

















## Team Members (Required)





**Reminder:** Make sure to provide **edit access** for this Milestone document to **everyone on your team!**

 Student Name: Dennys Antunish  
 Student Pronouns: He/Him  
 Student Email: [dantunish2@gmail.com](mailto:dantunish2@gmail.com)  
 Favorite Animal: Sharks

 Student Name: Angie Rivera  
 Student Pronouns: She/Her  
 Student Email: [angiervr9@gmail.com](mailto:angiervr9@gmail.com)  
 Favorite Park: McCarren Park

 Student Name: Navruz Asatullaev  
 Student Pronouns:   
 Student Email: [navruz.college@gmail.com](mailto:navruz.college@gmail.com)  
 Favorite Drink: Water

 Student Name: Evgeniia Yeroshkina  
 Student Pronouns: she/her  
 Student Email: [mironova.eug2016@gmail.com](mailto:mironova.eug2016@gmail.com)  
 Favorite Flavor: Vanilla

 Student Name: Aliya Jones  
 Student Pronouns: She/Her  
 Student Email: [aliya.jones@macaulay.cuny.edu](mailto:aliya.jones@macaulay.cuny.edu)  
 Favorite Game: Fortnite

[What are pronouns /](#)  
[Why are they included here?](#)

## Select one (or more) open-source Datasets to analyze (Required)

**Data Set Chosen:** The data set we have chosen to analyze for The Data Dig is...

**Name:** Intrusion detection evaluation dataset (CIC-IDS2017)

**Primary Link:** <https://www.unb.ca/cic/datasets/ids-2017.html>

**Data Set Description:** Where does the data come from? Who generated it? What kind of devices / technologies does it target? What format is the data in?

The CIC-IDS2017 dataset was created by the Canadian Institute for Cybersecurity (CIC) at the University of New Brunswick (UNB). It was developed by Iman Sharafaldin, Arash Habibi Lashkari, and Ali A. Ghorbani, and published in the paper *"Toward Generating a New Intrusion Detection Dataset and Intrusion Traffic Characterization"* (ICISSP 2018). The goal was to provide a modern and realistic dataset for evaluating intrusion detection systems. The data was captured over five days and includes both normal (benign) and attack traffic based on real-world scenarios such as DoS/DDoS, infiltration, web attacks, and botnets. The dataset targets various devices and technologies, including Windows (7, 8.1, 10, Vista), Linux (Ubuntu 12, 14.4, 16.4), macOS, routers, switches, and firewalls. It focuses on both traditional and modern cyberattacks and covers widely used protocols like HTTP, HTTPS, FTP, SSH, and email. The data is available in both PCAP format and CSV files.

**Hypothesis:** What are 3 things you expect to find when you analyze the data?

*Tip: You won't lose points if these hypotheses turn out to be wrong! Make educated guesses!*

**Finding #1:** We expect to find multiple victims on Friday on July 7, 2017

**Finding #2:** We expect to find a lot of logs of an attempted attack at a company

**Finding #3:** We expect to find continuous use of the same IP address for multiple attacks

## Select an incident-response playbook to follow (Required)

**Playbook Chosen:** The playbook we have decided to follow for The Data Dig is...

**Name:** GSPBC-1080 - Impact - Network Denial of Service.pdf

**Primary Link:** [https://github.com/guardsight/gsvsoc\\_cirt-playbook-battle-cards/blob/master/GSPBC-1080%20-%20Impact%20-%20Network%20Denial%20of%20Service.pdf](https://github.com/guardsight/gsvsoc_cirt-playbook-battle-cards/blob/master/GSPBC-1080%20-%20Impact%20-%20Network%20Denial%20of%20Service.pdf)

**Playbook Description:** Who wrote this playbook? Who is the target audience? Does it make any specific assumptions about the data set? If so, do those match your data, or will you have to adapt the playbook?

The playbook was created by GuardSight, a U.S.-based cybersecurity company specializing in managed detection and response (MDR), cybersecurity operations, and cyber incident response. It follows their CIRT (Computer Incident Response Team) methodology and is built on the PICERL model (Preparation, Identification, Containment, Eradication, Recovery, Lessons). The target audience would be Security Operations Center Analysts, Cybersecurity Engineers, or basically, anyone involved in defensive security. It does not assume any specific dataset or toolset. It includes DoS attacks and flow-based features which can be mapped to the playbook's identification and containment steps.

**Tools we Plan to Use:** Based on your dataset and playbook, what blue-team tools from this course will you use to analyze the incident? (MINIMUM of 2)

**Tool #1:** Splunk

**Tool #2:** VirusTotal

**Tool #3:** Abuseipdb

**Tool # 4:** Catayst

Answer each of the key aspect questions (Required)

**Instructions:** For each of the key aspects below, include a few sentences explaining how your project is demonstrating that aspect. Please include at least one specific example.



For a full definition of each of the key aspects, please view the Data Dig Project page on the Course Portal.

Monitoring Sources	
How it relates to our project:	<p>We had two zip files labeled Machine Learning and Generative Labeling which we then unzipped to download the Friday-WorkingHours-Afternoon-DDos.pcap_ISCX.csv which we entered into Splunk to monitor network logs. Network logs let us identify high volume traffic from many IPs toward a single destination – a typical DDoS pattern. We filtered based on "Label=DDoS" and "Destination IP"="192.168.10.50". Network logs were the most important data source for identifying this DDoS incident. These logs include detailed packet-level metadata such as:</p> <ul style="list-style-type: none"><li>• Source IP (e.g., the attacker)</li><li>• Destination IP (target of the attack)</li><li>• Destination Port (e.g., port 80, 443)</li></ul>

- Protocol (TCP, UDP, ICMP)
- Flow Duration and Timestamps

Example(s):

- In Splunk, we ran a search filtering by Label=DDoS, Source IP="\*" and Destination IP="". This revealed consistent attack behavior originating from a single source IP (172.16.0.1) targeting HTTP services on port 80.

New Search Save As Create Table View Close

```
1 index=main source="Friday-WorkingHours-Afternoon-DDos.pcap_ISCX.csv" Label=*
2 | stats count by Label,
```

✓ 451,490 events (before 8/2/25 9:45:16.000 PM) No Event Sampling

Events Patterns **Statistics (2)** Visualization

100 Per Page Format Preview

Label	count
BENIGN	195436
DDoS	256054

- **index=main**  
**source="Friday-WorkingHours-Afternoon-DDos.pcap\_ISCX.csv" Label=\***  
**| stats count by Label**
- This will output how many events are associated with each label (BENIGN, DDoS).

New Search Save As Create Table View Close

```
1 index=main source="Friday-WorkingHours-Afternoon-DDos.pcap_ISCX.csv" Label="DDoS"
2 | where isnotnull('Source IP') AND isnotnull('Destination IP') AND isnotnull('Destination Port')
3 | stats count by "Source IP", "Destination IP", "Destination Port"
4 | where count > 1
```

✓ 128,027 events (before 8/2/25 9:47:33.000 PM) No Event Sampling

Events Patterns **Statistics (1)** Visualization

100 Per Page Format Preview

Source IP	Destination IP	Destination Port	count
172.16.0.1	192.168.10.50	80	128024

- **index=main**  
**source="Friday-WorkingHours-Afternoon-DDos.pcap\_ISCX.csv" Label="DDoS"**  
**| where isnotnull('Source IP') AND isnotnull('Destination IP') AND isnotnull('Destination Port')**  
**| stats count by "Source IP", "Destination IP", "Destination Port"**  
**| where count > 1**
- This helped us spot repeated patterns of DDoS traffic going to the same IP/port from the same source
- By analyzing fields like Protocol and Flow Duration, we detected abnormally high traffic volumes, confirming the severity of the activity and validating its classification as

a DDoS attack.

New Search Save As Create T

```
1 index=main source="Friday-WorkingHours-Afternoon-DDos.pcap_ISCX.csv" Label="DDoS"
2 | where isnull('Source IP') AND isnull('Destination IP') AND isnull('Destination Port')
3 | stats count,
4   values(Protocol) as Protocols,
5   avg('Flow Duration') as Avg_Flow_Duration
6   by "Source IP", "Destination IP", "Destination Port"
7 | sort - count
```

✓ 128,027 events (before 8/2/25 10:37:45.000 PM) No Event Sampling

Events Patterns Statistics (4) Visualization

100 Per Page Format Preview

Source IP	Destination IP	Destination Port	count	Protocols	Avg_Flow_Duration
172.16.8.1	192.168.18.58	88	128024	6	
192.168.18.58	172.16.8.1	27836	1	6	
192.168.18.58	172.16.8.1	64869	1	6	
192.168.18.58	172.16.8.1	64873	1	6	

- **index=main**  
**source="Friday-WorkingHours-Afternoon-DDos.pcap\_ISCX.csv" Label="DDoS"**  
**| where isnull('Source IP') AND isnull('Destination IP') AND isnull('Destination Port')**  
**| stats count,**  
**values(Protocol) as Protocols,**  
**avg('Flow Duration') as Avg\_Flow\_Duration**  
**by "Source IP", "Destination IP", "Destination Port"**  
**| sort - count**
- This query groups DDoS-labeled flows by source IP, destination IP, and port, counting occurrences and averaging flow duration. It also lists the protocols used, helping identify patterns of suspicious high volume traffic.
- By organizing this data chronologically (`_time`), we were able to *track repeated attack attempts over a 5-year span*, which simulated what a real-world persistent threat might look like in a production environment.
- **index=main**  
**source="Friday-WorkingHours-Afternoon-DDos.pcap\_ISCX.csv" Label="DDoS"**  
**| where isnull('Source IP') AND isnull('Destination IP') AND isnull('Destination Port')**  
**| bucket \_time span=1d**  
**| stats count, values(Protocol) as Protocols, avg('Flow Duration') as Avg\_Flow\_Duration**  
**by \_time, "Source IP", "Destination IP", "Destination Port"**  
**| sort \_time**

splunkenterprise									
Search   Analytics   Datasets   Reports   Alerts   Dashboards									
New Search									
<pre>1 info-main source="Friday-Workington-Afternoon-Flas_prog_1931.csv" Label="00d" 2   where isnull(Source IP) AND isnull(Destination IP) AND isnull(Destination Port) 3   bucket _time span=1d 4   stats count, values(Protocol) as Protocols, avg(Flow Duration) as Avg_Flow_Duration 5   by _time, "Source IP", "Destination IP", "Destination Port" 6   sort _time</pre>									
✓ 128,027 events (before 8/2/25 10:48:48.000 PM)   No Event Sampling ▼   Job ▼									
Events   Patterns   Statistics (9)   Visualization									
_time s	Source IP s	✓	Destination IP s	✓	Destination Port s	✓	count s	✓	Protocols s
2020-02-13	172.16.0.1		192.168.10.50		80		531		
2020-03-21	172.16.0.1		192.168.10.50		80		106		
2020-04-28	172.16.0.1		192.168.10.50		80		566		
2020-06-04	172.16.0.1		192.168.10.50		80		1839		
2020-08-01	172.16.0.1		192.168.10.50		80		1		
2020-09-30	172.16.0.1		192.168.10.50		80		2583		
2020-10-01	172.16.0.1		192.168.10.50		80		1855		
2020-10-12	172.16.0.1		192.168.10.50		80		229		
2020-10-14	172.16.0.1		192.168.10.50		80		1		
2020-10-15	172.16.0.1		192.168.10.50		80		553		
2020-10-21	172.16.0.1		192.168.10.50		80		91		
2020-11-11	172.16.0.1		192.168.10.50		80		3623		
2020-11-15	172.16.0.1		192.168.10.50		80		2012		
2020-11-23	172.16.0.1		192.168.10.50		80		7299		
2020-12-10	172.16.0.1		192.168.10.50		80		2		
2020-12-30	172.16.0.1		192.168.10.50		80		2		
2021-01-04	172.16.0.1		192.168.10.50		80		1238		
2021-04-03	172.16.0.1		192.168.10.50		80		4455		
2021-05-26	172.16.0.1		192.168.10.50		80		5816		
2021-07-18	172.16.0.1		192.168.10.50		80		4147		
2021-08-21	172.16.0.1		192.168.10.50		80		2300		
2021-09-17	172.16.0.1		192.168.10.50		80		251		
2021-09-27	172.16.0.1		192.168.10.50		80		3002		
2021-10-01	172.16.0.1		192.168.10.50		80		2401		
2021-10-07	172.16.0.1		192.168.10.50		80		2021		
2021-10-13	172.16.0.1		192.168.10.50		80		88		
2021-10-18	172.16.0.1		192.168.10.50		80		2045		
2021-10-22	172.16.0.1		192.168.10.50		80		2017		
2021-10-26	172.16.0.1		192.168.10.50		80		2723		
2021-11-04	172.16.0.1		192.168.10.50		80		901		
2021-11-18	172.16.0.1		192.168.10.50		80		1920		
2021-11-27	172.16.0.1		192.168.10.50		80		2		
2021-12-01	172.16.0.1		192.168.10.50		80		986		
2021-12-30	172.16.0.1		192.168.10.50		80		452		
2022-01-14	172.16.0.1		192.168.10.50		80		2620		
2022-02-08	172.16.0.1		192.168.10.50		80		2094		
2022-03-17	172.16.0.1		192.168.10.50		80		1057		
2022-09-09	172.16.0.1		192.168.10.50		80		1660		
2022-10-09	172.16.0.1		192.168.10.50		80		1341		
2022-10-19	172.16.0.1		192.168.10.50		80		1908		
2022-11-05	172.16.0.1		192.168.10.50		80		2709		
2022-11-07	172.16.0.1		192.168.10.50		80		6		
2022-11-11	172.16.0.1		192.168.10.50		80		67		
2022-11-13	192.168.10.50		172.16.0.1		27836		1		
2022-11-13	192.168.10.50		172.16.0.1		64809		1		
2022-11-13	192.168.10.50		172.16.0.1		64873		1		
2022-11-19	172.16.0.1		192.168.10.50		80		4092		
2022-11-20	172.16.0.1		192.168.10.50		80		2437		
2022-12-01	172.16.0.1		192.168.10.50		80		511		
2022-12-04	172.16.0.1		192.168.10.50		80		268		
2023-01-16	172.16.0.1		192.168.10.50		80		2050		
2023-01-26	172.16.0.1		192.168.10.50		80		469		
2023-02-06	172.16.0.1		192.168.10.50		80		734		
2023-03-11	172.16.0.1		192.168.10.50		80		2		
2023-04-05	172.16.0.1		192.168.10.50		80		627		
2023-04-18	172.16.0.1		192.168.10.50		80		4118		
2023-06-29	172.16.0.1		192.168.10.50		80		425		
2023-08-09	172.16.0.1		192.168.10.50		80		3186		
2023-10-03	172.16.0.1		192.168.10.50		80		73		
2023-10-06	172.16.0.1		192.168.10.50		80		757		
2023-10-11	172.16.0.1		192.168.10.50		80		1416		
2023-10-20	172.16.0.1		192.168.10.50		80		1778		
2023-11-07	172.16.0.1		192.168.10.50		80		597		
2023-11-15	172.16.0.1		192.168.10.50		80		491		
2023-11-26	172.16.0.1		192.168.10.50		80		546		
2023-11-20	172.16.0.1		192.168.10.50		80		1304		
2023-12-06	172.16.0.1		192.168.10.50		80		688		
100 Per Page ▼   Format   Preview ▼									
_time s	Source IP s	✓	Destination IP s	✓	Destination Port s	✓	count s	✓	Protocols s
2024-06-12	172.16.0.1		192.168.10.50		80		432		
2024-07-28	172.16.0.1		192.168.10.50		80		2053		
2024-10-15	172.16.0.1		192.168.10.50		80		518		
2024-10-22	172.16.0.1		192.168.10.50		80		5328		
2024-10-23	172.16.0.1		192.168.10.50		80		4154		
2024-10-27	172.16.0.1		192.168.10.50		80		1306		
2024-11-02	172.16.0.1		192.168.10.50		80		6		
2024-11-04	172.16.0.1		192.168.10.50		80		1592		
2024-11-15	172.16.0.1		192.168.10.50		80		2577		
2024-11-18	172.16.0.1		192.168.10.50		80		8		
2024-11-20	172.16.0.1		192.168.10.50		80		802		
2024-11-20	172.16.0.1		192.168.10.50		80		127		
2024-12-05	172.16.0.1		192.168.10.50		80		83		
2024-12-29	172.16.0.1		192.168.10.50		80		822		
2025-01-11	172.16.0.1		192.168.10.50		80		18		
2025-01-19	172.16.0.1		192.168.10.50		80		1009		
2025-01-20	172.16.0.1		192.168.10.50		80		4		
2025-02-07	172.16.0.1		192.168.10.50		80		4539		
2025-03-07	172.16.0.1		192.168.10.50		80		485		
2025-03-27	172.16.0.1		192.168.10.50		80		702		
2025-04-02	172.16.0.1		192.168.10.50		80		1		
2025-05-24	172.16.0.1		192.168.10.50		80		583		

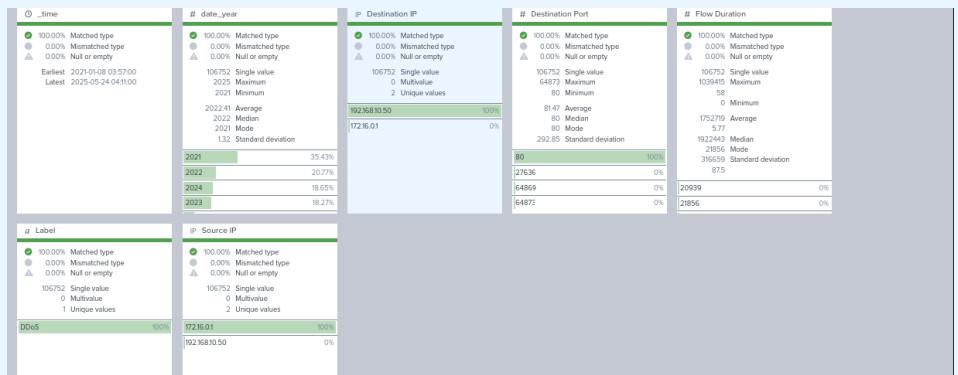
Identified Assets	
How it relates to our project:	In our project, we looked at a real DDoS attack using Splunk. The data shows us which systems were involved both the attackers and the victim. Understanding this helps us see what was affected, where weaknesses might be, and how tools like Splunk help spot and track these kinds of cyber attacks.
Example(s):	<p><b>Target Asset:</b></p> <ul style="list-style-type: none"> <li> <b>192.168.10.50</b>            This internal server is consistently targeted on port 80 (HTTP), making it the primary victim of the DDoS attack.           <ul style="list-style-type: none"> <li><i>Why it matters:</i> This server likely runs a web application, and it couldn't handle the huge amount of requests. It may need better protection against heavy traffic like DDoS attacks.</li> </ul> </li> </ul> <p><b>Attacker Assets:</b></p> <ul style="list-style-type: none"> <li> <b>Source IPs:</b> Mostly <b>172.16.0.1</b>, repeated many times in the dataset as the attacker's IP.           <ul style="list-style-type: none"> <li><i>Why it matters:</i> It was the machine (or one of many) sending the flood of traffic. Even one attacker can cause serious problems if not blocked.</li> </ul> </li> </ul> <p><b>Network Protocols &amp; Ports:</b></p> <ul style="list-style-type: none"> <li> <b>Protocol 6 (TCP)</b> with <b>Destination Port 80</b> is dominant, confirming that the attack leveraged standard web traffic to mask malicious behavior.           <ul style="list-style-type: none"> <li><i>Why it matters:</i> Shows potential weakness in perimeter defenses, where standard ports are often less scrutinized, making it easier for attackers</li> </ul> </li> </ul>



to slip DDoS traffic past filters.

### Systems & Applications Potentially Affected:


- Any **web application** or **HTTP service** running on 192.168.10.50.
- **Firewall or IDS systems**, if not properly tuned, could have failed to detect or block this attack.
- **Network infrastructure (switches, routers)** could experience performance degradation due to excessive traffic volume.



### Impact Analysis and Triage

How it relates to our project:

- **This project analyzes a labeled DDoS dataset (CIC-IDS2017) to show SOC detection, impact assessment, and triage. Using Splunk to query Friday-WorkingHours-Afternoon-DDos.pcap\_ISCX.csv and filter Label=DDoS for Destination IP=192.168.10.50, we discovered a high-volume disruption event. The impact analysis illustrates how we judged severity and prioritized mitigation, while triage describes how we scoped the attack and looked for further vulnerabilities.**
- **Impact Analysis — How we determined severity**  
**Data sources & metrics used**
  - Primary: Network flow/packet logs (CSV from CIC-IDS2017).
  - Key fields inspected: Source IP, Destination IP, Destination Port, Protocol, Timestamp, packet/flow

	<p>counts, flow duration.</p> <ul style="list-style-type: none"><li>● <b>Severity criteria and findings</b><ul style="list-style-type: none"><li>- <b>Event volume:</b> ~128,000 DDoS events identified — indicates <i>high</i> attack intensity.</li><li>- <b>Sustained spikes on timechart:</b> Repeated high counts per minute → indicates sustained service impact (not a short burst).</li><li>- <b>Single attacker IP:</b> Attack originated solely from 172.16.0.1</li><li>- <b>Targeted critical ports (e.g., 80/443):</b> If true, this increases business impact (web service disruption).</li><li>- <b>Conclusion: Severity = High</b> because of service availability impact and volume of traffic. Priority: <b>Immediate mitigation</b> to restore availability.</li></ul></li></ul>
Example(s):	<p> <b>Triage — How we scoped the incident and what we found</b></p> <p><b>Triage objectives</b></p> <ul style="list-style-type: none"><li>● Confirm the incident type (DDoS)</li><li>● Determine scope: isolated or widespread Check for signs of lateral movement or further compromise</li><li>● Identify attacker IPs for blocking/containment</li></ul> <p><b>Triage steps performed</b></p> <ul style="list-style-type: none"><li>● Filtered to:  <code>Label=DDoS "Destination IP"="192.168.10.50"</code></li><li>● Identified attacker:  <code>  top "Source IP"</code> → Result: Only <code>`172.16.0.1`</code></li><li>● Time-based analysis: <code>  timechart span=1m count</code></li></ul>

	<p>→ Showed sustained spikes</p> <ul style="list-style-type: none"> <li>• Checked if other systems were targeted:</li> </ul> <pre>index=main source="Friday-WorkingHours-Afternoon-DDos.pcap_ISCX.csv" Label=DDoS   stats count by "Destination IP"   sort - count</pre> <p>→ <b>Only 192.168.10.50</b> was affected</p> <ul style="list-style-type: none"> <li>• Hypothetical log analysis (not available in dataset): check system logs on 172.16.0.1 and 192.168.10.50 for performance issues, service crashes, or signs of compromise</li> <li>• Threat intelligence lookup (optional in a real-world SOC): Verify if 172.16.0.1 is internal and trusted or compromised</li> </ul> <p><b>Triage results summary</b></p> <ul style="list-style-type: none"> <li>• <b>Scope:</b> Single attacker (172.16.0.1) targeting single destination (192.168.10.50) <b>Attack type:</b> Single-source DDoS — continuous, high-traffic flood</li> <li>• <b>No lateral movement</b> or multiple internal hosts involved</li> <li>• <b>Key indicators captured:</b> Attacker IP, timeline, protocol/port usage</li> </ul>
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Threat Intelligence	
How it relates to our project:	<p>If you came across any relevant threat intelligence during your analysis, we will be sure to discuss this in your presentation. This might include information about the threat actors involved in the incident, the tactics, techniques, and procedures (TTPs) used, and any indicators of compromise (IOCs) that were identified.</p> <ul style="list-style-type: none"> <li>-Who owns the destination IP address?</li> <li>-targeting private company, the attacker used someone within the company</li> </ul>

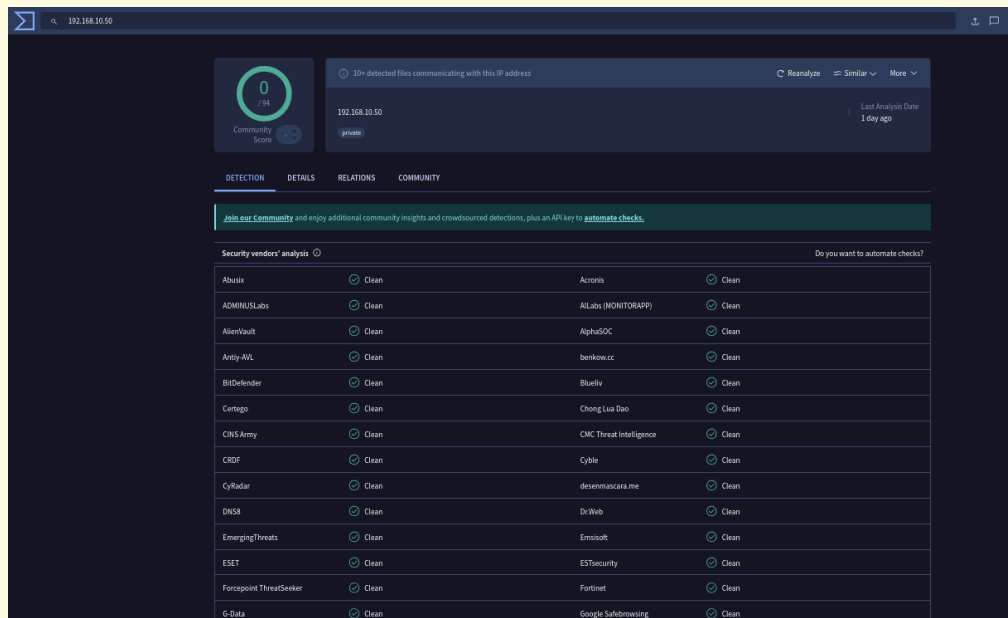
	<p>-VirusTotal and the other one</p> <p>destination ip - companies private ip address source ip - copying the company's ip address but covering it up</p>		
Example(s):			
	<b>Type</b>	<b>Example</b>	<b>Why it matters</b>
	Source IP	172.16.0.1	Repeated source of attack traffic
	Destination IP	192.168.10.50	Targeted internal server (likely hosting a web service)
	Destination Port	80	HTTP — often abused in DDoS due to open access
	Protocol	6 (TCP)	Used to mimic legitimate HTTP traffic
	Flow patterns	High counts, short duration	Indicates flood behavior (common in DDoS attacks)
	<b>Tactic</b>	<b>Technique</b>	<b>Details from Dataset</b>
	Initial Access	External Remote Services (T1133)	Traffic flooded from external IP toward open HTTP port
	Impact	Network Denial of Service (T1498.001)	Repeated, large-volume TCP traffic over port 80
	Evasion	Abuse of Legitimate Protocol (T1071.001)	Using HTTP/TCP to disguise the attack as regular web traffic
	From our analysis, we extracted key indicators like the attacker's IP address 172.16.0.1, the targeted server 192.168.10.50, and the fact that the attack used		

standard HTTP traffic on port 80. These are known as indicators of compromise and help security teams write firewall or IDS rules to block or flag such activity.

In terms of attacker behavior, or what's known as TTPs, we observed techniques commonly associated with denial-of-service campaigns.

In a real-world SOC scenario, this behavior would resemble a botnet-driven DDoS, where compromised machines (often part of a botnet like Mirai) flood a victim server. While we only saw one IP in this dataset, real DDoS attacks often involve hundreds or thousands of IPs coordinated across the globe.

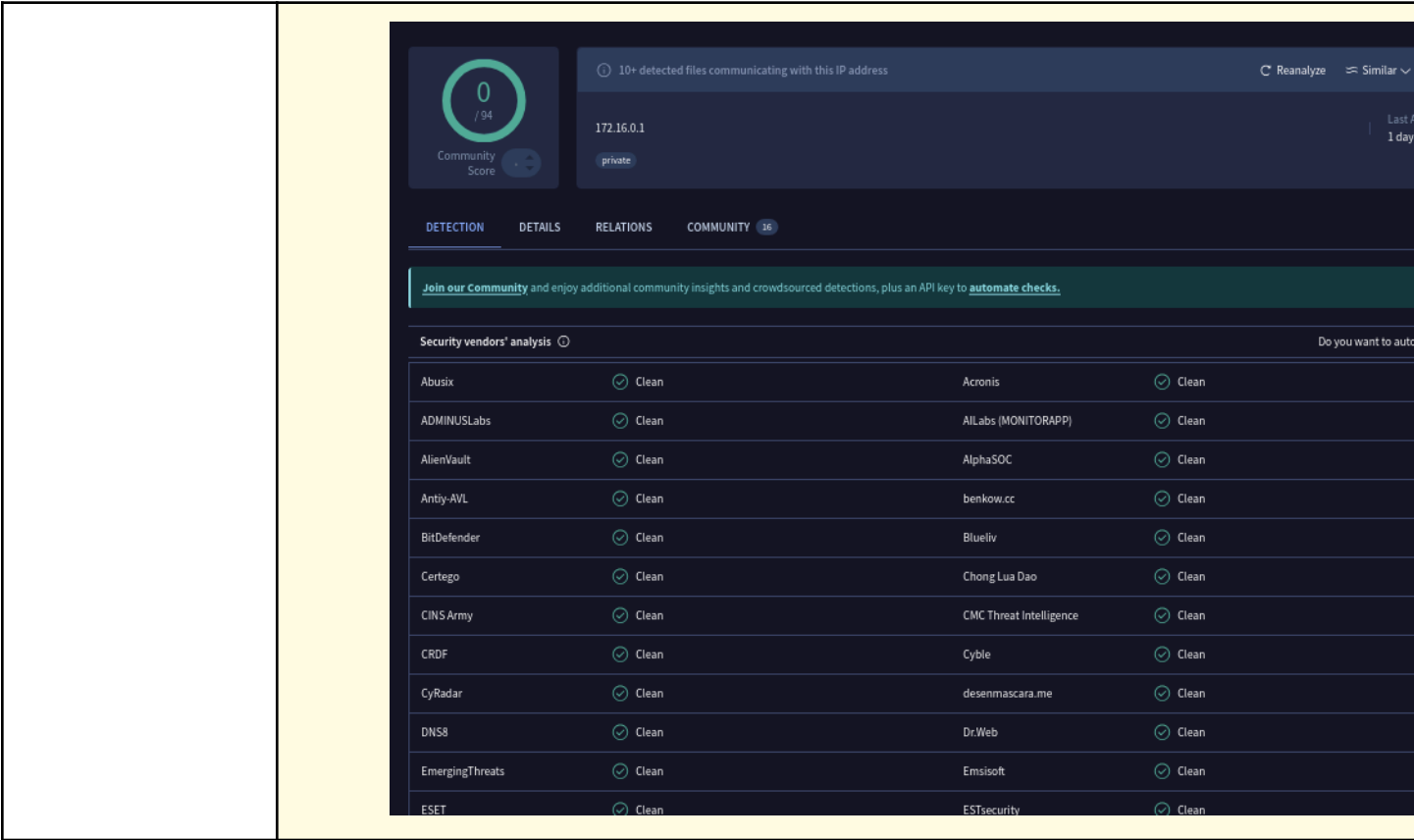
- Private IP **192.168.10.50**



The screenshot shows the VirusShare analysis page for the IP address 192.168.10.50. The page is dark-themed and displays various analysis results. At the top, there's a search bar with the IP address entered. Below the search bar, there's a summary section showing a 'Community Score' of 0/94 and a note that '10+ detected files communicating with this IP address'. The page is divided into tabs: DETECTION, DETAILS, RELATIONS, and COMMUNITY. The DETECTION tab is active, showing a table of security vendors' analysis results. The table has two columns for vendor names and their respective analysis results, all of which are 'Clean'. A 'Do you want to automate checks?' link is visible on the right side of the table.

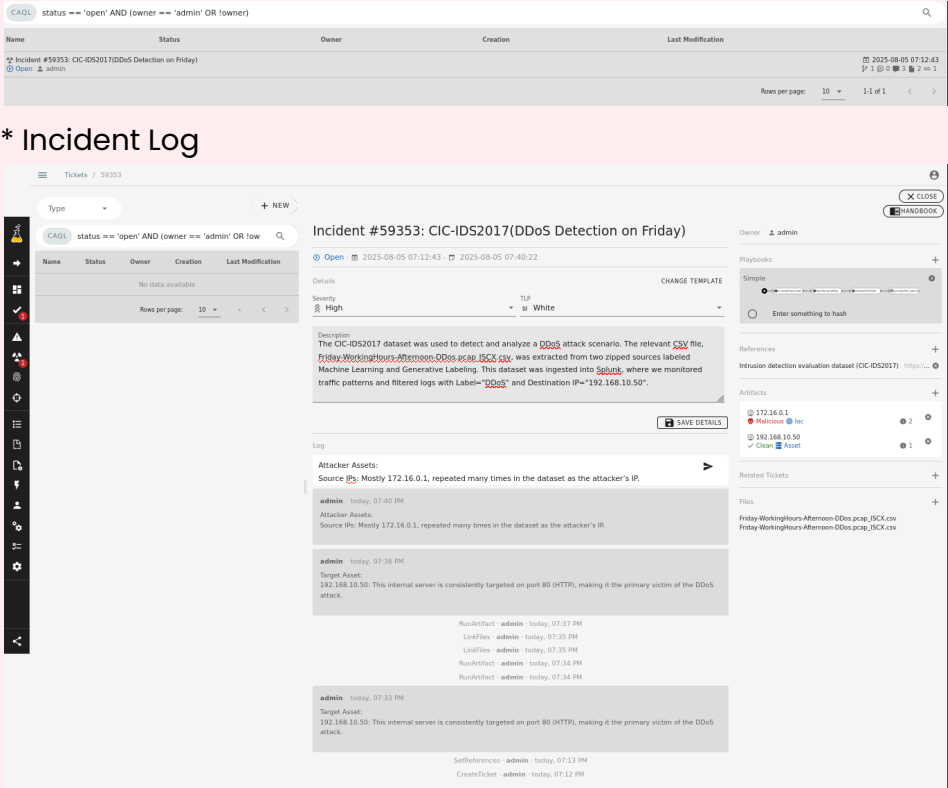
Security vendors' analysis	
Abusix	Clean
ADMINUSLabs	Clean
AlienVault	Clean
Antiy-AVL	Clean
BitDefender	Clean
Carango	Clean
CINIS Army	Clean
CIRIP	Clean
CyRadar	Clean
DNSA	Clean
EmergingThreats	Clean
ESET	Clean
Forcepoint ThreatSeeker	Clean
G-Data	Clean
Acronis	Clean
ALabs (MONITORAPP)	Clean
AlphaSOC	Clean
benkow.cc	Clean
Blueliv	Clean
Chong Lua Dao	Clean
CMC Threat Intelligence	Clean
Cyble	Clean
desenhamcara.me	Clean
Dr.Web	Clean
Emissoft	Clean
ESetSecurity	Clean
Fortinet	Clean
Google Safebrowsing	Clean

- Private IP **172.16.0.1**



Recommended Remediation	
How it relates to our project:	Based on what we found, we came up with the steps below to help protect the server and reduce the chance of this happening again in the future.
Example(s):	<div><b>Block Attacker IP (172.16.0.1)</b> Use a firewall rule to block the IP that generated the DDoS traffic.</div> <div><b>Limit HTTP Requests per IP</b> Set rules to block IPs making more than 100 HTTP requests per minute to prevent flooding.</div> <div><b>Deploy Cloudflare WAF</b> Add a Web Application Firewall in front of 192.168.10.50 to block suspicious HTTP traffic.</div> <div><b>Set Up Splunk Alert for DDoS Pattern</b> Set up Splunk alerts to notify the team when a single IP sends over 100 requests per minute or when there's a sudden spike in</div>

	<p>HTTP traffic.</p> <p><b>Force HTTPS and Disable HTTP on Port 80</b></p> <p>Force HTTPS by installing an SSL certificate, redirecting port 80 traffic to 443, and disabling HTTP if it's not needed.</p> <p><b>Update Devices</b></p> <p>Update the OS, firewall, and web server software on 192.168.10.50, and disable any open ports or services not being used.</p>
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<b>Case Management System</b> (and screenshots)	
How it relates to our project:	We used the Catalyst case management system to log each step in the incident response process. This made it easier to track analysis, SOCs, containment actions, and recovery.
Example(s):	<p><b>** CATALYST**</b></p>  <p>The screenshot displays the Catalyst Case Management System interface. At the top, there's a search bar with the query 'status == 'open' AND (owner == 'admin' OR lowner)'. Below this is a table listing tickets. The first ticket is Incident #59353, titled 'CIC-IDS2017DDoS Detection on Friday', created on 2025-08-05 at 07:12:43. The interface also shows a sidebar with navigation icons and a top navigation bar with 'Incident Log' and 'Incident #59353: CIC-IDS2017(DDoS Detection on Friday)'. The main content area provides details for this incident, including a description, severity (High), and a log of actions taken by 'admin' on 2025-08-05. The log shows that the incident was identified as a DDoS attack on 192.168.10.50, which was targeted on port 80 (HTTP).</p> <p><b>* OPEN</b></p>

Tickets / 59353

Type

+ NEW

CAQL status == 'open' AND (owner == 'admin' OR low)

Name	Status	Owner	Creation	Last Modification
No data available				

Rows per page: 10

Incident #59353: CIC-IDS2017(DDoS Detection on Friday)

Closed

2025-06-05 07:12:43

2025-06-05 07:41:45

Details

CHANGE TEMPLATE

Severity High

TLP White

Overview

The CIC-IDS2017 dataset was used to detect and analyze a DDoS attack scenario. The relevant CSV file, [Friday-Workinghours-Afternoon-DDos.pcap\\_ISCX.csv](#), was extracted from two zipped sources labeled Machine Learning and Generative Labeling. This dataset was ingested into [Splunk](#), where we monitored traffic patterns and filtered logs with Label="DDoS" and Destination IP="192.168.10.50".

SAVE DETAILS

Log

Attacker Assets:

Source IP: Mostly 172.16.0.1, repeated many times in the dataset as the attacker's IP.

admin today, 07:40 PM

Attacker Assets:

Source IP: Mostly 172.16.0.1, repeated many times in the dataset as the attacker's IP.

admin today, 07:38 PM

Target Asset:

192.168.10.50: This internal server is consistently targeted on port 80 (HTTP), making it the primary victim of the DDoS attack.

RunVulnFact - admin today, 07:37 PM

LoadFlux - admin today, 07:35 PM

LoadFlux - admin today, 07:35 PM

RunVulnFact - admin today, 07:34 PM

RunVulnFact - admin today, 07:34 PM

admin today, 07:33 PM

Target Asset:

192.168.10.50: This internal server is consistently targeted on port 80 (HTTP), making it the primary victim of the DDoS attack.

SetReferences - admin today, 07:33 PM

CreateTicket - admin today, 07:12 PM

Playbooks

Simple

Enter something to hash

References

Intrusion detection evaluation dataset (CIC-IDS2017)

Artifacts

172.16.0.1

Malicious

loc

192.168.10.50

Clean

Asset

Related Tickets

Files

Friday-Workinghours-Afternoon-DDos.pcap\_ISCX.csv

Friday-Workinghours-Afternoon-DDos.pcap\_ISCX.csv

\* Closed

Presentation Prep (Required)

**Presentation Plan:** What is your plan for the presentation? Please include a roadmap, flowchart, diagram, or outline.

Things to consider:

- ☐ What will you talk about, and in what order?
- ☐ Who will be talking at what times?
- ☐ What visual-aids will you use?

- Introduction - Dennys 15 sec
- Dataset - Dennys 30 sec
- Hypothesis - Aliya 15 sec
- Playbook - Aliya 30 sec
- Monitoring Sources - Angie 60 sec
- Identified Assets - Angie 60 sec
- Impact Analysis and Triage - Navruz 45 sec
- Threat Intelligence - Dennys 45 sec
- Recommended Remediation - Evgeniia 45 sec
- Case Management System - Angie 45 sec
- Conclusion - Aliya 30 sec
- We will have a PowerPoint presentation



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## Submission Checklist

👉 Check off each of the features you have completed. **You will only be graded on the features you check off.**

### Required Features

- ☒ ~~Select one (or more) open-source Datasets to analyze~~
  - ☒ ~~Data Set Chosen (Name & Link)~~
  - ☒ ~~Data Set Description~~
  - ☒ ~~3 Hypotheses Made~~
- ☒ ~~Select an incident-response playbook to follow~~
  - ☒ ~~Playbook Chosen (Name & Link)~~
  - ☒ ~~Playbook Description~~
  - ☒ ~~2+ Tools Identified~~
- ☒ ~~Answer each of the key aspect questions:~~
  - ☒ ~~Monitoring Sources~~
  - ☒ ~~Identified Assets~~
  - ☒ ~~Impact Analysis and Triage~~
  - ☒ ~~Threat Intelligence~~
  - ☒ ~~Recommended Remediation~~
  - ☒ ~~Case Management System~~
- ☒ ~~Your presentation plan: A roadmap, outline, or diagram~~

### Submit your work!