### Step 1: Import Data

```
Python > D598 - Task 1.py > ...
1    import pandas as pd
2
3    # Step 1: Import Data
4    data_file = 'C:/Users/kyleh/Downloads/D598 Data Set.xlsx'
5    df = pd.read_excel(data_file)
```

Purpose: Load the Excel data into a Pandas DataFrame (df).

**Explanation**: The pd.read\_excel function reads the Excel file located at the specified path into a DataFrame, a structured, table-like data format used in Python for analysis.

## Step 2: Identify Duplicate Rows

```
7  # Step 2: Identify Duplicate Rows
8  duplicates = df[df.duplicated()]
9  print("Duplicate Rows:")
10  print(duplicates)
```

**Purpose**: Identify and display rows in the dataset that are exact duplicates of other rows. **Explanation**: The duplicated() method returns a Boolean Series indicating whether each row is a duplicate of a previous row. Rows where True is returned are filtered out and stored in the duplicate DataFrame for inspection.

Step 3: Group by State and Calculate Descriptive Statistics

```
# Step 3: Group by State and Calculate Descriptive Statistics
grouped_stats = df.groupby('Business State').agg({
    'Total Long-term Debt': ['mean', 'median', 'min', 'max'],
    'Total Equity': ['mean', 'median', 'min', 'max'],
    'Total Liabilities': ['mean', 'median', 'min', 'max'],
    'Total Revenue': ['mean', 'median', 'min', 'max'],
    'Profit Margin': ['mean', 'median', 'min', 'max']
}).reset_index()
```

**Purpose**: Calculate summary statistics (mean, median, minimum, and maximum) for specific columns grouped by the 'Business State' column.

#### Explanation:

- The groupby method groups the data by the 'Business State' column.
- The agg method applies aggregation functions (mean, median, min, max) to the specified columns.
- The result is a new DataFrame (grouped\_stats) containing these statistics for each state.

#### Step 4: Filter Businesses with Negative Debt-to-Equity Ratios

```
# Step 4: Filter Businesses with Negative Debt-to-Equity Ratios
negative_debt_to_equity = df[df['Debt to Equity'] < 0]
print("Businesses with Negative Debt-to-Equity Ratios:")
print(negative_debt_to_equity)
```

**Purpose**: Add a new column to the DataFrame that calculates the Debt-to-Income ratio for each business.

**Explanation**: This ratio is computed by dividing the 'Total Long-term Debt' by 'Total Revenue', providing insight into how leveraged a business is relative to its income.

#### Step 5: Create Debt-to-Income Ratio

```
# Step 5: Create Debt-to-Income Ratio
df['Debt-to-Income'] = df['Total Long-term Debt'] / df['Total Revenue']
```

**Purpose**: Add a new column to the DataFrame that calculates the Debt-to-Income ratio for each business.

**Explanation**: This ratio is computed by dividing the 'Total Long-term Debt' by 'Total Revenue', providing insight into how leveraged a business is relative to its income.

#### Step 6: Concatenate the New DataFrame

```
# Step 6: Concatenate the New DataFrame
# Creating a new DataFrame with only Debt-to-Income Ratio

debt_to_income_df = df[['Business ID', 'Debt-to-Income']]

final_df = pd.merge(df, debt_to_income_df, on='Business ID')
```

**Purpose**: Merge the existing DataFrame with a new DataFrame that includes only the 'Business ID' and 'Debt-to-Income' columns.

## **Explanation**:

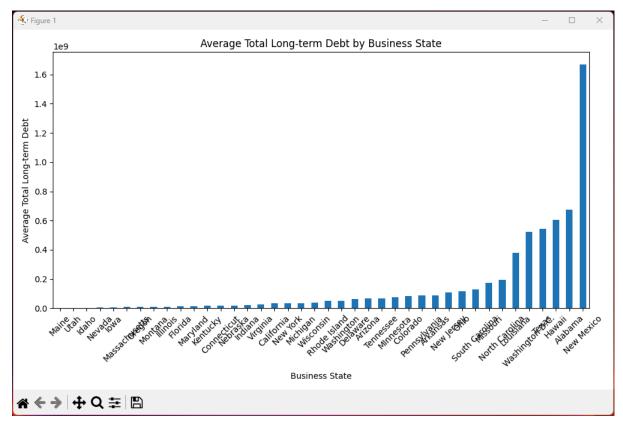
- The merge method combines df with debt\_to\_income\_df using 'Business
  ID' as the key.
- This ensures the final DataFrame (final\_df) contains all original columns and retains the Debt-to-Income ratio as part of its structure.

#### Step 7: Save the Final DataFrame

```
# Step 7: Save the Final DataFrame
final_df.to_excel('Final_Analysis_Output.xlsx', index=False)
print("Analysis Completed. Results saved as 'Final_Analysis_Output.xlsx'.")
```

**Purpose**: Export the final DataFrame to a new Excel file.

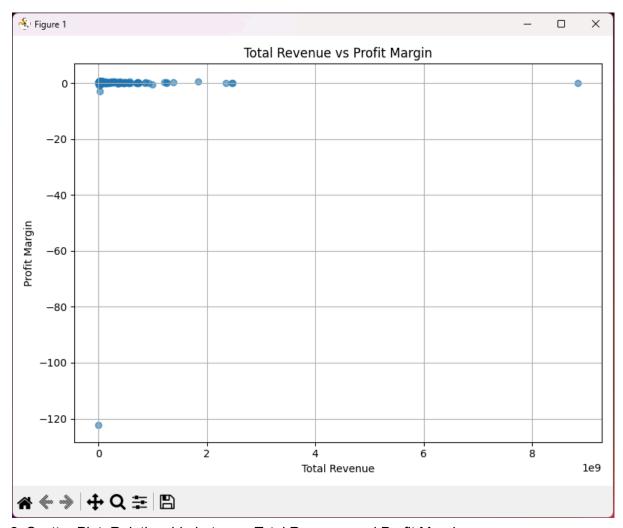
**Explanation**: The to\_excel method saves the DataFrame as an Excel file named Final\_Analysis\_Output.xlsx in the current working directory. The index=False argument ensures the row index is not included in the output.



## 1. Bar Chart: Average Total Long-term Debt by Business State

**Purpose**: Visualize the average Total Long-term Debt for businesses in each state. **Steps**:

- 1. The data was grouped by the column 'Business State' using groupby.
- 2. Calculated the mean of the 'Total Long-term Debt' for each group.
- 3. Sorted the values for better visual organization.
- 4. Plotted the data as a bar chart using the Pandas plot method.
- 5. Enhanced readability with labels, a title, and proper layout adjustments.



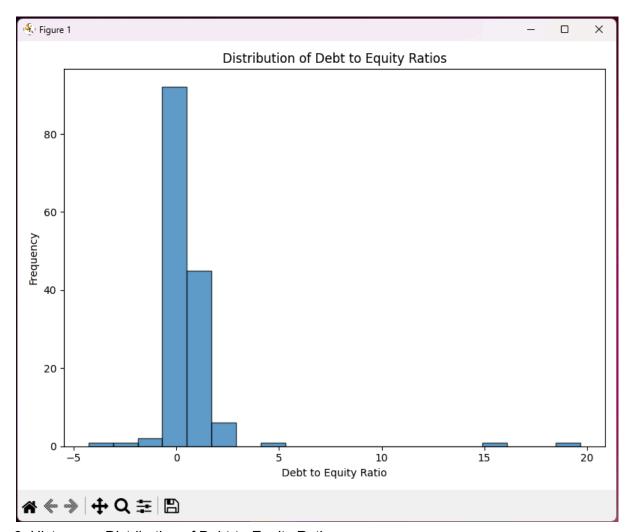
2. Scatter Plot: Relationship between Total Revenue and Profit Margin

```
# 2. Scatter Plot: Relationship between Total Revenue and Profit Margin
plt.figure(figsize=(8, 6))
plt.scatter(df['Total Revenue'], df['Profit Margin'], alpha=0.6)
plt.title('Total Revenue vs Profit Margin')
plt.xlabel('Total Revenue')
plt.ylabel('Profit Margin')
plt.grid(True)
plt.tight_layout()
plt.show()
```

**Purpose**: Show the relationship between a business's Total Revenue and its Profit Margin.

# Steps:

- 1. Used plt.scatter to create a scatter plot.
- 2. Mapped Total Revenue to the x-axis and Profit Margin to the y-axis.
- 3. Adjusted transparency with alpha for better visualization of overlapping points.
- 4. Added gridlines and axis labels for clarity.



3. Histogram: Distribution of Debt-to-Equity Ratios

```
# 3. Histogram: Distribution of Debt to Equity Ratios

plt.figure(figsize=(8, 6))

plt.hist(df['Debt to Equity'], bins=20, edgecolor='black', alpha=0.7)

plt.title('Distribution of Debt to Equity Ratios')

plt.xlabel('Debt to Equity Ratio')

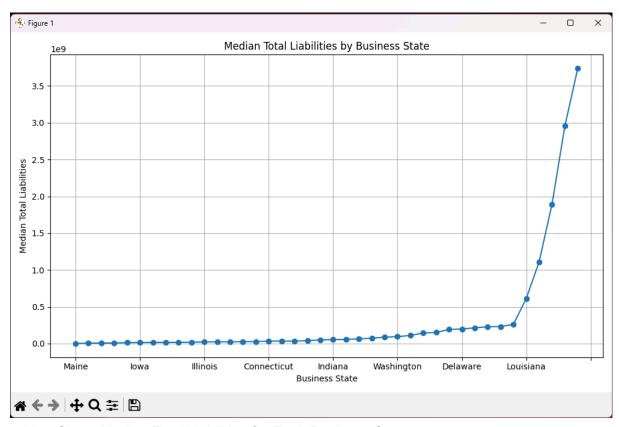
plt.ylabel('Frequency')

plt.tight_layout()

plt.show()
```

**Purpose**: Display the frequency distribution of Debt to Equity ratios across businesses. **Steps**:

- 1. Used plt.hist to create a histogram of the 'Debt to Equity' column.
- 2. Set the number of bins to 20 for more granularity.
- 3. Added edge color and adjusted transparency for better aesthetics.
- 4. Enhanced with labels and a title for better interpretation.



4. Line Chart: Median Total Liabilities for Each Business State

```
# 4. Line Chart: Median Total Liabilities for each Business State

median_liabilities_by_state = df.groupby('Business State')['Total Liabilities'].median().sort_values()

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

median_liabilities_by_state.plot(kind='line', marker='o', title='Median Total Liabilities by Business State')

plt.ylabel('Median Total Liabilities')

plt.xlabel('Business State')

plt.grid(True)

plt.tight_layout()

plt.show()

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```

**Purpose**: Show the median Total Liabilities for businesses in each state as a trend over the sorted states.

# Steps:

- 1. Grouped the data by 'Business State' and calculated the median of 'Total Liabilities'.
- 2. Sorted the results for logical ordering.
- 3. Used a line chart (kind='line') with markers (marker='o') to make data points visible.
- 4. Added gridlines, labels, and a title for better readability.

## References

- The Pandas Development Team. (n.d.). *Group by: Split-apply-combine*. Retrieved January 6, 2025, from https://pandas.pydata.org/docs/user\_guide/groupby.html
- The Matplotlib Development Team. (n.d.). *Pyplot tutorial*. Retrieved January 6, 2025, from https://matplotlib.org/stable/tutorials/introductory/pyplot.html
- D598 Course Materials. (2025). *Master of Data Engineering Task 2 assignment:* Financial data analysis and visualizations. Instructional content provided by Western Governors University.
- Colby, K. (2025). *Python code for data analysis and visualizations* [Personal knowledge and experience].