

# Training Sessions Outline on Python for Machine Learning and Neural Networks

## Session 1: Introduction to Python for Machine Learning

### Objectives:

- Set up and use virtual environments.
- Understand basic operations with NumPy.
- Perform data manipulation with Pandas.
- Create visualizations with Matplotlib.

### Agenda:

1. **Using Virtual Environments:**
    - Introduction to virtualenv and its benefits.
    - Setting up and using virtualenv in Windows and Ubuntu.
  2. **Introduction to Python Libraries:**
    - Basic operations with NumPy (simulated data and provided dataset).
    - Data manipulation with Pandas (simulated data and provided dataset).
    - Data visualization with Matplotlib (simulated data and provided dataset).
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## Session 2: Introduction to Supervised Learning

### Objectives:

- Understand the basics of supervised learning.
- Learn about different datasets (training, validation, testing).
- Apply a simple classifier on simulated data and evaluate using accuracy.

### Agenda:

1. **Introduction to Supervised Learning:**
  - What is supervised learning?
  - Importance of different datasets (training, validation, testing).
2. **Implementing a Simple Classifier:**
  - Build and train a simple classifier on simulated data.
  - Evaluate the classifier using accuracy.

### Homework:

- Implement a simple classifier on provided datasets.
  - Evaluate the classifier's performance.
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## Session 3: Data Preprocessing

### Objectives:

- Understand the importance of data preprocessing.
- Perform data cleaning, one-hot encoding, normalization, and centering.

### **Agenda:**

1. **Data Cleaning:**
  - Handling missing data.
  - Removing duplicates.
2. **Data Transformation:**
  - One-hot encoding for labels.
  - Normalization and centering.
  - Other relevant preprocessing steps.

### **Homework:**

- Clean and preprocess a provided dataset.
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## **Session 4: Cross-Validation Techniques**

### **Objectives:**

- Understand the importance of cross-validation.
- Learn different cross-validation techniques.
- Apply cross-validation to evaluate model performance.

### **Agenda:**

1. **Introduction to Cross-Validation:**
  - Definition and purpose of cross-validation.
  - Types of cross-validation (k-fold, stratified k-fold, etc.).
2. **Implementing Cross-Validation:**
  - Applying cross-validation to different models.
  - Evaluating model performance using cross-validation.

### **Homework:**

- Implement cross-validation on provided datasets and evaluate model performance.
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## **Session 5: Preventing Data Leakage**

### **Objectives:**

- Understand data leakage and its impact.
- Learn methods to prevent data leakage.

### **Agenda:**

1. **Introduction to Data Leakage:**
    - Definition and examples.
    - Impact on model performance.
  2. **Preventing Data Leakage:**
    - Proper train-test split.
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## Session 6: Feature Selection Methods

### Objectives:

- Understand the importance of feature selection.
- Implement feature selection methods.

### Agenda:

1. **Introduction to Feature Selection:**
  - Importance and benefits.
2. **Greedy Forward LDA-Based Feature Selection:**
  - Explanation and implementation.

### Homework:

- Implement greedy forward LDA-based feature selection on a provided dataset.
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## Session 7: Support Vector Machines (SVM)

### Objectives:

- Understand the theory behind SVM and RBF SVM.
- Implement and evaluate SVM models.

### Agenda:

1. **Introduction to SVM:**
  - Theory and concepts of SVM.
  - Importance and applications of SVM.
2. **Implementing SVM Models:**
  - Build and train SVM and RBF SVM on simulated data.
  - Evaluate the models.

### Homework:

- Implement and evaluate SVM models on provided datasets.
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## Session 8: Decision Trees and Random Forests

### Objectives:

- Understand the theory behind decision trees and random forests.
- Implement and evaluate tree-based models.

### Agenda:

1. **Introduction to Decision Trees:**
  - Theory and concepts of decision trees.
  - Importance and applications of decision trees.
2. **Implementing Decision Trees and Random Forests:**
  - Build and train decision trees and random forests on simulated data.
  - Evaluate the models.

### Homework:

- Implement and evaluate decision trees and random forests on provided datasets.
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## **Session 9: Introduction to Unsupervised Learning**

### **Objectives:**

- Understand the basics of unsupervised learning.
- Implement simple unsupervised learning models.

### **Agenda:**

1. **Introduction to Unsupervised Learning:**
  - What is unsupervised learning?
  - Types of unsupervised learning algorithms.
2. **Implementing Models:**
  - K-means clustering.
  - Principal Component Analysis (PCA).

### **Homework:**

- Implement and evaluate simple unsupervised learning models on provided datasets.
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## **Session 10: Neural Networks - MLP with Simulated Data**

### **Objectives:**

- Understand the basic structure and functioning of neural networks.
- Implement a simple MLP model on simulated data.

### **Agenda:**

1. **Basics of Neural Networks:**
  - Structure and activation functions.
  - Forward and backward propagation.
2. **Implementing MLP:**
  - Using TensorFlow/Keras to build a simple MLP.
  - Training and evaluating the MLP on simulated data.

### **Homework:**

- Build and train an MLP on provided datasets.
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## **Session 11: Neural Networks - MLP with Real Dataset**

### **Objectives:**

- Apply an MLP model to a real dataset.
- Evaluate the performance of the MLP model.

### **Agenda:**

**1. Implementing MLP on Real Data:**

- Build and train an MLP on the provided dataset.
- Evaluate the model's performance.

**Homework:**

- Fine-tune the MLP model and improve its performance on the provided dataset.
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## **Session 12: Regularization Techniques in Neural Networks**

**Objectives:**

- Understand the importance of regularization in neural networks.
- Implement dropout and batch normalization.

**Agenda:**

- 1. Introduction to Regularization:**
  - Overfitting and the need for regularization.
  - Techniques: Dropout and Batch Normalization.
- 2. Implementing Regularization:**
  - Adding dropout to neural networks.
  - Applying batch normalization.

**Homework:**

- Apply dropout and batch normalization to neural networks and evaluate the performance.
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## **Session 13: Convolutional Neural Networks (CNN) with Simulated Data**

**Objectives:**

- Understand the basics of CNNs.
- Implement a simple CNN model on simulated data.

**Agenda:**

- 1. Introduction to CNNs:**
  - Structure and functioning of CNNs.
  - Convolutional layers, pooling layers, and fully connected layers.
- 2. Implementing CNN:**
  - Using TensorFlow/Keras to build a simple CNN.
  - Training and evaluating the CNN on simulated data.

**Homework:**

- Build and train a CNN on provided datasets.
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## **Session 14: Convolutional Neural Networks (CNN) with Real Dataset**

**Objectives:**

- Apply a CNN model to a real dataset.
- Evaluate the performance of the CNN model.

**Agenda:**

1. **Implementing CNN on Real Data:**
  - Build and train a CNN on the provided dataset.
  - Evaluate the model's performance.

**Homework:**

- Fine-tune the CNN model and improve its performance on the provided dataset.