Hill and valley prediction using Logistic Regression.

Importing Pandas library, scikit library

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Defining Target variable(y) and feature variable(x)

Train Test Splitting of the data

8 rows × 101 columns

Scaling the data

In [47]: from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
X_train=sc.fit_transform(X_train)
print(X_train.shape)
X_test=sc.transform(X_test)
(909, 100)

Modelling

In [42]: from sklearn.linear_model import LogisticRegression
lg=LogisticRegression(random_state=0)
lg.fit(X_train,y_train)
y_pred=tg.predict(X_test)
print(y_pred.shape)

(303,)

Accuracy Score

In [43]: from sklearn.metrics import accuracy_score,confusion_matrix
print(accuracy_score(y_test,y_pred))
print(confusion_matrix(y_test,y_pred))

0.752475247525 [[151 2] [73 77]]

In []: #This project is based on the Dasa science using Python Programming language.

#The main purpose of creating this project is to develop a predictive model using logistic regression that can accurately classify geographical locations as #either hills or valleys based on a set of input features.

#The model will be trained on a labeled dataset of geographical features and their corresponding classifications,

#and then evaluated on a separate test dataset to measure its performance.

#The ultimate goal of this project is to provide a useful tool for identifying hills and valleys in various geographic locations,

#which can have important applications in fields such as geology, agriculture, and urban planning.

In []: