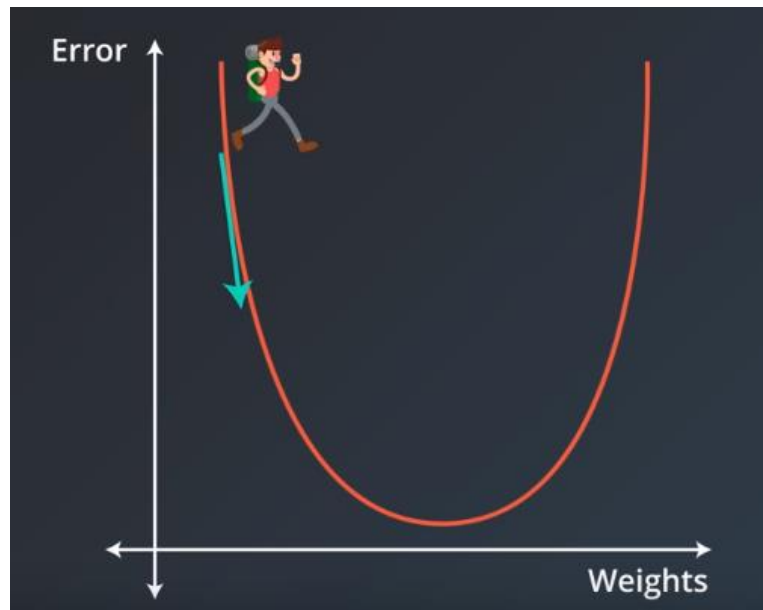
**How Should the Line move ?**



# Gradient Descent

Error Function

- Gradient of  
Error Function



Minimize  
the Error



Gradient  
Descent

# Gradient Descent

Error Function

- Gradient of  
Error Function

$$w_i \rightarrow w_i - \alpha \frac{\partial}{\partial w_i} \text{Error}$$



$$\frac{\partial}{\partial w_1} Error = -(y - \hat{y}) x$$

$$\frac{\partial}{\partial w_2} Error = -(y - \hat{y})$$

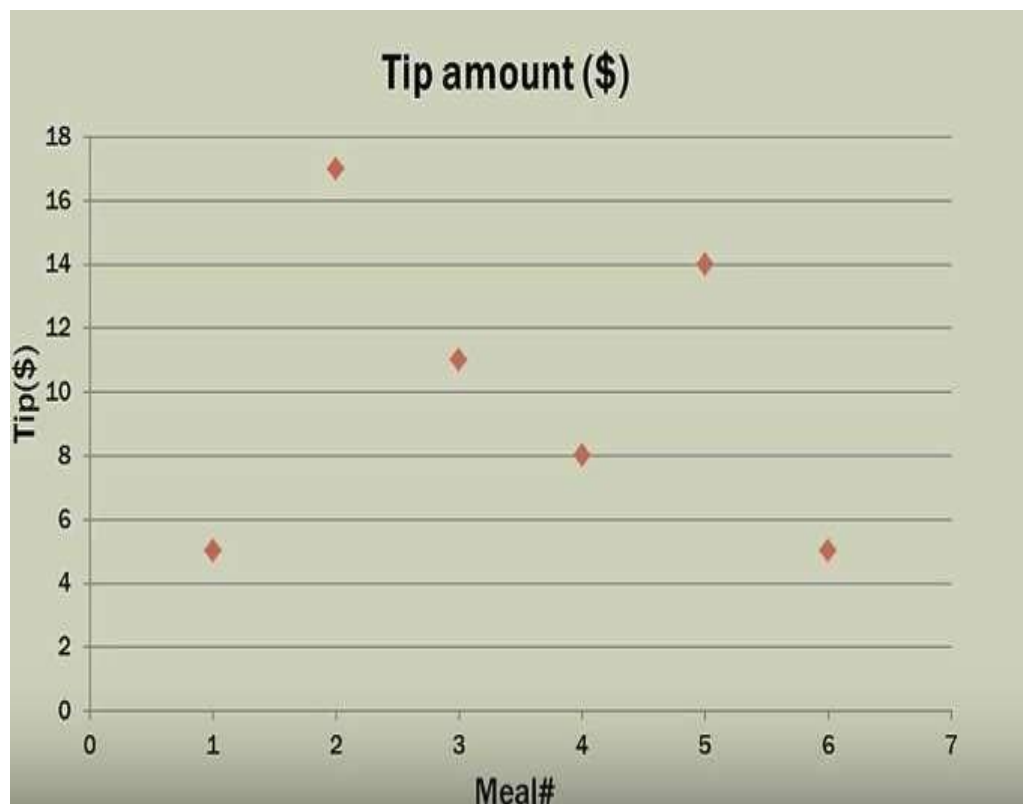
# One Variable

-No Independent variable

- **Problem:** A waiter wants to predict his next tip, but he forgot to record the bill amounts for previous tips.
- Here is a graph of his tips. The tips is the only variable. Let's call it the y variable.
- Meal# is not a variable. It is simply used to identify a tip.

Meal#	Tip amount (\$)
1	5.00
2	17.00
3	11.00
4	8.00
5	14.00
6	5.00

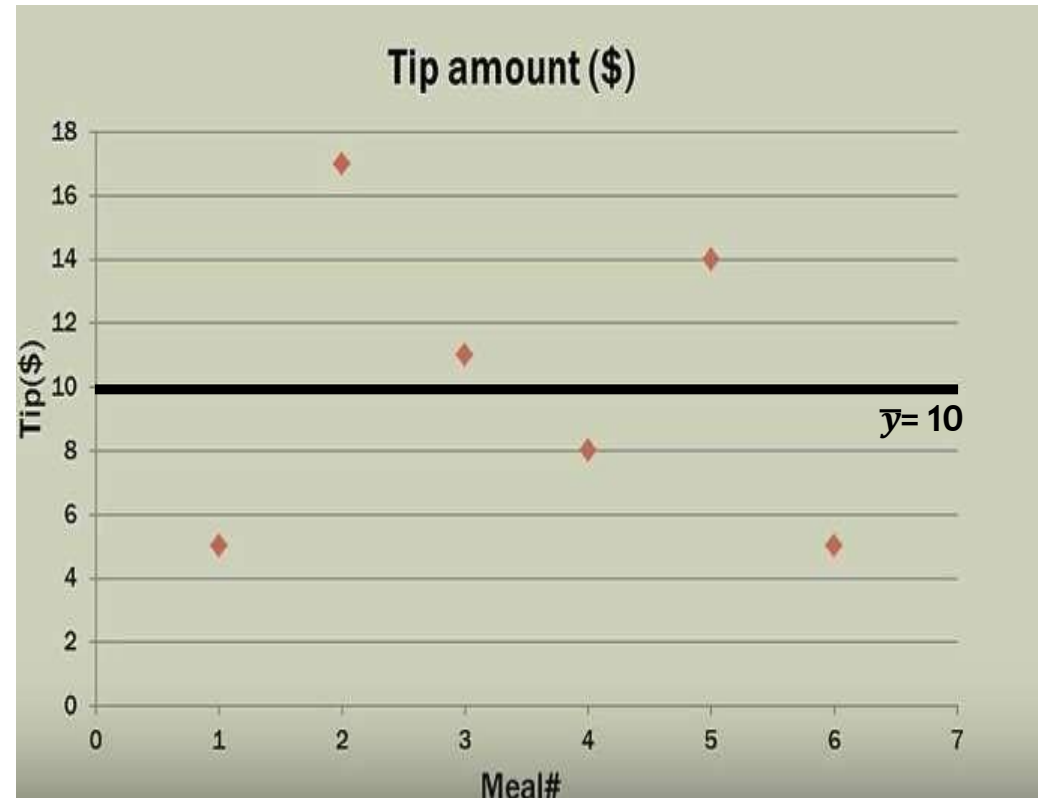
y variable



Can we come up with a model for this problem with only 1 variable?

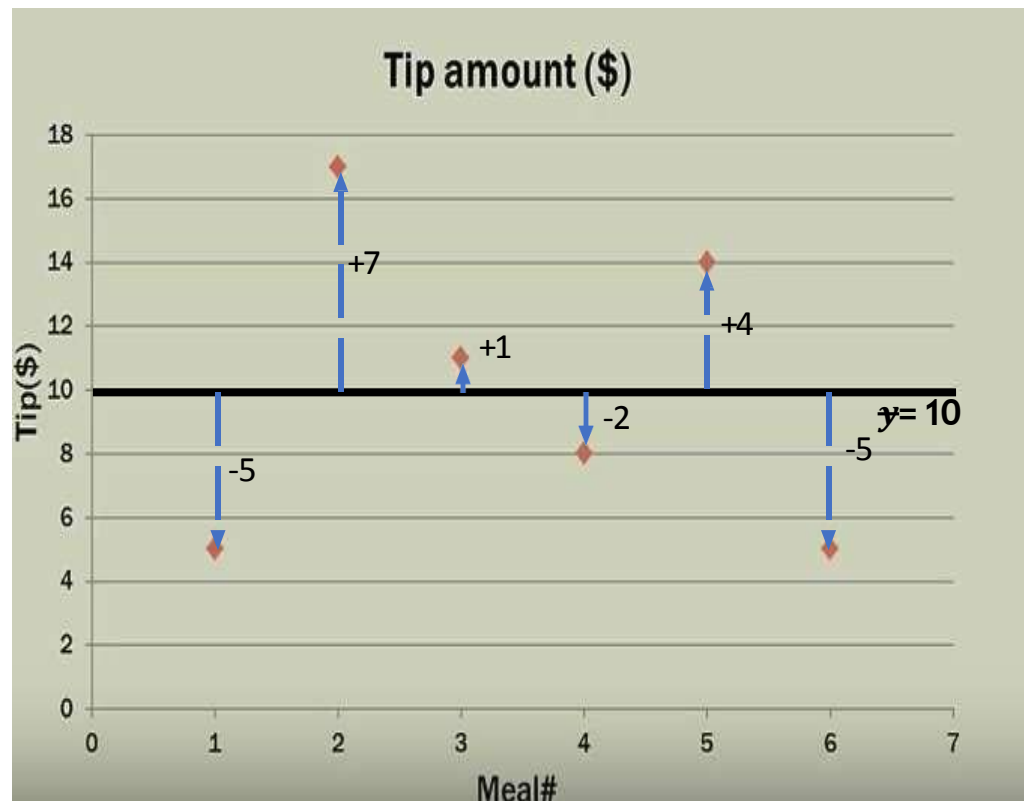
- The only option for our model is to use the mean of the Tips(\$)
- Tips are on the y axis. We would call the mean  $\bar{y}$  (y bar).
- The mean for the tip amounts is 10.
- The model for our problem is simply  $y = 10$ .
- $y = 10$  is our *best fit line* (represented by bold blackline).

Meal#	Tip amount (\$)
1	5.00
2	17.00
3	11.00
4	8.00
5	14.00
6	5.00



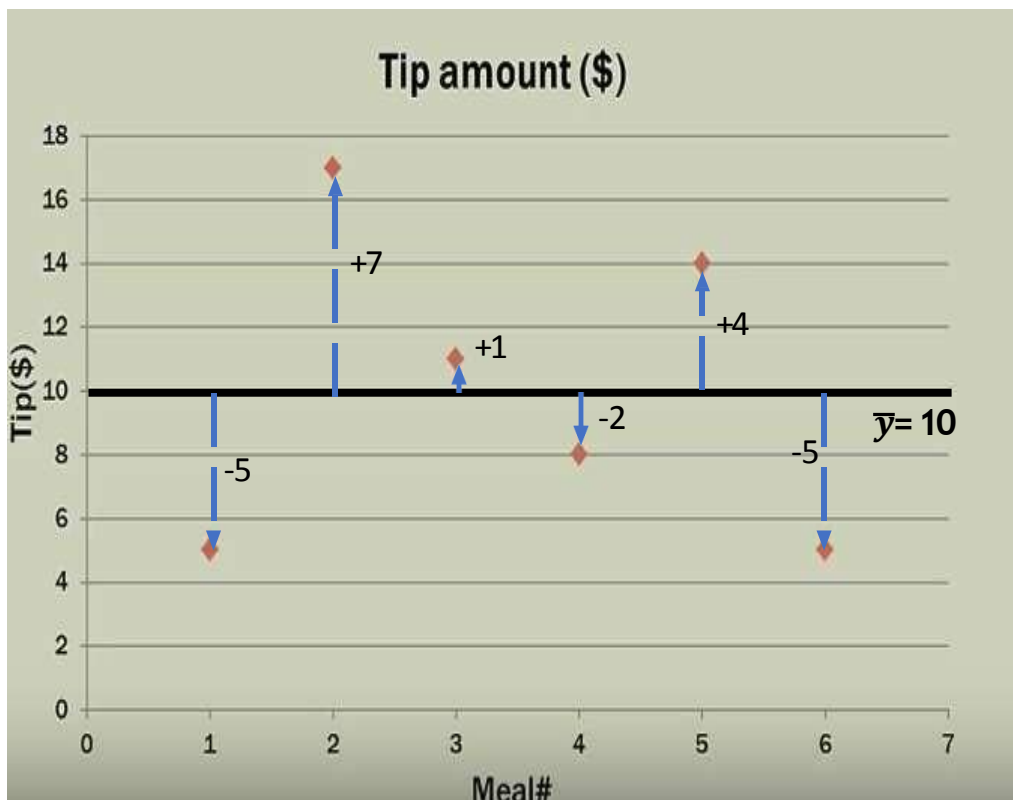
- Now, let's talk about goodness of fit. This will tell us how good our data points fit the line.
- We need to calculate the residuals (errors) for each point.

Meal#	Tip amount (\$)
1	5.00
2	17.00
3	11.00
4	8.00
5	14.00
6	5.00



- The best fit line is the one that minimizes the sum of the squares of the residuals (errors).
- The error is the difference between the actual data point and the point on the line.
- $SSE$  (Sum Of Squared Errors) =  $(-5)^2 + 7^2 + 1^2 + (-2)^2 + 4^2 + (-5)^2 = 120$

Meal#	Tip amount (\$)
1	5.00
2	17.00
3	11.00
4	8.00
5	14.00
6	5.00




- $SST$  (Sum Of Squared Total) =  $SSR$  (Sum Of Squared Regression) +  $SSE$  is the Sum Of Squares Equation.
- Since there is no regression line (as we only have 1 variable), we can not make the  $SSE$  any smaller than 120, because  $SSR = 0$ .



# Two Variables

- One Independent / Dependent variable

- **Repeating the Problem: As a waiter, how do we predict the tips we will receive for service rendered?**
- **Let's say, we didn't forget to record the bill amount.**

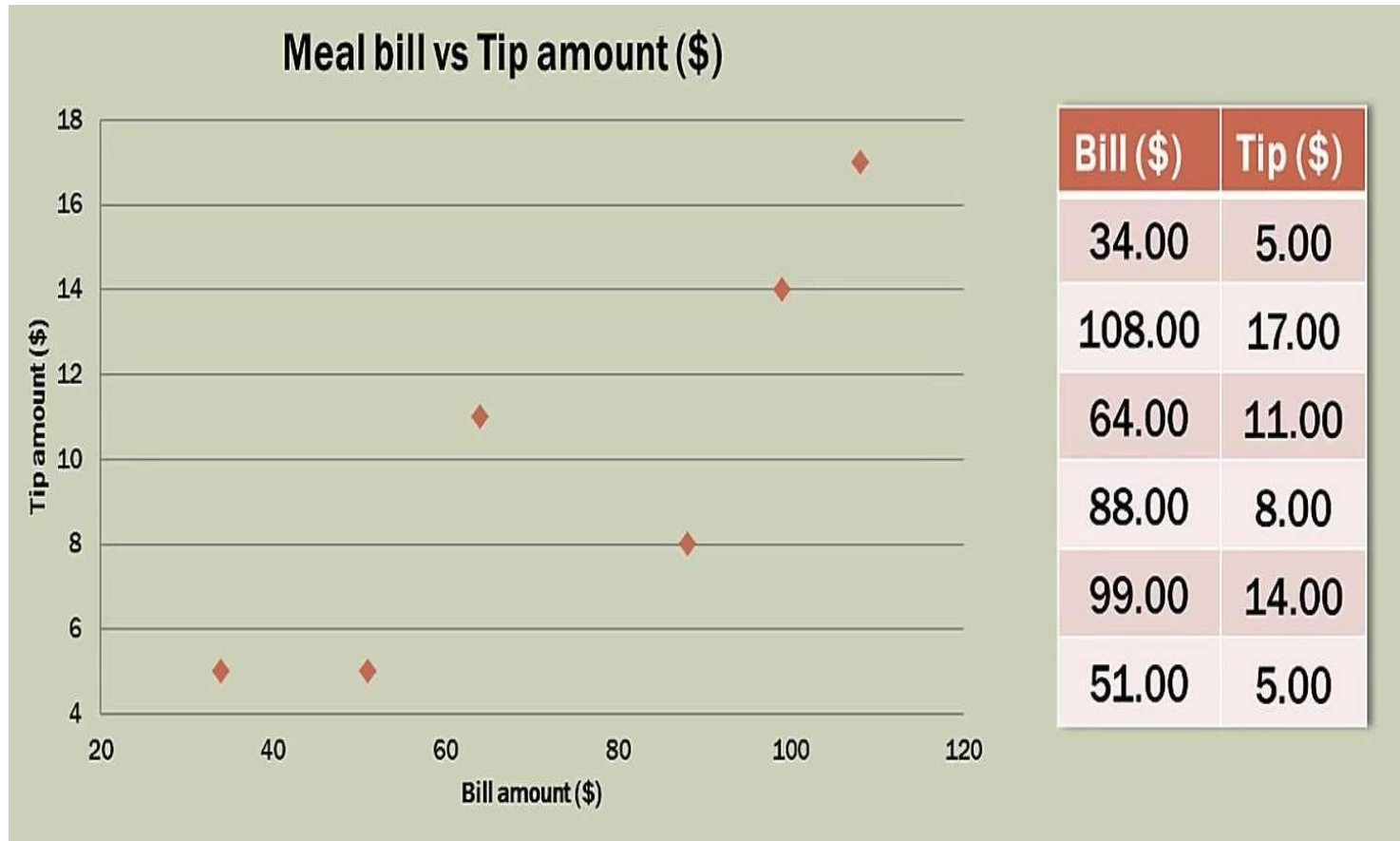


Independent Variable (x)

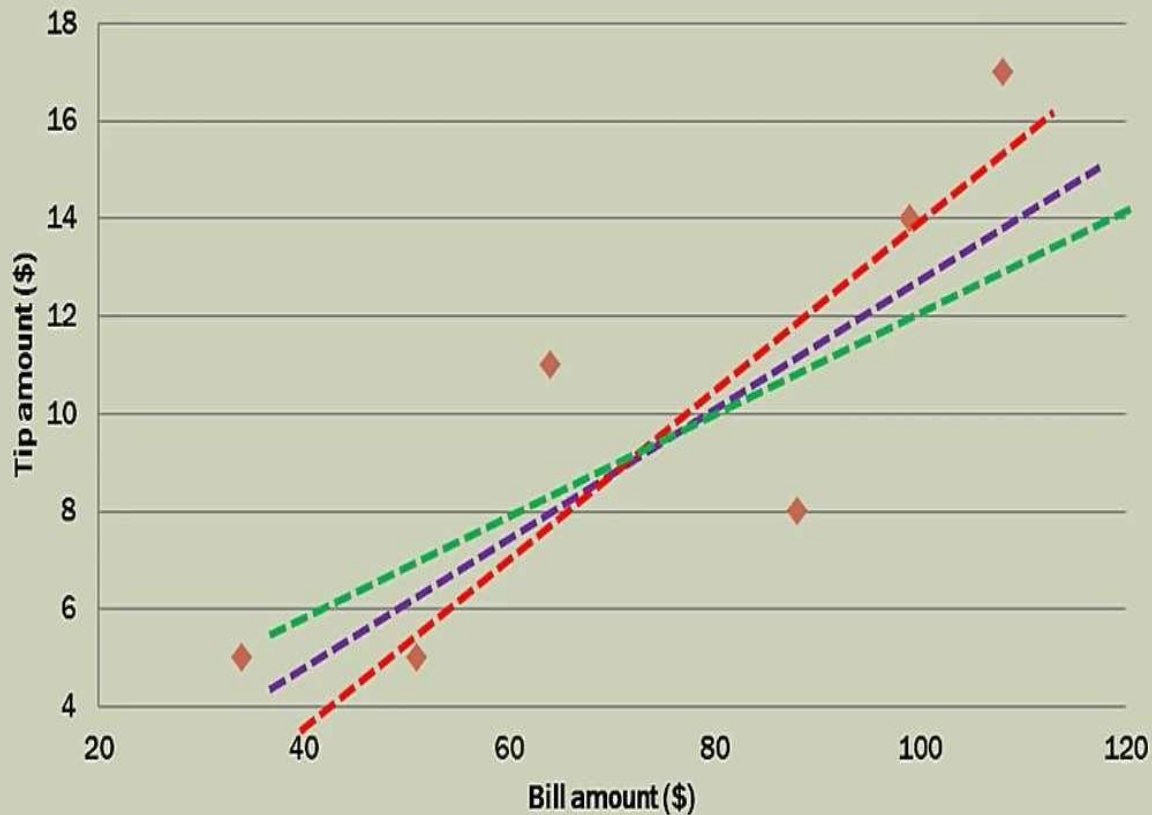
Dependent Variable (y)

Total bill (\$)	Tip amount (\$)
34.00	5.00
108.00	17.00
64.00	11.00
88.00	8.00
99.00	14.00
51.00	5.00

If we scale the graph according to the data points available, we can then plot the points.



## Meal bill vs Tip amount (\$)

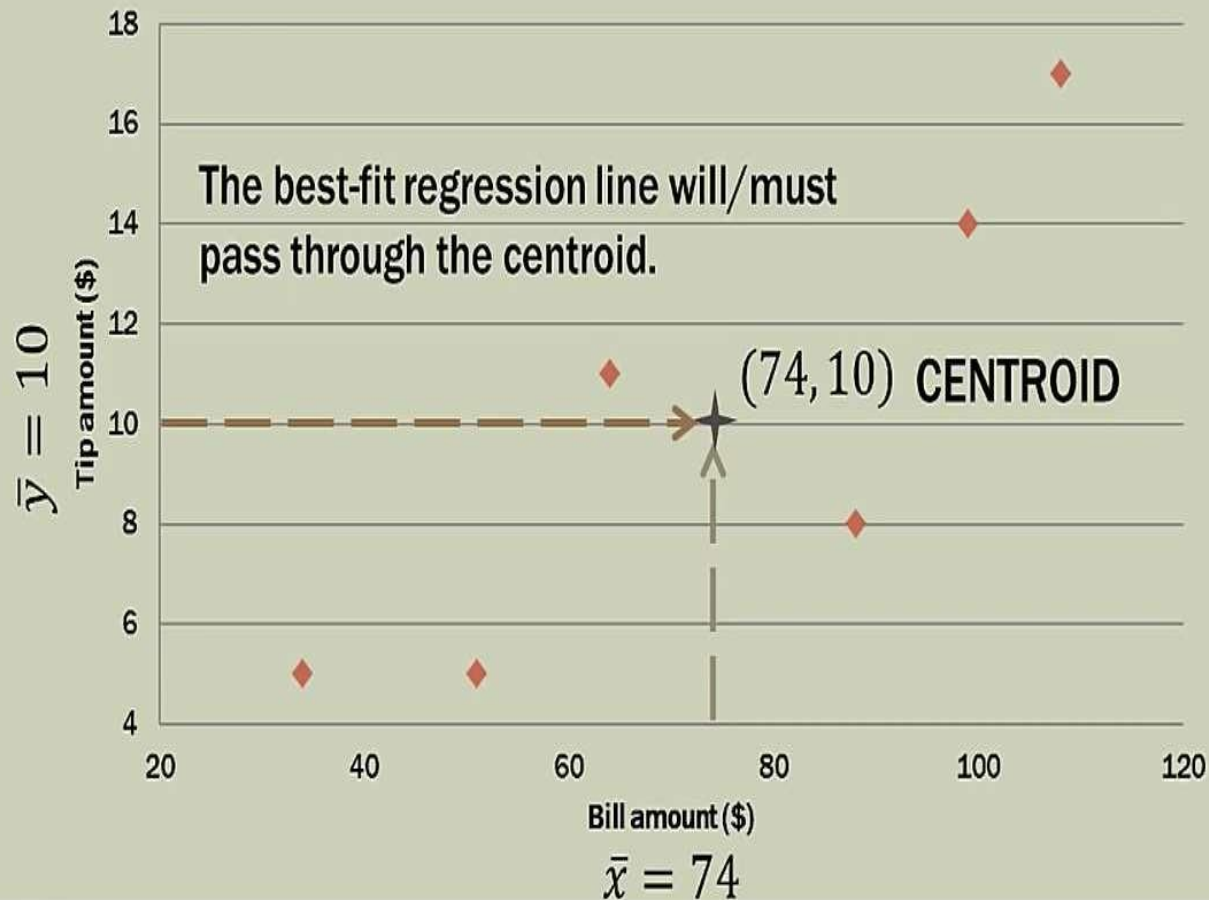


Does the data seem to fall along a line?

*In this case,  
YES! Proceed.*

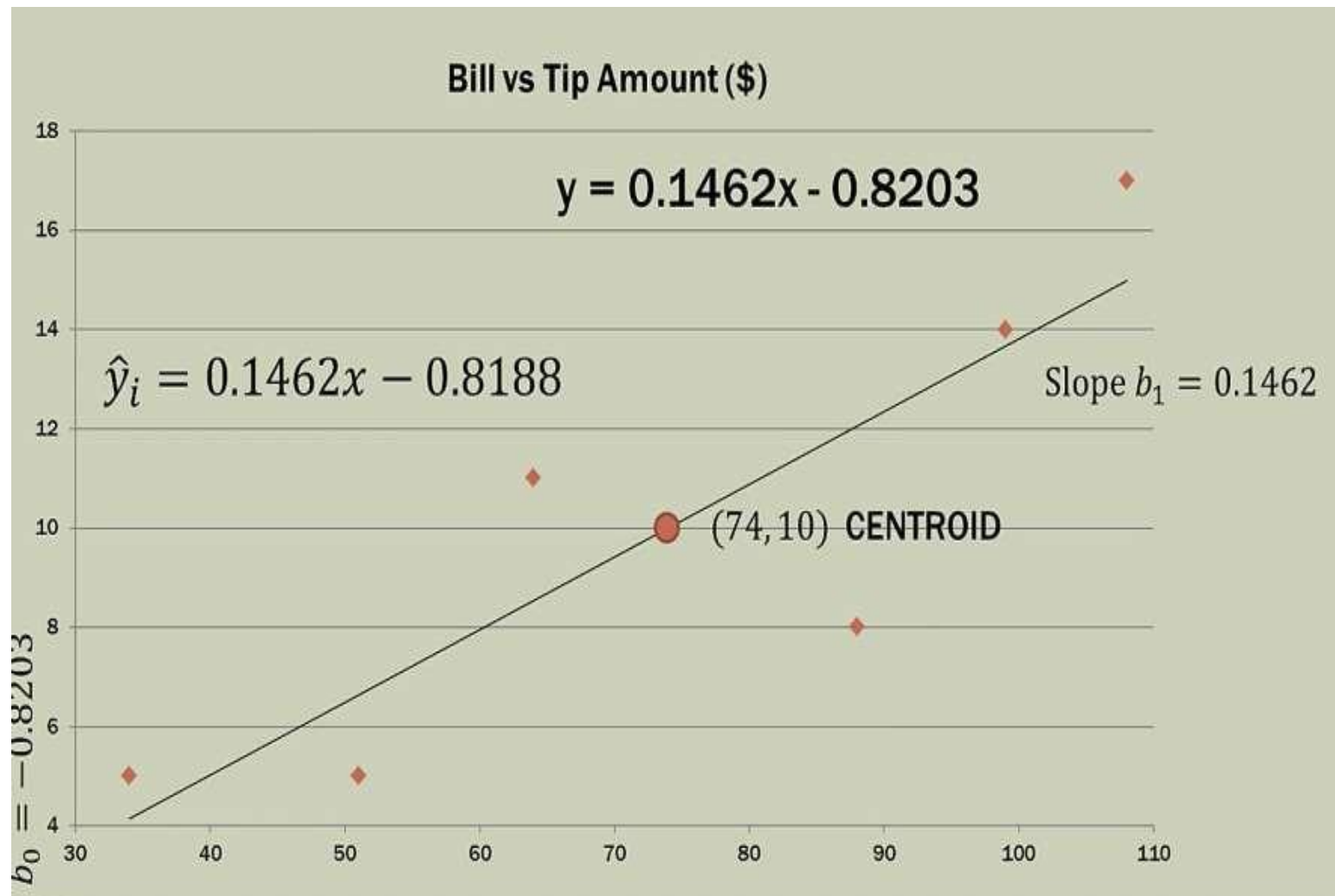
If not...if it's a BLOB with no linear pattern, then stop.

## Meal bill vs Tip amount (\$)



Bill (\$)	Tip (\$)
34.00	5.00
108.00	17.00
64.00	11.00
88.00	8.00
99.00	14.00
51.00	5.00
$\bar{x} = 74$	$\bar{y} = 10$

- (74,10) is the Centroid.
- We can calculate the linear regression in excel
- For comparison, Excel has calculated the regression equation very close to our manual calculation

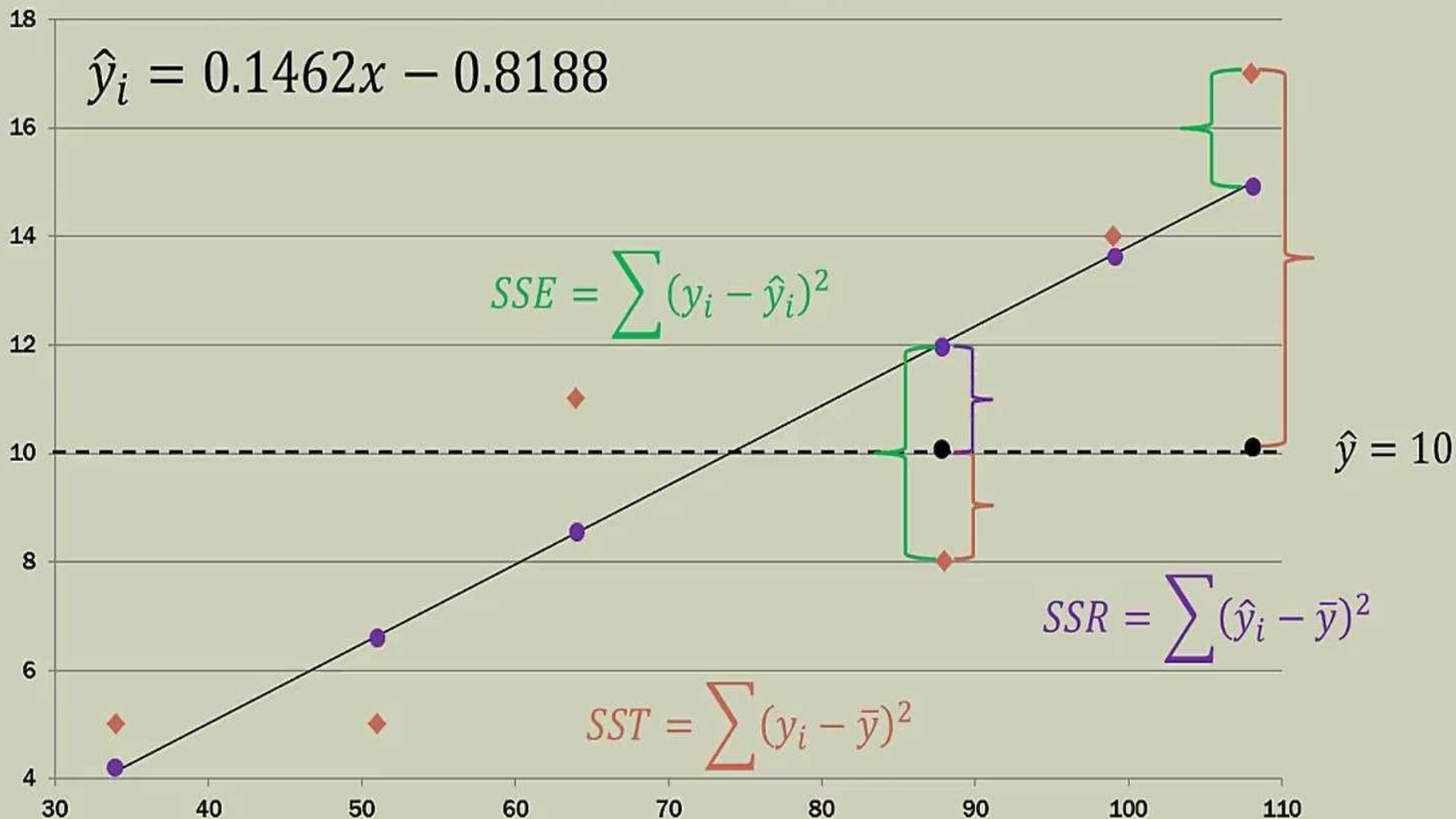


# Error Metrics

$$SST = SSR + SSE$$

Bill vs Tip Amount (\$)

3 Squared Differences





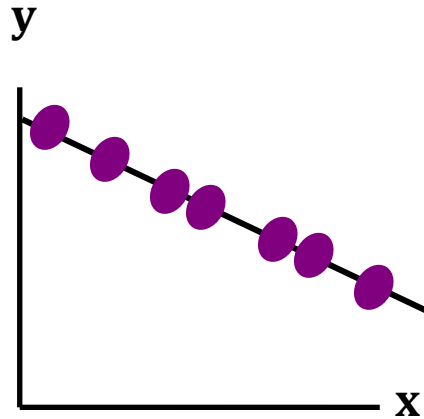
# Coefficient of Determination, $R^2$

- The **coefficient of determination** is the portion of the total variation in the dependent variable that is explained by variation in the independent variable
- The coefficient of determination is also called **R-squared** and is denoted as  $R^2$

$$R^2 = \frac{SSR}{SST}$$

where  $0 \leq R^2 \leq 1$

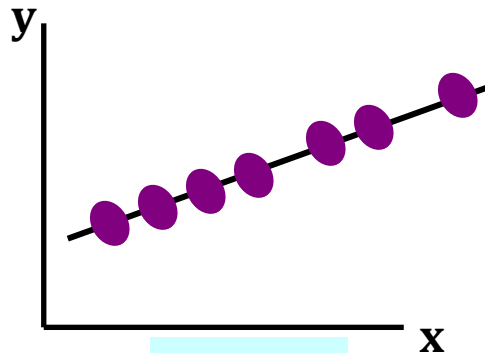
# Examples of Approximate $R^2$ Values



$$R^2 = 1$$

$$R^2 = 1$$

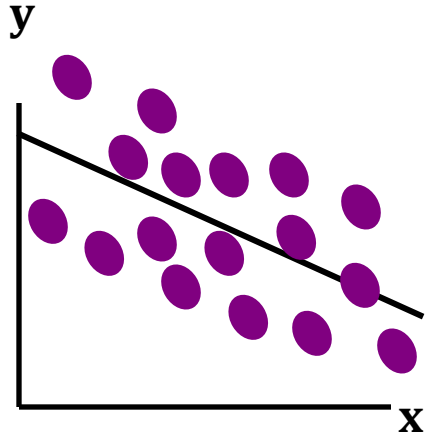
Perfect linear  
relationship between x  
and y:



$$R^2 = +1$$

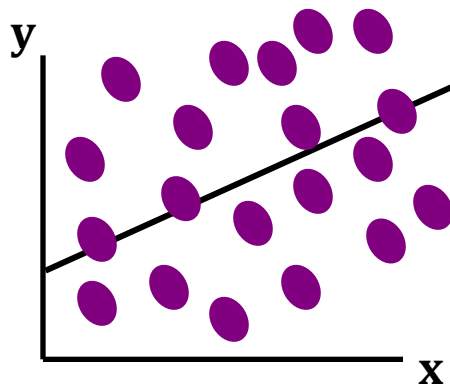
100% of the variation in  
y is explained by  
variation in x

# Examples of Approximate $R^2$ Values



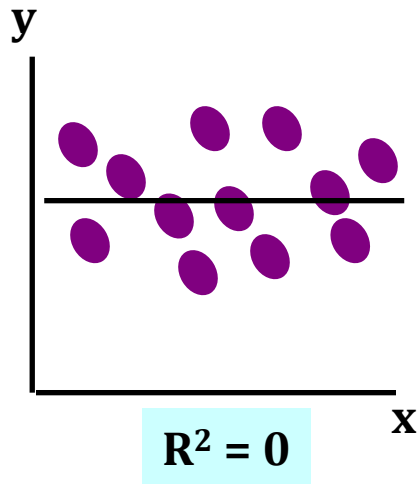
$$0 < R^2 < 1$$

**Weaker linear  
relationship between x  
and y:**



**Some but not all of the  
variation in y is  
explained by variation in  
x**

# Examples of Approximate $R^2$ Values

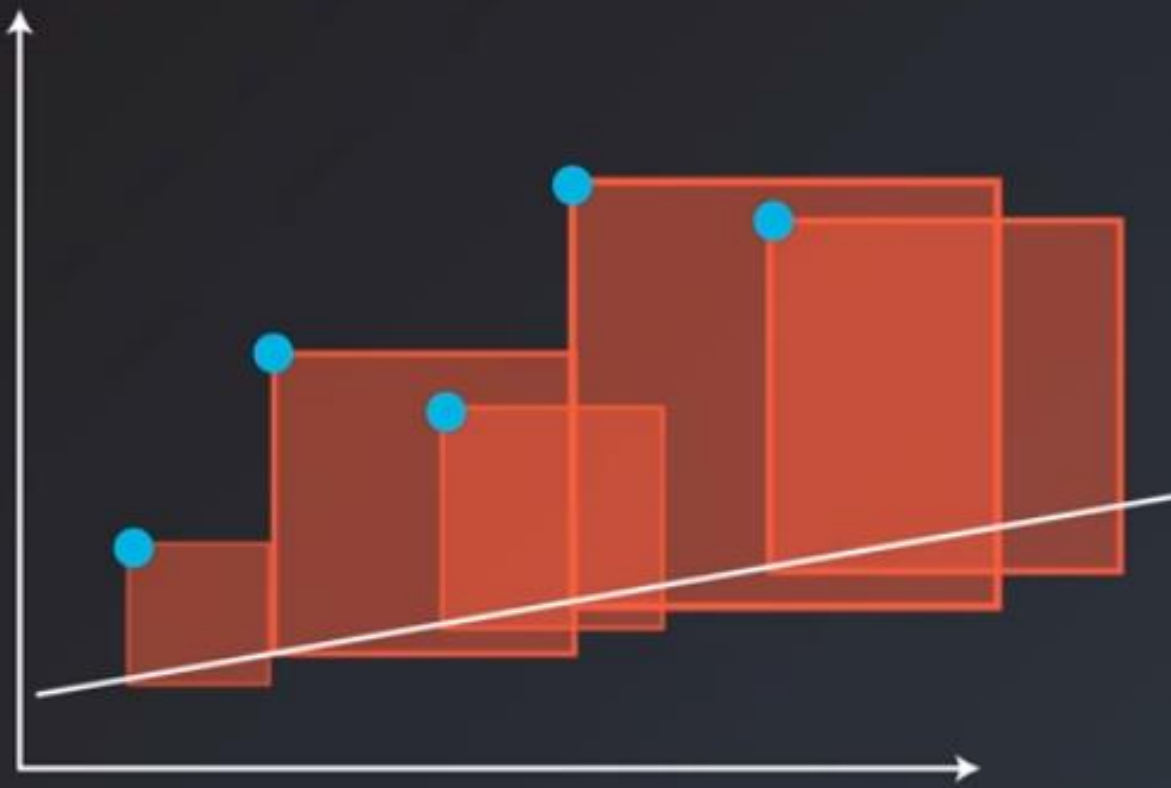


$$R^2 = 0$$

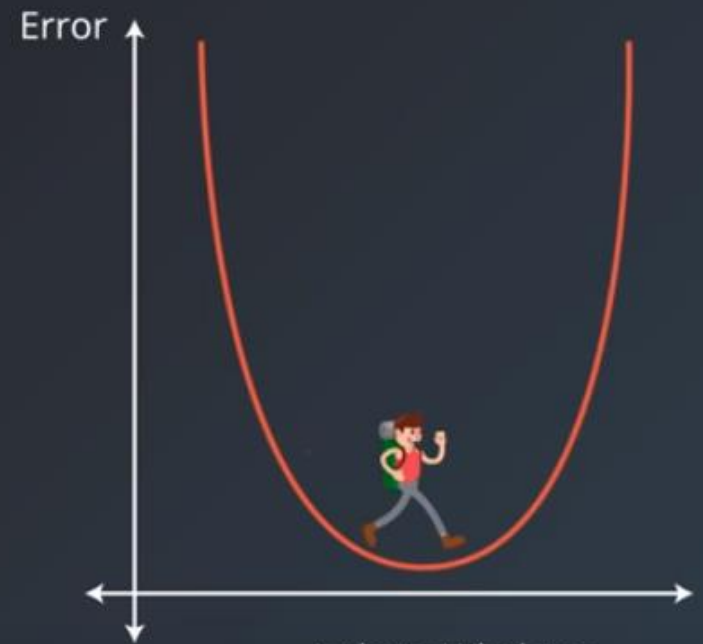
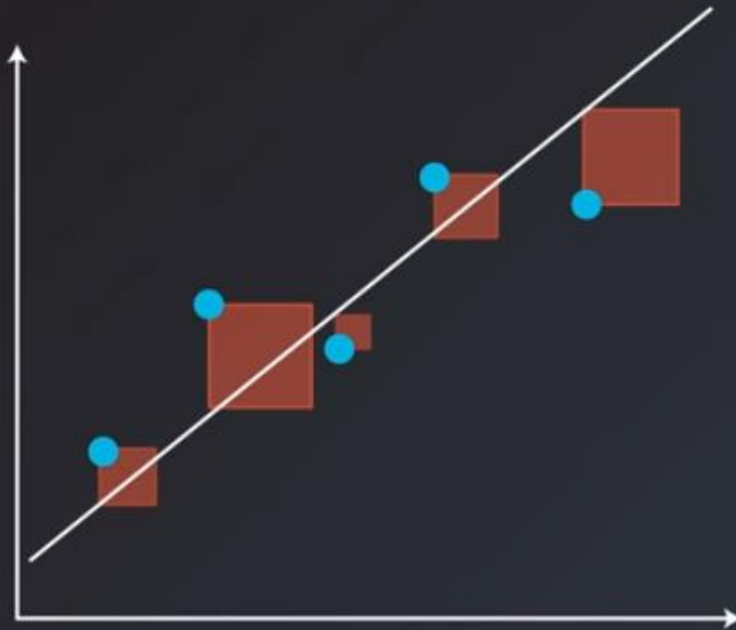
**No linear relationship  
between x and y:**

**The value of Y does not  
depend on x. (None of the  
variation in y is explained  
by variation in x)**

# Mean Squared Error



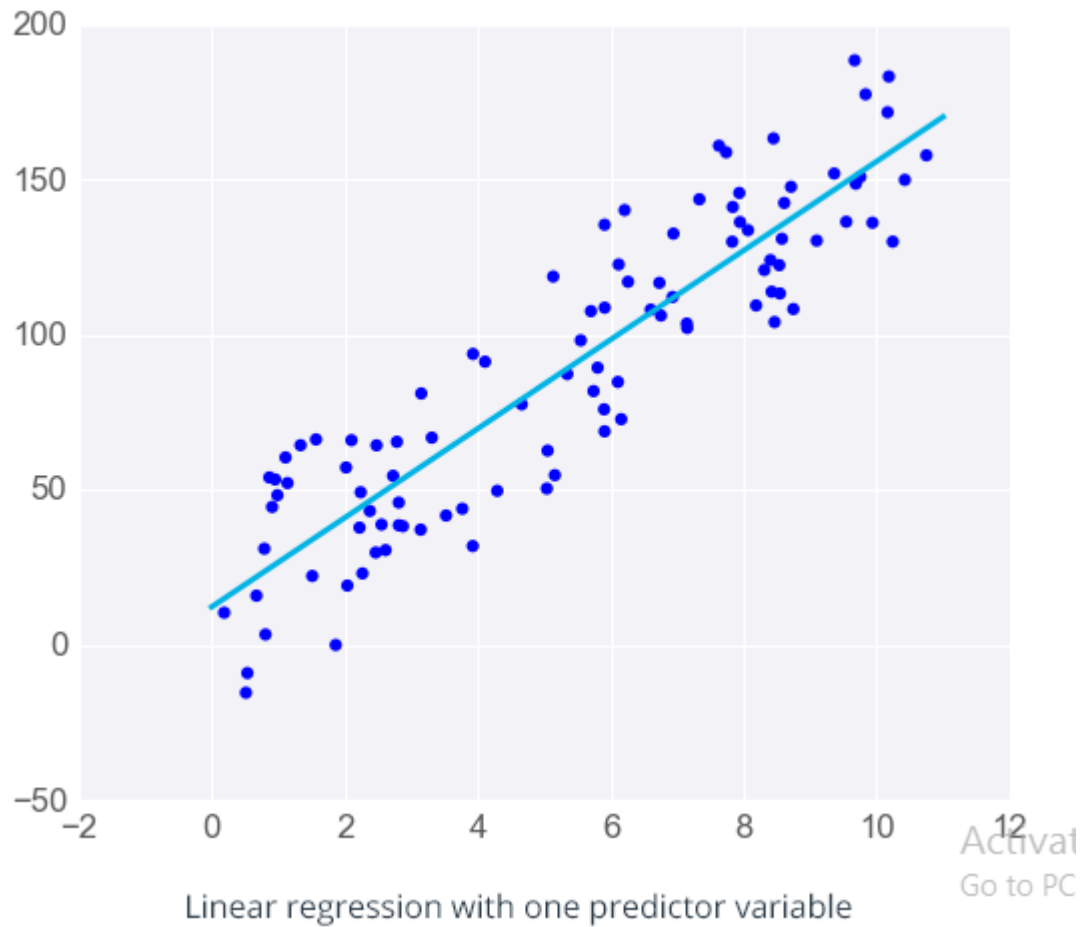
## Mean Squared Error



Activate Windows

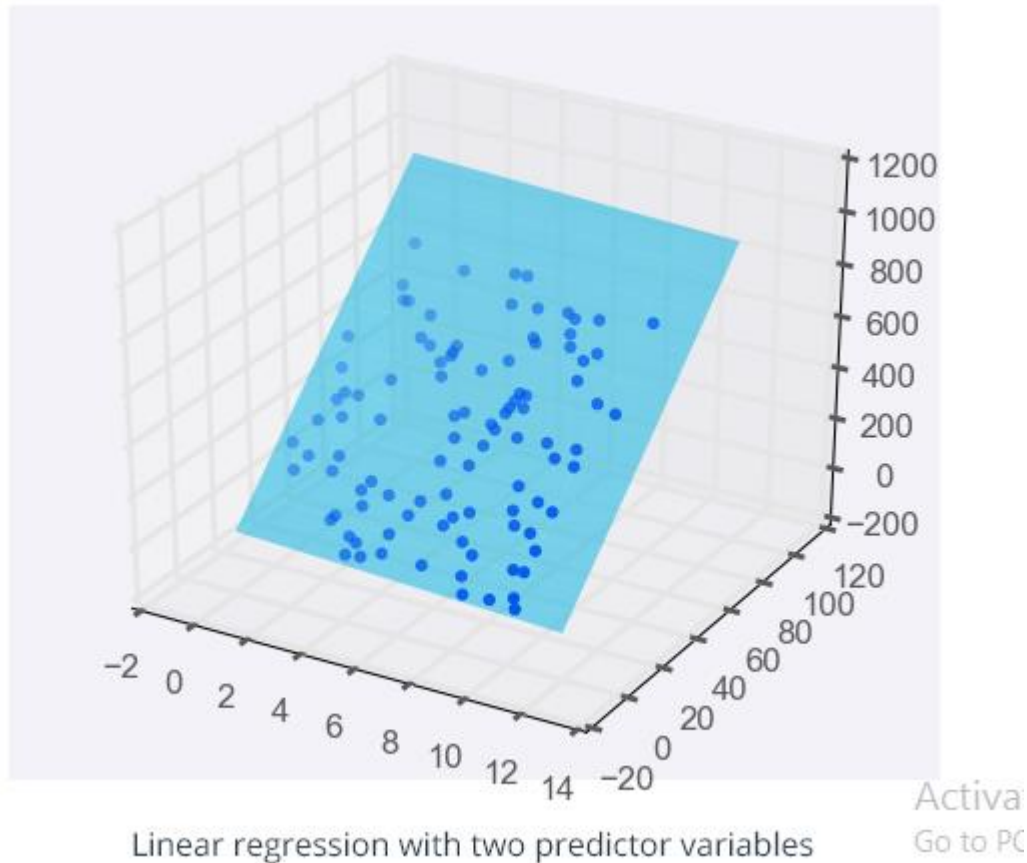
# **Visualization in N dimensions**

# Linear Regression – 1 Variable





# Linear Regression – 2 Variable



**When to use Linear Regression ?**

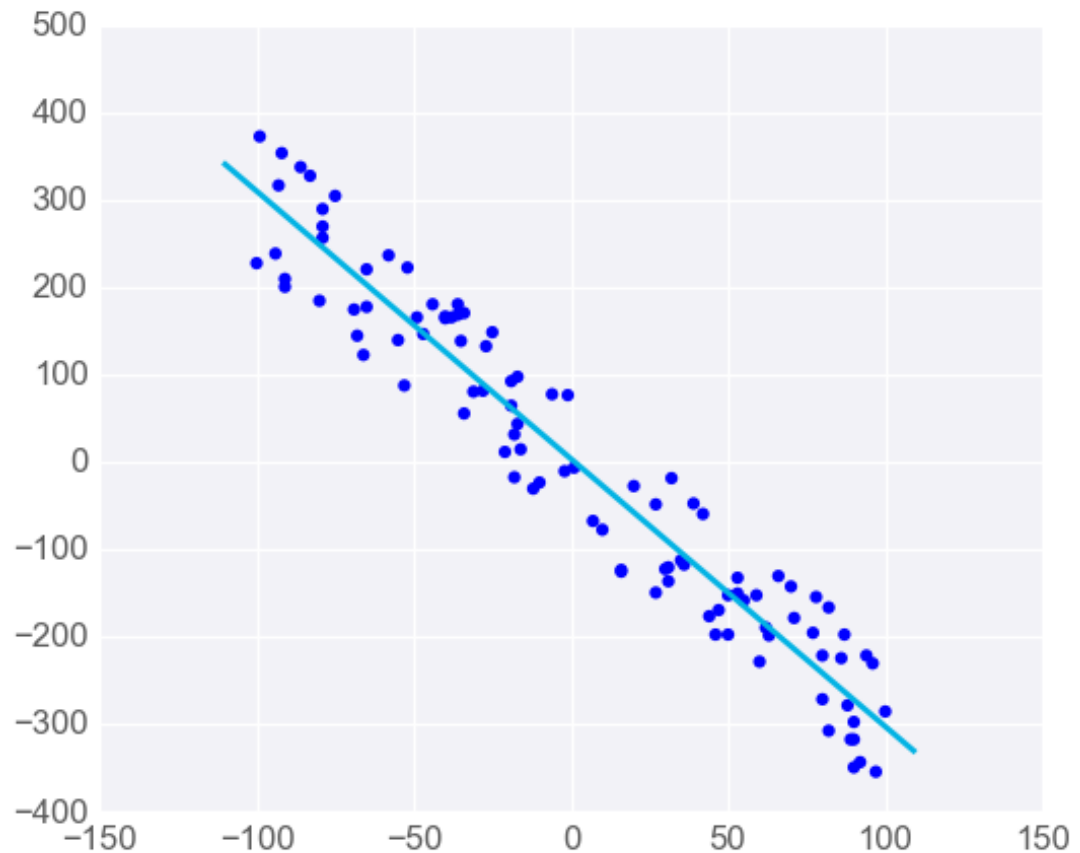
# Linear Regression Warnings

Linear Regression Works Best When the Data is Linear



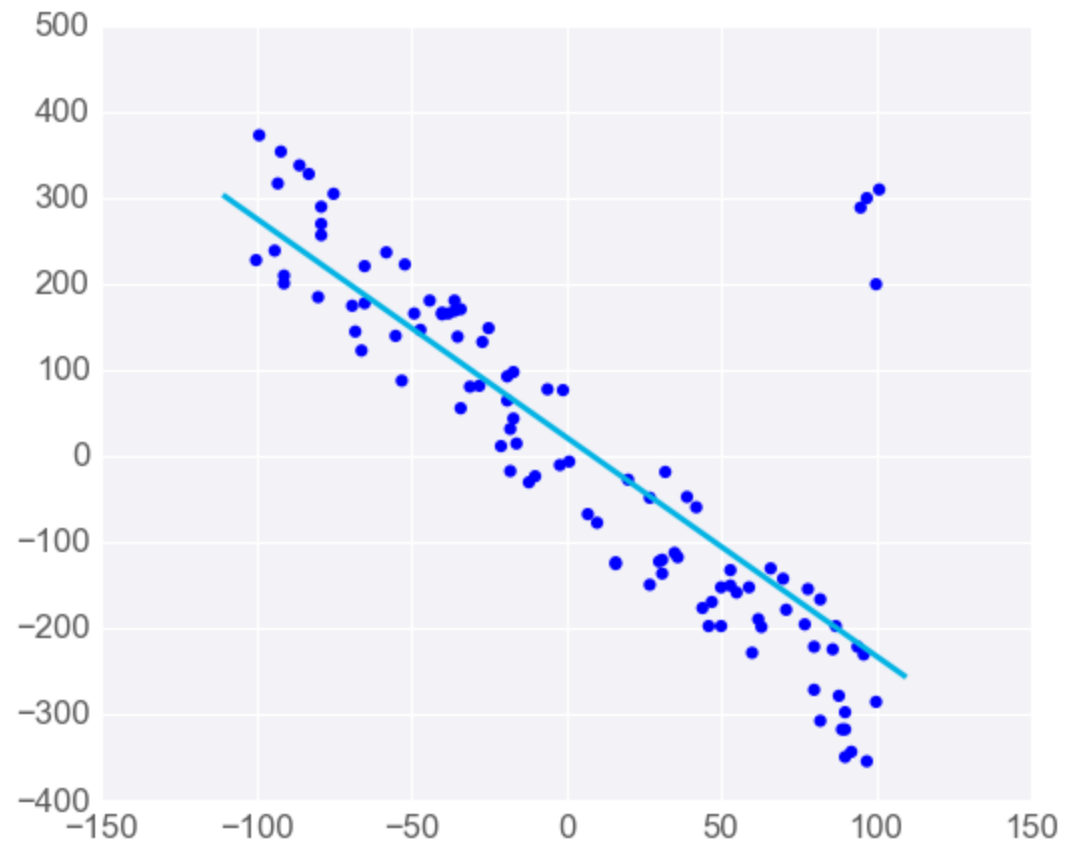
## Linear Regression Warnings

## Linear Regression is Sensitive to Outliers



# Linear Regression Warnings

## Linear Regression is Sensitive to Outliers



# Linear Regression

- Extended in case of Non Linearity

## **Polynomial Regression**

# Polynomial Regression



# Polynomial Regression

