Import Libaries and Load Data

```
In [11]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from statsmodels.tsa.holtwinters import ExponentialSmoothing
         from sklearn.metrics import mean squared error
         from sklearn.preprocessing import MinMaxScaler
         import warnings
         # Suppress the warning messages
         warnings.filterwarnings("ignore")
         # Read the CSV file into a pandas DataFrame
         file path = "london weather[1].csv"
         df = pd.read_csv(file_path)
         # Prepare the data
         df['date'] = pd.to_datetime(df['date'], format='%Y%m%d')
         df.set_index('date', inplace=True)
         mean_temp_data = df['mean_temp'].dropna()
```

Split Data into Train and Test Sets

```
In [12]: # Normalize the data
scaler = MinMaxScaler()
normalized_data = scaler.fit_transform(mean_temp_data.values.reshape(-1, 1))

# Split data into training and testing sets
train_size = int(0.8 * len(normalized_data))
train_data, test_data = normalized_data[:train_size], normalized_data[train_size:]
```

Fit ETS Model

```
In [14]: trend_list = ['add', None]
         seasonal_list = ['add', None]
         # Initialize variables to keep track of the best model
         best mse = np.inf
         best_ets_fit = None
         # Iterate through different trend and seasonal parameters
         for trend in trend list:
             for seasonal in seasonal_list:
                 # Fit ETS model to training data
                  ets_model = ExponentialSmoothing(train_data, trend=trend, seasonal=seasonal
                 ets_fit = ets_model.fit()
                  # Make predictions
                 test_predictions = ets_fit.forecast(len(test_data))
                  # Inverse transform the predictions
                 test_predictions = scaler.inverse_transform(test_predictions.reshape(-1, 1
                  # Calculate Mean Squared Error
                  mse = mean squared error(mean temp data.values[train size:], test prediction
```

```
# Compare MSE and update best model if needed
if mse < best_mse:
    best_mse = mse
    best_ets_fit = ets_fit</pre>
```

```
In [15]: # Make final predictions with the best model
  final_test_predictions = best_ets_fit.forecast(len(test_data))
  final_test_predictions = scaler.inverse_transform(final_test_predictions.reshape(-:
```

Plot Results

```
In [19]: # Plot the results of the best model
    plt.figure(figsize=(12, 6))
    plt.plot(mean_temp_data.index[train_size:], final_test_predictions, label='Test Proplet.plot(mean_temp_data.index, mean_temp_data.values, label='Actual Data', color='oplt.legend()
    plt.xlabel('Date')
    plt.ylabel('Temperature')
    plt.title('Best ETS Time Series Forecasting')
    plt.show()
```

