Data Science Intern at Data Glacier

Week 4: Deployment on Cloud

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1. Introduction

In this project, we will deploy a machine learning model (Linear Regression) using the Flask Framework on the cloud. As a demonstration, our model helps to predict Graduate admission. We will first build a machine learning model for prediction, then create an API for the model, using Flask, the Python micro-framework for building web applications, and lastly host the application on pythonwnywhere.com. This API allows us to utilize predictive capabilities through HTTP requests.

2. Data Information

This dataset is created for the prediction of Graduate Admissions from an Indian perspective. This dataset was built with the purpose of helping students in shortlisting universities with their profiles. The predicted output gives them a fair idea about their chances at a particular university. The dataset contains several parameters considered necessary during the application for Masters's Programs.

The parameters included are:

- 1. GRE Scores (out of 340)
- 2. TOEFL Scores (out of 120)
- 3. University Rating (out of 5)
- 4. Statement of Purpose and Letter of Recommendation Strength (out of 5)
- 5. Undergraduate GPA (out of 10)
- 6. Research Experience (either 0 or 1)

7. Chance of Admit (ranging from 0 to 1)

3. Building A Model

3.1 Importing Require Libraries and Dataset.

```
#Import Libraries and packages
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
#Import dataset taken from kaggle
df = pd.read_csv('Admission_Data.csv')
df.head()
   Serial No. GRE Score TOEFL Score University Rating SOP LOR CGPA Research Chance of Admit
                                              4 4.5 4.5
                                                           9.65
         2
                                              4 4.0 4.5
                                                           8.87
                                                                                   0.76
1
                                                                      1
                                              3 3.0 3.5
2
         3
                 316
                             104
                                                           8.00
                                                                                   0.72
                                              3 3.5 2.5
3
         4
                 322
                             110
                                                           8.67
                                                                       1
                                                                                   0.80
                                                           8.21
                                              2 2.0 3.0
#Information about dataset
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 9 columns):
                     Non-Null Count Dtype
# Column
0 Serial No. 500 non-null
1 GRE Score 500 non-null
2 TOEFL Score 500 non-null
                      500 non-null int64
                                      int64
                                      int64
float64
    University Rating 500 non-null
   SOP
                        500 non-null
                       500 non-null float64
                                      float64
   CGPA
                      500 non-null
    Research
                       500 non-null
                                        int64
8 Chance of Admit 500 non-null
                                        float64
dtypes: float64(4), int64(5)
memory usage: 35.3 KB
```

- There are a total of 8 useful features in this dataset, and each of the features has 500 records.
- There are no null or missing values in this dataset.

3.2 Build Model

We will split data into 80% for the training set and the remaining 20% for the test set. Here we implement a machine learning model to predict the applicant's chance of admission. For this

purpose, we will use a Multiple linear regression model using scikit-learn. After importing and initializing the LinearRegression model we fit it into the training dataset.

```
# split dependent and independent variables
X = df.iloc[:,:-1]
y = df.iloc[:,-1]

# split data into train and test set
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.20, random_state=42)

# Model Initilization & training
lm = LinearRegression()
lm.fit(X_train,y_train)

* LinearRegression
LinearRegression()

#Evaluating the Model Performance - Multiple Linear Regression
from sklearn.metrics import r2_score
predictions = lm.predict(X_test)
print('R^2 Score: ', r2_score(y_test,predictions))
R^2 Score: 0.8188432567829628
```

3.3 Save the Model

We saved our model using pickle.

```
import pickle
pickle.dump(lm, open("model.pkl","wb"))
```

4. Build a Flask application

Now we will develop a web application that consists of a simple web page with a form field that lets us enter a few applicants' scores. After submitting the features to the web application, it will tell us the applicants' chances of admission.

Flask is a micro web framework written in Python. Micro-framework because it does not require a particular tool or library. However, it supports extensions that can add application features. The flask is lightweight and particularly suitable for small and mid-size model deployment.

4.1 App.py

The app.py file contains the main code that will be executed by the Python interpreter to run the Flask web application, it uses t pickle to import the ML model for prediction.

```
from flask import Flask, jsonify, request, render_template
import pickle
import pandas as pd
import numpy as np

app = Flask(_name__)

model = pickle.load(open("model.pkl",'rb'))

@app.route('/')
def home():
    return render_template("index.html")

@app.route('/predict', methods = ['POST'])
def predict():
    float_features = [float(x) for x in request.form.values()]
    features = [np.array(float_features)]
    pred_price = model.predict(features)
    return render_template("index.html",prediction_text = "The Chances of Admissions are : {}".format(pred_price))

#driver function
if __name__ == '__main__':
    app.run(debug=True)
```

4.2 Index.html

The following are the contents of the index.html file that will render a text form where a user can enter the required attributes for prediction.

```
<!DOCTYPE html>
<!--From https://codepen.io/frytyler/pen/EGdtg-->
<head>
  <meta charset="UTF-8">
  <title>ML API</title>
<link href='https://fonts.googleapis.com/css?family=Open+Sans+Condensed:300' rel='stylesheet' type='text/css'>
</head>
 <div class="login">
    <h1>USA College Admission Prediction</h1>
    <input type= text name= !OEFL Score placeholder= !OEFL Score required= 'required '/'
<input type="text" name="University Rating" placeholder="University Rating" required="required" />
<input type="text" name="SOP" placeholder="SOP" required="required" />
<input type="text" name="LOR" placeholder="LOR" required="required" />
<input type="text" name="LOR" placeholder="LOR" required="required" />

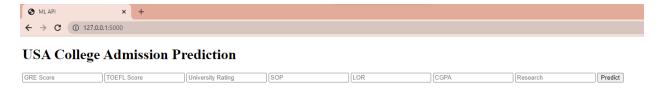
         <input type="text" name="CGPA" placeholder="CGPA" required="required"</pre>
         <input type="text" name="Research" placeholder="Research" required="required" />
         <button type="submit" class="btn btn-primary btn-block btn-large">Predict</button>
    </form>
   {{ prediction_text }}
</html>
```

4.3 Running Procedure

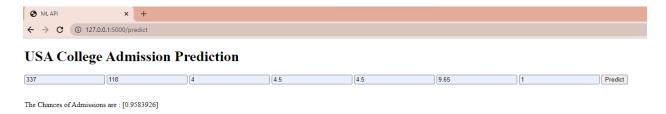
By default, the flask executes at port 5000 on the local host. When you execute this flask app and open the URL http://localhost:5000 in the browser, you will see the web app running. Once we have done all of the above, we can start running the API by executing the command from the Terminal:

```
chitr@Jarvis MINGW64 ~/Documents/DataGlacier/Flask/ModelDeployment_Flask (main)
$ python app.py
  * Serving Flask app "app" (lazy loading)
  * Environment: production
  WARNING: This is a development server. Do not use it in a production deployme
nt.
  Use a production WSGI server instead.
  * Debug mode: on
  * Restarting with watchdog (windowsapi)
  * Debugger is active!
  * Debugger PIN: 106-972-514
  * Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

Now we could open a web browser and navigate to http://127.0.0.1:5000/, we should see a simple website like shown below



Now we need to input the different parameters. After entering the input click the predict button and can see the result for chances of admission.



5. Host your application on the cloud (pythonwnywhere.com)

Now we will host our flask application using <u>pythonanywhere.com</u>. It is customized for python based framework deployment such as Django and Flask and is very easy & quick to deploy. Below are steps to host our flask app on <u>pythonanywhere.com</u>.

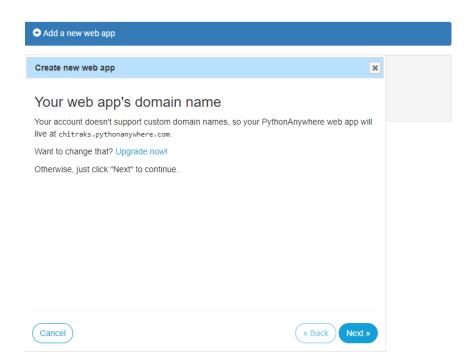
5.1 Setup virtual environment:

First, we need to set up a virtual environment and install the required dependencies. Then navigate to the console tab, start a new bash console, and execute the below commands in the terminal to install sklearn and flask.

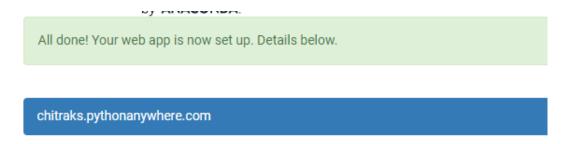
```
14:29 ~ $ mkvirtualenv --python=/usr/bin/python3.7 my-virtualenv
created virtual environment CPython3.7.13.final.0-64 in 31868ms
(my-virtualenv) 14:31 ~ $ pip install sklearn
(my-virtualenv) 14:32 ~ $ pip install flask
```

5.2 Add a new web app

I used Python 3.7 for this model deployment.



After creating the web app, we got a URL that points to the flask endpoint. By default, it will display a message saying "Hello, World!". Your endpoint looks something like this: [username].pythonanywhere.com





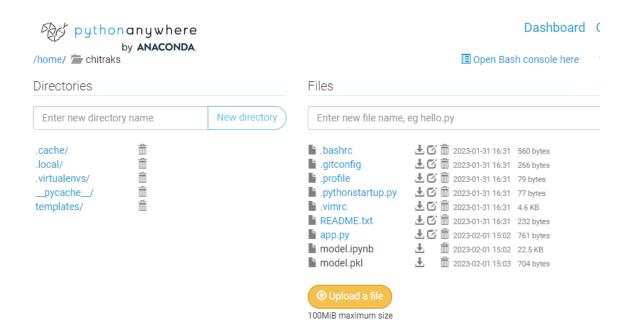
Hello, World!

This is the default welcome page for a PythonAnywhere hosted web application.

Find out more about how to configure your own web application by visiting the web app setup page

5.3. Upload the files

Now we need to upload the flask application file(app.py), saved model file (model.pkl), and template directory with index.html inside the specified project directory path.



5.4. Customizing web app for flask application :

Now we will customize the flask endpoint API to show our admission prediction model under the WSGI configuration file(under web > code), uncomment the flask portion, set the project directory path, and update the flask app name.

```
Code:

What your site is running.

Source code:

Enter the path to your web app source code

Working directory:

/home/chitraks/

→ Go to directory

WSGI configuration file:

/var/www/chitraks_pythonanywhere_com_wsgi.py

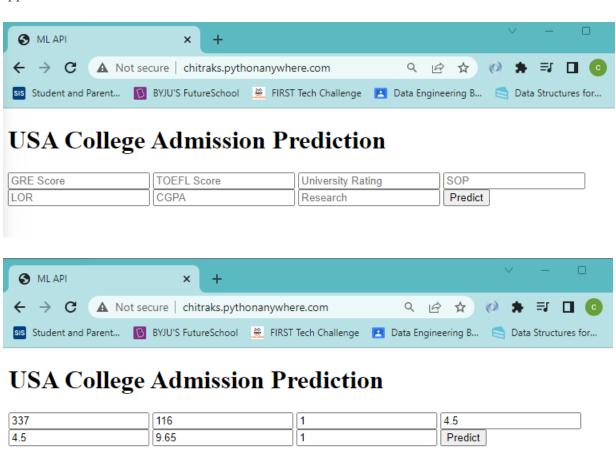
Python version:
```

Below is the extract of the WSGI configuration file for our flask application setup.

```
92
  93 # ++++++++ FLASK ++++++++
 94 # Flask works like any other WSGI-compatible framework, we just need
  95 # to import the application. Often Flask apps are called "app" so we
  96 - # may need to rename it during the import:
 97 #
 98
 99 import sys
 100
 101 # The "/home/chitraks" below specifies your home
 102 # directory -- the rest should be the directory you uploaded your Flask
 103 # code to underneath the home directory. So if you just ran
 104 # "git clone git@github.com/myusername/myproject.git"
 105 # ...or uploaded files to the directory "myproject", then you should
 106 # specify "/home/chitraks/myproject"
 107 path = '/home/chitraks/'
 108 - if path not in sys.path:
109
          sys.path.append(path)
110
111 from app import app as application # noga
112
```

5.5. Reload the web app

An endpoint will now act as an API to facilitate other applications. An example front-end application





USA College Admission Prediction



The Chances of Admissions are: [0.94469446]