

Week 4: ML model Deployment on Flask

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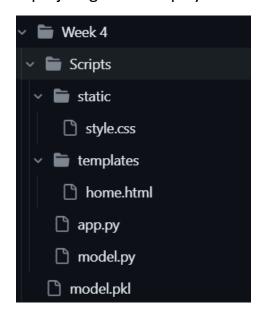
Submission Date: 27/10/2023

Submitted To: GitHub

1. Introduction

The aim of this project is to build a machine learning model and deploy it on Flask. For this purpose, we selected a diabetes dataset which is provided by Sklearn and used it to train a Linear Regression model for prediction of diabetes progression after 1 year of baseline.

The folder structure of the project goes as displayed below:



The project files, in order, of this project include:

- model.py Predicts the progression of diabetes and saves the model as model.pkl.
- app.py Contains APIs to fetch data input by user on webpage and calculate the predicted result.
- home.html Displays the UI for the user to input the values of each feature.
- **style.css** Stylesheet for the webpage.

2. Explaining the files

This section will describe **model.py**, **app.py**, and **home.html** file and their outcomes in detail.

a. Model.py

The dataset acquired from Sklearn contains 10 features and 1 target. The features include Age, Sex, Body Mass Index (BMI), Blood Pressure (BP), Total Serum Cholesterol (TC), Low-Density Lipoproteins (LDL), High-Density Lipoproteins (HDL), Total Cholesterol (TCH), Possibly log of serum triglycerides level (LTG), and Blood Sugar level (GLU). The target feature is a measure of the disease progression 1 year after the baseline. The values have already been mean centred and scaled by the standard deviation times of square root of number samples i.e., 442. Hence, the values lie between 1 and -1.

```
age sex bmi bp tc ldl hdl tch ltg glu target 0 0.038076 0.050680 0.061696 0.021872 -0.044223 -0.034821 -0.043401 -0.002592 0.019908 -0.017646 151.0 1 -0.001882 -0.044642 -0.051474 -0.026328 -0.008449 -0.019163 0.074412 -0.039493 -0.068330 -0.092204 75.0 2 0.085299 0.050680 0.044451 -0.005671 -0.045599 -0.034194 -0.032356 -0.002592 0.002864 -0.025930 141.0 3 -0.089063 -0.044642 -0.011595 -0.036656 0.012191 0.024991 -0.036038 0.034309 0.022692 -0.009362 206.0 4 0.005383 -0.044642 -0.036385 0.021872 0.003935 0.015596 0.008142 -0.002592 -0.031991 -0.046641 135.0
```

We split the dataset into train – to train the model on and test – to test the model on, and use Linear Regression for prediction.

The trained model is then saved using **Pickle** library. This model can be later loaded in **app.py** file for prediction of user inputs.

```
pkl.dump(lr_model, open('../model.pkl', 'wb'))
```

b. App.py

This is the most important python file which uses **Flask** for model deployment and creating the APIs. The model is loaded in this file for prediction of the user inputs.

```
# Load the model
model = pkl.load(open('..\model.pkl', 'rb'))
```

We import flask libraries such as **Flask**, **request** and **render_template**. The below code will direct to the homepage of the web application.

```
# Route to the homepage
@app.route('/')
def home():
    return render_template('home.html')
```

The next code below is triggered when the user has input all the required features on the form appearing on the homepage and submits the form. As described in the code, the form values are fetched via request.form.values() function and store the floated values as an array since the model takes the input features as an array. The loaded model predicts the input values and stores the value as result. We round the resulting value up to 3 decimal points. The render_template() then populates the result on the homepage under the submit button.

c. Home.html

This section describes the homepage of the web application. When we run the **app.py** python file, the console will generate the local url – 'http://127.0.0.1:5000/' for the web page.

```
* Serving Flask app 'app'

* Debug mode: on

WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead.

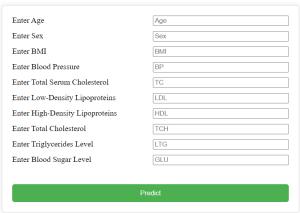
* Running on http://127.0.0.1:5000

Press CTRL+C to quit

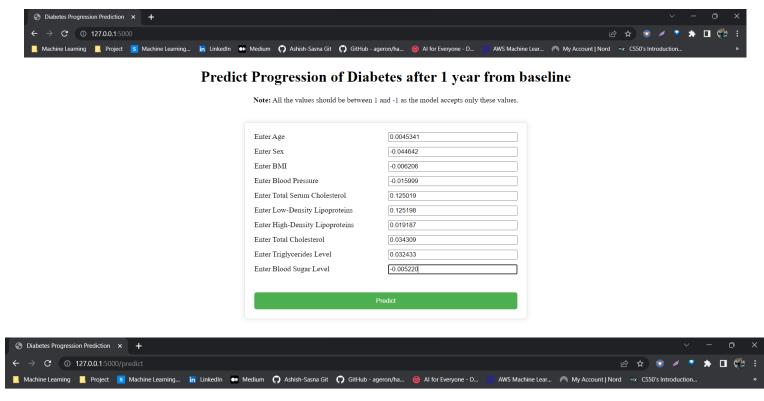
* Restarting with watchdog (windowsapi)
```

We can open the homepage by accessing this URL in a browser.





The webpage shows the form that the users can access to enter the values as displayed. Once the field are filled, the user can submit the form using **Predict** button at the bottom. This will trigger the predict API in the **app.py** file and calculate the results.



Predict Progression of Diabetes after 1 year from baseline

Note: All the values should be between 1 and -1 as the model accepts only these values. Enter Age 0.0045341 -0.044642 Enter Sex Enter BMI -0.006206 Enter Blood Pressure -0.015999 Enter Total Serum Cholesterol 0.125019 Enter Low-Density Lipoproteins 0.125198 Enter High-Density Lipoproteins 0.019187 Enter Total Cholesterol 0.034309 Enter Triglycerides Level 0.032433 Enter Blood Sugar Level -0.005220

Diabetes progression result is 138.001

As you can see, after hitting the **Predict** button, the result is displayed below. This can be explained by the html code in below figure. The **input** tag takes the input from the user for all the 10 features. After submitting the form, the **/predict** API will calculate the result and render the result at **{{prediction_result}}** section.

```
<div class="predict">
    <form action="/predict" method="POST">
        Enter Age <input type="text" name="age" placeholder="Age" value=
        Enter Sex <input type="text" name="sex" placeholder="Sex" value=
        Enter BMI <input type="text" name="bmi" placeholder="BMI" value=
        Enter Blood Pressure <input type="text" name="bp" placeholder="B
        Enter Total Serum Cholesterol <input type="text" name="tc" place
        Enter Low-Density Lipoproteins <input type="text" name="ldl" pla
        Enter High-Density Lipoproteins <input type="text" name="hdl" pl
        Enter Total Cholesterol <input type="text" name="tch" placeholde
        Enter Triglycerides Level <input type="text" name="ltg" placehol
        Enter Blood Sugar Level <input type="text" name="glu" placeholde
        <br>
        <button type="submit">Predict</button>
    </form>
</div>
<div class="result" id="predictionResult">
    {{ prediction_result }}
```