

```
import numpy as np
import pandas as pd

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px

import warnings
warnings.filterwarnings("ignore")
```

## Loading the Data

```
df = pd.read_csv('cybersecurity_attacks.csv')
df
```

	Timestamp	Source IP Address	Destination IP Address	\
0	2023-05-30 06:33:58	103.216.15.12	84.9.164.252	
1	2020-08-26 07:08:30	78.199.217.198	66.191.137.154	
2	2022-11-13 08:23:25	63.79.210.48	198.219.82.17	
3	2023-07-02 10:38:46	163.42.196.10	101.228.192.255	
4	2023-07-16 13:11:07	71.166.185.76	189.243.174.238	
...	...	...	...	
39995	2023-05-26 14:08:42	26.36.109.26	121.100.75.240	
39996	2023-03-27 00:38:27	17.21.163.81	196.108.134.78	
39997	2022-03-31 01:45:49	162.35.217.57	98.107.0.15	
39998	2023-09-22 18:32:38	208.72.233.205	173.79.112.252	
39999	2023-10-10 11:59:52	14.102.21.108	109.198.45.7	

	Source Port	Destination Port	Protocol	Packet Length	Packet
Type \					
0	31225	17616	ICMP	503	
Data					
1	17245	48166	ICMP	1174	
Data					
2	16811	53600	UDP	306	
Control					
3	20018	32534	UDP	385	
Data					
4	6131	26646	TCP	1462	
Data					
...	...	...	...	...	.
...					
39995	31005	6764	UDP	1428	
Control					
39996	2553	28091	UDP	1184	

Control				
39997	22505	25152	UDP	1043
Data				
39998	20013	2703	UDP	483
Data				
39999	50137	55575	ICMP	1175
Control				
Traffic Type		Payload Data		
...	\			
0	HTTP	Qui natus odio asperiores nam. Optio nobis ius...		
...				
1	HTTP	Aperiam quos modi officiis veritatis rem. Omni...		
...				
2	HTTP	Perferendis sapiente vitae soluta. Hic delectu...		
...				
3	HTTP	Totam maxime beatae expedita explicabo porro l...		
...				
4	DNS	Odit nesciunt dolorem nisi iste iusto. Animi v...		
...				
...	...	...		
...				
39995	HTTP	Quibusdam ullam consequatur consequuntur accus...		
...				
39996	HTTP	Quaerat neque esse. Animi expedita natus commo...		
...				
39997	DNS	Enim at aspernatur illum. Saepe numquam eligen...		
...				
39998	FTP	Officiis dolorem sed harum provident earum dis...		
...				
39999	HTTP	Eligendi omnis voluptate nihil voluptatibus do...		
...				
Action Taken		Severity Level	User Information \	
0	Logged	Low	Reyansh Dugal	
1	Blocked	Low	Sumer Rana	
2	Ignored	Low	Himmat Karpe	
3	Blocked	Medium	Fateh Kibe	
4	Blocked	Low	Dhanush Chad	
...	...	...	...	
39995	Logged	Medium	Adira Madan	
39996	Logged	High	Rati Dara	
39997	Blocked	Low	Samiha Joshi	
39998	Ignored	Low	Rasha Chauhan	
39999	Logged	Medium	Zaina Kumar	
Device Information Network				
Segment \				
0	Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...			Segment
A				

1	Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...	Segment
B		
2	Mozilla/5.0 (compatible; MSIE 9.0; Windows NT ...	Segment
C		
3	Mozilla/5.0 (Macintosh; PPC Mac OS X 10_11_5; ...	Segment
B		
4	Mozilla/5.0 (compatible; MSIE 5.0; Windows NT ...	Segment
C		
...	...	.
..		
39995	Mozilla/5.0 (iPad; CPU iPad OS 14_2_1 like Mac...	Segment
A		
39996	Mozilla/5.0 (Windows; U; Windows 98; Win 9x 4....	Segment
C		
39997	Mozilla/5.0 (Windows; U; Windows NT 4.0) Apple...	Segment
C		
39998	Mozilla/5.0 (X11; Linux i686) AppleWebKit/536....	Segment
B		
39999	Mozilla/5.0 (iPod; U; CPU iPhone OS 3_0 like M...	Segment
A		
Geo-location Data Proxy Information Firewall		
Logs \		
0	Jamshedpur, Sikkim	150.9.97.135 Log Data
1	Bilaspur, Nagaland	NaN Log Data
2	Bokaro, Rajasthan	114.133.48.179 Log Data
3	Jaunpur, Rajasthan	NaN NaN
4	Anantapur, Tripura	149.6.110.119 NaN
...	...	...
39995	Nashik, Manipur	NaN Log Data
39996	Vadodara, Mizoram	60.51.30.46 Log Data
39997	Mahbubnagar, Himachal Pradesh	NaN Log Data
39998	Rourkela, Arunachal Pradesh	137.76.130.8 Log Data
39999	Pudukkottai, West Bengal	112.169.115.139 Log Data
IDS/IPS Alerts Log Source		
0	NaN	Server
1	NaN	Firewall
2	Alert Data	Firewall

```

3      Alert Data  Firewall
4      Alert Data  Firewall
...
39995  Alert Data  Firewall
39996      NaN    Firewall
39997  Alert Data    Server
39998      NaN    Server
39999  Alert Data  Firewall

```

```
[40000 rows x 25 columns]
```

## Exploring the Dataset

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 40000 entries, 0 to 39999
```

```
Data columns (total 25 columns):
```

#	Column	Non-Null Count	Dtype
0	Timestamp	40000 non-null	object
1	Source IP Address	40000 non-null	object
2	Destination IP Address	40000 non-null	object
3	Source Port	40000 non-null	int64
4	Destination Port	40000 non-null	int64
5	Protocol	40000 non-null	object
6	Packet Length	40000 non-null	int64
7	Packet Type	40000 non-null	object
8	Traffic Type	40000 non-null	object
9	Payload Data	40000 non-null	object
10	Malware Indicators	20000 non-null	object
11	Anomaly Scores	40000 non-null	float64
12	Alerts/Warnings	19933 non-null	object
13	Attack Type	40000 non-null	object
14	Attack Signature	40000 non-null	object
15	Action Taken	40000 non-null	object
16	Severity Level	40000 non-null	object
17	User Information	40000 non-null	object
18	Device Information	40000 non-null	object
19	Network Segment	40000 non-null	object
20	Geo-location Data	40000 non-null	object
21	Proxy Information	20149 non-null	object
22	Firewall Logs	20039 non-null	object
23	IDS/IPS Alerts	19950 non-null	object
24	Log Source	40000 non-null	object

```
dtypes: float64(1), int64(3), object(21)
```

```
memory usage: 7.6+ MB
```

```
df.columns
Index(['Timestamp', 'Source IP Address', 'Destination IP Address',
      'Source Port', 'Destination Port', 'Protocol', 'Packet Length',
      'Packet Type', 'Traffic Type', 'Payload Data', 'Malware
Indicators',
      'Anomaly Scores', 'Alerts/Warnings', 'Attack Type', 'Attack
Signature',
      'Action Taken', 'Severity Level', 'User Information',
      'Device Information', 'Network Segment', 'Geo-location Data',
      'Proxy Information', 'Firewall Logs', 'IDS/IPS Alerts', 'Log
Source'],
      dtype='object')

df.shape
(40000, 25)
```

## Taking care of the Null Values

```
df.isnull().sum().sort_values(ascending=False)
```

Alerts/Warnings	20067
IDS/IPS Alerts	20050
Malware Indicators	20000
Firewall Logs	19961
Proxy Information	19851
Attack Type	0
Geo-location Data	0
Network Segment	0
Device Information	0
User Information	0
Severity Level	0
Action Taken	0
Attack Signature	0
Timestamp	0
Source IP Address	0
Anomaly Scores	0
Payload Data	0
Traffic Type	0
Packet Type	0
Packet Length	0
Protocol	0
Destination Port	0
Source Port	0
Destination IP Address	0
Log Source	0
dtype: int64	

There are **5** columns with null values:

- **Alerts/Warnings** = 20067
- **IDS/IPS Alerts** = 20050
- **Malware Indicators** = 20000
- **Firewall Logs** = 19961
- **Proxy Information** = 19851

[illegible]

```

else x)
df['Malware Indicators'].unique()
array(['IoC Detected', 'No Detection'], dtype=object)
df['Firewall Logs'] = df['Firewall Logs'].apply(lambda x: 'No Data'
                                                if pd.isna(x)
                                                else x)
df['Firewall Logs'].unique()
array(['Log Data', 'No Data'], dtype=object)
df['Proxy Information'].unique()
array(['150.9.97.135', nan, '114.133.48.179', ..., '60.51.30.46',
       '137.76.130.8', '112.169.115.139'], dtype=object)
df['Proxy Information'] = df['Proxy Information'].apply(lambda x: 'No
Proxy Data'
                                                        if pd.isna(x)
                                                        else x)
df['Proxy Information'].unique()
array(['150.9.97.135', 'No Proxy Data', '114.133.48.179', ...,
       '60.51.30.46', '137.76.130.8', '112.169.115.139'],
dtype=object)
df.isnull().sum().sort_values(ascending=False)
Timestamp                0
Attack Type              0
IDS/IPS Alerts           0
Firewall Logs            0
Proxy Information         0
Geo-location Data        0
Network Segment          0
Device Information       0
User Information         0
Severity Level           0
Action Taken             0
Attack Signature         0
Alerts/Warnings          0
Source IP Address        0
Anomaly Scores           0
Malware Indicators       0
Payload Data             0
Traffic Type             0
Packet Type              0
Packet Length            0
Protocol                 0
Destination Port         0

```

```
Source Port      0
Destination IP Address  0
Log Source      0
dtype: int64
```

**No more Null Values!!!**

```
df.columns
Index(['Timestamp', 'Source IP Address', 'Destination IP Address',
      'Source Port', 'Destination Port', 'Protocol', 'Packet Length',
      'Packet Type', 'Traffic Type', 'Payload Data', 'Malware
Indicators',
      'Anomaly Scores', 'Alerts/Warnings', 'Attack Type', 'Attack
Signature',
      'Action Taken', 'Severity Level', 'User Information',
      'Device Information', 'Network Segment', 'Geo-location Data',
      'Proxy Information', 'Firewall Logs', 'IDS/IPS Alerts', 'Log
Source'],
      dtype='object')
```

## Breaking Down Device Information

```
df['Device Information']

0      Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...
1      Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...
2      Mozilla/5.0 (compatible; MSIE 9.0; Windows NT ...
3      Mozilla/5.0 (Macintosh; PPC Mac OS X 10_11_5; ...
4      Mozilla/5.0 (compatible; MSIE 5.0; Windows NT ...
...
39995   Mozilla/5.0 (iPad; CPU iPad OS 14_2_1 like Mac...
39996   Mozilla/5.0 (Windows; U; Windows 98; Win 9x 4....
39997   Mozilla/5.0 (Windows; U; Windows NT 4.0) Apple...
39998   Mozilla/5.0 (X11; Linux i686) AppleWebKit/536....
39999   Mozilla/5.0 (iPod; U; CPU iPhone OS 3_0 like M...
Name: Device Information, Length: 40000, dtype: object
```

## **\*\* Creating a Device/OS Column\*\***

```
import re

devices = [
    r'Windows',
    r'Linux',
    r'Android',
```



```

r'iPad',
r'iPod',
r'iPhone',
r'Macintosh']

def device_os_finder(user_agent):
    for device in devices:
        match_device = re.search(device, user_agent, re.I)  # re.I
        makes the search case-insensitive
        if match_device:
            return match_device.group()
    return 'Unknown'

# Extract device or OS
df['Device/OS'] = df['Device Information'].apply(device_os_finder)
df['Device/OS'].head(10)

0      Windows
1      Windows
2      Windows
3  Macintosh
4      Windows
5      Linux
6      Linux
7  Macintosh
8  Macintosh
9      Windows
Name: Device/OS, dtype: object

df['Device/OS'].value_counts()

Device/OS
Windows      17953
Linux        8840
Macintosh    5813
iPod         2656
Android      1620
iPhone       1567
iPad         1551
Name: count, dtype: int64

```

## **\*\* Creating a Browser Column\*\***

```

df['Device Information']

0      Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...
1      Mozilla/5.0 (compatible; MSIE 8.0; Windows NT ...
2      Mozilla/5.0 (compatible; MSIE 9.0; Windows NT ...

```

```

3      Mozilla/5.0 (Macintosh; PPC Mac OS X 10_11_5; ...
4      Mozilla/5.0 (compatible; MSIE 5.0; Windows NT ...
...
39995   Mozilla/5.0 (iPad; CPU iPad OS 14_2_1 like Mac...
39996   Mozilla/5.0 (Windows; U; Windows 98; Win 9x 4....
39997   Mozilla/5.0 (Windows; U; Windows NT 4.0) Apple...
39998   Mozilla/5.0 (X11; Linux i686) AppleWebKit/536....
39999   Mozilla/5.0 (iPod; U; CPU iPhone OS 3_0 like M...
Name: Device Information, Length: 40000, dtype: object

df['Browser'] = df['Device Information'].str.split('/').str[0]
df['Browser']

0      Mozilla
1      Mozilla
2      Mozilla
3      Mozilla
4      Mozilla
...
39995   Mozilla
39996   Mozilla
39997   Mozilla
39998   Mozilla
39999   Mozilla
Name: Browser, Length: 40000, dtype: object

df['Browser'].value_counts()

Browser
Mozilla      31951
Opera         8049
Name: count, dtype: int64

```

## Creating Additional Time Columns

```

# Converting to datetime

df['Timestamp'] = pd.to_datetime(df['Timestamp'])
df['Timestamp'].info()

<class 'pandas.core.series.Series'>
RangeIndex: 40000 entries, 0 to 39999
Series name: Timestamp
Non-Null Count  Dtype
-----
40000 non-null  datetime64[ns]
dtypes: datetime64[ns](1)
memory usage: 312.6 KB

```

```

df['Year'] = df['Timestamp'].dt.year
df['Month'] = df['Timestamp'].dt.month
df['DayofWeek'] = df['Timestamp'].dt.dayofweek
df['Day'] = df['Timestamp'].dt.day
df['Hour'] = df['Timestamp'].dt.hour
df['Minute'] = df['Timestamp'].dt.minute
df['Second'] = df['Timestamp'].dt.second

```

```
df
```

	Timestamp	Source IP Address	Destination IP Address	\
0	2023-05-30 06:33:58	103.216.15.12	84.9.164.252	
1	2020-08-26 07:08:30	78.199.217.198	66.191.137.154	
2	2022-11-13 08:23:25	63.79.210.48	198.219.82.17	
3	2023-07-02 10:38:46	163.42.196.10	101.228.192.255	
4	2023-07-16 13:11:07	71.166.185.76	189.243.174.238	
...	...	...	...	
39995	2023-05-26 14:08:42	26.36.109.26	121.100.75.240	
39996	2023-03-27 00:38:27	17.21.163.81	196.108.134.78	
39997	2022-03-31 01:45:49	162.35.217.57	98.107.0.15	
39998	2023-09-22 18:32:38	208.72.233.205	173.79.112.252	
39999	2023-10-10 11:59:52	14.102.21.108	109.198.45.7	

	Source Port	Destination Port	Protocol	Packet Length	Packet
Type \					
0	31225	17616	ICMP	503	
Data					
1	17245	48166	ICMP	1174	
Data					
2	16811	53600	UDP	306	
Control					
3	20018	32534	UDP	385	
Data					
4	6131	26646	TCP	1462	
Data					
...	...	...	...	...	...
...					
39995	31005	6764	UDP	1428	
Control					
39996	2553	28091	UDP	1184	
Control					
39997	22505	25152	UDP	1043	
Data					
39998	20013	2703	UDP	483	
Data					
39999	50137	55575	ICMP	1175	
Control					

	Traffic Type	Payload Data
...	\	

```

0          HTTP Qui natus odio asperiores nam. Optio nobis ius...
...
1          HTTP Aperiam quos modi officiis veritatis rem. Omni...
...
2          HTTP Perferendis sapiente vitae soluta. Hic delectu...
...
3          HTTP Totam maxime beatae expedita explicabo porro l...
...
4          DNS  Odit nesciunt dolorem nisi iste iusto. Animi v...
...
...          ...
...
39995      HTTP Quibusdam ullam consequatur consequuntur accus...
...
39996      HTTP Quaerat neque esse. Animi expedita natus commo...
...
39997      DNS  Enim at aspernatur illum. Saepe numquam eligen...
...
39998      FTP  Officiis dolorem sed harum provident earum dis...
...
39999      HTTP Eligendi omnis voluptate nihil voluptatibus do...
...

```

Minute	Log	Source	Device/OS	Browser	Year	Month	DayofWeek	Day	Hour
0		Server	Windows	Mozilla	2023	5	1	30	6
33									
1		Firewall	Windows	Mozilla	2020	8	2	26	7
8									
2		Firewall	Windows	Mozilla	2022	11	6	13	8
23									
3		Firewall	Macintosh	Mozilla	2023	7	6	2	10
38									
4		Firewall	Windows	Mozilla	2023	7	6	16	13
11									
...		...	...	...	...	...	...	..	...
...									
39995		Firewall	iPad	Mozilla	2023	5	4	26	14
8									
39996		Firewall	Windows	Mozilla	2023	3	0	27	0
38									
39997		Server	Windows	Mozilla	2022	3	3	31	1
45									
39998		Server	Linux	Mozilla	2023	9	4	22	18
32									
39999		Firewall	iPod	Mozilla	2023	10	1	10	11
59									

Second

0	58
1	30
2	25
3	46
4	7
...	...
39995	42
39996	27
39997	49
39998	38
39999	52

[40000 rows x 34 columns]

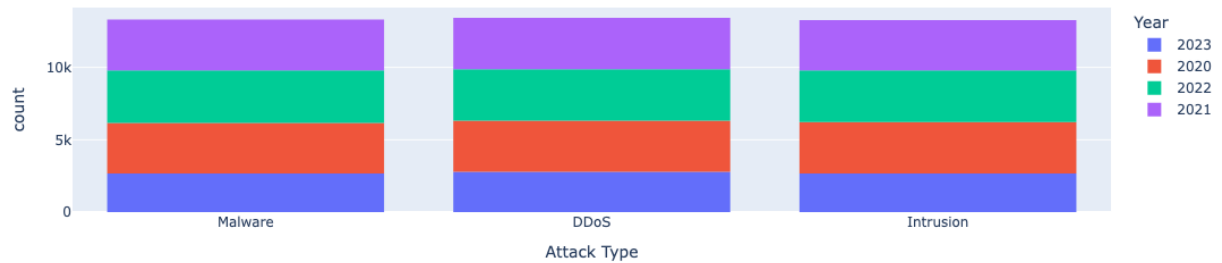
## Insights

```
df.columns
```

```
Index(['Timestamp', 'Source IP Address', 'Destination IP Address',
      'Source Port', 'Destination Port', 'Protocol', 'Packet Length',
      'Packet Type', 'Traffic Type', 'Payload Data', 'Malware
Indicators',
      'Anomaly Scores', 'Alerts/Warnings', 'Attack Type', 'Attack
Signature',
      'Action Taken', 'Severity Level', 'User Information',
      'Device Information', 'Network Segment', 'Geo-location Data',
      'Proxy Information', 'Firewall Logs', 'IDS/IPS Alerts', 'Log
Source',
      'Device/OS', 'Browser', 'Year', 'Month', 'DayofWeek', 'Day',
      'Hour',
      'Minute', 'Second'],
      dtype='object')
```

```
plt = px.histogram(df, x='Attack Type', color='Year', title='Number of
Attack Types by Year')
plt.show()
```

Number of Attack Types by Year



```
df.groupby(['Year'])['Attack Type'].value_counts()
```

```
Year  Attack Type
2020  Intrusion      3551
      DDoS           3533
      Malware        3489
2021  DDoS           3545
      Malware        3518
      Intrusion      3475
2022  Malware        3629
      Intrusion      3563
      DDoS           3558
2023  DDoS           2792
      Intrusion      2676
      Malware        2671
Name: count, dtype: int64
```

- In **2020**, **Intrusion** attacks were the most frequent
- In **2021** and **2023**, **DDoS** attacks were the most frequent
- In **2022**, **Malware** attacks were the most frequent

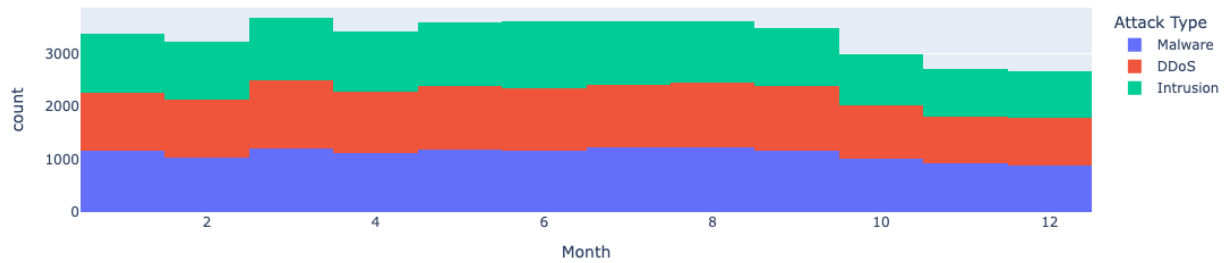
```
df.groupby(['Year'])['Attack Type'].count()
```

```
Year
2020    10573
2021    10538
2022    10750
2023     8139
Name: Attack Type, dtype: int64
```

**Least** amount of attack types in **2023**!

```
plt = px.histogram(df, x='Month', color='Attack Type', title='Number
of Attack Types by Month')
plt.show()
```

Number of Attack Types by Month



```
month_attacks = df.groupby(['Month'])['Attack Type'].count()
month_attacks.sort_values(ascending=False)
```

```
Month
3      3678
7      3623
8      3615
6      3609
5      3595
9      3482
4      3421
1      3378
2      3232
10     2989
11     2703
12     2675
Name: Attack Type, dtype: int64
```

**Most attacks happened in March and the least amount happened in December!**

```
attack_types_month = df.groupby(['Month'])['Attack Type'].value_counts()
attack_types_month.sort_values(ascending=False)
```

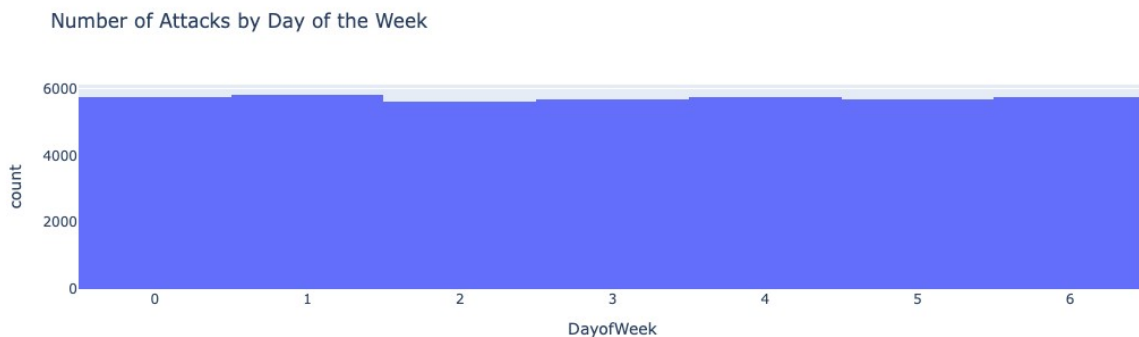
```
Month  Attack Type
3      DDoS          1299
6      Intrusion     1268
7      Malware       1236
8      DDoS          1226
7      Intrusion     1224
9      DDoS          1221
8      Malware       1216
5      Intrusion     1212
      DDoS          1200
3      Malware       1197
6      DDoS          1190
5      Malware       1183
```

3	Intrusion	1182
8	Intrusion	1173
9	Malware	1172
4	DDoS	1166
7	DDoS	1163
1	Malware	1163
6	Malware	1151
4	Intrusion	1140
1	Intrusion	1116
4	Malware	1115
2	Intrusion	1107
1	DDoS	1099
9	Intrusion	1089
2	DDoS	1085
	Malware	1040
10	DDoS	1015
	Malware	1013
	Intrusion	961
11	Malware	935
	Intrusion	899
12	DDoS	895
	Intrusion	894
	Malware	886
11	DDoS	869

Name: count, dtype: int64

- **March: Most DDoS Attacks**
- **June: Most Intrusion Attacks**
- **July: Most Malware Attacks**

```
plt = px.histogram(df, x='DayofWeek', title='Number of Attacks by Day of the Week')
plt.show()
```



**Monday** is the most popular day for an attack!



```
df['DayofWeek'].value_counts()
```

```
DayofWeek
```

```
1    5813
```

```
4    5753
```

```
0    5752
```

```
6    5744
```

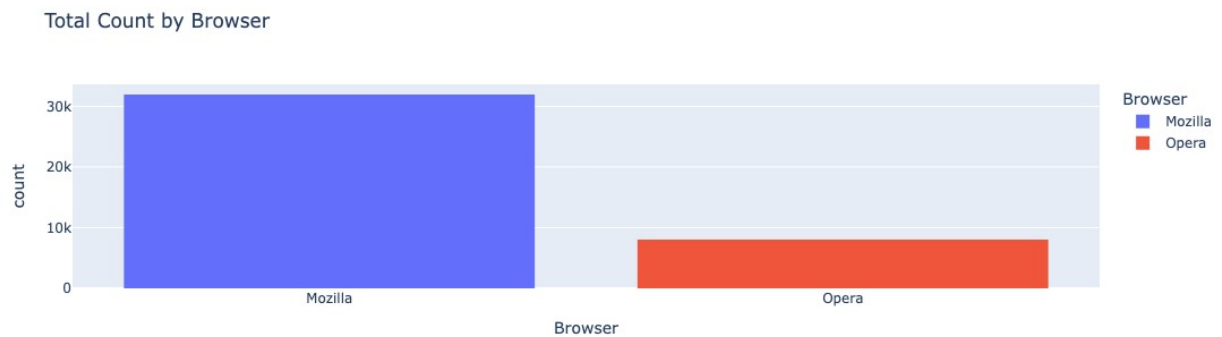
```
3    5676
```

```
5    5663
```

```
2    5599
```

```
Name: count, dtype: int64
```

```
plt = px.histogram(df ,x='Browser', color = 'Browser', title = 'Total  
Count by Browser')  
plt.show()
```



```
plt = px.pie(df ,names='Device/OS', title = 'Device/OS Types')  
plt.show()
```

Device/OS Types



Top 3 Device/OS Types:

1. **Windows**
2. **Linux**
3. **Macintosh**

```
plt = px.histogram(df, x='Traffic Type', color='Browser',
title='Traffic Type by Browser')
plt.show()
```



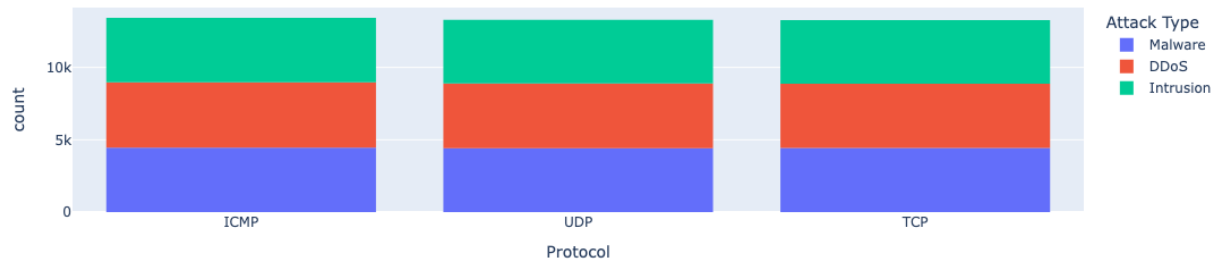
### Average Packet Length by Device/OS Type for each Browser

```
packet_length = df.groupby(['Browser', 'Device/OS']).agg({'Packet Length': 'mean'})
packet_length.sort_values('Packet Length', ascending=False)
```

		Packet Length
Browser	Device/OS	
Mozilla	iPad	800.304320
Opera	Windows	788.932635
Mozilla	Android	786.717284
	iPod	784.408886
	Macintosh	782.677963
	Windows	779.130513
Opera	Linux	778.100223
Mozilla	iPhone	777.880664
	Linux	774.952907

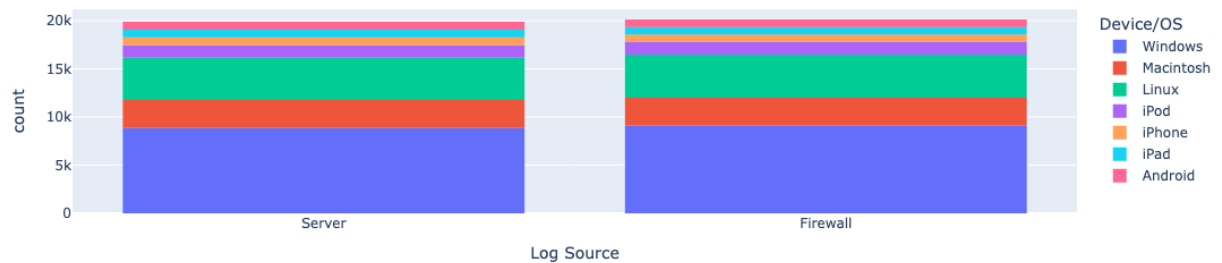
```
plt = px.histogram(df, x='Protocol', color = 'Attack Type', title =
'Number of Attack Type by Protocol')
plt.show()
```

Number of Attack Type by Protocol



```
plt = px.histogram(df, x='Log Source', color='Device/OS',
title='Number of Device/OS Types by Log Source')
plt.show()
```

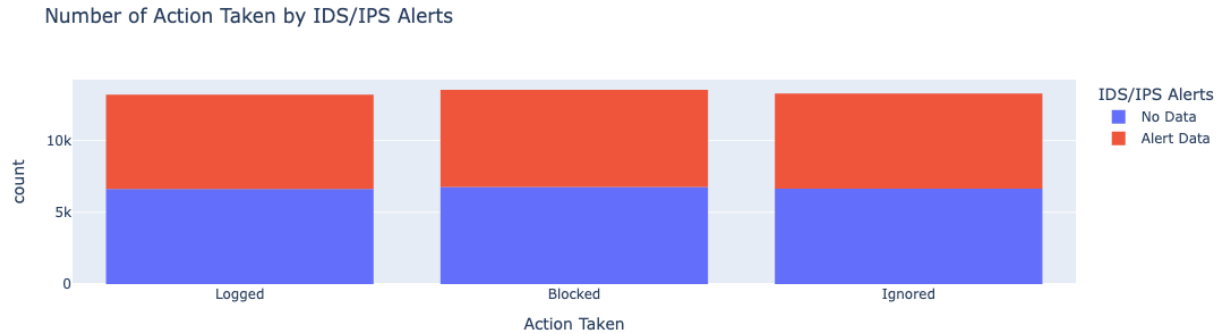
Number of Device/OS Types by Log Source



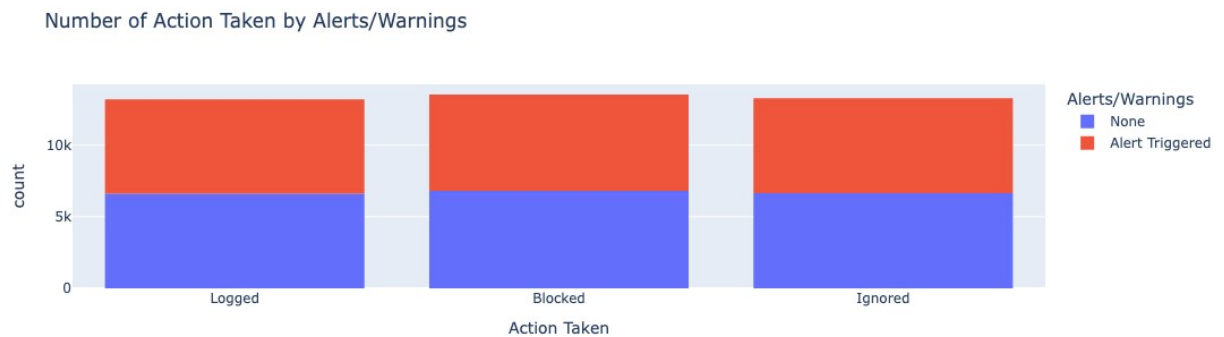
```
df.groupby(['Log Source'])['Device/OS'].value_counts()
```

```
Log Source  Device/OS
Firewall    Windows      9092
            Linux       4449
            Macintosh   2920
            iPod       1347
            Android      792
            iPad        765
            iPhone       751
Server      Windows     8861
            Linux       4391
            Macintosh   2893
            iPod       1309
            Android      828
            iPhone       816
            iPad        786
Name: count, dtype: int64
```

```
plt = px.histogram(df, x='Action Taken', color='IDS/IPS Alerts',
title= 'Number of Action Taken by IDS/IPS Alerts')
plt.show()
```



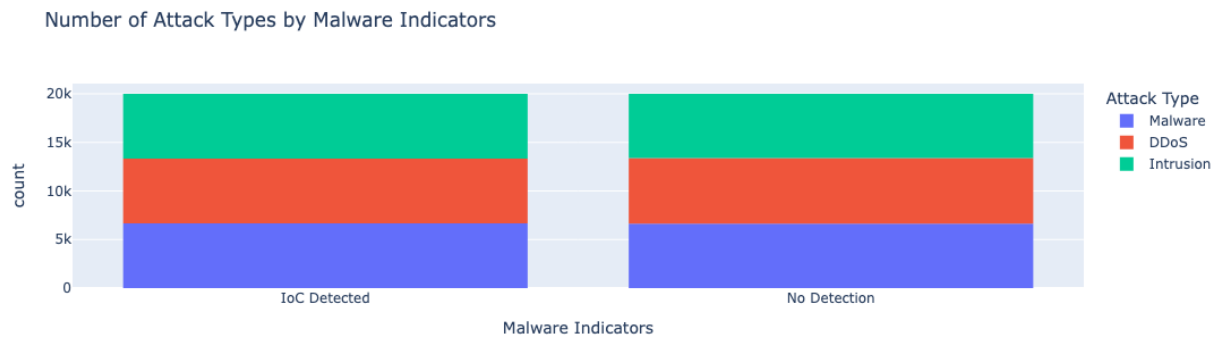
```
plt = px.histogram(df, x='Action Taken', color='Alerts/Warnings',
title= 'Number of Action Taken by Alerts/Warnings')
plt.show()
```



```
df['Action Taken'].value_counts()
```

```
Action Taken
Blocked    13529
Ignored    13276
Logged     13195
Name: count, dtype: int64
```

```
plt = px.histogram(df, x='Malware Indicators',color='Attack Type',
title='Number of Attack Types by Malware Indicators')
plt.show()
```



```
df['Malware Indicators'].value_counts()
```

Malware Indicators

IoC Detected 20000

No Detection 20000

Name: count, dtype: int64

```
df.groupby(['Malware Indicators', 'Attack Type']).agg({'Packet Length': 'mean'})
```

		Packet Length
Malware Indicators	Attack Type	
IoC Detected	DDoS	780.819958
	Intrusion	787.093473
	Malware	777.033678
No Detection	DDoS	789.797313
	Intrusion	774.694545
	Malware	779.070631

```
df.groupby(['Malware Indicators'])['Attack Type'].value_counts()
```

Malware Indicators Attack Type

IoC Detected Malware 6681

Intrusion 6665

DDoS 6654

No Detection DDoS 6774

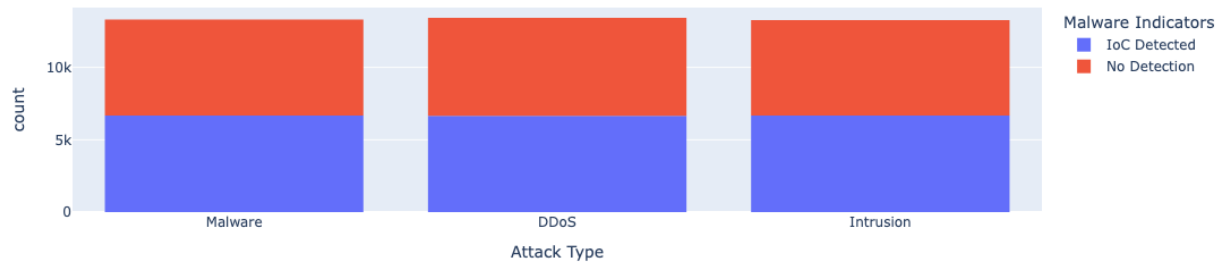
Malware 6626

Intrusion 6600

Name: count, dtype: int64

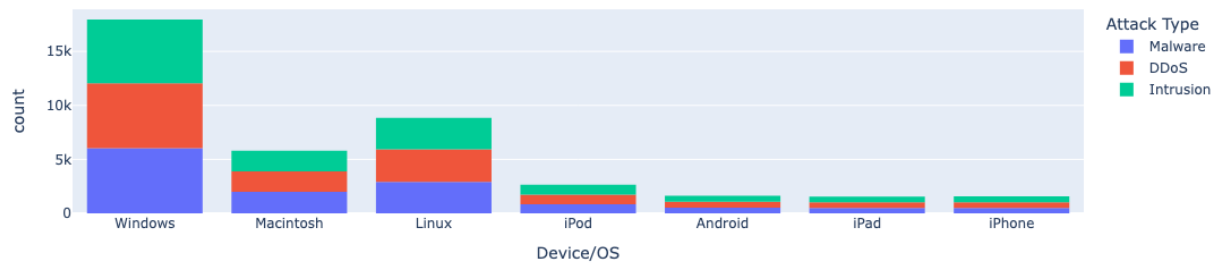
```
plt = px.histogram(df, x='Attack Type', color='Malware Indicators',
title='Number of Attack Types by Malware Indicators')
plt.show()
```

Number of Attack Types by Malware Indicators



```
plt = px.histogram(df, x= 'Device/OS', color = 'Attack Type', title =
'Number of Malware Attacks by Browser and Devices')
plt.show()
```

Number of Malware Attacks by Browser and Devices



```
plt = px.pie(df, names = 'Attack Type', title = 'Attack Type
Distribution')
plt.show()
```

Attack Type Distribution



```
plt = px.histogram(df ,x='Geo-location Data', color = 'Attack Type',
title = 'Number of Attack Types by Geo-Location',width=1000,
```

```
plt.show()
height=600)
```

Number of Attack Types by Geo-Location



```
geo = df.groupby(['Geo-location Data'])['Attack Type'].value_counts()
geo.sort_values(ascending=False).head(20)
```

Geo-location Data	Attack Type	count
Ghaziabad, Jharkhand	Intrusion	10
Aligarh, Chhattisgarh	Malware	9
Aurangabad, Nagaland	Malware	9
Srikakulam, Uttarakhand	Intrusion	8
Yamunanagar, Arunachal Pradesh	Malware	8
Rampur, Gujarat	Intrusion	8
Jalgaon, Mizoram	Malware	8
Amroha, Sikkim	Intrusion	8
Panvel, Jharkhand	Intrusion	8
Kochi, Tamil Nadu	DDoS	7
Ghaziabad, Meghalaya	Intrusion	7
Tenali, Madhya Pradesh	Malware	7
Ghaziabad, Nagaland	Malware	7
Pimpri-Chinchwad, Manipur	DDoS	7
Karnal, Tamil Nadu	Intrusion	7
Kottayam, Nagaland	Malware	7
Junagadh, Telangana	DDoS	7
Kadapa, Mizoram	Intrusion	7
Fatehpur, Gujarat	DDoS	7
Hospet, Gujarat	Malware	7

Name: count, dtype: int64