Week 4

Flask report and file details

4-8-2023 Virtual Internship

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Step 1 - CSV Data extraction

AirQuality.csv dataset used for this project was provided by FEDESORIANO (Saverio De Vito (saverio.devito '@' enea.it), ENEA - National Agency for New Technologies, Energy and Sustainable Economic Development), details of data collection and meaning of each value are described by FEDESORIANO online at: <u>Kaggle</u>

- 1. Load the dataset from raw storage on Github.
- 2. Drop unused columns (date and hour)
- 3. Replace invalid characters for numeric values
- 4. Convert to float all data
- 5. Drop last columns (Noise created after conversion)
- 6. Replace -200 with nan and drop nan values (See Kaggle description)
- 7. Split the dataset into 70% training, 30% validation

```
# Load dataset
data_frame = pd.read_csv('https://raw.githubusercontent.com/papitaAlgodonCplusplus/LISUM24/main/Week%204/dataset/AirQuality.csv', delimiter=";")
# Shuffle dataset
data_frame = data_frame.sample(frac = 1).reset_index(drop=True)

data_frame = data_frame.drop(columns=data_frame.columns[0])
data_frame = data_frame.drop(columns=data_frame.columns[0])

data_frame = data_frame.applymap(lambda x: x.replace(',', '.') if isinstance(x, str) else x)

data_frame = data_frame.astype(float)
data_frame = data_frame.drop(columns=data_frame.columns[-1])
data_frame = data_frame.drop(columns=data_frame.columns[-1])

data_frame = data_frame.replace(-200, float('nan'))
data_frame = data_frame.dropna()

X_train, X_test, y_train, y_test = train_test_split(data_frame.iloc[:, 1:], data_frame.iloc[:, 0], test_size=0.3, random_state=42)
```

Step 2 – Model Training and Plotting

By using scikit-learn library, I could easily implement this model so it allows plk saving (for flask deployment)

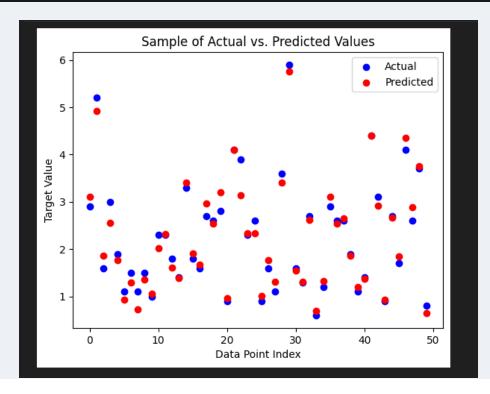
```
from sklearn.metrics import mean_squared_error

y_pred = linear_est.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)

... Mean Squared Error: 0.05878599159200155

> sample_size = 50

plt.scatter(range(sample_size), y_test[:sample_size], color='blue', label='Actual')
plt.scatter(range(sample_size), y_pred[:sample_size], color='red', label='Predicted')
plt.xlabel('Data Point Index')
plt.ylabel('Target Value')
plt.title('Sample of Actual vs. Predicted Values')
plt.legend()
plt.show()
```



Step 3 - Flask

Once the model is ready, flask deployment requires:

- 1. Saving the model as plk
- 2. From web_app.py, load the plk model
- 3. Create home and results html files with css style (optional)
 - 4. Declare and redirect user to each site given '/x' used
 - 5. Declare functions and variables for each site as needed
 - 6. Run the web_app.py from flask virtual enviorment

```
import joblib
  joblib.dump(linear_est, '../model/linear_regressor_model.pkl')
['../model/linear_regressor_model.pkl']
```

```
> from flask import Flask, request, jsonify, render_template, redirect, url_for...
 app = Flask("CO Prediction Web App")
 model = joblib.load('../model/linear_regressor_model.pkl')
 @app.route('/')
 def home():
     return render_template('index.html')
 @app.route('/predict', methods=['GET', 'POST'])
 def predict():
     if request.method == 'POST':
         float_features = [float(x) for x in request.form.values()]
         final features = [np.array(float features)]
         predicted_result = model.predict(final_features)
         # Redirect to the results page with the calculated prediction
         return redirect(url for('results', prediction=predicted result[0]))
     else:
         # If the method is GET, render the input form template
         return render_template('index.html')
 @app.route('/results', methods=['GET'])
 def results():
     # Get the prediction value from the URL parameter
     prediction = request.args.get('prediction')
     return render_template('results.html', prediction=prediction)
 if <u>name</u> == '_main_':
     app.run(host='0.0.0.0', port=5000)
```

Step 4 - HTML and CSS files

HTML indexing requires 2 files: Homepage and Results

HTML Homepage:

HTML Results:

```
<!DOCTYPE html>

<html>

link rel="stylesheet" type="text/css" href="{{ url_for('static', filename='style.css') }}">

<head>

<title>Result Page</title>

</head>

<body>

<h1>Result:</h1>

<h3>The calculated result is: {{ prediction }}
</hd>

</body>

</html>
```

CSS Style:

```
body {
font-family: Arial, sans-serif;
background-image: url("/static/background.jpg");
background-repeat: no-repeat;
background-size: cover;
} ....
```

Thanks