- What is Machine Learning
- Traditional Programming vs Machine Learning Simple Note
  - In a second of the second of th
  - Machine Learning
  - # In Short
- Al vs ML vs DL Simple Notes + Examples + Venn Diagram
  - AI Artificial Intelligence
  - ML Machine Learning

  - Set Venn Diagram
  - # In Short
- Types of Machine Learning Notes with Examples & Visuals
  - 1 Supervised Learning
  - Unsupervised Learning
  - Reinforcement Learning
  - Summary Table



## 👺 What is Machine Learning

Machine Learning is a way of teaching computers to learn from data — just like we humans learn from experience Instead of programming every step, we give machines lots of data and let them figure out patterns on their own.

### Some Real Life Examples of ML:

- YouTube Recommendations → Learns from your watch history
- Spam Detection in Gmail → Learns patterns in spam emails
- Voice Assistants (Siri, Alexa) → Learn how you speak
- Self-driving Cars → Learn to identify stop signs, pedestrians, and roads
- Face Unlock on Phones → Learns to recognize your face

# Traditional Programming vs Machine **Learning** — Simple Note



## Traditional Programming

- you the result (output)
- Process:
  - Programmer writes explicit logic
  - Follows fixed instructions
- Example: If age < 18 → label as "minor" If age ≥ 18 → label as "adult"</li>



## **Machine Learning**

- computer learns and creates the rules (model)
- Process:
  - No need for manual logic
  - Learns patterns and relationships from data automatically
- figures out the rule by itself



### In Short

- Traditional Programming: Rules + Data → Result
- Machine Learning: Data + Result → Rules (Model)

# Al vs ML vs DL — Simple Notes + Examples + Venn Diagram



- Definition: The broad field of making machines smart (mimic human intelligence) and reasoning).
- Key Point: Al = Any system that mimics human intelligence
- Scope: The "big umbrella" includes rule-based systems, logic, planning, as well as ML and DL.

### **Examples:**

- Playing chess like a human (Al chess bots)
- Talking to Alexa or Siri (voice assistants)
- Self-driving cars (autonomous navigation)
- Google Translate (language conversion)



## 🧗 ML — Machine Learning

- Definition: A subset of Al where machines learn from data and improve themselves over time (without being explicitly programmed for every task).
- Key Point: ML = Al that learns from data
- **Scope:** A branch inside Al. Focuses on algorithms that find patterns in data.

### **Examples:**

- YouTube recommending videos based on your watch history
- Netflix predicting your next favorite show
- Gmail filtering spam emails
- Credit card fraud detection



## 拳 DL — Deep Learning

- Definition: A subset of Machine Learning that uses neural networks (inspired by the human brain) to handle very complex patterns and big data (images, speech, etc.).
- Key Point: DL = ML using neural networks for big, complex data

• Scope: A specialized area within ML, excels at tasks like image, sound, and language understanding.

### **Examples:**

- Face recognition unlock on phones
- ChatGPT and other chatbots
- Self-driving car vision (image/video analysis)
- Real-time language translation (speech-to-speech)



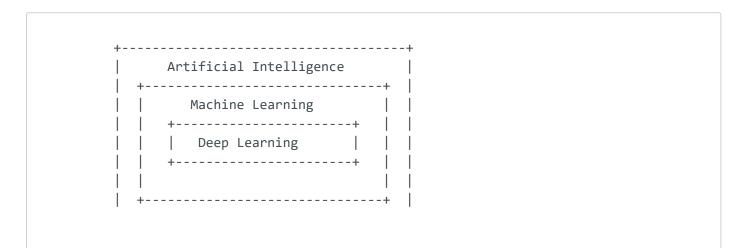
## **6** Summary Table

	Al	ML	DL
ls a	Field	Subset of Al	Subset of ML
Learns?	Not always	Yes, from data	Yes, via neural networks
Examples	Chess, Alexa, self- driving	YouTube, Netflix, Gmail	Face ID, ChatGPT, car vision
Key Idea	Mimics human intelligence	Learns from data	Learns via neural networks



## Set Venn Diagram

Below is a text-art representation. For beautiful diagrams, you can use tools like draw.io, Canva, or markdown with embedded images.



- Everything inside the largest rectangle is Al.
- ML is a subset of Al. (Focuses on Algorithm)
- **DL** is a subset of ML. (Complex Technique and algorithm)

## In Short

- AI: The big picture making machines smart.
- ML: The brain that learns from data (inside AI).
- Deep Learning: The super-powered brain (inside ML) for really tough, big-data problems!

# Types of Machine Learning — Notes with Examples & Visuals

# Supervised Learning

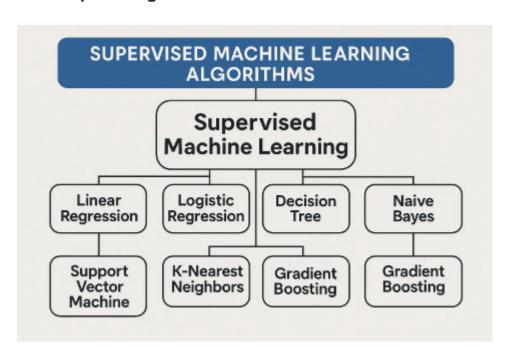
- What is it? Like a student learning from a teacher. The machine gets both input data and the correct answers (labels), and learns to predict outcomes.
- How it works:
  - o Both input and output variables are provided during training.
  - The model learns the relationship and can predict the output for new inputs.

Income(\$)	Credit Score	Loan
40,000	750	Yes
25,000	600	No
50,000	800	Yes
30,000	580	No

\*\*Example Table: \*\* </br>

Here, "Income" and "Credit Score" are inputs; "Loan" (Yes/No) is the output (label).

### • Popular Algorithms:



### • Real-life Examples:

- Email spam detection (input: email text, output: spam/not spam)
- Loan approval (input: applicant info, output: approve/deny)
- Image recognition (input: image, output: label)

# Unsupervised Learning

What is it? The machine gets input data only—there are no correct answers
provided. It tries to find patterns, group similar things, or detect outliers.

#### · How it works:

- The algorithm finds structure in data (like clustering).
- No labels or answers are given.
- What is Clustering? Clustering is about grouping similar data points together without knowing group labels in advance.
  - Imagine a scatterplot of dots: clustering draws circles around groups of dots that are close together.
  - For example, grouping customers based on their shopping habits, when you don't know categories ahead of time.

### Clustering Algorithms:

- K-Means Clustering
- Hierarchical Clustering
- DBSCAN

### Clustering Real-life Examples:

- Grouping customers by buying habits
- Organizing news articles into topics
- Detecting fraud (spotting unusual patterns)
- Photo apps grouping faces automatically

### Other Unsupervised Examples:

- Dimensionality reduction (PCA)
- Anomaly detection

# Reinforcement Learning

- What is it? Like training a dog: reward good behavior, discourage bad. The agent learns by trial and error, receiving feedback (rewards or penalties).
- How it works:

- Takes actions in an environment.
- Gets feedback (reward/penalty) and learns the best strategy over time.

### Real-life Examples:

- Game playing (chess, Go, video games)
- Robotics (robot learning to walk)
- Self-driving cars (learning to navigate)
- Recommender systems (learning best suggestions)



## Summary Table

Туре	Data Used	Goal	Example
Supervised Learning	Labeled	Predict output	Email spam detection, loan approval
Unsupervised Learning	Unlabeled	Find patterns/groups	Customer clustering, anomaly detection
Reinforcement Learning	Feedback	Maximize reward	Chess, robotics, self- driving cars

### **Key Points:**

- Supervised Learning: Learn from examples with answers (labels).
- Unsupervised Learning: Find structure in data without answers.
  - o Clustering is a main technique here: grouping data into clusters when you don't know the group labels in advance!
- Reinforcement Learning: Learn by trial and error, getting rewards or penalties.