

THALES

Novel anomaly detection and classification algorithms for IP and mobile networks

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Journée thématique du GT SSLR 2021 sur la sécurité des réseaux - 11 mai 2021

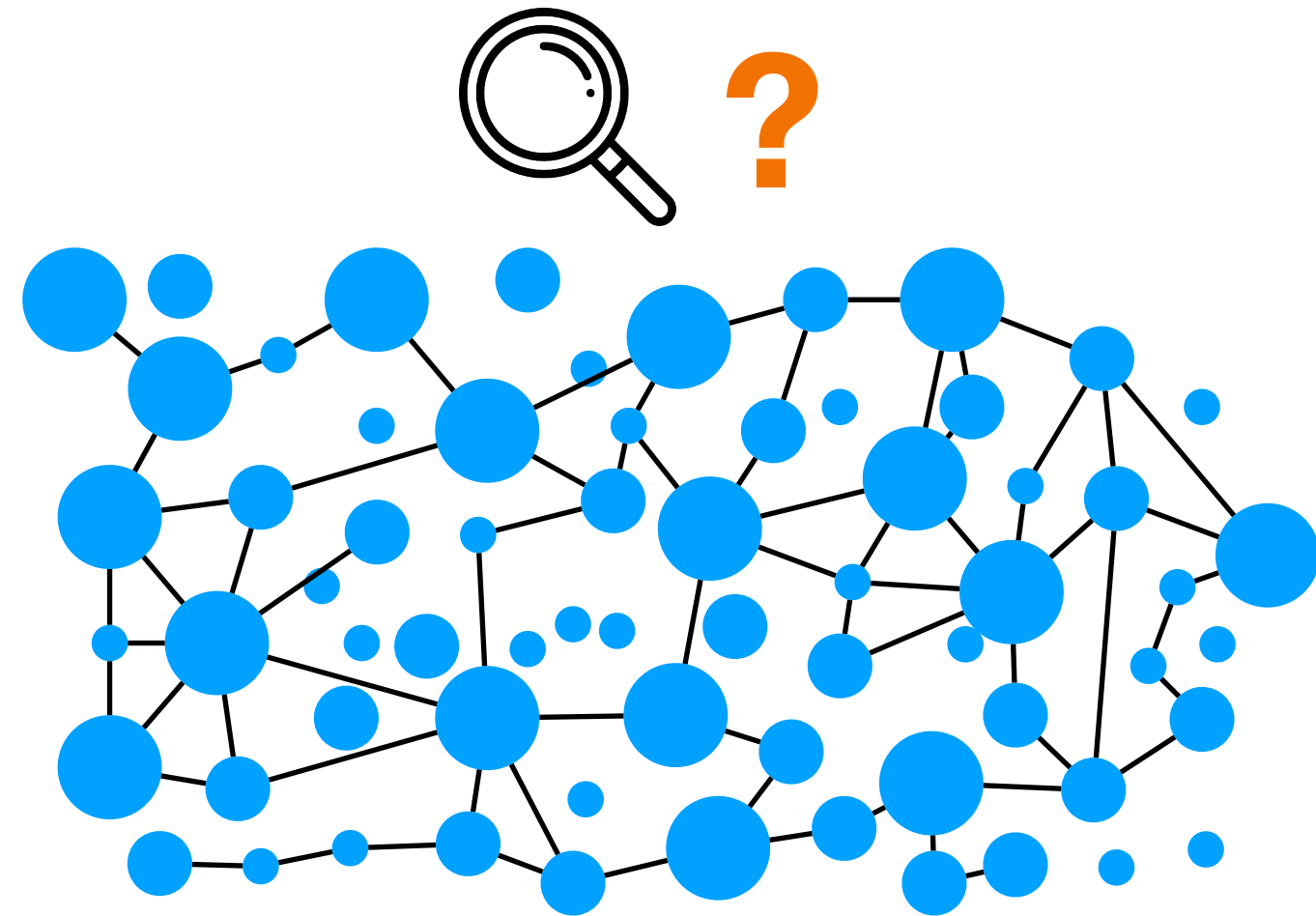
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Data analysis

Data: logs of communications, list of transactions, actions of the users, etc.

Potentially **thousands of logs** to handle each day

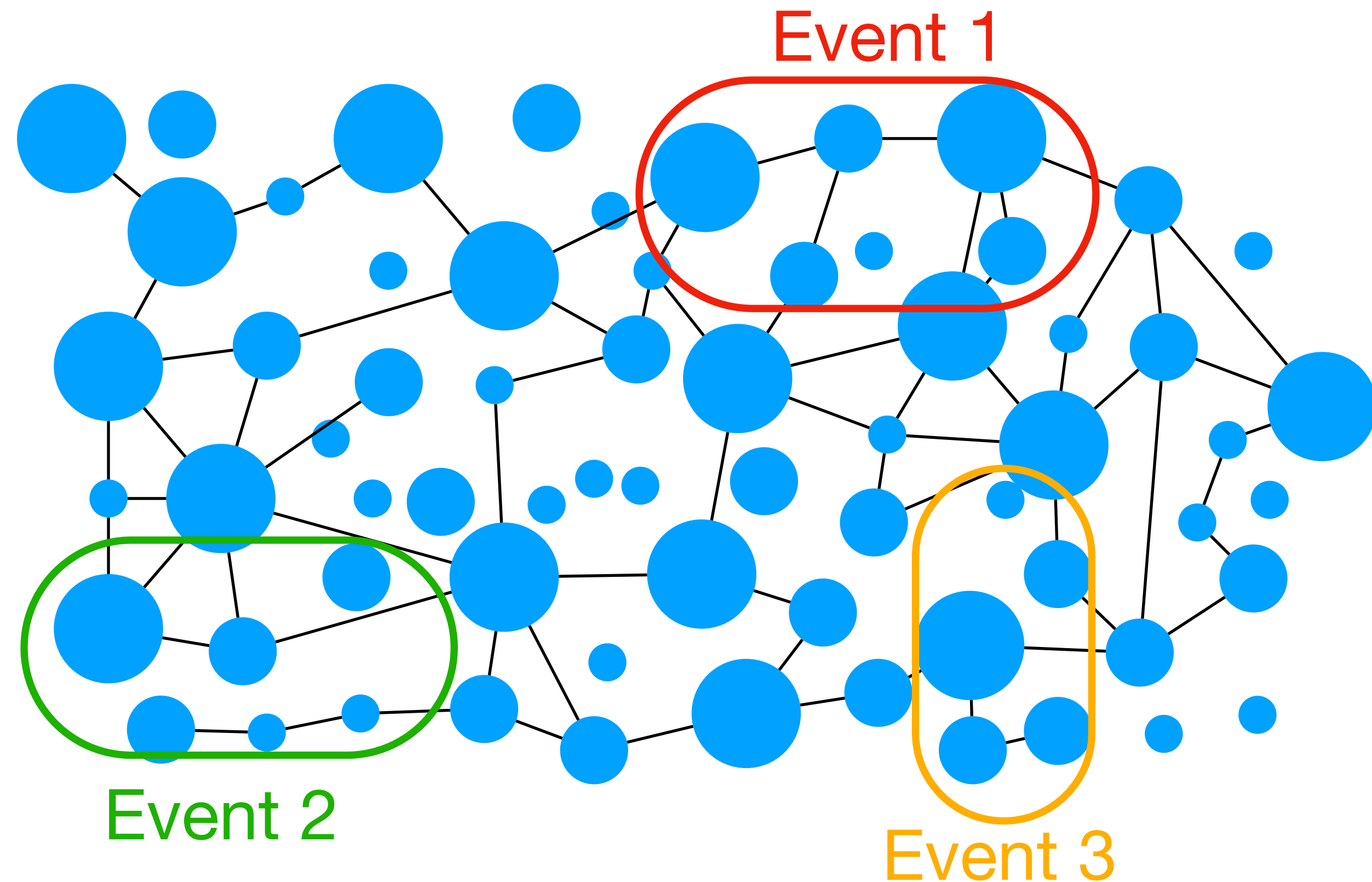


At first sight: **indecipherable**
and no obvious patterns

Knowledge discovery:

- ❖ Find underlying patterns
- ❖ Define generic model for learning

Data analysis techniques



Numerous anomalies

- ✦ Correlate them to find **events**
- ✦ Investigate **root causes, identity** of attackers, modus operandi...

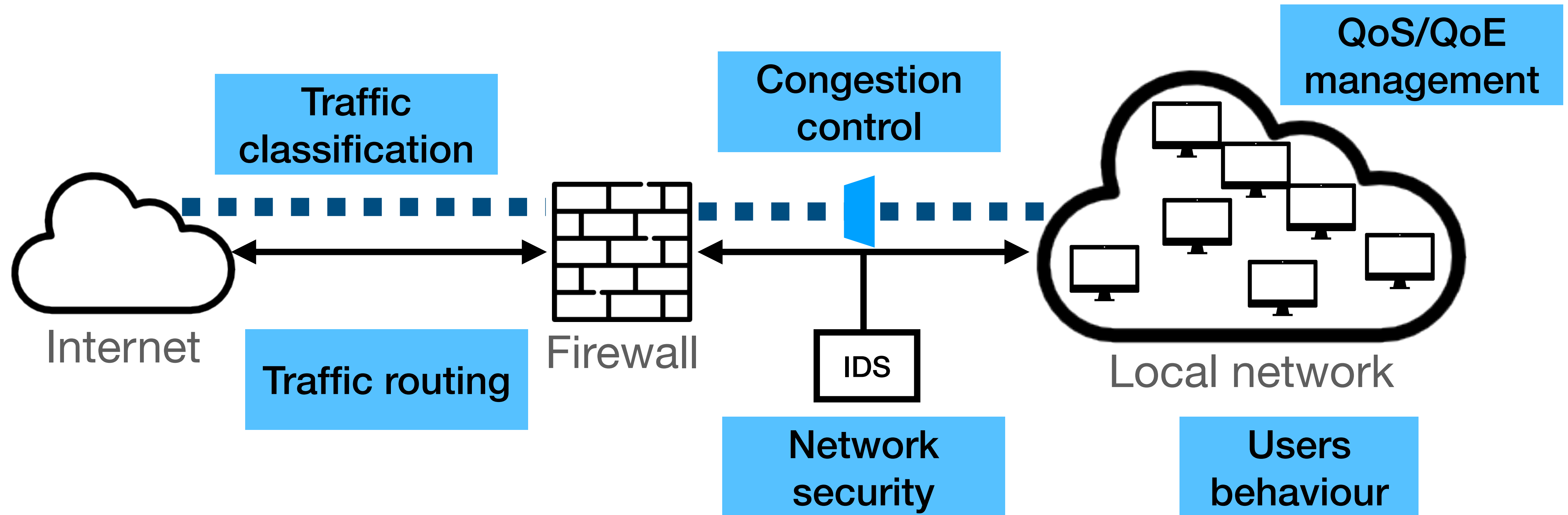
Supervised learning: learning based on example input-output pairs.

→ classification and regression techniques

Unsupervised learning: learns patterns from unlabelled data.

→ clustering and rule extraction techniques

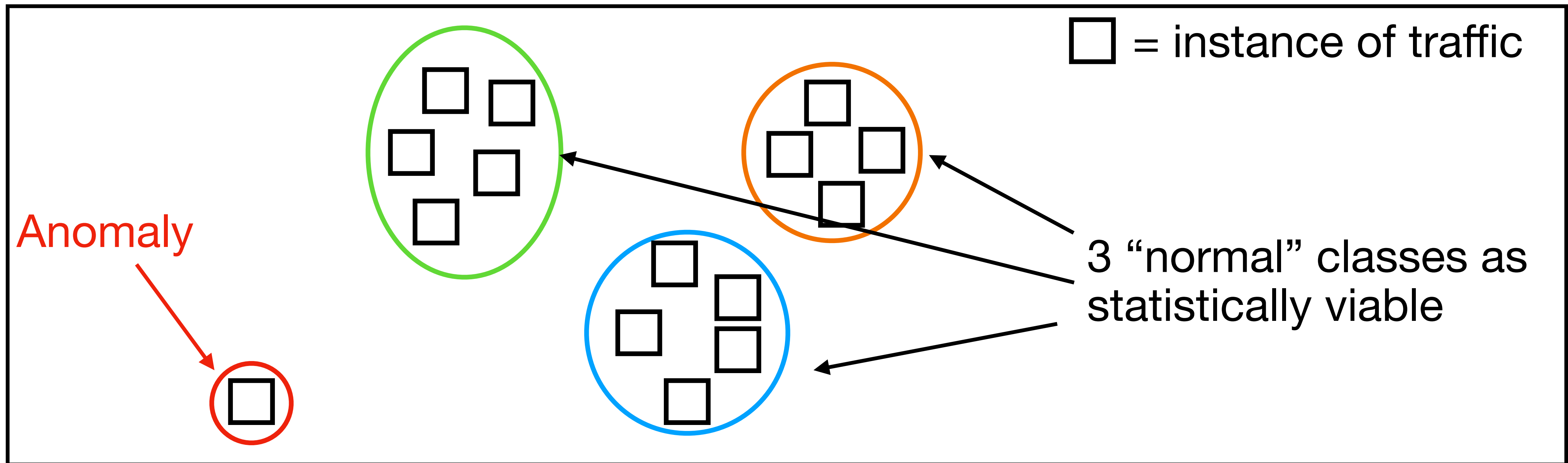
Network behaviour analysis



Targets of data analysis

- ❖ **Malicious behaviour** from users
 - ▶ Denial-of-service attacks, network scanning, click fraud, man-in-the-middle
- ❖ **Unusual behaviour** from users
 - ▶ Bursts of traffic, special events, point-to-multipoint communications
- ❖ **Operational events**
 - ▶ Outages from the network or cloud operator, hardware failures, bad configurations

Data analysis



1. Aggregation level

□ = host, flow?

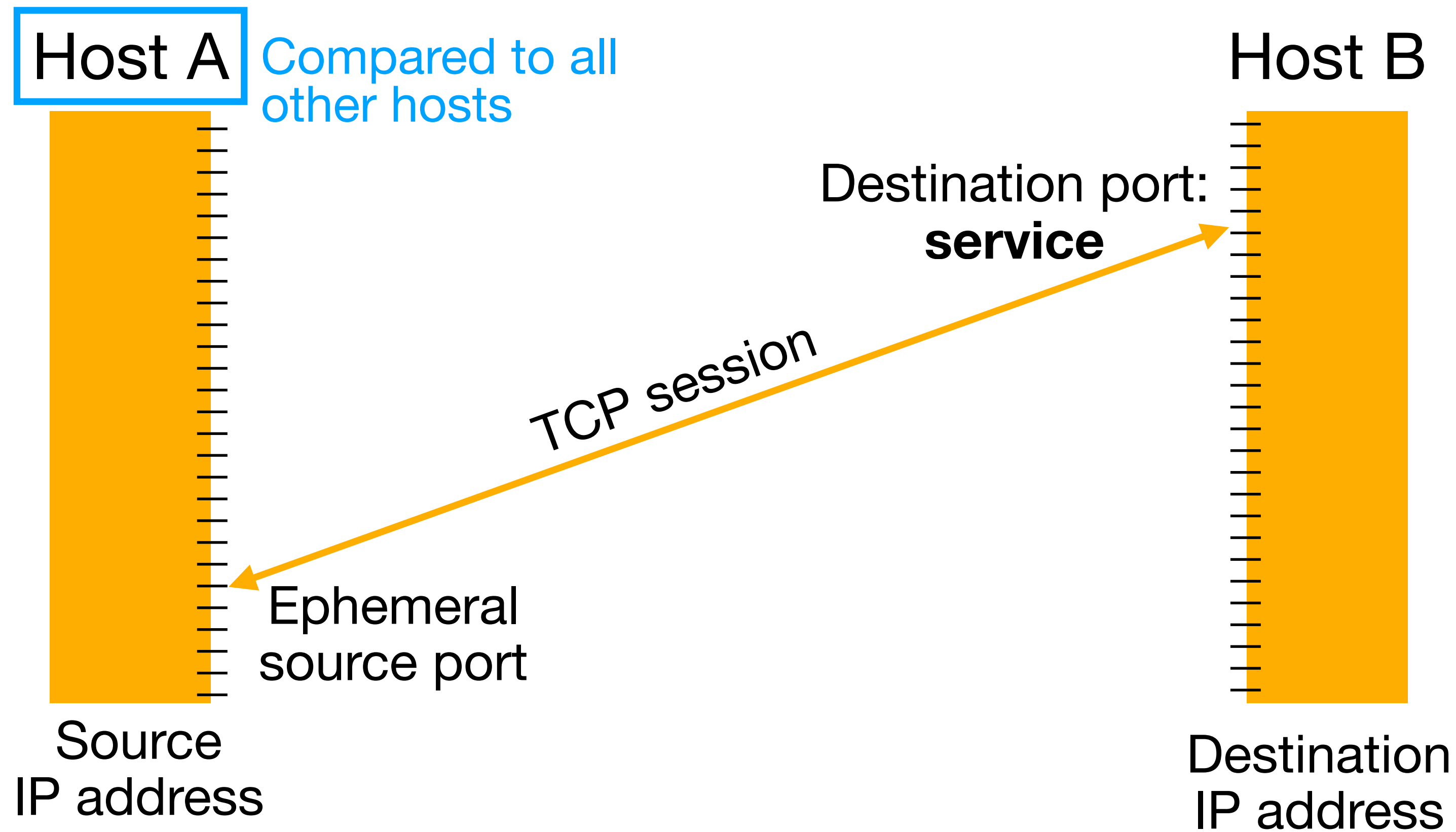
What to characterise?

2. Features choice

→ Attributes of the element

How to characterise it?

Aggregation levels



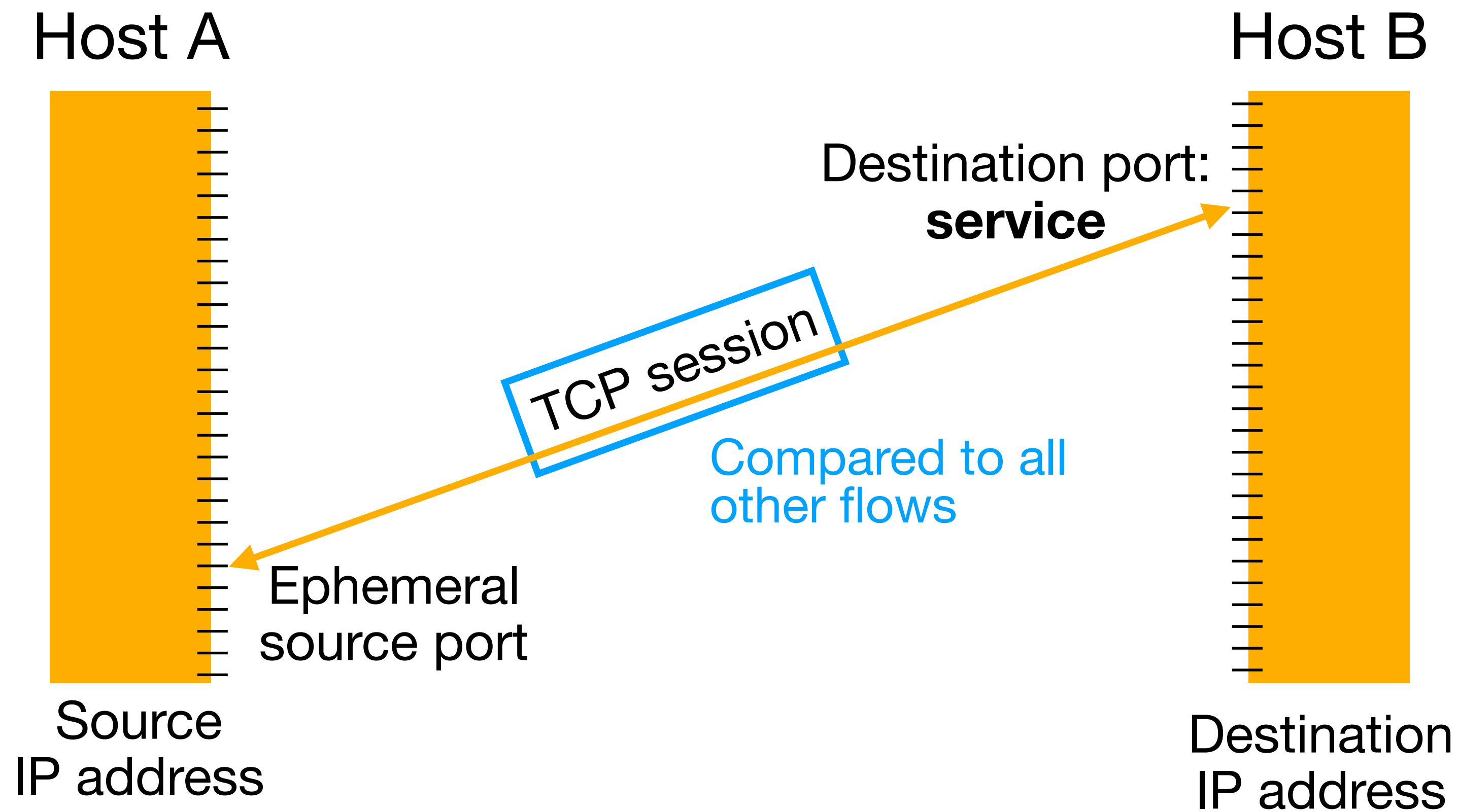
1. Aggregation level

Host behaviour

2. Features

Packet counts, frequency of communications, protocols

Aggregation levels



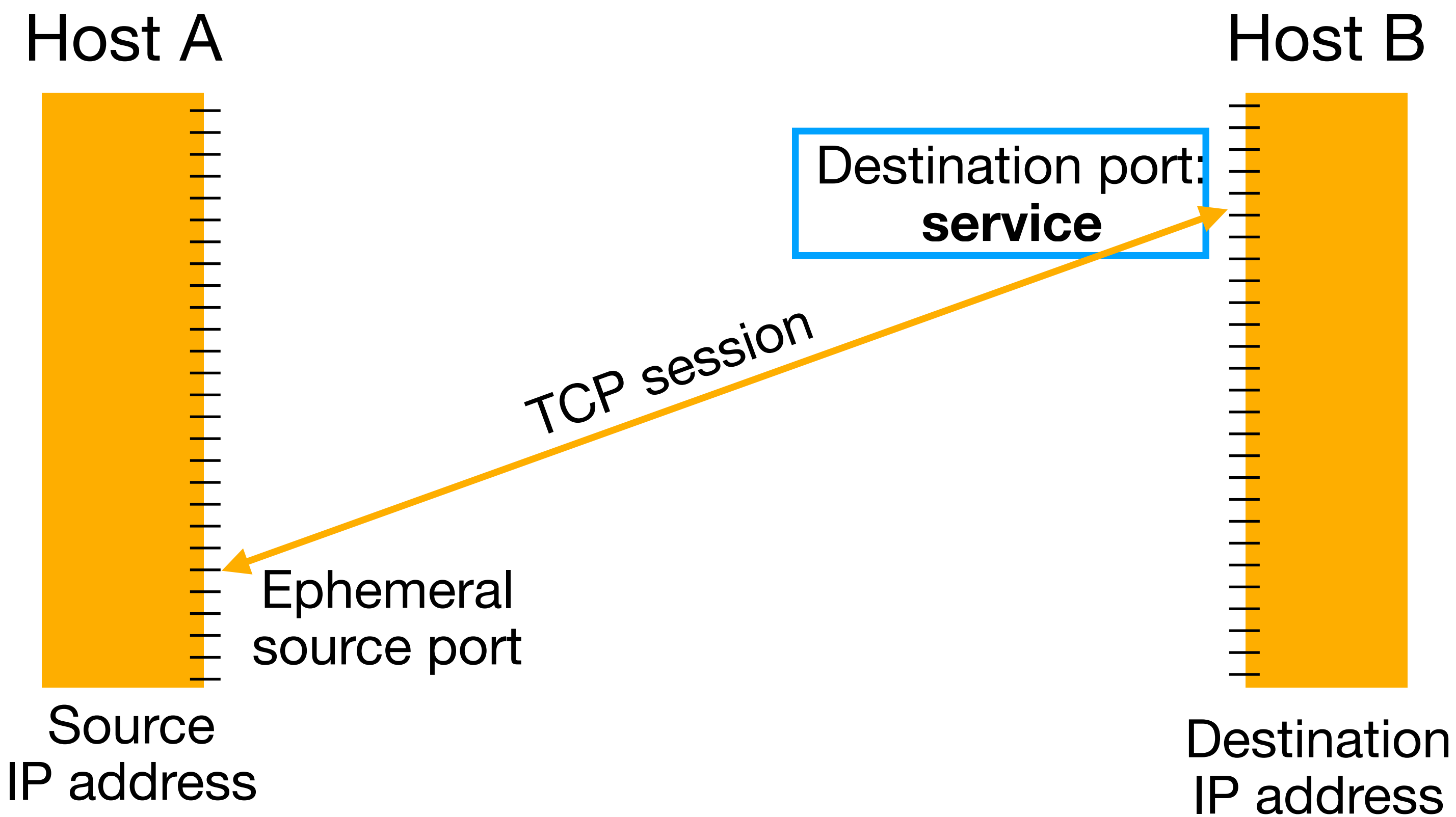
1. Aggregation level

Flow features

2. Features

Flow duration, flow volume, mean packet length, packet inter-arrival time, entropy

Aggregation levels



→ Port or service-level rarely analyzed

1. Aggregation level

Port number / service id

2. Features

Packet counts, diversity indices, protocols

Applications

Analysis of the usage of **services, applications and port numbers**

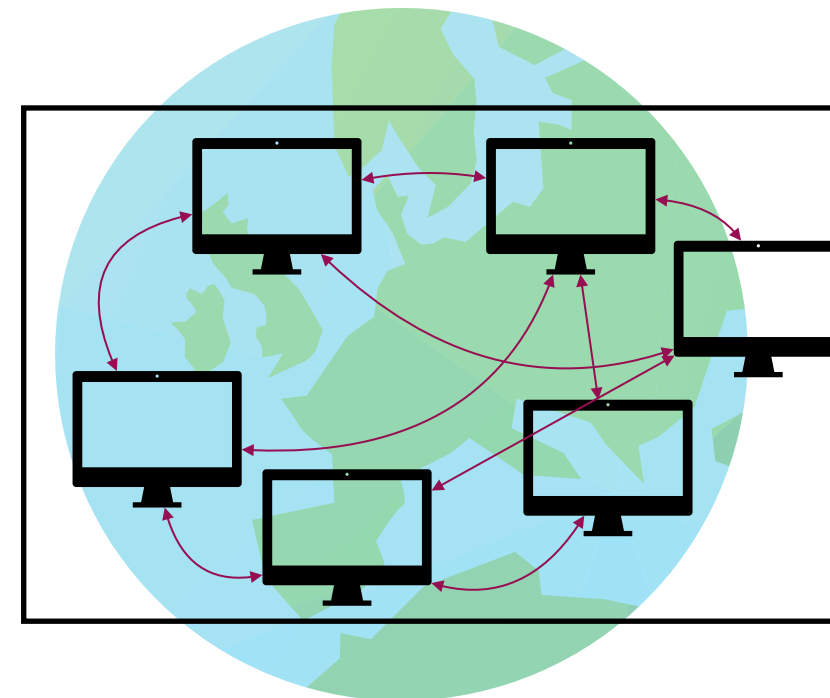
- ❖ **State-of the art:** reasons why unused technique
- ❖ **Objective:** assessing its **benefits** through **lightweight** techniques
- ❖ Examples in 3 different contexts:

Split-and-Merge



Internet-carrier level

BotFingerPrinting



Local (corporate) network

ASTECH



Cellular networks

Security aspects

Behavioural analysis

Per-service detection

Rather **underused** method:

- ❖ Numerous elements to analyse
 - ▶ In IP networks: 65,536 ports
 - ▶ In cellular networks: all services or mobile apps

→ Requires an algorithm of low-complexity

- ❖ Traffic obfuscation to avoid firewalls / encrypted traffic

→ Deep Packet Inspection to induce used applications

Per-service detection

Ports and applications universally and permanently used

Able to identify uncommon behaviours not seen with flows and IP addresses:

- ❖ **Long-term** detection as ports subsist over time

→ Detection of attackers **slowly spreading**

- ❖ **Several vantage points** as ports universally used

→ **Cross-validation**

- ❖ Application **failover** or **update**, **vulnerability scan** on a given port

→ Not visible by analysing IP addresses and flows

Our objectives

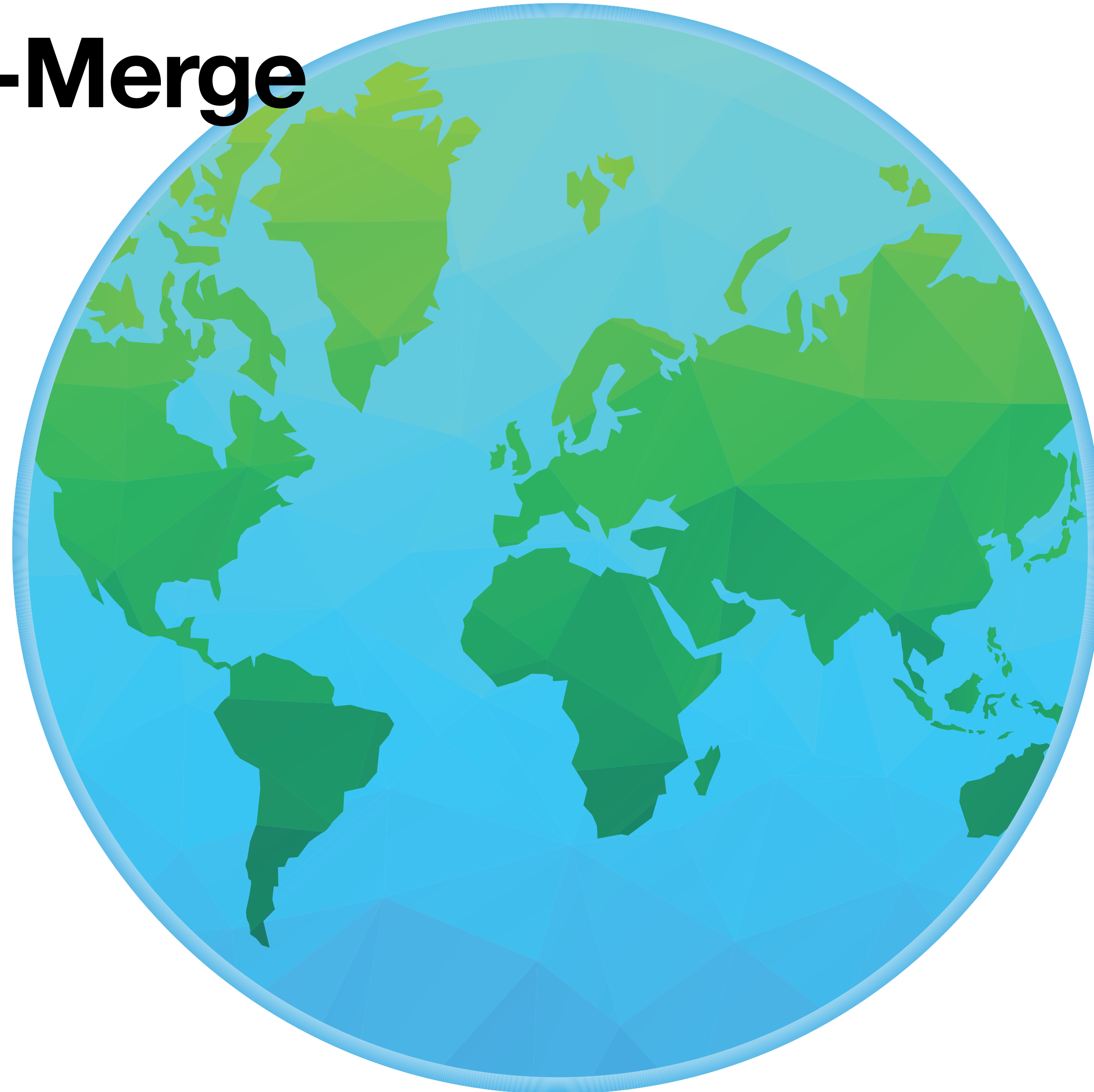
- ❖ State-of-the art: complex approaches, not fit for real networks

Objective: provide a **pragmatic approach**, lightweight, efficient and scalable

- ❖ Through the analysis of **ports, services and applications** usages
- ❖ Using **statistical and machine learning** techniques: classification, clustering, anomaly detection
- ❖ In various contexts: at IP-level, in local networks, in cellular networks

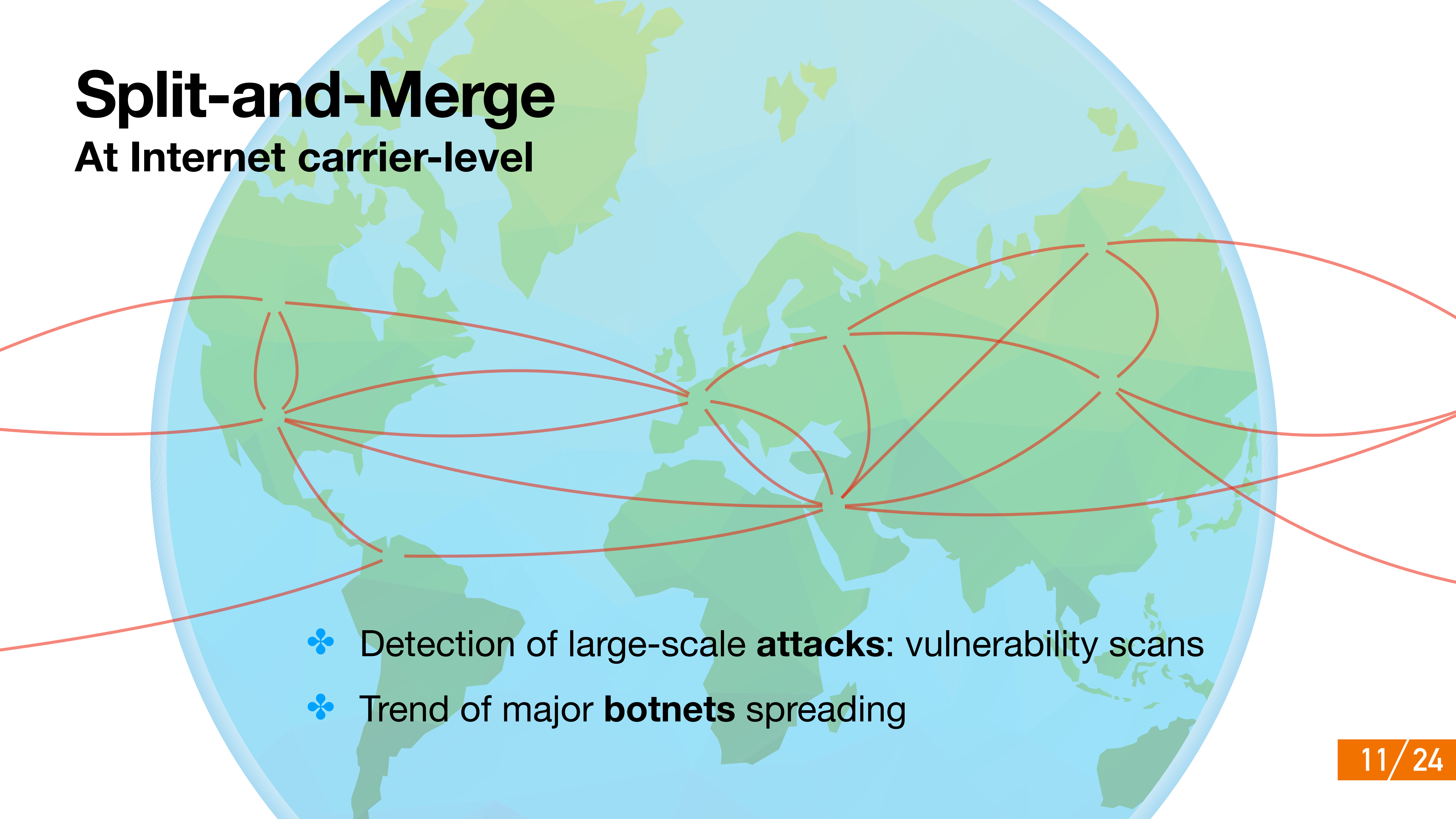
Split-and-Merge

Split-and-Merge



Split-and-Merge

At Internet carrier-level



- ❖ Detection of large-scale **attacks**: vulnerability scans
- ❖ Trend of major **botnets** spreading

Split-and-Merge

Challenge: major botnets spreading **not detected** by traditional Intrusion Detection Systems

↓

Our approach:

- ✿ **Long-term** analysis of ports usage
- ✿ **Cross-validation** in several subnetworks

Our contribution: detection of large-scale **vulnerability scans** and **botnets** spreading





Server vulnerabilities

Exposed to the Internet, open ports, no authentication

Common Vulnerabilities and Exposures:

- ❖ CVE-2018-1000115 (memcached) port 11211
- ❖ CVE-2017-17215 (Huawei HG532 routers) port 37215

IoT devices vulnerabilities

Low computational power to run **security functions**

- ❖ CVE-2018-7445 (MikroTik devices) port 8291
- ❖ CVE-2018-11653 & CVE-2018-11654 (Netwave IP cameras) port 8000

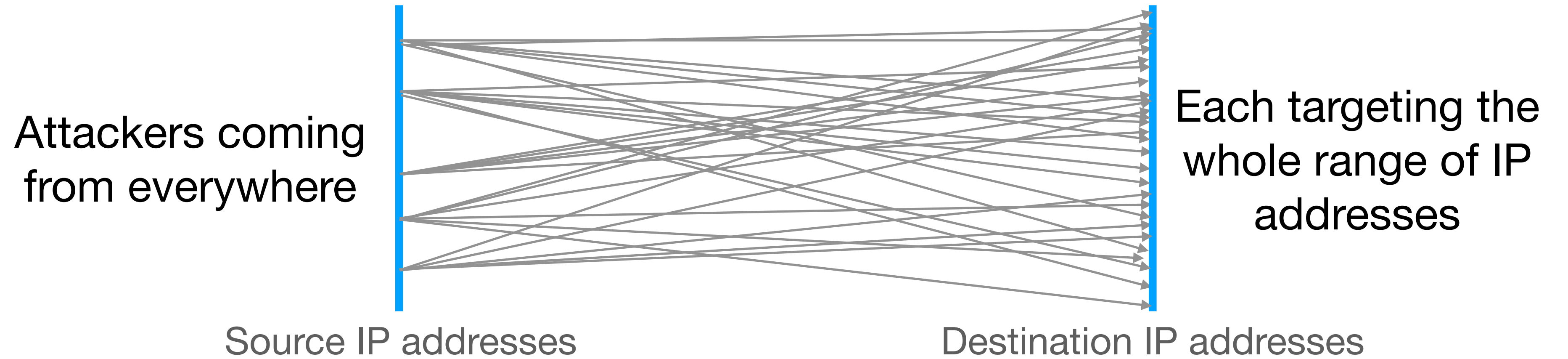
→ Identification of these services or devices by port number.



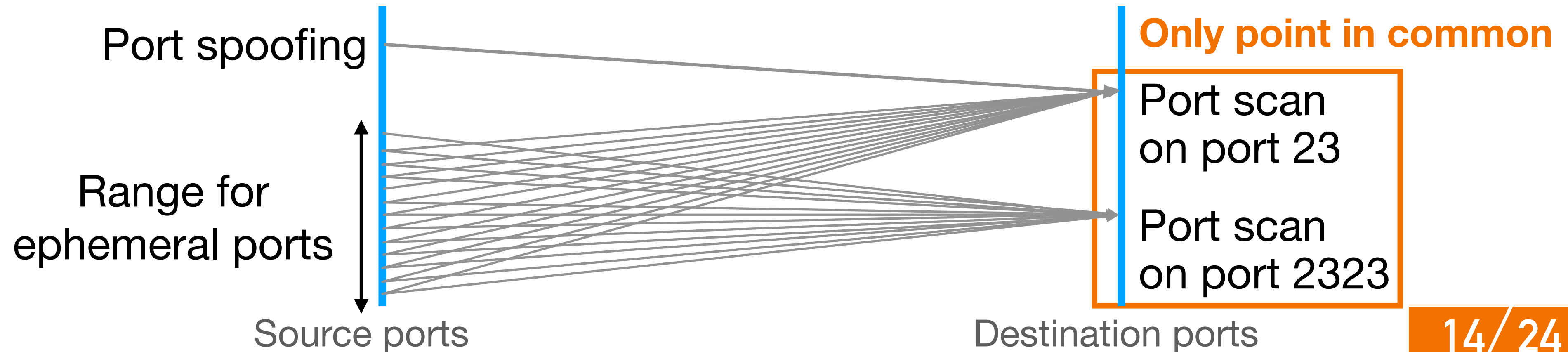
Vulnerability scan

Port scan to identify devices hosting **vulnerable services**

❖ IP addresses



❖ Port numbers





Split-and-Merge

Overview

- ❖ **Long-term analysis of the usage of ports:**
 - 1 - Features computation
 - 2 - Local anomaly detection
 - 3 - Central correlation
 - 4 - Fine-grained anomaly characterisation

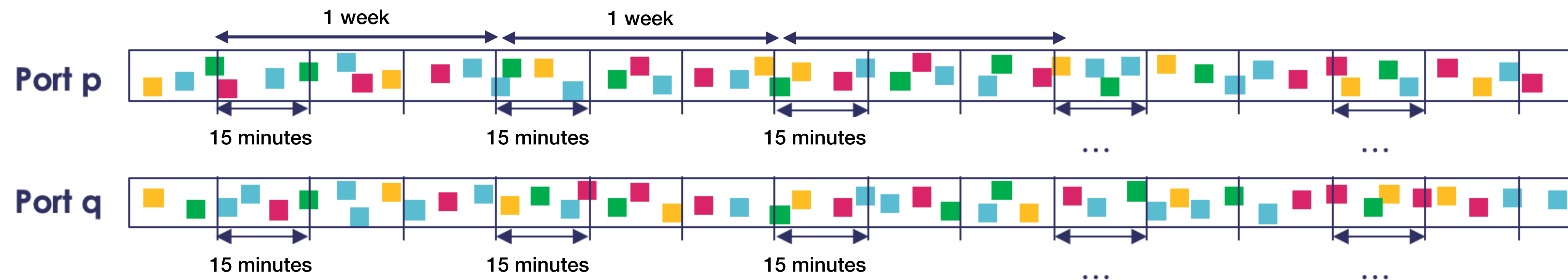


Split-and-Merge

1 - Features computation

For each port p :

- ❖ Source diversity index
- ❖ Destination diversity index
- ❖ Port diversity index
- ❖ Mean packet size
- ❖ Standard deviation of packet size
- ❖ Percentage of SYN packets





Split-and-Merge

2 - Local anomaly detection

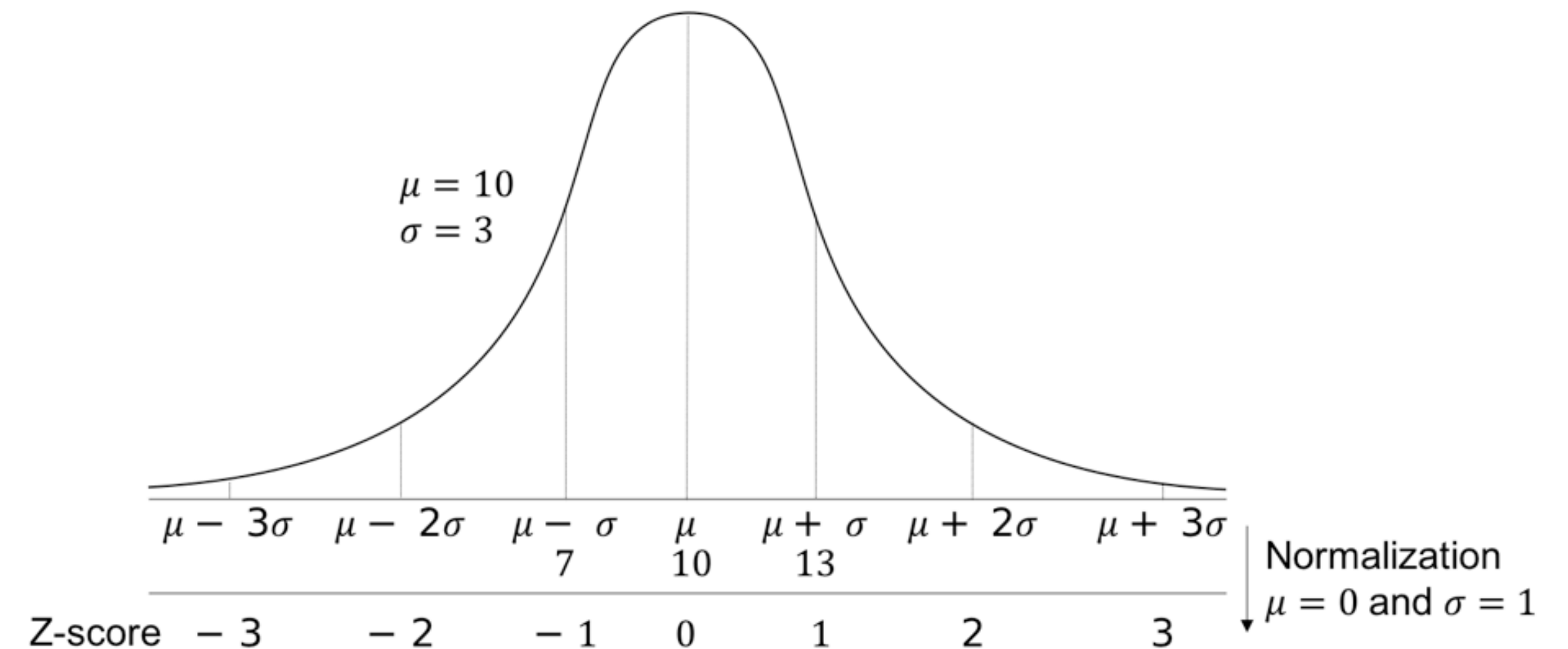
Time series $x \rightarrow$ normal distribution $\mathcal{N}(\mu, \sigma^2)$ of mean μ and std σ

port p	x_1	x_2	x_3
Feature	7	13	30
Feature	54	50	53

❖ Z-score of x_i : $Z = \frac{x_i - \mu}{\sigma}$

→ **not resistant to outliers**

❖ Modified Z-score using median and median std



If $M >$ threshold $T = 3.5 \rightarrow$ **anomaly**

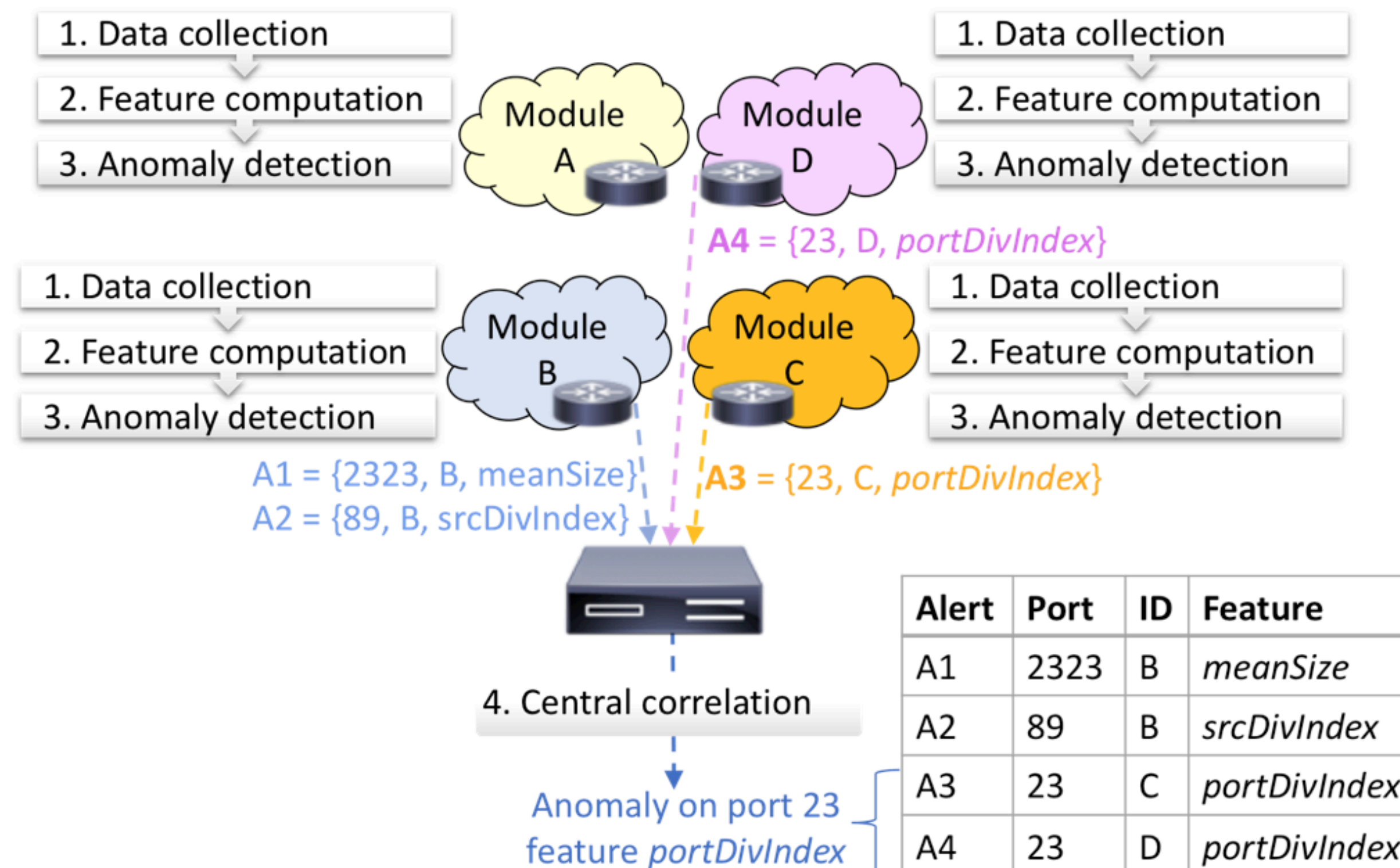


Split-and-Merge

3 - Central correlation

To reduce false positives: Split-and-Merge architecture

Central controller: **keep only distributed anomalies**





Split-and-Merge

4 - Fine-grained characterisation through expert rules

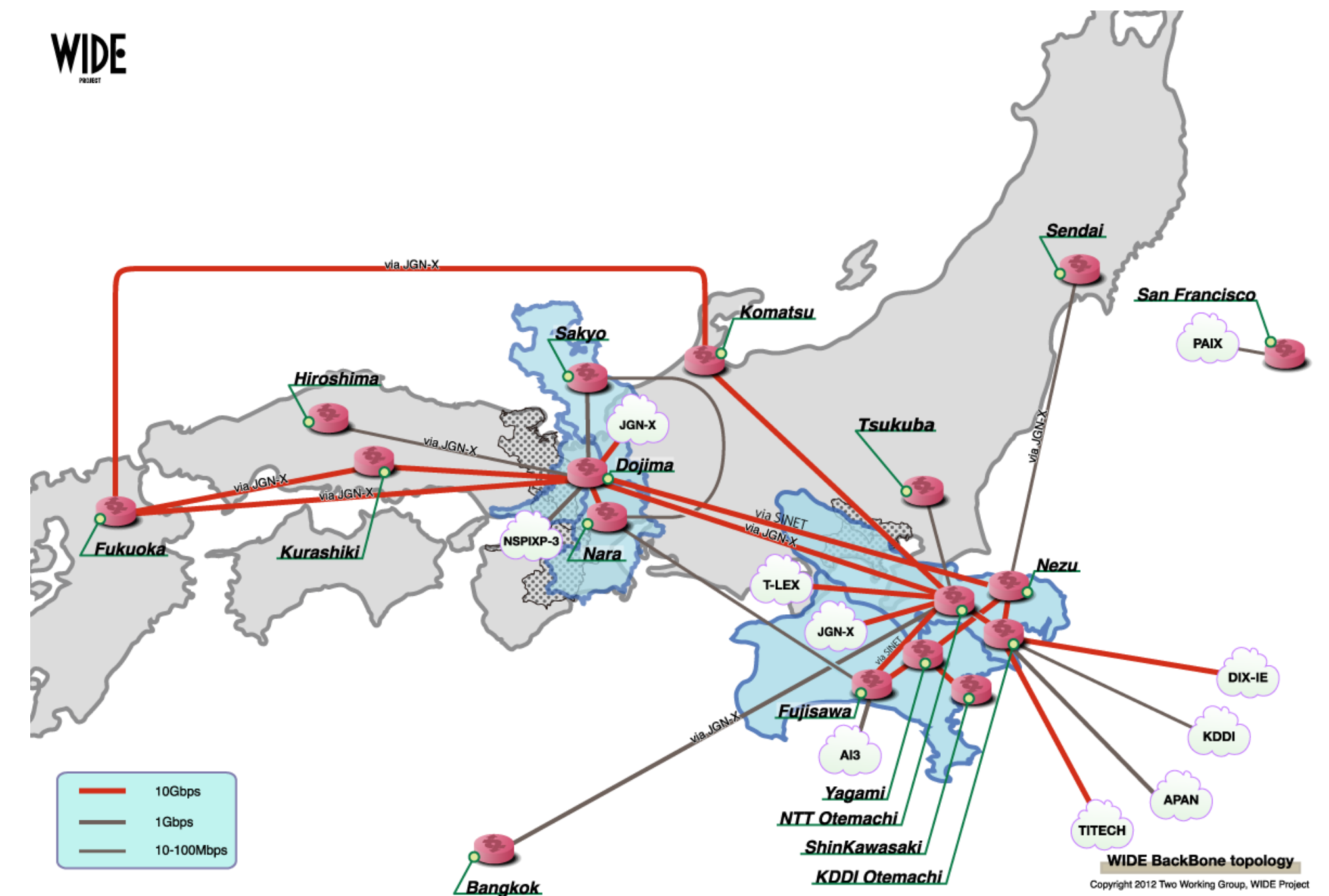
Classes	Characteristics
More normal packets	+meanSize, +stdSize
More forged packets	-meanSize, -stdSize
Large scan	-srcDivIndex, +destDivIndex, -meanSize
DDoS	+srcDivIndex, -destDivIndex
Botnet scan	+srcDivIndex, +destDivIndex, -meanSize
Botnet expansion	+srcDivIndex, +destDivIndex, -stdSize
Targeted scan	-srcDivIndex, -destDivIndex
Less botnet scan	-srcDivIndex, -destDivIndex, +meanSize, -stdSize



Evaluation on real-world traces

MAWI dataset (WIDE Project):

- ❁ **Daily files** of 15 minutes of traffic from a transpacific link
- ❁ Captured between the **MAWI network** and the **upstream ISP**

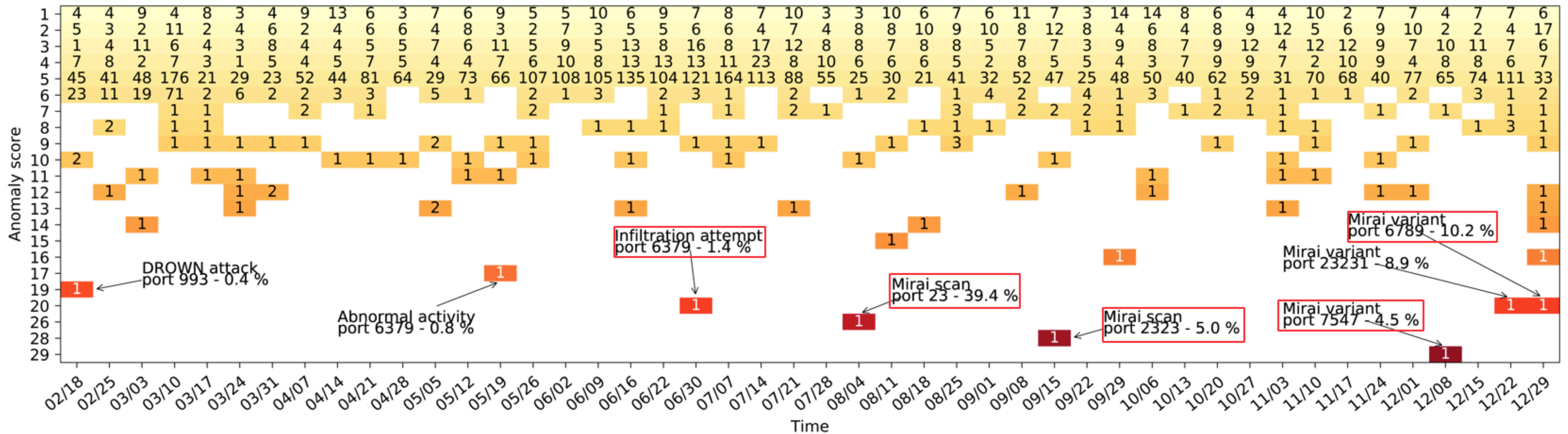




Evaluation (2016)

Anomaly score: number of anomalies for one port

→ Considering **all subnetworks** and **all features**



- ❁ Very low number of anomalies
- ❁ **Not detected** by traditional IDSs (MAWILab, ORUNADA)

MAWILab: combining diverse anomaly detectors for automated anomaly labeling and performance benchmarking, Co-NEXT, 2010.

Online and Scalable Unsupervised Network Anomaly Detection Method, IEEE Transactions on Networks and Service Management, 2016.



Retrospective of major botnets

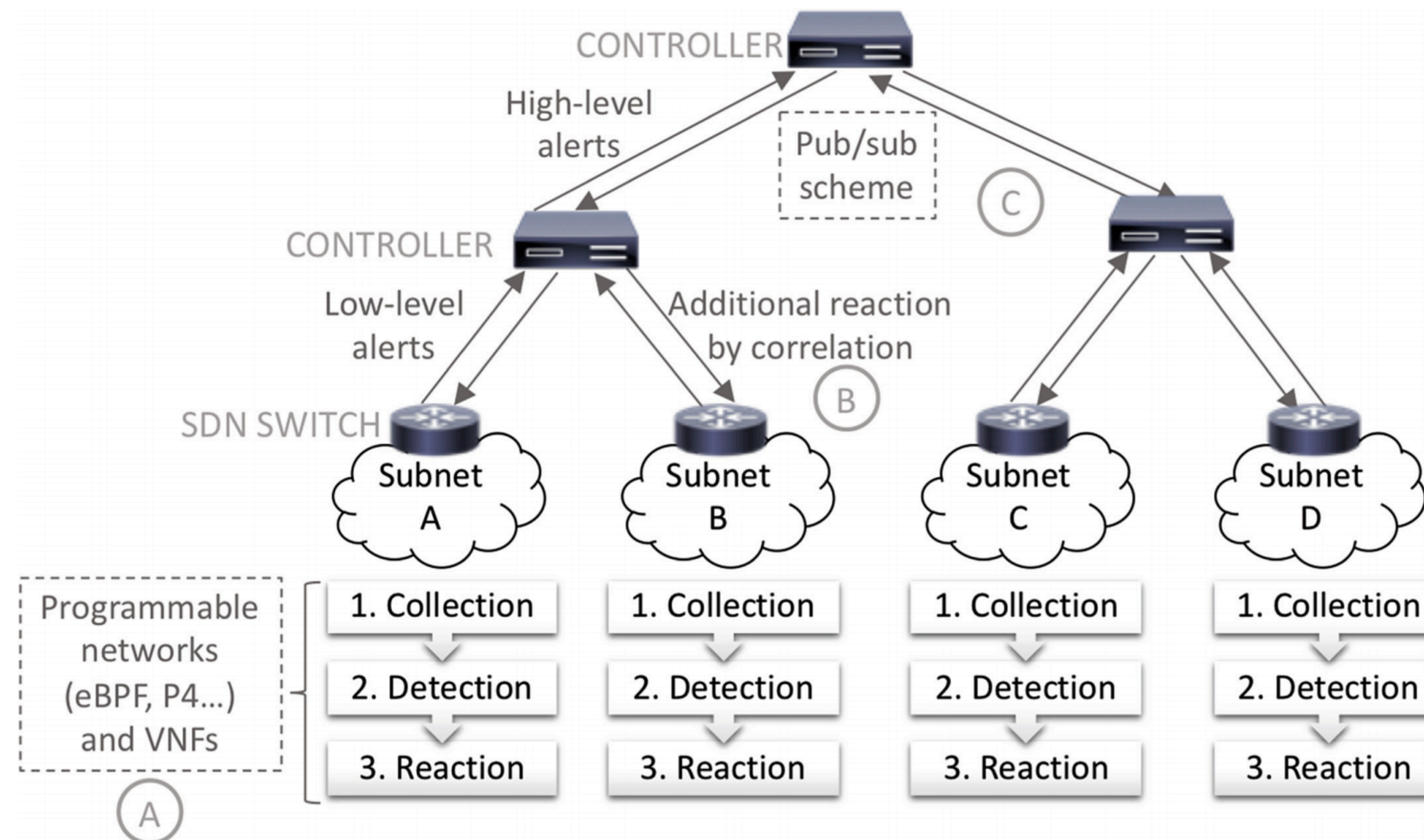
- ❖ Mirai (ports 23, 2323, 7547, 6789, 2222, 23231)
- ❖ Hajime (port 5358)
- ❖ Reaper (port 20480)
- ❖ Satori (ports 37215, 52869, 8000)
- ❖ ADB.Miner (port 5555)
- ❖ Memcached (port 11211)
- ❖ Wannacry (port 445)



Implementation

Local detection at the data plane enhanced by collaboration between ISPs

- ❖ **A:** data plane programming greatly easing the detection and prediction tasks
- ❖ **B:** controller aggregating high-level alerts to detect distributed attacks
- ❖ **C:** various controllers communicating using a pub/sub communication scheme





Split-and-Merge conclusion

Benefits of **per-port detection**:

- ❖ Focus on **port numbers**: detection of **world-wide attacks**, not seen by traditional IDS
- ❖ **Long-term** analysis: possible only when using **port numbers**
- ❖ **Cross-validation** in different subnetworks: very few **false positives**

Lightweight algorithm: ideally running at the switch-level

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