

# \* Machine Learning \*

## ① Artificial Intelligence (AI) :-

Artificial + Intelligence  
/                      \  
(Human / non-natural)      (Ability to understand /  
made thing.                      think)

study of how to train computers, so that comp can do things which humans can do.

↳ Adding human abilities to machines.

## ② Machine Learning (ML) :-

↳ Learning in which m/c can learn on its own, w/o explicitly programmed.

↳ Appl<sup>n</sup> of AI - provide system ability to automatically learn & improve from experience.

Learn from experience (E) w.r.t. some class of task (T) & ~~perfo~~ performance (P), if learner's performance at the task as measured by P improves by experiences.

## ① Difference bet<sup>n</sup> AI & ML :-

<u>Artificial Intelligence</u>	<u>Machine Learning</u>
① AI first used in 1956, by John McCarthy, hosted AI first conference.	① First used in 1952 by IBM comp scientist Arthur Samuel.
② Ability to acquire & apply knowledge.	② Acquisition of knowledge / skills.
③ contains ML & DL as subsets.	③ <del>AI</del> ML is subset of AI.
④ <u>Aim</u> - ↑ chance of success, not accuracy.	④ ↑ accuracy, not care about success.
⑤ Develop system able to perform complex tasks like decision-making.	⑤ Construct m/c able to perform tasks trained to them.
⑥ <u>Goal</u> - Simulate natural intelligence to solve the complex problem.	⑥ <u>Goal</u> - Learn from data on certain task to maximize performance.
⑦ Develop systems that mimics humans to solve problems.	⑦ Involves creating self-learning algorithms.
⑧ Find optimal solution.	⑧ Find a solution, whether optimal or not.



## Artificial Intelligence

⑨ Broad categories of AI:

- ↳ Narrow Intelligence
- ↳ General Intelligence
- ↳ Super Intelligence.

⑩ AI work with structured, semi-structured & unstructured data.

⑪ Uses of AI -

- ↳ Siri, chatbots.
- ↳ Expert Systems.
- ↳ Google Translate.
- ↳ Humanoid robots like Sophia.

⑫ Perform tasks like -

- ↳ Natural lang processing
- ↳ Recognize images.
- ↳ Making decisions
- ↳ Solve complex problems.

## Machine Learning

⑨ Broad categories of ML:

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

⑩ ML works with the structured & semi-structured data only.

⑪ Uses of ML -

- ↳ Google search algo.
- ↳ Bank fraud analysis
- ↳ Stock price forecast.
- ↳ Online recommendation system - e-commerce.

⑫ Train algos on data to -

- ↳ make decisions
- ↳ Predictions
- ↳ Recommendations.

## ● Machine Learning:-

↳ Enable computers to learn automatically from past data.

↳ ML uses algos for building math models & make predictions using historical data / info.

↳ Used for performing tasks like - image, speech recognition, speech recognition, email filtering, Facebook auto-tagging, recommenda<sup>n</sup> systems.

Machine has ability to learn, if it can improve its performance by gaining more data.

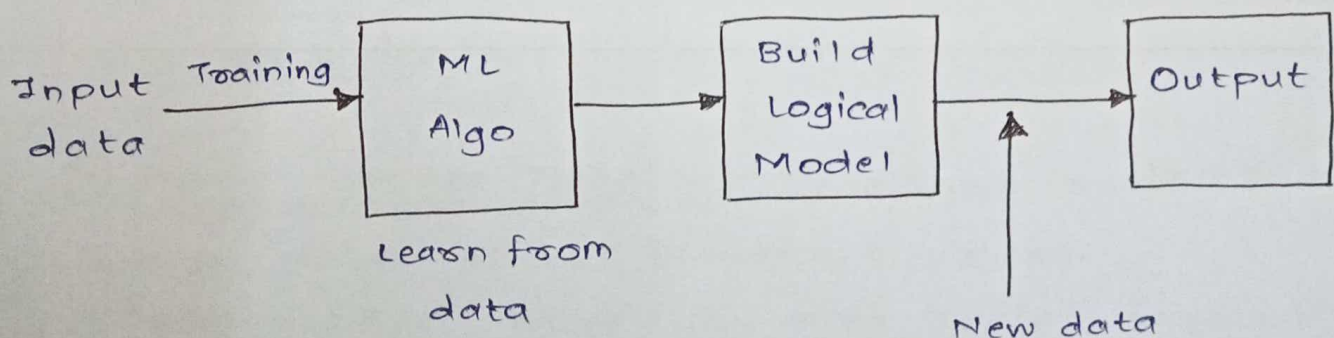
### ● How ML works —

① Learn from historical data.

② Build prediction models

③ Then receive a new data.

④ Predict o/p for new data.





## ① Applications of Machine Learning:-

① Image Recognition - classify images between cat & dog & evolve to Face Recognition and real-world use cases like - employee attendance.

② Speech Recognition - smart systems like Alexa & Siri, used for communication, convert voice instructions to text. Google speech searches by speaking.

③ Recommendation System - Analyze user preferences & search history  $\Rightarrow$  recommend content/service to them.

Ex. YouTube recommend videos. Netflix recommend movies & series.

④ Fraud Detection - detect fraud transac<sup>n</sup> & fraud activities.

⑤ Self Driving cars - cars drive without drivers.

Ex. Tesla cars.

⑥ Medical Diagnosis - perform health-related tasks. Detect disease & cure for that diseases.

Ex. Breast cancer classification, Parkinson's Disease classification & Pneumonia Detection.

⑦ Stock Market Trading - Intelligent systems predict future price trends & market values.

↳ Application based on - time series forecasting.

## ① Limitations of Machine Learning:-

① Data Quality - Ensure data accurate, complete & representative of problem-domain.

↳ Inaccurate data  $\Rightarrow$  inaccurate / biased models.

② Data Quantity - Large data unavailability, esp complex problems, require large data.

③ Bias & fairness - Sometimes bias & discrimination in training  $\Rightarrow$  unfair outcomes (minorities).

④ Overfitting & Underfitting -

Overfitting: Model too complex & fits training data very well, poor result for new data.

Underfitting: Model too simple & cannot capture all relevant patterns in data.

⑤ Privacy & Security - Interact sensitive info of people / organization  $\Rightarrow$  Raise privacy & security concern.

⑥ Interpretability - Some models like deep neural n/w, difficult interpret & understand, challenge for predictions & decisions.



## ① Types of Machine Learning:-

<u>Points</u> Aim	<u>Supervised</u> Calculate outcom.	<u>Unsupervised</u> Discovers patterns.	<u>Reinforcement</u> Learn series of action.
Input data	Labelled	Unlabelled.	Not pre-defined.
Problem	Learn pattern in i/p & their labels.	Divide data into classes.	Find best reward bet <sup>n</sup> start & end state.
Solution	Find mapping eq <sup>n</sup> on i/p data & its labels.	Find patterns in i/p & classify.	Maximize result by results of actions.
Model Building	First, build & train, then test.	Build & train, then test.	Train & test simultaneously.
Applications	classification & Regression problems.	clustering & association problems.	Deals with exploration & exploitation problems.
Algorithms used	<ul style="list-style-type: none"> <li>• Linear Regress<sup>n</sup>.</li> <li>• Decision tree.</li> <li>• K-nearest neighbors.</li> </ul>	<ul style="list-style-type: none"> <li>• k-means clustering.</li> <li>• k-medoids clustering</li> <li>• agglomerative clustering.</li> </ul>	<ul style="list-style-type: none"> <li>• Q-Learning.</li> <li>• SARSA</li> <li>• Deep Q n/w.</li> </ul>
Examples	<ul style="list-style-type: none"> <li>• Img Detec<sup>n</sup>.</li> <li>• Population Growth Detec<sup>n</sup>.</li> </ul>	<ul style="list-style-type: none"> <li>• Cust segmenta<sup>n</sup>.</li> <li>• Feature elicit<sup>n</sup>.</li> <li>• Targeted market.</li> </ul>	<ul style="list-style-type: none"> <li>• Driverless car.</li> <li>• self-navigating cleaners.</li> </ul>

# 1) Supervised Learning:-

↳ Teach / train m/c using well-labelled data.

↳ Then, new set of data provided to analyze training data & produce correct outcome.

↳ M/c learn from training data & then applies knowledge to test data.

## • 2 categories of algo:

① Classification - Outcome variable is categorical.

Ex. Red / Blue, Yes / No.

② Regression - Output variable is a real value.

Ex. Price, weight, population.

## • Advantages -

① Optimize performance with experience.

② Classification & Regression tasks.

③ Estimate / map result to new sample.

## Limitations -

① Big data classified challenge.

② Labelled data training large time required.

③ Require labelled data.

④ Cannot handle complex task.

## • Algorithms -

① Regression

② Logistic Regression

③ Classification

④ Naive Bayes classifiers

⑤ K-NN (K nearest neighbors)

⑥ Decision Tree

⑦ Support vector Machine (SVM)



## 2) Unsupervised Learning:-

↳ Training m/c using info not labelled/classified & allow act on them w/o guidance.

↳ Find underlying patterns in data without any prior training.

↳ Mainly deal with unlabelled data.

### • 2 categories -

① clustering - Discovers inherent groups in data, like grouping customers by purchasing.

② Association - Describe rules that describe large data, ex if customer buy X  $\Rightarrow$  also tend to buy Y.

### • clustering -

- ① Exclusive (partitioning)
- ② Agglomerative
- ③ Overlapping
- ④ Probabilistic.

### • Clustering Algos -

- ① Hierarchical clustering
- ② K-means clustering
- ③ Principal Component Analysis
- ④ Singular Value Decomposition
- ⑤ Independent Component Analysis

### • Advantages -

- ① No training data labelled.
- ② Dimensionality reduction.
- ③ Find unknown patterns.
- ④ Flexible - Apply on wide range of problems - cluster, anomaly detec<sup>n</sup> & association.
- ⑤ Exploration - Explore data & find patterns.
- ⑥ Low cost - Less expensive, because unlabelled data. Label - time & costly.

### • Limitations -

- ① Lack measure accuracy.
- ② No predefined answers.
- ③ Time to interpret & label classes in classific<sup>n</sup>.
- ④ Lack guidance - No guidance & feedback, difficult know patterns relevant & useful.
- ⑤ Scalability - complex problems, computationally expensive.
- ⑥ Sensitivity to data quality - Data include missing values, outliers & noise.