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COURSE NAME: Data Science with Python Career Program (ChatGPT

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ASSIGNMENT: Machine Learning (MAJOR)

Q1. Download the Oil Spill Dataset and perform Data cleaning and Data Pre-Processing if Necessary.

```
import pandas as pd
df = pd.read_csv('oil_spill.csv')

# Display the first few rows of the dataset
print("Initial dataset:")
print(df.head())
```

```
import pandas as pd
df = pd.read_csv('oil_spill.csv')

# Data Cleaning

# Check for missing values
print("\nMissing values:")
print(df.isnull().sum())

# Fill missing numerical values with mean
df.fillna(df.mean(), inplace=True)
```

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```
PS C:\Users\vaibh\Downloads> & C:\Users\vaibh/AppData/Local/Microsoft/WindowsApps/python3.11.exe "c:\Users\vaibh/Downloads/0i1 Spill"
 Missing values:
f_1 0
f_2 0
f_3 0
f_4 6
f_5 0
f_6 0
f_7 0
f_6 0
f_7 0
f_10 0
f_11 0
f_12 0
f_13 0
f_14 0
f_15 0
f_16 0
f_17 0
f_18 0
f_18 0
f_19 0
f_19 0
f_21 0
f_22 0
f_23 0
f_24 0
f_25 0
f_25 0
f_26 0
f_27 0
f_28 0
f_30 0
f_31 0
f_32 0
f_33 0
f_33 0
f_33 0
f_33 0
f_34 0
f_35 0
f_36 0
f_37 0
f_38 0
f_37 0
f_38 0
f_39 0
f_41 0
f_42 0
f_44 0
f_44 0
f_45 0
f_47 0
f_48 0
f_49 0
target int64
target int64
 PS C:\Users\vaibh\Downloads>
```

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Q2. Use various methods such as Handling null values, One-Hot Encoding, Imputation, and Scaling of Data Pre-Processing where necessary.

Answer:

a. Handling null values:

```
import pandas as pd

# Load the dataset
df = pd.read_csv('oil_spill.csv')

# Drop rows with null values
df.dropna(inplace=True)

# Display the dataset after handling null values
print(df.head())
```

```
2 3 115 1449.85 608.43 88 287500 40.42 7.34 ... 150.00 45.13 9.33 1 31692.84 65.81 7.84 1 3 4 1201 1562.53 295.65 66 3002500 42.40 7.97 ... 453.21 144.97 13.33 1 37696.21 65.67 8.07 1 4 5 312 950.27 440.86 37 780000 41.43 7.03 ... 512.54 109.16 2.58 0 29038.17 65.66 7.35 0 [5 rows x 50 columns]

PS C:\Users\vaibh\Downloads>
```

b. One-hot Encoding

```
import pandas as pd
from sklearn.preprocessing import OneHotEncoder

# Load the dataset
df = pd.read_csv('oil_spill.csv')

# Extract categorical columns
categorical_columns = df.select_dtypes(include=['object', 'category']).columns.tolist()

# Initialize OneHotEncoder
encoder = OneHotEncoder()

# Fit and transform the categorical columns
encoded_features = encoder.fit_transform(df[categorical_columns])

# Create a DataFrame for the encoded features
df_encoded = pd.DataFrame(encoded_features.toarray(), columns=encoder.get_feature_names_out(categorical_columns))

# Concatenate the encoded features with the original DataFrame
df = pd.concat([df.drop(columns=categorical_columns), df_encoded], axis=1)

print(df.head())
```

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```
PS C:\Users\vaibh\Downloads> & C:\Users\vaibh\AppData\Local\Microsoft\WindowsApps\python3.11.exe "c:\Users\vaibh\Downloads\Oil Spill" f_1 f_2 f_3 f_4 f_5 f_6 f_7 f_8 ... f_43 f_44 f_45 f_46 f_47 f_48 f_49 target 0 1 2558 1506.09 456.63 90 6395000 40.88 7.89 ... 763.16 135.46 3.73 0 33243.19 65.74 7.95 1 1 2 22325 79.11 841.03 180 55812500 51.11 1.21 ... 9593.48 1648.80 0.60 0 51572.04 65.73 6.26 0 2 3 115 1449.85 608.43 88 287500 40.42 7.34 ... 150.00 45.13 9.33 1 31692.84 65.81 7.84 1 3 4 1201 1562.53 295.65 66 3002500 42.40 7.97 ... 453.21 144.97 13.33 1 37696.21 65.67 8.07 1 4 5 312 950.27 440.86 37 780000 41.43 7.03 ... 512.54 109.16 2.58 0 29038.17 65.66 7.35 0 [5 rows x 50 columns]
PS C:\Users\vaibh\Downloads>
```

c. Imputation

```
import pandas as pd
from sklearn.impute import SimpleImputer

# Load the dataset
df = pd.read_csv('oil_spill.csv')

# Display the first few rows of the dataset
print("Initial dataset:")
print(df.head())

# Imputation

# Extract numerical columns
numerical_columns = df.select_dtypes(include=['int64', 'float64']).columns.tolist())

# Initialize SimpleImputer
imputer = SimpleImputer(strategy='mean')

# Fit and transform the numerical columns
df[numerical_columns] = imputer.fit_transform(df[numerical_columns])
print("\nDataset after imputation:")
print(df.head())
```

```
PS C:\Users\vaibh\Downloads> & C:\Users\vaibh/AppData/Local/Microsoft/WindowsApps/python3.11.exe "c:\Users\vaibh/Downloads/Oil Spill"
Initial dataset:
   f_1 f_2 f_3 f_4 f_5 f_6 f_7 f_8 ...
1 2558 1506.09 456.63 90 6395000 40.88 7.89 ...
                                                                                                   f_44 f_45 f_46
35.46 3.73 0
                                                                                       f_43
                                                                                                                                0 33243.19 65.74 7.95
                                                                                     763.16 135.46
          22325 79.11 841.03 180 55812590 51.11 1.21 ... 9593.48 1648.80 0.60 115 1449.85 608.43 88 287590 40.42 7.34 ... 150.00 45.13 9.33 1201 1562.53 295.65 66 3002500 42.40 7.97 ... 453.21 144.97 13.33 312 950.27 440.86 37 780000 41.43 7.03 ... 512.54 109.16 2.58
     2 22325
                                                                                                                       0 51572.04 65.73 6.26
                                                                                                                       1 31692.84 65.81 7.84
                                                                                                                       1 37696.21 65.67 8.07
[5 rows x 50 columns]
Dataset after imputation:
  f 1
            f 2
                       f_3
                                             f_5
                                                            f_6
                                                                                              f_43
                                                                                                          f_44
                                                                                                                   f_45 f_46
                                                                                                                                        90.0 6395000.0 40.88 7.89 ... 763.16
          2558.0 1506.09 456.63
                                                                                                       135.46 3.73 0.0 33243.19 65.74 7.95
  1.0
                                                                                                                                                                     1.0
                      79.11 841.03 180.0 55812500.0 51.11 1.21 ... 9593.48 1648.80
1 2.0 22325.0
                                                                                                                             0.0 51572.04 65.73 6.26
                                                                                                                   0.60
                                                                                                                                                                     0.0

    287500.0
    40.42
    7.34
    ...
    150.00
    45.13
    9.33
    1.0
    31692.84
    65.81
    7.84

    3002500.0
    42.40
    7.97
    ...
    453.21
    144.97
    13.33
    1.0
    37696.21
    65.67
    8.07

    780000.0
    41.43
    7.03
    ...
    512.54
    109.16
    2.58
    0.0
    29038.17
    65.66
    7.35

           115.0 1449.85 608.43
                                            88.0
2 3.0
                                                                                                                                                                     1.0
  4.0 1201.0 1562.53 295.65
                                            66.0
                                                                                                                                                                     1.0
4 5.0 312.0 950.27 440.86
                                            37.0
                                                                                                                                                                     0.0
[5 rows x 50 columns]
PS C:\Users\vaibh\Downloads>
```

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d. Scaling of Data Pre-Processing

```
import pandas as pd
from sklearn.preprocessing import StandardScaler

df = pd.read_csv('oil_spill.csv')

# Extract numerical columns
numerical_columns = df.select_dtypes(include=['int64', 'float64']).columns.tolist()

# Initialize StandardScaler
scaler = StandardScaler()

# Fit and transform the numerical columns
df[numerical_columns] = scaler.fit_transform(df[numerical_columns])

# dataset after scaling
print("\nDataset after scaling:")
print(df.head())
```

Q3. Derive some insights from the dataset.

```
# Summary Insights
print("\nSummary Insights:")
print("- Dataset was loaded, duplicates were removed.")
print("- Missing values in numerical columns were imputed with column means.")
print("- Categorical features were one-hot encoded.")
print("- Relevant features and target columns were selected.")
print("- Numerical features were scaled using StandardScaler.")
```

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```
Summary Insights:
- Dataset was loaded, duplicates were removed.
- Missing values in numerical columns were imputed with column means.
- Categorical features were one-hot encoded.
- Relevant features and target columns were selected.
- Numerical features were scaled using StandardScaler.
PS C:\Users\vaibh\Downloads>
```

Q4. Apply various Machine Learning techniques to predict the output in the target column, make use of Bagging and Ensemble as required, and find the best model by evaluating the model using Model evaluation techniques.

```
import pandas as pd
df = pd.read_csv("oil_spill.csv")
print(df.head())
df = df.drop_duplicates()
df.fillna(df.mean(), inplace=True)
from sklearn.preprocessing import StandardScaler, OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.impute import SimpleImputer
# Define feature columns and target column
features = ['f_1', 'f_2']
target = 'target'
X = df[features]
y = df[target]
numeric_features = ['f_1', 'f_2']
numeric_transformer = Pipeline(steps=[
   ('imputer', SimpleImputer(strategy='mean')),
   ('scaler', StandardScaler())])
categorical_transformer = Pipeline(steps=[
    ('imputer', SimpleImputer(strategy='constant', fill_value='missing')),
   ('onehot', OneHotEncoder(handle_unknown='ignore'))])
preprocessor = ColumnTransformer(
       ('num', numeric_transformer, numeric_features)])
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

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```
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
# Define the models
models = {
    'Logistic Regression': LogisticRegression(),
    'Random Forest': RandomForestClassifier(),
    'Gradient Boosting': GradientBoostingClassifier()
trained_models = {}
for name, model in models.items():
    pipeline = Pipeline(steps=[('preprocessor', preprocessor), ('classifier', model)])
    trained_models[name] = pipeline.fit(X_train, y_train)
from sklearn.metrics import accuracy_score, roc_auc_score
# Evaluate the models
for name, model in trained_models.items():
   y_pred = model.predict(X_test)
   accuracy = accuracy_score(y_test, y_pred)
    auc = roc_auc_score(y_test, model.predict_proba(X_test)[:, 1])
    print(f"{name} - Accuracy: {accuracy:.4f}, AUC: {auc:.4f}"
```

```
# Make predictions with the best model
best_model = trained_models['Random Forest']
y_pred = best_model.predict(X_test)

# Show predictions
print(y_pred)
```

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Q5. Save the best model and load the model.

```
# Save the best model
best_model = trained_models[best_model_name]
joblib_file = "best_model.pkl"
joblib.dump(best_model, joblib_file)
print(f"Best model saved as {joblib_file}")

# Load the best model
loaded_model = joblib.load(joblib_file)

# Use the loaded model to make predictions
y_pred_loaded = loaded_model.predict(X_test)

# Show predictions
print(y_pred_loaded)
```

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Q6. Take the original data set and make another dataset by randomly picking 20 datapoints from the oil spill data set and applying the saved model to the same.

```
# Train Multiple Models and Save the Best Model
models = {
    'Logistic Regression': LogisticRegression(),
    'Random Forest': RandomForestClassifier(),
    'Gradient Boosting': GradientBoostingClassifier()
trained_models = {}
best model name = None
best auc = 0
for name, model in models.items():
    pipeline = Pipeline(steps=[('preprocessor', preprocessor), ('classifier', model)])
    trained_models[name] = pipeline.fit(X_train, y_train)
   y_pred = pipeline.predict(X_test)
    accuracy = accuracy_score(y_test, y_pred)
    auc = roc_auc_score(y_test, pipeline.predict_proba(X_test)[:, 1])
    print(f"{name} - Accuracy: {accuracy:.4f}, AUC: {auc:.4f}")
    if auc > best_auc:
       best_auc = auc
        best_model_name = name
```

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```
print(f"Best model: {best model name} with AUC: {best auc:.4f}")
# Save the best model
best_model = trained_models[best_model_name]
joblib_file = "best_model.pkl"
joblib.dump(best_model, joblib_file)
print(f"Best model saved as {joblib_file}")
# 3. Randomly Select 20 Data Points
df_sample = df.sample(n=20, random_state=42)
# 4. Apply the Saved Model to the Selected Data Points
X_sample = df_sample[features]
# 5. Load the Best Model and Make Predictions
loaded_model = joblib.load(joblib_file)
y_pred_sample = loaded_model.predict(X_sample)
# Display Results
df_sample['predicted_target'] = y_pred_sample
print(df_sample)
```

```
predicted_target
                                                                                       f_47
                                                                                                         target
     29 105
              881.92 1128.79
                               83
                                     262500 38.90
                                                               84.74
                                                                       4.21
                                                                                    3425.75 65.97
     60 111 1153.32 1283.44
                                41
                                     277500
                                            41.25
                                                    5.98
                                                               140.92
                                                                       2.49
                                                                                A
                                                                                    5915.80 66.12
                                                                                                    7.34
                                                                                                                                A
209
     17 867 1059.49
                       581.31
                               46
                                    2167500
                                            31.08
                                                   8.26
                                                               408.01
                                                                       4.93
                                                                                a
                                                                                   5679.31 65.74
                                                                                                    7.42
                               140
                                                   11.28
                                                               87.51
                                                                                   6376.53 65.98
         85
                71.06
                       469.47
                                     688500
                                             70.85
                                                                       3.95
                                                                                0
                                                                                                    6.22
685
     38
         15
               32.47
                       582.13 156
                                     121500
                                                   12.11
                                                                84.66
                                                                       6.47
                                                                                   3285.95 66.11
                                                                                                    5.98
                                            73.27
                                                                                1 15720.91 66.30
                      1761.26
                                     215000
                                                                      14.93
         86
               769.73
                                55
                                             37.55
                                                    6.27
                                                                59.34
                                                                                                    6.71
               904.13
                      2689.99
                                     649687
                                             29.80
                                                    8.99
                                                                0.00
                                                                       0.00
                                                                                   40916.70
                                                                                            36.71
         128 1378.47
                                     320000
                                                                                   9183.53 65.98
                                                                                                   14.82
     38 294
                11.49 1559.36
                                40
                                     413437
                                             38.12
                                                                0.00
                                                                                0 10484.87
                                                                                            36.02
          98 1326.06
                      1109.08
                                     245000
                                            41.31
                                                                33.61
                                                                       3.71
                                                                                   7233.16 66.02
528
     34 151
              465.77
                      1736.15
                                73
                                     212343
                                            28.96
                                                    8.14
                                                                0.00
                                                                       0.00
                                                                                   8415.67 36.35
                                                                                                   14.83
247
    138 144 1341.72
                        78.22 110
                                                               60.43 11.49
                                     360000
                                            31.12
                                                    6.88
                                                                                   6824.45 65.55
                                                                                                    7.90
250
    156 260
             1080.89
                       833.29
                               111
                                     650000
                                             30.52
                                                    7.95
                                                               161.45
                                                                      5.93
                                                                                   4667.21 65.86
                                                                                                    7.36
485
                                     118125
     53
               575.19
                      1558.81
                                             30.94
                                                                0.00
                                                                                0 10674.79
                                                                                            36.41
                                                                                                   14.92
                               153
                                                    8.89
                                                                       0.00
467
               619.18 1622.32
                                     104062
                                             26.45
                                                    5.92
                                                                0.00
                                                                       0.00
                                                                                   11277.47
                                                                                            36.44
                                                                                                   14.90
                30.80
                        348.90
                                      81000
                                                                                   11172.62 65.80
483
               743.88 1250.60
                                      84375
                                             33.03
                                                   11.87
                                                                                   9370.56
                               127
                                                               109.90
                                                                       2.95
                                                                                            36.51
                                                                                                   15.08
886 154
          10
               182.50
                       460.00
                                90
                                     81000
                                             57.60
                                                   8.68
                                                                0.00
                                                                       4.00
                                                                                    6004.08 66.01
                                                                                                    6.58
              160.77
                       420.23
                                63
                                     105300
                                            51.15
                                                   10.66
                                                               56.92 20.62
                                                                                    3719.47 65.95
                                                                                                    6.55
244 118 308 1313.18
                       791.35
                                61
                                     770000 29.13
                                                   7.14 ... 181.66
                                                                      4.29
                                                                                0
                                                                                   6636.30 65.87
                                                                                                    7.63
[20 rows x 51 columns]
PS C:\Users\vaibh\Downloads>
```