Task - 1

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In [2]: import pandas as pd
        import numpy as np
        from sklearn.model selection import train test split
        from sklearn.metrics import mean squared error
        from sklearn.ensemble import RandomForestRegressor
        # Load the datasets
        train = pd.read csv('train.csv')
        test = pd.read_csv('test.csv')
        # Convert date column to datetime format
        train['date'] = pd.to datetime(train['date'])
        test['date'] = pd.to_datetime(test['date'])
        # Extract date features
        train['year'] = train['date'].dt.year
        train['month'] = train['date'].dt.month
        train['day'] = train['date'].dt.day
        train['day of week'] = train['date'].dt.dayofweek
        test['year'] = test['date'].dt.year
        test['month'] = test['date'].dt.month
        test['day'] = test['date'].dt.day
        test['day of week'] = test['date'].dt.dayofweek
        # Create lag features
        for lag in range(1, 8):
            train[f'lag {lag}'] = train.groupby('Item Id')['units'].shift(lag)
        # Drop rows with NaN values created by lag features before applying rolli
        train.dropna(subset=[f'lag {lag}' for lag in range(1, 8)], inplace=True)
        # Create rolling window features
        train['rolling_mean_7'] = train.groupby('Item_Id')['units'].transform(lam
        train['rolling_std_7'] = train.groupby('Item Id')['units'].transform(lamb
        # Drop rows with NaN values created by rolling window features
        train.dropna(subset=['rolling_mean_7', 'rolling_std_7'], inplace=True)
        # Handle any remaining NaN values by filling them with the mean of their
        train.fillna(train.mean(), inplace=True)
        # Check for infinite or excessively large values and replace them
        train.replace([np.inf, -np.inf], np.nan, inplace=True)
        train.fillna(train.mean(), inplace=True)
        # Define the feature columns and target column
        feature cols = ['ad spend', 'unit price', 'year', 'month', 'day', 'day of
                       [f'lag_{lag}' for lag in range(1, 8)] + \
                        ['rolling mean 7', 'rolling std 7']
        target_col = 'units'
        # Split the data into training and validation sets
        X train, X val, y train, y val = train test split(train[feature cols], tr
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# Initialize and train the model
model = RandomForestRegressor(n estimators=100, random state=42)
model.fit(X_train, y_train)
# Make predictions on the validation set
val_preds = model.predict(X_val)
# Evaluate the model
mse = mean_squared_error(y_val, val_preds)
print(f'Validation MSE: {mse}')
# Prepare the test set features
# Since 'units' column is not available in the test set, we'll use mean v
for lag in range(1, 8):
    test[f'lag_{lag}'] = train.groupby('Item Id')['units'].transform(lamb
# Ensure lag features align correctly in the test set
test.fillna(test.mean(), inplace=True)
# For rolling mean and std, we can't calculate directly on test set without
# We'll use mean values from the training set for these features as well
test['rolling_mean_7'] = train.groupby('Item Id')['units'].transform(lamb
test['rolling_std_7'] = train.groupby('Item Id')['units'].transform(lambd
# Ensure rolling window features align correctly in the test set
test.fillna(test.mean(), inplace=True)
# Check for infinite or excessively large values and replace them
test.replace([np.inf, -np.inf], np.nan, inplace=True)
test.fillna(test.mean(), inplace=True)
# Make predictions on the test set
test preds = model.predict(test[feature cols])
# Create a submission DataFrame
submission = test[['date', 'Item Id']].copy()
submission['predicted_units'] = test_preds
# Save the submission file
submission to csv('C:/Users/saicharan/OneDrive/Desktop/Nap Queens/Task1/s
print("Submission file created.")
```

Validation MSE: 1186.77385064152 Submission file created.

Task-2

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In [3]: # Convert date column to datetime format
    train['date'] = pd.to_datetime(train['date'])
    test['date'] = pd.to_datetime(test['date'])

# Extract date features
    train['year'] = train['date'].dt.year
    train['month'] = train['date'].dt.month
    train['day'] = train['date'].dt.day
    train['day_of_week'] = train['date'].dt.dayofweek

test['year'] = test['date'].dt.year
```

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test['month'] = test['date'].dt.month
test['day'] = test['date'].dt.day
test['day_of_week'] = test['date'].dt.dayofweek
# Create lag features
for lag in range(1, 8):
    train[f'lag {lag}'] = train.groupby('Item Id')['units'].shift(lag)
# Drop rows with NaN values created by lag features before applying rolli
train.dropna(subset=[f'lag_{lag}' for lag in range(1, 8)], inplace=True)
# Create rolling window features
train['rolling_mean_7'] = train.groupby('Item Id')['units'].transform(lam
train['rolling_std_7'] = train.groupby('Item Id')['units'].transform(lamb
# Drop rows with NaN values created by rolling window features
train.dropna(subset=['rolling_mean_7', 'rolling_std_7'], inplace=True)
# Handle any remaining NaN values by filling them with the mean of their
train.fillna(train.mean(), inplace=True)
# Check for infinite or excessively large values and replace them
train.replace([np.inf, -np.inf], np.nan, inplace=True)
train.fillna(train.mean(), inplace=True)
# Define the feature columns and target column, excluding 'ad spend'
feature_cols = ['unit_price', 'year', 'month', 'day', 'day_of_week'] + \
               [f'lag {lag}' for lag in range(1, 8)] + \
               ['rolling_mean_7', 'rolling_std_7']
target_col = 'units'
# Split the data into training and validation sets
X_train, X_val, y_train, y_val = train_test_split(train[feature_cols], tr
# Initialize and train the model
model = RandomForestRegressor(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
# Make predictions on the validation set
val_preds = model.predict(X_val)
# Evaluate the model
mse = mean squared error(y val, val preds)
print(f'Validation MSE: {mse}')
# Prepare the test set features
# Since 'units' column is not available in the test set, we'll use mean v
for lag in range(1, 8):
   test[f'lag {lag}'] = train.groupby('Item Id')['units'].transform(lamb
# Ensure lag features align correctly in the test set
test.fillna(test.mean(), inplace=True)
# For rolling mean and std, we can't calculate directly on test set without
# We'll use mean values from the training set for these features as well
test['rolling_mean_7'] = train.groupby('Item Id')['units'].transform(lamb
test['rolling std 7'] = train.groupby('Item Id')['units'].transform(lambd
# Ensure rolling window features align correctly in the test set
test.fillna(test.mean(), inplace=True)
```

```
# Check for infinite or excessively large values and replace them
test.replace([np.inf, -np.inf], np.nan, inplace=True)
test.fillna(test.mean(), inplace=True)

# Make predictions on the test set
test_preds = model.predict(test[feature_cols])

# Create a submission DataFrame
submission = test[['date', 'Item Id']].copy()
submission['predicted_units'] = test_preds

# Save the submission file
submission.to_csv('C:/Users/saicharan/OneDrive/Desktop/Nap Queens/Task2/s
print("Submission file created.")
```

Validation MSE: 3061.47883598121 Submission file created.