

CYBER ATTACK ANALYSIS: SYN FLOOD ATTACK

Network Traffic Threat
Intelligence Project

Yovel Hadari

Hakeriot - Mentoriot



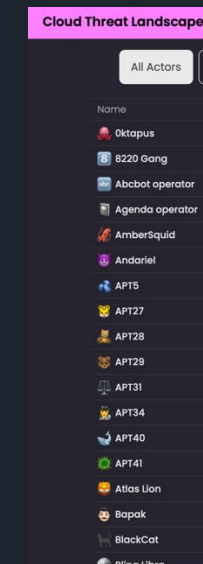
ABOUT

Analysis of a dataset comprising network traffic logs and packets to enhance hunting and detection capabilities using Python.

- Identifying anomalies and correlations between vectors and characteristics during an APT attack, specifically targeting DoS (Denial of Service) incidents.
- DoS/DDoS attacks can be integrated into APT (Advanced Persistent Threat) campaigns to: Distract security teams while executing a deeper intrusion.

DDoS Protection

Type	Threat Detection
D3FEND Tactic	Network Traffic Analysis (D3-NTA)
TLDR	Shields networks from attacks that overwhelm services with traffic.
Description	Measures and techniques designed to protect a network or server from distributed denial-of-service attacks, which aim to make a service unavailable by overwhelming it with traffic.
Techniques	

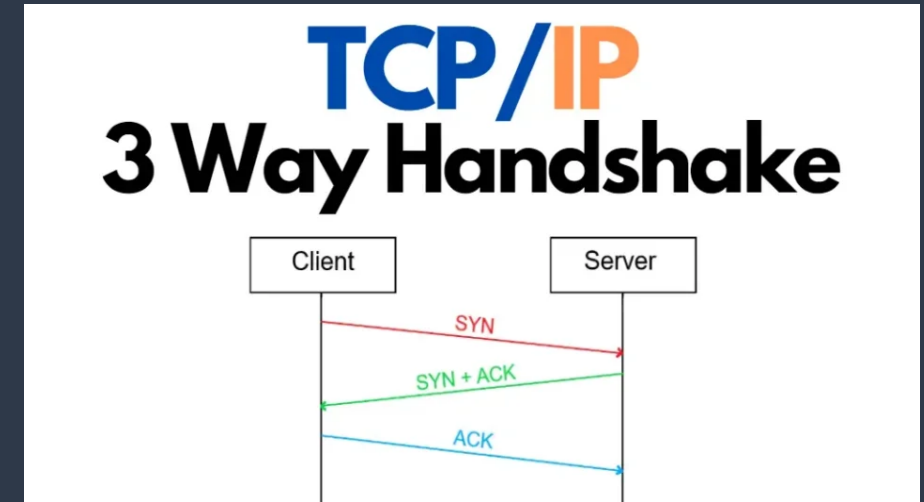


<https://threats.wiz.io/defenses/ddos-protection>

<https://threats.wiz.io/all-actors>

TCP Handshake Mechanism

- **SYN Flood Attack:**
 - SYN Flood exploits the TCP handshake process
 - Attacker sends multiple SYN requests but never completes the handshake
 - Server resources remain allocated, leading to exhaustion

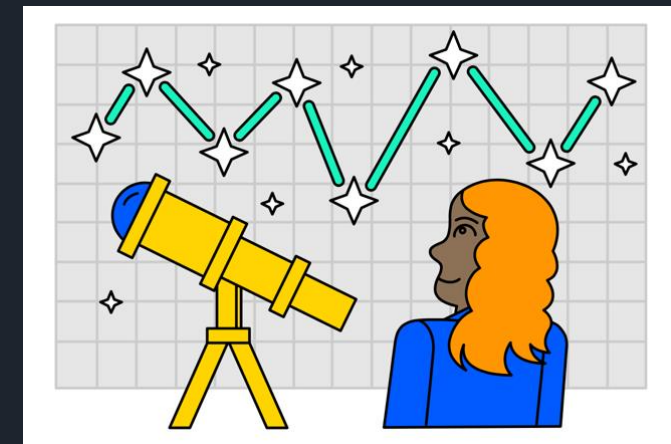


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- **SYN Flood Attack Analysis- Objective of this Report:**

attack_type	Dst IP	CountSrc_uniq	DateTime
SYN_Flood	206.207.50.50	1	19/07/2019 15:15

- Identify key attack characteristics
- Propose possible mitigation strategies



ABOUT the CODE

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

```
df['group_DosDetect_id'] = df.groupby(['Dst IP', pd.Grouper(freq='5min')]).ngroup()

# - 'grouped_summary': Contains aggregate statistics (e.g., counts, averages) per Dst IP and time interval.
grouped_summary = df.groupby(['group_DosDetect_id', 'Dst IP', pd.Grouper(freq='5min')]).agg(
    # Aggregate summarized statistics per group
    CountRequests=('Src IP', 'count'),
    CountSrc_uniq=('Src IP', 'nunique'), # Count unique source IPs
    Flow_Packets_s_avg=('Flow Packets/s', 'mean'),
    Flow_Bytes_s_avg=('Flow Bytes/s', 'mean'),
    SYN_count_sum=('SYN Flag Count', 'sum'),
    ACK_count_sum=('ACK Flag Count', 'sum'),
    Bwd_sum=('Total Bwd packets', 'sum'),
    Fwd_sum=('Total Fwd Packet', 'sum'),
).reset_index()

# Calculating ratios separately to avoid referencing within the aggregation function
grouped_summary['SYN_ACK_Ratio'] = grouped_summary['SYN_count_sum'] / ((grouped_summary['ACK_count_sum']) + 1)
grouped_summary['ACK_SYN_Ratio'] = grouped_summary['ACK_count_sum'] / (grouped_summary['SYN_count_sum'] + 1)
grouped_summary['Bwd_Fwd_Ratio'] = grouped_summary['Bwd_sum'] / (grouped_summary['Fwd_sum'] + 1)
```

Data Processing & Aggregation

- **Data Grouping: DST and slats of 5 min**
- **Unique Key Assignment:** Added a key (group_DosDetect_id) for linking further aggregations of Src details.
- **Detected Attack:** Identified DoS attack - 3 if's
- FP/FN/TP/TN

• Threshold

```
def detect_outliers_iqr(df, column):
    # Normalized Outlier
    Q1 = df[column].quantile(0.25)
    Q3 = df[column].quantile(0.75)
    iqr = Q3 - Q1
    lower_bound = Q1 - 1.5 * iqr
    upper_bound = Q3 + 1.5 * iqr

    outlier_dict = {
        "dict_name": "outlier_dict",
        # "df_pointer": point_to_df,
        "column": column,
        "lower_bound": lower_bound,
        "upper_bound": upper_bound,
        "iqr": iqr,
        "std": df[column].std()
    }
    return outlier_dict
```

• Protocols

```
# Global dictionary for protocol
mappings
PROTOCOL_MAPPING = {
    1: 'ICMP', # Internet Control
Message Protocol
    6: 'TCP', # Transmission Control
Protocol
    17: 'UDP', # User Datagram
Protocol
    50: 'ESP', # Encapsulating
Security Payload
    51: 'AH', # Authentication Header
    8: 'EGP', # Exterior Gateway
Protocol
    # Add other protocol mappings as
needed
}
```

• Ports

```
def categorize_ports(port):
    """
        Categorize ports into Well-Known Ports,
        Registered Ports, and Dynamic/Private Ports.

        - Well-Known Ports: 0-1023
        - Registered Ports: 1024-49151
        - Dynamic/Private Ports: 49152-65535
    """
    if port < 1024:
        return 'Well-Known Ports'
    elif 1024 <= port <= 49151:
        return 'Registered Ports'
    else:
        return 'Dynamic/Private Ports'

# Define protocol categories
```

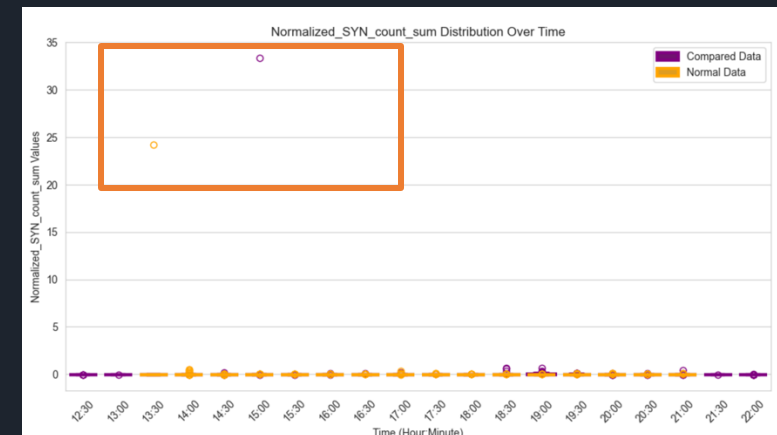
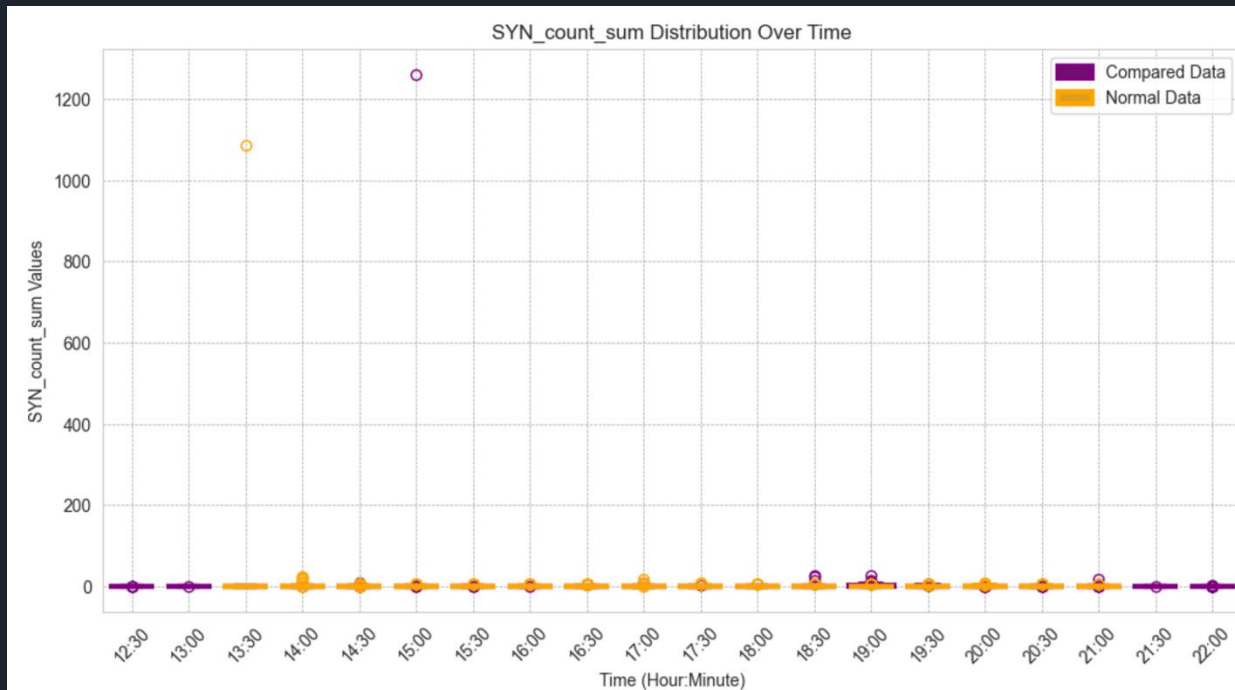
• Dict of DoS Attacks

```
if attack_type != 'Normal':  
    dict_dos_attacks[key] = {
```

```
"Attack_id": f"{key}-{row.group_DosDetect_id}",  
"group_DosDetect_id": row.group_DosDetect_id,  
"Source_dfName": df_filename,  
"attack_type": attack_type,  
f"{column}": Ratio,  
"Dst IP": row['Dst IP'],  
"CountRequests": row['CountRequests'],  
"CountSrc_uniq": row['CountSrc_uniq'], # Count unique source IPs  
"Datetime": row['Datetime'],  
"HourTime": row['HourTime'],  
# Calculate statistics for SYN Flag Count  
"SYN_count_sum": row.SYN_count_sum,  
# Calculate statistics for ACK Flag Count  
"ACK_count_sum": row.ACK_count_sum,  
# Calculate statistics for Fwd Packet Flag Count  
"Fwd_sum": row.Fwd_sum,  
# Calculate statistics for Bwd Packet Flag Count  
"Bwd_sum": row.Bwd_sum,  
# Avg Flow Packets/ Bytes  
"Flow_Packets_s_avg": row.Flow_Packets_s_avg,  
"Flow_Bytes_s_avg": row.Flow_Bytes_s_avg,  
# Source details  
"Src Details": row['Src Details'],  
"SrcIP_uniq": row['SrcIP_uniq'],  
"SrcPort_uniq": row.SrcPort_uniq,  
"SrcPort_categorical": row.SrcPort_categorical,  
"Protocol_uniq": row.Protocol_uniq,  
"Protocol_categorical": row.Protocol_categorical,  
# Port categorial count  
'Well_Known_Port_Count': row.Well_Known_Port_Count,  
'Registered_Port_Count': row.Registered_Port_Count,  
'Dynamic_Private_Port_Count': row.Dynamic_Private_Port_Count,  
# Protocol categorial count  
'Protocol_TCP_Count': row.Protocol_TCP_Count,  
'Protocol_UDP_Count': row.Protocol_UDP_Count,  
'Protocol_ICMP_Count': row.Protocol_TCP_Count,  
'Other_Protocol_Count': row.Other_Protocol_Count  
}
```

SYN Count Over Time

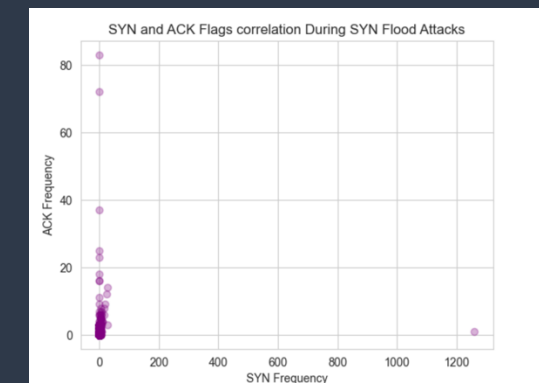
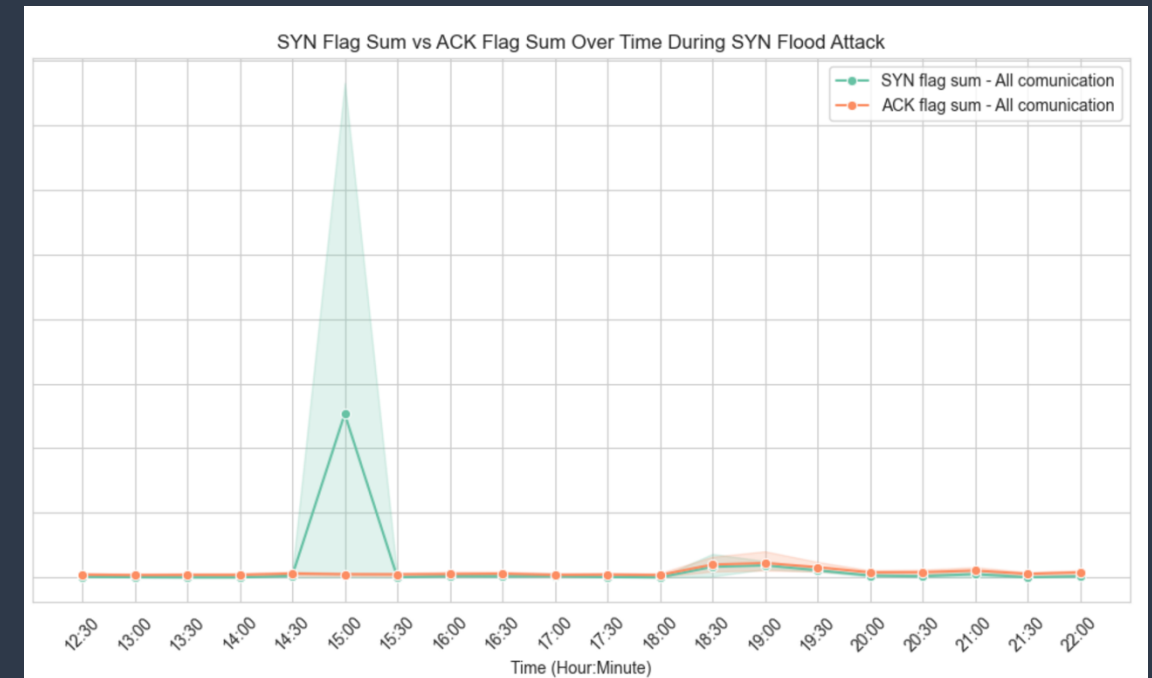
- Key Observations:
 - Unusual spike detected in SYN packet count at 15:15
 - Compared with normal traffic data
 - SYN count above 1200 sum, significantly deviating from normal levels



Anomalous SYN/ACK Ratio

- **Findings:**
 - High SYN to ACK ratio, indicating an incomplete handshake pattern
 - Normal traffic exhibits balanced SYN-ACK pairs
 - Attack traffic shows unmatched SYN packets

CountRequests	Syn_count_sum	ACK_count_sum	SYN_ACK_Ratio
1261	1260	1	630



Anomalous Packets/Bytes (Payload)

- **Findings:**
 - High Packet to Byte flow, indicating a lot of Syn Packets but no Payload

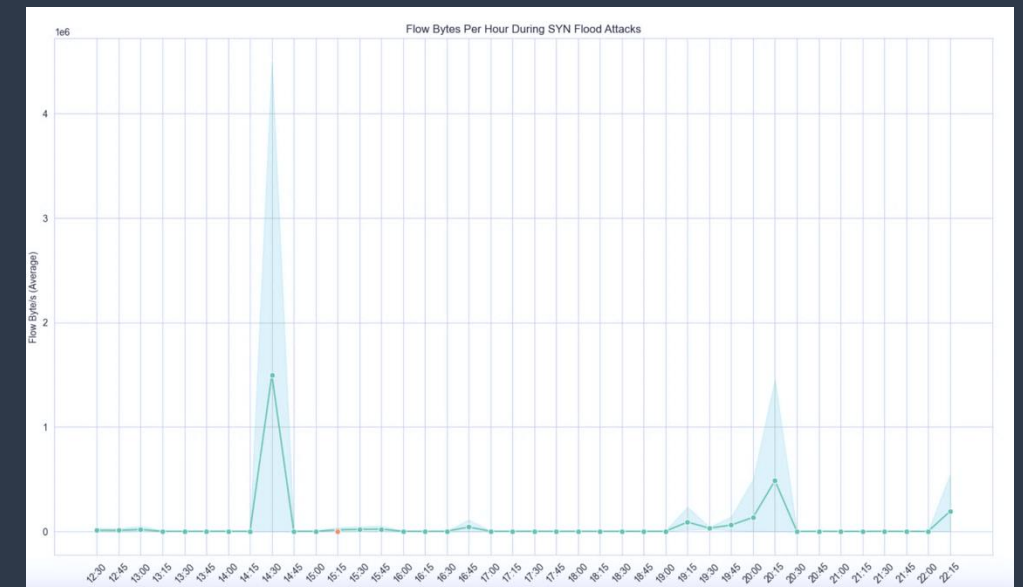
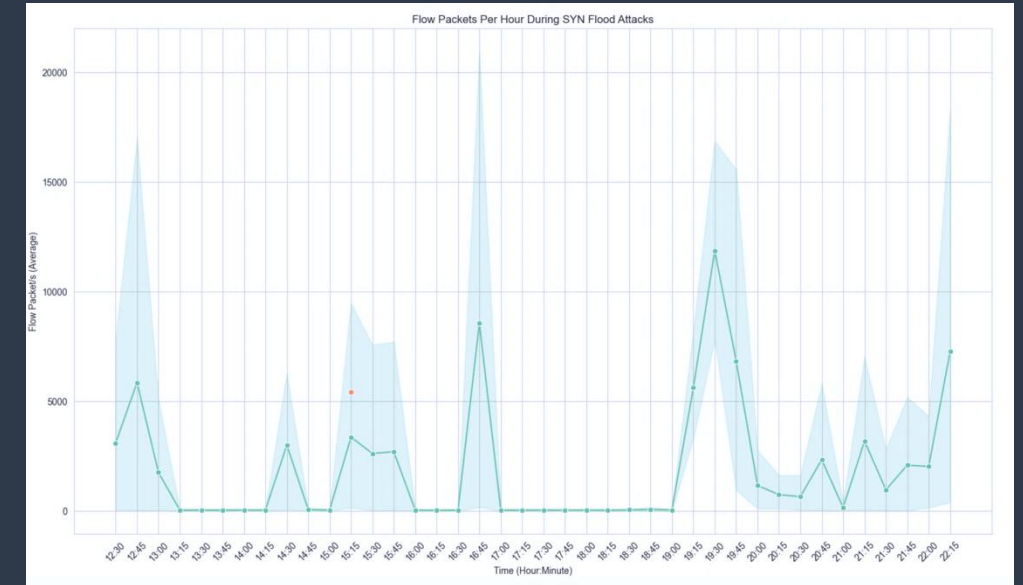
Flow Packets	Flow Bytes
5411	0

A pre-attack spike

may indicate reconnaissance or probing activity.

Decay Pattern

Network Anomaly Detection by Using a Time-Decay Closed Frequent Patterns



Comparison with Normal Traffic:

- Mean, std, Quarters, and max: Bytes > Packets
- Attack scenario shows sudden bursts followed by system slowdowns
- Ports and Protocols

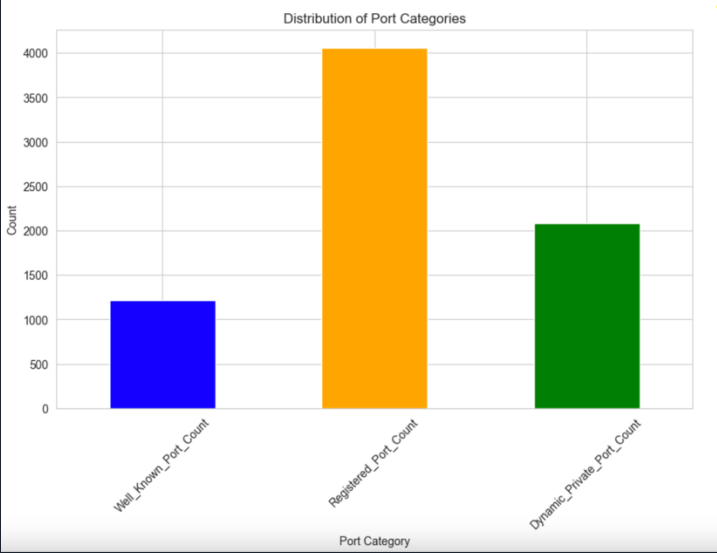
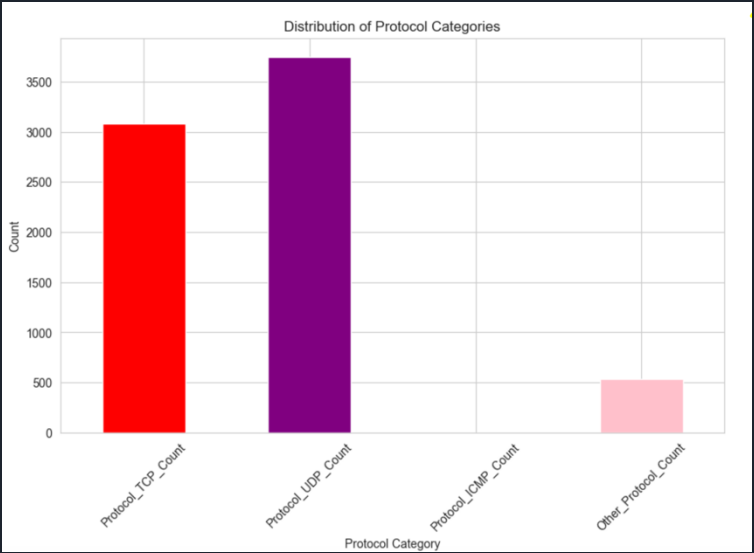
Normal Data

	Flow_Packets_s_avg	Flow_Bytes_s_avg
count	589	589
mean	1,625.344004	45,330.44
std	6,368.88894	742,403.20
min	0.025	0.00
25%	0.198407	0.00
50%	0.747536	36.68
75%	2.207692	187.92
max	131,166.0729	14,484,450.00

Compared Data

CountRequests	Syn_count_sum	ACK_count_sum	SYN_ACK_Ratio	Flow_Packets_s_avg	Flow_Bytes_s_avg	Protocol	Src Port
1261	1260	1	630	5411.156741	0	TCP, Other	[Registered Ports, Well-Known Ports]

Normal Data



Compared Data

Protocol_TCP_Count	Protocol_UDP_Count	Protocol_ICMP_Count	Other_Protocol_Count
1261	0	1261	0

Well_Known_Port_Count	Registered_Port_Count	Dynamic_Private_Port_Count
1	1261	0

Conclusions and what's next

Project Objectives:

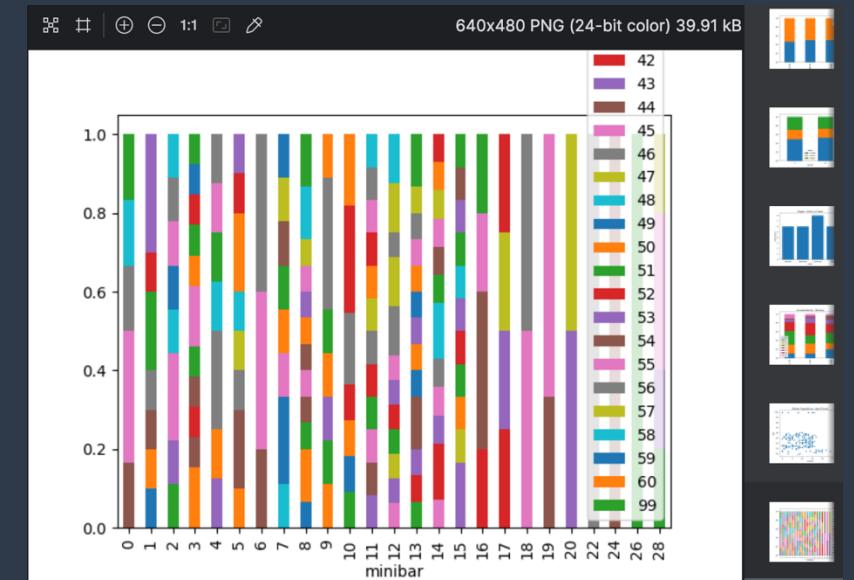
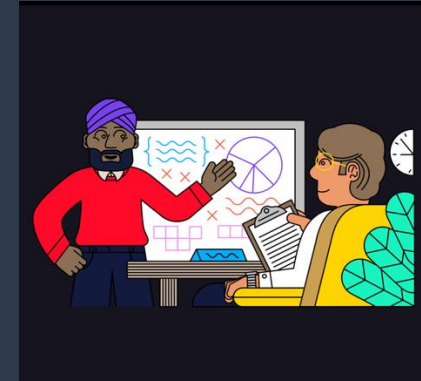
- Detect network anomalies and potential threats.
- Identify cyberattack behavioral patterns.
- Extract insights and correlations for threat intelligence.
- Improve detection accuracy using :
 - customer
 - AI services.

Key Takeaways:

- SYN Flood exploits the TCP handshake process.
- Analysis detected significant anomalies in SYN packet behavior.
- Proactive security measures are essential for network resilience.

Future Work:

- Improving response strategies for customer service and incident handling.
- Enhancing detection using AI-based anomaly detection techniques and unsupervised Data Science methods.



Bibliografy

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