

```
In [1]: import pandas as pd

# Load the data
file_path = "/Users/macbookpro/Desktop/ECON331 HW1/TÜİK 2013–2023 TR-DEU/Export.xlsx"
sheet_name = "Use this"
trade_data = pd.read_excel(file_path, sheet_name=sheet_name)

# Display the first few rows of the dataframe to understand its structure
print(trade_data.head())
```

	Year	TOTAL USD
0	2013	40430998510
1	2014	39757516843
2	2015	37226299584
3	2016	37812506839
4	2017	38637760895

```
In [2]: # Load Turkey GDP data
turkey_gdp_file_path = "/Users/macbookpro/Desktop/ECON331 HW1/TurkeyGDP.xls"
turkey_gdp_sheet_name = "Turkey GDP"
turkey_gdp_data = pd.read_excel(turkey_gdp_file_path, sheet_name=turkey_gdp_sheet_name)
```

```
In [3]: # Load Germany GDP data
germany_gdp_file_path = "/Users/macbookpro/Desktop/ECON331 HW1/GermanyGDP.xls"
germany_gdp_sheet_name = "Germany GDP"
germany_gdp_data = pd.read_excel(germany_gdp_file_path, sheet_name=germany_gdp_sheet_name)
```

```
In [4]: # Merge trade data with Turkey GDP data
merged_data = pd.merge(trade_data, turkey_gdp_data, on='Year', how='inner')

# Merge merged data with Germany GDP data
merged_data = pd.merge(merged_data, germany_gdp_data, on='Year', how='inner')
```

```
In [5]: # Rearrange columns
merged_data = merged_data[['Year', 'TOTAL USD', 'Germany GDP', 'Turkey GDP']]
```

```
In [6]: # Display the resulting dataframe
merged_data
```

Out[6]:

	Year	TOTAL USD	Germany GDP	Turkey GDP
0	2013	40430998510	3.733805e+12	9.577991e+11
1	2014	39757516843	3.889093e+12	9.389346e+11
2	2015	37226299584	3.357586e+12	8.643138e+11
3	2016	37812506839	3.469853e+12	8.696829e+11
4	2017	38637760895	3.690849e+12	8.589885e+11
5	2018	38888666525	3.974443e+12	7.789722e+11
6	2019	35897642733	3.889178e+12	7.610059e+11
7	2020	37711457138	3.887727e+12	7.203385e+11
8	2021	41037327626	4.278504e+12	8.198653e+11
9	2022	45174857021	4.082469e+12	9.071184e+11
10	2023	49763197518	4.509469e+12	1.112118e+12

```
In [7]: import numpy as np
import statsmodels.api as sm
```

```
In [8]: # Add Distance column with constant value of 2350 kilometers for the years 2013 to 2023
merged_data['Distance'] = 2350
```

```
In [9]: # Initial guess for parameters
initial_guess = [0.5, 1, 1]
```

```
In [10]: # Extract independent variables
X = merged_data[['Germany GDP', 'Turkey GDP', 'Distance']].values
```

```
In [11]: # Extract dependent variable
y = merged_data['TOTAL USD'].values
```

```
In [12]: from scipy.optimize import curve_fit
import numpy as np

# Define the gravity model function
def gravity_model(X, beta, n, constant):
    return constant + (beta * (X[:, 0] * X[:, 1])) / (X[:, 2] ** n)

# Rescale the variables
X_rescaled = X / np.max(X, axis=0)
y_rescaled = y / np.max(y)

# Initial guess for parameters
initial_guess = [0.5, 1, 1]

# Perform nonlinear least squares (NLS) regression
popt, _ = curve_fit(gravity_model, X_rescaled, y_rescaled, initial_guess)

# Extract estimated parameters
beta, n, constant = popt

# Print the estimated parameters
print("Estimated beta:", beta)
print("Estimated n:", n)
print("Estimated constant:", constant)
```

Estimated beta: 0.6076928959513533  
 Estimated n: 1.0  
 Estimated constant: 0.39608622960928297

/opt/anaconda3/lib/python3.9/site-packages/scipy/optimize/minpack.py:833: OptimizeWarning: Covariance of the parameters could not be estimated  
 warnings.warn('Covariance of the parameters could not be estimated',

```
In [13]: from scipy.stats import t
```

```
In [14]: # Calculate standard errors for parameters
beta_std_err = np.sqrt(np.diag(np.linalg.inv(np.dot(X_rescaled.T, X_rescaled))))
t_values = beta / beta_std_err
```

```
In [15]: # Degrees of freedom
df = len(y_rescaled) - len(initial_guess)
```

```
In [16]: # Calculate p-values for variables
p_values = 2 * (1 - t.cdf(np.abs(t_values), df))
```

```
In [17]: # Print p-values for variables
print("P-values for Variables:")
print("Product of GDPs:", p_values[0])
print("Distance:", p_values[1])
```

P-values for Variables:  
 Product of GDPs: 0.8973610826324747  
 Distance: 0.8644004376758858

```
In [18]: # Compute predicted values
y_pred_rescaled = gravity_model(X_rescaled, beta, n, constant)
```

```
In [19]: # Compute R-squared (R^2)
ss_total = np.sum((y_rescaled - np.mean(y_rescaled)) ** 2)
ss_residual = np.sum((y_rescaled - y_pred_rescaled) ** 2)
rsquared = 1 - (ss_residual / ss_total)
print("R-squared (R^2):", rsquared)
```

R-squared (R^2): 0.8829831023533649

```
In [20]: # Compute Mean Squared Error (MSE)
mse = np.mean((y_rescaled - y_pred_rescaled) ** 2)
print("Mean Squared Error (MSE):", mse)
```

Mean Squared Error (MSE): 0.0006897210443292836

```
In [21]: import matplotlib.pyplot as plt

# Set figure size
plt.figure(figsize=(10, 6)) # Adjust width and height as needed

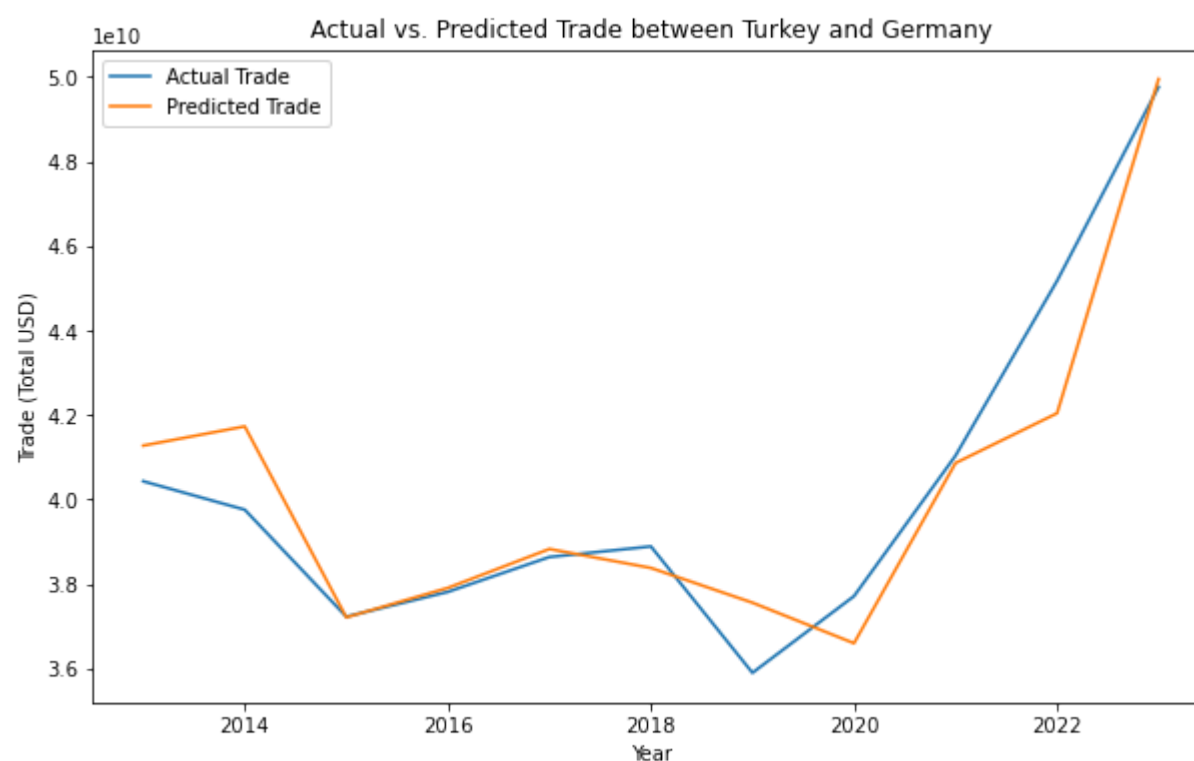
# Denormalize the predicted trade values
y_pred = y_pred_rescaled * np.max(y)

# Plot actual trade values
plt.plot(merged_data['Year'], merged_data['TOTAL USD'], label='Actual Trade')

# Plot predicted trade values
plt.plot(merged_data['Year'], y_pred, label='Predicted Trade')

# Set plot labels and title
plt.xlabel('Year')
plt.ylabel('Trade (Total USD)')
plt.title('Actual vs. Predicted Trade between Turkey and Germany')
plt.legend()

# Display the plot
plt.show()
```



Including MATR on the RHS

```
In [22]: # Load the MATR index data
matr_data = pd.read_excel("/Users/macbookpro/Desktop/ECON331 HW1/econ331 matr.xlsx")
```

```
In [23]: # Encode "yes" as 1 and "no" as 0
matr_data['Status'] = matr_data['Status'].map({'yes': 1, 'no': 0})
```

```
In [24]: # Sum "yes" for each country for each year
matr_summary = matr_data.groupby(['Year', 'Country'])['Status'].sum().reset_index()
```

```
In [25]: # Pivot the data to get MATR index for both countries throughout the years
matr_pivot = matr_summary.pivot(index='Year', columns='Country', values='Status').reset_index()
```

```
In [26]: # Rename the columns for clarity
matr_pivot.columns.name = None # Remove the 'Country' label from the columns
matr_pivot.columns = ['Year', 'MATR_Germany', 'MATR_Turkey']
```

```
In [27]: # Output the table
print(matr_pivot)
```

	Year	MATR_Germany	MATR_Turkey
0	2013	19	34
1	2014	19	34
2	2015	18	35
3	2016	18	35
4	2017	18	35
5	2018	18	38
6	2019	18	38
7	2020	18	36
8	2021	18	37

```
In [28]: import matplotlib.pyplot as plt

# Extract data
years = matr_pivot['Year']
matr_index_germany = matr_pivot['MATR_Germany']
matr_index_turkey = matr_pivot['MATR_Turkey']

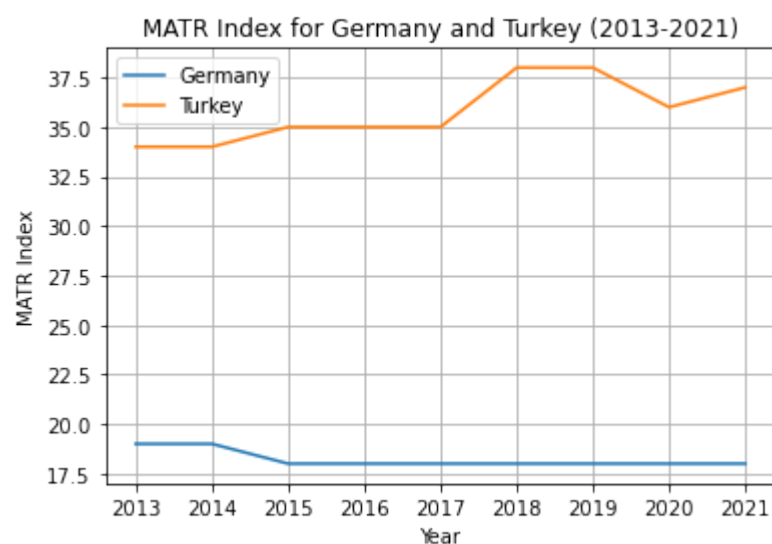
# Plot MATR index values for Germany
plt.plot(years, matr_index_germany, label='Germany')

# Plot MATR index values for Turkey
plt.plot(years, matr_index_turkey, label='Turkey')

# Add labels and title
plt.xlabel('Year')
plt.ylabel('MATR Index')
plt.title('MATR Index for Germany and Turkey (2013-2021)')

# Add legend
plt.legend()

# Show plot
plt.grid(True)
plt.show()
```



```
In [29]: # Merge MATR index data with merged data
matr_merged_data = pd.merge(merged_data, matr_pivot, on='Year', how='inner')
```

```
In [30]: matr_merged_data
```

Out[30]:

	Year	TOTAL USD	Germany GDP	Turkey GDP	Distance	MATR_Germany	MATR_Turkey
0	2013	40430998510	3.733805e+12	9.577991e+11	2350	19	34
1	2014	39757516843	3.889093e+12	9.389346e+11	2350	19	34
2	2015	37226299584	3.357586e+12	8.643138e+11	2350	18	35
3	2016	37812506839	3.469853e+12	8.696829e+11	2350	18	35
4	2017	38637760895	3.690849e+12	8.589885e+11	2350	18	35
5	2018	38888666525	3.974443e+12	7.789722e+11	2350	18	38
6	2019	35897642733	3.889178e+12	7.610059e+11	2350	18	38
7	2020	37711457138	3.887727e+12	7.203385e+11	2350	18	36
8	2021	41037327626	4.278504e+12	8.198653e+11	2350	18	37

```
In [31]: # Calculate MATR index product
matr_merged_data['MATR_Product'] = matr_merged_data['MATR_Germany'] * matr_merged_data['MATR_Turkey']
```

```
In [32]: # Extract independent variables
X = matr_merged_data[['Germany GDP', 'Turkey GDP', 'Distance', 'MATR_Product']].values

# Extract dependent variable
y = matr_merged_data['TOTAL USD'].values
```

```
In [33]: # Define the gravity model function
def gravity_model_with_matr(X, beta1, n, constant, beta2):
    return constant + (beta1 * (X[:, 0] * X[:, 1]) / (X[:, 2] ** n)) + beta2 * X[:, 3]
```

```
In [34]: # Initial guess for parameters
initial_guess = [0.5, 1, 1, 0.5]
```

```
In [35]: # Perform nonlinear least squares (NLS) estimation using curve_fit
params, covariance = curve_fit(gravity_model_with_matr, X, y, p0=initial_guess)

# Extract estimated parameters
beta1, n, constant, beta2 = params
```

```
In [36]: # Print the estimated parameters
print("Estimated beta1:", beta1)
print("Estimated n:", n)
print("Estimated constant:", constant)
print("Estimated beta2:", beta2)
```

```
Estimated beta1: 1.15406745732168e-05
Estimated n: 2.7933607126181954
Estimated constant: 27192357824.325413
Estimated beta2: -4120659.767225214
```

```
In [37]: # Calculate standard errors for parameters
beta_std_err = np.sqrt(np.diag(np.linalg.inv(np.dot(X.T, X))))
t_values = params / beta_std_err
```

```
In [38]: # Degrees of freedom
df = len(y) - len(params)
```

```
In [39]: # Calculate p-values for parameters
p_values = 2 * (1 - t.cdf(np.abs(t_values), df))
```

```
In [40]: # Print p-values for parameters with 3 decimals
print("P-values for Parameters (with 3 decimals):")
print("Estimated beta1:", format(p_values[0], '.3f'))
print("Estimated n:", format(p_values[1], '.3f'))
print("Estimated constant:", format(p_values[2], '.3f'))
print("Estimated beta2:", format(p_values[3], '.3f'))
```

```
P-values for Parameters (with 3 decimals):
Estimated beta1: 0.000
Estimated n: 0.000
Estimated constant: 0.000
Estimated beta2: 0.000
```

```
In [41]: # Predicted values
predicted_values = gravity_model_with_matr(X, *params)
```

In [42]: `from sklearn.metrics import r2_score, mean_squared_error`

```
# Calculate R-squared
r_squared = r2_score(y, predicted_values)

# Calculate mean squared error
mse = mean_squared_error(y, predicted_values)

# Print R-squared and MSE
print("R-squared (R^2):", r_squared)
print("Mean Squared Error (MSE):", mse)
```

R-squared (R<sup>2</sup>): 0.7256402921164997  
Mean Squared Error (MSE): 6.49841164470292e+17

In [43]: `# Set figure size
plt.figure(figsize=(10, 6)) # Adjust width and height as needed

# Plot actual trade data
plt.plot(matr_merged_data['Year'], matr_merged_data['TOTAL USD'], label='Actual Trade Data', marker='o')

# Plot fitted values
plt.plot(matr_merged_data['Year'], predicted_values, label='Fitted Values', marker='x')

# Add labels and title
plt.xlabel('Year')
plt.ylabel('Trade (Total USD)')
plt.title('Fit of Gravity Model with MATR Index to Actual Trade Data')
plt.legend()

# Show plot
plt.grid(True)
plt.show()`

