

# **FLOOD PREDICTION**

**Name : G B V K UDAY**

**Reg No : 39110325**

**College : Sathyabama Institute of Science and Technology**

**Branch : B.E , CSE. Project Title : Flood prediction using Machine Learning.**

## **INDEX**

### **1. INTRODUCTION**

- a. Overview
- b. Purpose

### **2. LITERATURE SURVEY**

- a. Existing Solution
- b. Proposed solution

### **3. THEORETICAL ANALYSIS**

### **4. HARDWARE / SOFTWARE REQUIREMENTS**

### **5. FLOW CHART**

### **6. RESULT**

### **7. ADVANTAGES AND DISADVANTAGES**

### **8. APPLICATIONS**

### **9. CONCLUSION**

### **10. FEATURE SCOPE**

### **11. BIBILOGRAPHY**

### **12. APPENDIX**

- a. Source code

b. UI output Screen-shots

## **INTRODUCTION :**

Floods are inevitable, but with timely alerts, their effects can be minimized. There are many people who die every year due to devastating floods, the number of people become homeless and many people die due to lack of proper help after a flood. The lack of timely alerts has always been an issue concerning it. Delay in alerts in flood-prone areas is the biggest loophole of an economy. Conventional systems run a little low in forecasting floods at the right time so that proper actions could be taken before any disaster. By using machine learning we can predict floods or forecast floods with better accuracy. This project aims at building predictive modeling based on the historical weather data of particular areas in order to predict the occurrence of floods. The predictive model is built on different machine learning algorithms. The concerned authority monitor this flood prediction system through a web application.

## **LITERATURE SURVEY :**

For this project, we need to use classification algorithms as the output we will get is categorical values which is either floods will occur or not so, we need to build the most accurate classification algorithm to get the desired output

In this project, we will be using classification algorithms such as Decision tree, Random forest, KNN, and xgboost. We will train and test the data with these algorithms. From this best model is selected and saved in save format. We will be doing flask integration and cloud deployment.

## **THEORETICAL ANALYSIS:**

1. Install Required Libraries.
2. Data collections
  - . Collect the dataset or Create the dataset
3. DataPreprocessing
  - Import the Libraries.
  - Importing the dataset.
  - Understanding Data Type and Summary of features.
  - Take care of missing data
  - Data Visualization.
  - Drop the column from DataFrame & replace the missing value.
  - Splitting the Dataset into Dependent and Independent variables
  - Splitting Data into Train and Test.
4. Model Building
  - Training and testing the model
  - Evaluation of Model
  - Saving the Model
5. Application Building
  - Create an HTML file
  - Build a Python Code
6. Final UI
  - Dashboard Of the flask app.

## **HARDWARE / SOFTWARE REQUIREMENTS :**

- > Processor : Intel(R) Core(TM) i3 - 6100U 2.30GHz / Equivalent processor
- > RAM : 4Gb or more
- >Storage : A basic hard disk of storing files.

Software requirements :

- >Anaconda Software
- >Visual Studio code / Spyder Software

## **EXPERIMENTAL INVESTIGATIONS:**

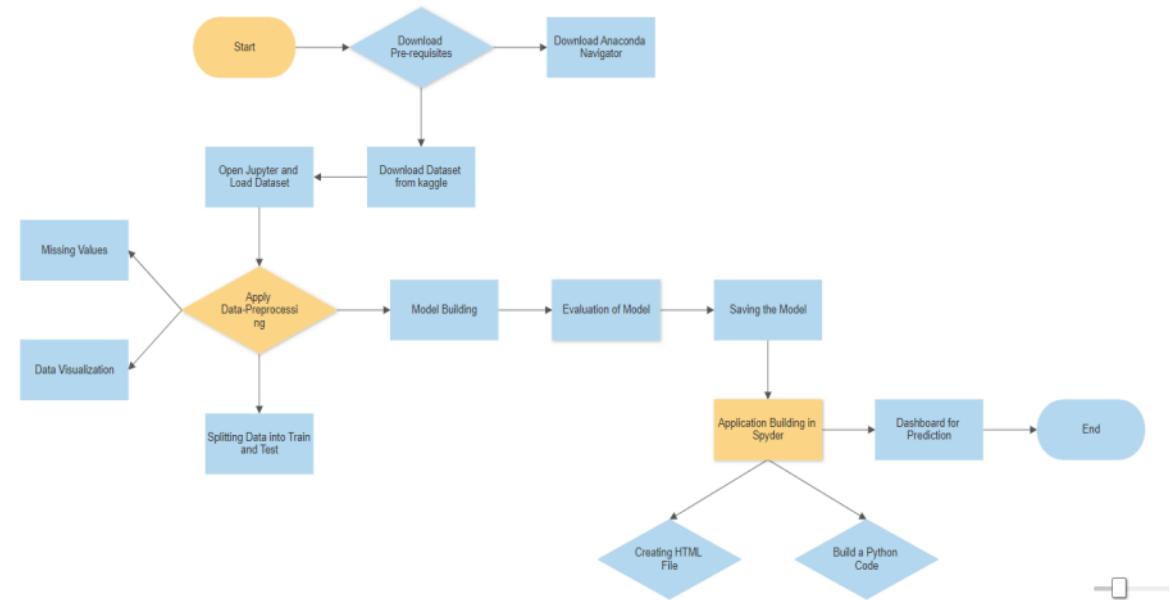
Category: Machine Learning

Skills Required:

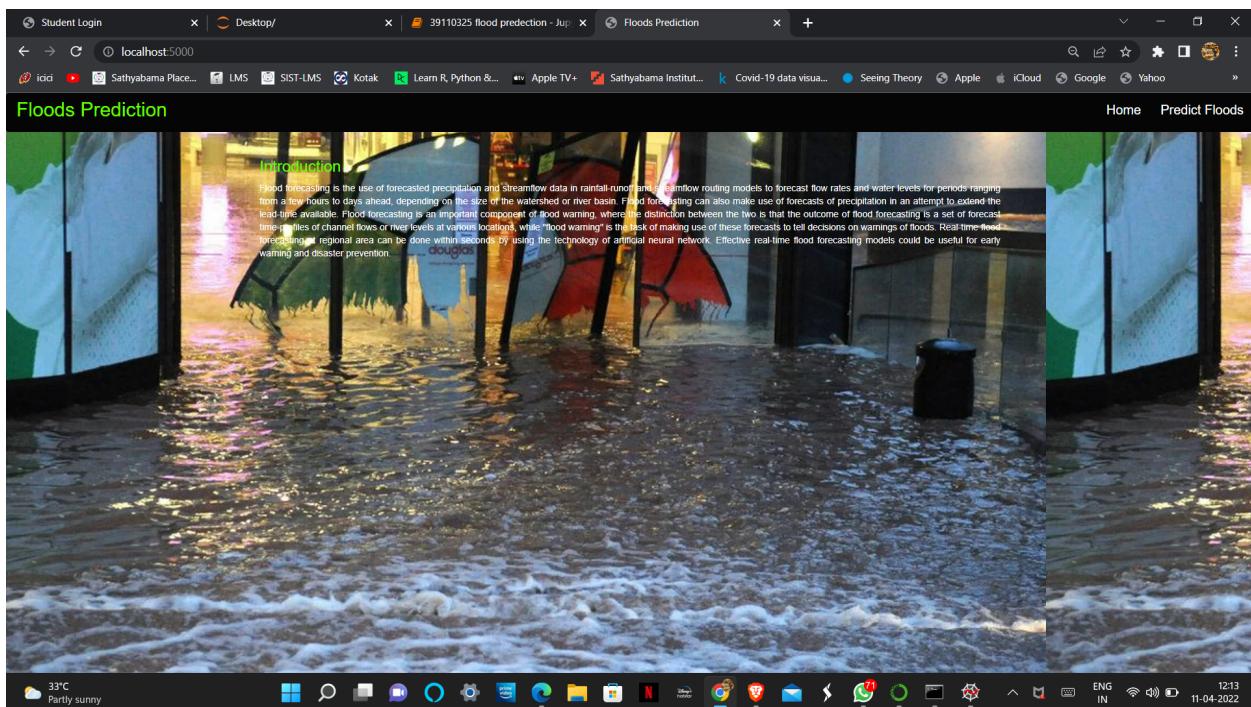
Python,Python Web Frame Works,Python For Data Analysis,Exploratory Data Analysis,Data Preprocessing Techniques,Machine Learning

While working on the solution, I have seen the dataset of the floods and its quite frequent to see the whether forecast change drastically and cloud cover increases , rainfalls in the months April to June and lastly floods occur. There is interconnection of all the factors which leads to floods.

## FLOW CHART :



## RESULT:



**Floods Prediction**

Cloud Cover      Annual Rain Fall

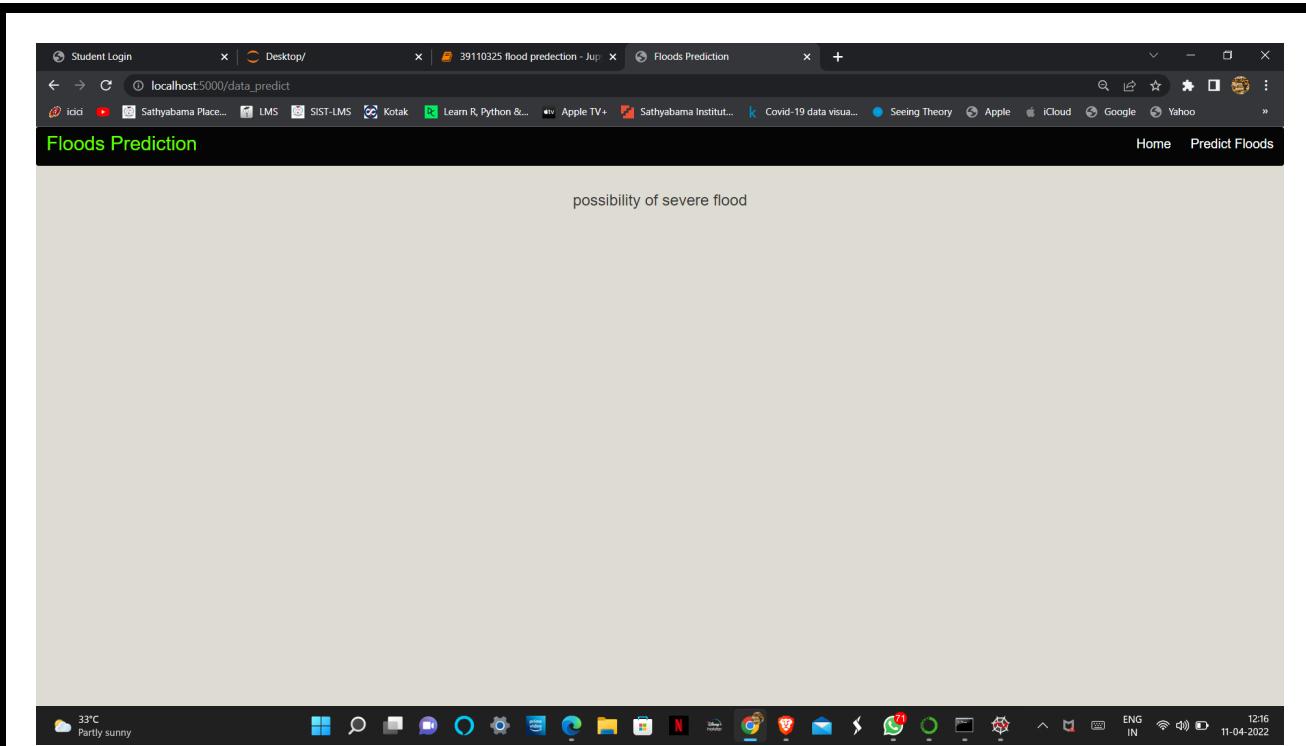
Cloud Cover Percentage (%)      Enter Annual Rain Fall in mm

Jan-Feb Rainfall      June-September

Jan-Feb Rainfall in mm      June-September Rainfall in mm

March-May Rainfall      Predict

No possibility of severe flood



## ADVANTAGES AND DISADVANTAGES ADVANTAGES :

- >The main purpose of flood warning is to save life by allowing people, support and emergency services time to prepare for flooding.
- >The model which I built is 96 % accurate so one can rely on the prediction done by the model. ->With the help of this model , you can determine and manage environmental and water resource systems.
- >Knowledge about the characteristics of a river's drainage basin, such as soil-moisture conditions, ground temperature, snowpack, topography, vegetation cover, and impermeable land area, which can help to predict how extensive and damaging a flood might become.

## **DISADVANTAGES :**

->One of the algorithm used in this model is Random Forest, where it can become a plot as it gets prediction accuracy on complex problems is usually inferior to Gradient - Boosted Trees.

## **APPLICATIONS :**

The applications in flood prediction can be classified according to flood resource variables, i.e., water level, river flood, soil moisture, rainfall–discharge, precipitation, river inflow, peak flow, river flow, rainfall–runoff, flash flood, rainfall, streamflow, seasonal stream flow.

## **CONCLUSION :**

The summary of the project is that even though Floods are inevitable, they can be prevented with accurate measures like predicting the outcomes, changes to be taken to be on the safe side like building dams, reservoirs etc..and especially the technology has taken huge turn so, it will be matter of time before we gain control on what we are challenged.

## **FUTURE SCOPE :**

There can be future changes can be made to my model by adding the data like water level, soil moisture, average temperature of the area, rainfall - discharge , precipitation , drainage systems in the area etc.... in the dataset makes even more accurate and since the data will be enhanced, the analysis becomes the key part in the prediction.

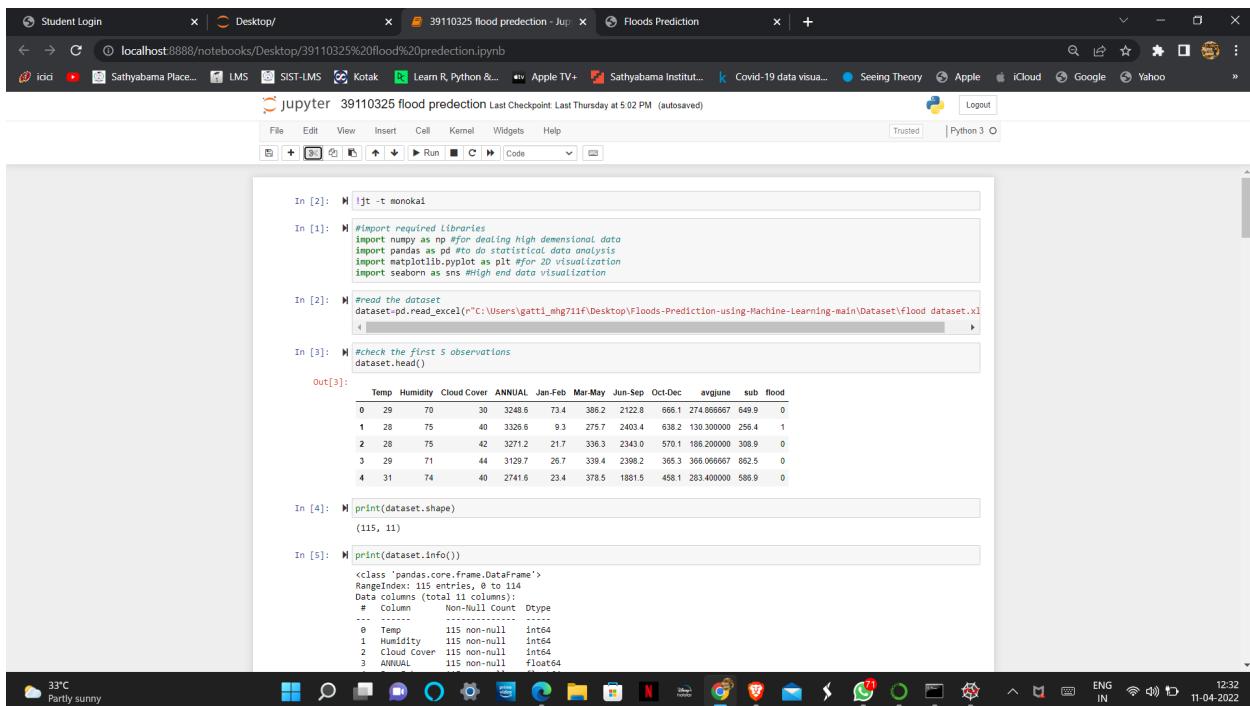
## BIBILOGRAPHY:

There are some references which are to be considered while making the project.

URL - 1 : Flood prediction - Oxford Reference

URL - 2 : Flood Forecasting - an overview | ScienceDirect Topics

## APPENDIX - SOURCE CODE :



The screenshot shows a Jupyter Notebook interface running on a local host. The code in the notebook is as follows:

```
In [2]: % jt -t monokai
In [1]: #import required Libraries
import numpy as np #for dealing high dimensional data
import pandas as pd #to do statistical data analysis
import matplotlib.pyplot as plt #for 2D visualization
import seaborn as sns #high end data visualization

In [2]: #read the dataset
dataset=pd.read_excel(r"C:\Users\gatti_mhg71if\Desktop\Floods-Prediction-using-Machine-Learning-main\Dataset\flood dataset.xlsx")

In [3]: #check the first 5 observations
dataset.head()

Out[3]:
   Temp  Humidity  Cloud Cover  ANNUAL  Jan-Feb  Mar-May  Jun-Sep  Oct-Dec  avgjune  sub_flood
0    29       70          30    3248.6     73.4    388.2    2122.8      666.1    274.866667      849.9      0
1    28       75          40    3326.6     9.3    275.7    2403.4      638.2    130.300000      256.4      1
2    28       75          42    3271.2    21.7    336.3    2343.0      570.1    188.200000      308.9      0
3    29       71          44    3129.7    26.7    339.4    2308.2      365.3    366.066667      862.5      0
4    31       74          40    2741.6    23.4    378.5    1881.5      458.1    283.400000      586.9      0

In [4]: print(dataset.shape)
(115, 11)

In [5]: print(dataset.info())
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 115 entries, 0 to 114
Data columns (total 11 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Temp            115 non-null    int64  
 1   Humidity         115 non-null    int64  
 2   Cloud Cover     115 non-null    int64  
 3   ANNUAL          115 non-null    float64
```

Student Login   Desktop/   39110325 flood prediction - Jupyter Notebook   Floods Prediction

In [9]: `#Correlation  
dataset.corr()`

```
Out[9]:
      Temp Humidity Cloud Cover ANNUAL Jan-Feb Mar-May Jun-Sep Oct-Dec avgJune sub flood
Temp 1.000000 -0.012727 -0.045568 -0.053014 -0.080076 0.099515 -0.081965 -0.063034 -0.019751 -0.088331 -0.080946
Humidity -0.012727 1.000000 0.085824 -0.054767 -0.185985 -0.101232 -0.029583 0.059739 0.017656 0.029981 0.082050
Cloud Cover -0.045568 0.085824 1.000000 0.051166 0.044376 0.096845 0.010833 0.020965 -0.086843 -0.16645 0.086901
ANNUAL -0.053014 -0.054767 0.051166 1.000000 0.038359 0.387790 0.081190 0.220964 0.474644 0.220000 0.628674
Jan-Feb -0.080076 -0.185985 0.004376 0.038359 1.000000 0.068479 0.001178 -0.143670 0.164691 0.201268 -0.084446
Mar-May 0.099515 -0.101232 0.096845 0.387790 0.068479 1.000000 -0.029007 0.171805 0.019183 -0.475750 -0.071598
Jun-Sep -0.081965 -0.029583 0.010833 0.081190 0.001178 -0.029007 1.000000 -0.144487 0.511113 0.431997 0.705202
Oct-Dec -0.063034 0.059739 0.020965 0.232069 -0.143670 0.171805 -0.144487 1.000000 -0.028055 -0.050862 -0.024852
avgJune -0.019751 0.017656 -0.08843 0.474644 0.164691 0.019183 0.511113 -0.028055 1.000000 0.780445 0.349828
sub -0.088331 0.029981 -0.106455 0.220964 0.201268 -0.475750 0.431997 -0.050862 0.780445 1.000000 0.349828
flood -0.080946 0.020250 0.088081 0.628874 -0.084446 -0.017588 0.075202 -0.024852 0.379778 0.349828 1.000000
```

In [10]: `import seaborn as sns  
fig=plt.gcf()  
fig.set_size_inches(15,15)  
fig=sns.heatmap(dataset.corr(),annot=True,cmap='summer',  
 linewidths=1,linecolor='k',square=True,  
 mask=False,vmin=-1,vmax=1,  
 cbar_kws={"orientation": "vertical"},cbar=True)`

```
In [11]: M dataset.drop(["Oct-Dec"],axis=1,inplace=True)
In [12]: M dataset.head()
Out[12]:
   Temp  Humidity  Cloud Cover  ANNUAL  Jan-Feb  Mar-May  Jun-Sep  avgjune  sub_flood
0    29       70        30    3248.6    73.4    388.2    2122.8  274.866667      649.9      0
1    28       75        40    3328.6    9.3    275.7    2403.4  130.300000     256.4      1
2    28       75        42    3271.2   21.7    338.3    2343.0  188.200000     308.9      0
3    29       71        44    3129.7   26.7    339.4    2398.2  366.000000     862.5      0
4    31       74        40    2741.6   23.4    378.5    1881.5  283.400000     588.9      0

In [13]: M dataset['Flood'].value_counts()
Out[13]:
0    99
1    16
Name: flood, dtype: int64

In [14]: M #independent features
x=dataset.iloc[:,2:7].values

In [15]: M #dependent feature
y=dataset.iloc[:,9].values

In [16]: M x.shape
Out[16]: (115, 5)

In [17]: M y.shape
Out[17]: (115, 1)

In [18]: M #split the data into train and test set from our x and y
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25,random_state=10)

In [19]: M #checking the shape of our 4 variables
print(x_train.shape)
```

```
In [19]: M #checking the shape of our 4 variables
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
Out[19]:
(86, 5)
(29, 5)
(86, 1)
(29, 1)

In [20]: M #import StandardScaler
from sklearn.preprocessing import StandardScaler
#Create object to StandardScaler class
sc=StandardScaler()
x_train=sc.fit_transform(x_train)
x_test=sc.fit_transform(x_test)

In [21]: M #import dump class from joblib
from joblib import dump
dump(sc,"transform.save")
Out[21]: ['transform.save']

In [22]: M from sklearn.ensemble import GradientBoostingClassifier
xg_cla = GradientBoostingClassifier(learning_rate = 0.1,
max_depth = 5, n_estimators = 10)

In [23]: M #fit the model
xg_cla.fit(x_train,y_train)
C:\Users\gati\eng\21fwanaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
return f(*args, **kwargs)
Out[23]: GradientBoostingClassifier(max_depth=5, n_estimators=10)

In [24]: M #predictions with unseen data by model
y_pred_xgb = xg_cla.predict(x_test)
y_pred_xgb
Out[24]: array([1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0,
0, 0, 0, 0, 0, 0], dtype=int64)
```

```
In [25]: #checking the accuracy score
from sklearn.metrics import accuracy_score,confusion_matrix
acc=accuracy_score(y_test,y_pred_xgb)
acc

Out[25]: 0.9655172413793104

In [26]: #summary of predictions
cm2=confusion_matrix(y_test,y_pred_xgb)
cm2

Out[26]: array([[25,  1],
       [ 0,  3]], dtype=int64)

In [27]: from sklearn.metrics import mean_squared_error
rmse = np.sqrt(mean_squared_error(y_test, y_pred_xgb))
print("RMSE: %.4f" % (rmse))

RMSE: 0.185695

In [28]: dataset.head()
Out[28]:
```

	Temp	Humidity	Cloud Cover	ANNUAL	Jan-Feb	Mar-May	Jun-Sep	avginne	sub	flood
0	29	70	30	3248.6	73.4	388.2	2122.8	274.866667	649.9	0
1	28	75	40	3226.6	9.3	275.7	2402.4	130.200000	356.4	1
2	28	75	42	3271.2	21.7	338.3	2343.0	186.200000	308.9	0
3	29	71	44	3129.7	26.7	339.4	2398.2	368.666667	862.5	0
4	31	74	40	2741.6	23.4	378.5	1881.5	283.400000	586.9	0

```
In [29]: rand_pred = xg_cla.predict(sc.transform([[30,3248.6,73.4,386.2,2122.8]]))
rand_pred

Out[29]: array([0], dtype=int64)

In [30]: rand_pred = xg_cla.predict(sc.transform([[40,3326.6,9.3,275.7,2403.4]]))
rand_pred

Out[30]: array([1], dtype=int64)
```

```
In [31]: from sklearn.tree import DecisionTreeClassifier
dtc = DecisionTreeClassifier()
dtc.fit(x_train,y_train)
Out[31]: DecisionTreeClassifier()

In [32]: y_predict = dtc.predict(x_test)
y_predict

Out[32]: array([1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0,
       0, 0, 0, 0, 0, 0], dtype=int64)

In [33]: from sklearn.metrics import accuracy_score,confusion_matrix

In [34]: acc=accuracy_score(y_test,y_predict)
acc

Out[34]: 0.9655172413793104

In [35]: acc=accuracy_score(y_test,y_pred_xgb)
acc

Out[35]: 0.9655172413793104

In [36]: cm1=confusion_matrix(y_test,y_predict)
cm1

Out[36]: array([[25,  1],
       [ 0,  3]], dtype=int64)

In [37]: cm2=confusion_matrix(y_test,y_pred_xgb)
cm2

Out[37]: array([[25,  1],
       [ 0,  3]], dtype=int64)

In [38]: #saving the file
from joblib import dump
dump(xg_cla,'floods.save')
Out[38]: ['floods.save']
```

Spyder (Python 3.8)

File Edit Search Source Run Debug Consoles Projects Tools View Help

C:\Users\gatti\_mhg71lf\Desktop\Floods-Prediction-using-Machine-Learning-main\Flask\app.py

```

1  from flask import Flask, render_template, request
2  # used to run/serve our application
3  # render_template is used for rendering the html pages
4  # import load from joblib to load the saved model file
5  from joblib import load
6
7  app=Flask(__name__) # our flask app
8  #load model file
9  model =load('C:/Users/gatti_mhg71lf/Desktop/Floods-Prediction-using-Machine-Learning-main/Flask'
10 sc=load('C:/Users/gatti_mhg71lf/Desktop/Floods-Prediction-using-Machine-Learning-main/Flask/tr
11
12
13 @app.route('/') # rendering the html template
14 def home():
15     return render_template('home.html')
16 @app.route('/predict') # rendering the html template
17 def index():
18     return render_template("index.html")
19
20
21 @app.route('/data_predict', methods=['POST']) # route for our prediction
22 def predict():
23     temp = request.form['temp']
24     Hum = request.form['Hum']
25     db = request.form['db']
26     ap = request.form['ap']
27     aa1 = request.form['aa1']
28
29     data = [[float(temp),float(Hum),float(db),float(ap),float(aa1)]]
30     prediction = model.predict(sc.transform(data))
31     output=prediction[0]
32     if(output==0):
33         return render_template('noChance.html', prediction='No possibility of severe flood')
34     else:
35         return render_template('chance.html', prediction='possibility of severe flood')
36
37 if __name__ == '__main__':
38     app.run(debug=False)

```

Variable explorer Help Plots Files

Console 1/A

Python 3.8.8 (default, Apr 13 2021, 15:08:03) [MSC v.1916 64 bit (AMD64)]  
Type "copyright", "credits" or "license" for more information.  
IPython 7.22.0 -- An enhanced Interactive Python.  
In [1]:

34°C Partly sunny

Spyder (Python 3.8)

File Edit Search Source Run Debug Consoles Projects Tools View Help

C:\Users\gatti\_mhg71lf\Desktop\Floods-Prediction-using-Machine-Learning-main\Flask\app.py

```

1  from flask import Flask, render_template, request
2  # used to run/serve our application
3  # render_template is used for rendering the html pages
4  # import load from joblib to load the saved model file
5  from joblib import load
6
7  app=Flask(__name__) # our flask app
8  #load model file
9  model =load('C:/Users/gatti_mhg71lf/Desktop/Floods-Prediction-using-Machine-Learning-main/Flask'
10 sc=load('C:/Users/gatti_mhg71lf/Desktop/Floods-Prediction-using-Machine-Learning-main/Flask/tr
11
12
13 @app.route('/') # rendering the html template
14 def home():
15     return render_template('home.html')
16 @app.route('/predict') # rendering the html template
17 def index():
18     return render_template("index.html")
19
20
21 @app.route('/data_predict', methods=['POST']) # route for our prediction
22 def predict():
23     temp = request.form['temp']
24     Hum = request.form['Hum']
25     db = request.form['db']
26     ap = request.form['ap']
27     aa1 = request.form['aa1']
28
29     data = [[float(temp),float(Hum),float(db),float(ap),float(aa1)]]
30     prediction = model.predict(sc.transform(data))
31     output=prediction[0]
32     if(output==0):
33         return render_template('noChance.html', prediction='No possibility of severe flood')
34     else:
35         return render_template('chance.html', prediction='possibility of severe flood')
36
37 if __name__ == '__main__':
38     app.run(debug=False)

```

Variable explorer Help Plots Files

Console 1/A

WARNING: This is a development server. Do not use it in a production deployment.  
Use a production WSGI server instead.  
\* Debug mode: off  
C:\Users\gatti\_mhg71lf\anaconda3\lib\site-packages\sklearn\base.py:310: UserWarning: Trying to unpickle estimator DummyClassifier from version 0.23.0 when using version 0.24.1. This might lead to breaking code or invalid results. Use at your own risk.  
warnings.warn(C:\Users\gatti\_mhg71lf\anaconda3\lib\site-packages\sklearn\base.py:310: UserWarning: Trying to unpickle estimator DecisionTreeRegressor from version 0.23.0 when using version 0.24.1. This might lead to breaking code or invalid results. Use at your own risk.  
warnings.warn(C:\Users\gatti\_mhg71lf\anaconda3\lib\site-packages\sklearn\base.py:310: UserWarning: Trying to unpickle estimator GradientBoostingClassifier from version 0.23.0 when using version 0.24.1. This might lead to breaking code or invalid results. Use at your own risk.  
warnings.warn(C:\Users\gatti\_mhg71lf\anaconda3\lib\site-packages\sklearn\base.py:310: UserWarning: Trying to unpickle estimator StandardScaler from version 0.23.0 when using version 0.24.1. This might lead to breaking code or invalid results. Use at your own risk.  
warnings.warn(\* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)

Run file

34°C Partly sunny