

# Assignment 5: MLP for Adult Income Classification

## Overview

This assignment builds upon Assignment #4 to design and train a Multi-Layer Perceptron (MLP) network for binary classification of the Adult Income dataset. The focus is on network architecture optimization and comprehensive performance evaluation.

Score: 90/100 ★

## Objectives

- Design an MLP network for income classification
- Optimize network architecture and hyperparameters
- Evaluate model performance using multiple metrics
- Visualize training progress with learning rate history

## Dataset

**Adult Income Dataset:** Based on preprocessed data from Assignment #4.

## Task Requirements

### Data Preparation

- **Skip Step 6** (train\_test\_split) from Assignment #4
- **Use Step 7** (StratifiedKFold) from Assignment #4
- Use the **first split** of training and test data for model training and validation

### Feature Engineering

Determine and justify:

- Which features to use in the network
- Transformation methods for each feature:
  - Label encoding
  - One-hot encoding
  - Standardization

### Model Training Configuration

#### Training Parameters

- **Epochs:** 500 epochs
- **Optimizer:** SGD (mandatory - do not use other optimizers)

## Network Architecture Optimization

Find the best configuration by experimenting with:

- Number of neurons per layer
- Number of hidden layers
- Activation functions (ReLU, LeakyReLU, Tanh, etc.)
- Learning rate schedulers

## Performance Evaluation

After training, display the following metrics on **test data**:

1. **Confusion Matrix**
2. **Accuracy**
3. **Precision**
4. **Sensitivity (Recall)**
5. **Specificity**
6. **F1 Score**
7. **AUC (Area Under Curve)**

## Visualization Requirements

Plot accuracy and loss curves with learning rate history overlay:

- Training accuracy/loss over epochs
- Validation/test accuracy/loss over epochs
- Learning rate changes over epochs (overlaid)

Reference: See slide 23 from "07\_PyTorch\_Custom Dataset" lecture material.

## Code Quality Standards

### Required

- Clean and concise code
- Remove unnecessary code
- Show data shapes throughout
- Use provided template and fill all fields

## Not Required

- ✗ Do not show intermediate results
- ✗ Keep only final outputs

## Deliverables

- Jupyter notebook containing:
  - Data preprocessing and feature selection
  - MLP model architecture definition
  - Training loop for 500 epochs
  - All evaluation metrics
  - Accuracy/loss plots with learning rate history
  - Clear documentation of best configuration

## Important:

- ✅ Include all results in the notebook
- ✗ Do NOT compress the notebook file
- ⚠️ **Plagiarism Warning:** Both copier and copied will receive 0 score with no second chances

## Requirements

```
python  
torch  
pandas  
numpy  
scikit-learn  
matplotlib  
seaborn #for confusion matrix visualization
```

## Installation

```
bash  
  
pip install torch pandas numpy scikit-learn matplotlib seaborn
```

# Key Components

## Model Architecture

Design considerations:

- Input layer size (based on selected features)
- Hidden layers (experiment with depth)
- Output layer (binary classification)
- Activation functions

## Optimizer

```
python  
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate)
```

## Evaluation Metrics

```
python  
from sklearn.metrics import (  
    confusion_matrix,  
    accuracy_score,  
    precision_score,  
    recall_score,  
    f1_score,  
    roc_auc_score  
)
```

For specificity, calculate manually:

```
Specificity = TN / (TN + FP)
```

## Example Network Configurations to Try

1. **Baseline:** 2 hidden layers (128, 64) + ReLU + Constant LR
2. **Deep:** 3 hidden layers (256, 128, 64) + ReLU + StepLR
3. **Wide:** 2 hidden layers (512, 256) + LeakyReLU + ExponentialLR
4. **Custom:** Experiment with your own configuration

## Results

Successfully designed and trained an MLP classifier achieving strong performance on the Adult Income dataset. Comprehensive evaluation metrics demonstrate model effectiveness in binary income classification.

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*Assignment completed as part of Deep Learning coursework*