ADV EXP - 7

Aim - To create Visualizations using D3.js on a Finance Dataset.

Objectives:

- 1. To explore and visualize a dataset related to Finance/Banking/Insurance/Credit using D3.js.
- 2. To create basic visualizations (Bar chart, Pie chart, Histogram, Timeline chart, Scatter plot, Bubble plot) to understand data distribution and trends.
- 3. To create advanced visualizations (Word chart, Box and Whisker plot, Violin plot, Regression plot, 3D chart, Jitter) for deeper insights and complex relationships.
- 4. To perform hypothesis testing using the Pearson correlation coefficient to evaluate relationships between numerical variables in the dataset.

Implementation:

Barplot:

```
.attr("width", width + margin.left + margin.right)
            .attr("height", height + margin.top + margin.bottom)
            .append("g")
${margin.top})`);
Amount as an integer
            const debitData = data.filter(d => d["Transaction Type"]
=== "debit")
                                 .map(d => ({ ...d, Amount: +d.Amount
}));
            const categoryTotals = d3.rollups(
                v \Rightarrow d3.sum(v, d \Rightarrow d.Amount),
                d => d.Category
            );
select the top 5
            const top5Spends = categoryTotals.sort((a, b) => b[1] -
a[1]).slice(0, 5);
            const top5SpendsFormatted = top5Spends.map(([category,
amount]) => ({ category, amount }));
            console.log(top5SpendsFormatted);
            const x = d3.scaleLinear()
d.amount)])
                         .range([0, width]);
```

```
const y = d3.scaleBand()
                         .domain(top5SpendsFormatted.map(d =>
d.category))
                         .range([0, top5SpendsFormatted.length *
barHeight])
                        .padding(0.1);
            d3.select("svg").attr("height", top5SpendsFormatted.length
            svg.selectAll("rect")
               .data(top5SpendsFormatted)
               .enter()
               .append("rect")
               .attr("x", 0)
               .attr("y", d => y(d.category))
               .attr("width", d => x(d.amount))
               .attr("fill", "steelblue");
            svg.append("text")
                .attr("y", -margin.top / 2)
                .style("font-size", "16px")
                .style("font-weight", "bold")
                .text("Most Amount spent categories");
            svg.append("g")
               .attr("transform", `translate(0,
${top5SpendsFormatted.length * barHeight})`)
               .call(d3.axisBottom(x).ticks(5))
               .selectAll("text")
               .style("font-size", "12px");
            svg.append("g")
               .call(d3.axisLeft(y))
               .selectAll("text")
```

```
.style("font-size", "12px");

}).catch(error => {
        console.error("Error loading the CSV file:", error);
    });
   </script>

</body>
</html>
```

Credit Card Payment – Mortgage & Rent – Home Improvement – Groceries –

Most Amount spent categories

10,000 15,000 20,000 25,000 30,000

This bar chart shows that the highest spending category is "Credit Card Payment," followed by "Mortgage & Rent" and "Home Improvement."

Categories such as "Groceries" and "Utilities" account for significantly lower spending amounts compared to the top categories.

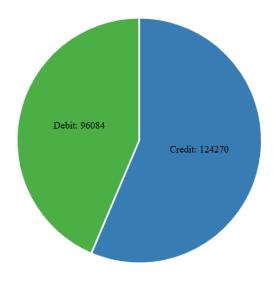
Pie Chart:

Utilities -

```
const debitTotal = Math.round(d3.sum(data.filter(d =>
d["Transaction Type"] === "debit"), d => +d.Amount));
            const creditTotal = Math.round(d3.sum(data.filter(d =>
d["Transaction Type"] === "credit"), d => +d.Amount));
                { label: "Debit", value: debitTotal },
            const width = 300, height = 500, radius = Math.min(width,
height) / 2;
           const svg = d3.select("svg")
                          .append("g")
${height / 2 }) `);
            const pie = d3.pie().value(d => d.value);
            const arc = d3.arc().innerRadius(0).outerRadius(radius);
            const color = d3.scaleOrdinal()
                            .domain(pieData.map(d => d.label))
                            .range(["#4daf4a", "#377eb8"]);
            svg.selectAll("path")
               .data(pie(pieData))
               .enter()
               .append("path")
               .attr("d", arc)
               .style("stroke-width", "2px");
```

```
svg.selectAll("text")
       .data(pie(pieData))
       .enter()
       .append("text")
       .text(d => `${d.data.label}: ${d.data.value}`)
       .attr("transform", d => `translate(${arc.centroid(d)})`)
       .style("text-anchor", "middle")
       .style("font-size", "12px");
    svg.append("text")
        .attr("x", -1)
        .attr("y", -200)
        .style("font-size", "16px")
        .style("font-weight", "bold")
        .text("Credit vs Debit");
}).catch(error => {
});
```

Credit vs Debit



This pie chart illustrates the comparison between credit and debit amounts, with credit transactions totaling 124,270 and debit transactions totaling 96,084. Credit transactions make up a larger portion, indicating higher inflow compared to outflow.

Timeline:

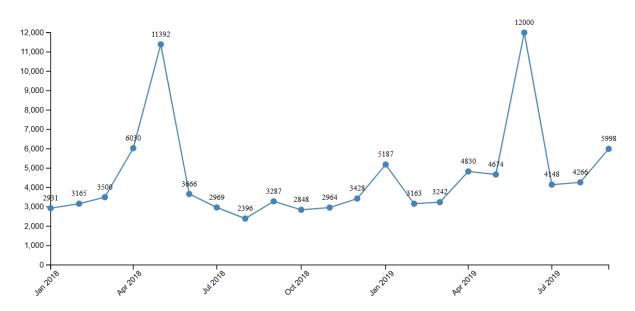
```
!DOCTYPE html>
Chtml lang="en">
   <meta charset="UTF-8">
   <title>Timeline Chart</title>
   <svg width="800" height="400"></svg>
Amount as an integer
            const debitData = data.filter(d => d["Transaction Type"]
=== "debit")
                                .map(d => ({ ...d, Amount: +d.Amount
}));
            debitData.forEach(d => {
                d.Date = d3.timeParse("%m-%d-%Y")(d.Date);
                d.Amount = +d.Amount;
            });
month
                d3.group(debitData, d =>
d3.timeFormat("%Y-%m")(d.Date)),
                ([key, values]) => ({
                    Date: d3.timeParse("%Y-%m")(key), // Convert key
back to date
                    Amount: Math.round(d3.sum(values, d => d.Amount))
            );
```

```
const margin = { top: 100, right: 30, bottom: 100, left: 50
                  width = 800 - margin.left - margin.right,
                  height = 500 - margin.top - margin.bottom;
            const svg = d3.select("svg")
                          .attr("width", width + margin.left +
margin.right)
margin.bottom)
                          .append("g")
translate(${margin.left},${margin.top})`);
            const xScale = d3.scaleTime()
                             .domain(d3.extent(monthlyData, d =>
d.Date))
                             .range([0, width]);
d.Amount)])
                             .range([height, 0]);
            svg.append("g")
               .attr("transform", `translate(0, ${height})`)
               .call(d3.axisBottom(xScale).tickFormat(d3.timeFormat("%b
               .selectAll("text")
               .attr("transform", "rotate(-45)")
               .style("text-anchor", "end");
            svg.append("g")
               .call(d3.axisLeft(yScale));
            const line = d3.line()
                           .x(d => xScale(d.Date))
```

```
.y(d => yScale(d.Amount));
    svg.append("path")
       .attr("fill", "none")
       .attr("stroke-width", 1.5)
    svg.selectAll("circle")
       .data(monthlyData)
       .enter()
       .append("circle")
       .attr("cy", d => yScale(d.Amount))
    svg.selectAll("text.label")
       .data(monthlyData)
       .enter()
       .append("text")
       .attr("x", d => xScale(d.Date))
       .attr("y", d => yScale(d.Amount) - 10)
       .attr("text-anchor", "middle")
       .text(d => d.Amount);
    svg.append("text")
        .attr("x", width/2)
        .attr("text-anchor", "middle")
        .style("font-size", "16px")
        .style("font-weight", "bold")
}).catch(error => {
});
```

```
</script>
</body>
</html>
```

Expenses per month



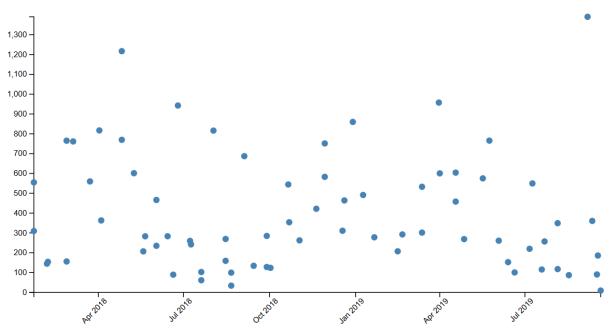
This timeline chart shows monthly expenses over time, with significant peaks in April 2018 and May 2019, where spending exceeded 10,000. The pattern suggests periodic spikes in expenditure, while most other months maintain a relatively stable range around 3,000 to 5,000.

Scatter Plot:

```
data.forEach(d => {
                d.Date = d3.timeParse("%m-%d-%Y") (d.Date);
                d.Amount = +d.Amount;
            const creditData = data.filter(d => d["Transaction Type"]
=== "credit" & d['Amount']<1500);
            const margin = { top: 100, right: 30, bottom: 50, left: 50
                  width = 800 - margin.left - margin.right,
            const svg = d3.select("svg")
margin.right)
margin.bottom)
                          .append("g")
                          .attr("transform",
 translate(${margin.left},${margin.top})`);
            const xScale = d3.scaleTime()
                             .domain(d3.extent(creditData, d =>
d.Date))
                             .range([0, width]);
            const yScale = d3.scaleLinear()
                             .domain([0, d3.max(creditData, d =>
d.Amount)])
                             .range([height, 0]);
            svg.append("g")
               .call(d3.axisBottom(xScale).tickFormat(d3.timeFormat("%b
               .selectAll("text")
```

```
.style("text-anchor", "end");
    svg.append("g")
       .call(d3.axisLeft(yScale));
    svg.selectAll("circle")
       .data(creditData)
       .enter()
       .append("circle")
       .attr("cx", d => xScale(d.Date))
       .attr("cy", d => yScale(d.Amount))
    svg.append("text")
        .attr("x", width/2)
        .attr("y", -50)
        .attr("text-anchor", "middle")
        .style("font-size", "16px")
        .style("font-weight", "bold")
        .text("Credits overtime");
});
```

Credits overtime



The scatter plot shows no noticeable upward or downward trend in credit amounts over time, indicating a relatively stable pattern. Credit transactions are mostly scattered below the 1,000 mark, with a few higher outliers over the timeline. The distribution of credits suggests sporadic, high-value transactions rather than consistent monthly increases or decreases.

Hypothesis Testing:

```
import pandas as pd
from scipy.stats import pearsonr

# Load dataset
data = pd.read_csv('personal_transactions.csv')
data.head()

# Get the required data
reqdata = []
for index, rec in data.iterrows():
    if rec['Transaction Type'] == 'credit':
        reqdata.append(rec)

# Get date and covert it to numeric
date = []
for rec in reqdata:
    date.append(rec['Date'])
```

```
date numeric = pd.to datetime(date, format='%m-%d-%Y').astype('int64')
print(date_numeric)
Index([1514937600.0, 1515715200.0, 1516320000.0, 1516579200.0, 1516579200.0,
       1517529600.0, 1517788800.0, 1517875200.0, 1518739200.0, 1519603200.0,
       1565913600.0, 1566000000.0, 1567123200.0, 1567728000.0, 1568160000.0,
       1568332800.0, 1568592000.0, 1568678400.0, 1568937600.0, 1569542400.0],
      dtype='float64', length=118)
credit = []
for rec in reqdata:
    credit.append(round(rec['Amount']))
print(credit)
[2298, 2000, 2000, 555, 310, 2000, 145, 154, 2000, 765, 156, 2000, 762, 2000,
corr, p value = pearsonr(date numeric, credit)
print(f"Pearson Correlation Coefficient: {corr}")
print(f"P-Value: {p value}")
 Pearson Correlation Coefficient: -0.014825326904847773
 P-Value: 0.8734020555859093
```

Pearson Correlation Coefficient (-0.0148): This value is very close to zero, which suggests that there is no linear relationship between the date and the credit amount. **P-Value** (0.8734): A high p-value (typically above 0.05) means that the result is not statistically significant.