```
In [1]: import numpy as np
        import pandas as pd
        from sklearn.datasets import make_classification
        from sklearn.model_selection import train test split
        from sklearn.metrics import log loss
        import math
        import matplotlib.pyplot as plt
        import warnings
        warnings.filterwarnings("ignore")
In [2]: X, y = make classification(n samples=50000, n features=15, n informative=10, n redundant=5,
                                   n_classes=2, weights=[0.7], class_sep=0.7, random_state=15)
        X_train, X_test, y_train, y_test = train_test_split(X, y, test size=0.25, random state=15)
In [3]: class EpochLoss:
          def init (self, epoch, loss, weight, intercept):
            self.epoch = epoch
            self.loss = loss
            self.weight = weight
            self.intercept = intercept
        class SGDLogisticRegression:
            def sigmoid(self, x, w, b):
                z = np.dot(x, w.T) + b
                sig = 1 / (1 + np.exp(-z))
                return sig
            def fit(self, X train, X test, y train, y test, eta0, alpha, num iteration):
                N train = len(X train)
                N \text{ test} = len(X \text{ test})
                lst train = []
                lst_test = []
                #Initial w and b
                w = np.zeros_like(X[0])
                b = 0
                for epoch in range(1, num iteration):
                    #-----#
                    lst y pred train = []
                    lst y pred test = []
                    for i in range(N_train):
                        #calculating sigmoid of z
                        '''y pred train = self.sigmoid(X train[i], w, b)
                        lst y pred train.append(y pred train)
                        calculating sigmoid of z
                        y pred test = self.sigmoid(X_test[i], w, b)
                        lst y pred test.append(y pred test)'''
                        y pred train = self.sigmoid(X train[i], w, b)
                        #subtracting y pred train from y train to get error
                        error = y train[i] - y pred train
                        #Updating weights and intercept
                        w = (1 - (alpha * eta0)/N_train) * w + alpha * X_train[i] * error
                        b = (b + alpha * error)
                    #Calculating log loss for Train
                    lst_y_pred_train = np.array([self.sigmoid(X_train[i], w, b) for i in range(N_train)])
                    avg_log_loss_train = log_loss(y_train, lst_y_pred_train)
                    #Calculating log loss for Test
                    lst_y_pred_test = np.array([self.sigmoid(X_test[i], w, b) for i in range(N_test)])
                    avg log loss test = log loss(y test, lst y pred test)
                    #Checking if previous epoch for train data is having the same value then breaking the ep
        och loop
                    round_upto = 4
                    if len(lst_train) > 0:
                      found = False
                      for loss in 1st train:
                        if loss.loss == round(avg_log_loss_train, round_upto):
                          found = True
                          break
                      if found == False:
                        lst_train.append(EpochLoss(epoch, round(avg_log_loss_train, round_upto), w, b))
                        lst_test.append(EpochLoss(epoch, round(avg_log_loss_test, round_upto), w, b))
                      else:
                        break
                    else:
                        lst train.append(EpochLoss(epoch, round(avg log loss train, round upto), w, b))
                        lst_test.append(EpochLoss(epoch, round(avg_log_loss_test, round_upto), w, b))
                    print('Epoch=%d, Bias=%.3f, Avg. Loss=%.4f' % (epoch, b, avg_log_loss_train))
                self.coef = w
                self.intercept = b
                self.lstEpochLoss_train = lst_train
                self.lstEpochLoss_test = lst_test
                return self
            def pred(self, w, b, X):
              N = len(X)
              predict = []
              for i in range(N):
                if self.sigmoid(w, X[i], b) >= 0.5:
                  predict.append(1)
                else:
                  predict.append(0)
              return np.array(predict)
In [4]: eta0 = 0.0001 #lambda
        alpha = 0.0001 #learning rate
        num_iterations = 20
        sgd logistic reg = SGDLogisticRegression()
        model = sgd_logistic_reg.fit(X_train, X_test, y_train, y_test, eta0, alpha, num_iterations)
        Epoch=1, Bias=-0.315, Avg. Loss=0.4040
        Epoch=2, Bias=-0.472, Avg. Loss=0.3884
        Epoch=3, Bias=-0.581, Avg. Loss=0.3831
        Epoch=4, Bias=-0.660, Avg. Loss=0.3808
        Epoch=5, Bias=-0.718, Avg. Loss=0.3796
        Epoch=6, Bias=-0.762, Avg. Loss=0.3790
        Epoch=7, Bias=-0.794, Avg. Loss=0.3786
        Epoch=8, Bias=-0.819, Avg. Loss=0.3785
        Epoch=9, Bias=-0.837, Avg. Loss=0.3784
        Epoch=10, Bias=-0.851, Avg. Loss=0.3783
In [5]: print('\n--- Weight ---')
        print(model.coef)
        print('\n--- Intercept ---')
        print(model.intercept)
        #Train Loss and epoch
        loss train = [o.loss for o in model.lstEpochLoss train]
        epoch_train = [o.epoch for o in model.lstEpochLoss_train]
         #Test Loss and epoch
        loss test = [o.loss for o in model.lstEpochLoss test]
        epoch_test = [o.epoch for o in model.lstEpochLoss_test]
        #Plotting curve
        plt.plot(epoch train, loss train, label='Train Loss curve')
        plt.plot(epoch_test, loss_test, label='Test Loss curve')
        #Adding legends, label, title and grid to the plot
        plt.legend()
        plt.xlabel("Epoch")
        plt.ylabel('Loss')
        plt.title('Train and Test loss against each epoch')
        plt.grid()
        plt.show()
        --- Weight ---
        [-0.42483228 \quad 0.19148808 \quad -0.1465521 \quad 0.33813581 \quad -0.21434406 \quad 0.56645689
         -0.44530973 -0.09124939 0.21893754 0.17080465 0.19611785 0.00156244
         --- Intercept ---
        -0.8608553931749093
                     Train and Test loss against each epoch
           0.405
                                              Train Loss curve
                                              Test Loss curve
           0.400
           0.395
        O.390
           0.385
           0.380
                                  Epoch
        Implementing SGDClassifier from sklearn
In [6]: from sklearn import linear model
        clf = linear_model.SGDClassifier(eta0=0.0001, alpha=0.0001, loss='log', random_state=15, penalty='l
        2', tol=1e-3, verbose=2, learning_rate='constant')
        clf.fit(X=X_train, y=y_train)
        -- Epoch 1
        Norm: 0.76, NNZs: 15, Bias: -0.314605, T: 37500, Avg. loss: 0.455801
        Total training time: 0.01 seconds.
        -- Epoch 2
        Norm: 0.92, NNZs: 15, Bias: -0.469578, T: 75000, Avg. loss: 0.394737
        Total training time: 0.01 seconds.
        -- Epoch 3
        Norm: 0.98, NNZs: 15, Bias: -0.580452, T: 112500, Avg. loss: 0.385561
        Total training time: 0.02 seconds.
        Norm: 1.02, NNZs: 15, Bias: -0.660824, T: 150000, Avg. loss: 0.382161
```

## Total training time: 0.04 seconds. -- Epoch 6 Norm: 1.06, NNZs: 15, Bias: -0.761816, T: 225000, Avg. loss: 0.379481 Total training time: 0.04 seconds.

learning\_rate='constant', loss='log', max\_iter=None, n\_iter=None,
n\_jobs=1, penalty='l2', power\_t=0.5, random\_state=15, shuffle=True,

Norm: 1.04, NNZs: 15, Bias: -0.717218, T: 187500, Avg. loss: 0.380474

Total training time: 0.02 seconds.

Convergence after 6 epochs took 0.04 seconds

tol=0.001, verbose=2, warm\_start=False)

-- Epoch 5

0.9547466666666666

0.95224

--- Accuracy on X\_test ---

Comparing weight vector and intercept of sklearn SGDClassifier and sgd\_logistic\_reg

In [7]: model.coef - clf.coef\_, model.intercept - clf.intercept\_

## 

```
In [8]: print('\n--- Accuracy on X_train ---')
    y_hat_train = sgd_logistic_reg.pred(model.coef, model.intercept, X_train)
    print(1- np.sum(y_train - y_hat_train) / len(X_train))

    print('\n--- Accuracy on X_test ---')
    y_hat_test = sgd_logistic_reg.pred(model.coef, model.intercept, X_test)
    print(1-np.sum(y_test - y_hat_test)/len(X_test))

--- Accuracy on X_train ---
```