

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats

df = pd.read_csv("/content/drive/MyDrive/Metro_Interstate_Traffic_Volume.csv", parse_dates=["date_time"])

print("Rows, Cols:", df.shape)
display(df.head())
print(df.info())
display(df.describe(include='all'))

Rows, Cols: (48204, 9)
   holiday    temp  rain_1h  snow_1h  clouds_all weather_main weather_description      date_time  traffic_volume
0     NaN  288.28       0.0       0.0        40      Clouds  scattered clouds 2012-10-02 09:00:00       5545
1     NaN  289.36       0.0       0.0        75      Clouds  broken clouds 2012-10-02 10:00:00       4516
2     NaN  289.58       0.0       0.0        90      Clouds  overcast clouds 2012-10-02 11:00:00       4767
3     NaN  290.13       0.0       0.0        90      Clouds  overcast clouds 2012-10-02 12:00:00       5026
4     NaN  291.14       0.0       0.0        75      Clouds  broken clouds 2012-10-02 13:00:00       4918
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48204 entries, 0 to 48203
Data columns (total 9 columns):
 #   Column          Non-Null Count  Dtype    
--- 
 0   holiday         61 non-null    object    
 1   temp            48204 non-null  float64  
 2   rain_1h         48204 non-null  float64  
 3   snow_1h         48204 non-null  float64  
 4   clouds_all      48204 non-null  int64    
 5   weather_main    48204 non-null  object    
 6   weather_description 48204 non-null  object    
 7   date_time       48204 non-null  datetime64[ns]
 8   traffic_volume  48204 non-null  int64    
dtypes: datetime64[ns](1), float64(3), int64(2), object(3)
memory usage: 3.3+ MB
None
   holiday    temp  rain_1h  snow_1h  clouds_all weather_main weather_description      date_time  traffic_volume
count      61  48204.000000  48204.000000  48204.000000      48204  48204.000000      48204  48204.000000
unique       11      NaN      NaN      NaN      NaN      11      38      NaN      NaN
top    Labor Day      NaN      NaN      NaN      NaN      15164     11665      NaN      NaN
freq        7      NaN      NaN      NaN      NaN      15164     11665      NaN      NaN
mean    281.205870  0.334264  0.000222  49.362231      NaN      NaN  2016-01-05 10:46:16.773711616  3259.818355
min     0.000000  0.000000  0.000000  0.000000      NaN      NaN  2012-10-02 09:00:00  0.000000
25%    272.160000  0.000000  0.000000  1.000000      NaN      NaN  2014-02-06 11:45:00  1193.000000
50%    282.450000  0.000000  0.000000  64.000000      NaN      NaN  2016-06-11 03:30:00  3380.000000
75%    291.806000  0.000000  0.000000  90.000000      NaN      NaN  2017-08-11 06:00:00  4933.000000
max    310.070000  9831.300000  0.510000  100.000000      NaN      NaN  2018-09-30 23:00:00  7280.000000
std     13.38232  44.789133  0.008168  39.015750      NaN      NaN  1986.860670

```

Non-Graphical EDA

```

tv = df["traffic_volume"].dropna()

mean_tv = tv.mean()
median_tv = tv.median()
mode_tv = tv.mode().iloc[0] if not tv.mode().empty else np.nan
std_tv = tv.std()
min_tv = tv.min()
max_tv = tv.max()
range_tv = max_tv - min_tv
skew_tv = tv.skew()
kurt_tv = tv.kurtosis()

print(f"Mean: {mean_tv:.2f}")
print(f"Median: {median_tv:.2f}")
print(f"Mode: {mode_tv}")
print(f"Std: {std_tv:.2f}")
print(f"Range: {range_tv:.2f}")
print(f"Skewness: {skew_tv:.3f}")
print(f"Kurtosis (excess): {kurt_tv:.3f}")

Mean: 3259.82
Median: 3380.00
Mode: 357
Std: 1986.86
Range: 7280
Skewness: -0.089
Kurtosis (excess): -1.309

```

[إنما كان الوسيط أكبر من المتوسط هذا يشير إلى توزيع متحرف لليسار.

الفرق بين الوسيط والمتوسط يدل على وجود قيم متطرفة تؤثر على المتوسط.

```

group_weather = df.groupby("weather_main")["traffic_volume"].agg(["mean","std","count"]).sort_values("mean", ascending=False)
display(group_weather)

```

	mean	std	count
weather_main			
Clouds	3618.449749	1906.202879	15164
Haze	3502.101471	1873.511057	1360
Rain	3317.905501	1982.228705	5672
Drizzle	3290.727073	1997.901830	1821
Smoke	3237.650000	1978.015893	20
Clear	3055.908819	1987.101411	13391
Snow	3016.844226	1900.185883	2876
Thunderstorm	3001.620890	1988.296012	1034
Mist	2932.956639	2073.025670	5950
Fog	2703.720395	2125.525863	912
Scall	2061.750000	1950.070490	4

Task C: Multivariate Analysis via Grouping

The weather condition associated with the highest average traffic is Clouds.

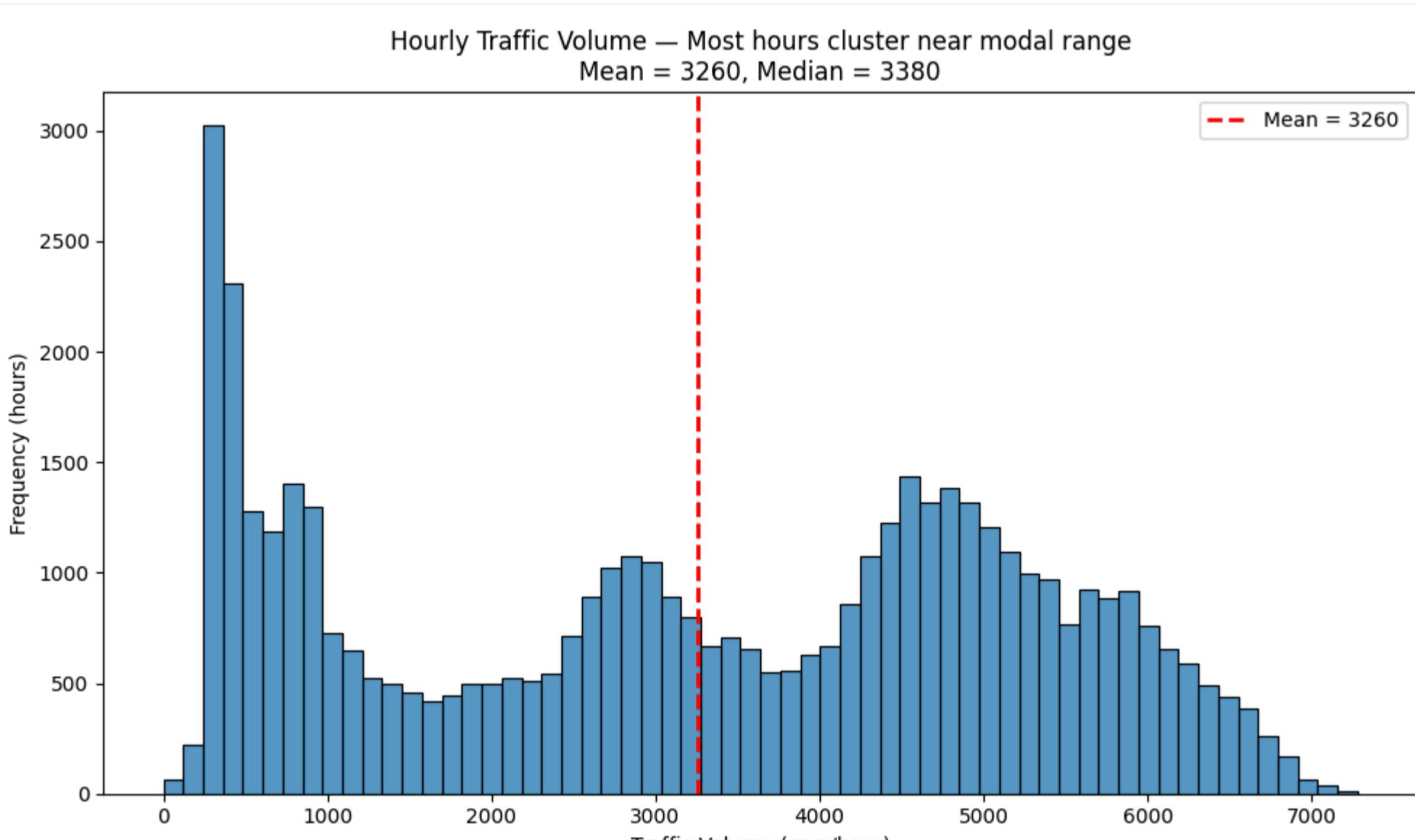
The road condition with the highest variance is Thunderstorm.

Graphical EDA

```

plt.figure(figsize=(10,6))
sns.histplot(tv, bins=60, kde=False)
plt.title("Hourly Traffic Volume — Most hours cluster near modal range\nMean = 3260, Median = 3380")
plt.xlabel("Hourly Traffic Volume (cars/hour)")
plt.ylabel("Frequency (hours)")
plt.legend()
plt.tight_layout()
plt.show()

```



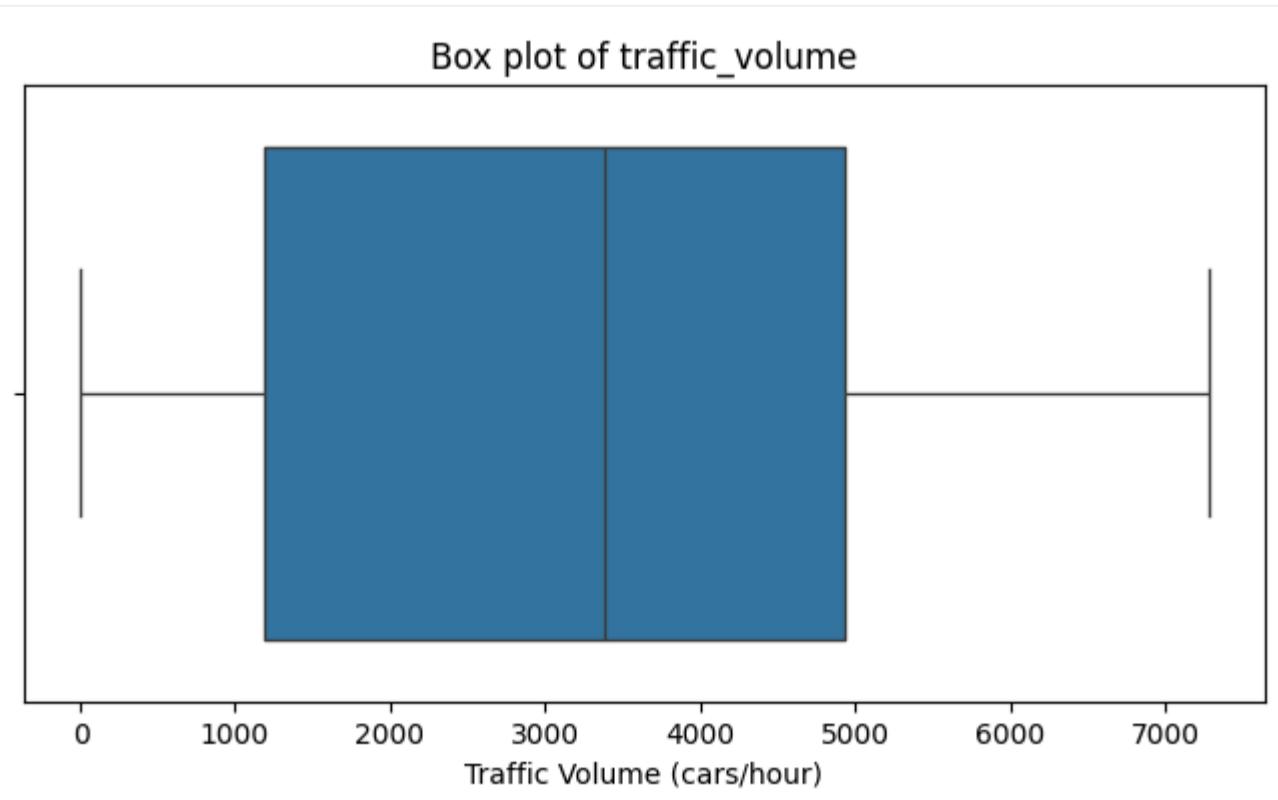
B. Box plot and calculating IQR

```

plt.figure(figsize=(8,4))
sns.boxplot(x=tv)
plt.title("Box plot of traffic_volume")
plt.xlabel("Traffic Volume (cars/hour)")
plt.show()

# calculate IQR
Q1 = tv.quantile(0.25)
Q3 = tv.quantile(0.75)
IQR = Q3 - Q1
print(f"Q1={Q1}, Q3={Q3}, IQR={IQR}")

```



.Interquartile range (IQR) measures the variation in the middle of the distribution (the middle 50% of the data)

.The IQR is smaller than the overall range, indicating that outliers increase the overall variance.

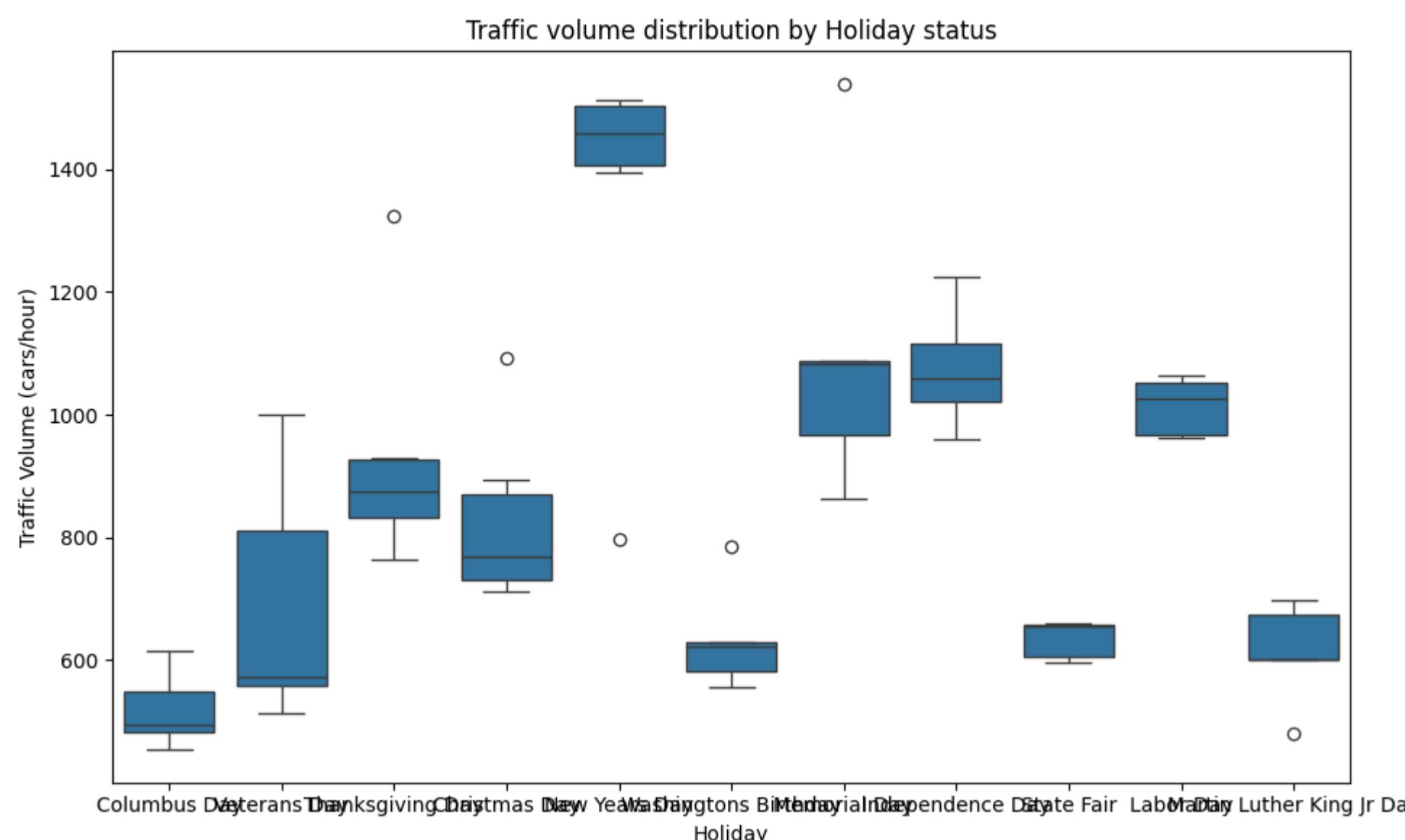
C. Grouped box plots by holiday (Comparison of holidays vs regular days)

```

plt.figure(figsize=(10,6))
sns.boxplot(x="holiday", y="traffic_volume", data=df)
plt.title("Traffic volume distribution by Holiday Status")

```

```
plt.xlabel("Holiday")
plt.ylabel("Traffic Volume (cars/hour)")
plt.tight_layout()
plt.show()
```

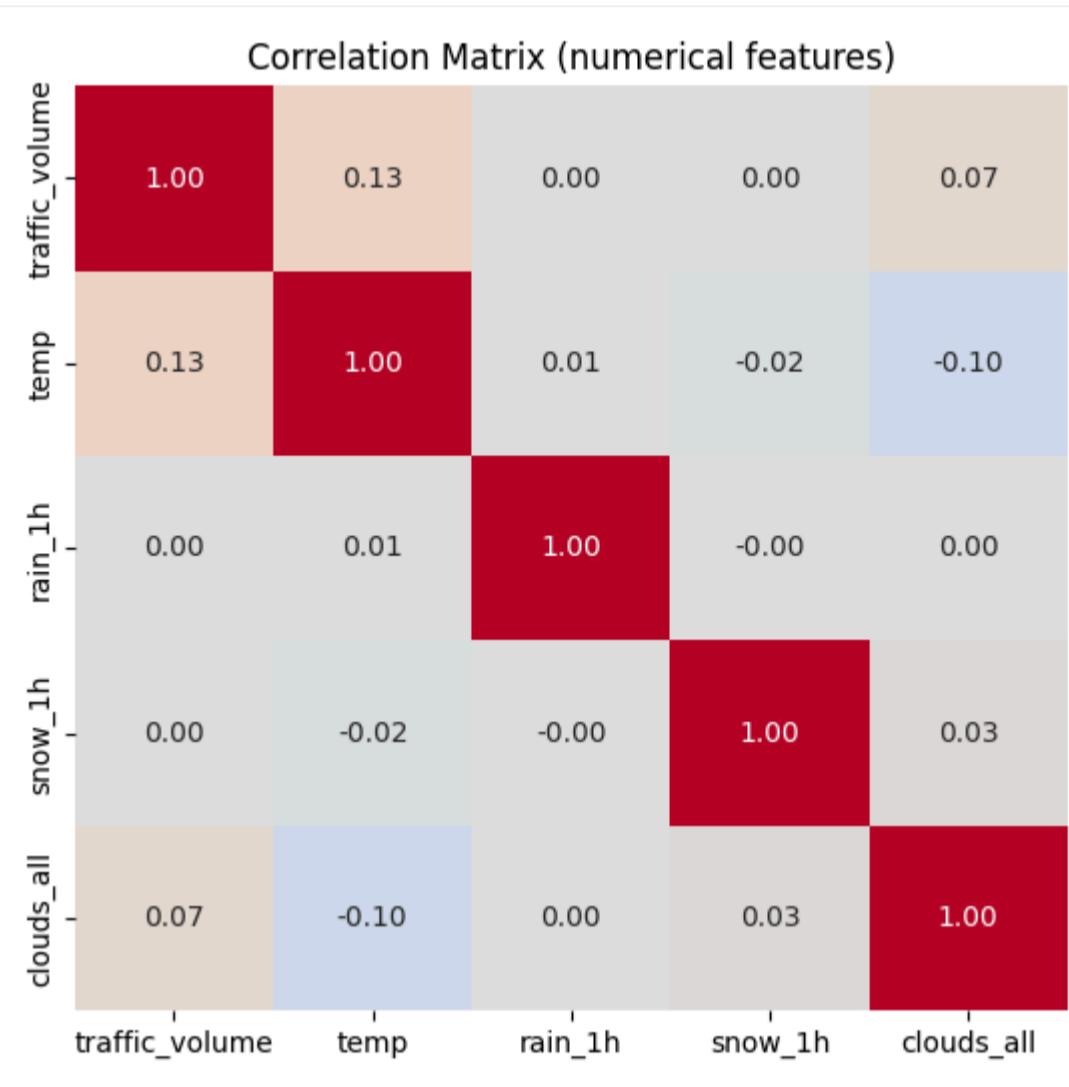
**Note:**

Yes, there is a big difference in distribution between regular days and holidays.

Ordinary days have a higher mean and lower variance compared to holidays

Correlation Matrix and Heatmap

```
num_cols = ["traffic_volume", "temp", "rain_1h", "snow_1h", "clouds_all"]
corr = df[num_cols].corr()
plt.figure(figsize=(8,6))
sns.heatmap(corr, annot=True, fmt=".2f", cmap="coolwarm", center=0)
plt.title("Correlation Matrix (numerical features)")
plt.show()
```

**Task A: Correlation Matrix and Heatmap**

1_In the summer (high heat), people may reduce driving during the hot peak hours.

2_The causality paradox:

.People are on vacation and may travel out of town .Using alternative means of transportation: In good weather, people may prefer walking or cycling.

.Peak hours: traffic is usually higher in the morning and evening when the temperature is moderate.

```
import plotly.express as px
fig = px.scatter(df.sample(5000, random_state=1), x="temp", y="traffic_volume",
                  hover_data=["weather_description", "date_time", "holiday"],
                  title="Traffic Volume vs Temperature (sampled points)")
fig.update_layout(xaxis_title="Temp (Kelvin)", yaxis_title="Traffic Volume (cars/hour)")
fig.show()
```

Traffic Volume vs Temperature (sampled points)

