Data Intake Report Cloud and API deployment

Project Name: Cloud and API deployment

Report Date: 06/04/2024

Internship Batch: LISUM33

Version: 1.0

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Data Intake Reviewer: Data Glacier

Data Storage Location:

https://github.com/Marinatsv07/Data_Glacier_Internship/tree/main/Week_5

Model Setup

df = pd.read parquet('https://d37ci6vzurychx.cloudfront.net/trip-data/yellow tripdata 2024-01.parquet')

```
In [6]:
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# Calculate 'duration' as the difference between drop-off and pick-up times

df['duration'] = (df['tpep_dropoff_datetime'] - df['tpep_pickup_datetime'])

# Convert duration from timedelta to minutes

df['duration_min'] = df['duration'].dt.total_seconds() / 60

# Calculate the standard deviation of trip durations in January (in minutes)

std_duration_january = df['duration_min'].std()

print("Standard deviation of trips duration in January (in minutes):", std_duration_january)

# Filtering df to remove outliers, keeping only durations between 1 and 60 minutes

original_count = len(df)

df_filtered = df[(df['duration_min'] >= 1) & (df['duration_min'] <= 60)].copy()

filtered_count = len(df_filtered)

# Determine what fraction of records remains after dropping outliers

fraction_left = filtered_count / original_count
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print("Fraction of records left after dropping outliers:", fraction left)

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# Convert location IDs to strings
df_filtered['PULocationID'] = df_filtered['PULocationID'].astype(str)
df filtered['DOLocationID'] = df filtered['DOLocationID'].astype(str)
# Use only the filtered duration in minutes for further processing
df filtered['duration'] = df filtered['duration min']
# Convert DataFrame to a list of dictionaries for feature encoding
data dicts = df filtered[['PULocationID', 'DOLocationID']].to dict(orient='records')
# Create and apply DictVectorizer
vectorizer = DictVectorizer(sparse=True)
feature matrix = vectorizer.fit transform(data dicts)
# Number of columns in the feature matrix
print("Dimensionality of the feature matrix (number of columns):", feature matrix.shape[1])
# Target variable
y = df filtered['duration']
# Split data into training and test sets
X_train, X_test, y_train, y_test = train_test_split(feature_matrix, y, test_size=0.2, random_state=42)
# Create and train a linear regression model
model = LinearRegression()
model.fit(X_train, y_train)
# Make predictions on the training data
y_train_pred = model.predict(X_train)
# Calculate RMSE on the training data
rmse_train = np.sqrt(mean_squared_error(y_train, y_train_pred))
print("RMSE on the train data:", rmse_train)
Standard deviation of trips duration in January (in minutes): 34.851053592192876
Fraction of records left after dropping outliers: 0.9778326020432945
Dimensionality of the feature matrix (number of columns): 518
RMSE on the train data: 7.948999308349224
                                                                                                     In [7]:
# Save the trained model using pickle
model filename = 'linear regression model.pkl'
with open(model filename, 'wb') as file:
  pickle.dump(model, file)
print(f"Model saved as {model filename}")
Model saved as linear_regression_model.pkl
```

```
In [8]:
#save the vectorizer too
vectorizer filename = 'vectorizer.pkl'
with open(vectorizer_filename, 'wb') as file:
  pickle.dump(vectorizer, file)
print(f"Vectorizer saved as {vectorizer_filename}")
Vectorizer saved as vectorizer.pkl
                                                                                                       In [9]:
# Load the saved model
with open(model filename, 'rb') as file:
  loaded model = pickle.load(file)
# Load the saved vectorizer
with open(vectorizer filename, 'rb') as file:
  loaded_vectorizer = pickle.load(file)
                                                                                                       In [6]:
def preprocess_data(data_path: str) -> pd.DataFrame:
  Preprocess the data from the given Parquet file.
  Parameters:
  data_path (str): Path to the input data file (Parquet format).
  Returns:
  pd.DataFrame: Preprocessed DataFrame.
  print("Loading data...")
  df = pd.read_parquet(data_path)
  print("Data loaded. Calculating duration...")
  df['duration'] = (df['tpep_dropoff_datetime'] - df['tpep_pickup_datetime'])
  df['duration min'] = df['duration'].dt.total seconds() / 60
  print("Filtering outliers...")
  df filtered = df[(df['duration min'] >= 1) & (df['duration min'] <= 60)].copy()
  print("Converting location IDs to strings...")
  df filtered['PULocationID'] = df filtered['PULocationID'].astype(str)
  df filtered['DOLocationID'] = df filtered['DOLocationID'].astype(str)
  print("Data preprocessing complete.")
  return of filtered
```

```
data_path = 'https://d37ci6vzurychx.cloudfront.net/trip-data/yellow_tripdata_2024-02.parquet'
df_filtered = preprocess_data(data_path)
print(df_filtered.head())
```

Loading data...

Data loaded. Calculating duration...

Filtering outliers...

Converting location IDs to strings...

Data preprocessing complete.

trip_distance RatecodeID store_and_fwd_flag PULocationID DOLocationID \

U	4.39	1.0	IN	00	230
1	7.71	1.0	N	48	243
2	28.69	2.0	N	132	261
3	1.10	1.0	N	161	163
4	2.60	1.0	N	246	79

payment type ... extra mta tax tip amount tolls amount \

0	1	1.0	0.5	1.28	0.00
1	1	1.0	0.5	9.00	0.00
2	2	0.0	0.5	0.00	6.94
3	1	3.5	0.5	2.85	0.00
4	2	3.5	0.5	0.00	0.00

improvement surcharge total amount congestion surcharge Airport fee \

0	1.0	26.78	2.5	0.00
1	1.0	45.00	2.5	0.00
2	1.0	82.69	2.5	1.75
3	1.0	17.15	2.5	0.00
4	1.0	20.60	2.5	0.00

duration duration_min

[5 rows x 21 columns]

```
Load the DictVectorizer from the given file.
  Parameters:
  vectorizer path (str): Path to the saved DictVectorizer file (Pickle format).
  Returns:
  DictVectorizer: Loaded DictVectorizer.
  print("Loading vectorizer...")
  with open(vectorizer_path, 'rb') as file:
     vectorizer = pickle.load(file)
  print("Vectorizer loaded.")
  return vectorizer
vectorizer_path = 'vectorizer.pkl'
vectorizer = load vectorizer(vectorizer path)
data_dicts = df_filtered[['PULocationID', 'DOLocationID']].to_dict(orient='records')
print("Transforming data...")
X = vectorizer.transform(data dicts)
print("Data transformed. Shape:", X.shape)
Loading vectorizer...
Vectorizer loaded.
Transforming data...
Data transformed. Shape: (2938060, 518)
                                                                                                      In [8]:
def load_model(model_path: str) -> LinearRegression:
  Load the model from the given file.
  Parameters:
  model path (str): Path to the saved model file (Pickle format).
  Returns:
  LinearRegression: Loaded model.
  print("Loading model...")
  with open(model path, 'rb') as file:
     model = pickle.load(file)
  print("Model loaded.")
  return model
# Example usage
model path = 'linear regression model.pkl'
model = load model(model path)
```

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y = df filtered['duration min']
print("Making predictions...")
y_pred = model.predict(X)
print("Predictions made. First 10 predictions:", y pred[:10])
Loading model...
Model loaded.
Making predictions...
Predictions made. First 10 predictions: [13.24544287 20.48169216 37.42164098 13.46284077
13.12928572 12.15326858
12.69852248 13.03233581 13.81530867 10.55141448]
                                                                                                 In [9]:
def calculate_rmse(y_true: pd.Series, y_pred: np.ndarray) -> float:
  Calculate the Root Mean Squared Error (RMSE).
  Parameters:
  y true (pd.Series): True values.
  y_pred (np.ndarray): Predicted values.
  Returns:
  float: RMSE value.
  print("Calculating RMSE...")
  rmse = np.sqrt(mean_squared_error(y_true, y_pred))
  print("RMSE calculated:", rmse)
  return rmse
# Example usage
rmse = calculate rmse(y, y pred)
Calculating RMSE...
RMSE calculated: 8.124017135620836
```

Heroku Deployment





