

# Machine Learning for Robotics

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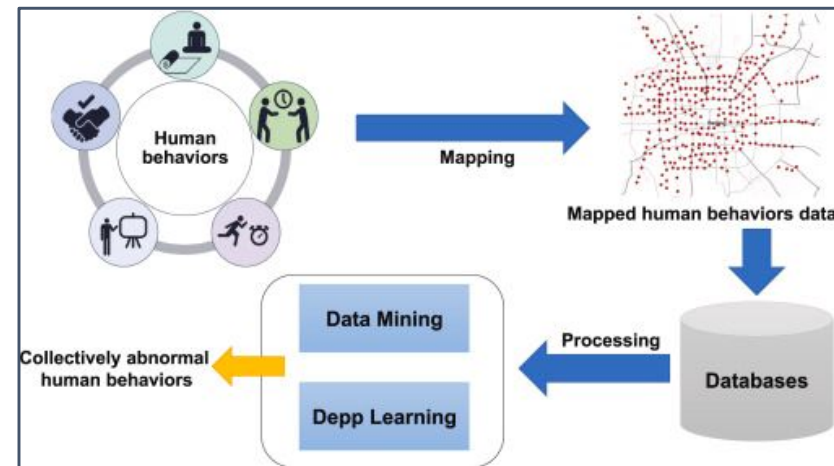
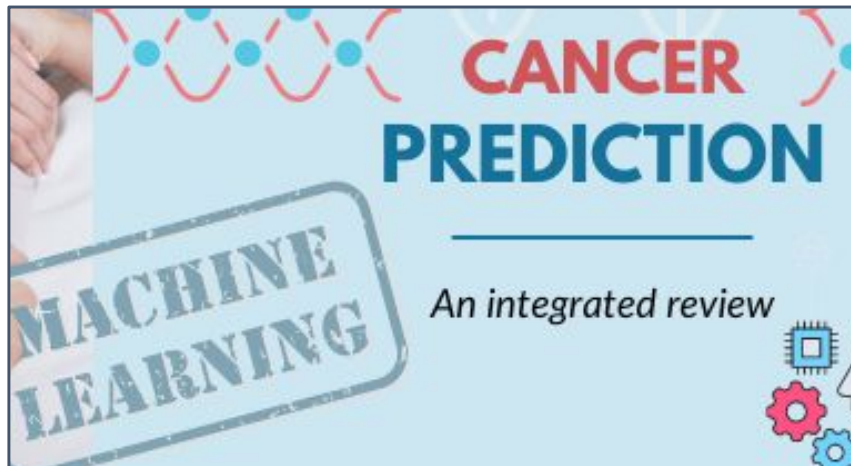
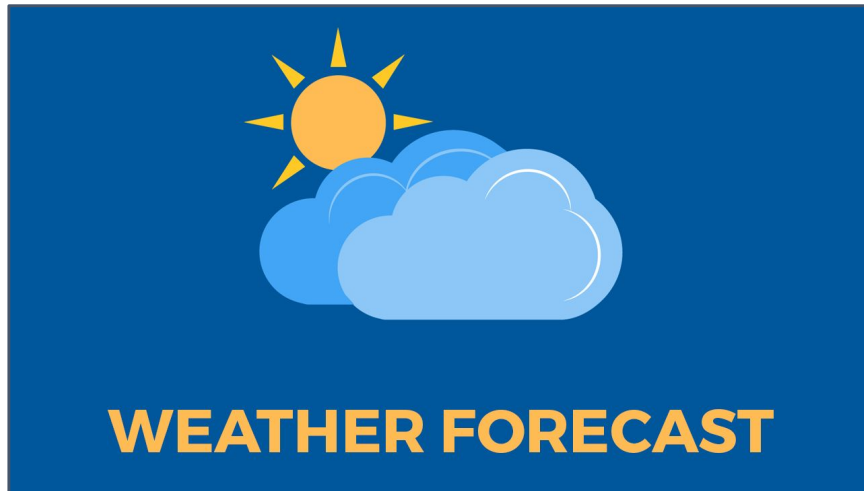
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# **Class 01 - Introduction to Machine Learning**

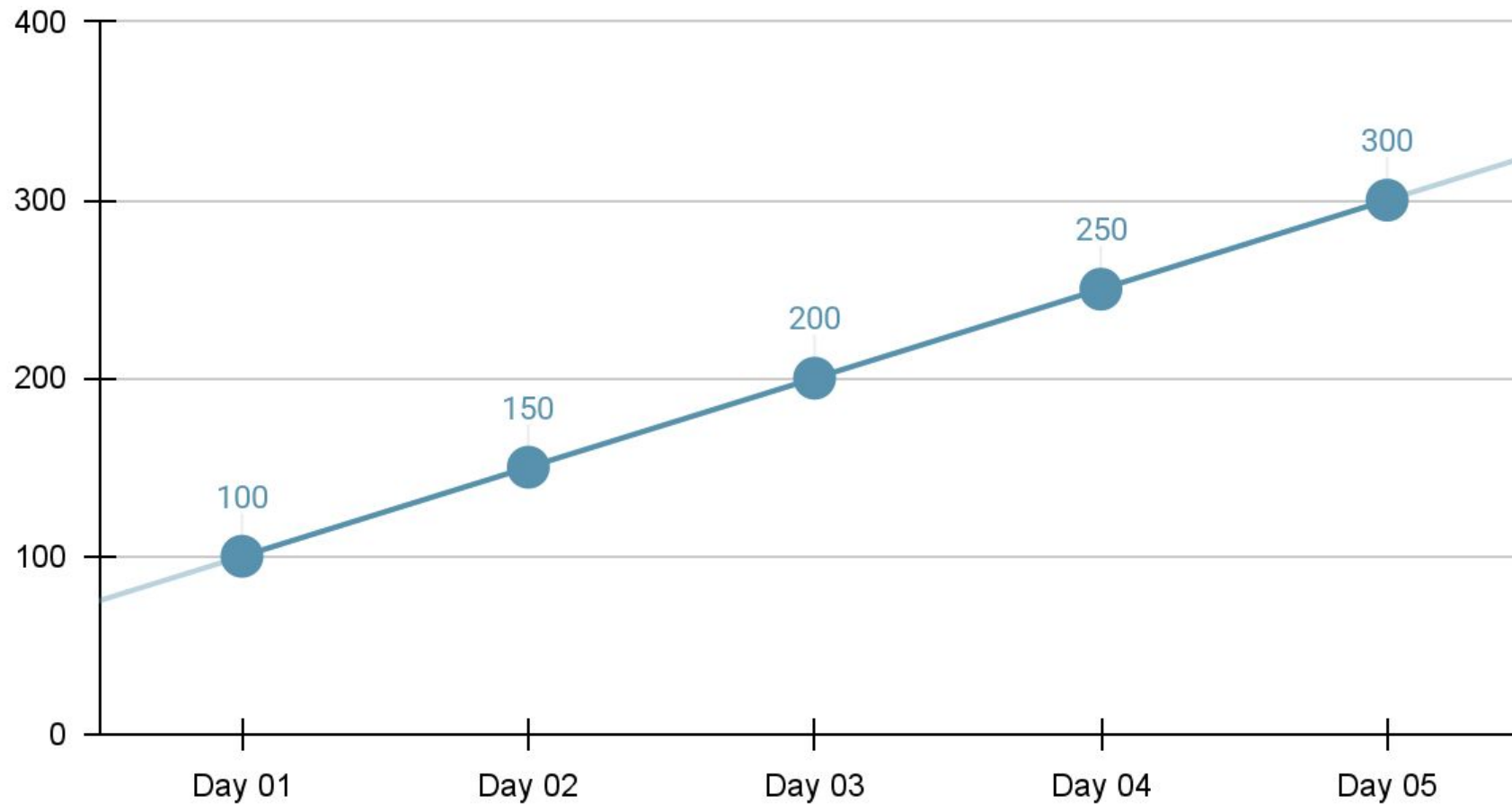
# | Machine Learning

Machine Learning is all about **prediction**.



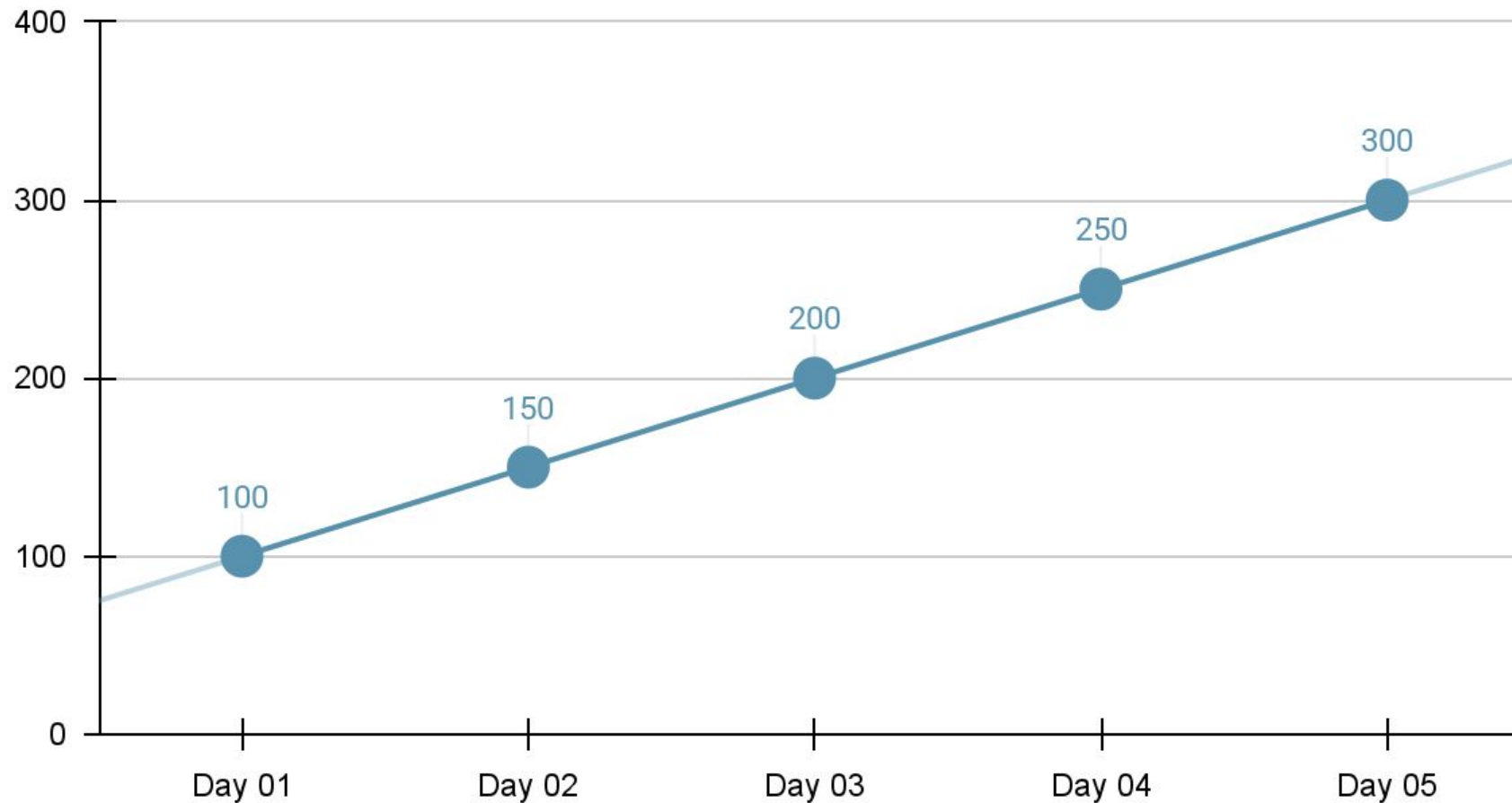
# Machine Learning

Price of Per KG Apple



# Machine Learning

Price of Per KG Apple

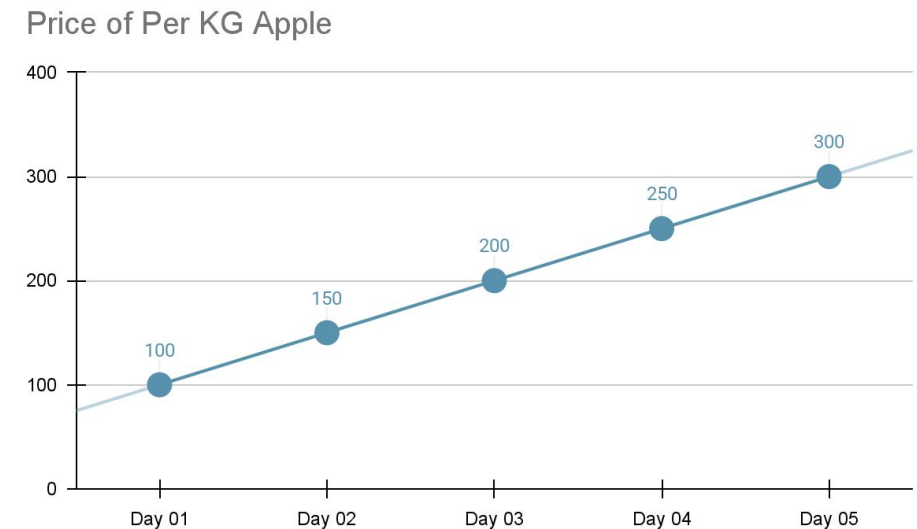


**What will be the price of Apple at Day 06?**

# Machine Learning

To calculate the price of Day - 6,

- we can use  $y = m.x + b$  formula (ideal equation of a straight line)
- we know that  $x = 6$  (day) and we need to calculate  $y$  (price)
- but, first we need to figure out the value of  $m$  (slope) and  $b$  (intercept)



What will be the price of Apple at Day 06?

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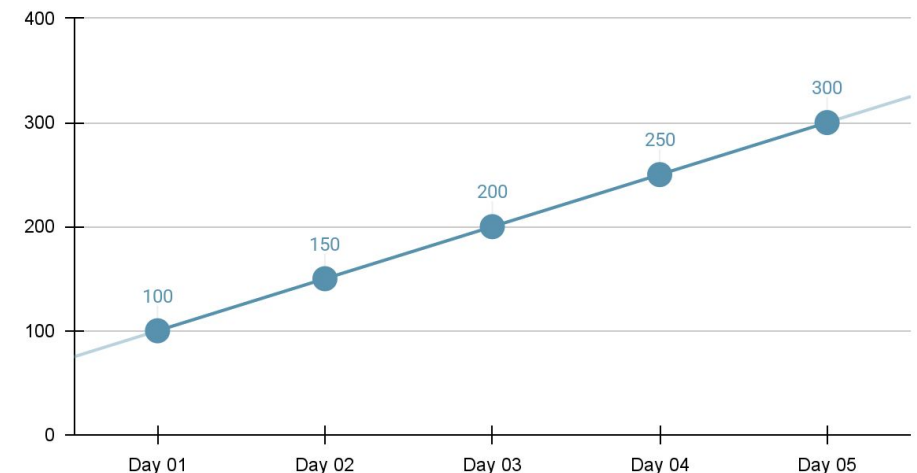
To calculate the value of  $m$ ,  $c$ :

- We know if two points  $(x_1, y_1)$  and  $(x_2, y_2)$  are given, then we can use

$$m = (y_2 - y_1) / (x_2 - x_1)$$

$$b = y_1 - m.x_1$$

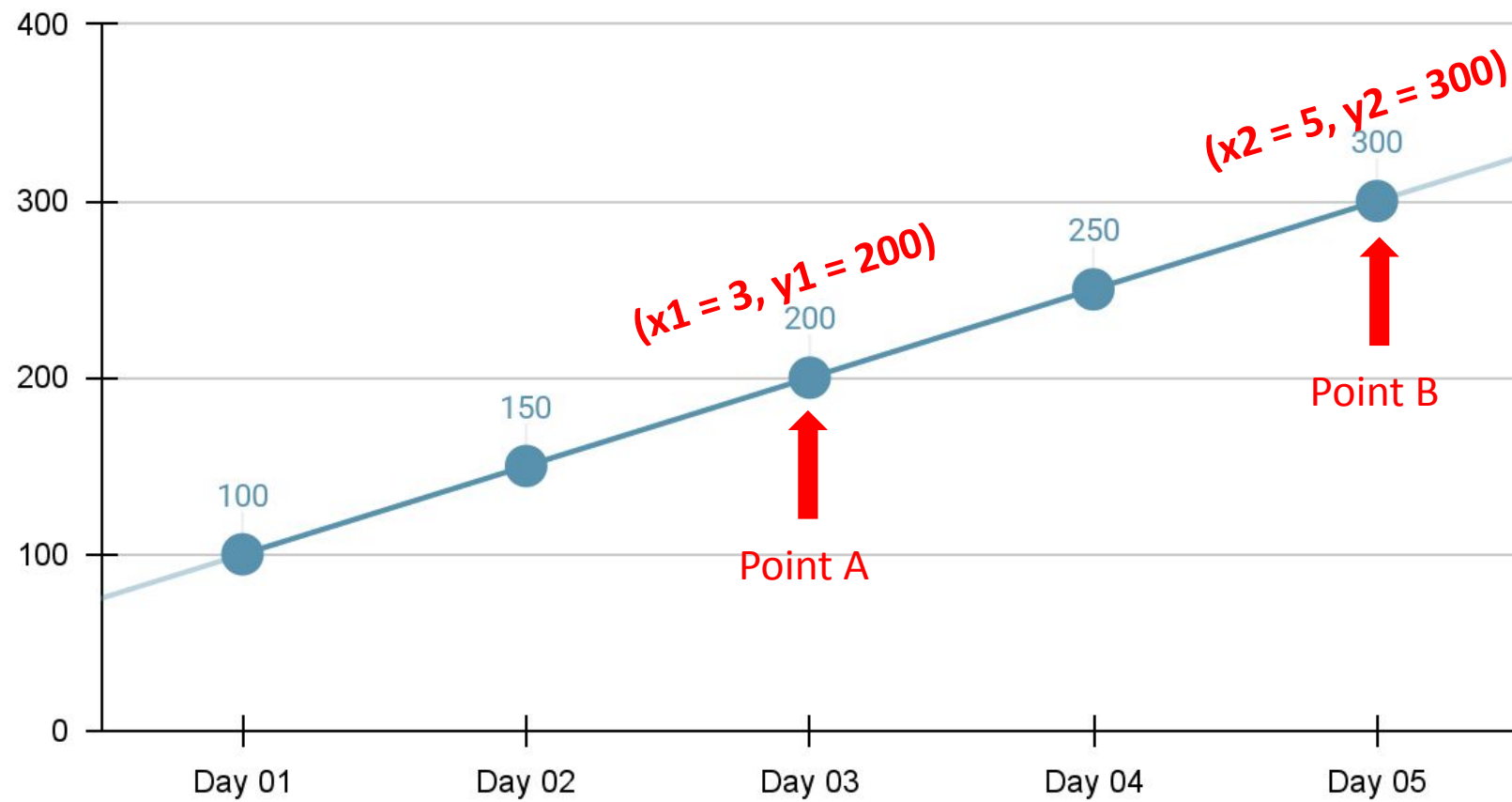
Price of Per KG Apple



What will be the price of Apple at Day 06?

# Machine Learning

Price of Per KG Apple

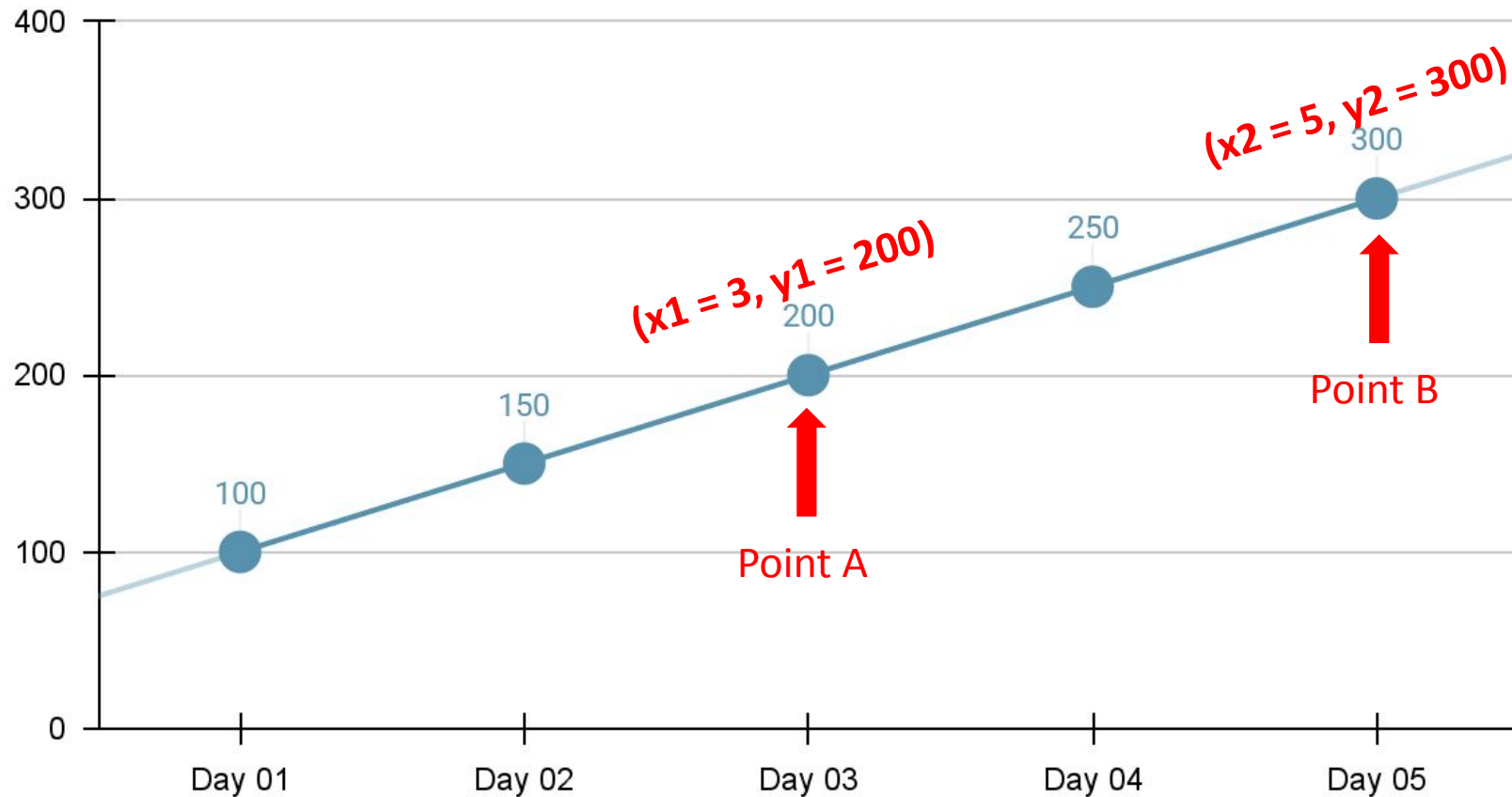




# Machine Learning

$$m = (y_2 - y_1) / (x_2 - x_1)$$
$$m = (300 - 200) / (5 - 2)$$
$$m = 50$$

Price of Per KG Apple

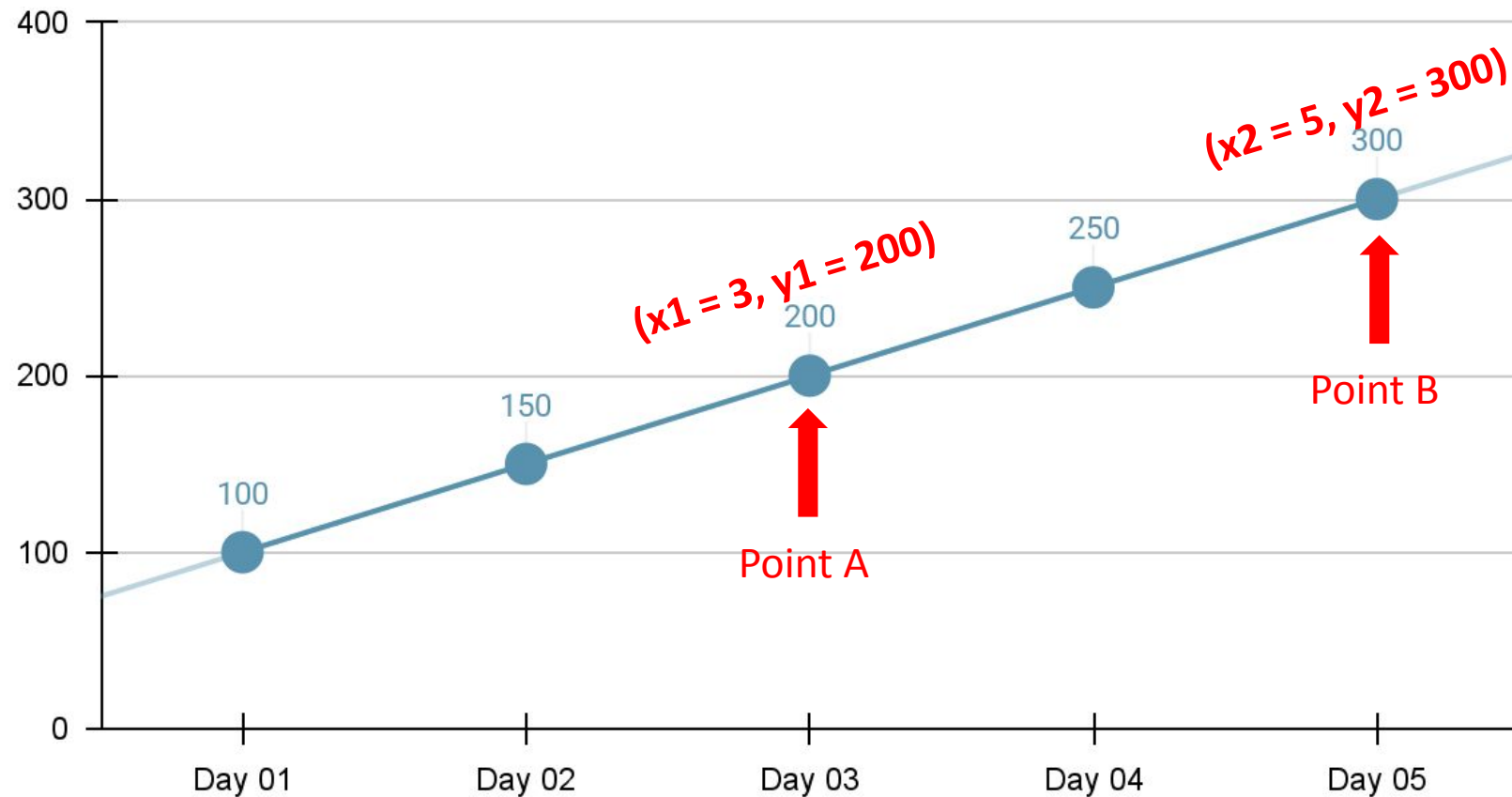


# Machine Learning

$$m = (y_2 - y_1) / (x_2 - x_1)$$
$$m = (300 - 200) / (5 - 2)$$
$$m = 50$$

$$b = y_1 - m.x_1$$
$$b = 200 - 50.3$$
$$b = 150$$

Price of Per KG Apple



# **| Machine Learning**

**So,  $x = 6$  (day),  $m = 50$  (slope),  $b = 50$  (intercept)**

**Our Equation:  $y = m.x + b$**

# | Machine Learning

So,  $x = 6$  (day),  $m = 50$  (slope),  $b = 50$  (intercept)

Our Equation:  $y = m.x + b$

$$y = 50.6 + 50$$

$$y = 350$$

# | Machine Learning

So,  $x = 6$  (day),  $m = 50$  (slope),  $b = 50$  (intercept)

Our Equation:  $y = m.x + b$

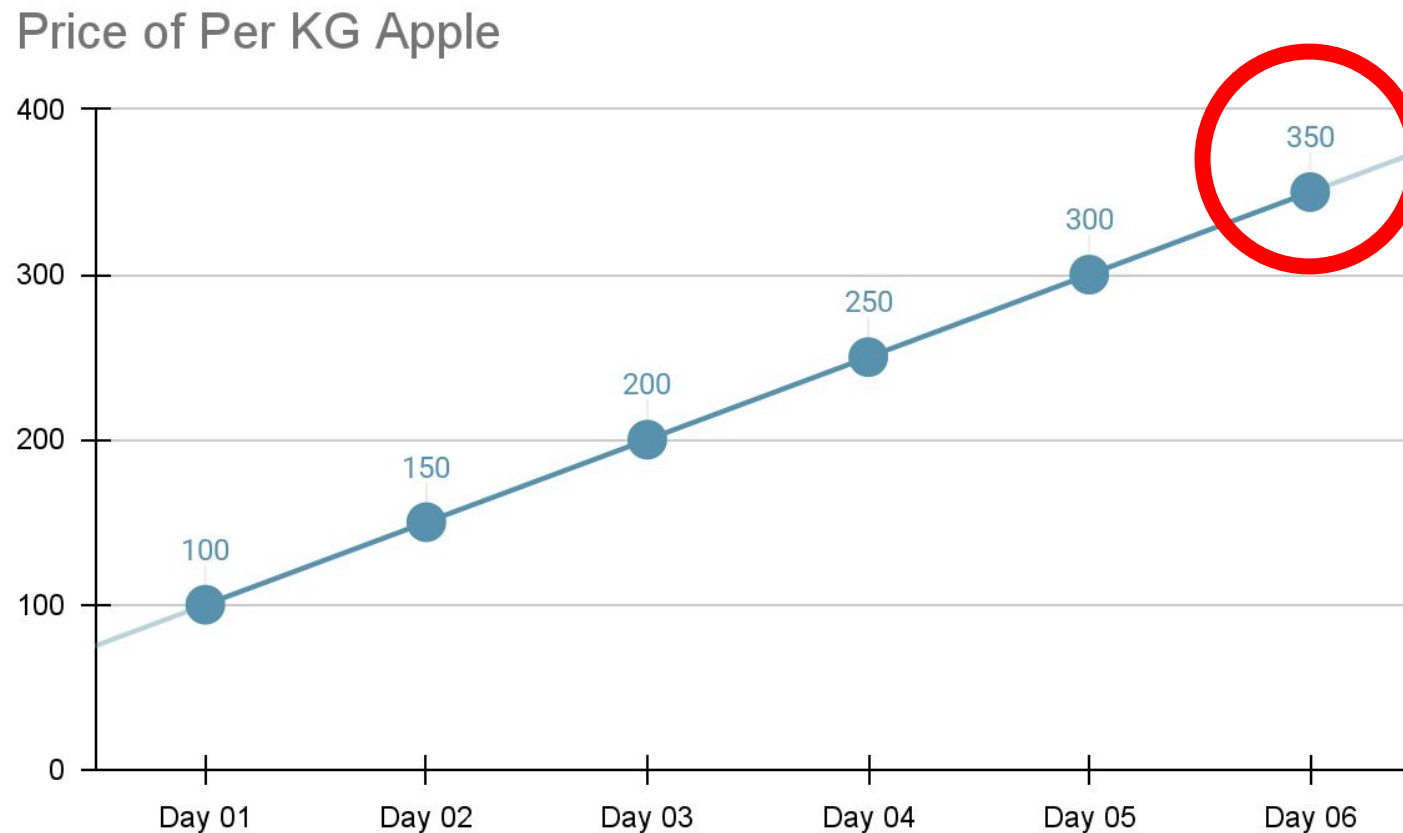
$$y = 50.6 + 50$$

$$y = 350$$

**The price of apple on Day 06 is 350 tk**

# Machine Learning

The price of apple on Day 06 is 350 tk



# | Machine Learning



Now, We can predict the price of Apple for any DAY!!!

# | Machine Learning

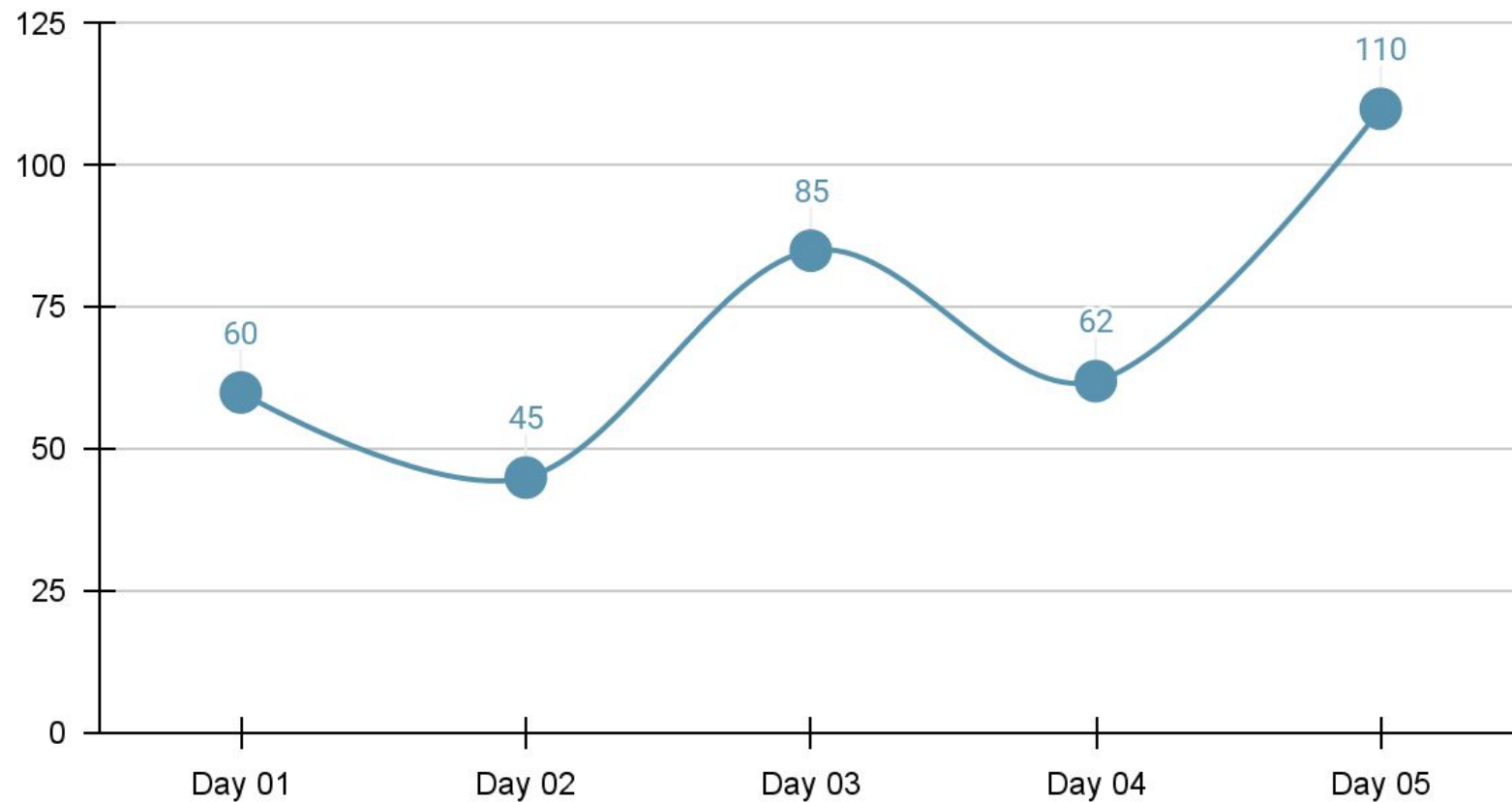
*Let's predict the price of Mango!!!*



# Machine Learning

*Let's predict the price of Mango!!!*

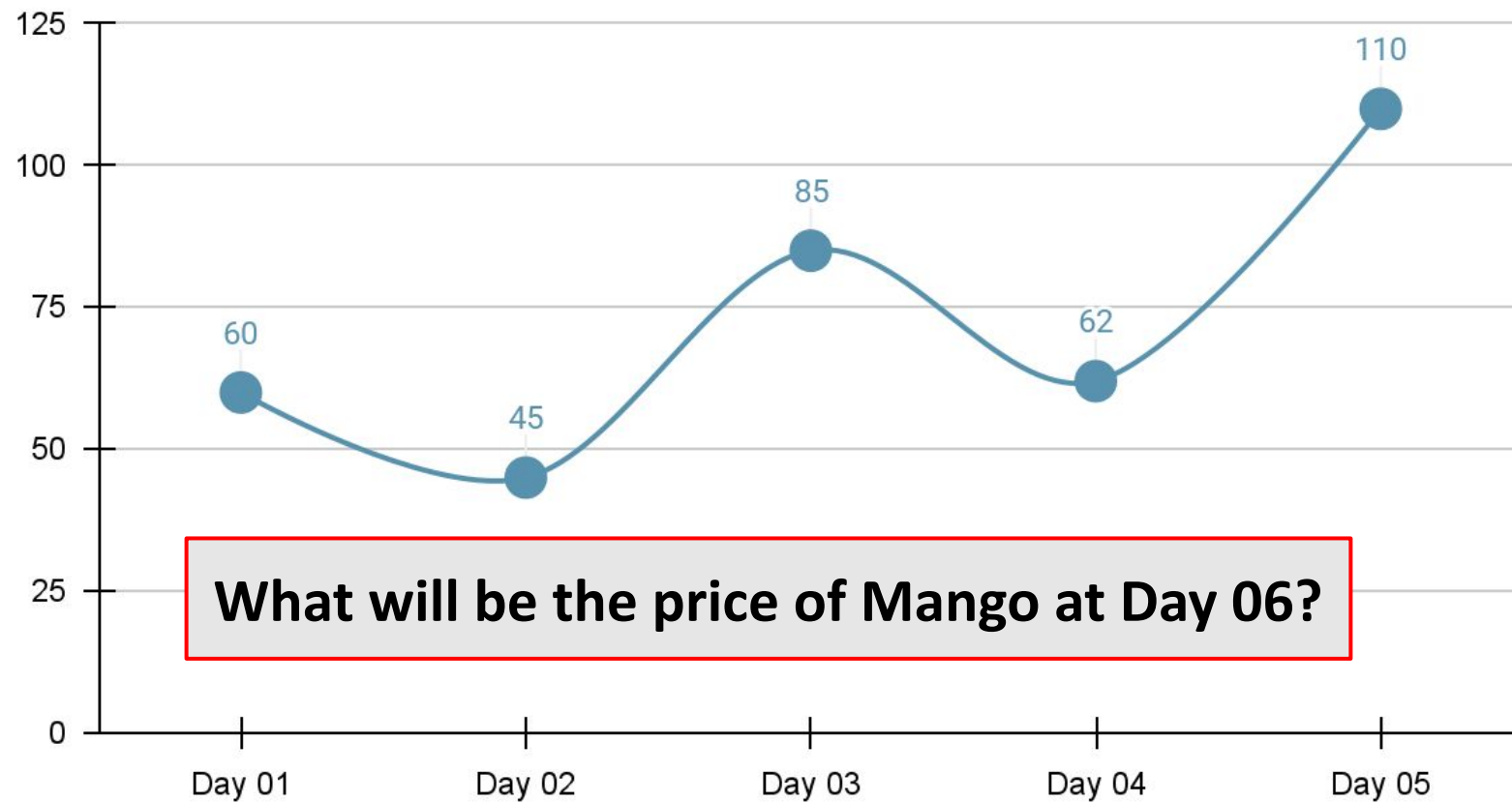
Price of Per KG Mango



# Machine Learning

*Let's predict the price of Mango!!!*

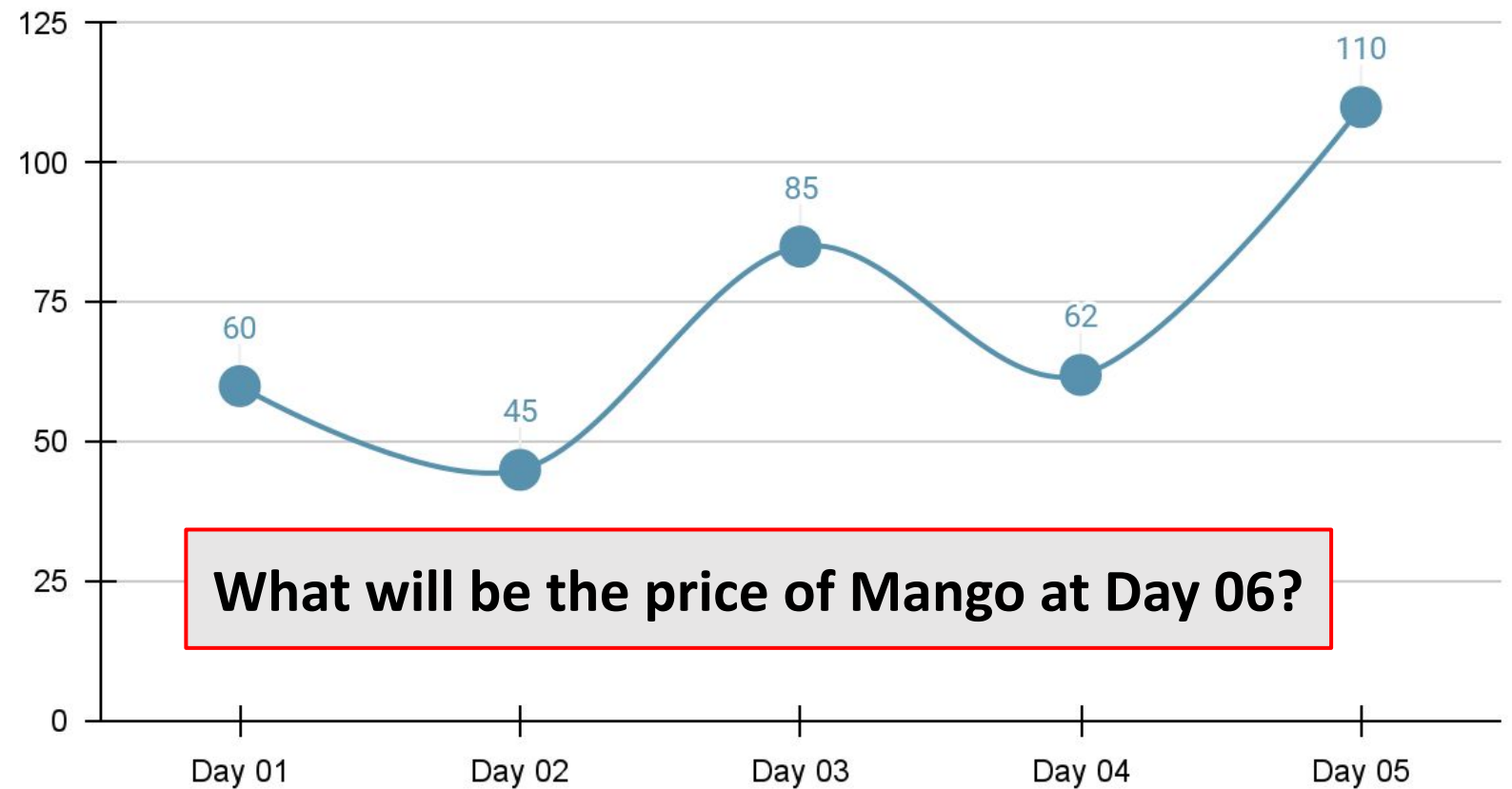
Price of Per KG Mango



# Machine Learning

*Let's analyse the data from graph!*

Price of Per KG Mango

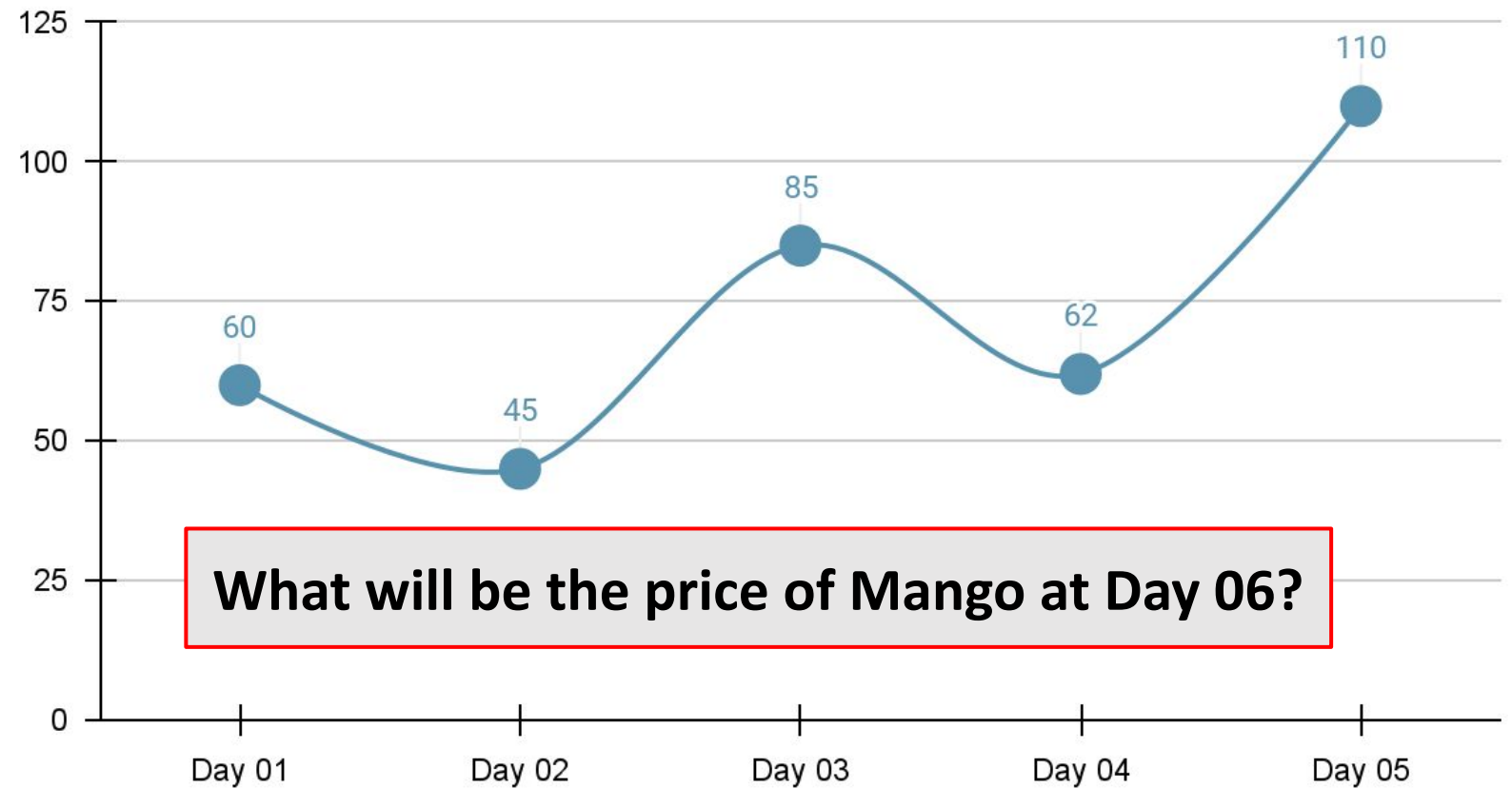


# Machine Learning

*Let's analyse the data from graph!*

- The data points are not on a straight line.

Price of Per KG Mango

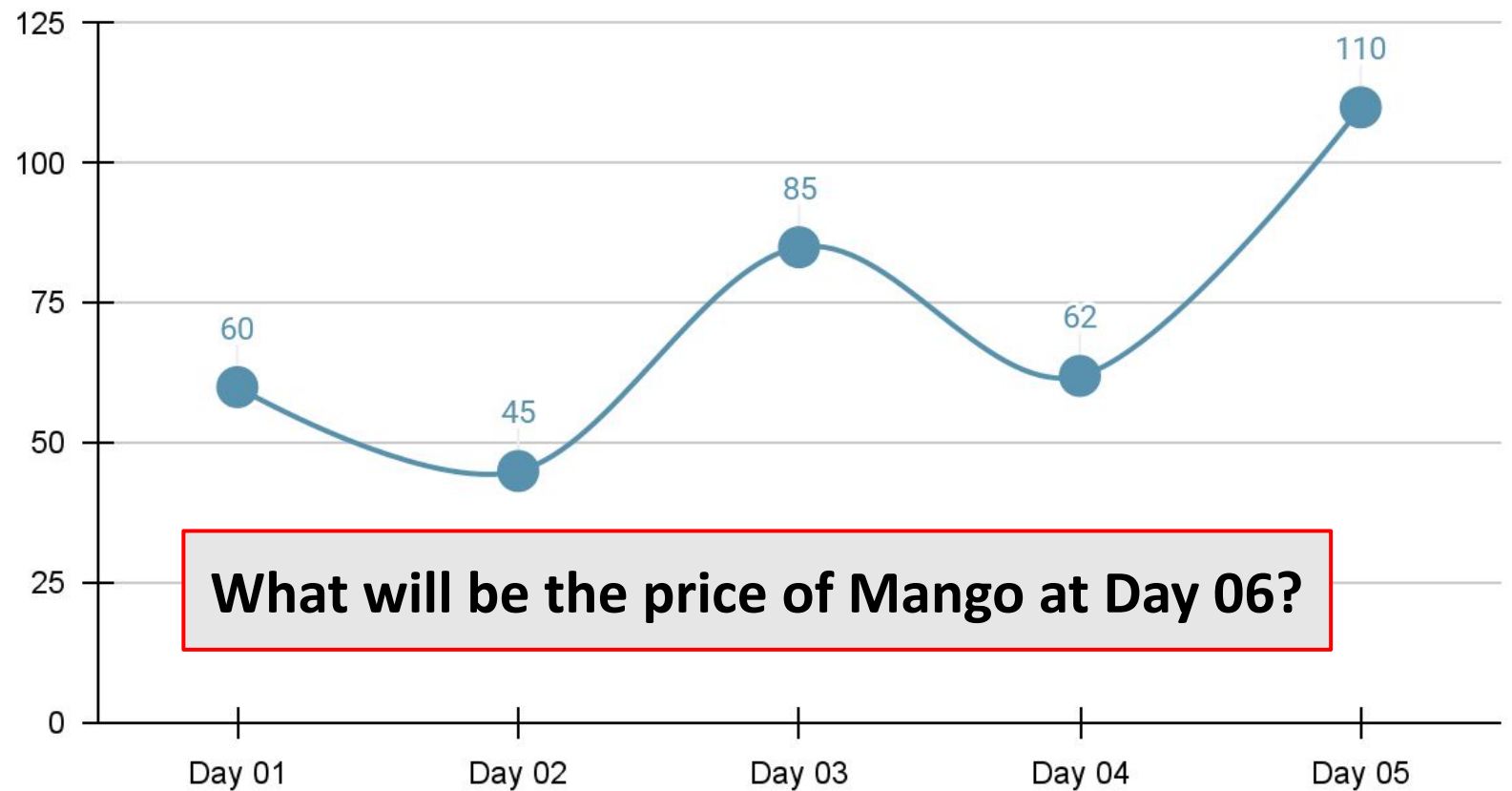


# Machine Learning

*Let's analyse the data from graph!*

- The data points are not on a straight line.
- The data points do not have a straightforward linear relationship.

Price of Per KG Mango

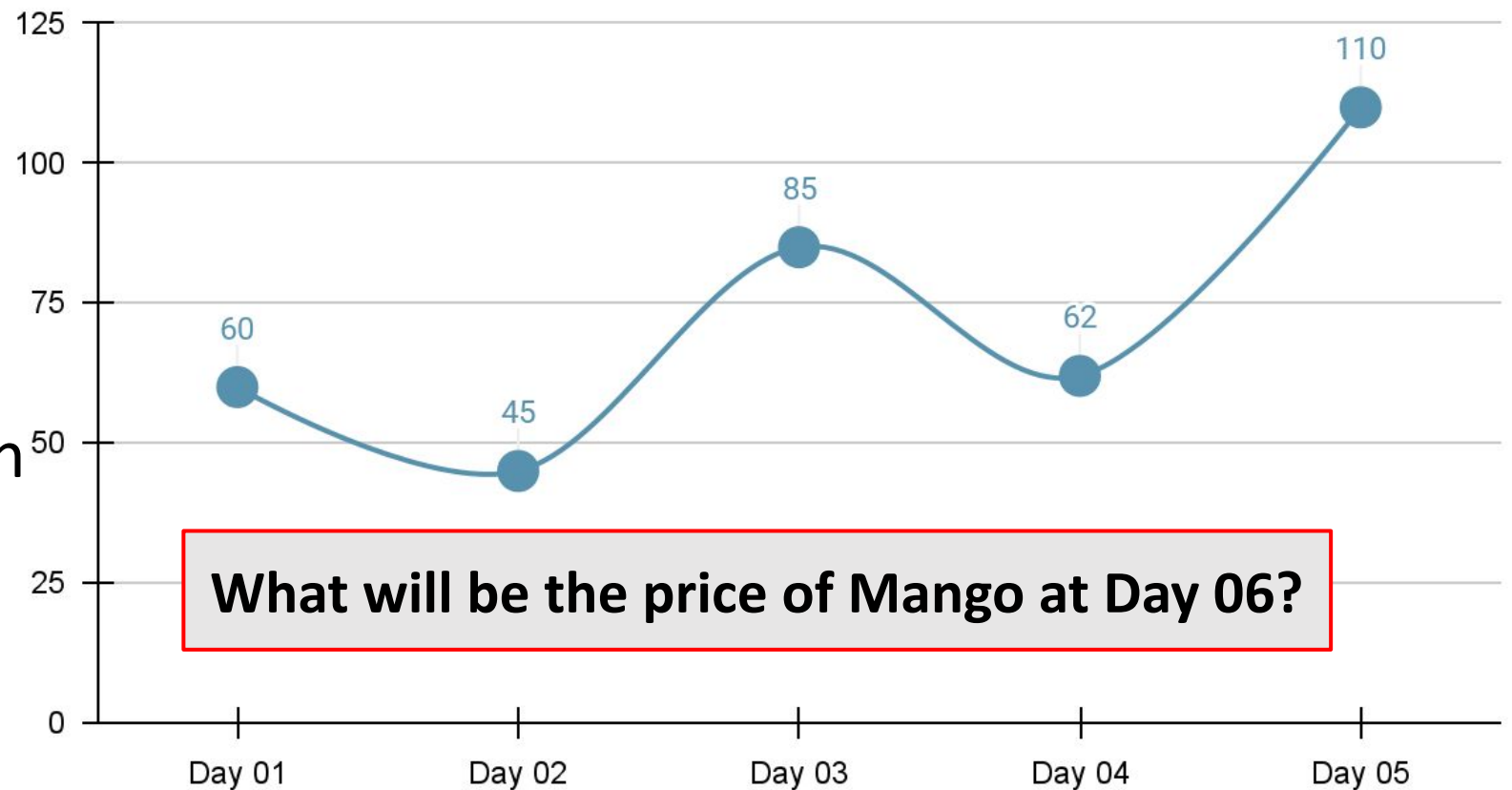


# Machine Learning

*Let's analyse the data from graph!*

- The data points are not on a straight line.
- The data points do not have a straightforward linear relationship.
- So, we need to find a Linear Relation between the data points.

Price of Per KG Mango

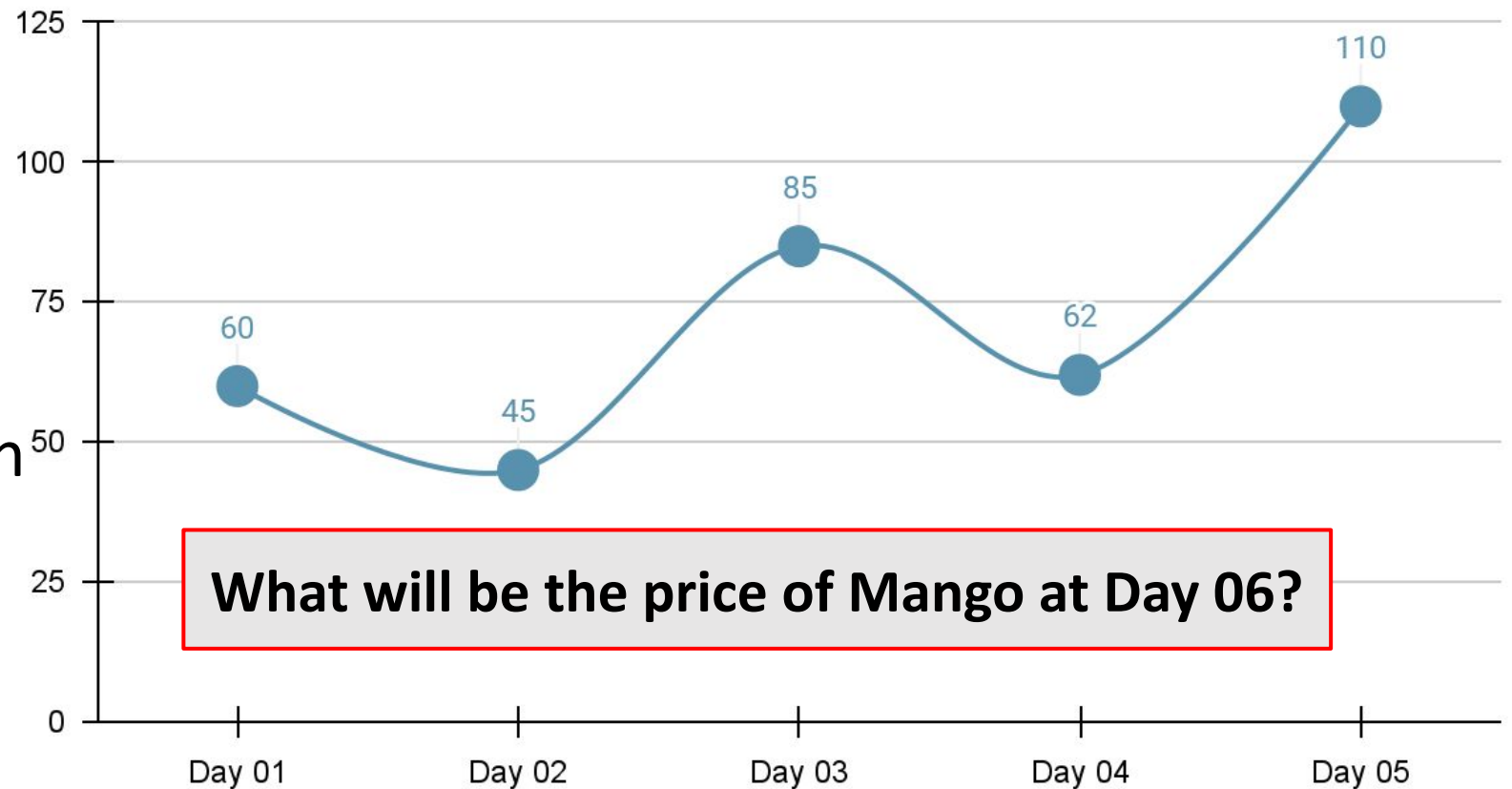


# Machine Learning

*Let's analyse the data from graph!*

- The data points are not on a straight line.
- The data points do not have a straightforward linear relationship.
- So, we need to find a Linear Relation between the data points.
- Can you tell what is our main purpose in ML?

Price of Per KG Mango



# **| Machine Learning**

***Remember, Our main goal is to predict (guess) the value, not calculate the actual value.***



# | Machine Learning

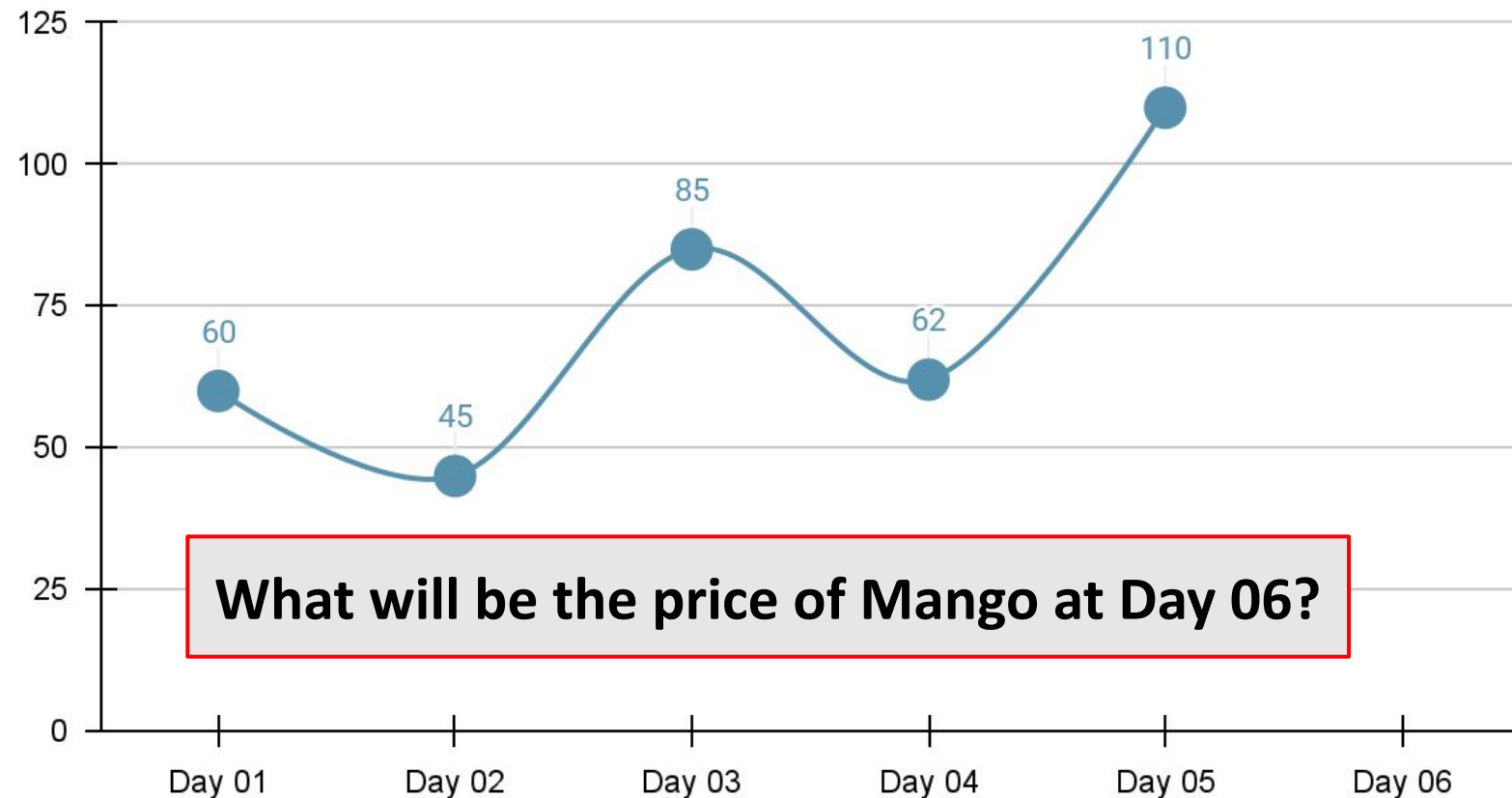
*Remember, Our main goal is to predict (guess) the value, not calculate the actual value.*

***When we guess something, there is always room for error.***

# Machine Learning

*Can we guess the price for day - 6?*

Price of Per KG Mango

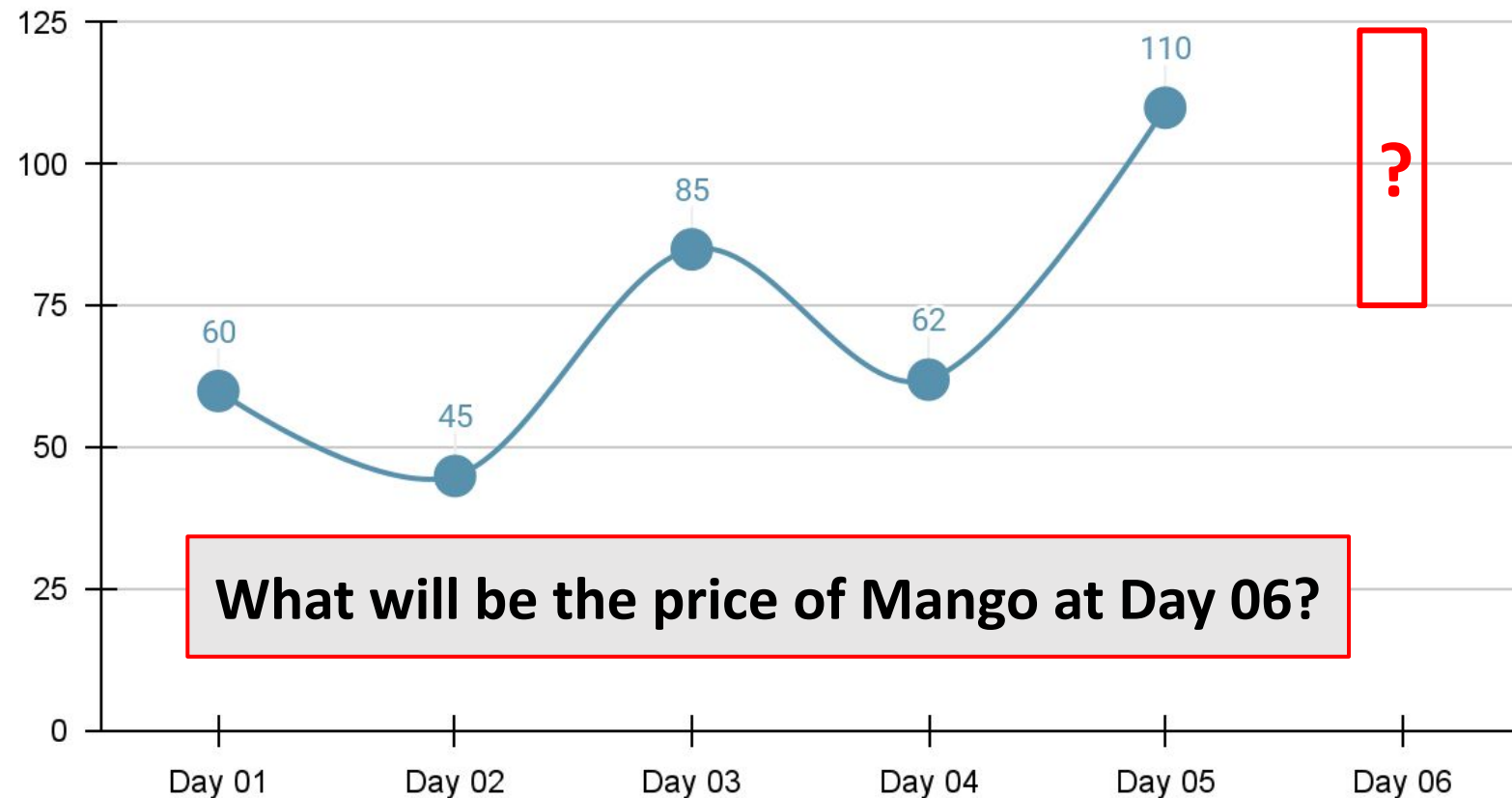


**What will be the price of Mango at Day 06?**

# Machine Learning

*Can we guess the price for day - 6?*

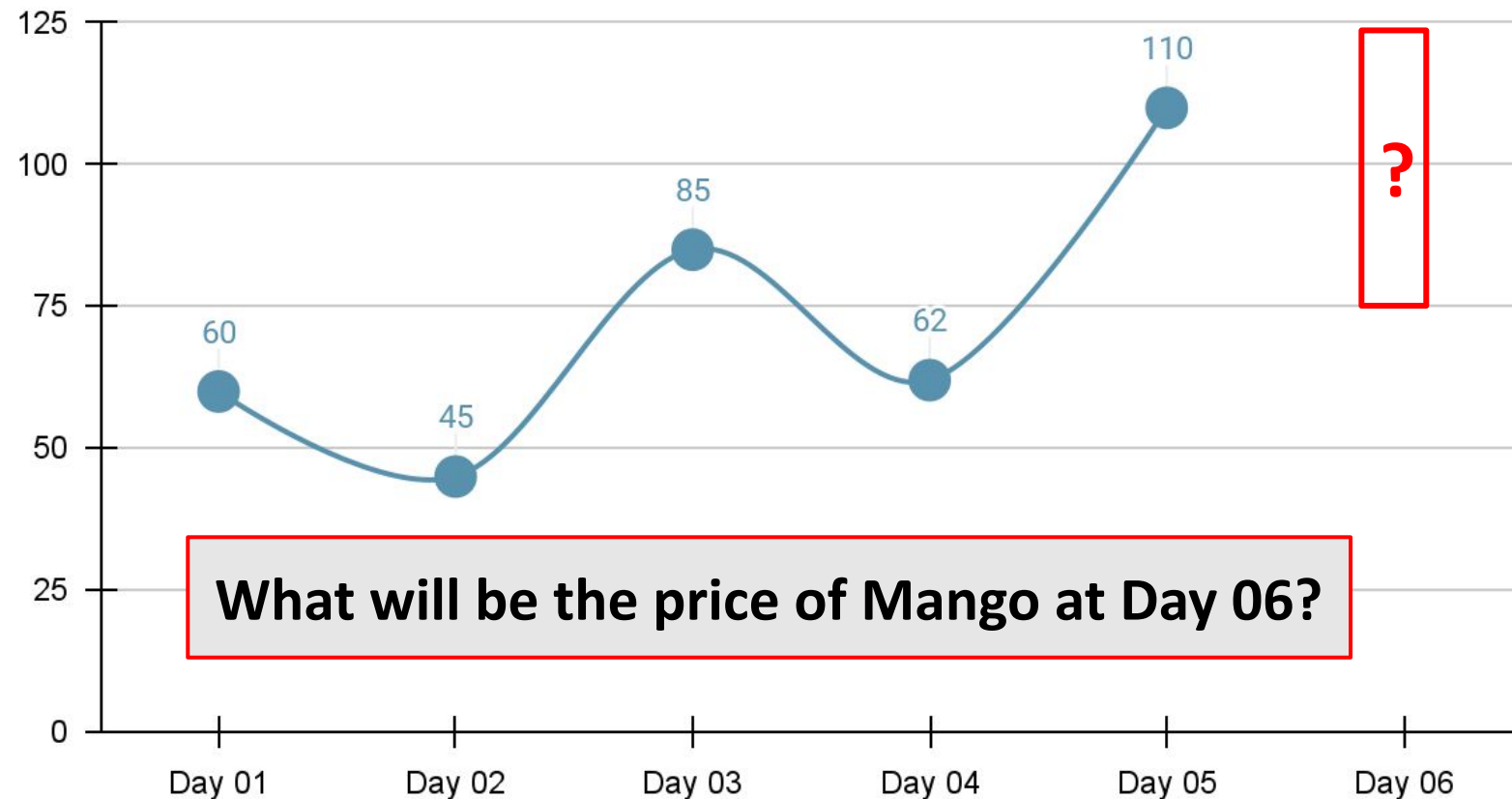
Price of Per KG Mango



# Machine Learning

*Price should be between **75 to 120 tk***

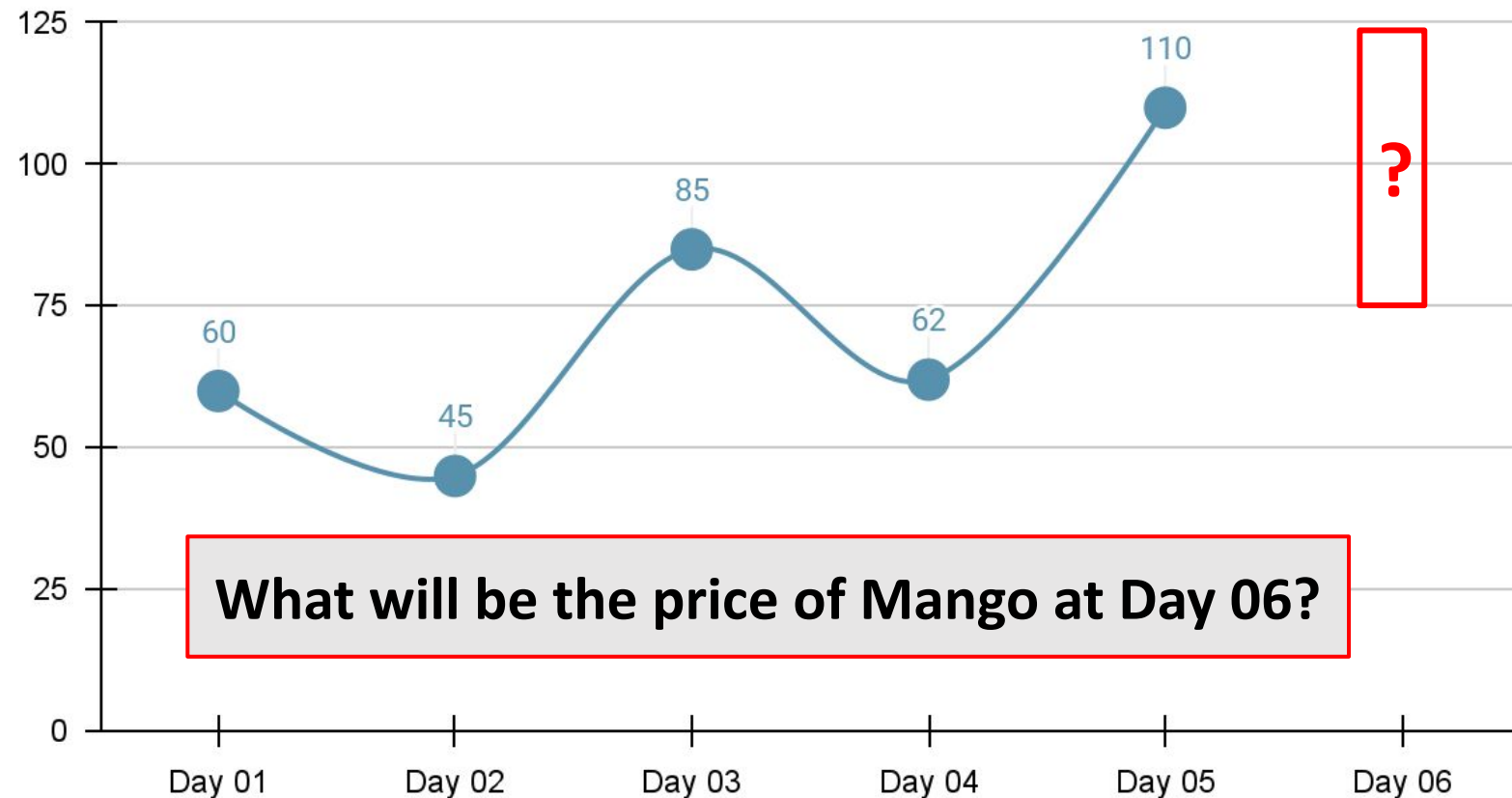
Price of Per KG Mango



# Machine Learning

*How did we predict this range?*

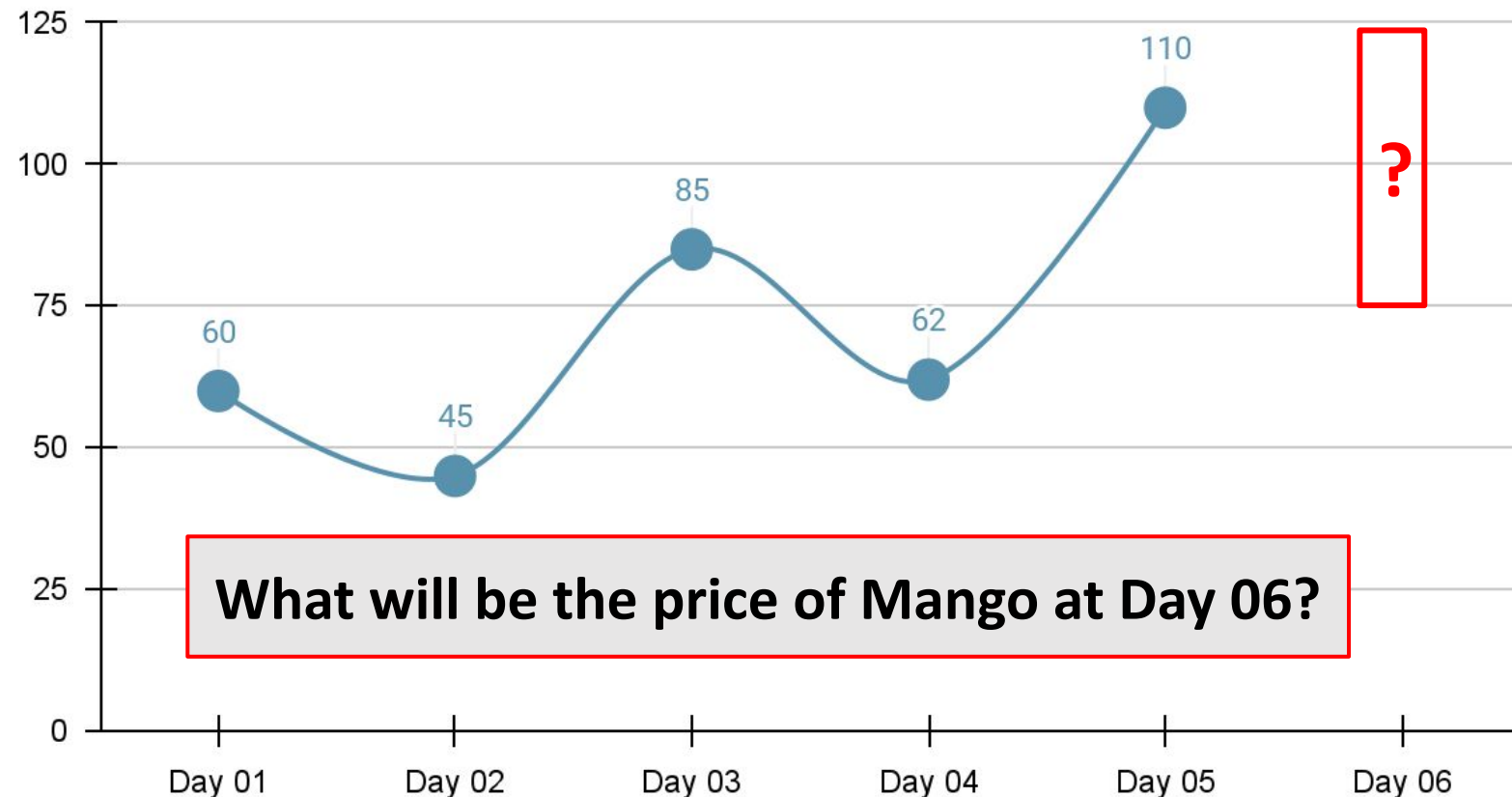
Price of Per KG Mango



# Machine Learning

*Because we have imagined a straight area that passes through the data points.*

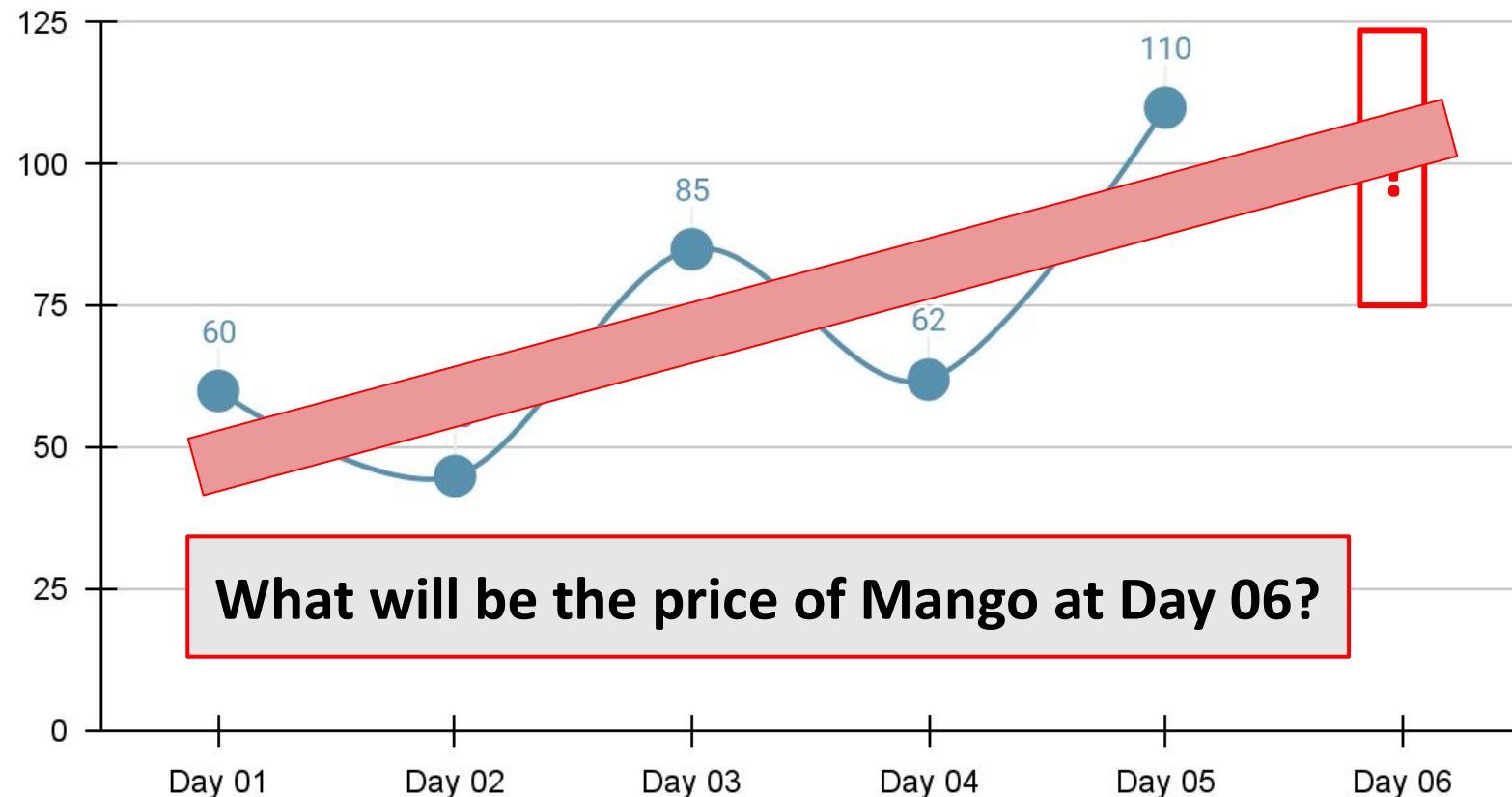
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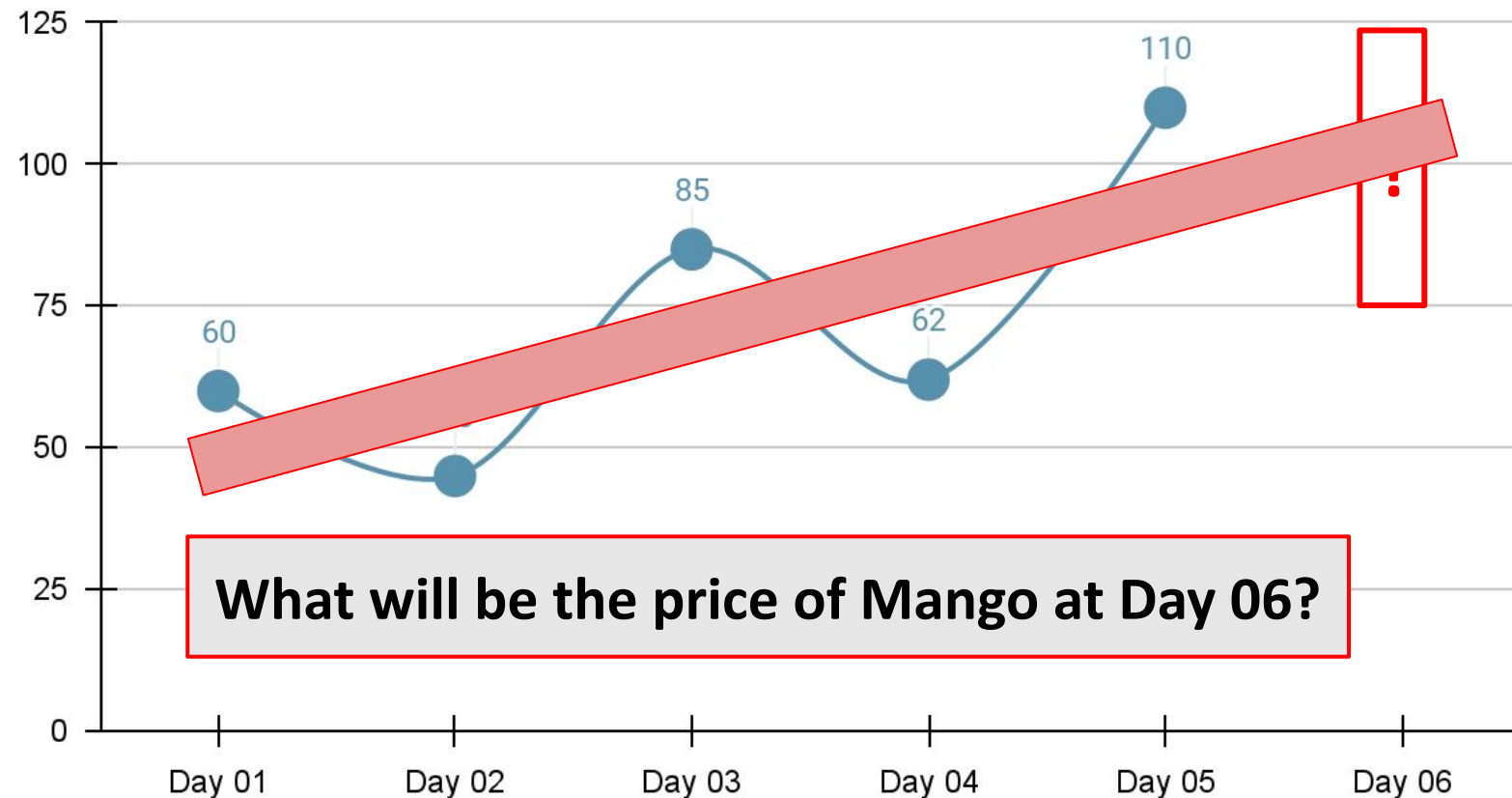
Price of Per KG Mango



# Machine Learning

*Now, instead of a straight area, we have to draw a straight line so that the prediction error can be minimized as much as possible.*

Price of Per KG Mango

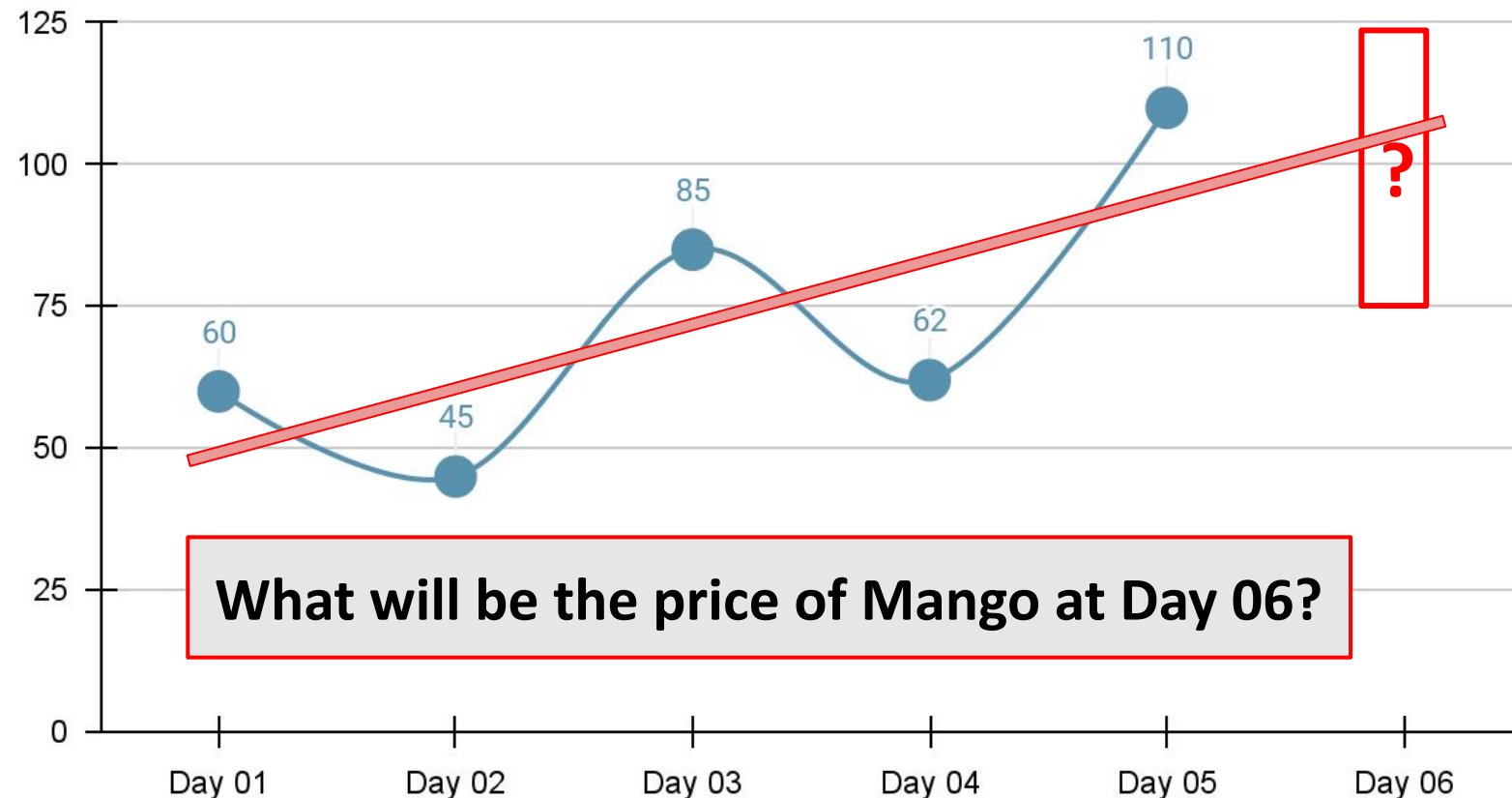




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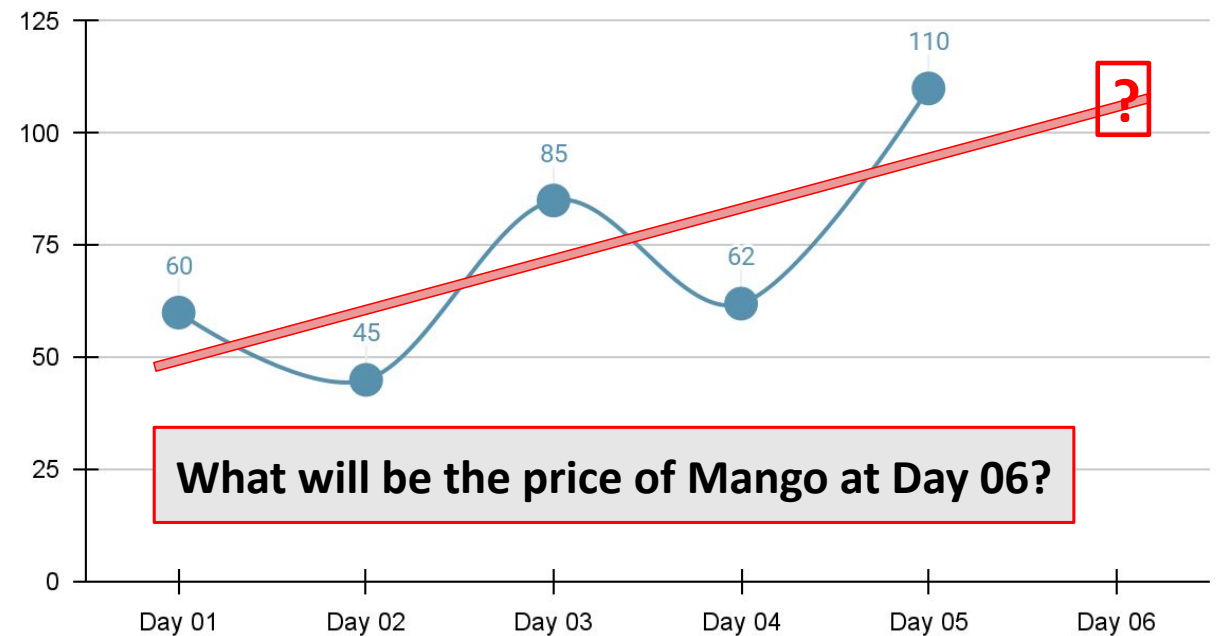
Price of Per KG Mango



# Machine Learning

*How can we find this straight line?*

Price of Per KG Mango

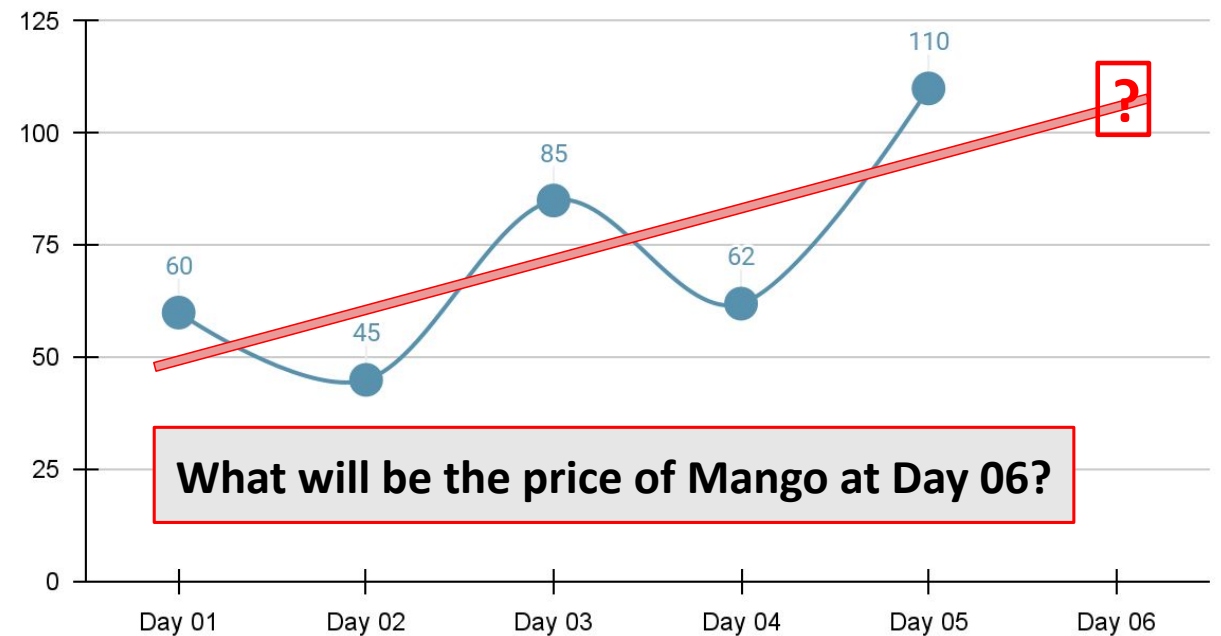


# Machine Learning

*How can we find this straight line?*

- We know, equation of a straight line is:  $y = mx + b$

Price of Per KG Mango

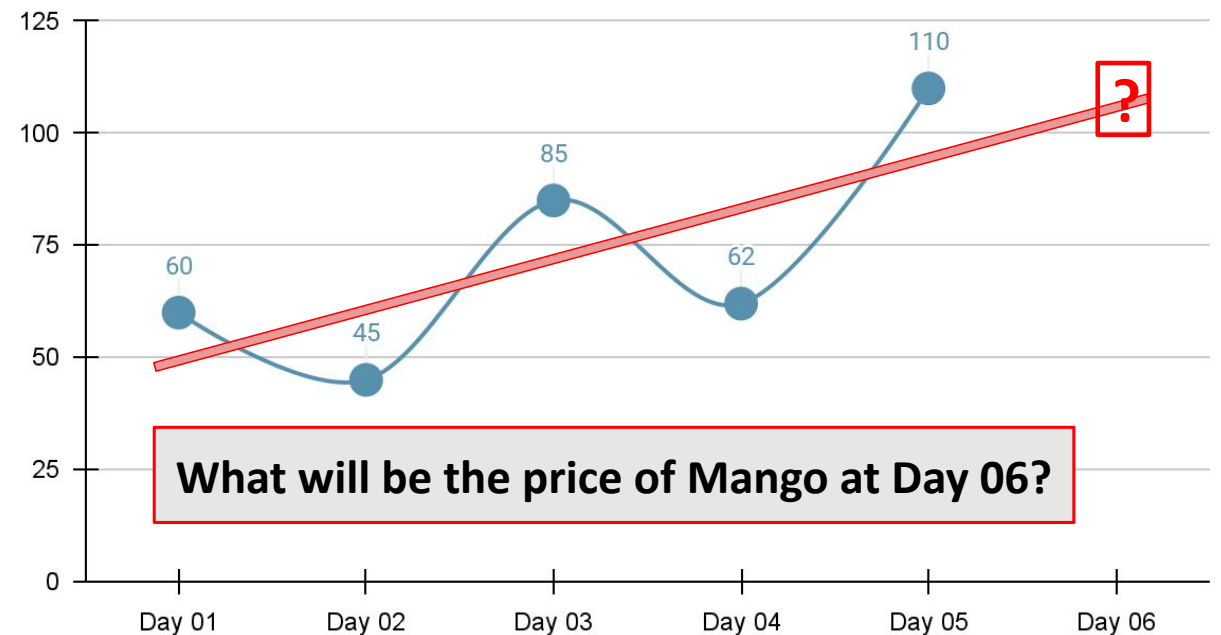


# Machine Learning

*How can we find this straight line?*

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- **Here,  $x = 6$  and we need to calculate  $y$ .**

Price of Per KG Mango

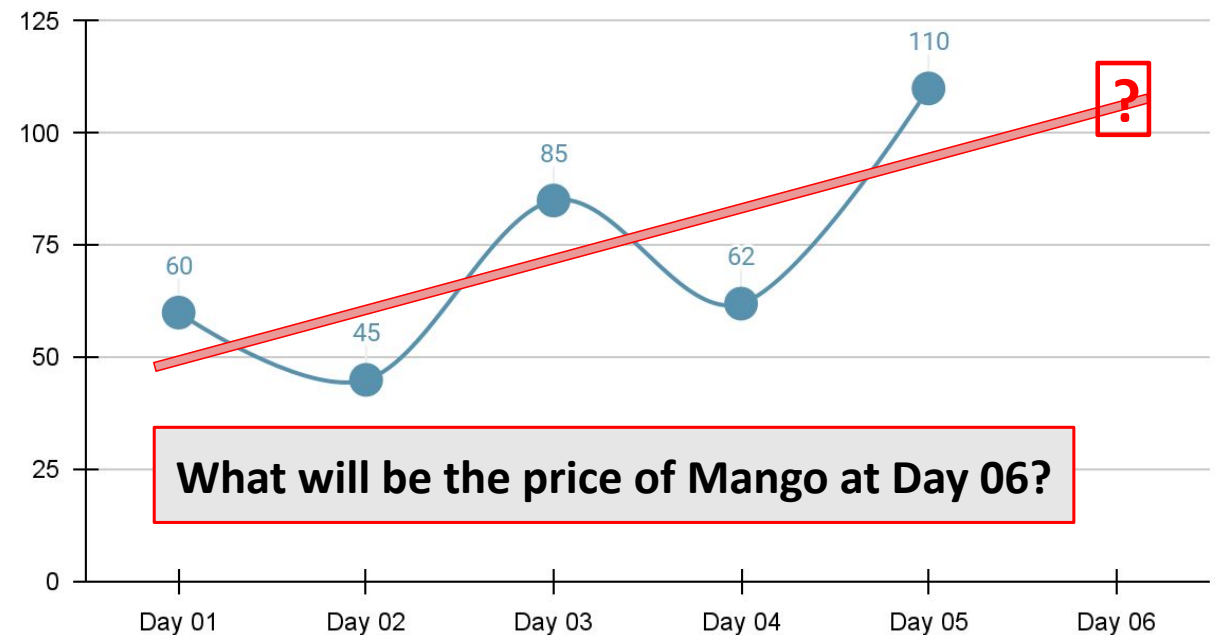


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- **Like before, the values of  $m$  and  $b$  need to be calculated.**

Price of Per KG Mango

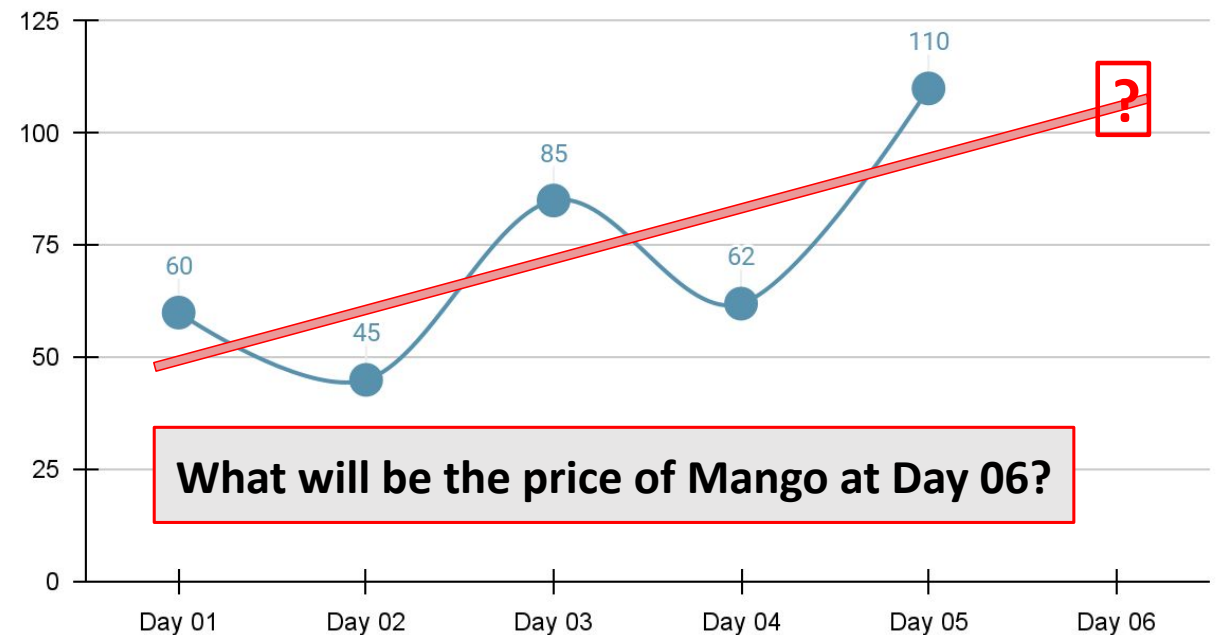


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- **But, in this case we can not directly calculate the values of  $m, b$**

Price of Per KG Mango

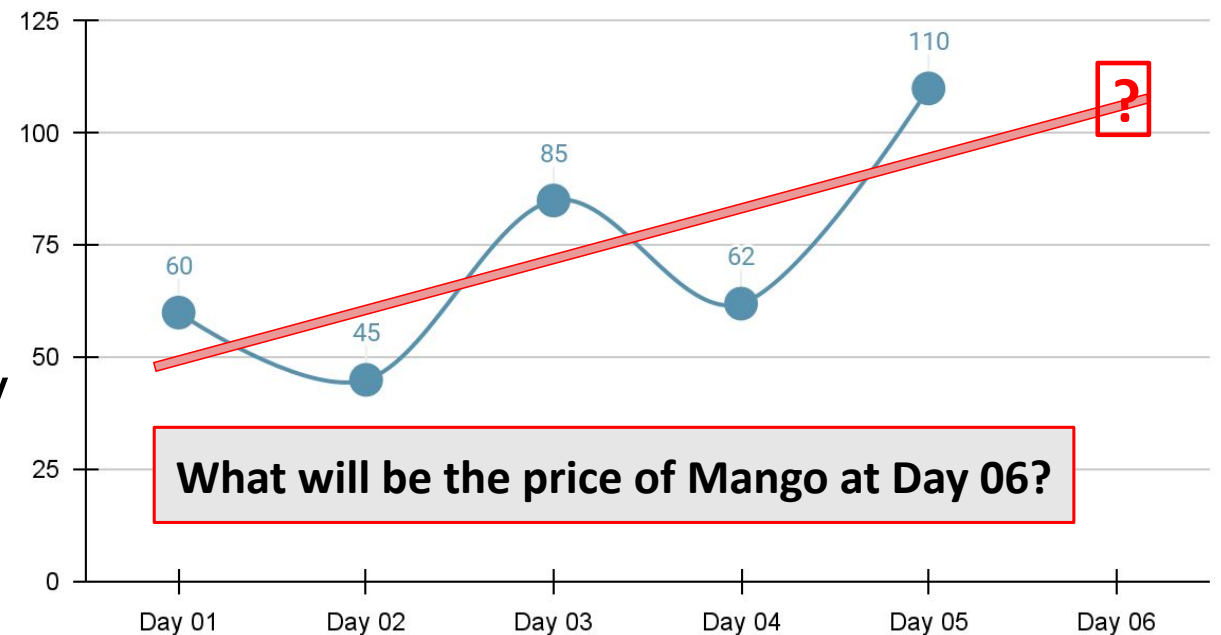


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- **So, again we need to guess the values for  $m, b$**

Price of Per KG Mango

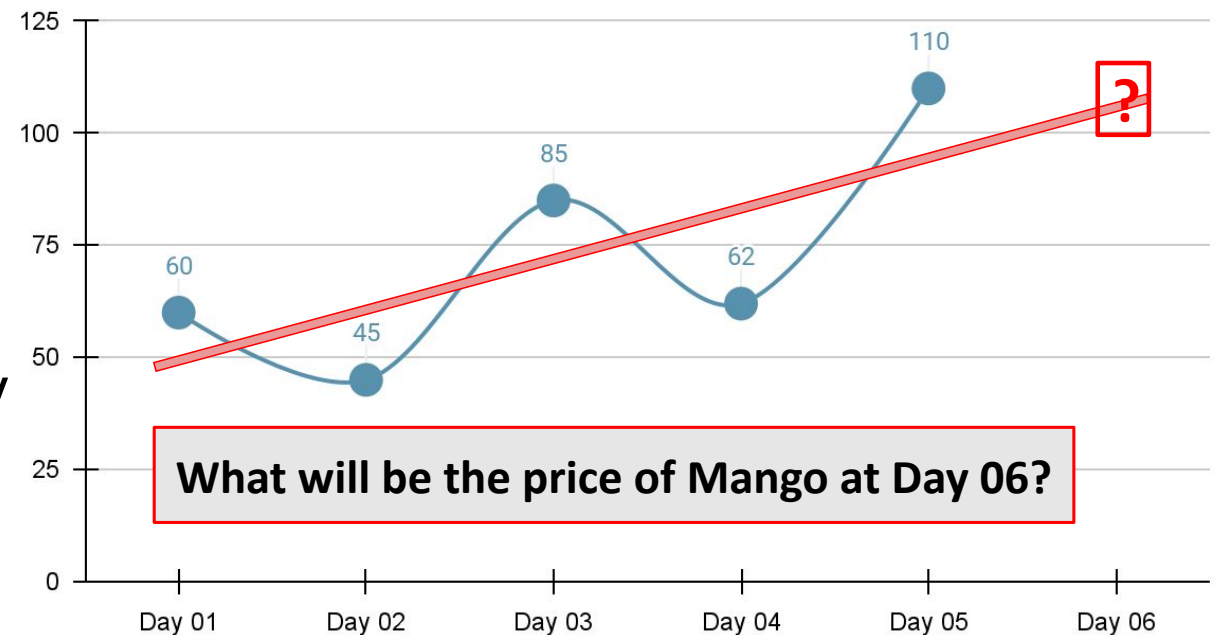


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- So, again we need to guess the values for  $m, b$
- **We can guess infinite values and draw infinite lines.**

Price of Per KG Mango

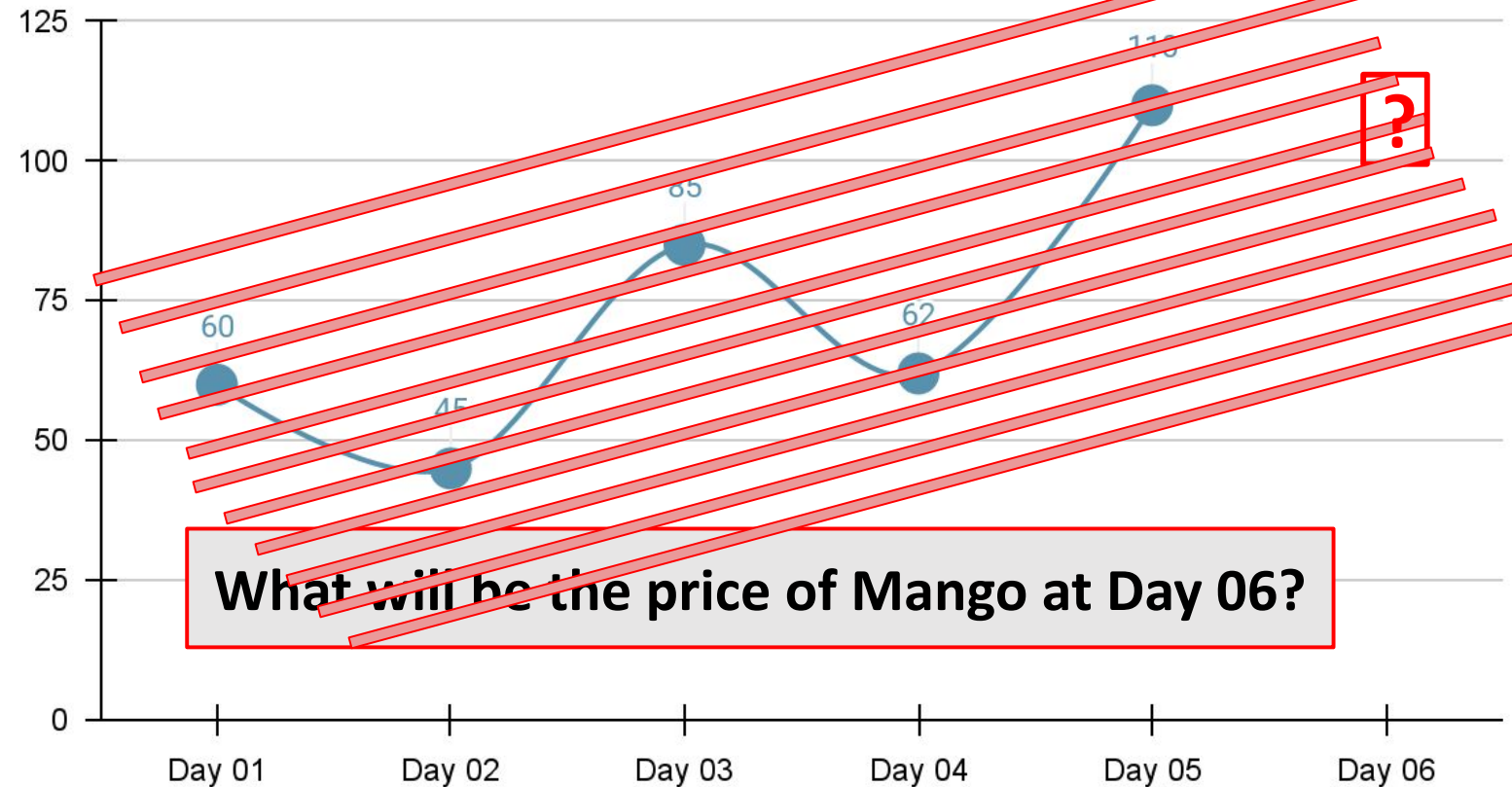




# Machine Learning

- We can guess infinite values and draw infinite lines.

Price of Per KG Mango

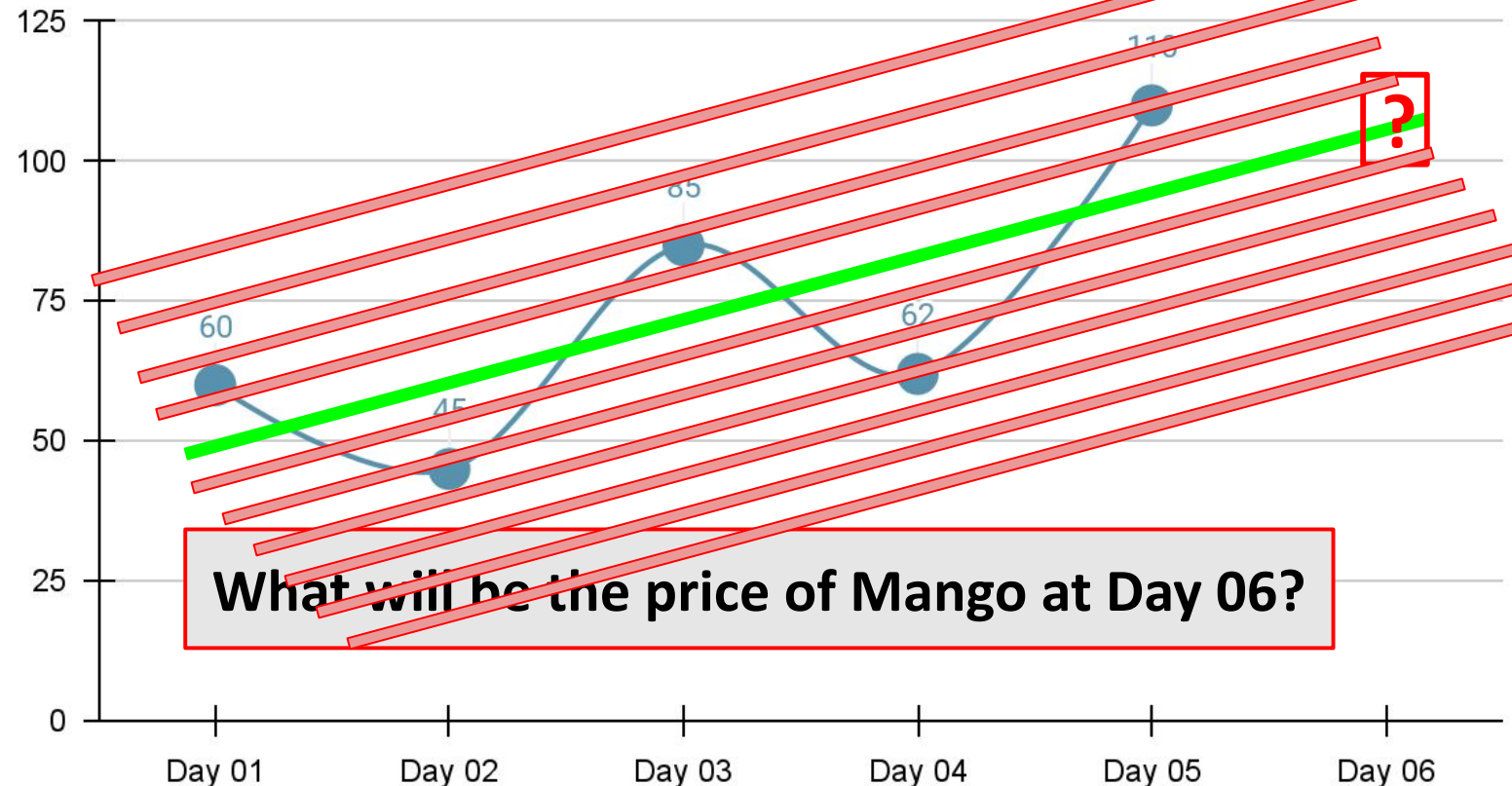


# Machine Learning

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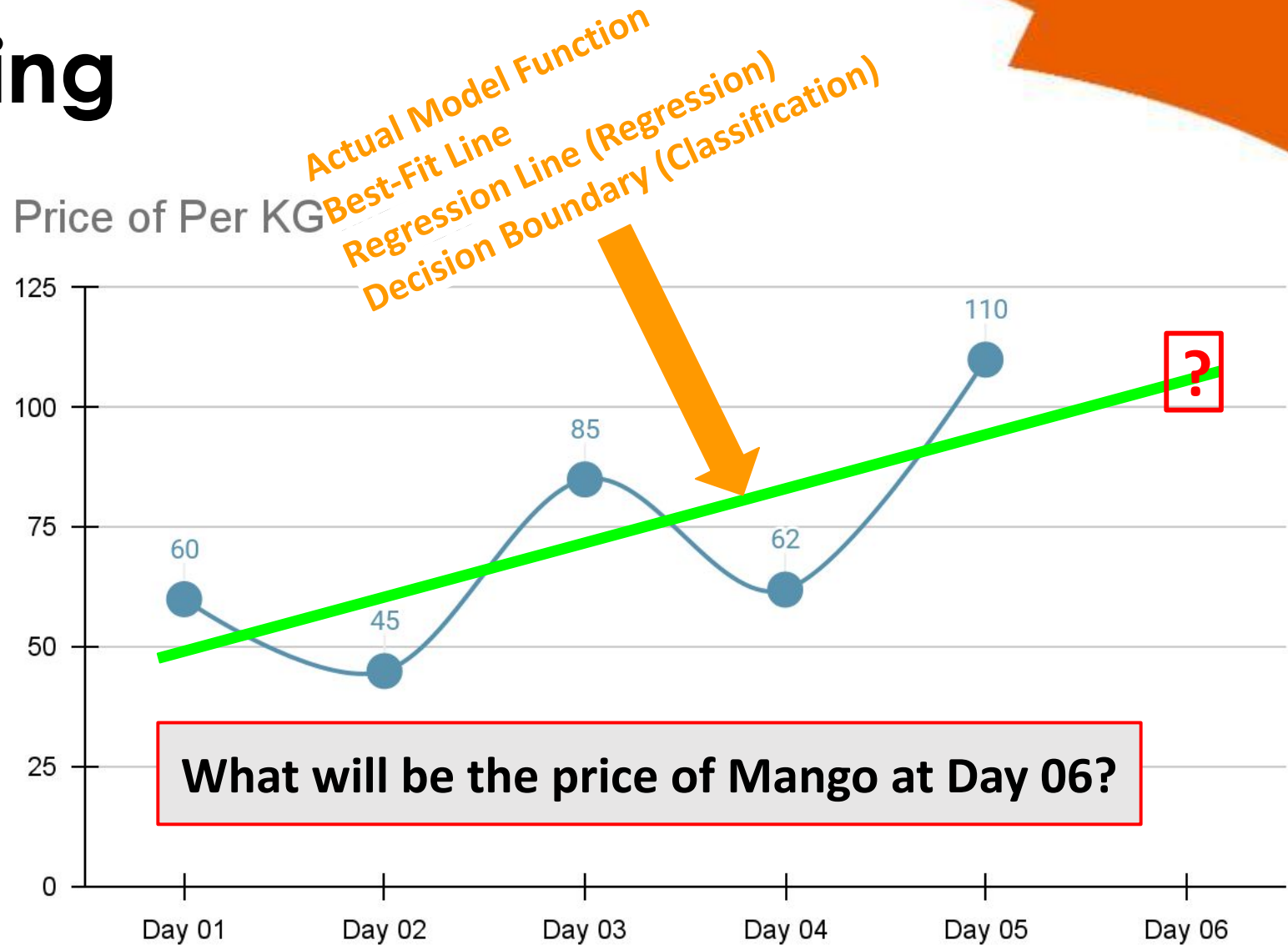
- **Condition:** We need to calculate  $m$  and  $b$  in such a way that our prediction error is minimized as much as possible.

Price of Per KG Mango



# Machine Learning

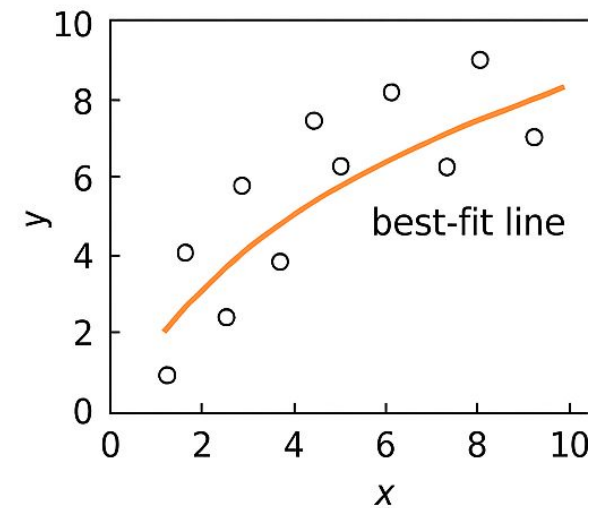
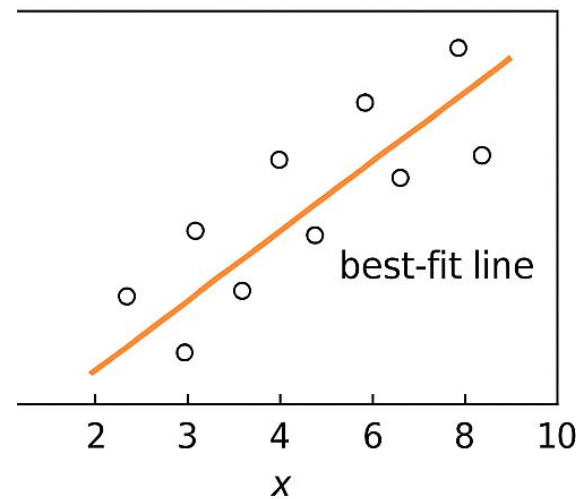
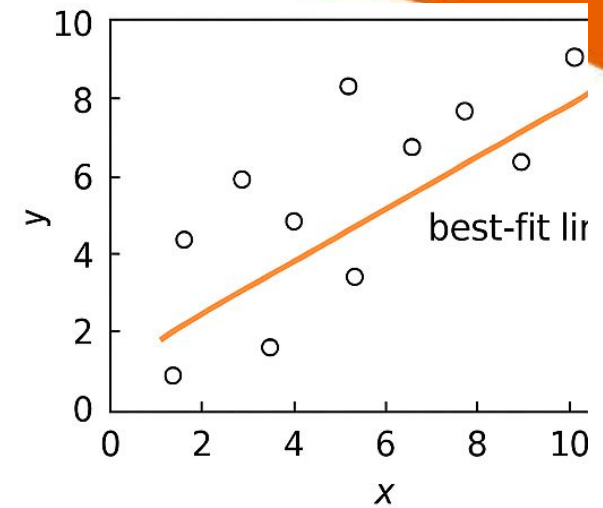
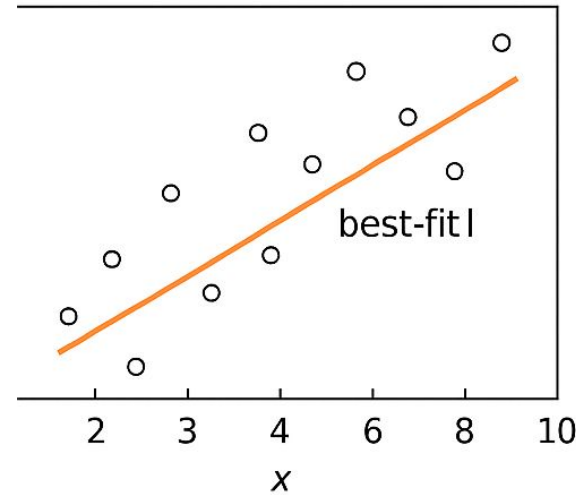
- In general, we just call it the model or hypothesis function (the function the algorithm learns).
- In regression, that line is usually called the regression line or line of best fit.
- In classification, a similar line is often called the decision boundary.



# Machine Learning

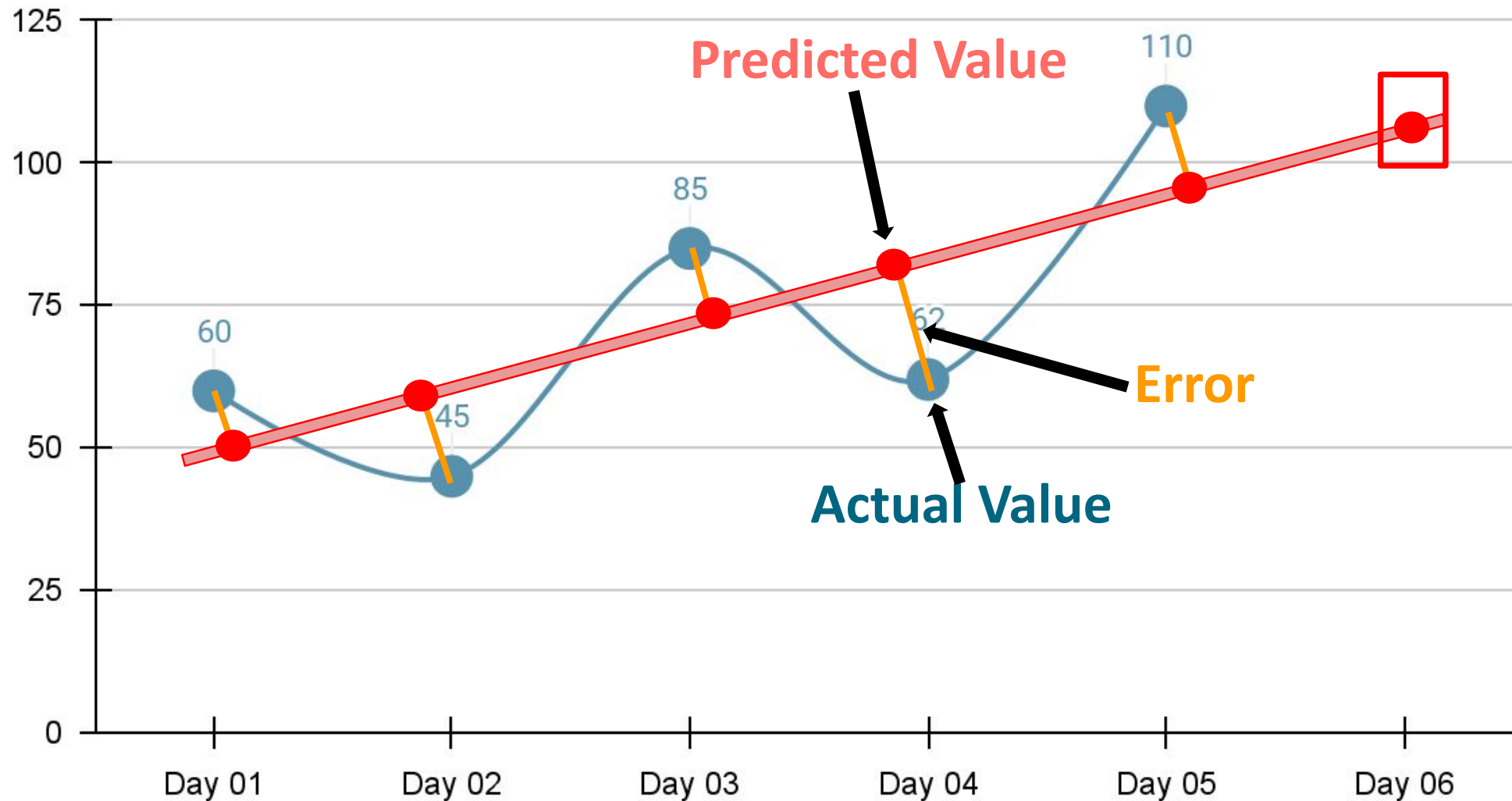
Data যেমনই হোক, আমাদের  
জীবনের লক্ষ্য, উদ্দেশ্য, সাধনা  
একটাই-

**Best-Fit Line খুঁজে বের করা**



# Machine Learning

*The Error: Difference between Actual Value and Predicted Value*



# Machine Learning

- “**Error**” is usually formalized as **loss function** which measures error for one example.

Input	Actual Result	Predicted Result	Loss Function
Day 1	60	56	4
Day 2	45	49	-4
Day 3	85	79	6
Day 4	62	61	1
Day 5	110	95	15

# Machine Learning

- **“Cost Function”** is usually the average (or sum) of the loss over the whole dataset.

Input	Actual Result	Predicted Result	Loss Function
Day 1	60	56	4
Day 2	45	49	-4
Day 3	85	79	6
Day 4	62	61	1
Day 5	110	95	15
Cost			4.4 (avg)

# | Machine Learning

- **Cost function** (average loss over many samples) is what really indicates the overall performance of the model.
- **Loss** tells you how good or bad a single prediction is.

**\*\*\*Mathematical formula of Cost function changes depending on the machine learning model.**




# Machine Learning

Finding the BEST-FIT Line:

1. Numerical Analysis (least squares) Method :

$$m = \frac{\overline{xy} - \bar{x}\bar{y}}{\overline{x^2} - (\bar{x})^2}$$

$$b = \bar{y} - m \cdot \bar{x}$$

- $\bar{x}$  is the mean of  $x$ .
- $\bar{y}$  is the mean of  $y$ .
- $\overline{xy}$  is the mean of the product of  $x$  and  $y$ .
- $(\bar{x})^2$  is the square of the mean of  $x$ .
- $\overline{x^2}$  is the mean of  $x^2$ . 

# Machine Learning

## 1. Numerical Analysis (least squares) Method :

### 1. Mean of $x$

$$\bar{x} = \frac{1 + 2 + 3 + 4 + 5}{5} = \frac{15}{5} = 3$$

### 2. Mean of $y$

$$\bar{y} = \frac{60 + 45 + 85 + 62 + 110}{5} = \frac{362}{5} = 72.4$$

Input	Actual Result
x	y
Day 1	60
Day 2	45
Day 3	85
Day 4	62
Day 5	110

# Machine Learning

## 1. Numerical Analysis (least squares) Method :

### 3. Mean of $xy$

First compute  $xy$  for each pair:

- Day 1:  $1 \times 60 = 60$
- Day 2:  $2 \times 45 = 90$
- Day 3:  $3 \times 85 = 255$
- Day 4:  $4 \times 62 = 248$
- Day 5:  $5 \times 110 = 550$

Sum:

$$60 + 90 + 255 + 248 + 550 = 1203$$

Mean:

$$\overline{xy} = \frac{1203}{5} = 240.6$$

Input	Actual Result
x	y
Day 1	60
Day 2	45
Day 3	85
Day 4	62
Day 5	110

# Machine Learning

## 1. Numerical Analysis (least squares) Method :

### 4. Mean of $x^2$

First compute  $x^2$ :

- $1^2 = 1$
- $2^2 = 4$
- $3^2 = 9$
- $4^2 = 16$
- $5^2 = 25$

Sum:

$$1 + 4 + 9 + 16 + 25 = 55$$

Mean:

$$\overline{x^2} = \frac{55}{5} = 11$$

Input	Actual Result
x	y
Day 1	60
Day 2	45
Day 3	85
Day 4	62
Day 5	110

# Machine Learning

1. Numerical Analysis (least squares) Method :

$$m = \frac{\overline{xy} - \bar{x}\bar{y}}{\overline{x^2} - (\bar{x})^2} = \frac{240.6 - (3)(72.4)}{11 - 3^2} = \frac{240.6 - 217.2}{11 - 9} = \frac{23.4}{2} = 11.7$$

$$b = \bar{y} - m \cdot \bar{x} = 72.4 - (11.7)(3) = 72.4 - 35.1 = 37.3$$

- $\bar{x} = 3$
- $\bar{y} = 72.4$
- $\overline{xy} = 240.6$
- $\overline{x^2} = 11$

# | Machine Learning

1. Numerical Analysis (least squares) Method :

Price of Mango on Day 6 is:

$$\text{price} = 11.7 * 6 + 37.3 \Rightarrow 107.5 \text{ tk}$$

# | Machine Learning

Finding the BEST-FIT Line:

1. Numerical Analysis (least squares) Method

# | Machine Learning

Finding the BEST-FIT Line:

1. Numerical Analysis (least squares) Method
2. Gradient Descent Algorithm (most important)



# | Machine Learning

## 2. Gradient Descent Algorithm

### What it is:

Gradient descent is an **optimization algorithm** used to **find the best parameters** (like  $m$  and  $b$  in a line) that **minimize a cost function**.

### Core idea:

1. Start with a random guess of the parameters.
2. Compute the **gradient** (the direction of steepest increase of the cost).
3. Move a small step in the **opposite** direction (downhill) to reduce the cost.
4. Repeat this process until the cost stops decreasing much.

# | Machine Learning



CONGRATULATIONS

We have just started  
our Machine Learning  
Journey!

# | Machine Learning

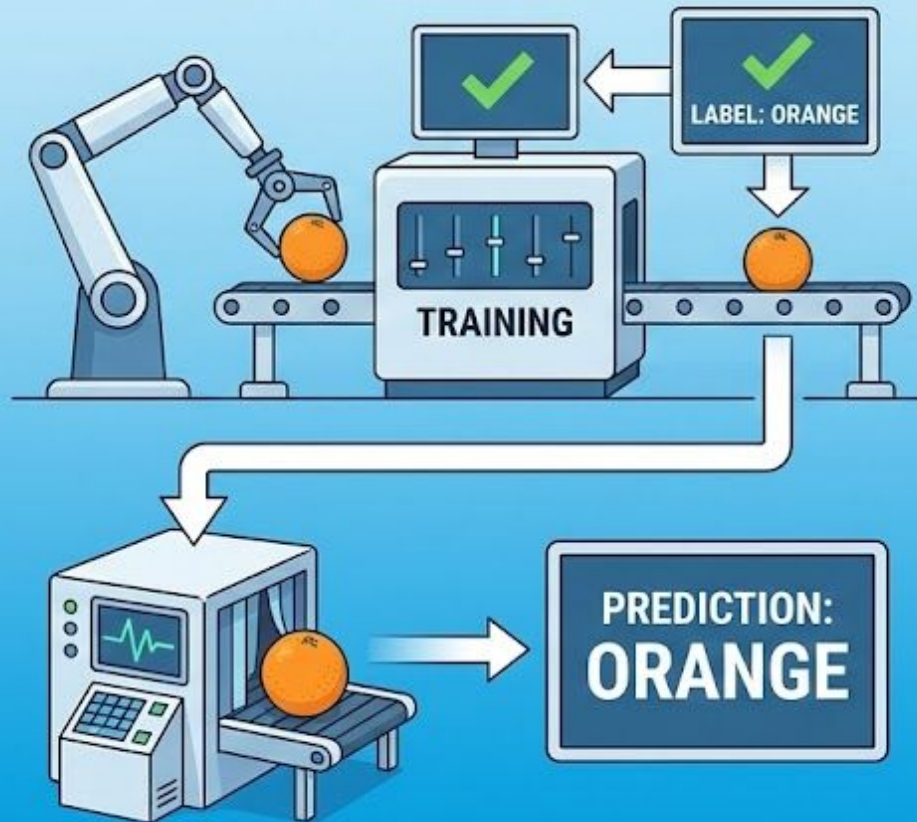
## *Three Main Types of Learning:*

- **Supervised Learning:** Learn from input–output pairs (features, label)
- **Unsupervised Learning:** Learn patterns from data without labels
- **Reinforcement Learning (RL):** Learn by interacting with an environment and receiving rewards

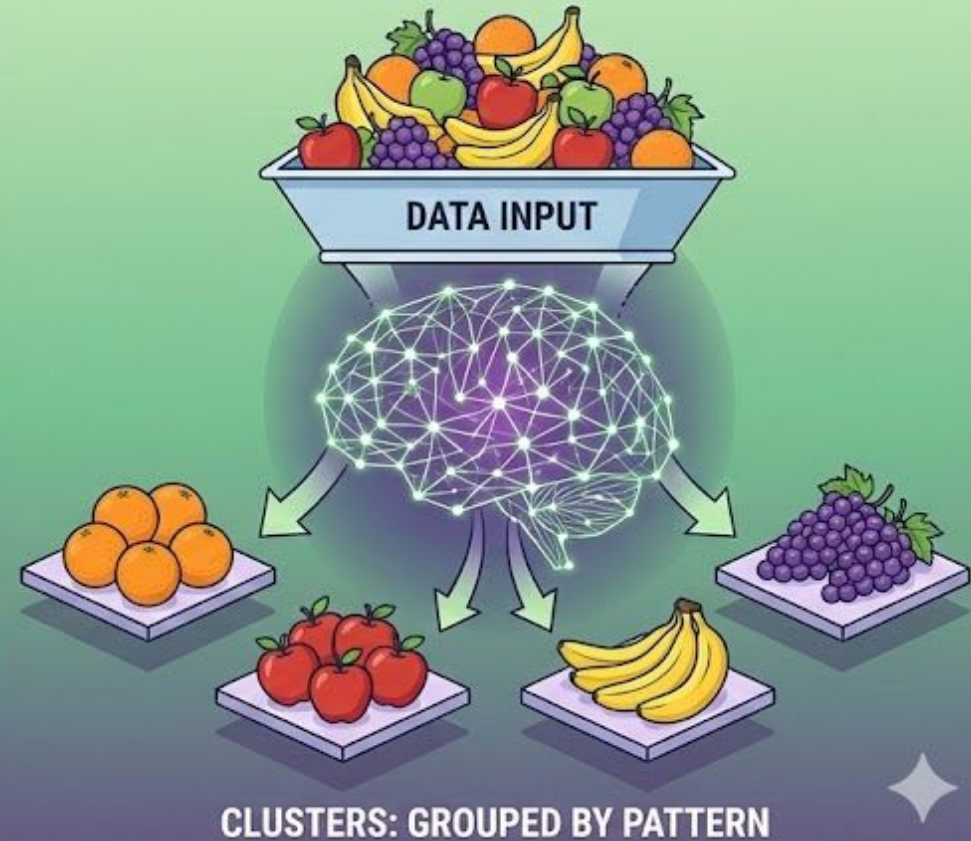
# Machine Learning

## MACHINE LEARNING: TWO APPROACHES

### SUPERVISED LEARNING (Labeled Data)

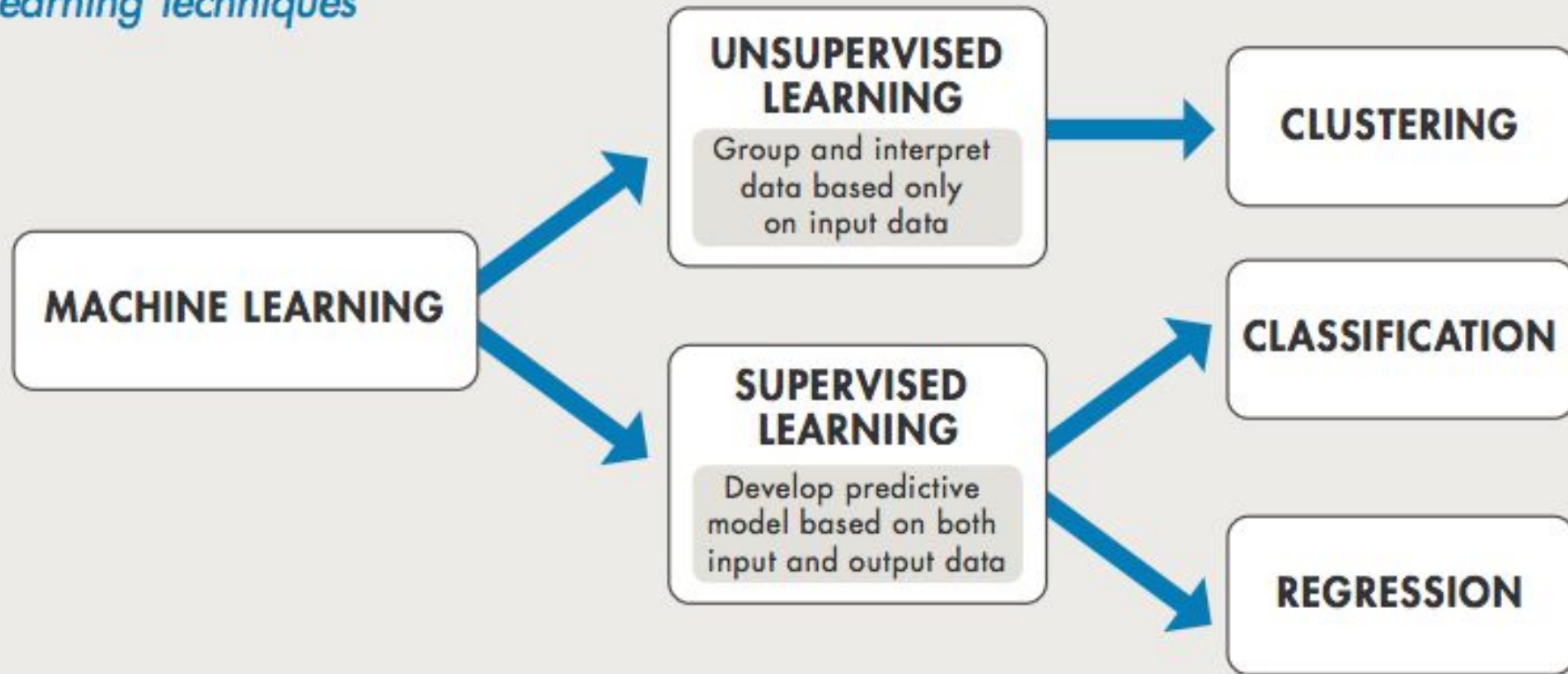


### UNSUPERVISED LEARNING (Unlabeled Data)



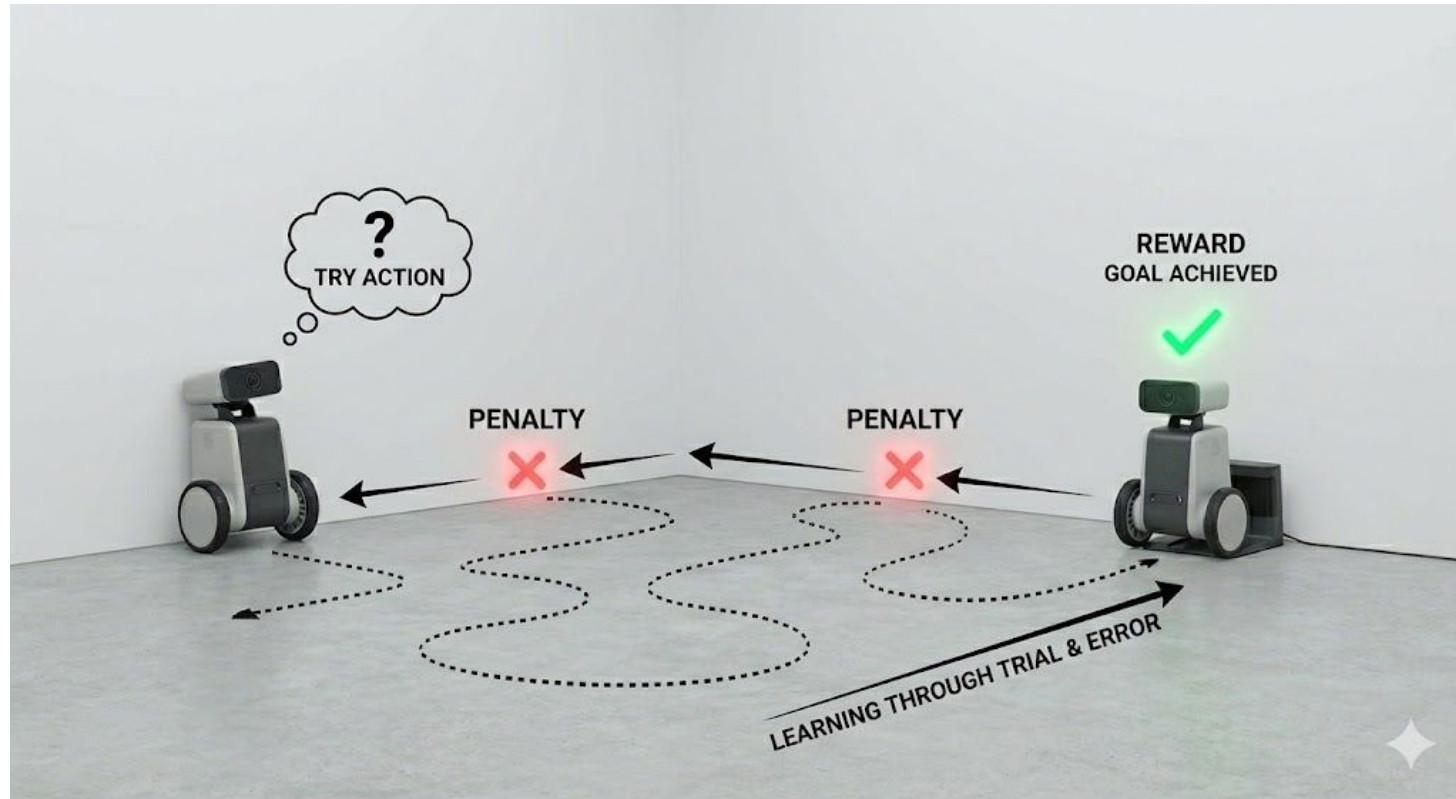
# Machine Learning

## *Machine Learning Techniques*





# Machine Learning



- Reinforcement Learning (RL)  
ভুল থেকে শিক্ষা নেওয়া!!

# | Domain of AI

## Artificial Intelligence

Any technique that enables computers to mimic human behaviour



1956

## Machine Learning

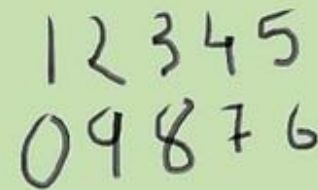
Ability to learn without explicitly being programmed



1997

## Deep Learning

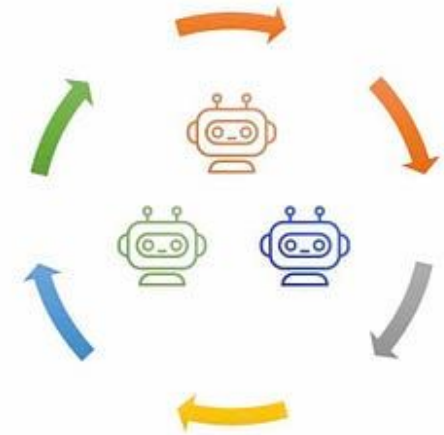
Extract patterns from Data using neural networks



2012

## Agentic AI

A software that can **perceive**, **reason**, & then **act** autonomously.

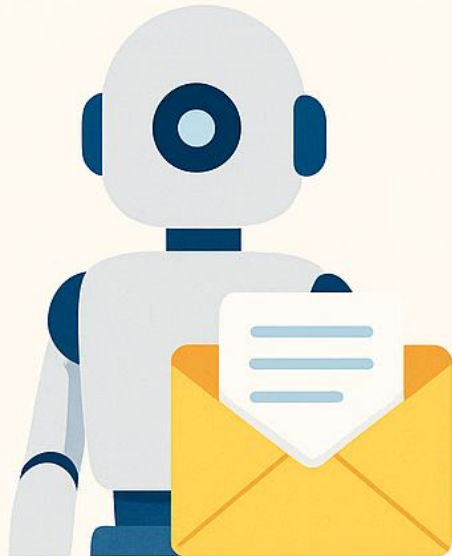


2025

# |Types of AI

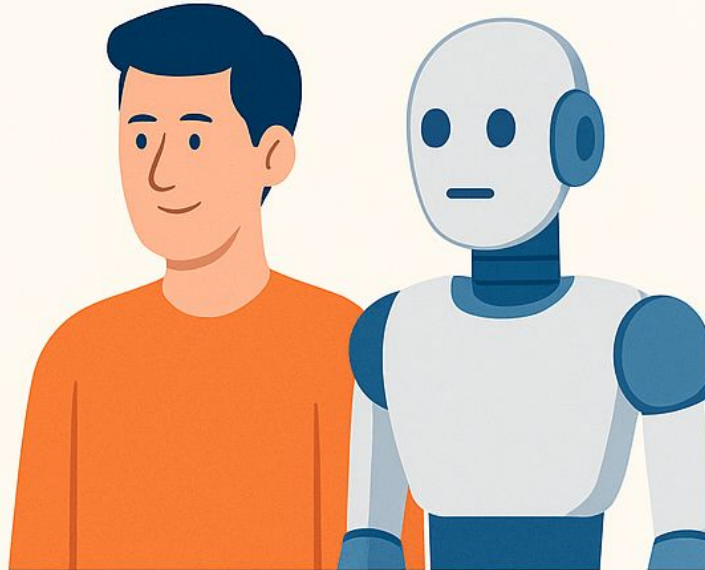
## Narrow AI (ANI)

focused on  
a specific task



## General AI (AGI)

human-level flexible  
intelligence  
(future goal)



## Super AI (ASI)

beyond human  
intelligence  
(theoretical)





# Intelligent Robots: When AI Meets Robotics





**ANY QUESTIONS ?**

# THANK YOU

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