

Machine Learning

Week 1 • Class 2

Core Terminologies

Features, Labels, Models, Training vs Testing

Class Objective

- Understand key machine learning terminologies
- Identify features and labels in a dataset
- Explain what a model is
- Distinguish between training and testing phases

Why Terminology Matters

Machine learning relies heavily on precise terminology.

Understanding core terms helps you:

- Read ML papers and documentation
- Communicate effectively with teams
- Design correct ML pipelines
- Avoid common beginner mistakes

What Is Data in Machine Learning?

In machine learning, **data** is the foundation.

Data typically consists of:

- Inputs (features)
- Outputs (labels)
- Examples (rows or records)

The quality of data directly impacts model performance.

Features

What Are Features?

Features are individual measurable properties or characteristics of data.

They represent the **inputs** given to a machine learning model.

Examples of Features

Problem	Features
House price prediction	Area, number of rooms, location
Email spam detection	Word frequency, sender, subject
Student performance	Attendance, assignment scores

Features are usually represented as columns in a dataset.

Feature Types (High-Level)

Common feature types include:

- Numerical (age, salary, marks)
- Categorical (gender, city, product type)
- Binary (yes/no, true/false)
- Text and image-based features

Feature selection is a critical ML step.

Labels

What Are Labels?

Labels are the outcomes or target values the model is trying to predict.

They represent the **correct answer** for supervised learning.

Examples of Labels

Problem	Label
House price prediction	House price
Email classification	Spam or Not Spam
Disease detection	Disease present or not

Labels are usually the final column in a dataset.

Features vs Labels

Features	Labels
Input variables	Output variable
Given to model	Predicted by model
Independent	Dependent
Multiple per dataset	Usually one

Understanding this distinction is essential.

What Is a Model?

A **machine learning model** is a mathematical representation that:

- Learns patterns from data
- Maps features to labels
- Makes predictions on new data

The model contains learned parameters.

Examples of Models

Common machine learning models include:

- Linear Regression
- Decision Trees
- k-Nearest Neighbors
- Support Vector Machines
- Neural Networks

Different problems require different models.

Training a Model

What Is Training?

Training is the process where a model:

- Learns from historical data
- Adjusts internal parameters
- Minimizes prediction error

Training uses labeled data.

Training Data

Training data:

- Is the largest portion of the dataset
- Contains both features and labels
- Is used to teach the model patterns

Typically 70–80% of data is used for training.

Testing a Model

What Is Testing?

Testing evaluates how well the model performs on **unseen data**.

The model does not learn during testing.

Testing Data

Testing data:

- Is kept separate from training data
- Contains known labels
- Measures model accuracy and reliability

Typically 20–30% of data is used for testing.

Training vs Testing

Training	Testing
Model learns	Model evaluates
Uses majority of data	Uses held-out data
Adjusts parameters	No parameter updates
Risk of overfitting	Checks generalization

Why Split Data?

Data is split to:

- Prevent memorization
- Measure real-world performance
- Detect overfitting
- Ensure fairness in evaluation

Without splitting, results are misleading.

Overfitting (Concept Preview)

Overfitting occurs when:

- Model performs very well on training data
- Model performs poorly on testing data

This will be covered in detail in later classes.

Class Summary

- Features are input variables
- Labels are target outputs
- Models learn patterns from data
- Training teaches the model
- Testing evaluates performance

Next Class

Types of Learning Problems – Classification