**Data Science Intern AT Data Glacier** 

Week 5: Cloud API development

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# Model Deployment using Heroku

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### 1.Introduction

In this project I will deploy the Logistic regression model using Flask. My ML model predicts if someone is diabetic or not. The goal of this project is to create API for the ML model using Flask. Flask is a framework for building a web application.

### 2.Data information

Researchers at the Bio-Solutions lab want to get better understanding of diabetes disease among women and are planning to use machine learning models that will help them to identify patients who are at risk of diabetes. The "Pima Indians Diabetes.csv" dataset was collected. All patients here are females at least 21 years old of Pima Indian heritage.

Our primarily ML model shows the most important features to make the prediction; Therefore, I filtered the dataset to include only the important features then we rebuild the model on the filtered Dataset.

Pregnancie	int64					
Glucose:	768 non-null	float64				
BMI:	768 non-null	float64				
Pedigree:	768 non-null	float64				
Class:	768 non-null	int64				
dtypes:	dtypes: float64(3), int64(2)					
memory usage: 30.1 KB						

### **Data Description:**

- Pregnancies: Number of times pregnant
- Glucose: Plasma glucose concentration over 2 hours in an oral glucose tolerance test
- BMI: Body mass index (weight in kg/(height in m)^2)
- Pedigree: Diabetes pedigree function A function that scores likelihood of diabetes based on family history.
- Class: Class variable (0: the person is not diabetic or 1: the person is diabetic)

Index	Pregnancies	Glucose	BMI	Pedigree	Class
566	1	99.0	38.6	0.412	0
207	5	162.0	37.7	0.151	1
58	0	146.0	40.5	1.781	0
686	3	130.0	23.1	0.314	0

751	1	121.0	39.0	0.261	0

### 3. The Machine learning model

### **Import the libraries:**

```
# To filter the warnings
import warnings
warnings.filterwarnings("ignore")
# Libraries to help with reading and manipulating data
import pandas as pd
import numpy as np
# libaries to help with data visualization
import matplotlib.pyplot as plt
import seaborn as sns
# Library to split data
from sklearn.model selection import train test split
from sklearn.linear_model import LogisticRegression
# To build linear model for statistical analysis and prediction
# To get diferent metric scores
from sklearn import metrics
from sklearn.metrics import accuracy score, roc curve, confusion matrix, roc auc score, recall score
import pickle
Read the dataset
data = pd.read csv("pima-indians-diabetes.csv")
# defining the most important columns and replace 0 with NaN
cols = ["Glucose", "BMI", "Pedigree"]
# replacing 0 with NaN
data[cols] = data[cols].replace(0, np.nan)
# Let's impute missing values using mean value
data[cols] = data[cols].fillna(data[cols].mean())
#Target and independent features.
X = data.drop(["Class"], axis=1)
Y = data["Class"]
# split data into training and test
X_train,X_test,y_train,y_test=train_test_split(X,Y,test_size=0.2,random_state=1,stratify=Y)
Logistic Regression model
lg=LogisticRegression(penalty='elasticnet',l1_ratio=0.9,max_iter=500,solver='saga',class_weight='balan
ced')
lg.fit(X train,y train)
# convert the model pickle.
pickle.dump(lg, open('model.pkl','wb'))
```

# 4. Applying ML Model to flask framework

### A. Creating application files

We create a folder for the application. The directory tree inside this folder as follow:

# week4 | app.py | logistic\_regression\_diabetes.py | Logistic\_Regression\_Hands-On.ipynb | model.pkl | pima-indians-diabetes.csv | request.py | -----static | script.js | style.css | templates | index.html

### o APP.Py

App.py contains the main code that will be run by python. It includes the Machine leaning model and route the Index.html. the route decorator used to map the URL to a return value that means connect url to return value of a function.

```
from webbrowser import get
     import numpy as np
 3
     import pandas as pd
    import pickle
    from flask import Flask, request, render_template, jsonify
7
     app=Flask(__name__)
     model=pickle.load(open('model.pkl','rb'))
9
     @app.route('/') # take to the out page
10
    def home():
     return render_template('index.html')
11
     @app.route('/predict',methods=['post','get'])
12
13
     def predict():
14
         For rendering results on HTML GUI
15
16
         float_features = [float(x) for x in request.form.values()]
17
18
        final features = [np.array(float features)]
         prediction = model.predict(final features)
19
20
21
        output = round(prediction[0], 2)
22
         if output == 1:
             output_text='Diabetic'
23
24
         elif output == 0:
25
            output_text='Not Diabetic'
26
        return render_template('index.html',prediction_text=f'You are {output_text}')
27
28
     if __name__ == "__main__":
29
        app.run(debug=True)
```

### o Index.html

The index file is a html file. Index.html contains HTML code or tags with some CSS and JavaScript code. It should be in templates folder. When the web site is requested, by default index.html file is returned.

```
templates > O index.html
      <!DOCTYPE html>
  1
      <html >
  2
      <head>
 3
  4
       <title>Machine Learning Model</title>
  5
      <link rel="stylesheet" href="{{ url_for('static', filename='css/style.css') }}"></link>
  6
      <script type="test/javascript" src="{{ url_for('static', filename='script/script.js') }}">
  8
      </script>
  9
      </head>
 10
 11
      <body>
 12
       <div class="container">
 13
        <h1>Predict Diabetes</h1>
 14
          <!-- Main Input For Receiving Query to our ML -->
 15
          <form action="{{ url_for('predict')}}"method="post">
 16
            <input type="text" name="Pregnancies" placeholder="Pregnancies" required="required" />
 17
             <input type="text" name="Glucose" placeholder="Glucose" required="required" />
 18
          <input type="text" name="BMI" placeholder="BMI" required="required" />
 19
              <input type="text" name="Pedigree" placeholder="Pedigree" required="required" />
 20
 21
 22
              <button type="submit" class="btn btn-primary btn-block btn-large">Predict</button>
          </form>
 23
 25
         <br>
         <br>
 27
        {{ prediction_text }}
 28
       </div>
 29
 30
 31
 32
      </body>
```

### o style.css

CSS file is necessary to determine how the HTML API looks and it must be saved in static folder.

```
# stylecss > % container
@Import url(https://fonts.googleapis.com/css?family=Open+Sans);
.btn { display: inline-block; *display: inline; *zoom: 1; padding: 4px 10px 4px; margin-bottom: 0; font-size: 13px; line-height: .btn:hover, .btn:active, .btn.active, .btn.disabled, .btn[disabled] { background-color: □ #e66666; }
.btn-large (padding) px 14px; font-size: 28px; line-height: nomeal; webkit-border-radius: 5px; more-border-radius: .btn:hover { color: ■#33333; text-decoration: none; background-color: □ #e66666; background-spsition: 0 -15px; -webkit-transiti .btn-primary, .btn-primary: .btn-primary:
```

### B. Running App.py

We can run the application either by double click and app.py or by run the command in terminal.

Running with debugger shows the following results.

```
* Serving Flask app 'app' (lazy loading)

* Environment: production

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

Use a production WSGI server instead.

* Debug mode: on

* Restarting with stat

* Debugger is active!

* Debugger PIN: 808-315-324

* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

o The output looks like this:



O We fill the input values, and the output will be like this.



# 6. Model deployment using Heroku

After building the machine leaning model and application using flask, we are ready to deploy the model using Heroku. Heroku is a cloud platform supporting several programming languages. One of the first cloud platforms and it support most of the well known programing languages.

To Deploy python code in Heroku. We need two main files: Requirements.txt/setup.txt and Profile.

### > requirements file

it contains all the python packages required to execute the application.

```
asttokens==2.0.8
 attrs==22.1.0
backcall==0.2.0
backports.functools-lru-cache==1.6.4
bleach==5.0.1
 certifi==2022.9.14
charset-normalizer==2.1.1
click==8.0.4
colorama==0.4.5
commonmark==0.9.1
coverage==6.4.4
debugpy==1.6.3
decorator==5.1.1
 docopt==0.6.2
docutils==0.19
entrypoints==0.4
ERAlchemy==1.2.10
executing==1.0.0
Flask==2.1.3
fonttools==4.37.1
greenlet==1.1.3
gunicorn==20.1.0
heroku==0.1.4
 idna==3.3
imblearn==0.0
importlib-metadata==4.12.0
iniconfig==1.1.1
ipykernel==6.15[3]
 ipython==8.4.0
itsdangerous==2.0.1
jaraco.classes==3.2.2
jedi==0.18.1
Jinja2==3.0.3
  ioblib==1.2.6
 jupyter_client==7.3.5
jupyter_core==4.11.1
keyring==23.9.1
 kiwisolver==1.4.4
```

### > Profile file.

We need first to install gunicorn library then build the Profile contains the application file and programing language. Gunicorn is used to maintain the web application multiple instance executions, distributing incoming requests across those instances, and communicate with the web server.

App is the flask application file name. in our case is app.py, could be different name.

web: gunicorn app:app

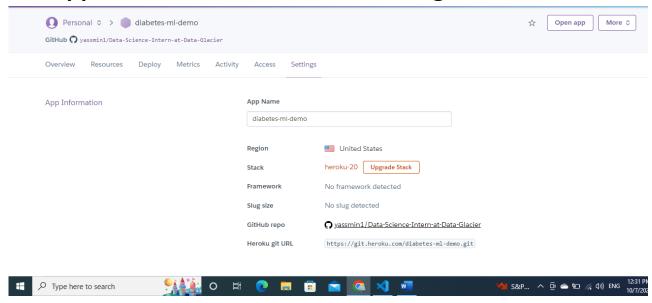
runtime.txt python -3.9.13

# Steps to deploy MI model in Heroku:

**First step**: We must create an account in Heroku and create an app as follow with stack 20 **Second step**: Connecting the Heroku app to Github repository containing the application files.

**Third step**: Choosing the language and then deploy the application. The output of this step as follows.

# The application information and setting:



# The application can be found at

https://diabetes-ml-demo.herokuapp.com/

