

AI in Ramadan

Session 1 :
Beginner's guide

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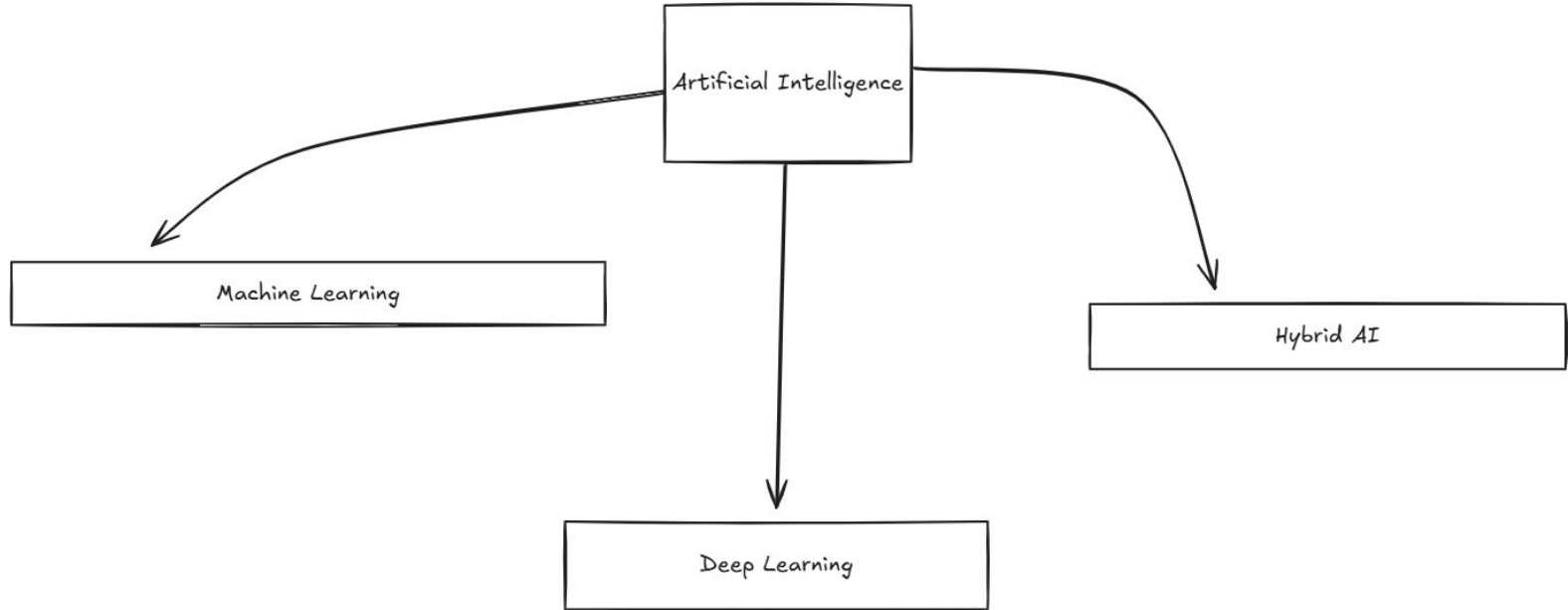
Introduction

A brief history

Introduction

Artificial Intelligence (AI) is the field of computer science focused on creating machines that can perform tasks requiring **human-like intelligence**, such as learning, reasoning, and problem-solving.

AI approaches



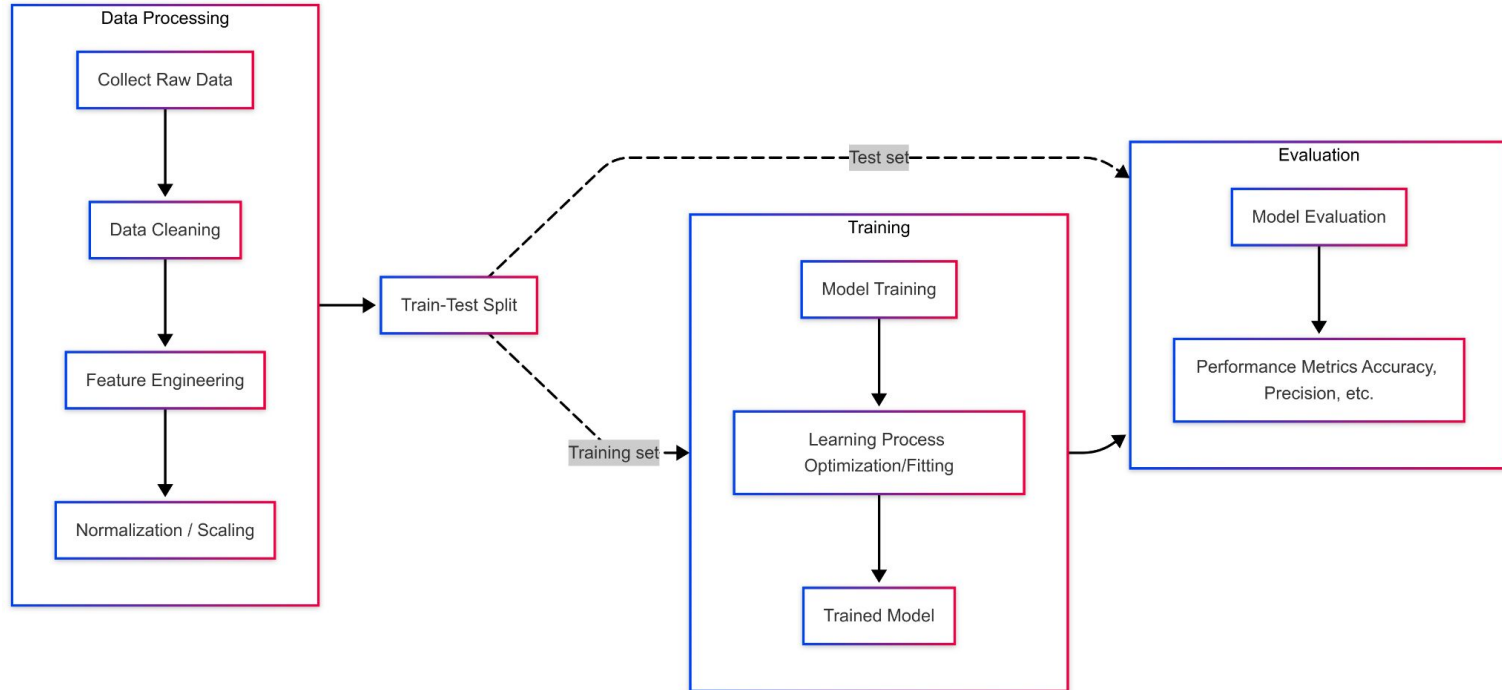
Brief History

- 1950-1960 : The birth of AI as a research field
- 1960-2000 : Formalization of machine learning algorithms, the boom of back-propagation and the introduction and the fall of neural networks
- 2000-2006: The focus shifts to Machine learning approaches and optimization algorithms
- 2006-present: Deep Learning rebirth, AlexNet boom and the emergence of the Transformer architecture

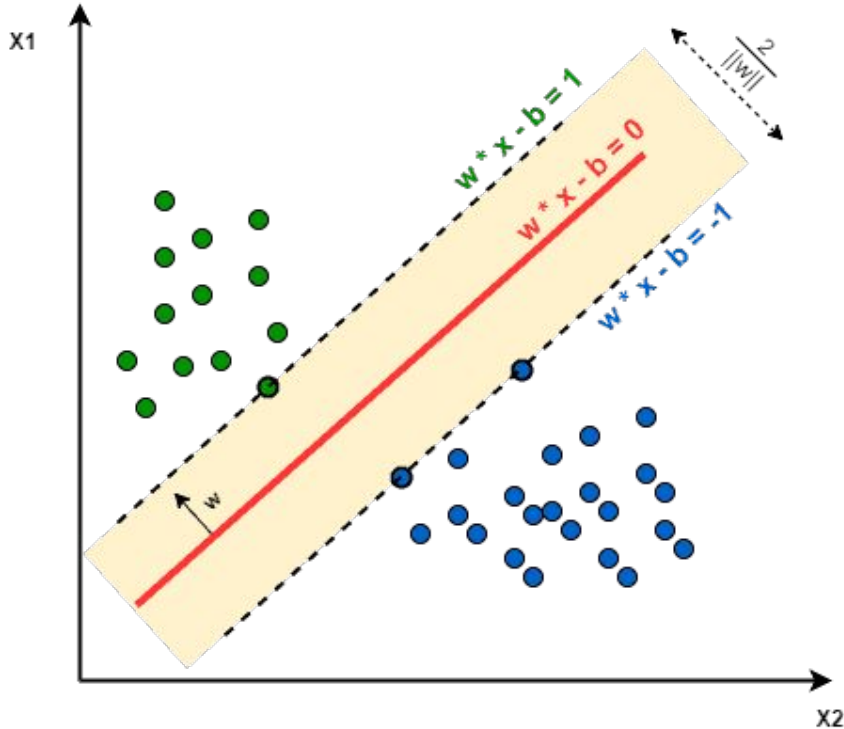
ML models

An overview

AI workflow



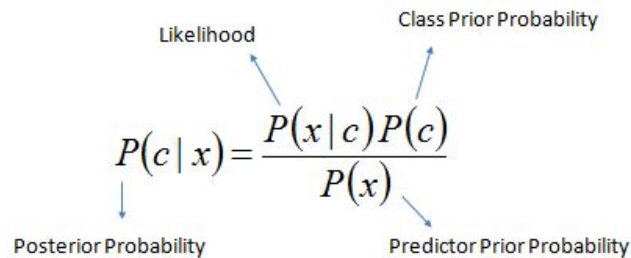
Support Vector Machine (SVM)



Support Vector Machine (SVM) is a supervised learning model that aims to find the best boundary—or **hyperplane**—that separates data into different classes. It does this by **maximizing** the **margin** between the **closest points** of each class

Naive Bayes

Naive Bayes is a simple yet effective classification technique based on probability theory and Bayes' theorem. It assumes that the features in a dataset are independent of each other, which simplifies the computation and makes the algorithm very fast.



A diagram showing the components of Bayes' theorem. The equation $P(c|x) = \frac{P(x|c)P(c)}{P(x)}$ is centered. Four blue arrows point from labels to parts of the equation: 'Likelihood' points to $P(x|c)$, 'Class Prior Probability' points to $P(c)$, 'Posterior Probability' points to $P(c|x)$, and 'Predictor Prior Probability' points to $P(x)$.

$$P(c|x) = \frac{P(x|c)P(c)}{P(x)}$$

Likelihood

Class Prior Probability

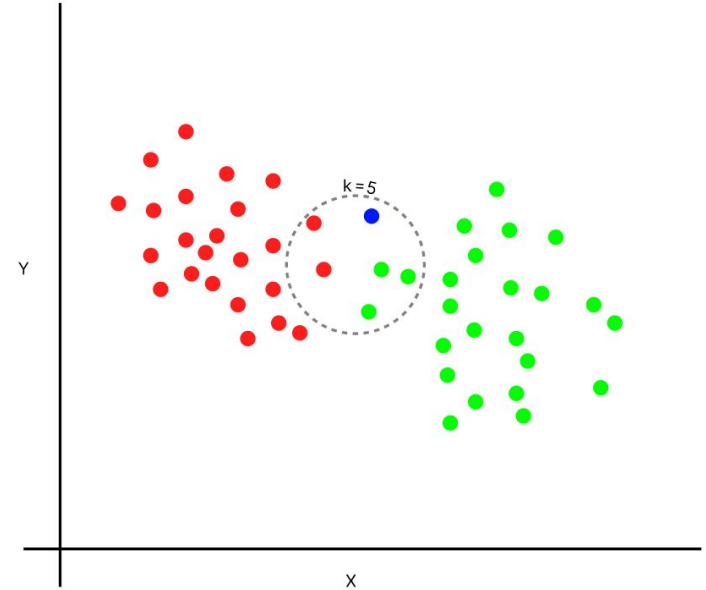
Posterior Probability

Predictor Prior Probability

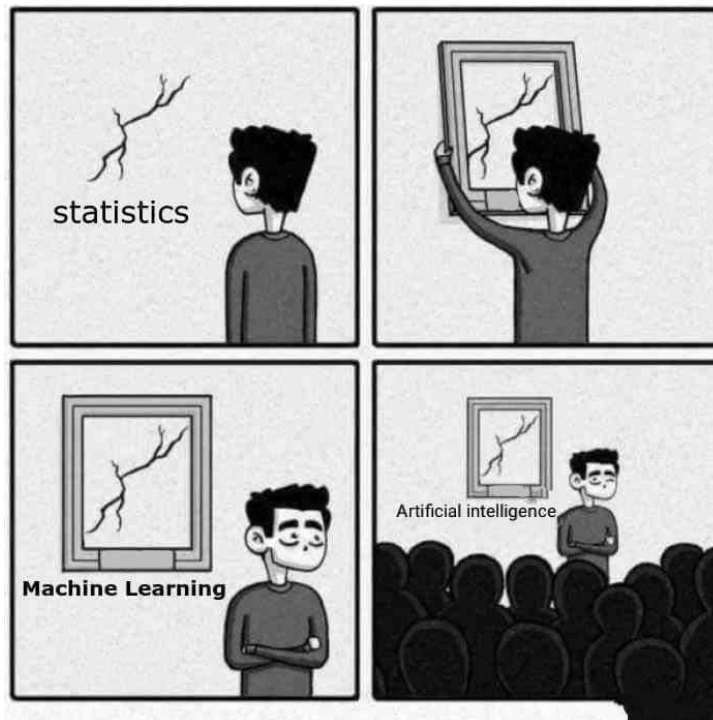
$$P(c|X) = P(x_1|c) \times P(x_2|c) \times \dots \times P(x_n|c) \times P(c)$$

K- Nearest Neighbour

K-Nearest Neighbors (KNN) is a straightforward and intuitive classification algorithm that classifies new data points based on the **closest** examples in the training data. It works by looking at the '**k**' **nearest neighbors** of a new data point, and assigning the **most common** class among those neighbors.



Sooooo, basically



Metrics

$$Accuracy = \frac{T_{pos} + T_{neg}}{T_{pos} + T_{neg} + F_{pos} + F_{neg}} \quad (4)$$

$$Recall = \frac{T_{pos}}{T_{pos} + F_{neg}} \quad (5)$$

$$Precision = \frac{T_{pos}}{T_{pos} + F_{pos}} \quad (6)$$

$$F_1 - Score = 2 * \frac{Precision * Recall}{Precision + Recall} \quad (7)$$

Where T_{pos} is the number of true positives, T_{neg} the number of true negatives, F_{pos} the number of false positive and F_{neg} the number of false negatives.

		Actual Values	
		Positive (1)	Negative (0)
Predicted Values	Positive (1)	TP	FP
	Negative (0)	FN	TN

Let's collab



Questions?

See you
next time

