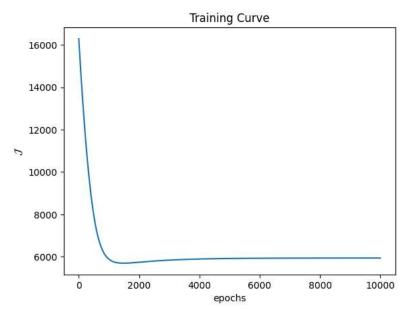
```
1 import pandas as pd
 2 import numpy as np
 3 import matplotlib.pyplot as plt
1
    from matplotlib.colors import ListedColormap
    cmap_bold = ListedColormap(['#FF0000','#0000FF'])
    cmap_light = ListedColormap(['#FFBBBB','#BBBBFF'])
1
    # Step 1: Load the dataset
    data = pd.read csv("/content/Churn Modelling - Churn Modelling.csv")
 1
    # Step 2: Data preprocessing
    # Encode categorical variables
 2
    data['Gender'] = data['Gender'].map({'Female': 0, 'Male': 1}) # Encode Gender
    data = pd.get_dummies(data, columns=['Geography']) # One-hot encoding for Geography
4
 6
    # Select features and target variable
    7
              'Geography_France', 'Geography_Germany', 'Geography_Spain']].values
9
    y = data['Exited'].values.reshape(-1, 1) # Reshape y for matrix operations
10
11
    # Normalize features
12
    X = X.astype(np.float64)
13
14
    X = (X - np.mean(X, axis=0)) / np.std(X, axis=0)
15
1
    def sigmoid(h):
2
        return 1/(1+np.exp(-h))
3
4
    def bin_cross_entropy(y,p_hat):
 5
        return -(1/len(y))*np.sum(y*np.log(p_hat)+(1-y)*np.log(1-p_hat))
 6
7
    def accuracy(y,y_hat):
 8
        return np.mean(y==y_hat)
1
    class LogisticRegression():
        def init (self,thresh=0.5):
2
 3
            self.thresh=thresh
            self.W = None
4
 5
 6
 7
        def fit(self,X,y,eta=1e-3,epochs=1e3,show_curve=True):
 8
            epochs = int(epochs)
9
            N,D = X.shape
10
11
            #Initailize Weights and Biases
            self.W = np.random.randn(D)
12
            self.b = np.random.randn(1)
13
            J = np.zeros(epochs)
14
15
16
            #Stochastic Gradient Descent
            for epoch in range(epochs):
17
                p_hat = self.__forward__(X)
18
19
                J[epoch] = bin_cross_entropy(y,p_hat)
20
21
                #WeightUpdate Rules
22
                error = (p_hat - y.ravel())
23
                self.W -= eta*(1/N)*X.T@error
                self.b -= eta*(1/N)*np.sum(error)
24
25
            if show_curve:
26
                plt.figure()
27
                plt.plot(J)
                plt.xlabel("epochs")
28
29
                plt.ylabel("$\mathcal{J}$")
30
                plt.title("Training Curve")
31
                plt.show()
32
        def __forward__(self,X):
33
            return sigmoid(X@self.W+self.b)
34
35
36
        def predict(self,X):
            raturn (salf forward (X) >= salf thrash) astuna(nn int3)
```

→

1 log_reg = LogisticRegression()
2 log_reg.fit(X,y, epochs =1e4, eta=1e-2, show_curve=True)
3 y_hat = log_reg.predict(X)
4 print(f"Training Accuracy: {accuracy(y,y_hat):0.4f}")
5 print(log_reg.W,log_reg.b)



Training Accuracy: 0.7528
[-0.06457684 0.76219181 -0.04609227 0.16450587 -0.05903227 -0.02034948
-0.53715938 0.07763329 -0.26304535 -0.35818206 0.02522075 -0.294035441 [-1.65302541

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