# **Regression Method**

```
In [1]: # Necessary packages
   import pandas as pd
   import numpy as np
   from numpy.linalg import inv
   import matplotlib.pyplot as plt

# Importing the CSV file
   df_reg = pd.read_csv("treasury_data_all.csv")
   df_reg.head()
```

#### Out[1]:

|   | CUSIP    | ISSUE_DATE | MATURITY_DATE | COUPON_RATE | BID_PRICE  | ASK_PRICE  | TIME_TO_MATURITY |
|---|----------|------------|---------------|-------------|------------|------------|------------------|
| 0 | 912810EB | 22/11/1988 | 15/11/2018    | 9.000       | 103.093750 | 103.218750 | 0.5              |
| 1 | 912828M6 | 16/11/2015 | 15/11/2018    | 1.250       | 99.609375  | 99.648438  | 0.5              |
| 2 | 912828JR | 17/11/2008 | 15/11/2018    | 3.750       | 100.734375 | 100.757812 | 0.5              |
| 3 | 912796QJ | 17/05/2018 | 15/11/2018    | 0.000       | 99.069000  | 99.073667  | 0.5              |
| 4 | 912828R4 | 16/05/2016 | 15/05/2019    | 0.875       | 98.679688  | 98.718750  | 1.0              |

This is an overview over the different variables which is located in this dataframe

- CUSIP: A unique security-level identifier
- ISSUE\_DATE: The date the Treasury bond was issued
- MATURITY\_DATEThe date the Treasury bond matures
- COUPON RATEThe bond's coupon rate in percent
- BID\_PRICEThe price at which dealers are willing to buy
- ASK\_PRICEThe price at which dealers are willing to sell
- TIME\_TO\_MATURITYThe time-to-maturity measured in years

### Out[2]:

|   | CUSIP    | ISSUE_DATE | MATURITY_DATE | COUPON_RATE | BID_PRICE  | ASK_PRICE  | TIME_TO_MATURITY | Mid_Price  |
|---|----------|------------|---------------|-------------|------------|------------|------------------|------------|
| 0 | 912810EB | 22/11/1988 | 15/11/2018    | 9.000       | 103.093750 | 103.218750 | 0.5              | 103.156250 |
| 1 | 912828M6 | 16/11/2015 | 15/11/2018    | 1.250       | 99.609375  | 99.648438  | 0.5              | 99.628906  |
| 2 | 912828JR | 17/11/2008 | 15/11/2018    | 3.750       | 100.734375 | 100.757812 | 0.5              | 100.746094 |
| 3 | 912796QJ | 17/05/2018 | 15/11/2018    | 0.000       | 99.069000  | 99.073667  | 0.5              | 99.071333  |
| 4 | 912828R4 | 16/05/2016 | 15/05/2019    | 0.875       | 98.679688  | 98.718750  | 1.0              | 98.699219  |

```
In [3]: ## Creating the Cash Flow Matrix
        unique_maturities = sorted(df_reg["TIME_TO_MATURITY"].unique())
        # Initialize the cash flow matrix with zeroes
        cash_flow_matrix = np.zeros((len(df_reg), len(unique_maturities)))
        # Iterative process to create a vector row of each bonds casf flows at different maturities
        for i, bond in df reg.iterrows():
            for j, maturity in enumerate(unique maturities):
                # Coupon payment for maturities before, and at the bonds maturity
                cash_flow = 100*bond["COUPON_RATE"] / 100 / 2 # Assuming semiannual
                if maturity == bond["TIME_TO_MATURITY"]:
                    cash flow += 100 # Adding principal
                cash_flow_matrix[i, j] = cash_flow
        # We convert the cash flow matrix to Dataframe
        C = pd.DataFrame(cash_flow_matrix, columns = unique_maturities)
        C.head()
Out[3]:
```

|   | 0.5      | 1.0      | 1.5    | 2.0    | 2.5    | 3.0    | 3.5    | 4.0    | 4.5    | 5.0    | 5.5    | 6.0    | 6.5    | 7.0    | 7.5    | 8.0    |    |
|---|----------|----------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|----|
| 0 | 104.5000 | 4.5000   | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4. |
| 1 | 100.6250 | 0.6250   | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0.6250 | 0. |
| 2 | 101.8750 | 1.8750   | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1.8750 | 1. |
| 3 | 100.0000 | 0.0000   | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0. |
| 4 | 0.4375   | 100.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0. |
|   |          |          |        |        |        |        |        |        |        |        |        |        |        |        |        |        | •  |

```
In [4]: ## Step 2. Prepare the Price Vector
        price_vector = df_reg["Mid_Price"].values
```

```
In [5]: # Transpose and multiply the cash flow matrix
CT_C = np.dot(C.T, C)

# Invert the resulting matrix
inv_CT_C = inv(CT_C)

# Multiply by tranposed cash flow matrix and price vector
Z_0 = np.dot(np.dot(inv_CT_C, C.T), price_vector)

# Convert the discount factors to DataFrame
df_reg_discount = pd.DataFrame(Z_0)
df_reg_discount["Discount_Factors"] = df_reg_discount[0]
df_reg_discount = df_reg_discount.drop(df_reg_discount.columns[0], axis = 1)
df_reg_discount.head()
```

#### Out[5]:

|   | Discount_Factors |
|---|------------------|
| 0 | 0.733487         |
| 1 | 0.830802         |
| 2 | 0.816418         |
| 3 | 0.683638         |
| 4 | 0.811555         |

```
In [6]: ## Task 2 is to calculate the continously compounded zero-coupon bond yields for each maturity

# We will be using the same discount factors as calculated above

df_reg_discount["Continously_Compounded_Yield"] = -np.log(df_reg_discount["Discount_Factors"]) / df_reg_disco

# Convert the yield to percent

df_reg_discount["Continously_Compounded_Yield_in%"] = df_reg_discount.Continously_Compounded_Yield * 100

# Plot the term structure

plt.figure(figsize = (10, 6))

plt.plot(df_reg_discount.index, df_reg_discount["Continously_Compounded_Yield_in%"], marker = "o")

plt.title("Term structure of Zero-Coupon Yields")

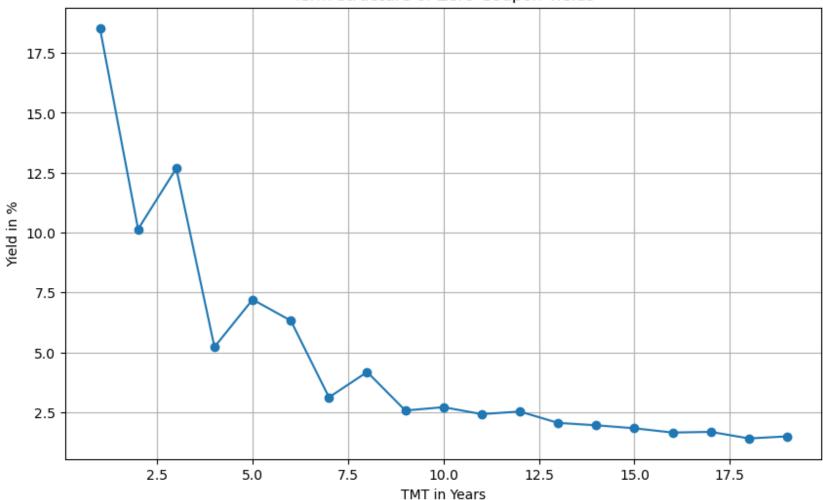
plt.xlabel("TMT in Years")

plt.ylabel("Yield in %")

plt.grid(True)

plt.show()
```





# **Nelson Siegel**

In [7]: # Necessary package to optimize
from scipy.optimize import minimize

```
In [8]: # We define the df_NS as same as the df_reg
    df_NS = df_reg

# Define the NS model functions
def nelson_siegel_yield(theta_0, theta_1, theta_2, lambda_, T):
        # Calculates the yield for a given maturity T based on NS
        return theta_0 + (theta_1 + theta_2)*(1-np.exp(-T / lambda_)) / (T / lambda_) - theta_2 * np.exp(-T / lam

def discount_factor(theta_0, theta_1, theta_2, lambda_, T):
    # Calculates the discount factor for a given maturity
    r = nelson_siegel_yield(theta_0, theta_1, theta_2, lambda_, T)
    return np.exp(-r * T)
```

```
In [9]: # Defining the objective function for Optimization
        def objective_function(params, df_NS):
            theta_0, theta_1, theta_2, lambda_ = params
            squared_errors = []
            for index, row in df_NS.iterrows():
                T = row["TIME TO MATURITY"]
                coupon = row["COUPON RATE"] / 100
                mid price = row["Mid Price"]
                face value = 100 # We assume that the standard face value is 100
                # Calculate the price of the bond under the NS model
                # We will assume semi-annual coupons
                cash_flows = [coupon * face_value / 2 for _ in range(int(2*T))]
                cash flows[-1] += face value # Add the Face value as the last payment
                model_price = sum(discount_factor(theta_0, theta_1, theta_2, lambda_, t/2) * cf for t, cf in enumerat
                # Calculate the squared error between the model price and the observed mid price
                squared errors.append((model price - mid price) ** 2)
            return np.sum(squared_errors)
```

```
In [10]: # We will initialize the optimization model by a guess for our thetas, lambda and T initial_guess = [0.03, 0, 0, 3]

# Start the optimization result = minimize(objective_function, initial_guess, args = (df_NS, ), method = 'L-BFGS-B')

# Print out the result print("Optimized Parameters: 00 = {:.4f}, 01 = {:.4f}, 02 = {:.4f}, \lambda = {:.4f}".format(*result.x))
```

Optimized Parameters:  $\theta\theta = 0.0300$ ,  $\theta1 = -0.0092$ ,  $\theta2 = 0.0054$ ,  $\lambda = 3.0000$ 

In [11]: ### Next task is to se the estimated model parameters to calculate the zero-coupon bond yield for each maturi:
 ## We extract our optimized parameters so it could be later used
 optimized\_params = result.x
 theta\_0, theta\_1, theta\_2, lambda\_ = optimized\_params

```
In [12]: # Extract the unique maturity values
    unique_maturities = df_NS["TIME_TO_MATURITY"].unique()

# We then calculate yeuekds for each maturity
    zero_cp_yields = {T: nelson_siegel_yield(theta_0, theta_1, theta_2, lambda_, T) for T in unique_maturities}

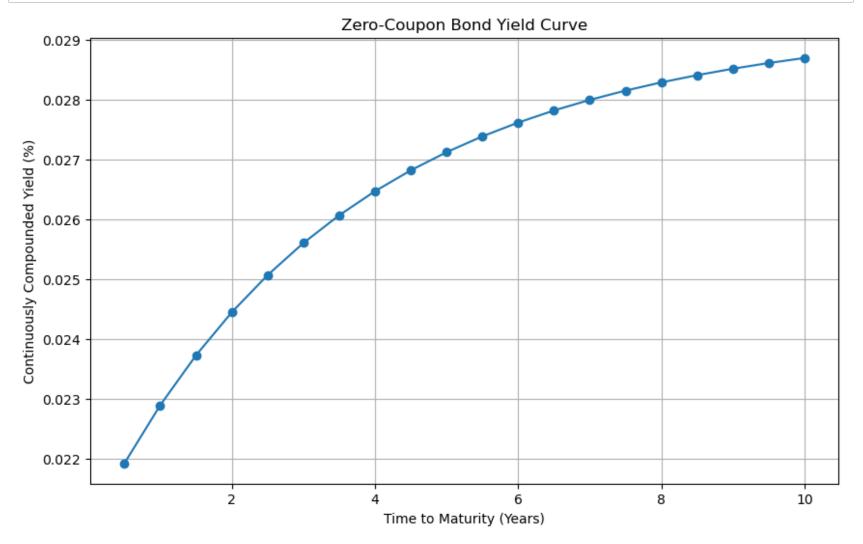
# We convert the dictionary to a DF
    df_NS_model = pd.DataFrame(list(zero_cp_yields.items()), columns = ["Maturity", "Continously_Compounded_Yield
    df_NS_model.head()
```

#### Out[12]:

|   | Maturity | Continously_Compounded_Yield_in% |
|---|----------|----------------------------------|
| 0 | 0.5      | 0.021928                         |
| 1 | 1.0      | 0.022898                         |
| 2 | 1.5      | 0.023732                         |
| 3 | 2.0      | 0.024451                         |
| 4 | 2.5      | 0.025071                         |

```
In [13]: # We can then plot the term structure of interest rates obtained in our calculation

plt.figure(figsize=(10, 6))
plt.plot(df_NS_model['Maturity'], df_NS_model['Continously_Compounded_Yield_in%'], marker='o')
plt.title('Zero-Coupon Bond Yield Curve')
plt.xlabel('Time to Maturity (Years)')
plt.ylabel('Continuously Compounded Yield (%)')
plt.grid(True)
plt.show()
```



# **Bootstrap Method**

```
In [14]: df_bot = pd.read_csv("treasury_data_bootstrap.csv")
df_bot.head()
```

#### Out[14]:

|   | CUSIP    | ISSUE_DATE | MATURITY_DATE | COUPON_RATE | BID_PRICE  | ASK_PRICE  | TIME_TO_MATURITY |
|---|----------|------------|---------------|-------------|------------|------------|------------------|
| 0 | 912810EB | 22/11/1988 | 15/11/2018    | 9.000       | 103.093750 | 103.218750 | 0.5              |
| 1 | 912828R4 | 16/05/2016 | 15/05/2019    | 0.875       | 98.679688  | 98.718750  | 1.0              |
| 2 | 912828LY | 16/11/2009 | 15/11/2019    | 3.375       | 101.382812 | 101.429688 | 1.5              |
| 3 | 912828X9 | 15/05/2017 | 15/05/2020    | 1.500       | 98.218750  | 98.257812  | 2.0              |
| 4 | 9128283G | 15/11/2017 | 15/11/2020    | 1.750       | 98.195312  | 98.234375  | 2.5              |

In [15]: df\_bot["Mid\_Price"] = (df\_bot.ASK\_PRICE + df\_bot.BID\_PRICE)/2
df\_bot.head()

#### Out[15]:

|   | CUSIP    | ISSUE_DATE | MATURITY_DATE | COUPON_RATE | BID_PRICE  | ASK_PRICE  | TIME_TO_MATURITY | Mid_Price  |
|---|----------|------------|---------------|-------------|------------|------------|------------------|------------|
| 0 | 912810EB | 22/11/1988 | 15/11/2018    | 9.000       | 103.093750 | 103.218750 | 0.5              | 103.156250 |
| 1 | 912828R4 | 16/05/2016 | 15/05/2019    | 0.875       | 98.679688  | 98.718750  | 1.0              | 98.699219  |
| 2 | 912828LY | 16/11/2009 | 15/11/2019    | 3.375       | 101.382812 | 101.429688 | 1.5              | 101.406250 |
| 3 | 912828X9 | 15/05/2017 | 15/05/2020    | 1.500       | 98.218750  | 98.257812  | 2.0              | 98.238281  |
| 4 | 9128283G | 15/11/2017 | 15/11/2020    | 1.750       | 98.195312  | 98.234375  | 2.5              | 98.214844  |

```
## Step 1. Creating the Cash Flow Matrix
In [16]:
         unique_maturities = sorted(df_bot["TIME_TO_MATURITY"].unique())
         # Initialize the cash flow matrix with zeroes
         cash_flow_matrix = np.zeros((len(df_bot), len(unique_maturities)))
         # Iterative process to create a vector row of each bonds casf flows at different maturities
         for i, bond in df_bot.iterrows():
             for j, maturity in enumerate(unique maturities):
                 # Coupon payment for maturities before, and at the bonds maturity
                 cash_flow = 100*bond["COUPON_RATE"] / 100 / 2 # Assuming semiannual
                 if maturity == bond["TIME_TO_MATURITY"]:
                     cash flow += 100
                                       # Adding principal
                 cash_flow_matrix[i, j] = cash_flow
         # We convert the cash flow matrix to Dataframe
         df_cash_flow_matrix = pd.DataFrame(cash_flow_matrix, columns = unique_maturities)
         df_cash_flow_matrix.head()
```

#### Out[16]:

|   | 0.5      | 1.0      | 1.5      | 2.0      | 2.5      | 3.0    | 3.5    | 4.0    | 4.5    | 5.0    | 5.5    | 6.0    | 6.5    | 7.0    | 7.5    |      |
|---|----------|----------|----------|----------|----------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|------|
| 0 | 104.5000 | 4.5000   | 4.5000   | 4.5000   | 4.5000   | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.5000 | 4.50 |
| 1 | 0.4375   | 100.4375 | 0.4375   | 0.4375   | 0.4375   | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.4375 | 0.43 |
| 2 | 1.6875   | 1.6875   | 101.6875 | 1.6875   | 1.6875   | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6875 | 1.6  |
| 3 | 0.7500   | 0.7500   | 0.7500   | 100.7500 | 0.7500   | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7500 | 0.7  |
| 4 | 0.8750   | 0.8750   | 0.8750   | 0.8750   | 100.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8750 | 0.8  |
| 4 |          |          |          |          |          |        |        |        |        |        |        |        |        |        |        |      |

```
In [17]: ## Step 2. Prepare the Price Vector
         price_vector = df_bot["Mid_Price"].values
```

#### Out[18]:

|   | Discount_Factors |
|---|------------------|
| 0 | 0.335418         |
| 1 | 0.919311         |
| 2 | 0.753008         |
| 3 | 0.866359         |
| 4 | 0.846787         |

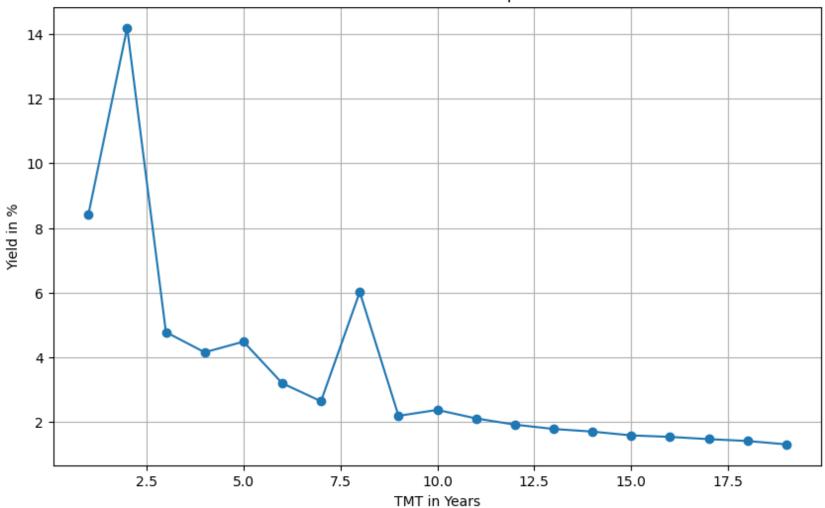
```
In [19]: ## Task 2 is to calculate the continously compounded zero-coupon bond yields for each maturity

# We will be using the same Discount_Factors as calculated above
df_bootstrap["Continously_Compounded_Yield"] = -np.log(df_bootstrap["Discount_Factors"]) / df_bootstrap.index

# Convert the yield to percent
df_bootstrap["Continously_Compounded_Yield_in%"] = df_bootstrap.Continously_Compounded_Yield * 100

# Plot the term structure
plt.figure(figsize = (10, 6))
plt.plot(df_bootstrap.index, df_bootstrap["Continously_Compounded_Yield_in%"], marker = "o")
plt.title("Term structure of Zero-Coupon Yields")
plt.xlabel("TMT in Years")
plt.ylabel("Yield in %")
plt.grid(True)
plt.show()
```

## Term structure of Zero-Coupon Yields



## **Plotting BootStrap and Regression**

```
In [20]: # Initialize the figure
    plt.figure(figsize=(12,10))

# Plotting the Bootstrap
    plt.plot(df_bootstrap.index, df_bootstrap["Continously_Compounded_Yield_in%"], label = "Bootstrap Method", ma

# Plotting Regression
    plt.plot(df_reg_discount.index, df_reg_discount["Continously_Compounded_Yield_in%"], label = "Regression Method"

# Adding titles and Labels
    plt.title('Term Structures from Different Methods')
    plt.ylabel('Time to Maturity (Years)')
    plt.ylabel('Yield (%)')
    plt.legend()
    plt.grid(True)

# Show plot
    plt.show()
```

### Term Structures from Different Methods

