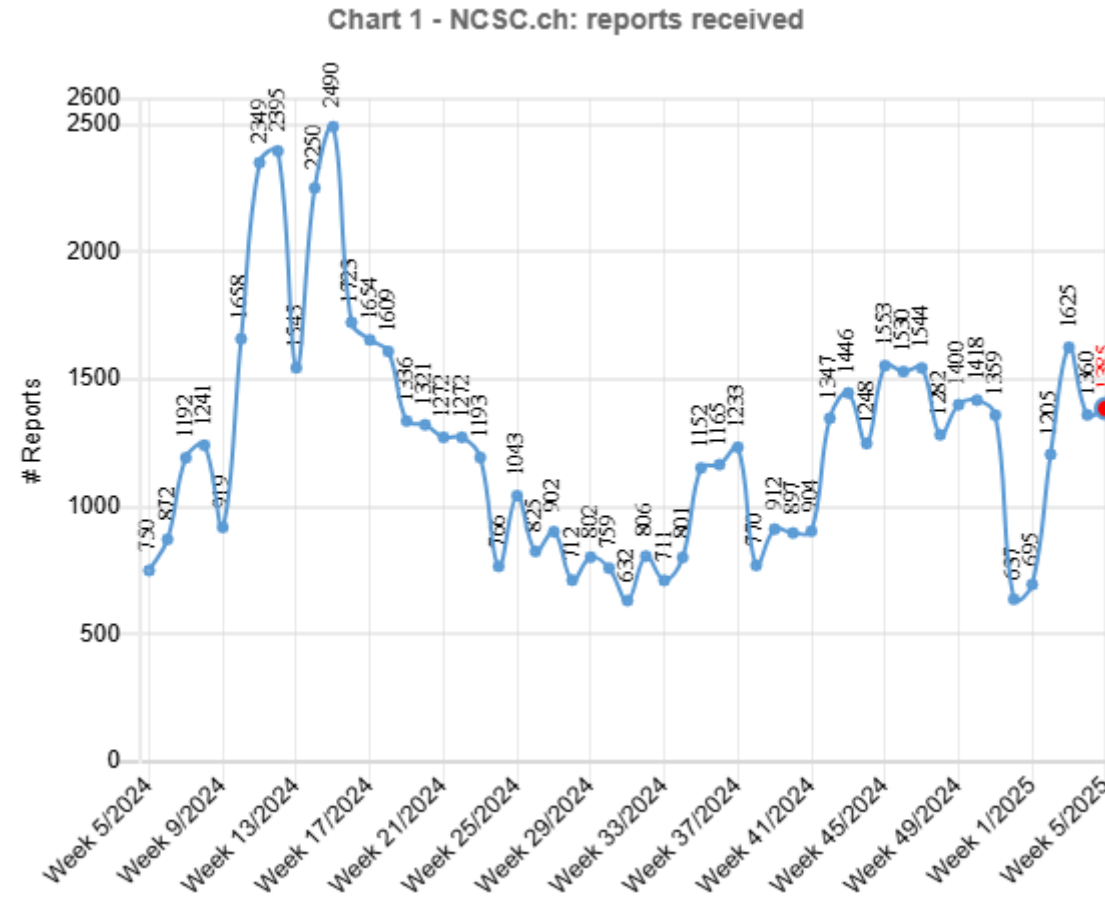


# Towards a fully automated Blue Team at Locked Shields

Roland Meier  
AMLD 2025, February 13, 2025



# Every week, the NCSC receives >1`000 reports about cyber incidents





# Every week, the NCSC receives >1`000 reports about cyber incidents



## E-mails with malware in the name of debt collection agencies and health insurance companies

02.12.2024 - The NCSC is currently receiving numerous reports of e-mails that claim to come from a debt collection agency or a health insurance company. They concern an alleged claim or reminder. Do not click on the link, as this is an attempt to distribute malware to Windows users.



## Update: Even after the conclusion of the high-level conference on peace in Ukraine, the overload attacks on websites of organisations involved continue

17.06.2024 - As expected, the overload attacks continue even after the conclusion of the high-level conference on peace in Ukraine. The websites of the organisations involved in the conference are still being targeted. The National Cyber Security Centre is monitoring the situation and is in contact with the organisations concerned.



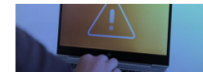
## Critical vulnerability in Palo Alto firewalls

18.04.2024 - The NCSC warns of the security vulnerability in Palo Alto's Next-Generation Firewall (NGFW). These firewalls are mainly used by companies and public authorities. They have a critical vulnerability that is already being exploited by cyber criminals. The attackers exploit the vulnerability to execute commands. The NCSC has already received corresponding reports from organisations in Switzerland. The NCSC recommends installing the security updates as quickly as possible or even reinstalling the NGFW if possible.



## Critical vulnerability in file transfer software «MOVEit»: Apply Patch quickly

02.06.2023 - The file transfer software called «MOVEit», which is mainly used by businesses, has a critical vulnerability that is already being exploited by cybercriminals. The attackers are exploiting the vulnerability to steal files from the file transfer software. The NCSC started to receive corresponding reports from organisations in Switzerland on 1 June. The NCSC recommends applying the security patch as quickly as possible.



## Update: Still over 2,000 unsecured Microsoft Exchange servers in Switzerland

01.12.2022 - Just over a fortnight ago, the NCSC called for the security patches provided by Microsoft to be installed in order to fix the ProxyNotShell vulnerability. Despite the urgency, there are still some operators that have failed to heed this call. Therefore, the NCSC has sent more than 2,000 registered letters to those affected, urging them to act now.




# There are not enough experts in this field



**Forbes**




FORBES > LEADERSHIP > CAREERS


## Nearly 4 Million Cybersecurity Jobs Are Vacant: Here's Why You Should Consider Breaking Into This Sector

**Jack Kelly** Senior Contributor   
*Jack Kelly covers career growth, job market and workplace trends.*

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  Aug 16, 2024, 06:00am EDT



Cybersecurity consistently ranks among the top areas for job growth and demand within the broader ... [\[+\]](#) GETTY

# AI is everywhere...

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... why not in cyber defense?





## Cyber defenders fight hackers in high-tech Estonia war

Attacks on vital systems and fake news

© Fri, Apr 28, 2017, 01:00



Daniel McLaughlin in Tallinn



The attack on the airbase began with a salvo of fake news. “A report appeared saying drones were using nerve gas,” said Lauri Luht, crisis management chief for the cyber security department of Estonia’s information system authority.



## Cyber defenders fight hackers in high-tech Estonia war games

Attacks on vital systems and fake news are all part of Locked Shields exercise

© Fri, Apr 28, 2017, 01:00



Daniel McLaughlin in Tallinn



Locked Shields, now taking place in Estonia involving 20 teams from Europe and the US, is the world's most advanced live-fire cyber defence exercise. Photograph: Daniel McLaughlin



The attack on the airbase began with a salvo of fake news. “A report appeared saying drones were using nerve gas,” said Lauri Luht, crisis management chief for the cyber security department of Estonia’s information system authority.



# Locked Shields is the largest live-fire global cyber defense exercise



Dr. Roland Meier

Picture: NATO CCDCOE



# Locked Shields is the largest live-fire global cyber defense exercise

- Red Team vs. Blue Team exercise
  - Attackers      Defenders
  - 1 Team      1 Team / country
- > 1'000 experts from 30 nations
- > 4'000 systems
- > 2'500 attacks





# Locked Shields is organized by the NATO CCDCoE

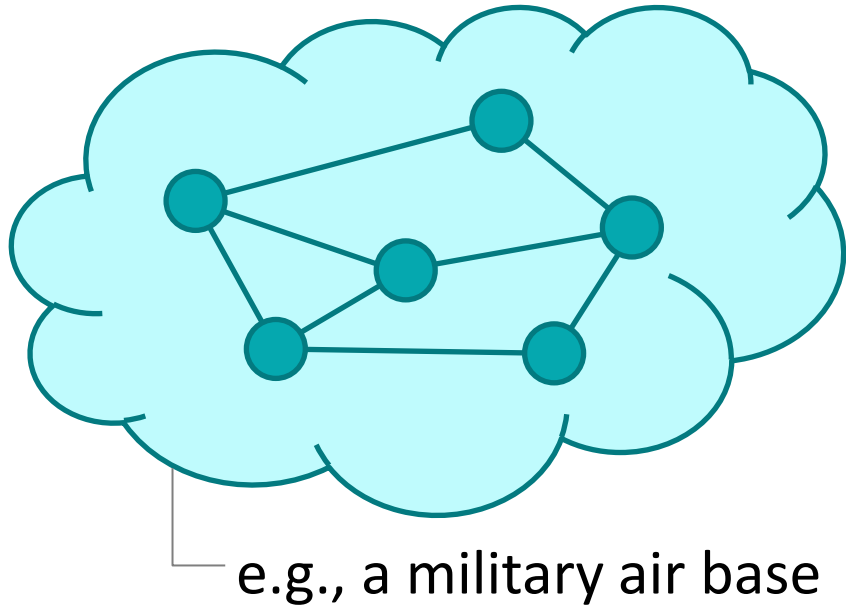


**Sponsoring Nations**

**Contributing Participants**

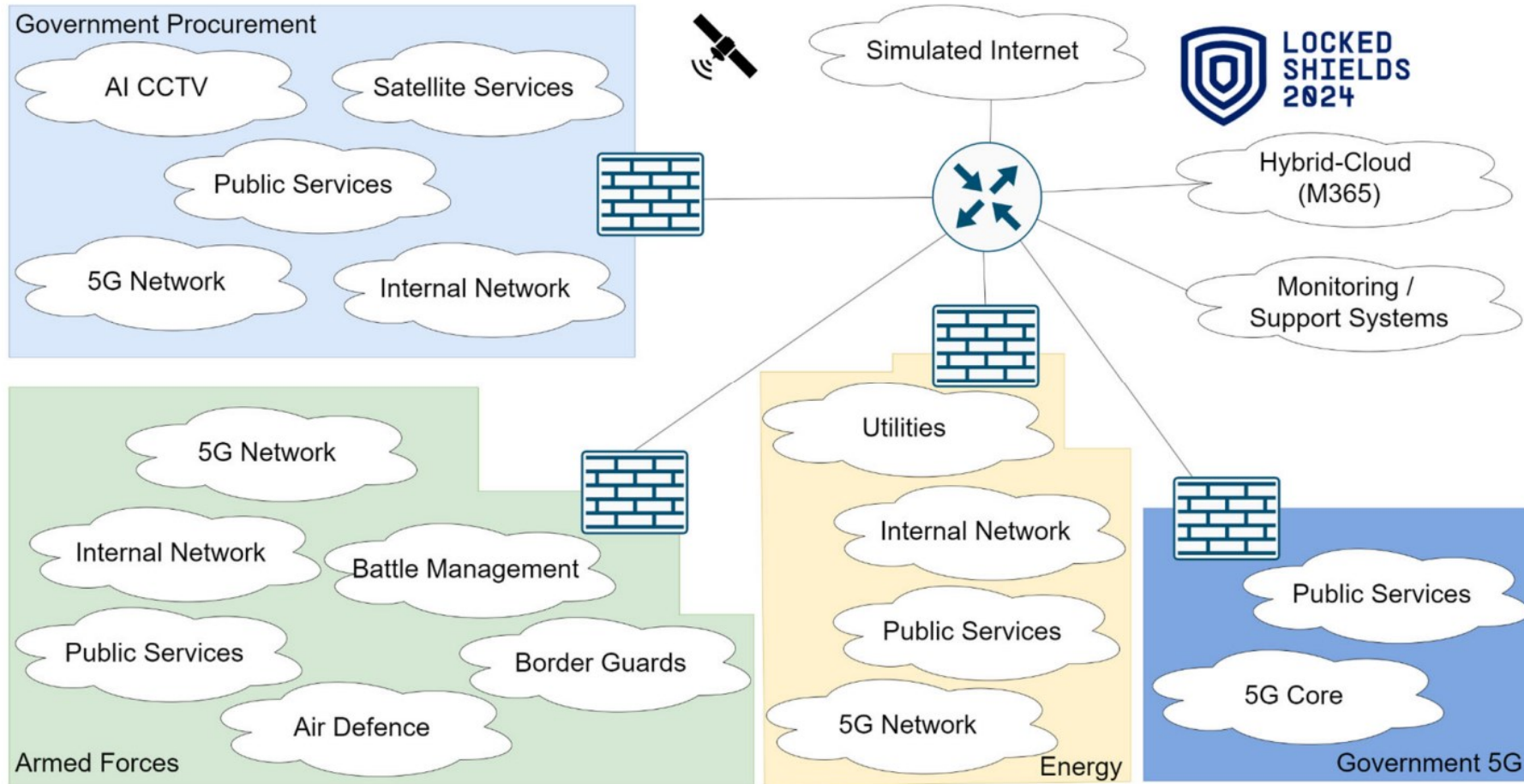


# Each Blue Team is responsible for its network (“Gamenet”)



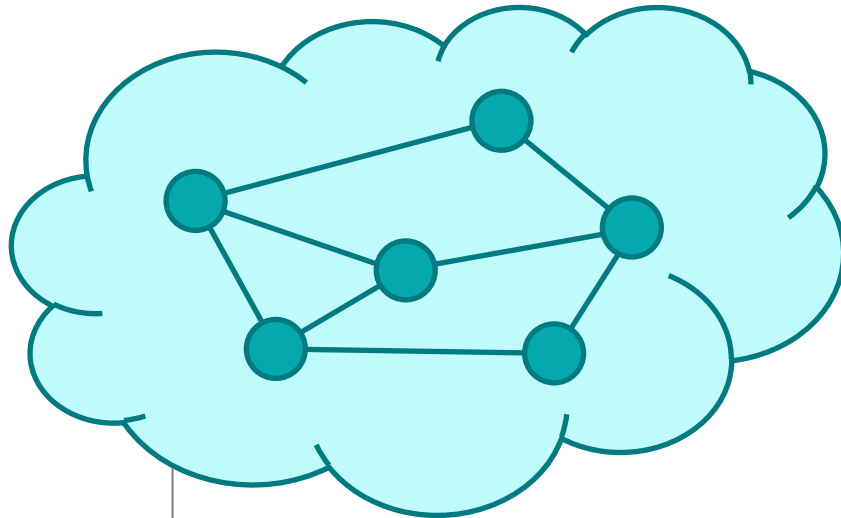


# The Gamenet consists of a large variety of systems





# Each Blue Team is responsible for its network (“Gamenet”)



e.g., a military air base

3 main tasks:

- Perform initial hardening
- Defend against attacks
- Communicate with other teams

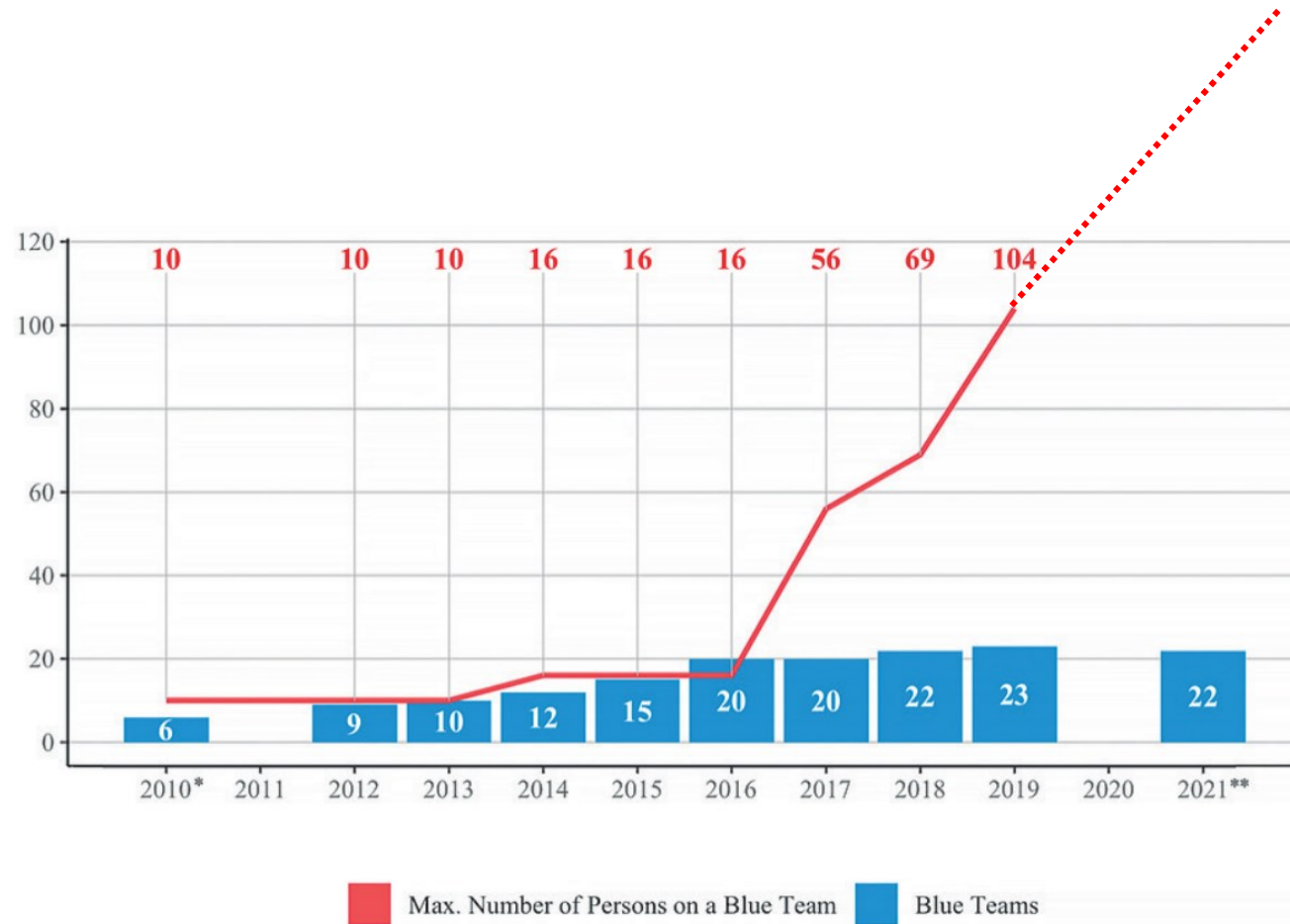


# Besides defending its network, a Blue Team needs to communicate with other teams

User Simulation Team	Read, address and respond to support tickets
Yellow Team	Provide periodic reports
White team	Voice or video calls through the Gamenet
Green Team	Gamenet status, reverting of devices



# The number of people required in a Blue Team continuously increases



[Smeets, Max. "The role of military cyber exercises: A case study of Locked Shields." CyCon 2022]





# Research goals

How can automation / AI help for cyber defense?

*And eventually...*

What would it take to have a fully automated Blue Team  
in a future iteration of Locked Shields?



# The history

100th Anniversary Conference on Cyber Security  
17-18 October 2019, ETH Zurich, Switzerland  
19-20 October 2019, ETH Zurich, Switzerland

## Machine Learning-based Detection of C&C Channels with a Focus on the Locked Shields Cyber Defense Exercise

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**Vincent Lenders**  
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**Abstract:** The diversity of applications and devices in enterprise networks combined with large traffic volumes make it inherently challenging to quickly identify malicious traffic. When incidents occur, emergency response teams often lose precious time in reconfiguring the network topology and configuration before they can focus on malicious activities and digital forensics.

100th Anniversary Conference on Cyber Security  
17-18 October 2019, ETH Zurich, Switzerland  
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## Defeating and Improving Network Flow Classifiers Through Adversarial Machine Learning

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**Abstract:** Recent work has shown that machine learning models can be vulnerable to an adversary crafting targeted inputs designed to cause mispredictions. This is critical in security-related applications such as network intrusion detection systems. While past attacks such as memory or gradient-based attacks are able to efficiently generate adversarial examples, they require potentially large input modifications, which is not effective at defeating network flow classifiers.

In this work, we show that small modifications to the input (e.g. the traffic that the attacker generates) are enough to manipulate the outcome of a classifier. We focus on manually-crafted adversarial examples to defeat tree-ensemble-based network flow classifiers. We develop an attack that builds on a previous attack introduced by Kandaswamy et al. in 2016, which formulates evasion for tree ensembles as a Mixed Integer Linear Program, and which we extend by supporting discrete and categorical features, implementing per feature evasion costs and modeling single feature dependencies. This makes our attack more applicable to the network flow classification problem, which typically uses discrete and independent input features.

We demonstrate our attack on the network flow classifier developed by König et al. in 2019, which was trained to detect command and control (C&C) channels in the



## DOCTORAL THESIS Automating Defences against Cyber Operations in Computer Networks

Mauno Pihelgas  
TALLINN TECHNICAL UNIVERSITY  
TALINN UNIVERSITY OF TECHNOLOGY  
TALINN 2021

## Towards an Active, Autonomous and Intelligent Cyber Defense of Military Systems: the NATO AICA Reference Architecture

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**Abstract:** While the future Global Information Grid (GIG) enables operational autonomy, network defense capabilities are still limited. This thesis presents an active and autonomous cyber defense architecture, which leverages machine learning and artificial intelligence to detect, analyze, and respond to cyber threats. The architecture is designed to be scalable and adaptable to various network environments. It consists of several key components: a threat intelligence module, a network monitoring module, a threat detection module, a threat analysis module, and a threat response module. The architecture is implemented and evaluated using a real-world network environment. The results show that the architecture is able to detect and respond to a wide range of cyber threats, including advanced persistent threats (APTs). The architecture is also able to adapt to changes in the network environment, such as the addition of new hosts and services. The architecture is designed to be scalable and adaptable to various network environments. It consists of several key components: a threat intelligence module, a network monitoring module, a threat detection module, a threat analysis module, and a threat response module. The architecture is implemented and evaluated using a real-world network environment. The results show that the architecture is able to detect and respond to a wide range of cyber threats, including advanced persistent threats (APTs). The architecture is also able to adapt to changes in the network environment, such as the addition of new hosts and services.

100th Anniversary Conference on Cyber Security  
17-18 October 2019, ETH Zurich, Switzerland  
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## Towards an AI-powered Player in Cyber Defence Exercises

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**Abstract:** Cyber attacks are becoming increasingly frequent, sophisticated, and costly. This makes it harder for cyber defense teams to keep up. One way to mitigate this risk is to use artificial intelligence (AI) to automate some of the tasks that cyber defense teams perform. This paper presents an AI-powered player for cyber defense exercises. The player is designed to be able to detect, analyze, and respond to cyber threats. It consists of several key components: a threat intelligence module, a network monitoring module, a threat detection module, a threat analysis module, and a threat response module. The player is implemented and evaluated using a real-world network environment. The results show that the player is able to detect and respond to a wide range of cyber threats, including advanced persistent threats (APTs). The player is also able to adapt to changes in the network environment, such as the addition of new hosts and services.

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17-18 October 2019, ETH Zurich, Switzerland  
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## Towards Generalizing Machine Learning Models to Detect Command and Control Attack Traffic

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**Abstract:** Identifying compromised hosts from network traffic traces has become challenging because benign and malicious traffic is overlapped, and both use the same protocols and ports. Machine learning based anomaly detection models have been proposed to address this challenge by classifying malicious traffic based on network flow features learned from historical patterns. Previous work has shown that such models successfully identify compromised hosts in the same network environment as which they were trained. However, cyber incident response teams often have to look for intrusions in foreign networks, and we have found that learned models often fail to generalize to different network conditions. In this paper, we analyze the root cause of this problem using five network traces collected from different states and teams of Locked Shields, the world's largest live-fire cyber defense exercise. We then explore techniques to make machine learning models generalize better to unknown network environments and reduce their secrecy.

**Keywords:** machine learning, traffic classification, network security, command and control, Locked Shields



## LSPR23: A novel IDS dataset from the largest live-fire cybersecurity exercise

**Alain Hugi**, **Erwin Hiltbrunner**, **Corrado Tridello**, **Alan Schi**, **Mauno Pihelgas**, **Roland Meier**  
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**Abstract:** This paper presents a novel dataset for intrusion detection systems (IDS) derived from the largest live-fire cybersecurity exercise, Locked Shields 2023. The dataset consists of network traffic traces collected from a large number of hosts and services. The dataset is designed to be scalable and adaptable to various network environments. It consists of several key components: a threat intelligence module, a network monitoring module, a threat detection module, a threat analysis module, and a threat response module. The dataset is implemented and evaluated using a real-world network environment. The results show that the dataset is able to detect and respond to a wide range of cyber threats, including advanced persistent threats (APTs). The dataset is also able to adapt to changes in the network environment, such as the addition of new hosts and services.

**1. Introduction**  
With the increasing frequency and complexity of cyber attacks, the need for robust and effective intrusion detection systems (IDS) is becoming more and more urgent. This paper presents a novel dataset for IDS derived from the largest live-fire cybersecurity exercise, Locked Shields 2023. The dataset consists of network traffic traces collected from a large number of hosts and services. The dataset is designed to be scalable and adaptable to various network environments. It consists of several key components: a threat intelligence module, a network monitoring module, a threat detection module, a threat analysis module, and a threat response module. The dataset is implemented and evaluated using a real-world network environment. The results show that the dataset is able to detect and respond to a wide range of cyber threats, including advanced persistent threats (APTs). The dataset is also able to adapt to changes in the network environment, such as the addition of new hosts and services.



## Defeating and Improving Network Flow Classifiers Through Adversarial Machine Learning

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**Abstract:** Recent work has shown that machine learning models can be vulnerable to an adversary crafting targeted inputs designed to cause mispredictions. This is critical in security-related applications such as network intrusion detection systems. While past attacks such as memory or gradient-based attacks are able to efficiently generate adversarial examples, they require potentially large input modifications, which is not effective at defeating network flow classifiers.

In this work, we show that small modifications to the input (e.g. the traffic that the attacker generates) are enough to manipulate the outcome of a classifier. We focus on manually-crafted adversarial examples to defeat tree-ensemble-based network flow classifiers. We develop an attack that builds on a previous attack introduced by Kandaswamy et al. in 2016, which formulates evasion for tree ensembles as a Mixed Integer Linear Program, and which we extend by supporting discrete and categorical features, implementing per feature evasion costs and modeling single feature dependencies. This makes our attack more applicable to the network flow classification problem, which typically uses discrete and independent input features.

We demonstrate our attack on the network flow classifier developed by König et al. in 2019, which was trained to detect command and control (C&C) channels in the



## Defeating and Improving Network Flow Classifiers Through Adversarial Machine Learning

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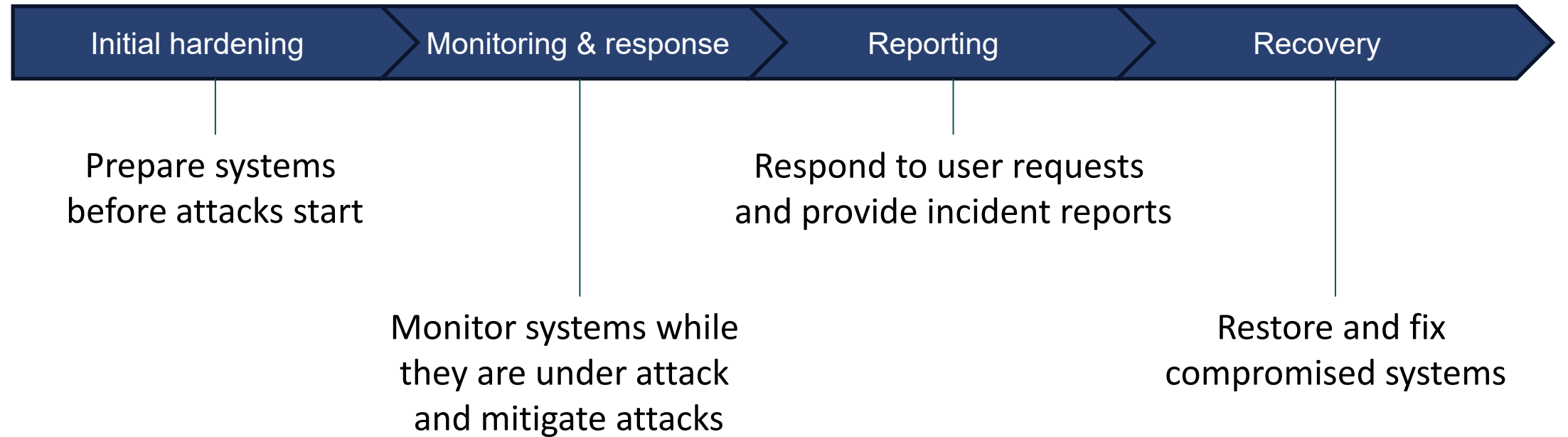
**Abstract:** Recent work has shown that machine learning models can be vulnerable to an adversary crafting targeted inputs designed to cause mispredictions. This is critical in security-related applications such as network intrusion detection systems. While past attacks such as memory or gradient-based attacks are able to efficiently generate adversarial examples, they require potentially large input modifications, which is not effective at defeating network flow classifiers.

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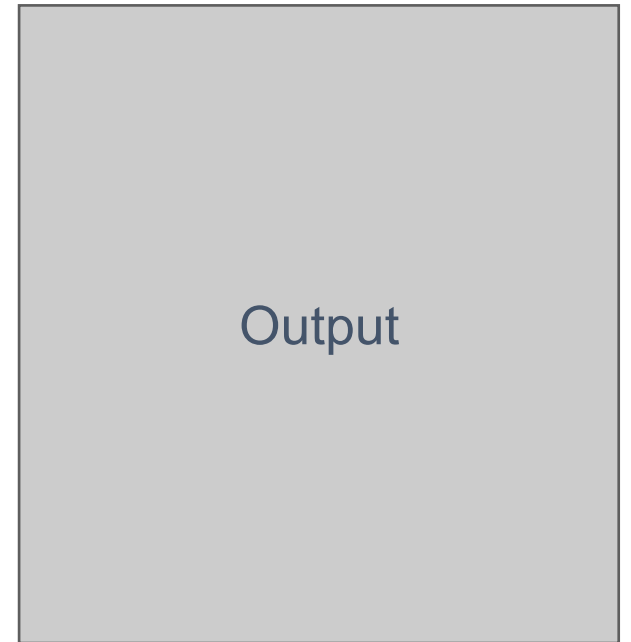
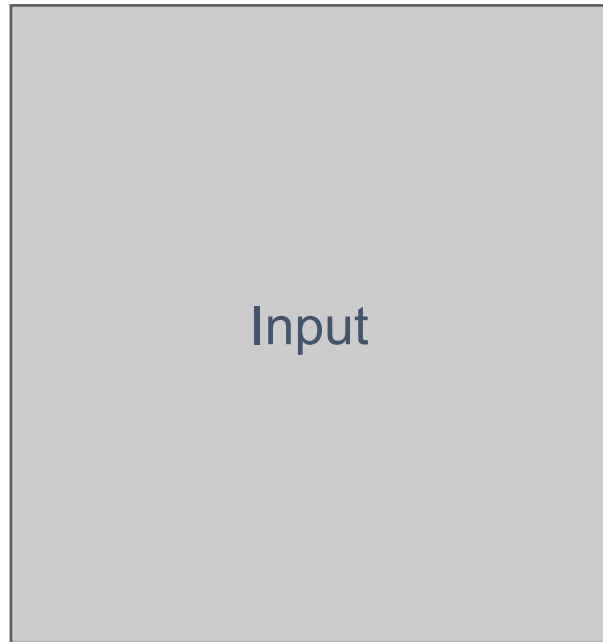


# Four stages in the exercise



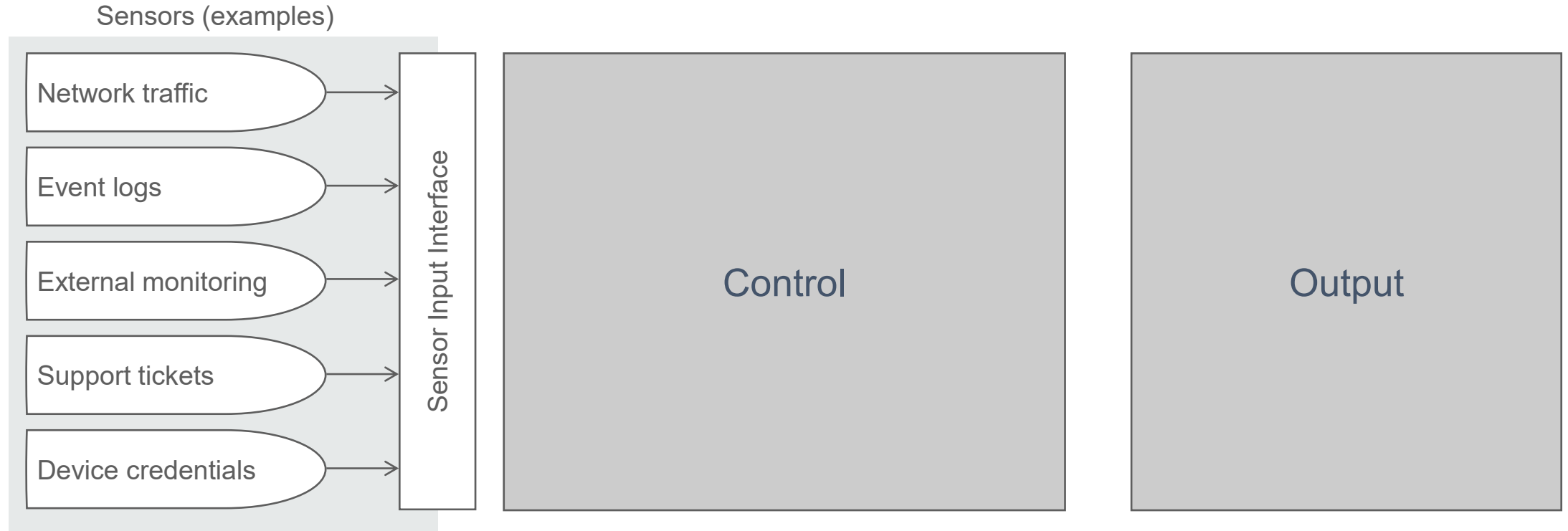


# Automated Blue Team framework overview



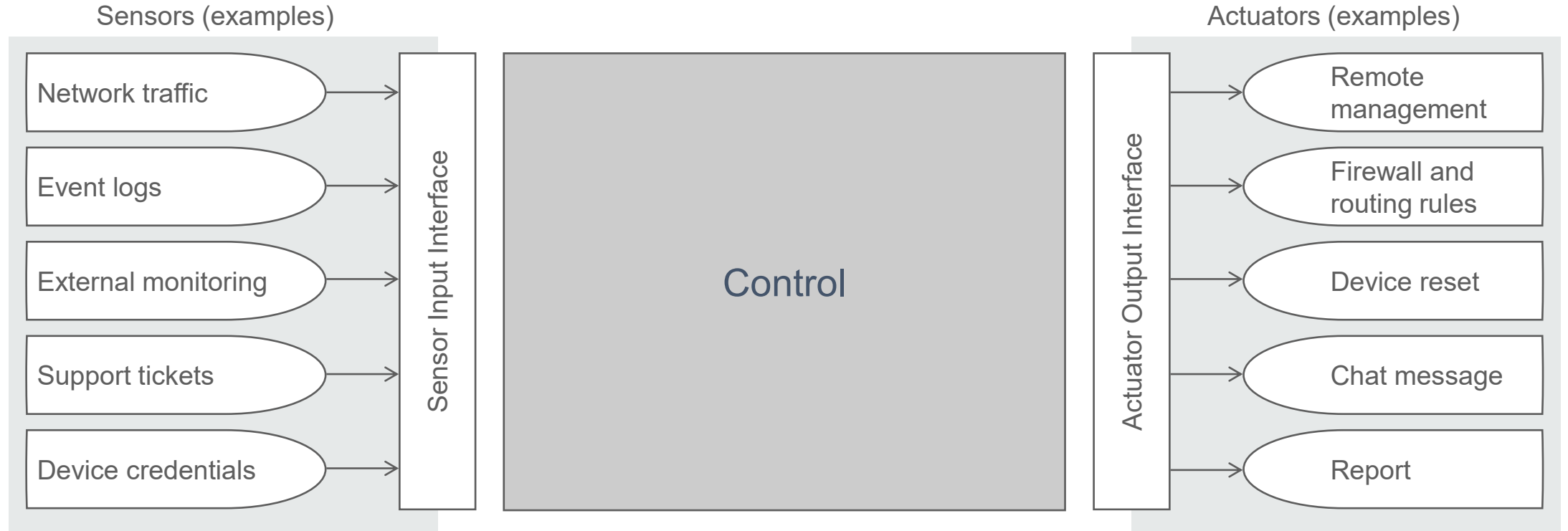


# Automated Blue Team framework overview



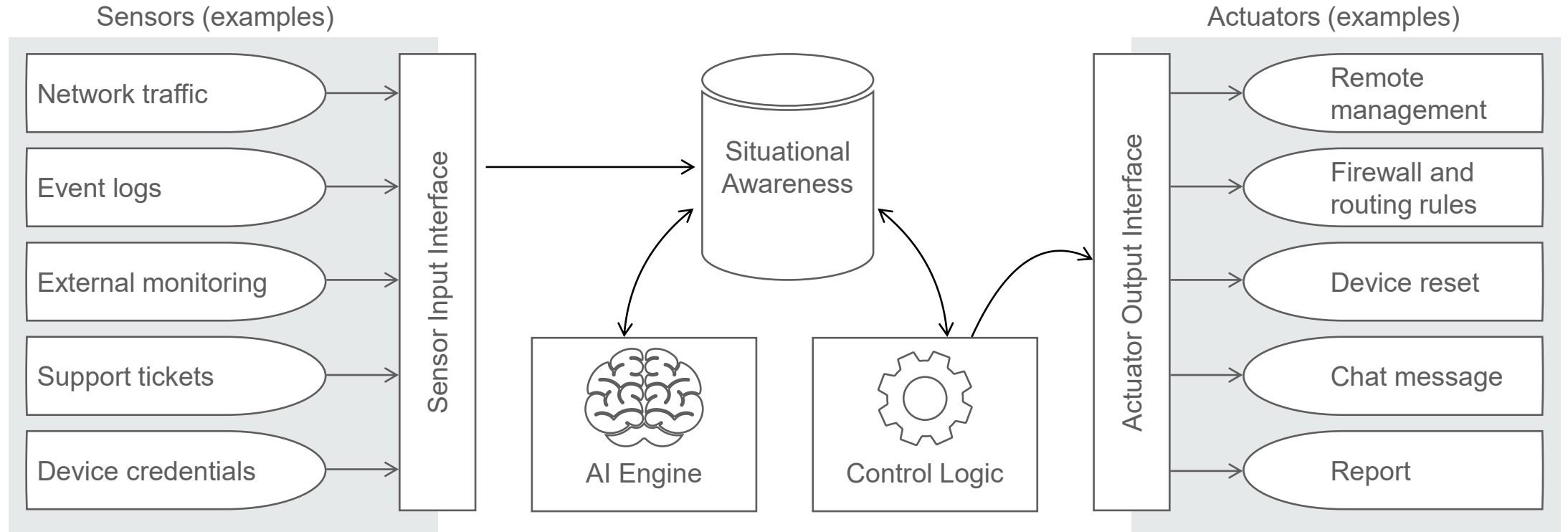


# Automated Blue Team framework overview



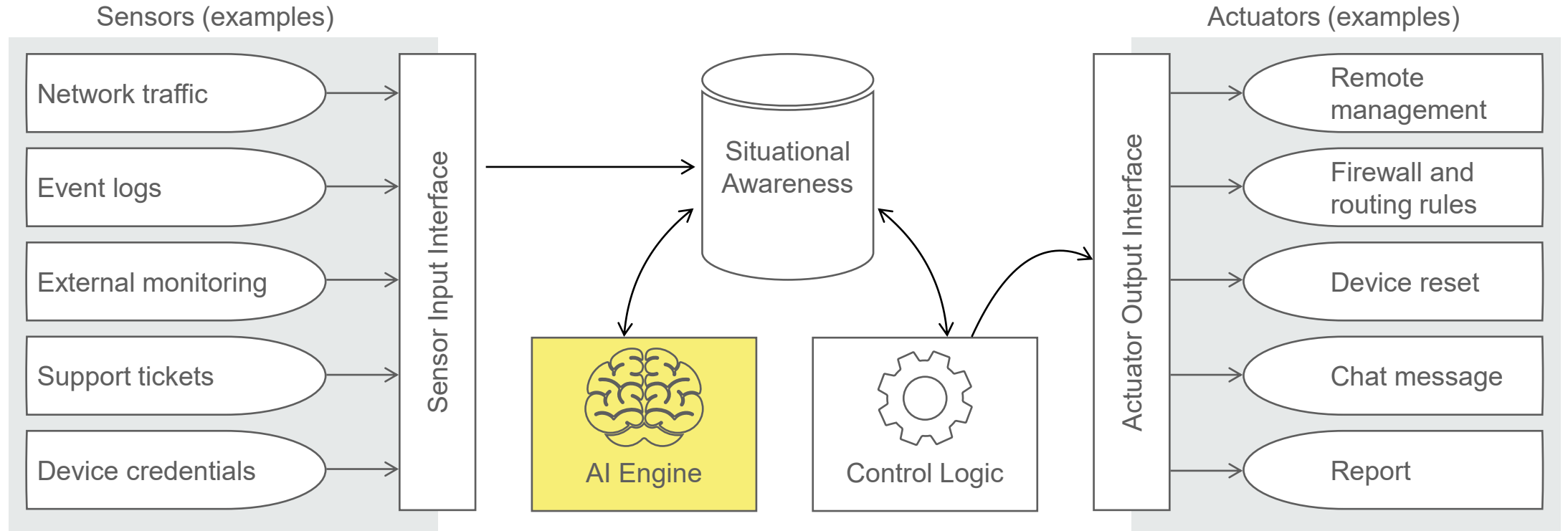


# Automated Blue Team framework overview





# Automated Blue Team framework overview







# Five tasks for AI

Identification / classification

What is it?

Categorisation

What belongs together?

Assessment

What is important?

Recommendation

What should be done?

Prediction

What will happen?



# AI for initial hardening



Identification / classification

Categorisation

Find groups of similar devices

Assessment

Recommendation

Prediction



# AI for monitoring and response



Identification / classification

Detect malicious network traffic

Categorisation

Detect malicious patterns in log files

Assessment

Recommendation

Prediction



# AI for reporting



Identification / classification

Categorisation

Assessment

Recommendation

Prediction

Link support tickets and monitoring alerts

Prioritise support tickets

Formulate response to support tickets

Predict impact on the scoring



# AI for recovery



Identification / classification

Find devices that need to be recovered

Categorisation

Find similar devices as a template for recovery

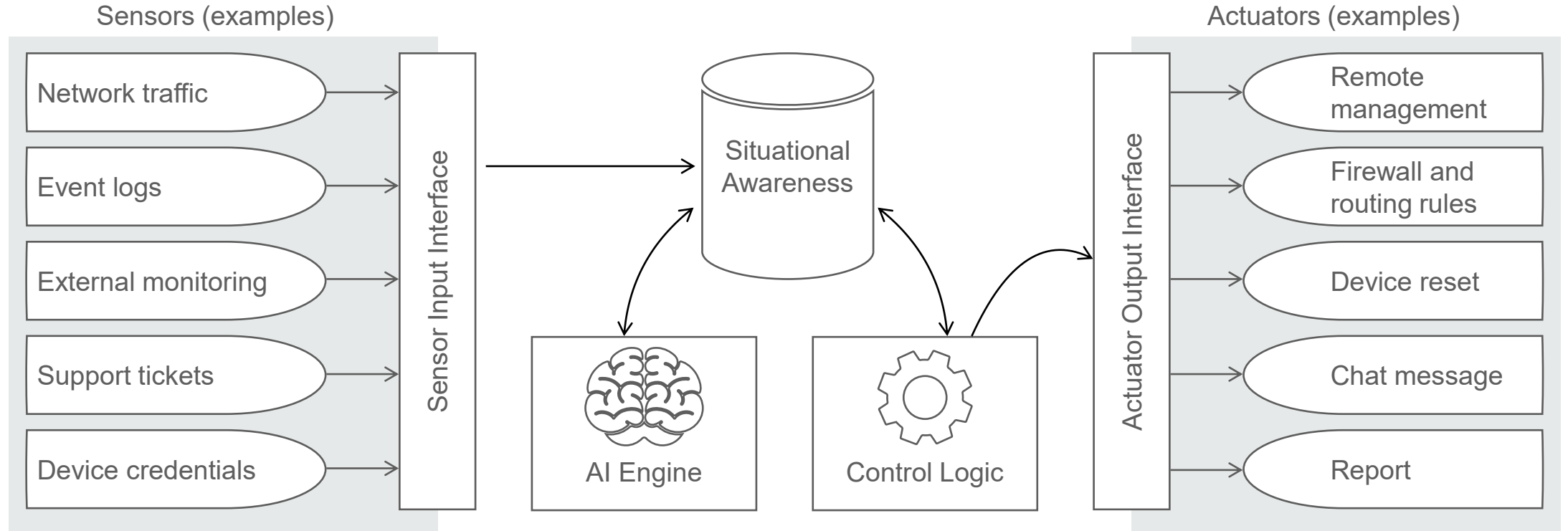
Assessment

Recommendation

Prediction

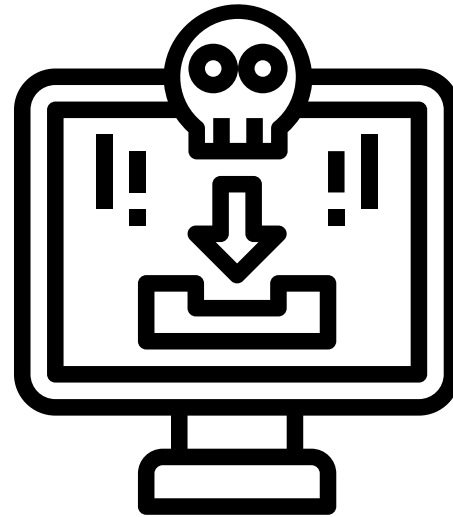


# Automated Blue Team framework overview



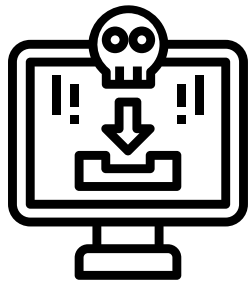


# Case study: A user mistakenly downloads and executes a malicious file





# Case study

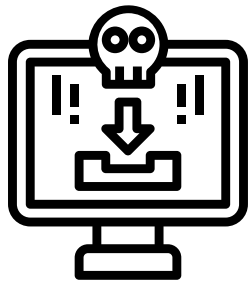


- Configure all clients to send HTTP(s) traffic via a proxy
- Enable detailed logging and send logs to a central server
- Set up recording of all network traffic and feature extraction





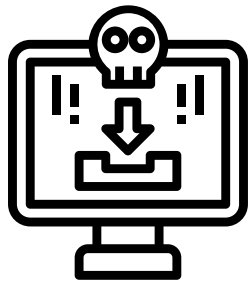
# Case study



- Proxy detects the malicious payload
- Logging reports the execution of an unknown file
- Sniffer detects connection to C&C server



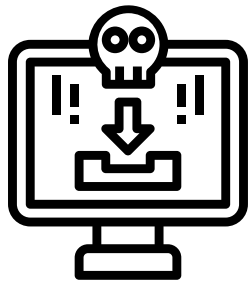
# Case study



- Proxy detects the malicious payload → Remove payload
- Logging reports the execution of an unknown file → Block execution
- Sniffer detects connection to C&C server → Drop packets



# Case study



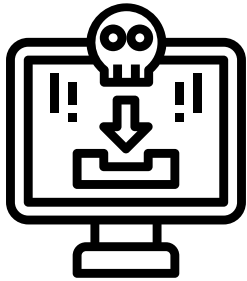
- Human-readable report with information about the incident
  - Malware source (compromised webserver)
  - Malware type
  - ...



# Case study

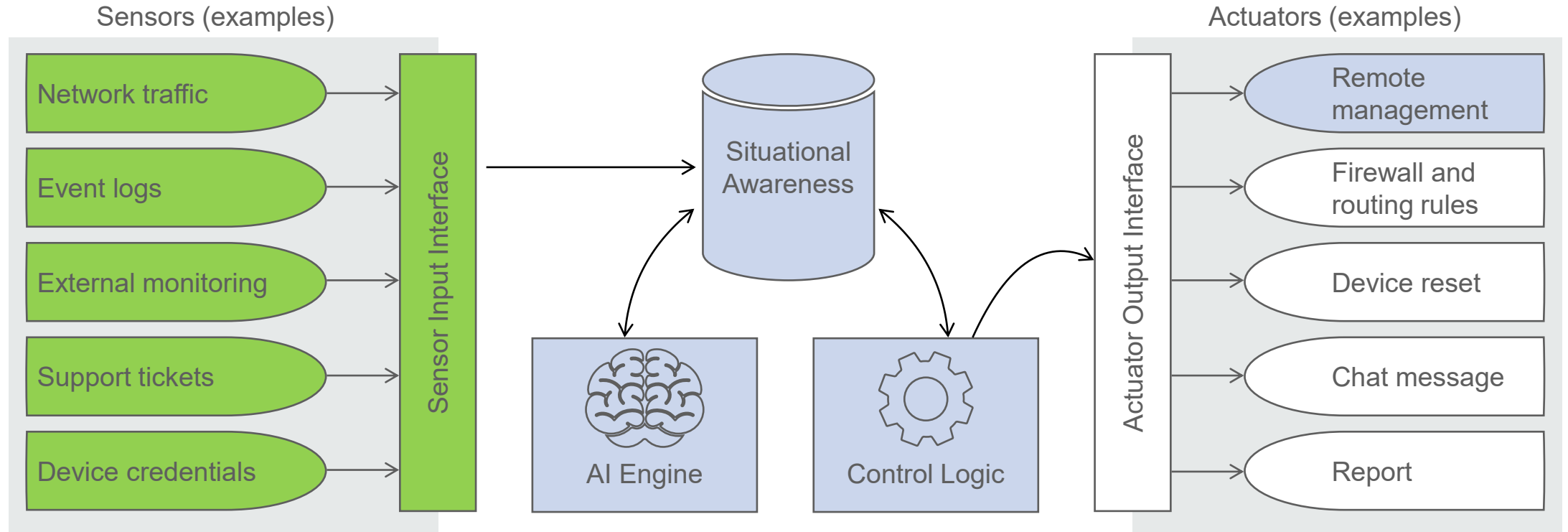


- Restore a device if the malware was executed  
E.g., from a backup



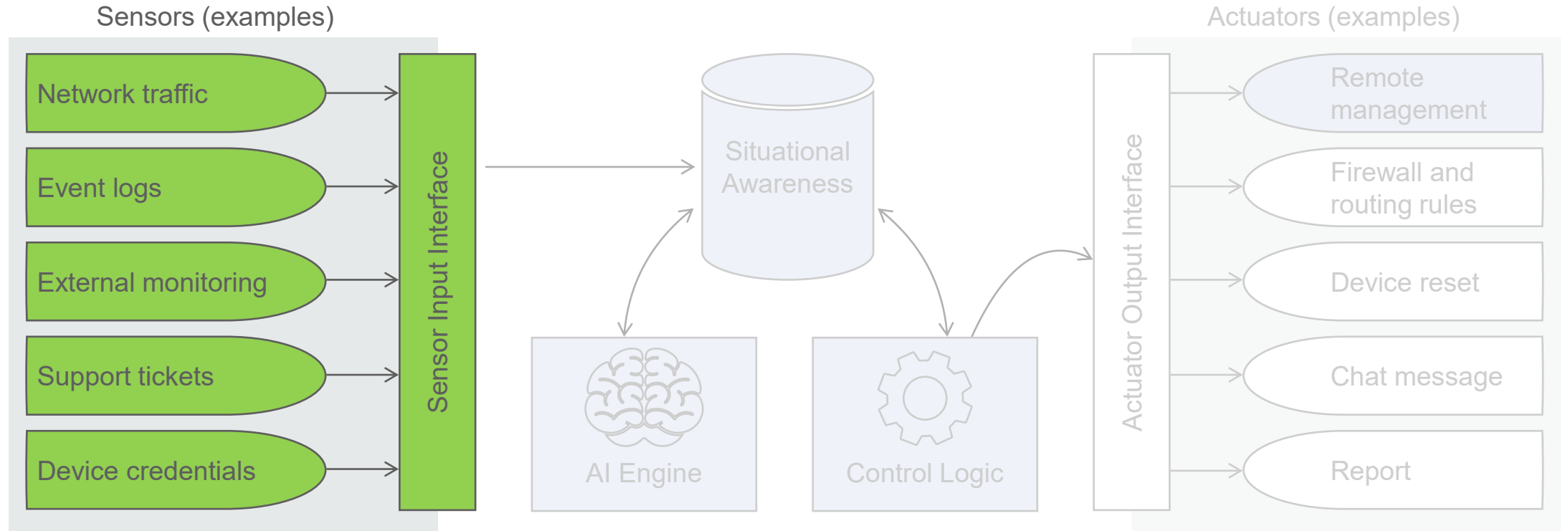


# Current status of the project





# Current status of the project





# Training and testing AI models requires high-quality datasets

- We work with data from the Swiss Blue Team and other collaborating nations
- Since 2023, we (researchers) participate as a separate Blue Team in Locked Shields for data collection
- Datasets collected in 2023 and 2024 are (soon) publicly available





# The LSPR23 dataset

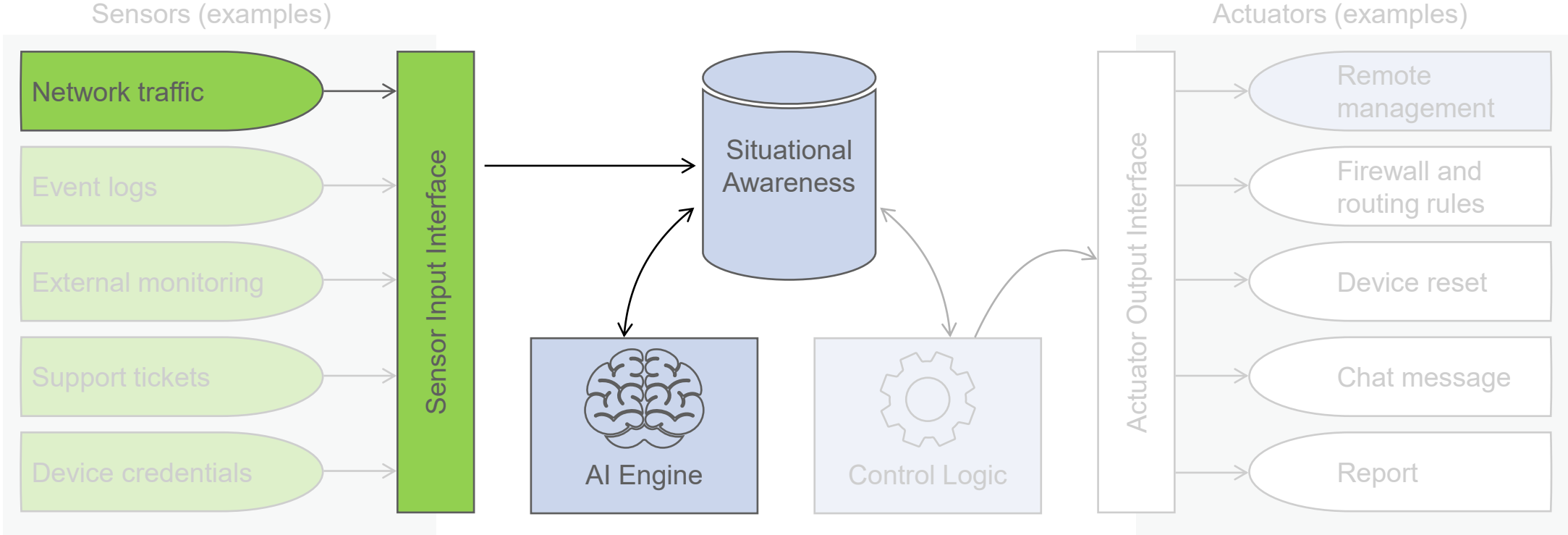
Total Flows	16,353,511	
Labels:	Source	Destination
Blue Flows	12,493,826	10,821,533
Red Flows	13,880	1,630,719
Green Flows (not scoring)	17,256	263,318
Green Flows (scoring bots)	3,415,280	136,321
External Flows	413,269	3,501,620
Benign Flows*	14,708,912	
Malicious Flows*	1,644,599	
Network segments:		
Berilia Energy Group	7,566,863	6,755,988
Berilia Airforce	4,573,708	3,849,478
Bank Of Berilia	277,705	188,984
Berilia Airforce 5G	68,369	27,649
Other	3,866,866	5,531,412

Attack labels "Goals"	
Privilege Escalation	63
System Compromise	58
Data Theft	52
Website Defacement	47
Non Destructive	18
Attack labels "Methods"	
Remote Code Execution	35
Authorize with Default credentials	28
Remote Desktop	23
Authorize with RT credentials	10
Authorize with Stolen credentials	9





# Current status of the project








# We used supervised learning to detect “Command and Control” traffic


- Initial results showed that the approach works well in some cases (when training and testing was in similar settings), but it does not generalize well


Test data


LS17  LS18 

Training data

LS17 

LS18 

LS19 

LS21 

0.993	0.966
0.945	0.993
0.743	0.928
0.952	0.918






F1 scores







# We used supervised learning to detect “Command and Control” traffic

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Test data

LS17  LS18  LS19  LS21  LS21 

Training data

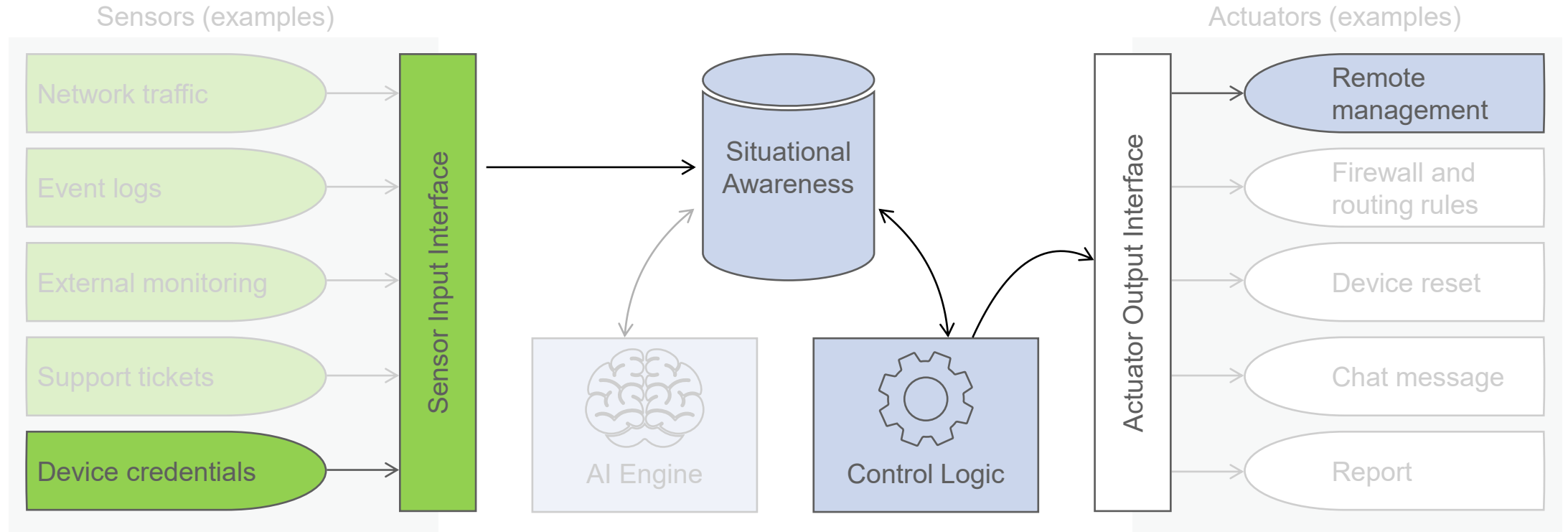
LS17  LS18  LS19  LS21 

0.993	0.966	0.007	0.856	0.215
0.945	0.993	0.060	0.806	0.167
0.743	0.928	0.791	0.351	0.000
0.952	0.918	0.038	0.986	0.158

F1 scores



# Current status of the project





# We detect and fix common misconfigurations automatically

- Currently two tools:
  - Automatically change all login credentials
  - Scan the network for nginx web servers, analyze their configuration and fix some misconfigurations automatically

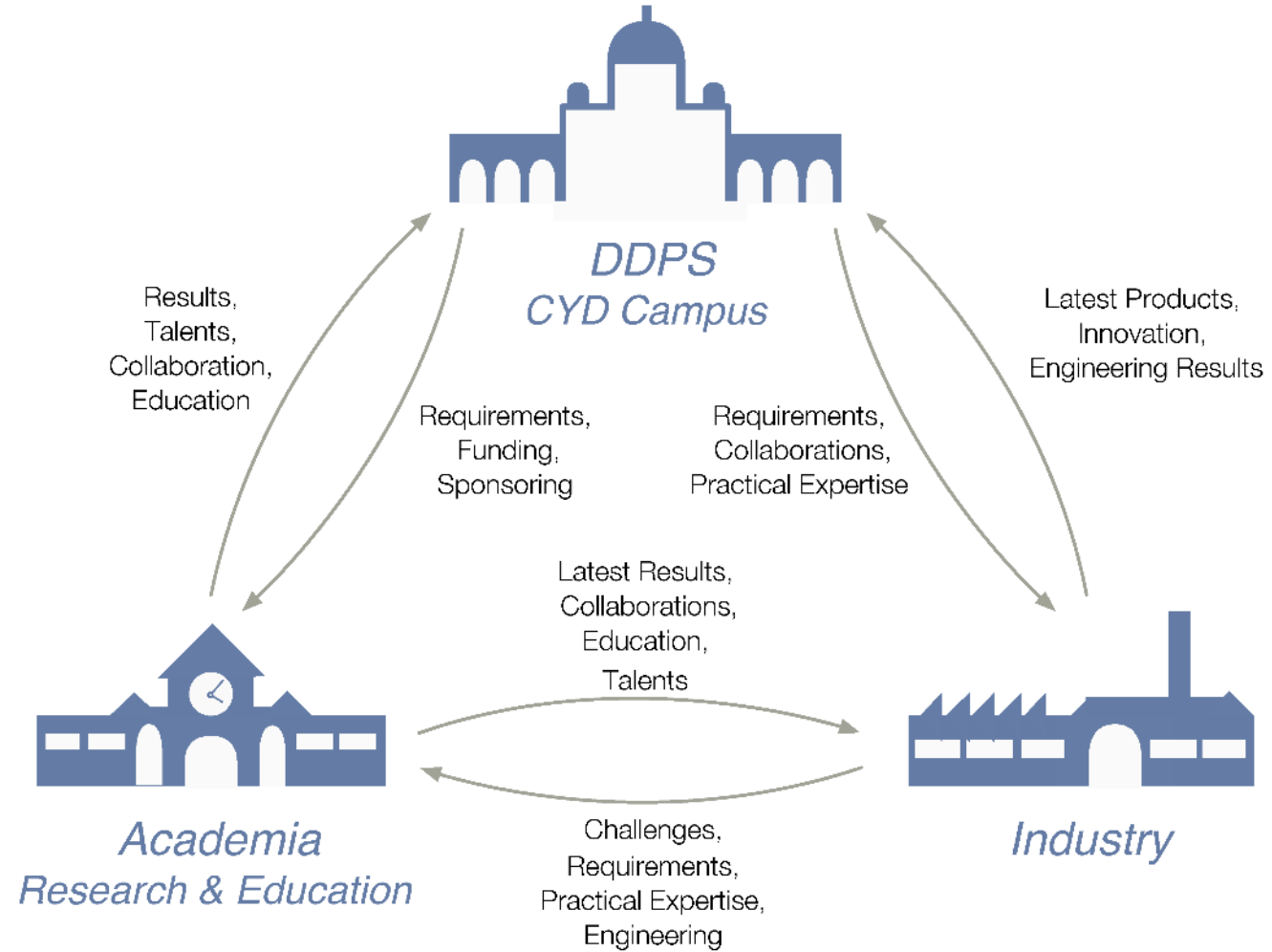


# Other topics we are currently (planning to) investigate

- Using Generative AI / LLMs
  - Parse support tickets
  - Generate reports
  - Generate code or configuration
- Improve existing models such that they generalize better
- Building an additional training and testing environment



# The Cyber-Defence Campus connects government, academia and industry





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**Thank you for your attention!**

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