notebook

April 15, 2021

1 Web App Deployment Using Flask

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Date: 2021-03-22

1.1 Introduction

This report presents my approach in deploying my web app using Flask, the micro Web app frame work on python. As a demonstration, the app asks the client to select a polynomial degree from a top down list, and subsequently plot a polynomial regression on the airquality data set.

1.2 Importing libraries

```
[]: from flask import Flask, request, render_template #Web app framework import numpy as np import pandas as pd #Scikit learn for regression model from sklearn.model_selection import train_test_split from sklearn.metrics import mean_squared_error, r2_score from sklearn.linear_model import LinearRegression from sklearn.preprocessing import PolynomialFeatures import matplotlib.pyplot as plt #To create plots from io import BytesIO #To encode plots and pass them to render_template import base64
```

1.3 Home page

The home page prompts the client to choose an order for the degree polynomial to build the model. A HTTP GET request is issued (refer to layout.html), and the url is then parsed using the imported function request() from Flask, as seen below.

```
[]: #Create app using flask
app = Flask(__name__)

@app.route('/')
```

```
def index():
    #Parse html for "GET" request with client's input
    poly_order = request.args.get("poly_order", 1, type=int)
    #Load data
    airquality = pd.read_csv('airquality.csv')
    airquality.dropna(inplace=True)
    data = airquality[['Temp', 'Ozone']]
    #Build model
    plot_url, r2 = build_model(poly_order, data)
    return render_template('layout.html', tables=[data.head(6).
    →to_html(classes='data', header=True)], r2=r2, plot_url=plot_url.
    →decode('utf8'), chosen_order=poly_order)
```

1.4 Model building

The client's selection is then passed to the function *build_model()* where the polynomial regression model is built as seen below.

```
[]: def build model(poly order, data):
         x = data['Ozone']
         x = x[:, np.newaxis]
         y = data['Temp']
         y = y[:, np.newaxis]
         #Create polynomial terms out of 1D array
         poly_features = PolynomialFeatures(degree = poly_order)
         x_poly = poly_features.fit_transform(x)
         #Build the model
         model = LinearRegression()
         model.fit(x_poly, y)
         y_poly_pred = model.predict(x_poly)
         #Training Error
         r2 = r2_score(y, y_poly_pred)
         #Sort axis and create plot
         plt.scatter(data['Ozone'], data['Temp'], s=10)
         sorted_zip = sorted(zip(data['Ozone'], y_poly_pred))
         x, y_poly_pred = zip(*sorted_zip)
         plt.plot(x, y_poly_pred, color='m')
         plt.xlabel('Ozone (ppb)')
         plt.ylabel('Temp (Fahrenheit)')
         #Save plot to return to html
         img = BytesIO()
         plt.savefig(img, format='png')
         plt.close()
         img.seek(0)
```

```
#Encode figure and pass to html in index function
plot_url = base64.b64encode(img.getvalue())
return(plot_url, r2)

#Run app with debug
if __name__ == "__main__":
app.run(debug=True)
```

2 Results

Running the app using Flask from VScode IDE.

```
[1]: from IPython.display import Image
Image(filename= "snap1.jpg", width=900)
```

[1]:

```
| The Life Selection View Go Run Terminal Help | Namency Visual States Code (Administrator) | Namency | Na
```

The image below shows the web app. The user is prompted by the selection box surrounded by **red box** to choose a polynomial degree, and the output is the regression plot surrounded by **green rectangle**.

```
[3]: Image(filename= "webapp.jpg", width=900)
[3]:
```

