

1. Data pre-processing operations such as outliers and/or inconsistent data value management.
 - a. Load the Titanic dataset from a CSV file or another data source into a Pandas DataFrame.
 - b. Identify and handle missing values in the dataset
 - c. Create new features if necessary, e.g., 'FamilySize' by combining 'SibSp' and 'Parch.'
 - d. Encode categorical variables like 'Sex' and 'Embarked' as numerical values using one-hot encoding or label encoding.
 - e. Check for and remove duplicate rows if they exist in the dataset.
 - f. Detect outliers (Z-Score, DBSCAN) in numerical columns such as 'Fare' or 'Age'. Also analyze the outliers generated by different methods.
2. Implement Feed forward neural network with three hidden layers for classification on CIFAR-10 dataset.
 - a. Write code to load the CIFAR-10 dataset and display a few sample images.
 - b. Display the dimensions of the images and the number of classes in the dataset.
 - c. Implement data preprocessing steps, including normalizing pixel values and one-hot encoding the labels
 - d. Implement the architecture of a Feedforward Neural Network with three hidden layers
 - e. Compile, Train and evaluate the model performance using metrics such as accuracy, loss, by applying different optimizers (SGD, Adam) and loss functions.
3. Implement a feed forward neural network with three hidden layers for the CIFAR-10 dataset. Record the accuracy and loss during training.
 - a. Repeat the training process with Xavier initialization for weight initialization. Compare the convergence speed and accuracy of the network with the baseline results.
 - b. Analyze the impact of Xavier initialization on the network's performance.
 - c. Repeat the training process with Kaiming initialization for weight initialization. Compare the convergence speed and accuracy of the network with the baseline results.
 - d. Analyze the impact of Kaiming initialization on the network's performance.
4. Implement a feed forward neural network with three hidden layers for the CIFAR-10 dataset. Record the accuracy and loss during training.
 - a. Implement dropout regularization by applying dropout to the hidden layers of the network.
 - b. Train the network with dropout regularization and compare its performance with the baseline results.
 - c. Analyze the impact of dropout on the network's performance in terms of accuracy and overfitting

5. Implement a feed forward neural network with three hidden layers for the CIFAR-10 dataset. Record the accuracy and loss during training.
 - a. Implement L1 or L2 regularization techniques by adding a regularization term to the loss function during training.
 - b. Train the network with regularization and compare its performance with the baseline results.
 - c. Analyze the impact of regularization on the network's performance in terms of accuracy and prevention of overfitting

6. Implement CNN for digit classification using MNIST dataset
 - a. Write code to load the MNIST dataset and display a few sample images. Discuss the dimensions of the images and the number of classes in the dataset.
 - b. Provide code for preprocessing the MNIST dataset before feeding it into the CNN. Include normalization and reshaping of the input images
 - c. Implement the architecture of a Convolutional Neural Network (CNN) for digit classification and train it
 - d. Implement code to evaluate the trained CNN on a test set. Calculate and display relevant evaluation metrics such as accuracy.

7. Implement a simple RNN for review classification using IMDB dataset
 - a. Load the IMDB dataset using the Keras library.
 - b. Preprocess the data by limiting the number of words (e.g., max_features=10,000) and setting a maximum sequence length for reviews (e.g., max_len=500).
 - c. Include a SimpleRNN layer with an appropriate number of units.
 - d. Add a Dense layer with a single neuron and a sigmoid activation function for binary classification.
 - e. Train the RNN model on the training dataset for a reasonable number of epochs (e.g., 5 to 10).
 - f. Evaluate the trained model on the testing dataset and report the test accuracy.

8. Implement a LSTM for review classification using IMDB dataset
 - a. Load the IMDB dataset with only the top 10,000 most frequent words
 - b. Pad or truncate the review sequences to a fixed length of 500 words
 - c. Create lists to store model performance results
 - d. Compile and train the model
 - e. Evaluate the model on the test data
 - f. Visualize the performance change
 - g. Print the test loss and accuracy for different models.

9. Implement a GRU for review classification using IMDB dataset
 - a. Load the IMDB dataset with only the top 10,000 most frequent words
 - b. Pad or truncate the review sequences to a fixed length of 500 words
 - c. Create lists to store model performance results
 - d. Compile and train the model
 - e. Evaluate the model on the test data
 - f. Visualize the performance change
 - g. Print the test loss and accuracy for different models.

10. Implement time series forecasting prediction for NIFTY-50 dataset.
 - a. Load appropriate dataset
 - b. Scale the dataset and convert into time series using TimeseriesGenerator
 - c. Implement SimpeRNN with activation function 'tanh', loss function 'MSE', optimizer 'Adam'
 - d. Fit the model and evaluate the model on the test data
 - e. Print the prediction at any time

11. Implement time series forecasting prediction for Google_Stock_Price dataset.
 - a. Load appropriate dataset
 - b. Scale the dataset and convert into time series using TimeseriesGenerator
 - c. Implement SimpeRNN with activation function 'tanh', loss function 'MSE', optimizer 'Adam'
 - d. Fit the model and evaluate the model on the test data
 - e. Print the prediction at any time

12. Implement time series forecasting prediction for Sunspots dataset.
 - a. Load appropriate dataset
 - b. Scale the dataset and convert into time series using TimeseriesGenerator
 - c. Implement SimpeRNN with activation function 'tanh', loss function 'MSE', optimizer 'Adam'
 - d. Fit the model and evaluate the model on the test data
 - e. Print the prediction at any time