# Lab 15 Box Plot

April 7, 2025

## 1 Box Plots

Estimated time needed: 45 minutes

In this lab, you will focus on the visualization of data. The dataset will be provided through an RDBMS, and you will need to use SQL queries to extract the required data.

## 1.1 Objectives

In this lab you will perform the following:

- Visualize the distribution of data.
- Visualize the relationship between two features.
- Visualize data composition and comparisons using box plots.

### 1.1.1 Setup: Connecting to the Database

## 1. Download the Database File

--2025-04-07 11:30:28-- https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/QR9YeprUYhOoLafzlLspAw/survey-results-public.sqlite Resolving cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud (cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud)... 169.63.118.104 Connecting to cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud (cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud)|169.63.118.104|:443... connected.

200 OKequest sent, awaiting response...

Length: 211415040 (202M) [application/octet-stream]

Saving to: 'survey-results-public.sqlite.2'

survey-results-publ 100%[===========] 201.62M 69.1MB/s in 2.9s

2025-04-07 11:30:34 (69.1 MB/s) - 'survey-results-public.sqlite.2' saved [211415040/211415040]

#### 2. Connect to the Database Install the needed libraries

# [2]: !pip install pandas Requirement already satisfied: pandas in /opt/conda/lib/python3.12/site-packages Requirement already satisfied: numpy>=1.26.0 in /opt/conda/lib/python3.12/sitepackages (from pandas) (2.2.4) Requirement already satisfied: python-dateutil>=2.8.2 in /opt/conda/lib/python3.12/site-packages (from pandas) (2.9.0.post0) Requirement already satisfied: pytz>=2020.1 in /opt/conda/lib/python3.12/sitepackages (from pandas) (2024.2) Requirement already satisfied: tzdata>=2022.7 in /opt/conda/lib/python3.12/sitepackages (from pandas) (2025.2) Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.12/sitepackages (from python-dateutil>=2.8.2->pandas) (1.17.0) [3]: !pip install matplotlib !pip install seaborn Requirement already satisfied: matplotlib in /opt/conda/lib/python3.12/sitepackages (3.10.1) Requirement already satisfied: contourpy>=1.0.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (1.3.1) Requirement already satisfied: cycler>=0.10 in /opt/conda/lib/python3.12/sitepackages (from matplotlib) (0.12.1) Requirement already satisfied: fonttools>=4.22.0 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (4.57.0) Requirement already satisfied: kiwisolver>=1.3.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (1.4.8) Requirement already satisfied: numpy>=1.23 in /opt/conda/lib/python3.12/sitepackages (from matplotlib) (2.2.4) Requirement already satisfied: packaging>=20.0 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (24.2) Requirement already satisfied: pillow>=8 in /opt/conda/lib/python3.12/sitepackages (from matplotlib) (11.1.0) Requirement already satisfied: pyparsing>=2.3.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (3.2.3) Requirement already satisfied: python-dateutil>=2.7 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (2.9.0.post0) Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.12/sitepackages (from python-dateutil>=2.7->matplotlib) (1.17.0) Requirement already satisfied: seaborn in /opt/conda/lib/python3.12/sitepackages (0.13.2) Requirement already satisfied: numpy!=1.24.0,>=1.20 in /opt/conda/lib/python3.12/site-packages (from seaborn) (2.2.4) Requirement already satisfied: pandas>=1.2 in /opt/conda/lib/python3.12/sitepackages (from seaborn) (2.2.3) Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in

```
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     Requirement already satisfied: contourpy>=1.0.1 in
     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
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     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
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     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
     Requirement already satisfied: packaging>=20.0 in
     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
     Requirement already satisfied: pillow>=8 in /opt/conda/lib/python3.12/site-
     packages (from matplotlib!=3.6.1,>=3.4->seaborn) (11.1.0)
     Requirement already satisfied: pyparsing>=2.3.1 in
     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
     (3.2.3)
     Requirement already satisfied: python-dateutil>=2.7 in
     /opt/conda/lib/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
     (2.9.0.post0)
     Requirement already satisfied: pytz>=2020.1 in /opt/conda/lib/python3.12/site-
     packages (from pandas>=1.2->seaborn) (2024.2)
     Requirement already satisfied: tzdata>=2022.7 in /opt/conda/lib/python3.12/site-
     packages (from pandas>=1.2->seaborn) (2025.2)
     Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.12/site-
     packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.4->seaborn) (1.17.0)
[12]: import sqlite3
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      # Connect to the SQLite database
      conn = sqlite3.connect('survey-results-public.sqlite')
     1.2 Demo: Basic SQL Queries
     Demo 1: Count the Number of Rows in the Table
 [5]: QUERY = "SELECT COUNT(*) FROM main"
      df = pd.read_sql_query(QUERY, conn)
      print(df)
        COUNT(*)
     0
           65437
```

Demo 2: List All Tables

```
[6]: QUERY = """
    SELECT name as Table_Name
    FROM sqlite_master
    WHERE type = 'table'
    """
    pd.read_sql_query(QUERY, conn)
```

[6]: Table\_Name
0 main

## Demo 3: Group Data by Age

```
[7]: QUERY = """
    SELECT Age, COUNT(*) as count
    FROM main
    GROUP BY Age
    ORDER BY Age
    """
    df_age = pd.read_sql_query(QUERY, conn)
    print(df_age)
```

```
Age count
     18-24 years old 14098
0
     25-34 years old 23911
1
2
     35-44 years old 14942
3
     45-54 years old
                     6249
4
     55-64 years old
                       2575
   65 years or older
                        772
5
  Prefer not to say
                        322
7 Under 18 years old
                       2568
```

## 1.3 Visualizing Data

#### 1.3.1 Task 1: Visualizing the Distribution of Data

#### 1. Box Plot of CompTotal (Total Compensation)

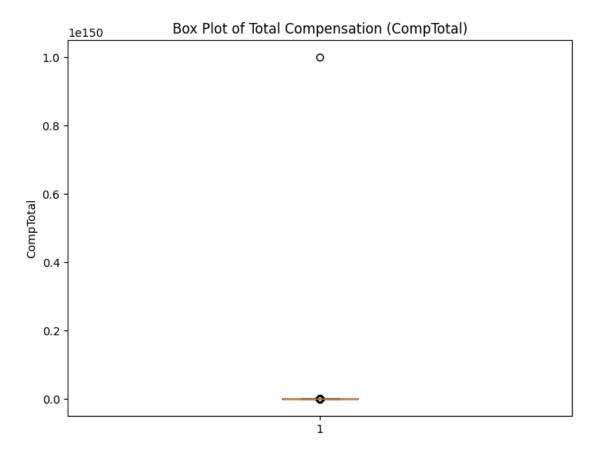
Use a box plot to analyze the distribution and outliers in total compensation.

```
[8]: query_check = "SELECT CompTotal FROM main WHERE CompTotal IS NOT NULL LIMIT 5"
    df_check = pd.read_sql(query_check, conn)
    print(df_check)

if not df_check.empty:
    query = "SELECT CompTotal FROM main WHERE CompTotal IS NOT NULL"
    df = pd.read_sql(query, conn)
plt.figure(figsize=(8, 6))
```

```
plt.boxplot(df['CompTotal'])
  plt.title('Box Plot of Total Compensation (CompTotal)')
  plt.ylabel('CompTotal')
  plt.show()
else:
    print("No data found for CompTotal.")
```

```
CompTotal
0 2040000.0
1 28000.0
2 85000.0
3 50000.0
4 110000.0
```



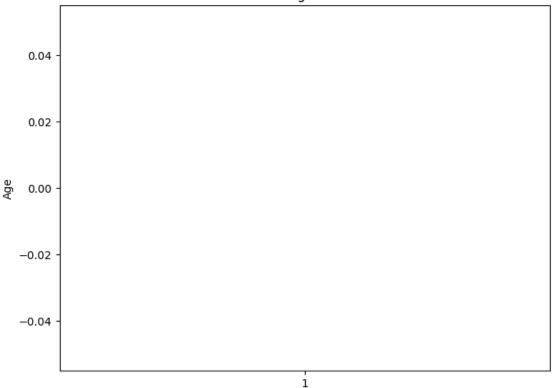
## 2. Box Plot of Age (converted to numeric values)

Convert the Age column into numerical values and visualize the distribution.

```
[13]: # Check the first few rows to inspect the Age column
query_check = "SELECT Age FROM main LIMIT 5"
df_check = pd.read_sql(query_check, conn)
```

Age
0 Under 18 years old
1 35-44 years old
2 45-54 years old
3 18-24 years old
4 18-24 years old





## 1.3.2 Task 2: Visualizing Relationships in Data

#### 1. Box Plot of CompTotal Grouped by Age Groups:

Visualize the distribution of compensation across different age groups.

```
[14]: # Load Age and CompTotal columns
    query = "SELECT Age, CompTotal FROM main"
    df = pd.read_sql(query, conn)

# Close the connection
    conn.close()

# Convert 'Age' to numeric and drop rows with NaNs
    df['Age'] = pd.to_numeric(df['Age'], errors='coerce')
    df['CompTotal'] = pd.to_numeric(df['CompTotal'], errors='coerce')
    df = df.dropna(subset=['Age', 'CompTotal'])

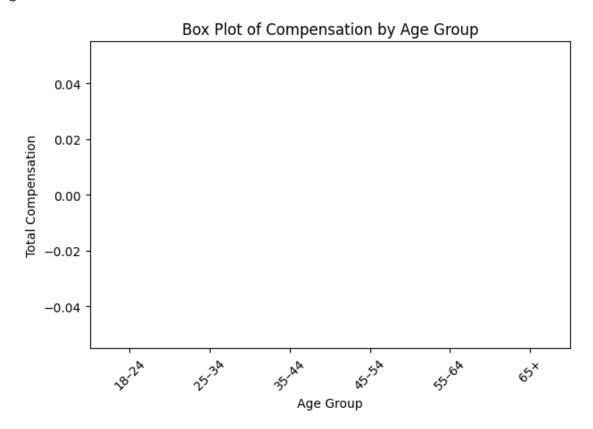
# Define age groups
bins = [18, 25, 35, 45, 55, 65, 100]
labels = ['18-24', '25-34', '35-44', '45-54', '55-64', '65+']
```

```
df['AgeGroup'] = pd.cut(df['Age'], bins=bins, labels=labels, right=False)

# Drop rows with undefined age groups (if any)
df = df.dropna(subset=['AgeGroup'])

# Plot box plot
plt.figure(figsize=(10, 6))
df.boxplot(column='CompTotal', by='AgeGroup', grid=False)
plt.title('Box Plot of Compensation by Age Group')
plt.suptitle('')
plt.xlabel('Age Group')
plt.ylabel('Total Compensation')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

<Figure size 1000x600 with 0 Axes>



## 2. Box Plot of CompTotal Grouped by Job Satisfaction (JobSatPoints\_6):

Examine how compensation varies based on job satisfaction levels.

```
[]: # your code goes here
```

#### 1.3.3 Task 3: Visualizing the Composition of Data

#### 1. Box Plot of ConvertedCompYearly for the Top 5 Developer Types:

Analyze compensation across the top 5 developer roles.

[]: # your code goes here

## 2. Box Plot of CompTotal for the Top 5 Countries:

Analyze compensation across respondents from the top 5 countries.

[]: # your code goes here

## 1.3.4 Task 4: Visualizing Comparison of Data

## 1. Box Plot of CompTotal Across Employment Types:

Analyze compensation for different employment types.

[]: # your code goes here

## 2. Box Plot of YearsCodePro by Job Satisfaction (JobSatPoints\_6):

Examine the distribution of professional coding years by job satisfaction levels.

[]: # your code goes here

#### 1.3.5 Final Step: Close the Database Connection

After completing the lab, close the connection to the SQLite database:

[]: conn.close()

#### 1.4 Summary

In this lab, you used box plots to visualize various aspects of the dataset, focusing on:

- Visualize distributions of compensation and age.
- Explore relationships between compensation, job satisfaction, and professional coding experience.
- Analyze data composition across developer roles and countries.
- Compare compensation across employment types and satisfaction levels.

Box plots provided clear insights into the spread, outliers, and central tendencies of various features in the dataset.

#### 1.5 Authors:

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# 1.5.1 Other Contributors:

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