The **Machine Learning (ML) Life Cycle** is the step-by-step process used to build, train, evaluate, and deploy machine learning models. Here's a detailed breakdown of each stage:

1. Problem Definition

- Understand the business or research problem.
- Define the goal of the ML system.
- Example: Predict student dropout from an online course.

2. Data Collection

- Gather data from sources like databases, sensors, web, etc.
- The data should be relevant and sufficient.
- Example: Collect student attendance, grades, and activity logs.

3. Data Preprocessing / Cleaning

- Handle missing values, noise, duplicates, etc.
- Convert raw data into a usable format (e.g., numerical encoding).
- Normalize or scale features.

4. Exploratory Data Analysis (EDA)

- Visualize data patterns.
- Identify relationships and trends.
- Example: Correlation heatmaps, boxplots for outlier detection.

🛠 5. Feature Engineering

- Select, create, or transform features to improve model performance.
- Example: Convert a date into day, month, and weekday.



6. Model Selection

- Choose suitable algorithms (e.g., Linear Regression, Decision Trees, SVM, etc.).
- Split data into training and testing sets.

🏋 7. Model Training

- Feed training data to the algorithm.
- Learn patterns and create a model.

📊 8. Model Evaluation

- Use testing data to measure performance using metrics:
 - Classification: Accuracy, Precision, Recall, F1-score
 - Regression: MAE, RMSE, R²
- Example: Confusion matrix for a spam email detector.

9. Hyperparameter Tuning

- Adjust parameters like learning rate, tree depth, etc., for better results.
- Use techniques like Grid Search or Random Search.

10. Model Deployment

- Deploy the trained model to a production environment (web app, mobile app, etc.).
- Example: Flask API that predicts student dropout risk.

11. Model Monitoring & Maintenance

Monitor real-world performance.

• Retrain with new data if the model degrades (called model drift).