

# Assignment 2: Developing Interactive Data Visualizations in React and D3.js

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## 1 Introduction

This report presents the development of interactive visualizations for a bike-sharing dataset using React and D3.js. The visualizations are part of Assignment 2 for the course "High Performance Data Analytics" at the University of Luxembourg. The two visualizations are a scatterplot with a 2D brush interaction and a histogram, both of which are synchronized to highlight selected data subsets simultaneously. The implementation leverages modern JavaScript frameworks, React for modular UI components and D3.js for high-performance data manipulation, to create a seamless and interactive data exploration experience.

The primary objective is to facilitate data exploration, allowing users to interactively select subsets of data points and immediately observe their impact across different dimensions of the dataset. This report details the design decisions, implementation considerations, and a discussion of the pros and cons of the histogram visualization.

## 2 Input Data

The provided dataset contains 8,760 records of bike-sharing-related data from Seoul. It includes various attributes that can be analyzed to understand patterns in bike usage throughout different times and weather conditions.

### 2.1 Quantitative Attributes

The dataset includes the following quantitative attributes:

1. **Date:** The date on which the data was recorded.
2. **RentedBikeCount:** The number of bikes rented.
3. **Hour:** The hour of the day when data was collected.
4. **Temperature:** Temperature in degrees Celsius.
5. **Humidity:** Humidity percentage.
6. **WindSpeed:** Wind speed measured in meters per second.
7. **Visibility:** Visibility distance in meters.
8. **DewPointTemperature:** Dew point temperature in degrees Celsius.
9. **SolarRadiation:** Solar radiation in MJ/m<sup>2</sup>.
10. **Rainfall:** Amount of rainfall in millimeters.
11. **Snowfall:** Amount of snowfall in centimeters.

These attributes represent various quantitative aspects of bike rentals, weather conditions, and temporal data. The dataset allows us to explore how external conditions such as temperature, humidity, and rainfall affect bike-sharing usage across different times of the day.

### 3 System Implementation

The application comprises two synchronized visualizations:

- **Scatterplot with 2D brush interaction:** Allows users to select data points dynamically based on two attributes, such as "Temperature" versus "Humidity".
- **Histogram with 1D brush interaction:** Enables users to explore the frequency distribution of any chosen attribute, with selections linked to the scatterplot.

React components and Redux are used for state management, providing a modular approach to maintain consistency across the application. D3.js manages the visual rendering and brush interactions. Redux ensures synchronization between the scatterplot and histogram by sharing state updates, enabling real-time highlighting across both visualizations.

### 4 Scatterplot

The scatterplot visualization is designed to explore relationships between two selected attributes. Users can dynamically change the X and Y attributes from a dropdown menu, providing the flexibility to explore various correlations within the dataset, such as between "Temperature" and "RentedBikeCount".

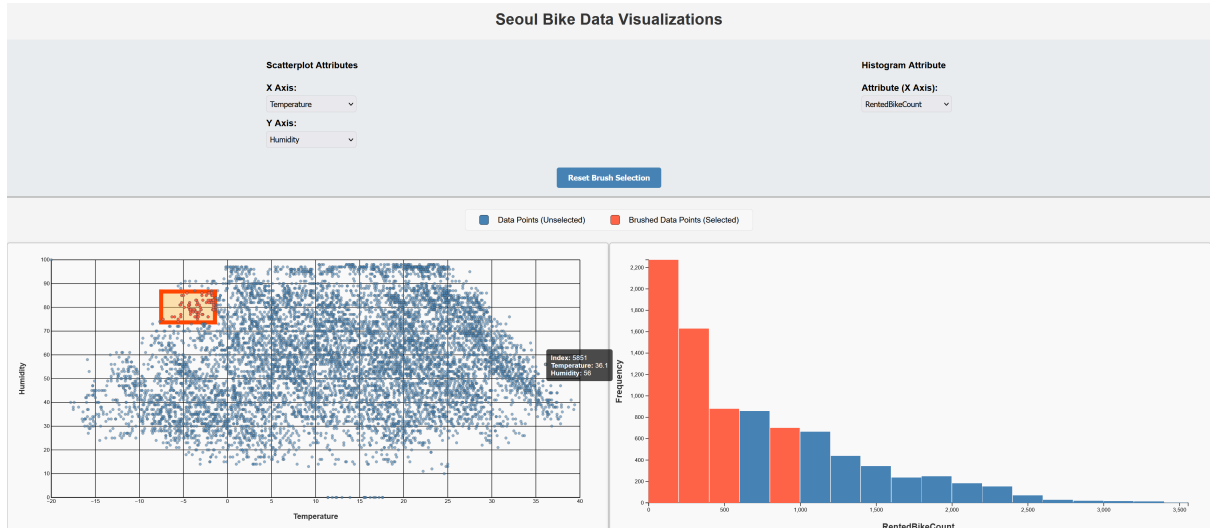


Figure 1: Scatterplot of Temperature vs. Humidity with a highlighted brushed selection.

Figure 1 shows an example of the scatterplot with temperature on the X-axis and humidity on the Y-axis. The scatterplot allows for intuitive exploration using D3's brush API to create a **2D brush interaction**, enabling the selection of data subsets. This provides users with a powerful way to dynamically filter the dataset and observe how subsets relate to other attributes in the histogram.

### 5 Histogram Design Justification

The histogram was selected as the second visualization due to its ability to effectively summarize data distributions. The histogram features a dynamic X-axis that allows users to explore the distribution of any chosen attribute, while the Y-axis remains fixed to represent frequency, ensuring consistency and ease of interpretation.

#### 5.1 Pros of the Histogram Visualization

**1. Effective Summarization of Data Distribution:** Histograms are well-suited for summarizing large datasets by grouping data points into discrete bins, thereby representing the distribution of an attribute in an aggregated format. In the context of the bike-sharing dataset, attributes such as "RentedBikeCount" are more comprehensible when visualized as a histogram (Figure 2). Users can easily

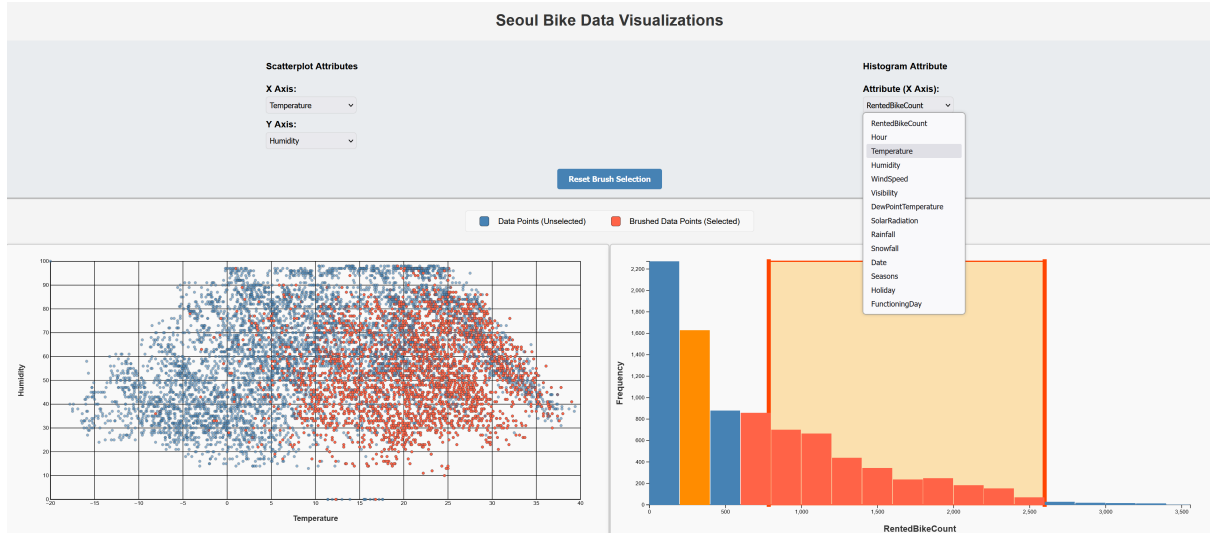


Figure 2: Histogram of RentedBikeCount showing the frequency distribution.

understand which ranges are most common, allowing for quick identification of central tendencies and outliers.

**2. Dynamic Exploration through Attribute Selection:** The ability to dynamically change the X-axis attribute provides flexibility for users to explore various aspects of the dataset. For example, selecting "Temperature" or "Humidity" as the X-axis reveals patterns in bike rentals based on weather conditions. This adaptability helps users develop hypotheses and find correlations more effectively.

**3. Consistency and Clarity with a Fixed Frequency Y-Axis:** Keeping the Y-axis fixed to represent frequency ensures consistency across different attributes. Users do not need to adjust their interpretation based on changing axes, reducing cognitive load. This is particularly useful for less experienced users who benefit from the predictable representation of data distributions, thereby making the visualization more accessible.

**4. Linked Interaction with Scatterplot:** The **1D brush interaction** on the histogram is synchronized with the scatterplot via Redux, enabling linked filtering. When users select a range on the histogram, the scatterplot automatically highlights the corresponding data points, thus providing a coordinated overview-detail analysis. This interaction fosters an integrated exploration of the dataset—users can select data ranges of interest and immediately view how they relate to other variables.

**5. Insight into Data Skews and Outliers:** Histograms allow users to quickly identify data skews and the presence of outliers. For instance, if bike rentals are disproportionately high on specific days or during particular temperature ranges, this trend becomes immediately visible. Such information is critical for decision-making and further analysis, as it helps uncover patterns that might indicate user behavior trends or data collection issues.

**6. Performance Efficiency for Large Datasets:** Given the large size of the dataset, histograms offer a performance-efficient way to visualize aggregated information without overwhelming users with individual data points. The data binning approach reduces visual clutter while retaining the overall distribution characteristics. This makes the histogram a robust choice for displaying high-level insights in a concise format.

## 5.2 Cons of the Histogram Visualization

While histograms have numerous advantages, there are certain drawbacks to consider:

**1. Limited Detail Compared to Scatterplots:** Unlike scatterplots, histograms do not show relationships between multiple attributes at the individual data point level. They are limited to depicting the frequency of one attribute, making it less effective for identifying correlations between attributes.

**2. Bin Size Sensitivity:** The representation in a histogram depends on the choice of bin size. Too many bins can result in a noisy and cluttered visualization, while too few bins may oversimplify the distribution. Careful selection of bin size is crucial to accurately represent the dataset while avoiding information loss.

Despite these drawbacks, the histogram provides an essential high-level summary of attribute distributions, making it highly complementary to the scatterplot in this application. The pros of flexibility, ease of interpretation, and linked brushing outweigh the limitations, particularly in the context of exploratory data analysis where both aggregated and relational insights are needed.

## 6 Conclusion

The interactive visualizations developed for this assignment offer a powerful approach to exploring the bike-sharing dataset. The scatterplot provides a detailed view of correlations between two attributes, while the histogram presents an aggregated perspective on the distribution of a chosen attribute. Together, they form an effective overview-detail system that supports both high-level exploration and deep-dive analysis of relationships in the data.

The use of React, D3.js, and Redux has resulted in a modular, scalable, and high-performance implementation. The linked brushing feature ensures that the user's interactions are meaningful across both visualizations, providing an intuitive and seamless exploration experience. The histogram, in particular, with its dynamic X-axis and fixed frequency Y-axis, is a thoughtfully chosen design that serves the purpose of summarizing and exploring large datasets effectively.

## 7 References

- Mike Bostock, D3.js - Data-Driven Documents. Available: <https://d3js.org/>
- React Documentation. Available: <https://reactjs.org/>
- Redux Documentation. Available: <https://redux.js.org/>