Appendix-2: Logistic Regression

Logistic Regression is a binary classification algorithm that can be extended to multiclass classification by using one-vs-rest or softmax. It is a linear model that learns a weight vector and a bias term to map the input features to a probability score.

For our implementation, we used scikit-learn's LogisticRegression classifier with the default L2 regularization.

```
In [1]: # import library dependencies
  import numpy as np
  from sklearn.linear_model import LogisticRegression
  from sklearn.metrics import accuracy_score, precision_score, recall_score, f
  from sklearn.model_selection import train_test_split
  from sklearn.model_selection import learning_curve
  import matplotlib.pyplot as plt
  import joblib
```

Import Data

```
In [2]: ROOT_PATH='../'
In [3]: # function to open pickle file
        def unpickle(file):
            import pickle
            with open(file, 'rb') as fo:
                dict = pickle.load(fo, encoding='bytes')
            return dict
In [4]: # store each pickle files in individual batches
        batch1 = unpickle(ROOT_PATH+"cifar-10-batches-py/data_batch_1")
        batch2 = unpickle(ROOT_PATH+"cifar-10-batches-py/data_batch_2")
        batch3 = unpickle(ROOT_PATH+"cifar-10-batches-py/data_batch_3")
        batch4 = unpickle(ROOT PATH+"cifar-10-batches-py/data batch 4")
        batch5 = unpickle(ROOT_PATH+"cifar-10-batches-py/data_batch_5")
        test_batch = unpickle(ROOT_PATH+"cifar-10-batches-py/test_batch")
In [5]: # function to create labels and images from data
        def load data0(btch):
            labels = btch[b'labels']
            imgs = btch[b'data'].reshape((-1, 32, 32, 3))
            res = []
            for ii in range(imgs.shape[0]):
                img = imgs[ii].copy()
                img = np.fliplr(np.rot90(np.transpose(img.flatten().reshape(3,32,32))
                res.append(img)
            imgs = np.stack(res)
            return labels, imgs
```

```
In [6]: # function to load data into training and test set
        def load_data():
             x_{train_l} = []
             y train l = []
             for ibatch in [batch1, batch2, batch3, batch4, batch5]:
                 labels, imgs = load_data0(ibatch)
                 x train l.append(imgs)
                 y_train_l.extend(labels)
             x_train = np.vstack(x_train_l)
             y train = np.vstack(y train 1)
             x_{test_l} = []
             y \text{ test } l = []
             labels, imgs = load_data0(test_batch)
             x_test_l.append(imgs)
            y_test_l.extend(labels)
             x_test = np.vstack(x_test_1)
             y_test = np.vstack(y_test_1)
             return (x_train, y_train), (x_test, y_test)
```

Preprocess Data

```
In [7]: # create training and test set
         (x_train, y_train), (x_test, y_test) = load_data()
In [8]: print('x_train shape:', x_train.shape)
         print('y_train shape:', y_train.shape)
         print('x_test shape:', x_test.shape)
         print('y_test shape:', y_test.shape)
         x_train shape: (50000, 32, 32, 3)
         y train shape: (50000, 1)
         x_test shape: (10000, 32, 32, 3)
         y_test shape: (10000, 1)
In [9]: print(x train.shape[0], 'train samples (x)')
         print(y_train.shape[0], 'train samples (y)')
         50000 train samples (x)
         50000 train samples (y)
In [10]: print(x_test.shape[0], 'test samples (x)')
         print(y_test.shape[0], 'test samples (y)')
         10000 test samples (x)
         10000 test samples (y)
In [11]: # Flatten the images
         X_train = x_train.reshape(x_train.shape[0], -1)
         X_test = x_test.reshape(x_test.shape[0], -1)
In [12]: # Normalize the data
         X train = X train.astype('float32') / 255
         X test = X test.astype('float32') / 255
```

```
In [13]: # Reshape y train and y test to 1d arrays
         y_train = y_train.ravel()
         y_test = y_test.ravel()
In [14]: # Split the data into training and validation sets
         X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_siz
         Define model and train
In [15]:
         # Define the model
         lr = LogisticRegression(solver='saga', multi_class='multinomial', verbose=1,
In [16]: # Train the model
         lr.fit(X_train, y_train)
         [Parallel(n_jobs=-1)]: Using backend ThreadingBackend with 8 concurrent work
         ers.
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Epoch 503, change: 0.00128441
Epoch 504, change: 0.00119309
Epoch 505, change: 0.00130530
Epoch 506, change: 0.00140873
Epoch 507, change: 0.00154390
Epoch 508, change: 0.00129209
Epoch 509, change: 0.00125857
Epoch 510, change: 0.00128727
Epoch 511, change: 0.00146484
Epoch 512, change: 0.00144715
Epoch 513, change: 0.00150014
Epoch 514, change: 0.00135295
Epoch 515, change: 0.00119341
Epoch 516, change: 0.00145061
Epoch 517, change: 0.00119237
Epoch 518, change: 0.00126799
Epoch 519, change: 0.00140673
Epoch 520, change: 0.00119249
Epoch 521, change: 0.00128651
Epoch 522, change: 0.00131331
Epoch 523, change: 0.00119235
Epoch 524, change: 0.00119426
Epoch 525, change: 0.00148332
Epoch 526, change: 0.00119262
Epoch 527, change: 0.00119214
Epoch 528, change: 0.00119241
Epoch 529, change: 0.00119131
Epoch 530, change: 0.00119223
Epoch 531, change: 0.00119152
Epoch 532, change: 0.00119212
Epoch 533, change: 0.00119197
Epoch 534, change: 0.00119110
Epoch 535, change: 0.00119164
Epoch 536, change: 0.00119134
Epoch 537, change: 0.00119453
Epoch 538, change: 0.00118896
Epoch 539, change: 0.00119143
Epoch 540, change: 0.00124383
Epoch 541, change: 0.00119126
Epoch 542, change: 0.00119099
Epoch 543, change: 0.00119328
Epoch 544, change: 0.00119152
Epoch 545, change: 0.00119215
Epoch 546, change: 0.00118962
Epoch 547, change: 0.00119132
Epoch 548, change: 0.00119129
Epoch 549, change: 0.00118992
Epoch 550, change: 0.0max iter reached after 3547 seconds
```

Save and load model

```
In [17]: # Save the model to a file
    joblib.dump(lr, 'logistic_regression_model_saga.sav')
Out[17]: ['logistic_regression_model_saga.sav']
In [18]: # Load the saved model from a file
    loaded_model = joblib.load('logistic_regression_model_saga.sav')
```

Evaluate the model

```
In [19]:
        # Evaluate the model
         train_acc = accuracy_score(y_train, loaded_model.predict(X_train))
         val_acc = accuracy_score(y_val, loaded_model.predict(X_val))
         test acc = accuracy score(y test, loaded model.predict(X test))
         print(f"Train Accuracy: {train acc:.4f}")
         print(f"Val Accuracy: {val acc:.4f}")
         print(f"Test Accuracy: {test_acc:.4f}")
         Train Accuracy: 0.5047
         Val Accuracy: 0.3848
         Test Accuracy: 0.3864
In [20]: # Calculate precision, recall, and F1 score on test set
         test pred = loaded model.predict(X test)
         precision = precision score(y test, test pred, average='weighted')
         recall = recall_score(y_test, test_pred, average='weighted')
         f1 = f1 score(y test, test pred, average='weighted')
         print(f"Test Precision: {precision:.4f}")
         print(f"Test Recall: {recall:.4f}")
         print(f"Test F1 score: {f1:.4f}")
         Test Precision: 0.3833
         Test Recall: 0.3864
         Test F1 score: 0.3843
```