Machine Learning Training Session Outline

1. Introduction to Machine Learning

What is Machine Learning?

- Machine Learning (ML) is a subset of Artificial Intelligence (AI) that enables systems
 to learn from data and improve their performance without being explicitly
 programmed.
- Instead of writing hard-coded rules, we train models on data to make predictions or decisions.

Why Machine Learning?

- Automation: ML automates decision-making (e.g., spam detection, recommendation systems).
- Handling Complex Data: Finds patterns in large datasets where manual analysis is impossible.
- Adaptability: Models improve over time with new data.

Example Applications:

- **Supervised Learning:** Email spam detection, house price prediction.
- **Unsupervised Learning:** Customer segmentation, anomaly detection.
- **Reinforcement Learning:** Self-driving cars, game-playing AI (AlphaGo).

2. Types of Machine Learning

- 1. Supervised Learning (Learning from Labeled Data)
- The model learns from **input-output pairs** (X → y).

- **Goal:** Predict the output for new, unseen data.
- Types:
- Regression (Continuous Output): Predict house prices, stock prices.
- o Classification (Discrete Output): Spam vs. not spam, image recognition.

2. Unsupervised Learning (Finding Patterns in Unlabeled Data)

- The model learns **hidden structures** from data without labels.
- **Goal:** Discover patterns, groupings, or anomalies.
- Types:
- Clustering: Grouping similar customers (e.g., K-Means).
- Dimensionality Reduction: Reducing features (e.g., PCA).
- Anomaly Detection: Fraud detection.

3. Reinforcement Learning (Learning by Trial & Error)

- The model learns by interacting with an environment and receiving rewards/punishments.
- Goal: Maximize cumulative reward (e.g., game AI, robotics).

4. Semi-Supervised & Self-Supervised Learning (Brief Mention)

Uses a mix of labeled and unlabeled data (common in real-world scenarios).

3. How Machine Learning Works

Key Steps in ML:

- 1. **Data Collection:** Gather relevant data (structured/tabular, images, text).
- 2. **Data Preprocessing:** Clean, normalize, handle missing values (using Pandas, NumPy).
- 3. **Feature Engineering:** Select/extract meaningful features (e.g., converting text to numbers).

- 4. **Model Selection:** Choose an algorithm (Linear Regression, Decision Trees, Neural Networks).
- 5. **Training:** Fit the model on training data (adjusting weights to minimize error).
- 6. **Evaluation:** Test on unseen data (using metrics like accuracy, MSE).
- 7. **Deployment:** Use the model in real-world applications.

Example Workflow (House Price Prediction):

- Input Features (X): Size, location, bedrooms.
- Output (y): Price.
- **Model:** Linear Regression → Predicts price for new houses.

4. Common Machine Learning Algorithms

Supervised:

- **Regression:** Linear Regression, Decision Trees.
- Classification: Logistic Regression, SVM, Random Forest.

Unsupervised:

- Clustering: K-Means, DBSCAN.
- Dimensionality Reduction: PCA, t-SNE.

Deep Learning (Brief Intro):

• Neural Networks for complex tasks (image, speech recognition).