



# Assignment: Modern ML & Neural Networks

## Objective

The objective of this assignment is to:

- Understand **classical machine learning models**
- Implement **core algorithms from scratch**
- Compare **manual implementations vs library implementations**
- Analyze **model performance using metrics**
- Practice **clean OOP-based coding** and **experimental reporting**

You will work primarily on a **movie completion prediction dataset**, and partially on a **music genre classification dataset**.



## Academic Integrity

- Plagiarism is strictly prohibited
- You may use libraries **only where allowed**
- All manual implementations must be your own



## Datasets

### 1. Movie Completion Dataset

<https://www.kaggle.com/datasets/cmaigrot/movies-and-series-rating-from-imdb>

### 2. Music Genre Classification Dataset

<https://www.kaggle.com/datasets/andradaolteanu/gtzan-dataset-music-genre-classification>



## Part 1: Decision Tree & Random Forest

### 1.1 Decision Tree using Library

As demonstrated in class, use a **Decision Tree classifier from a standard ML library** to predict whether a user completes a movie.

You must:

- Train a Decision Tree model
- Try **at least 3 different hyperparameter combinations**, such as:
  - `max_depth`
  - `min_samples_split`
  - `criterion` (gini / entropy)
- Evaluate each model using:
  - Accuracy
  - F1-score

## 1.2 Comparison Table (Mandatory)

Create a **tabular comparison** in your report with the following columns:

Model	max_depth	min_samples_split	criterion	Accuracy	F1 Score
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## ★ Bonus (Optional): Random Forest from Scratch

- Implement the **Random Forest algorithm manually**
- You **may use Decision Trees from a library**, but:
  - Bootstrapping
  - Feature sampling
  - Aggregation (majority voting)

**must be written by you**

- Compare:
  - Decision Tree vs Random Forest
  - Accuracy
  - F1 Score

## Part 2: Logistic Regression

### 2.1 Logistic Regression from Scratch

- Implement **Logistic Regression manually**
- Use:
  - Gradient Descent
  - Sigmoid activation

- Binary cross-entropy loss
- Follow **OOP principles** (use a class)

Evaluate:

- Accuracy
  - F1 Score
- on the movie dataset.
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## 2.2 Logistic Regression using Library

- Train Logistic Regression using a standard ML library
  - Evaluate:
    - Accuracy
    - F1 Score
  - Compare results with your manual implementation
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## ★ Bonus (Theory): Ensemble Methods with Logistic Regression

In the report, **theoretically explain**:

Which ensemble method (Bagging, Boosting, Stacking) would be more suitable when using Logistic Regression for this movie dataset, and why?

Support your answer using:

- Bias-variance intuition
  - Dataset characteristics
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## Part 3: Neural Networks

### 3.1 Neural Network from Scratch (Movie Dataset)

- Implement a **small Neural Network from scratch**
- Example architecture:
  - 3 layers
  - 5 neurons per layer
- Use:
  - Forward propagation
  - Backpropagation

- Gradient descent
- Follow **OOP design** (class-based implementation)

Train on:

- Movie completion dataset

Report:

- Accuracy
  - F1 Score
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### 3.2 Neural Network using Library (Music Genre Dataset)

- Use a deep learning library (TensorFlow / PyTorch)
  - Train a Neural Network on **music genre classification**
  - Evaluate:
    - Accuracy
    - F1 Score
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### 3.3 Hyperparameter Tuning (Mandatory)

- Study basic concepts of **hyperparameter tuning**
- Try **at least 5 different hyperparameter combinations**, such as:
  - Learning rate
  - Number of layers
  - Number of neurons
  - Batch size
  - Activation functions

Report:

- Best model configuration
  - Corresponding Accuracy & F1 Score
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### Visualization Requirements (Mandatory)

Wherever model training is involved, include:

- Training loss vs epochs
- Validation loss vs epochs

Plots must be:

- Clearly labeled
  - Properly explained in the report
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## Coding Guidelines (Strict)

- Any **manual implementation** must:
    - Use **Python classes**
    - Follow **OOP principles**
  - Code should be:
    - Clean
    - Modular
    - Well-commented
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## Submission Format

You must submit a zip file containing:

### 1 Jupyter Notebook ( `.ipynb` )

- All code cells must be:
  - Executed
  - Outputs visible
- Proper section headings
- No broken cells

### 2 Report (separate PDF)

The report must include:

- Explanation of each model
  - Hyperparameter choices
  - Tables of results
  - Graphs and interpretations
  - Your **methodology and thought process**
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## Evaluation Criteria

Component	Weight
Correct implementation	40%
Experimental comparison	20%
Report clarity & explanation	20%
Code quality & OOP usage	10%
Bonus sections	10%