Visual Data Analysis of Fraudulent Transactions

Your CFO has also requested detailed trends data on specific card holders. Use the starter notebook to query your database and generate visualizations that supply the requested information as follows, then add your visualizations and observations to your markdown report.

```
In []:
    # Initial imports
    import pandas as pd
    import calendar
    import hvplot.pandas
    from sqlalchemy import create_engine
    %matplotlib inline
```

```
In [ ]:
         # Create a connection to the database
         def postgres connect str(
             uname = 'Daniel',
             pass env = 'POSTGRES PASSWORD',
             host = 'pg-2e8191e-instructors-1f45.aivencloud.com',
             database = 'daniel',
             port
                    = 18645,
         ):
             from dotenv import load dotenv
             import os
             load dotenv()
             password = os.getenv(pass env)
             return f"postgresql://{uname}:{password}@{host}:{port}/{database}?sslmode=require"
         postgres connect str()
         engine = create engine(postgres connect str())
```

Data Analysis Question 1

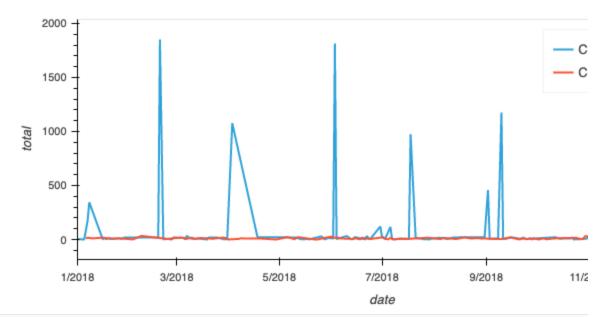
The two most important customers of the firm may have been hacked. Verify if there are any fraudulent transactions in their history. For privacy reasons, you only know that their cardholder IDs are 2 and 18.

• Using hvPlot, create a line plot representing the time series of transactions over the course of the year for each cardholder separately.

- Next, to better compare their patterns, create a single line plot that containins both card holders' trend data.
- What difference do you observe between the consumption patterns? Does the difference suggest a fraudulent transaction? Explain your rationale in the markdown report.

```
In [ ]:
         # Write function that locates outliers using standard deviation
         sql=f'''SELECT card holder id, date trunc('day',date) as date, SUM(amount) as total FROM "Transactions" t
                 INNER JOIN "Credit card" c
                 ON t."card number" = c."cardNumber"
                 WHERE card holder id = 2 or card holder id = 18
                  GROUP BY date_trunc('day',date), card_holder_id
                 ORDER BY date trunc('day',date)
         1.1.1
         df=pd.read sql query(sql, engine)
         df.index = pd.to datetime(df["date"])
         plot1 = df[
                 df["card holder id"] == 18 # where card holder id is 18
         [["total", "date"]].hvplot(label="Customer id 18")
         plot2 = df[
                 df["card holder id"] == 2 # where card holder id is 18
         [["total", "date"]].hvplot(label="Customer id 2") # select all but card_holder_id
         plot1 * plot2
```

Out[]:



```
In []:
    std2= df['total'].std()*2
    len(df[df['total'] >= std2])

Out[]:
    7
```

Data Analysis Question 2

The CEO of the biggest customer of the firm suspects that someone has used her corporate credit card without authorization in the first quarter of 2018 to pay quite expensive restaurant bills. Again, for privacy reasons, you know only that the cardholder ID in question is 25.

- Using Plotly Express, create a box plot, representing the expenditure data from January 2018 to June 2018 for cardholder ID 25.
- Are there any outliers for cardholder ID 25? How many outliers are there per month?
- Do you notice any anomalies? Describe your observations and conclusions in your markdown report.

Out[]:



```
200 200 2/2018 3/2018 4/2018 5/2018 date
```

```
In []:
    df.describe()
    std2up = df.std()*2 + df.mean()
    std2low= df.std()*2 - df.mean()
```

/Users/dan/opt/anaconda3/envs/dev3/lib/python3.7/site-packages/ipykernel_launcher.py:3: FutureWarning: DataFrame.mean and DataFrame.median with numeric_only=None will include datetime64 and datetime64tz columns in a future version.

This is separate from the ipykernel package so we can avoid doing imports until
/Users/dan/opt/anaconda3/envs/dev3/lib/python3.7/site-packages/ipykernel_launcher.py:4: FutureWarning: DataFrame.mean and DataFrame.median with numeric only=None will include datetime64 and datetime64tz columns in a future version.

```
In [ ]:
    df.groupby([df.index.month, df.index.day]).sum()
```

Out[]: sum

after removing the cwd from sys.path.

date	date	
1	2	1.46
	5	10.74
	7	2.93
	10	1.39
	14	17.84
	16	1.65
	18	15.86
	21	2.22
	30	1177.00
	31	2.75
2	2	10.75
	5	10.81

		sum
date	date	
	7	5.97
	12	3.69
	18	16.70
	23	14.90
	28	2.09
3	2	12.42
	5	16.58
	6	1334.00
	7	2.88
	9	2.04
	11	13.57
	12	14.83
	16	4.20
	17	2.56
	18	18.28
	31	21.04
4	1	102.62
	2	24.23
	8	1083.21
	9	269.00
	10	10.24
	18	7.39
	19	6.01
	20	20.03
	26	28.47
	29	16.50
5	6	1.10
	13	1046.00

date date

17 12.15

19 2.27

29

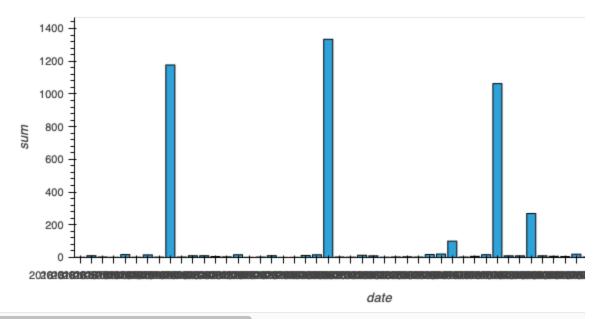
5.97

```
In []:
    monthly_df = df.copy()
    #monthly_df.rename(columns={"sum":"monthlySum"}, inplace=True)
    monthly_df.groupby([monthly_df.index.month, monthly_df.index.day]).sum()

    monthly_df["Month"] = monthly_df["date"].apply(lambda x: calendar.month_name[x.month])
    monthly_df["Day"] = monthly_df["date"].apply(lambda x: x.day)

    monthly_df[["sum", "Month", "Day"]].drop(columns=["Day"]).hvplot.bar()
```

Out[]:



```
In []: # loop to change the numeric month to month names

In []: # Creating the six box plots using plotly express
```