

FLASK INTEGRATION DOCUMENT

Flask is a web application framework written in Python.

Directory Structure:

Project Directory >>

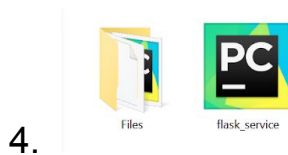
- flask_service.py
- **Files >>**
 - roll_no.pkl (Model Pickle File)
 - roll_no.py (Preprocessing and Feature Engineering Python File)

Flask Zip File:

- First and foremost download the Flask zip file from the noticeboard.
- Extract them and store it in the Project Directory.
- Rename all the files in the **FILES** directory by your roll_number.
 - Eg. 1405043.py → 17*****.py
 - And the class name as well in that file.
 - class _1405043(): → class _17*****():
- Before you proceed with Flask Deployment, if you have applied multiple models in your notebook. Select only one and remember the features passed in that.
 - Eg. Linear Regression:
 - Features List: amount, days, mean, median.
 - During Deployment, remember the features list and how to recreate them only. No Feature selection in Deployment.
- Go to your final notebook, select which fitted model you are going to use. Create a pickle file. Follow the Code snippet below.
- And store the Pickle (.pkl) file in the directory mentioned above.

Installation:

1. Open CMD / Conda Prompt / Jupyter Notebook (If CMD not working, try CONDA Prompt)
2. Type “pip install Flask” in CMD/Conda Prompt
(<https://pypi.org/project/Flask/>)
3. After installation, please cross check your Project Directory Structure



5. Inside **Files** Folder, there should be two things present ->
 - a. Pickle File (Incase you don't have pickle library, pip install pickle):

```
import pickle
filename = 'roll_number.pkl'
pickle.dump(model, open(filename, 'wb'))
```
 - b. Preprocessing & Feature Engineering Script: roll_number.py

Integration:

6. Just like any python program, In your CMD or Anaconda Prompt whichever available run ***python flask_service.py***. This ***flask_service.py*** command runs a Python file and sets `__name__ == "__main__"`. If the main block calls `app.run()`, it will run the development server.
7. Flask_service.py is the flask application. Inside it we have a route named ***'/predict'***, which will be called from the client side. The route is responsible for returning the output from the model in the form of a json object.
8. Inside ***'predict'*** we instantiate an object of ***'_rollno'*** class and call `getPredictions` function which will return us the result.

9. Inside `_rollno` class we have written different transformation functions which will do the preprocessing and feature engineering , followed by obtaining predictions from the prebuilt model(`roll_no.pkl`) and return the predicted value to `Flask_service`.
10. Once the Flask Server is up and running you can call it from React by using the below function:

```
export function prediction(data) {  
  return axios.post(  
    'http://127.0.0.1:5000/predict?',  
    {},  
    {  
      headers: { 'Content-Type': 'application/json' },  
      params: {  
        data: data,  
      },  
    }  
  );  
}
```

11. The format of the data sent should be:

```
{  
  "id": "<ROLL_NO>",  
  "data": [  
    {  
      .....  
    },  
    {  
      .....  
    }  
  ]  
}
```

12. The column id to display the response from Flask should be 'predictions' and 'predicted_payment_type'.