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BATCH CODE: LISUM02

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## DEPLOYMENT OF ML MODEL ON FLASK

1. Select the dataset

I have selected a small dataset for housing price prediction from Kaggle

2. Open IDE Spyder and work on a file named model.py, cleaning the data, conducting EDA on the dataset and finally model building from the dataset

```
C:\Users\Sayoj\.spyder-py3\Deployment-flask-new\model.py
temp.py × model.py* × app.py ×
       # Importing the libraries
       #import numpy as np
       #import matplotlib.pyplot as plt
       import pandas as pd
       import pickle
       #Reading the dataset
       dataset = pd.read_csv('housing price.csv')
       #The required columns are considered important for prediction (after EDA)
feature_names = ['LotArea', 'YearBuilt', 'TotRmsAbvGrd']
       #X is assigned dataset with independent variables
       X = dataset[feature_names]
       #Y is assigned for the dependent variable
       y = dataset['SalePrice']
        #Splitting Training and Test Set
        from sklearn.ensemble import RandomForestRegressor
        regressor = RandomForestRegressor(random_state=1)
        #Fitting model with trainig data
        regressor.fit(X, y)
```

3. Load the model to the disk using pickle (model.pkl) and compare the results

```
# Saving model to disk
pickle.dump(regressor, open('model.pkl','wb'))

# Loading model to compare the results
model = pickle.load(open('model.pkl','rb'))

print(model.predict(X))
```

4. The app.py file is created, app.py file contains the python code that can redirect to execute the model from model.py

```
lacksquare temp.py 	imes model.py 	imes app.py 	imes
       import numpy as np
from flask import Flask, request, jsonify, render_template
        import pickle
        app = Flask(__name__)
       model = pickle.load(open('model.pkl', 'rb'))
        @app.route('/')
       def home():
            return render_template('index.html')
  12
        @app.route('/predict',methods=['POST'])
       def predict():
            For rendering results on HTML GUI
            int_features = [int(x) for x in request.form.values()]
            final_features = [np.array(int_features)]
prediction = model.predict(final_features)
  18
19
            output = round(prediction[0], 2)
            return render_template('index.html', prediction_text='House pricing sho
           p.route('/predict_api',methods=['POST'])
     ▼ def predict_api():
            For direct API calls trought request
            data = request.get_json(force=True)
            prediction = model.predict([np.array(list(data.values()))])
            output = prediction[0]
```

return jsonify(output)

 if \_\_name\_\_ == "\_\_main\_ app.run(debug=True)

- 5. The static folder contains the css file (style file), these are used to control the appearance of the UI
- 6. The templates folder contains the html pages which will be used to create the UI.
- 7. On Anaconda Prompt, ensure that you are in the project home directory. Create the machine learning model by running below command python model.pyThis would create a serialized version of our model into a file model.pkl
- 8. Run app.py using python app.py command to start Flask API

```
base) C:\Users\Sayoj\.spyder-py3\Deployment-flask-new>python app.py
* 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 windowsapi reloader
* Debugger is active!
* Debugger PIN: 165-361-565
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

9. Use the url received in the browser and the result page as below:

