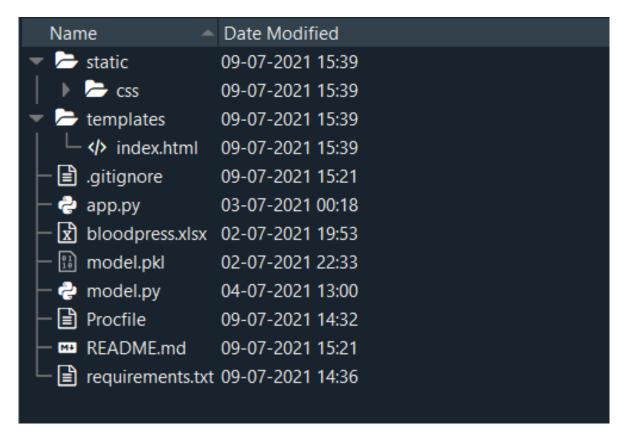
Week 5: Cloud and API Deployment

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Submission Date	09 July 2021
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Directory Structure



Steps Involved in the Deployment

Step 1:

- 1. Train the Blood Pressure Predictor model
- 2. Use the pickle operation to serialize the machine learning algorithm (Linear Regression) and dumping the serialized format to the file **model.pkl** in a binary write format.
- 3. Save the file in **model.py**

```
app.py × model.py × index.html ×
         # -*- coding: utf-8 -*-
         Created on Thu Jul 1 19:15:34 2021
         @author: RUNA
         import pandas as pd
         import pickle
         from sklearn.linear_model import LinearRegression
         df = pd.read_excel('bloodpress.xlsx')
         #print(df.isnull().values.sum())
         X = df.iloc[:,1:]
         y = df.iloc[:,0:1]
         regressor = LinearRegression()
         regressor.fit(X,y)
         # Make predictions using the testing set
         bp_predict = regressor.predict(X)
         # Saving model to disk
         pickle.dump(regressor,open('model.pkl','wb'))
```

Step 2: Creating the HTML file and saving in index.html and CSS (for styling) file.

```
app.py × model.py × index.html ×
            <!DOCTYPE html>
            <html >
<!--From https://codepen.io/frytyler/pen/EGdtg-->
               <meta charset="UTF-8">
                <title>Blood Pressure Prediction ML API</title>
                   <link rel="stylesheet" href="https://www.w3schools.com/w3css/4/w3.css">
<link rel="stylesheet" href="https://fonts.googleapis.com/css?family=Raleway">
<link rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-awesome/4.7.0/css/font-awesome.min.css">
                   <link rel="stylesheet" href="{{ url_for('static', filename = 'css/style9.css') }}">
           <div class="w3-container w3-light-grey">
                 <h1 class="w3-center", style="padding:128px 16px">PREDICT YOUR BLOOD PRESSURE</h1>
                 <!-- Main Input For Receiving Query to our ML -->
<form action="{{ url_for('predict')}}"method="post">
                       rm action="{{ url_for("predict")}} "method="post">
<input class="w3-input w3-border" type="text" name="age" placeholder="Age(years)" required="required" />
<input class="w3-input w3-border" type="text" name="weight" placeholder="Weight(kg)" required="required" />
<input class="w3-input w3-border" type="text" name="bsa" placeholder="Body Surface Area(sq m)" required="required" />
<input class="w3-input w3-border" type="text" name="hypertension" placeholder="Duration of Hypertension(years)" required="p><input class="w3-input w3-border" type="text" name="pulse" placeholder="Basal Pulse(beats per minute)" required="required" />
<input class="w3-input w3-border" type="text" name="stress" placeholder="Stress Index" required="required" />
                        <div class = "center">
                               cbutton type="submit" class="w3-button w3-black" type = "submit">
<i class="fa fa-heart"></i>
                               Predict Blood Pressure
                                </button>
               {{ prediction_text }}
```

```
app.py × model.py × index.html × style9.css ×
        body, html {
          height: 100%;
          line-height: 1.8;
       }
        /* Full height image header */
        .bgimg-1 {
          background-position: center;
          background-size: cover;
background-image: url("bp.jpg");
       min-height: 100%;
   11
        .w3-bar .w3-button {
         padding: 16px;
       }
        .center {
          align-items: center;
          height: 200px;
```

Step 3: Creating the web application file using **Flask**.

```
lacktriangledown app.py 	imes model.py 	imes index.html 	imes style9.css 	imes
          Created on Fri Jul 2 10:49:56 2021
           @author: RUNA
          import numpy as np
from flask import Flask, request, render_template
          app = Flask(__name__)
model = pickle.load(open('model.pkl', 'rb'))
          @app.route('/')
   18
            return render_template('index.html')
          @app.route('/predict',methods=['POST'])
def predict():
                For rendering results on HTML GUI
               int_features = [int(float(x)) for x in request.form.values()]
              final_features = [np.array(int_features)]
prediction = model.predict(final_features)
               output = np.round(prediction[0], 2)
               return render_template('index.html', prediction_text='The predicted Blood Pressure is {} mm Hg'.format(output))
          if <u>name</u>
                __name__ == "__main__
app.run(debug=True)
```

- 1. Creating the object of class Flask.
- 2. Whenever user visits that URL '/', index() method would be called automatically, and the index() method returns our main HTML page called index.html.
 The flask.render_template() looks for the this index.html file in the templates folder and dynamically generates/renders a HTML page for the end user
- 3. @app.route ('/predict') maps the predict() method with the /predict URL, this takes the input given by the user, does all the pre-processing and generates the final feature vector, runs the model on it and gets the final prediction.

Step 4: Create a requirement.txt and Procfile (without any extensions)

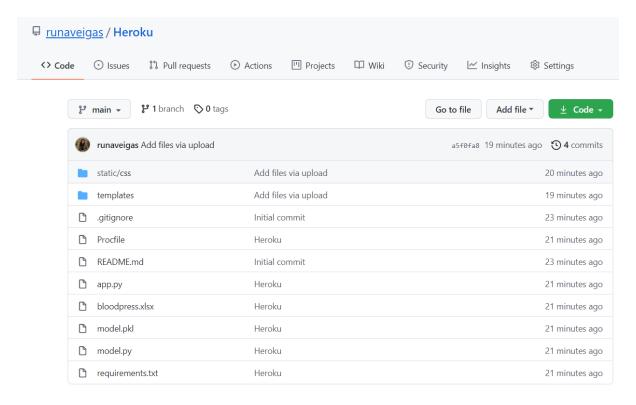
Procfile tells **heroku** about a set of processes/commands that needs to be run beforehand.



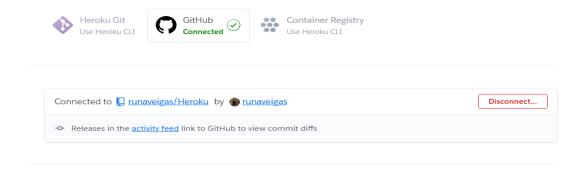
requirements.txt tells **Heroku** the set of applications or libraries needed for the deployment of the application

```
| Tequirements.txt | Flask==1.1.1 | 2 | gunicorn==19.9.0 | 3 | itsdangerous==1.1.0 | 4 | Jinja2==2.10.1 | 5 | MarkupSafe==1.1.1 | 6 | Werkzeug==0.15.5 | 7 | numpy>=1.9.2 | 8 | scipy>=0.15.1 | 9 | scikit-learn>=0.18 | 10 | matplotlib>=1.4.3 | 11 | pandas>=0.19
```

Step 5: Commit the changes to git-hub



Step 6: Link and connect Github to Heroku.



Step 7: Deploy the application.

Deploy a GitHub branch

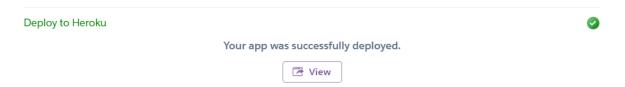
This will deploy the current state of the branch you specify below. Learn more.

Choose a branch to deploy

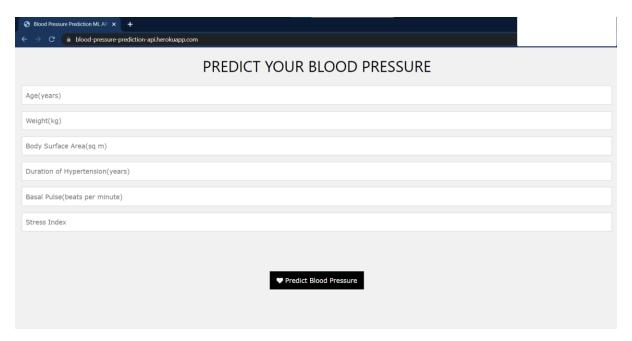
property main

Deploy Branch

Step 8: After successful deployment we can view the application by clicking on **view**.



Step 9: Final Result.



***********Thank You*********