

# Introduction to Supervised Learning: Concepts and Applications

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Community Researcher of RL & LLM  
Cohere for AI

## About Me



# Gusti Triandi Winata

### Latest Work Experiences:

- Researcher, Cohere for AI 2024 - present
- MLE Consultant, Freeport Indonesia 2023 - 2024
- Mid. MLE, eFishery 2022 - 2023
- OSS Fellowship with Adobe, Major League Hacking 2021

### Education:

- Bandung Institute of Technology Graduated at 2021  
*Bachelor of Electrical and Computer Engineering*  
*Was a Bangkit Graduate of the first batch (2020)!*

# Ground Rules

Observe the following rules to ensure a supportive, inclusive, and engaging classes



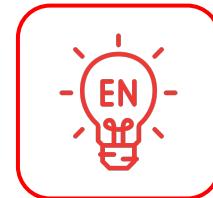
Give full attention  
in class



Mute your microphone  
when you're not talking



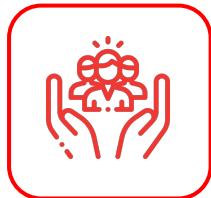
Keep your  
camera on



Turn on the CC Feature  
on Meet

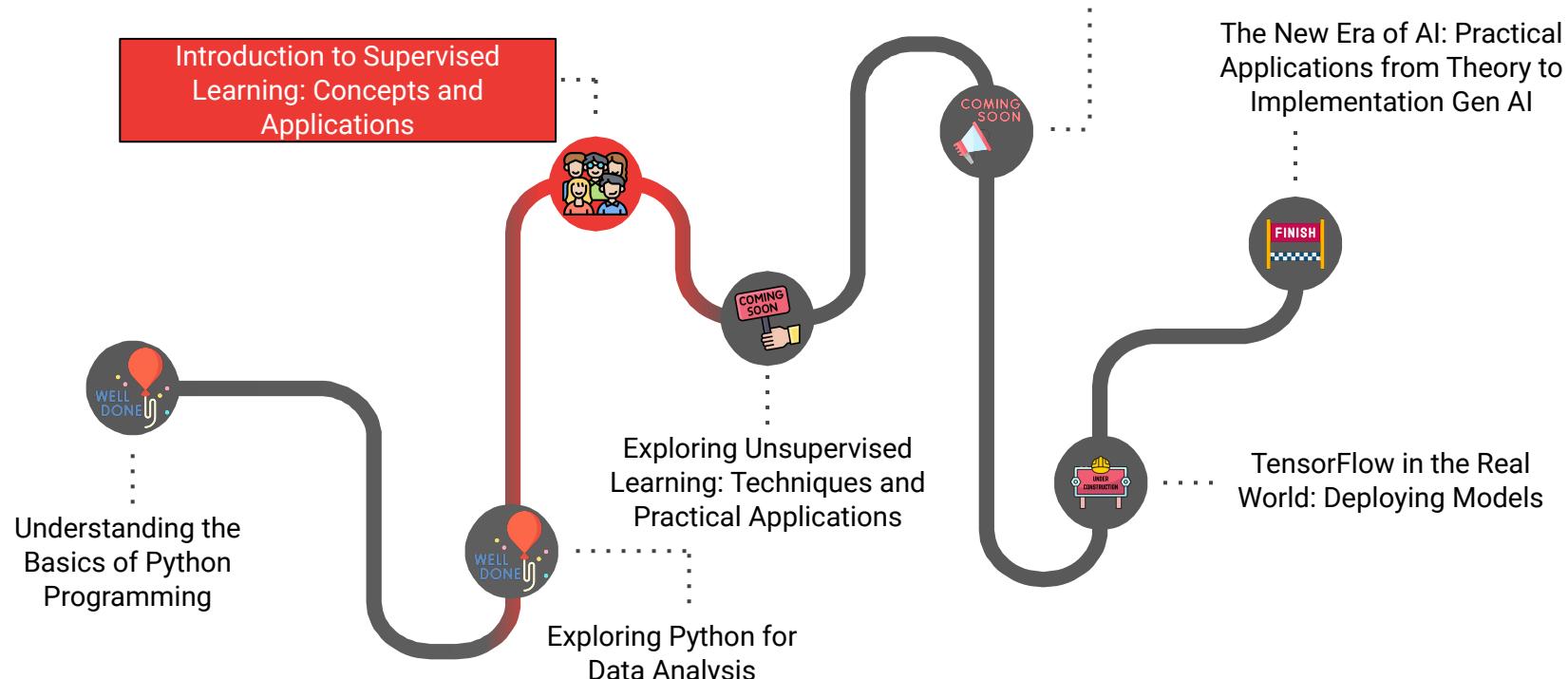


Use raise hand or chat  
to ask questions



Make this room a safe place  
to learn and share

# Where Are We Now?



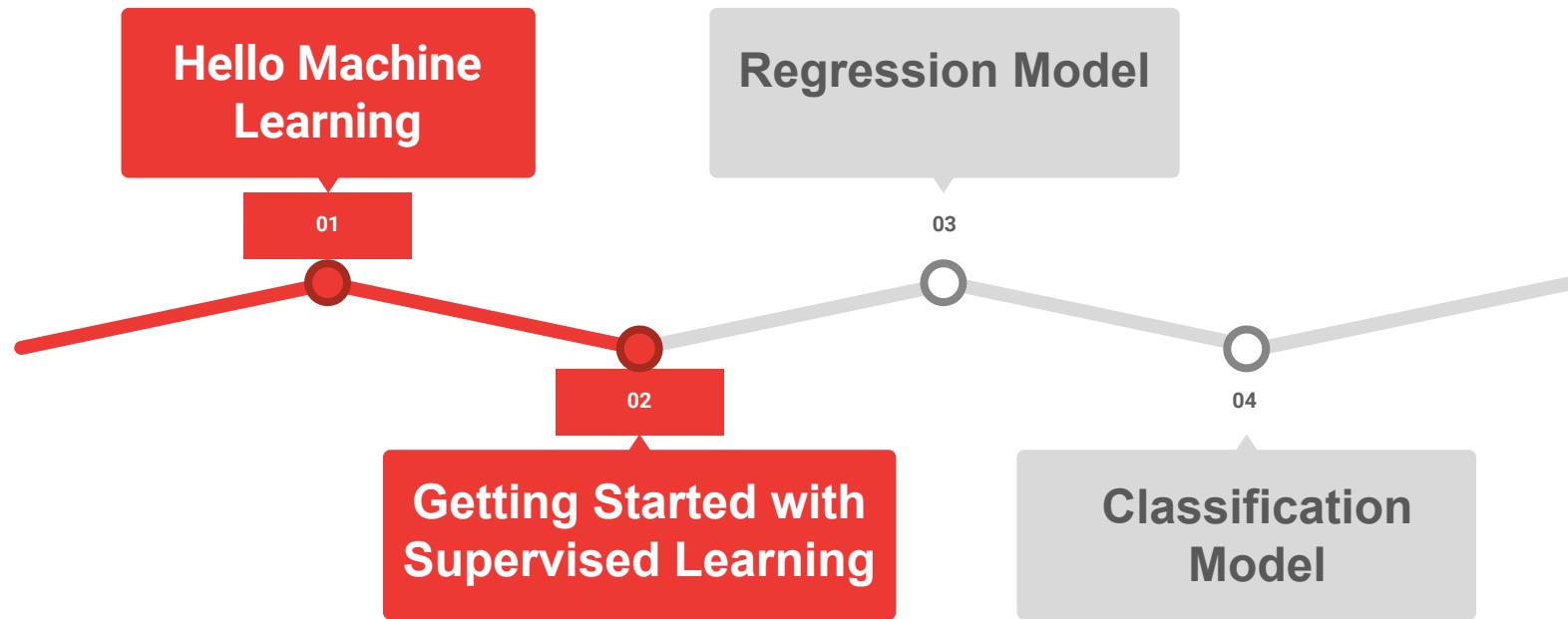
# Learning Objectives

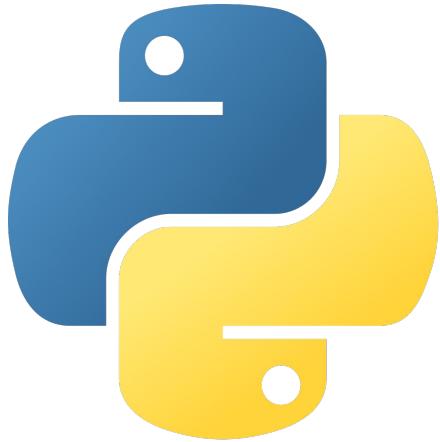
Identify basic concepts in machine learning

Conceptualising supervised learning

Comparing algorithm in supervised learning

# Today's Agenda





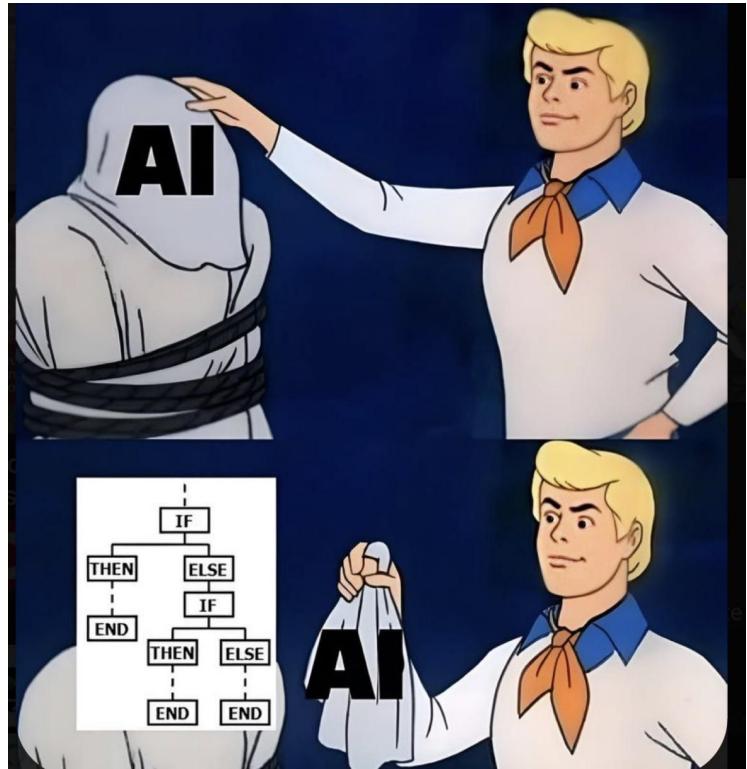
Condition of  
Comparison

## Condition Evaluations

```
if hour < 12:  
    print("Good morning!")  
  
def check(number):  
    if number > 0:  
        return "Positive"  
    elif number == 0:  
        return "Zero"  
    else:  
        return "Negative"
```

## Traditional Programming





# DATA TEAMS



Generative AI

Machine Learning

Business Intelligence

Data Literacy



Going  
back to the basics  
will strengthen  
your  
foundations...

# Hello to Machine Learning!

# Say Hello to Machine Learning!

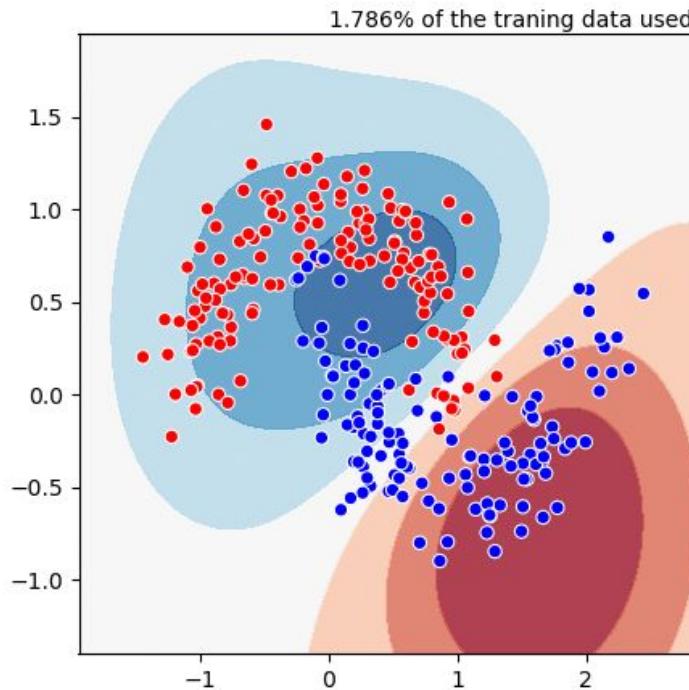
Imagine you are given the task of creating a rule based on this data.

bas	acc	feta	ute	ligh	sev	pro	ab	r	pe	m	hist	hist	hist	hi	hi	hi	hi	hi	hi	hi	hi	f
120.0	0.0	0.0	0.0	0.0	0.0	0.0	73.0	0.5	43.0	2.4	64.0	62.0	126.0	2.0	0.0	120.0	137.0	121.0	73.0	1.0	2.0	
132.0	0.006	0.0	0.006	0.003	0.0	0.0	17.0	2.1	0.0	10.4	130.0	68.0	198.0	6.0	1.0	141.0	136.0	140.0	12.0	0.0	1.0	
133.0	0.003	0.0	0.008	0.003	0.0	0.0	16.0	2.1	0.0	13.4	130.0	68.0	198.0	5.0	1.0	141.0	135.0	138.0	13.0	0.0	1.0	
134.0	0.003	0.0	0.008	0.003	0.0	0.0	16.0	2.4	0.0	23.0	117.0	53.0	170.0	11.0	0.0	137.0	134.0	137.0	13.0	1.0	1.0	
132.0	0.007	0.0	0.008	0.0	0.0	0.0	16.0	2.4	0.0	19.9	117.0	53.0	170.0	9.0	0.0	137.0	136.0	138.0	11.0	1.0	1.0	
134.0	0.001	0.0	0.01	0.009	0.0	0.002	26.0	5.9	0.0	0.0	150.0	50.0	200.0	5.0	3.0	76.0	107.0	107.0	170.0	0.0	3.0	
134.0	0.001	0.0	0.013	0.008	0.0	0.003	29.0	6.3	0.0	0.0	150.0	50.0	200.0	6.0	3.0	71.0	107.0	106.0	215.0	0.0	3.0	
122.0	0.0	0.0	0.0	0.0	0.0	0.0	83.0	0.5	6.0	15.6	68.0	62.0	130.0	0.0	0.0	122.0	122.0	123.0	3.0	1.0	3.0	
122.0	0.0	0.0	0.002	0.0	0.0	0.0	84.0	0.5	5.0	13.6	68.0	62.0	130.0	0.0	0.0	122.0	122.0	123.0	3.0	1.0	3.0	
122.0	0.0	0.0	0.003	0.0	0.0	0.0	86.0	0.3	6.0	10.6	68.0	62.0	130.0	1.0	0.0	122.0	122.0	123.0	1.0	1.0	3.0	
151.0	0.0	0.0	0.001	0.001	0.0	0.0	64.0	1.9	9.0	27.6	130.0	56.0	186.0	2.0	0.0	150.0	148.0	151.0	9.0	1.0	2.0	
150.0	0.0	0.0	0.001	0.001	0.0	0.0	64.0	2.0	8.0	29.5	130.0	56.0	186.0	5.0	0.0	150.0	148.0	151.0	10.0	1.0	2.0	
131.0	0.005	0.072	0.008	0.003	0.0	0.0	28.0	1.4	0.0	12.9	66.0	88.0	154.0	5.0	0.0	135.0	134.0	137.0	7.0	1.0	1.0	
131.0	0.009	0.222	0.006	0.002	0.0	0.0	28.0	1.5	0.0	5.4	87.0	71.0	158.0	2.0	0.0	141.0	137.0	141.0	10.0	1.0	1.0	
130.0	0.006	0.408	0.004	0.005	0.0	0.001	21.0	2.3	0.0	7.9	107.0	67.0	174.0	7.0	0.0	143.0	125.0	135.0	76.0	0.0	1.0	
130.0	0.006	0.38	0.004	0.004	0.0	0.001	19.0	2.3	0.0	8.7	107.0	67.0	174.0	3.0	0.0	134.0	127.0	133.0	43.0	0.0	1.0	
130.0	0.006	0.441	0.005	0.005	0.0	0.0	24.0	2.1	0.0	10.9	125.0	53.0	178.0	5.0	0.0	143.0	128.0	138.0	70.0	1.0	1.0	
131.0	0.002	0.383	0.003	0.005	0.0	0.002	18.0	2.4	0.0	13.9	107.0	67.0	174.0	5.0	0.0	134.0	125.0	132.0	45.0	0.0	2.0	
130.0	0.003	0.451	0.006	0.004	0.0	0.001	23.0	1.9	0.0	8.8	99.0	59.0	158.0	6.0	0.0	133.0	124.0	129.0	36.0	1.0	1.0	

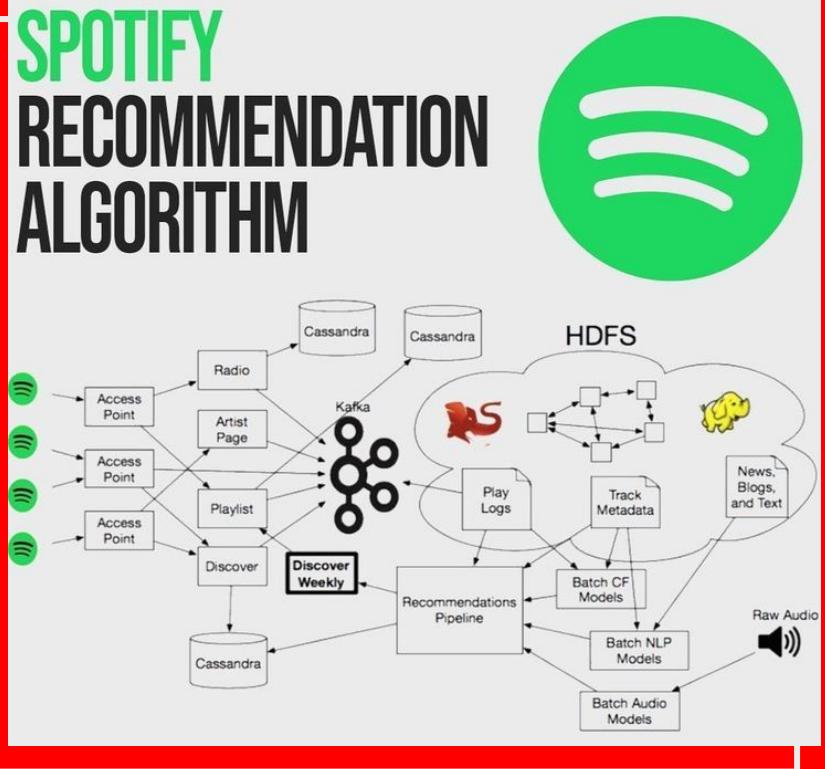
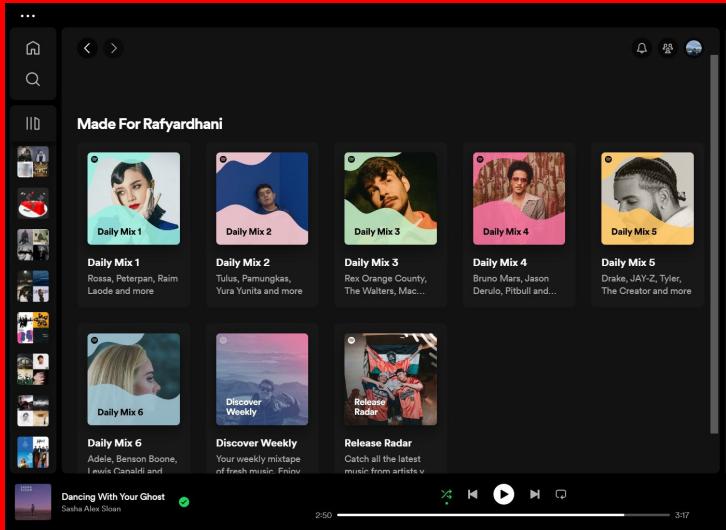
# Say Hello to Machine Learning!

With machine learning, we don't need to create a rules explicit.

They will find the patterns based on our data.



# Say Hello to Machine Learning!



# What is Machine Learning?

“Field of study that **gives computers the ability to learn** without being explicitly programmed.”

# Example: Predict Housing Price

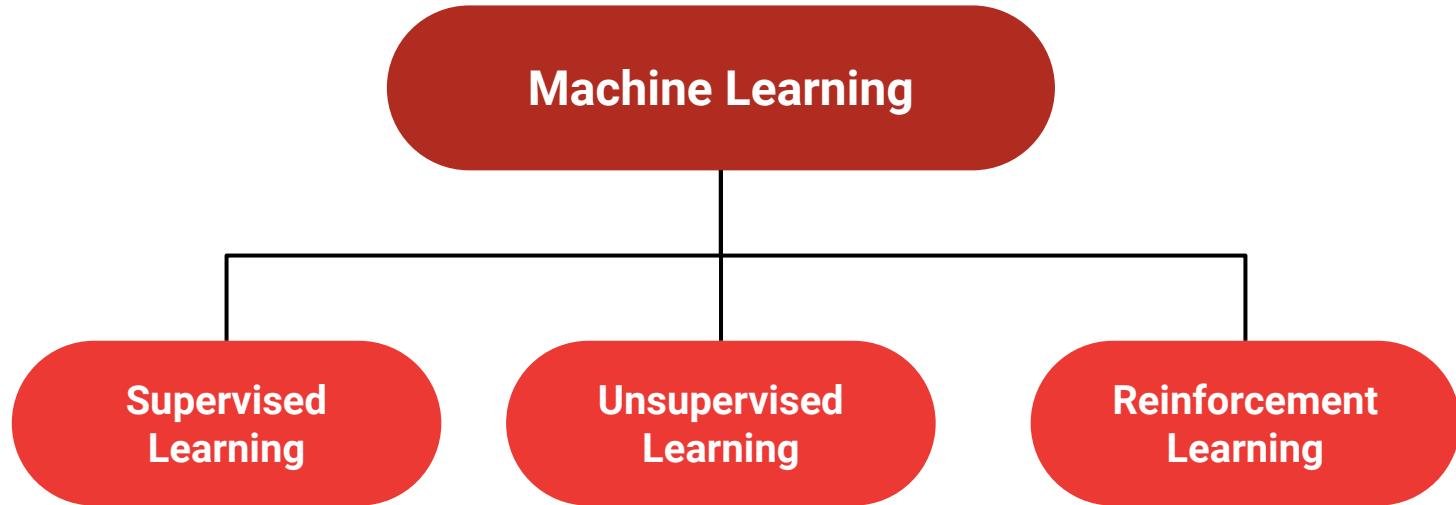


LA	BA	BD	BR	Price
1100	700	5	6	28.000.000.000
824	800	4	4	19.000.000.000
500	400	4	3	4.700.000.000
251	300	5	4	4.900.000.000
1340	575	4	5	28.000.000.000

# Why Do We Need Machine Learning?



# Types of Machine Learning



# Getting Started with Supervised Learning

# What is Supervised Learning?

- Supervised learning refers to algorithms that **learn how** to map **input X** to **output Y**.
- The critical characteristic of supervised learning is that we give learning algorithm **example data** and **include the right answers** to learn.
- The learning algorithm will **discover the connections** between the example data that produce the correct answers.





# Supervised

$X_1$	$X_2$	$X_p$	$Y$

Target

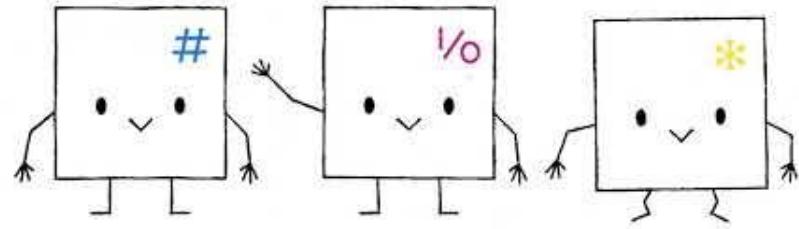


# Type of Supervised Learning

The most common use cases for supervised learning are **regression** and **classification**:

- Regression model predicts a **numeric value** from **infinitely many possible numbers**.
- Classification model predict the **likelihood** that something belongs to a **category**.

Regression      Binary Classification      Multiclass Classification



# Type of Supervised Learning

## Regression



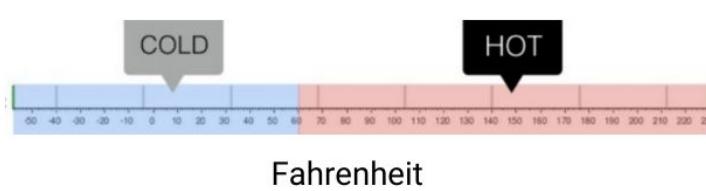
What will be the temperature tomorrow?



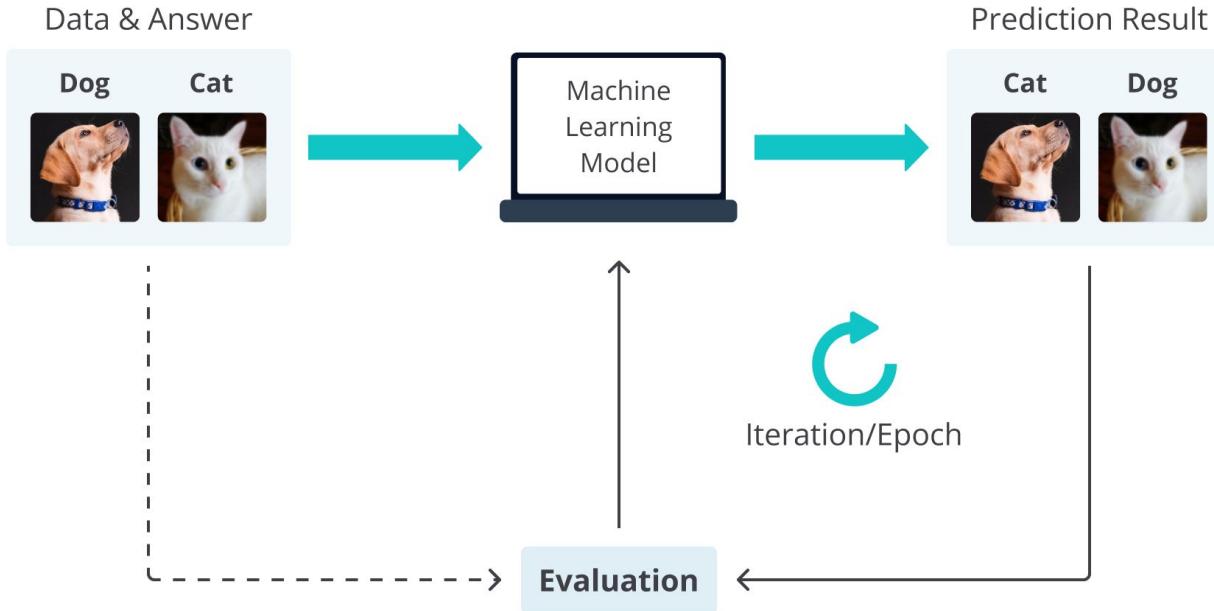
## Classification



Will it be hot or cold tomorrow?



# How Supervised Learning **Learn**?



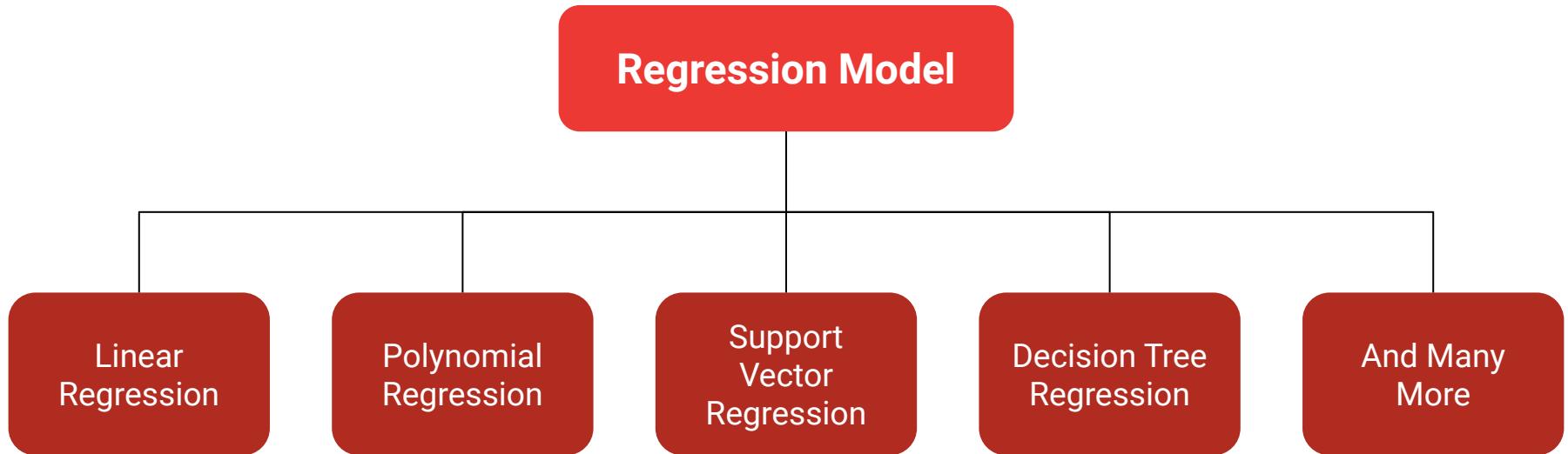
## Mini Quiz

Consider a scenario where you are tasked with predicting the precise amount of rainfall (in millimeters) for the upcoming day in a specific location within a city.

In the context of employing a machine learning algorithm for this prediction, would you categorize this problem as regression or classification, and what factors contribute to your decision?

# Regression Model

# Varieties of Regression Model



# Linear Regression Model

Linear regression is a **linear model** that assumes a **linear relationship** between the **input variables (x)** and the single **output variable (y)**.

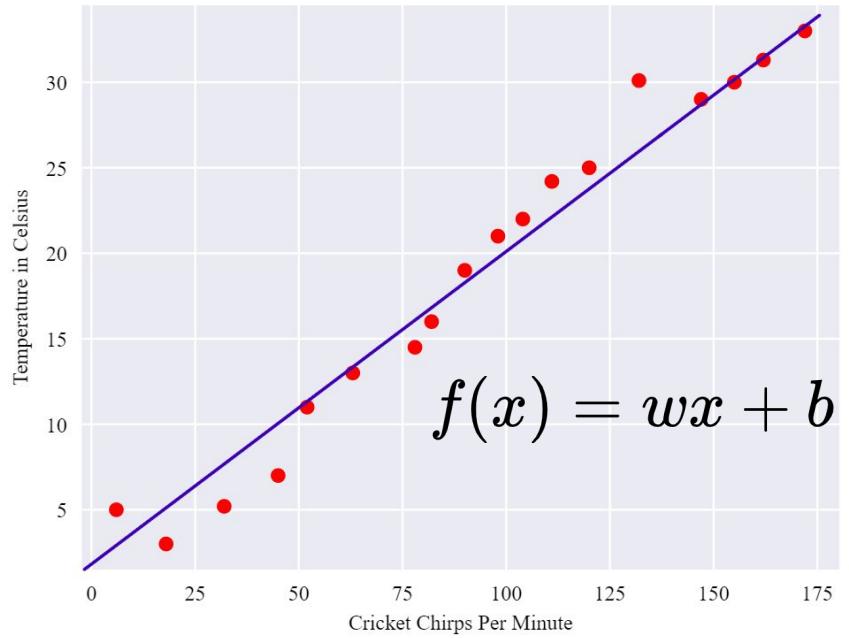
**Simple linear regression**

**Multiple linear regression**

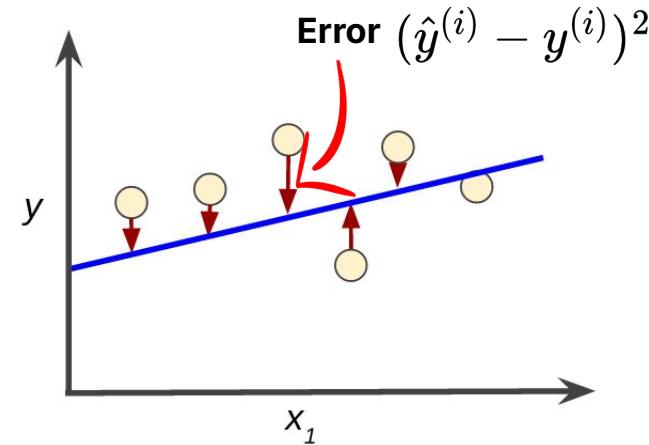
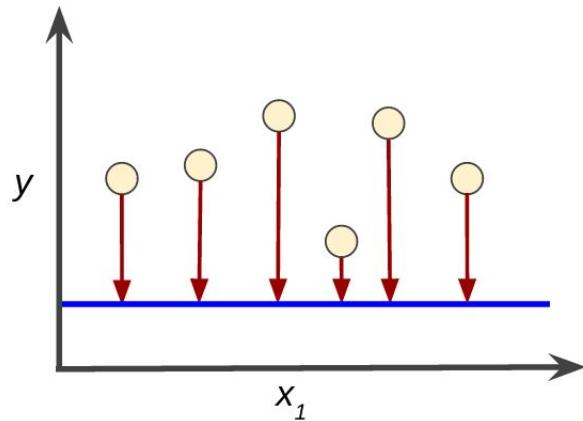


# Understand the Notation

- The  $x$  is the **input variabel**.
- The  $y$  is the **output or target variabel**.
- The  $\hat{y}$  is the **prediction result**.
- The  $f$  is the **hypothesis function** to make the predictions.
- The  $w$  and  $b$  is the **weight & bias** parameters.



# Cost Function



$$f(x^{(i)}) = \hat{y}^{(i)} = wx^{(i)} + b$$

$$J(w, b) = \frac{1}{2m} \sum_1^m (\hat{y}^{(i)} - y^{(i)})^2$$

# Gradient Descent

Gradient descent is an algorithm that you can use to try to **minimize** any function, including the **cost function**.

- Start with some  $w, b$
- **Update the parameters** to reduce the cost function.
- Until we get or near the **global minimum**.

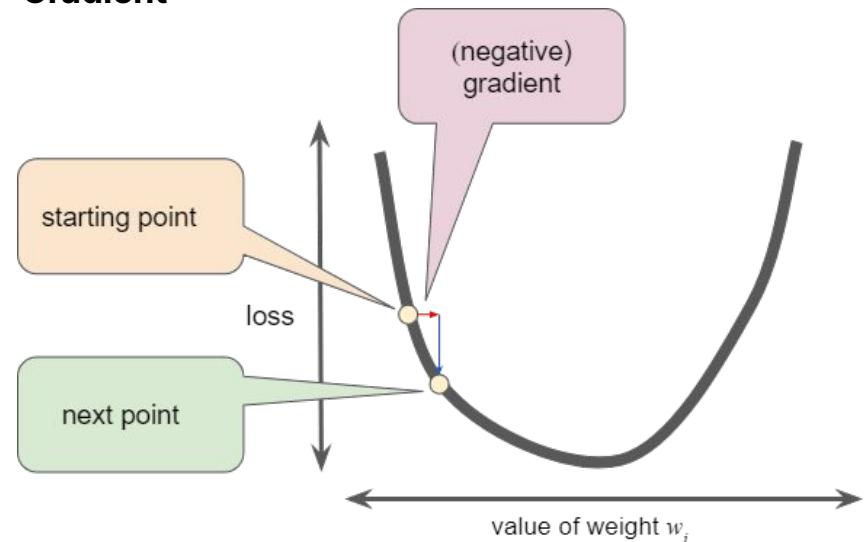


# Gradient Descent

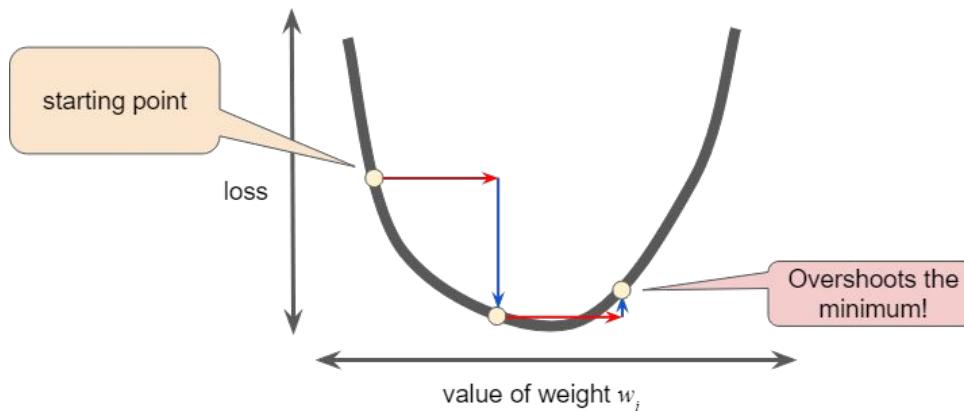
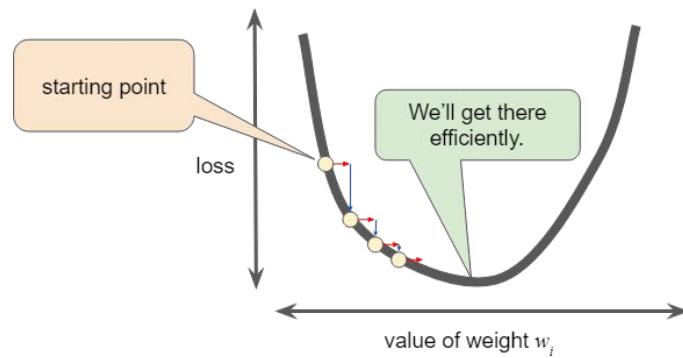
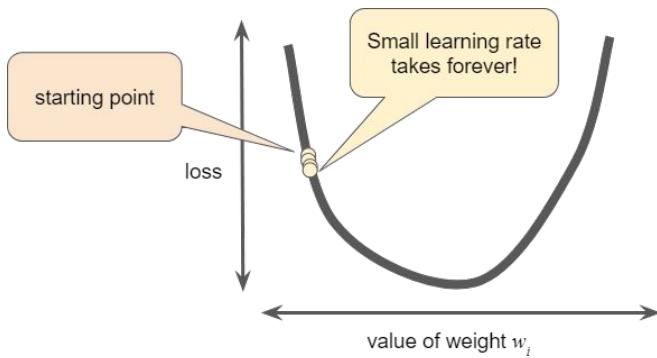
$$w_{new} = w - \alpha \frac{\partial}{\partial w} J(w, b)$$

Learning rate      Gradient

$$b_{new} = b - \alpha \frac{\partial}{\partial b} J(w, b)$$

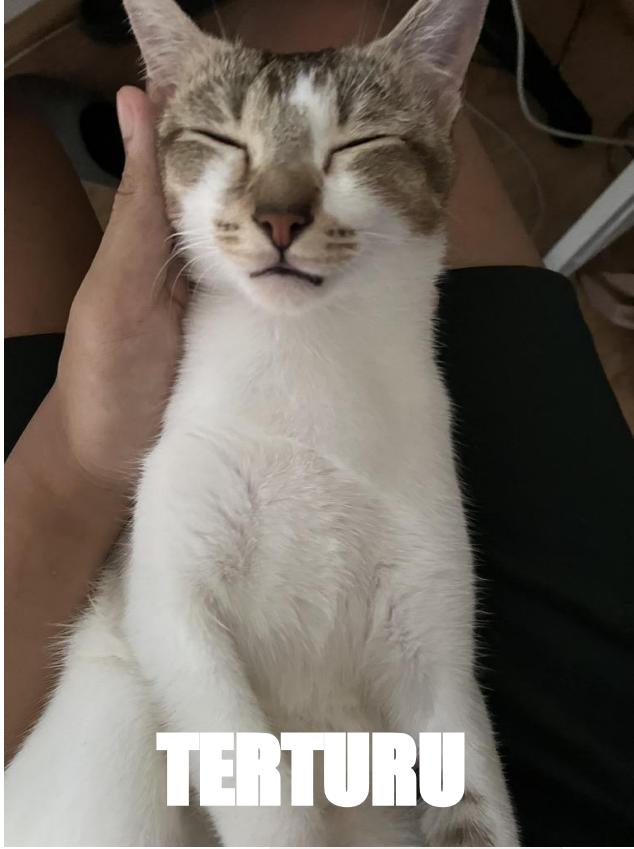


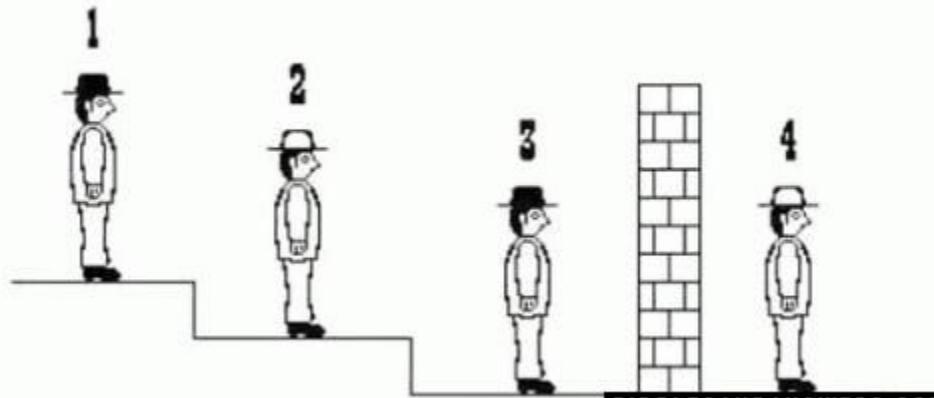
# How to Choose the Learning Rate?



# Diving into the World of Regression Algorithms

# Break Time!



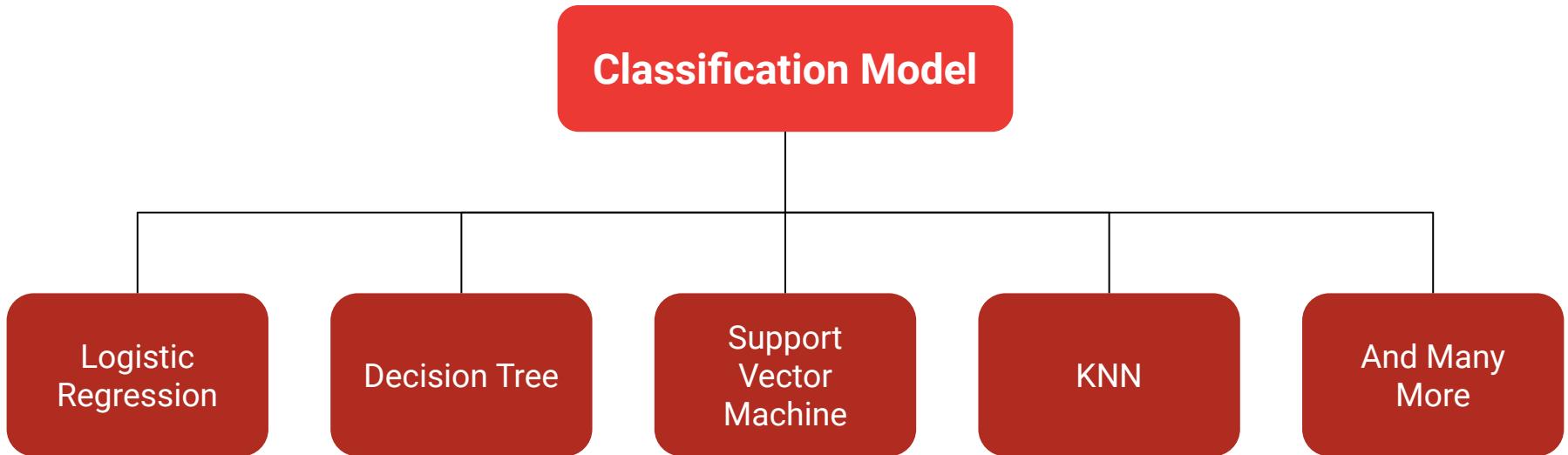


## Four Prisoner Hat

Can you solve this riddle?

# Classification Model

# Varieties of Classification Model

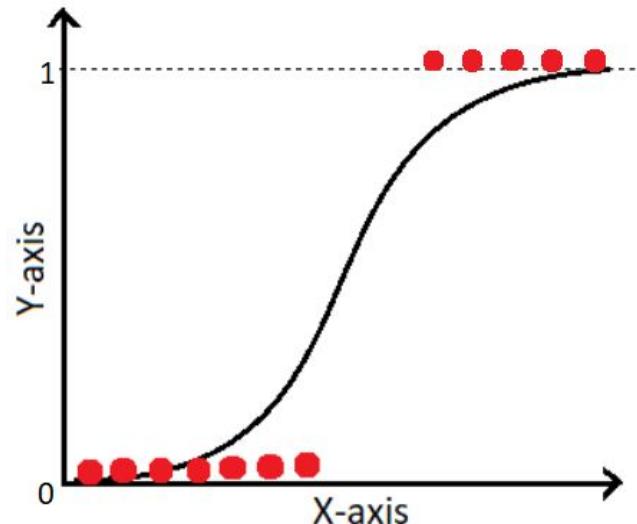


# Logistic Regression Model

Logistic regression is a **classification algorithm** that is used to predict the probability of a categorical dependent variable.

Typically used to model a **binary classification**

Used **sigmoid function**



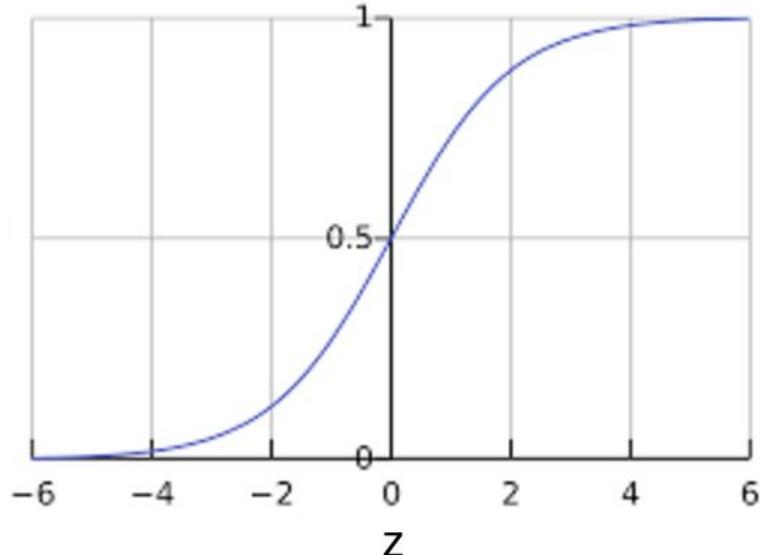
# Sigmoid Function

- Sigmoid is a mathematical function that takes **any real number** and **maps it to a probability between 1 and 0.**
- This function assists the logistic regression model to **squeeze the values from  $(-n,+n)$  to  $(0,1)$**



## Sigmoid Function

$$f(x) = wx + b = z \xrightarrow{\text{g(z)}} g(z) = \frac{1}{1+e^{-z}}$$



$$0 < g(z) < 1$$

# Evaluation Metrics for Classification

$$\text{Accuracy} = \frac{\text{TP} + \text{TN}}{\text{Total}}$$

The proportion of correct predictions to the total predictions made.

$$\text{Recall} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$

The proportion of positive data that is correctly predicted by the model.

$$\text{Precision} = \frac{\text{TP}}{\text{TP} + \text{FP}}$$

The proportion of correct positive predictions compared to total positive predictions.

$$\text{F1-Score} = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

Harmonic mean of precision and recall, useful when there is an imbalance between classes.

# **Classify with Style: An Exciting Voyage through Algorithmic Waters**

# Other Important Things

# Feature Scaling

Feature scaling is a method used to **normalize the range of independent variables or features of data.**

- **Min-Max scaling** is a scaling technique in which values are **shifted and rescaled between 0 and 1**.
- **Z-score normalization** is scaling technique where the values are **centered around the mean with a unit standard deviation**.

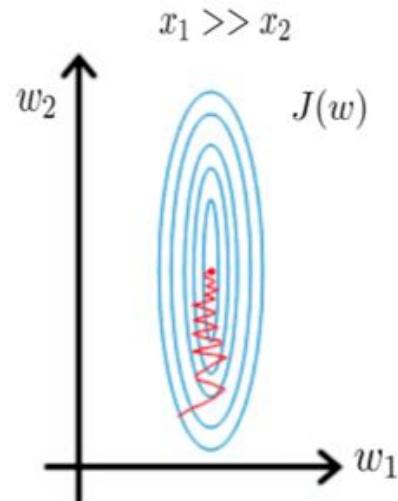


# Feature Scaling

$$x' = \frac{x - x_{\max}}{x_{\max} - x_{\min}}$$

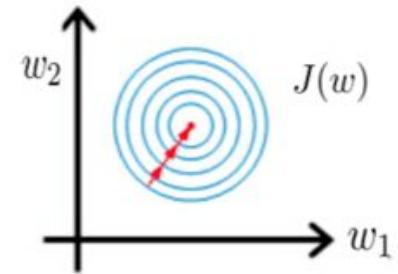
$$x' = \frac{x - \mu}{\sigma}$$

Gradient descent  
without scaling



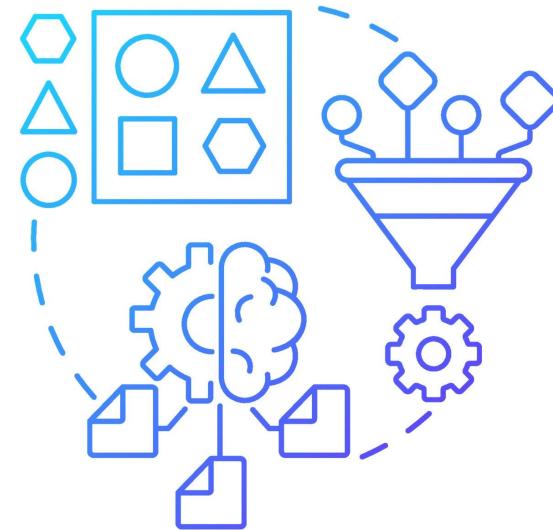
Gradient descent  
after scaling variables

$$0 \leq x_1 \leq 1$$
$$0 \leq x_2 \leq 1$$

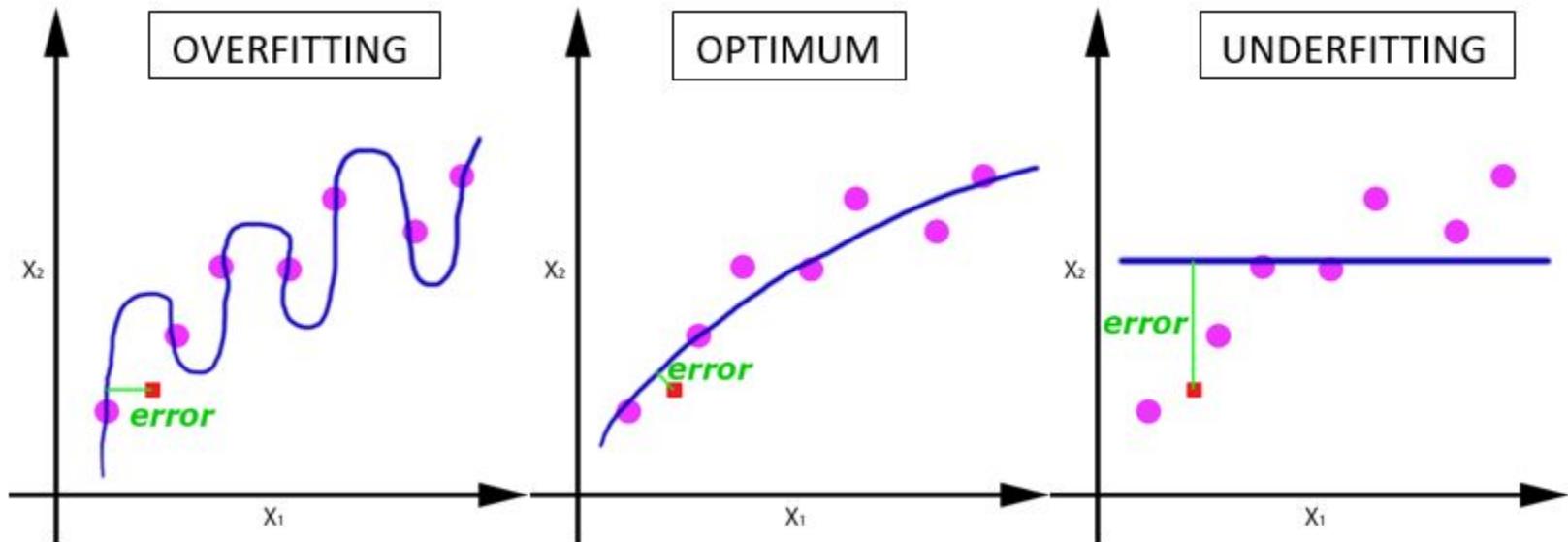


# Feature Engineering

- Feature engineering is a technique that **leverages data to create a new feature** that isn't in the training set.
- The goal is to **simplify** and **speed up** the training process while **enhancing model accuracy**.



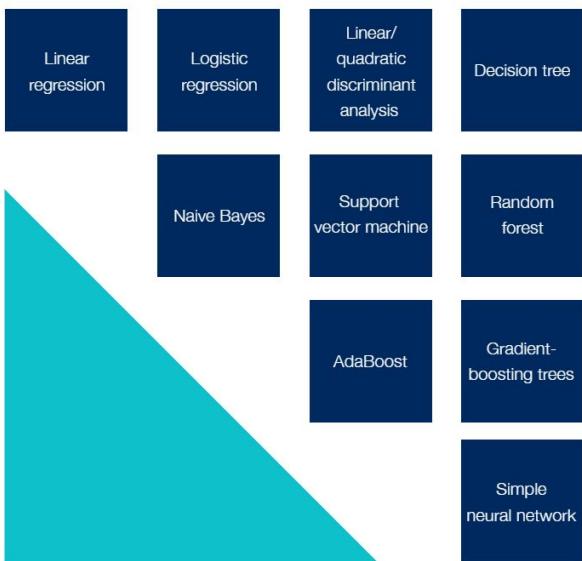
# Overfitting & Underfitting



# Industry Impact/Example

<https://www.mckinsey.com/capabilities/quantumblack/our-insights/an-executives-guide-to-ai>

## Supervised Learning



### Linear regression

Highly interpretable, standard method for modeling the past relationship between independent input variables and dependent output variables (which can have an infinite number of values) to help predict future values of the output variables

**Business use cases**

- Understand product-sales drivers such as competition prices, distribution, advertisement, etc
- Optimize price points and estimate product-price elasticities

### Decision tree

Highly interpretable classification or regression model that splits data-feature values into branches at decision nodes (eg, if a feature is a color, each possible color becomes a new branch) until a final decision output is made

**Business use cases**

- Understand product attributes that make a product most likely to be purchased
- Provide a decision framework for hiring new employees

# Industry Impact/Example

<https://www.mckinsey.com/capabilities/quantumblack/our-insights/an-executives-guide-to-ai>

## Reinforcement Learning



Optimize the trading strategy for an options-trading portfolio



Balance the load of electricity grids in varying demand cycles



Stock and pick inventory using robots



Optimize the driving behavior of self-driving cars



Optimize pricing in real time for an online auction of a product with limited supply

# Discussions



**NOOB BERTANYA,  
LORD MENJAWAB**

# Quiz and Feedback

BINTANG 5 PLIS



OH TUHAN



BERI HAMBA

# Conclusion

You have learned **basic concepts** in  
**machine learning**, how **regression**  
and **classification models** work

These knowledges will give you a solid  
fundamental concept to **becoming a**  
**machine learning practitioner**



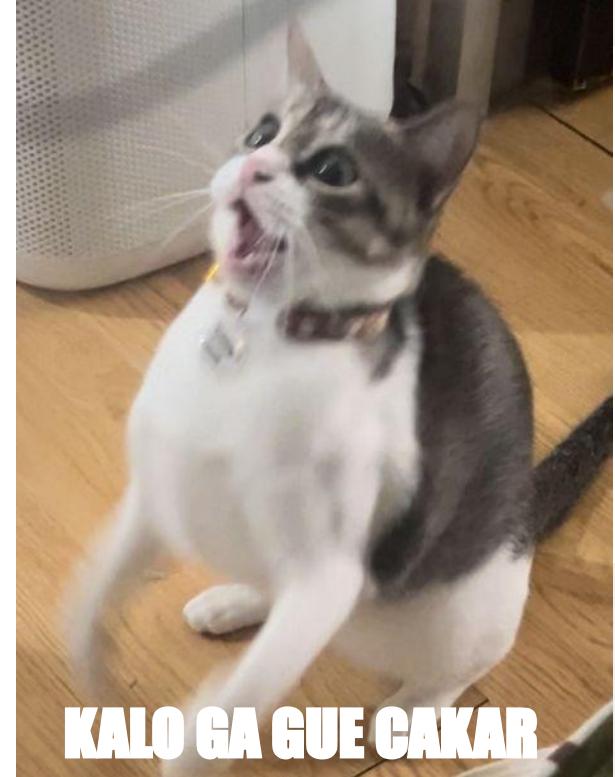
# Thank You



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 bangkit



KALO GA GUE CAKAR

LU