

Correlation is a statistical measure that describes the extent to which two variables change together. If two variables tend to increase or decrease in tandem, they are said to be correlated. Correlation is widely used in various fields such as finance, economics, medicine, and social sciences to identify and quantify the strength and direction of relationships between variables.

Key Concepts of Correlation

Direction of Correlation:

Positive Correlation: When both variables move in the same direction. As one variable increases, the other also increases, and as one decreases, the other also decreases. For example, height and weight typically have a positive correlation.

Negative Correlation: When variables move in opposite directions. As one variable increases, the other decreases, and vice versa. For example, the amount of exercise and body weight might have a negative correlation.

No Correlation: When there is no consistent relationship between the variables. Changes in one variable do not predict changes in the other.

Strength of Correlation:

The strength of a correlation is measured by a correlation coefficient, which ranges from -1 to 1.

- 1 indicates a perfect positive correlation.
- -1 indicates a perfect negative correlation.
- 0 indicates no correlation.

Types of Correlation Coefficients

1. Pearson Correlation Coefficient (r):
 - Measures the linear relationship between two continuous variables.
 - Calculated as:

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n \sum x^2 - (\sum x)^2][n \sum y^2 - (\sum y)^2]}}$$

- Assumes that the data is normally distributed and the relationship between the variables is linear.

2. Spearman's Rank Correlation Coefficient (ρ or r_s):
 - Measures the strength and direction of the monotonic relationship between two ranked variables.
 - Calculated based on the ranks of the data rather than the actual data values.
 - Does not require the assumption of normality and can be used for ordinal data.
3. Kendall's Tau (τ):
 - Another measure of rank correlation.
 - Evaluates the ordinal association between two measured quantities.
 - Based on the number of concordant and discordant pairs of observations.

Interpretation of Correlation Coefficients

- to ± 0.1 : Negligible correlation.
- ± 0.1 to ± 0.3 : Weak correlation.
- ± 0.3 to ± 0.5 : Moderate correlation.
- ± 0.5 to ± 0.7 : Strong correlation.
- ± 0.7 to ± 1.0 : Very strong correlation.

In cybersecurity, correlation methods can be used to detect anomalous behavior in network traffic, which may indicate potential security threats such as intrusion attempts, malware activities, or data breaches. By analyzing the relationships between different features of network traffic, we can identify patterns that are indicative of malicious activity.

We'll create a synthetic dataset that simulates normal and anomalous network traffic. Then, we'll use a correlation-based approach to detect anomalies.

No.	Time	Source	Destination	Protocol	Length	Info
1	0.24046800	2409:40f4:	UDP	195	443 > 54762 Len=133	
2	2.24E-07	2404:6800	2409:40f4:	UDP	112	443 > 54762 Len=50
3	0.000274	2409:40f4:	2404:6800	UDP	107	54762 > 443 Len=45
4	0.000356	2409:40f4:	2404:6800	UDP	102	54762 > 443 Len=40
5	0.058865	2404:6800	2409:40f4:	UDP	87	443 > 54762 Len=25
6	0.20445	192.168.2.224	0.0.25	MDNS	103	Standard query 0x0012 PTR _CASE8412_sub._googlecast._tcp.local, "QM" question PTR _googlecast._tcp.local, "QM" question
7	1.081791	2409:40f4:	64:ff9b::2	TLSv1.2	229	Application Data
8	1.081872	2409:40f4:	2404:6800	TLSv1.2	158	Application Data
9	1.081916	2409:40f4:	64:ff9b::2	TLSv1.2	444	Application Data
10	1.083393	2409:40f4:	64:ff9b::2	TLSv1.2	125	Application Data
11	1.083409	2409:40f4:	2404:6800	TLSv1.2	125	Application Data
12	1.083413	2409:40f4:	64:ff9b::2	TLSv1.2	125	Application Data
13	1.171444	64:ff9b::2	2409:40f4:	TCP	86	443 > 60214 [ACK] Seq=1 Ack=144 Win=945 Len=0 TSval=784177837 TSecr=2706552420
14	1.211287	2404:6800	2409:40f4:	TCP	86	443 > 39058 [ACK] Seq=1 Ack=73 Win=268 Len=0 TSval=3699953245 TSecr=3475464165
15	1.302936	2404:6800	2409:40f4:	TLSv1.2	181	Application Data
16	1.302936	2404:6800	2409:40f4:	TLSv1.2	1294	Application Data
17	1.302947	64:ff9b::2	2409:40f4:	TCP	86	443 > 54476 [ACK] Seq=1 Ack=359 Win=289 Len=0 TSval=429568206 TSecr=2529154042
18	1.302947	64:ff9b::2	2409:40f4:	TCP	86	443 > 54476 [ACK] Seq=1 Ack=398 Win=289 Len=0 TSval=429568211 TSecr=2529154044
19	1.302947	64:ff9b::2	2409:40f4:	TLSv1.2	125	Application Data
20	1.30312	2409:40f4:	2404:6800	TCP	86	39058 > 443 [ACK] Seq=112 Ack=1304 Win=502 Len=0 TSval=3475464386 TSecr=3699953314
21	1.303912	2404:6800	2409:40f4:	TLSv1.2	425	Application Data
22	1.303912	2404:6800	2409:40f4:	TLSv1.2	117	Application Data
23	1.303912	2404:6800	2409:40f4:	TLSv1.2	125	Application Data
24	1.304559	2409:40f4:	2404:6800	TCP	86	39058 > 443 [ACK] Seq=112 Ack=1713 Win=502 Len=0 TSval=3475464388 TSecr=3699953316
25	1.305193	2409:40f4:	2404:6800	TLSv1.2	121	Application Data
26	1.305221	2409:40f4:	2404:6800	TLSv1.2	125	Application Data
27	1.306045	2404:6800	2409:40f4:	TCP	86	443 > 39058 [ACK] Seq=1713 Ack=112 Win=268 Len=0 TSval=3699953340 TSecr=3475464167
28	1.306046	2404:6800	2409:40f4:	TLSv1.2	125	Application Data
29	1.324154	64:ff9b::2	2409:40f4:	TCP	86	443 > 60214 [ACK] Seq=1 Ack=183 Win=945 Len=0 TSval=784177993 TSecr=2706552422
30	1.326105	64:ff9b::2	2409:40f4:	TLSv1.2	125	Application Data
31	1.346404	2409:40f4:	64:ff9b::2	TCP	86	54476 > 443 [ACK] Seq=398 Ack=40 Win=502 Len=0 TSval=2529154307 TSecr=429568211
32	1.346431	2409:40f4:	2404:6800	TCP	86	39058 > 443 [ACK] Seq=186 Ack=1752 Win=502 Len=0 TSval=3475464430 TSecr=3699953340
33	1.366494	2409:40f4:	64:ff9b::2	TCP	86	60214 > 443 [ACK] Seq=183 Ack=40 Win=502 Len=0 TSval=2706552705 TSecr=784177993
34	1.376112	2404:6800	2409:40f4:	TCP	86	443 > 39058 [ACK] Seq=1752 Ack=186 Win=268 Len=0 TSval=3699953410 TSecr=3475464389
35	1.451687	64:ff9b::2	2409:40f4:	TLSv1.2	148	Application Data
36	1.451715	2409:40f4:	64:ff9b::2	TCP	86	60214 > 443 [ACK] Seq=183 Ack=102 Win=502 Len=0 TSval=2706552790 TSecr=784178120

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

file_path = 'late.csv' # Replace with your actual file path

df = pd.read_csv(file_path)

print(df.head())

df_numeric = df[['Time', 'Length']]
```

```
print(df_numeric.head())
```

```
correlation_matrix = df_numeric.corr(method='pearson')
```

```
print("Correlation Matrix:\n", correlation_matrix)
```

```
plt.figure(figsize=(8, 6))
```

```
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', vmin=-1, vmax=1)
```

```
plt.title('Correlation Heatmap of Network Traffic Features')
```

```
plt.show()
```