

# INTERNAL NETWORK MONITORING AND ANOMALY DETECTION THROUGH HOST CLUSTERING

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# SHARED RESEARCH PROGRAM – CYBER SECURITY

## ROADMAP MONITORING AND DETECTION

### › Shared data

- › Partners provide (anonymized) data to evaluate developed techniques

### › Shared working

- › Project teams are staffed all partners

### › Shared results

- › The project results are shared among all partners



# TARGETED ATTACKS

## › Targeted

- › Small number of victims
- › Making use of very specific knowledge and vulnerabilities of the target

## › Persistent

- › Patient in gathering information
- › Great effort in staying undetected

## › Highly skilled and well organized attackers

- › Including state actors



**Dell SecureWorks attack lifecycle**

# TARGETED ATTACKS



**Political motives**  
(e.g. Stuxnet, 2010)



**Financial motives**  
(e.g. Carbanak, 2015)



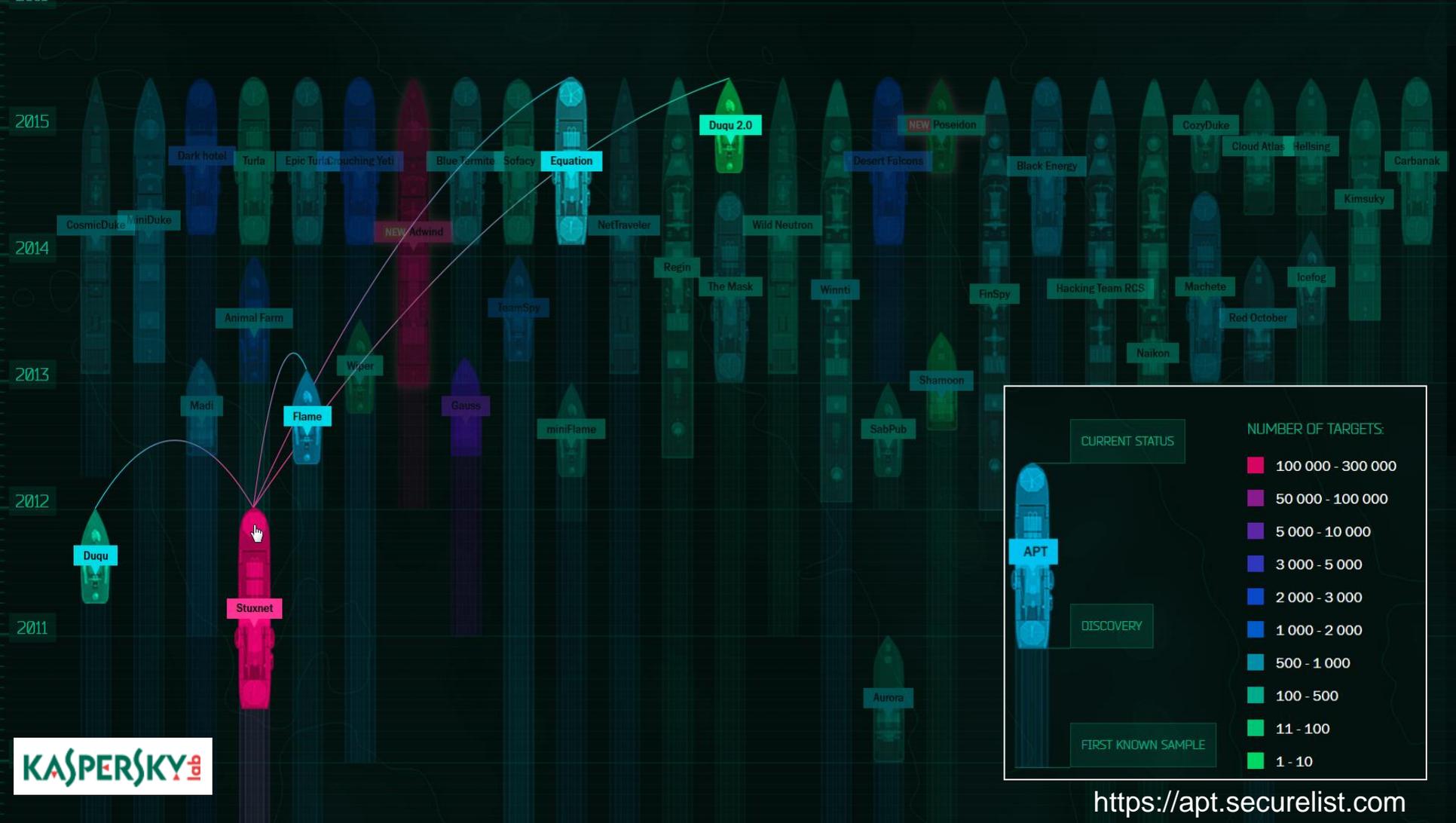
**Intelligence gathering**  
(e.g. Duqu, 2011)



**Shaming**  
(e.g. Sony Hack, 2014)



**Terrorism**  
(next?)



# EXAMPLE: COMMAND AND CONTROL TRENDS (1/4)

## A) FOCUS ON RESILIENCE - TRENDS

- › Huge networks of p0wned systems
  - › IOT will extend this
  - › For CnC host, proxies, staging, etc.
  - › Semi-volatile
- › Two-stage / multi-stage CnC channels
- › Attacks more targeted on *specific* organisations
  - › Specific objective → Low noise
  - › Organisation-specific IPs and domains → IOCs become useless
  - › Spread over long periods





# EXAMPLE: COMMAND AND CONTROL TRENDS (3/4)

## *C) HIDING THE CHANNEL - TRENDS*

- › Encryption
  - › Everything communication, storage, malware
  - › Multiple layers of encryption
- › Use of decoys
- › Hiding in common usage patterns
  - › Low frequency interaction
  - › Lined up with system/network usage patterns



# EXAMPLE: COMMAND AND CONTROL TRENDS (4/4)

## *D) HIDING THE ATTACKERS - TRENDS*

- › More encryption
- › Attacks leave smaller forensic footprints (everything in memory)
- › Detection sandboxes, virtualisation, debugging
- › Falsified footprint information
  - › Copying from other attacks
  - › Different languages
- › Multiple attacker organisations cooperate



# GENERAL TARGETED ATTACK TRENDS (1/2)

- › Attackers are more professional
  - › Better prepared, better tools, better organised
  - › More state actors
  - › C2aaS, BaaS
- › Lots of attacks are seemingly connected
  - Possible organised groups carrying out multiple attacks
- › Increasing use of standardised toolkits for every stage
  - › Tailored for each attack



## GENERAL TARGETED ATTACK TRENDS (2/2)

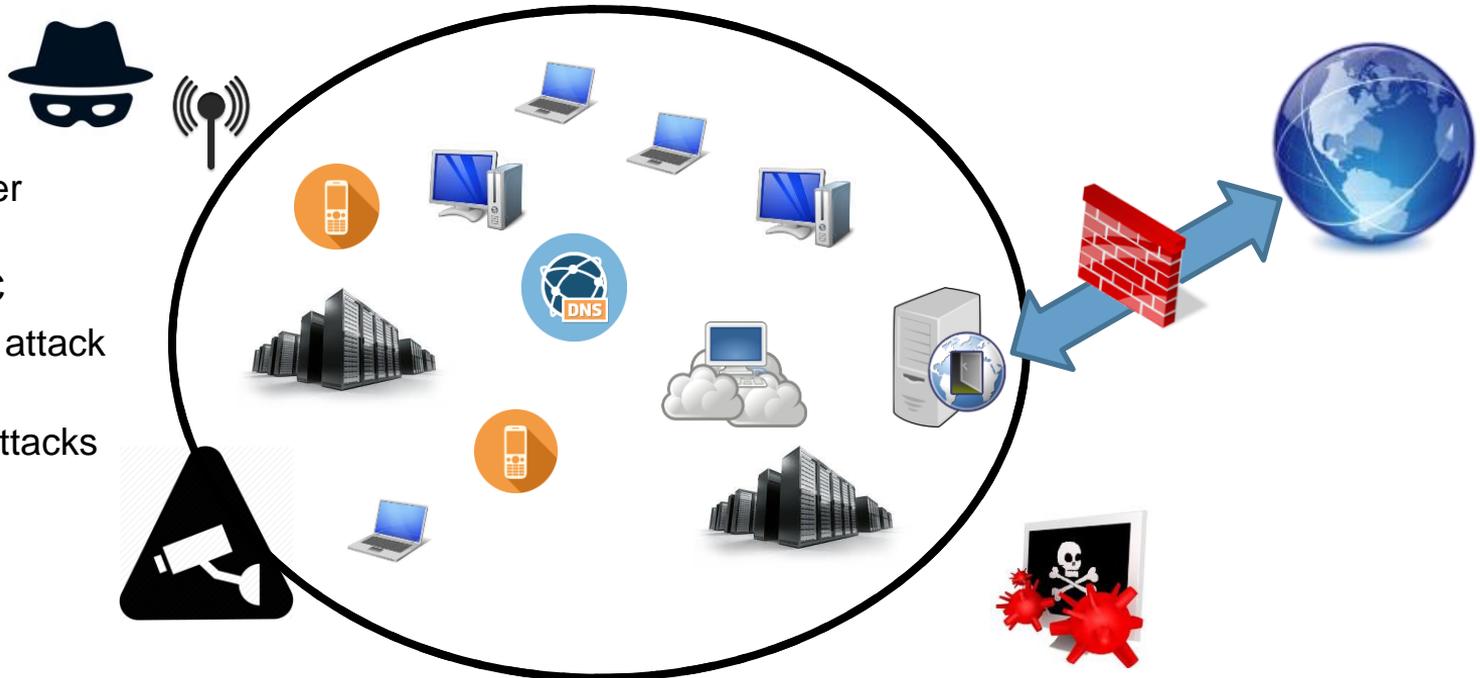
- › Development pace increases
- › Attacks, infrastructure, techniques diversify
- › Attacks persist for long times → evolve gradually
- › Supply chains are attacked as well
  - › Pre-installed rootkits
  - › Malicious hardware components



# HARD OUTER LAYER & SOFT INNER LAYER

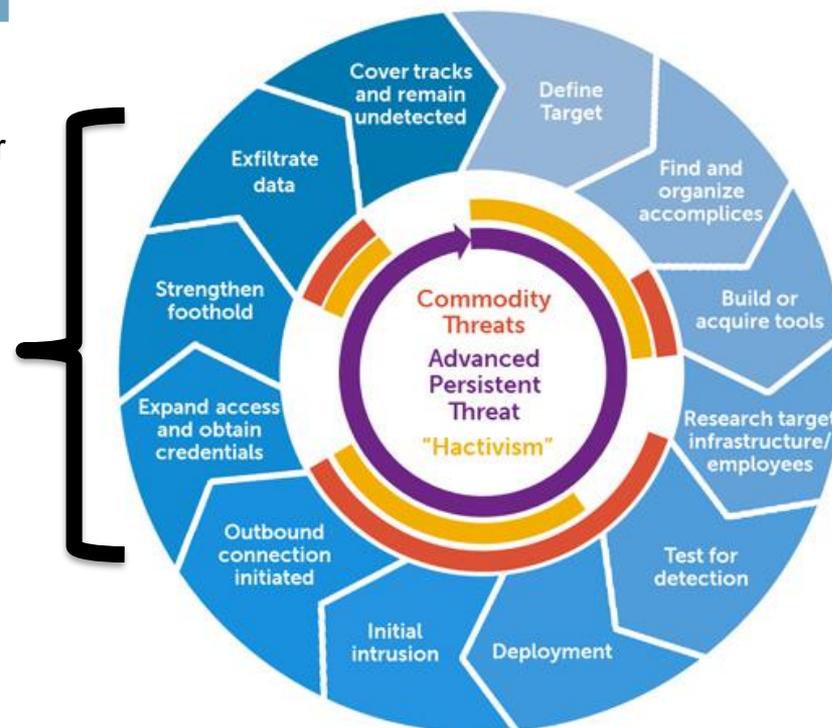
## NEED FOR INTERNAL NETWORK MONITORING AND DETECTION

- › Advanced cyber attacks
- › Advanced CnC
- › IoT: increasing attack surface
- › Supply chain attacks
- › ...

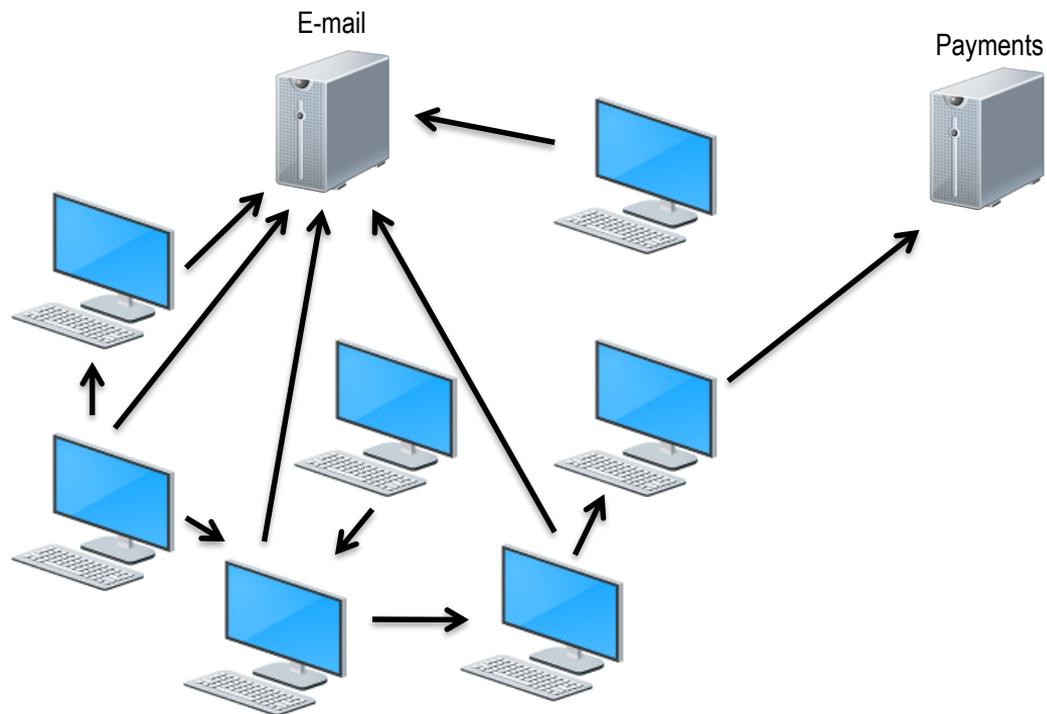


# INTERNAL NETWORK TRAFFIC IS A VALUABLE SOURCE OF INFORMATION

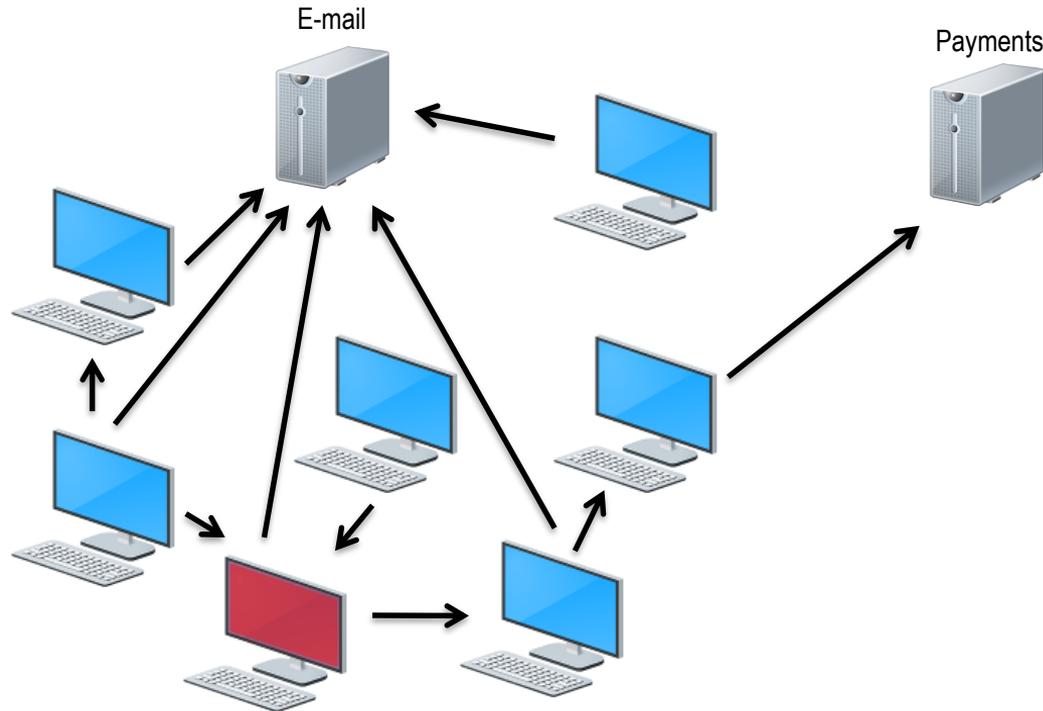
- › Undervalued source of information
  - › Many detection mechanism focus on the border of the network
- › Targeted attack behaviour is visible in the internal network traffic
  - › Probing hosts for vulnerabilities
  - › P2P C&C channel
  - › Local proxies
  - › Staging servers
  - › ...



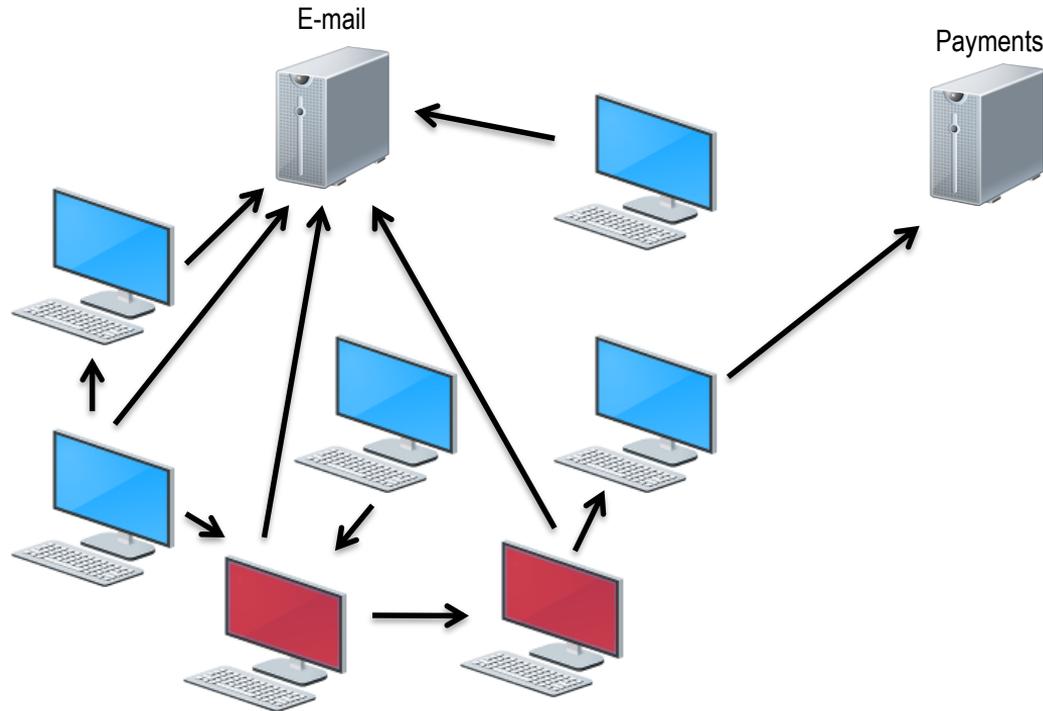
# INTERNAL NETWORK ACTIVITY IN CARBANAK (1/4)



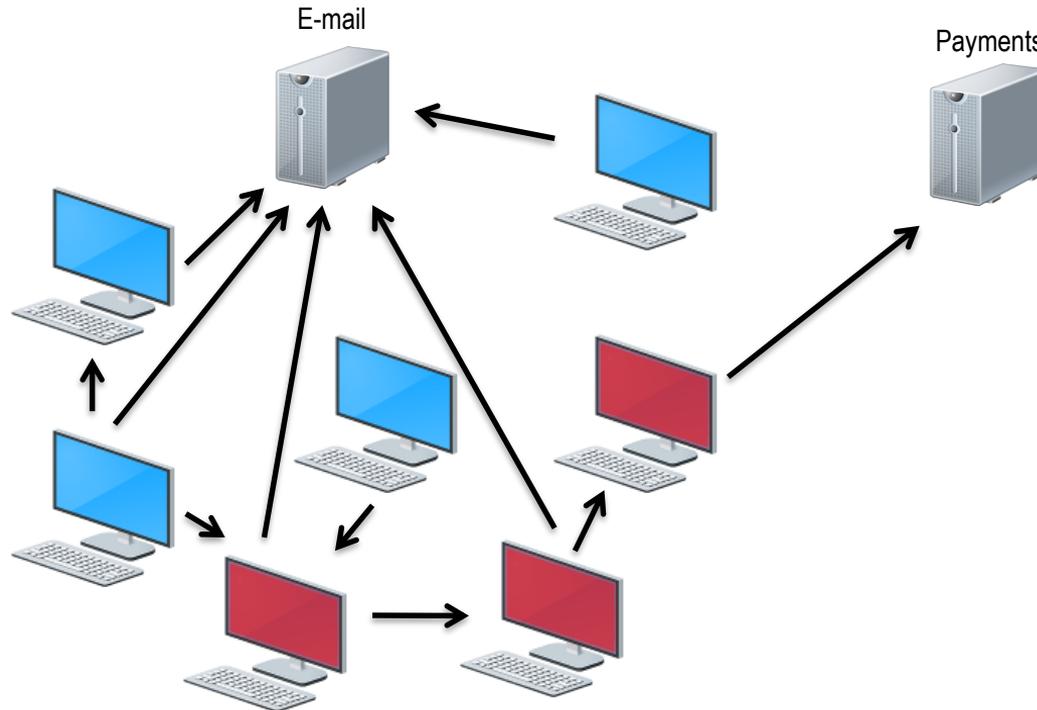
# INTERNAL NETWORK ACTIVITY IN CARBANAK (2/4)



# INTERNAL NETWORK ACTIVITY IN CARBANAK (3/4)



# INTERNAL NETWORK ACTIVITY IN CARBANAK (4/4)



- Online-banking**  
Money was transferred to fraudsters' accounts
- E-payment systems**  
Money was transferred to banks in China and the US
- Inflating account balances**  
The extra funds were pocketed via a fraudulent transaction
- Controlling ATMs**  
Orders to dispense cash at a pre-determined time

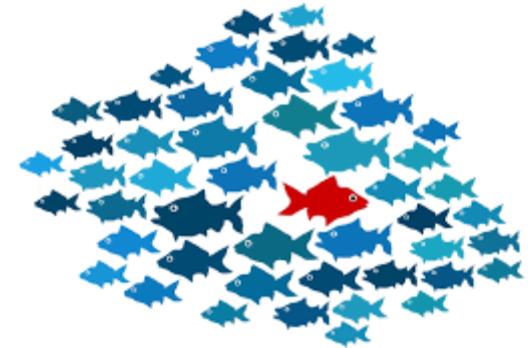
# ANOMALY DETECTION APPROACH TO BE ABLE TO DETECT NEW THREATS

## › Misuse-based

- › Uses signatures
- › Low false positive rate
- › Crucial because of the modular approach in which parts of attacks are re-used
- › Can only discover known threats

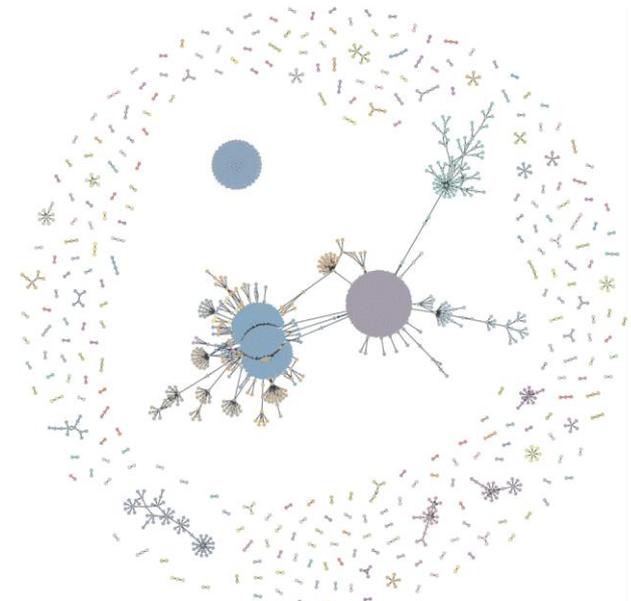
## › Anomaly-based

- › Uses heuristics
- › Often high false-positive rate
- › Capable of discovering new and organization specific threats



# INFORMATION CONCENTRATORS ARE USED TO EFFICIENTLY EXTRACT RELEVANT NETWORK DATA

- › Full PCAP
  - › The amount of data is huge
  - › Difficult to monitor the entire network
- › Alternatively, we aim for information concentrators within the network
  - › DNS servers
  - › IAM servers
  - › Internal NetFlow
  - › ...



**Bank network**  
**Internal DNS data**

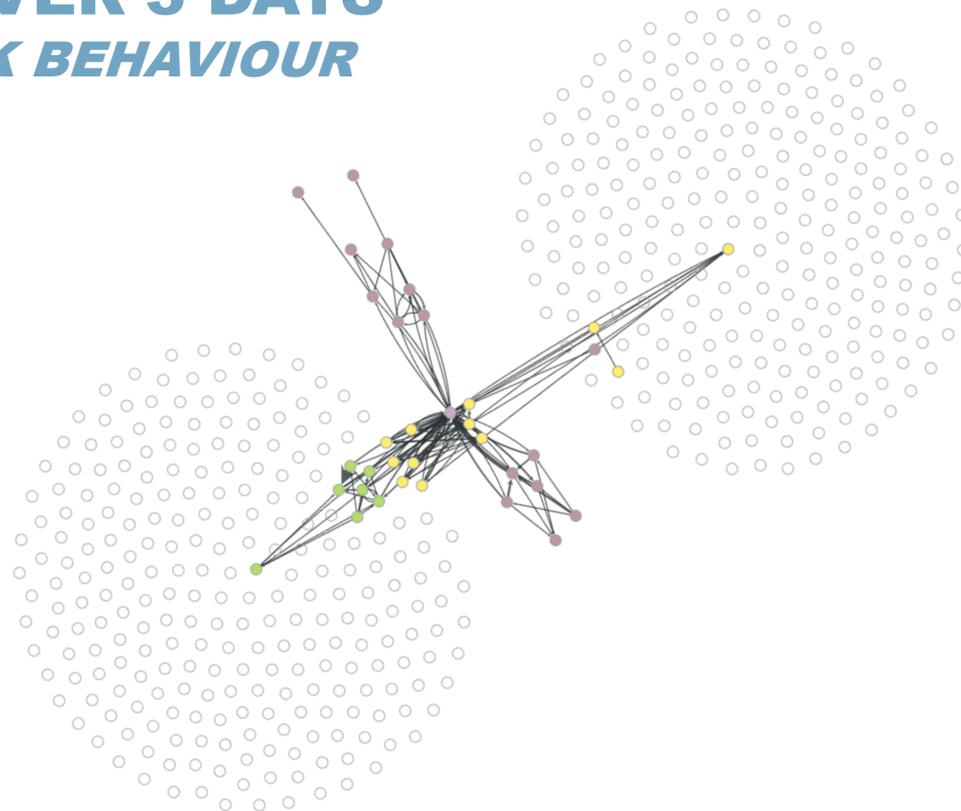
# ANOMALY DETECTION OVER 3 DAYS *BASED ON INTERNAL NETWORK BEHAVIOUR*

› Friday



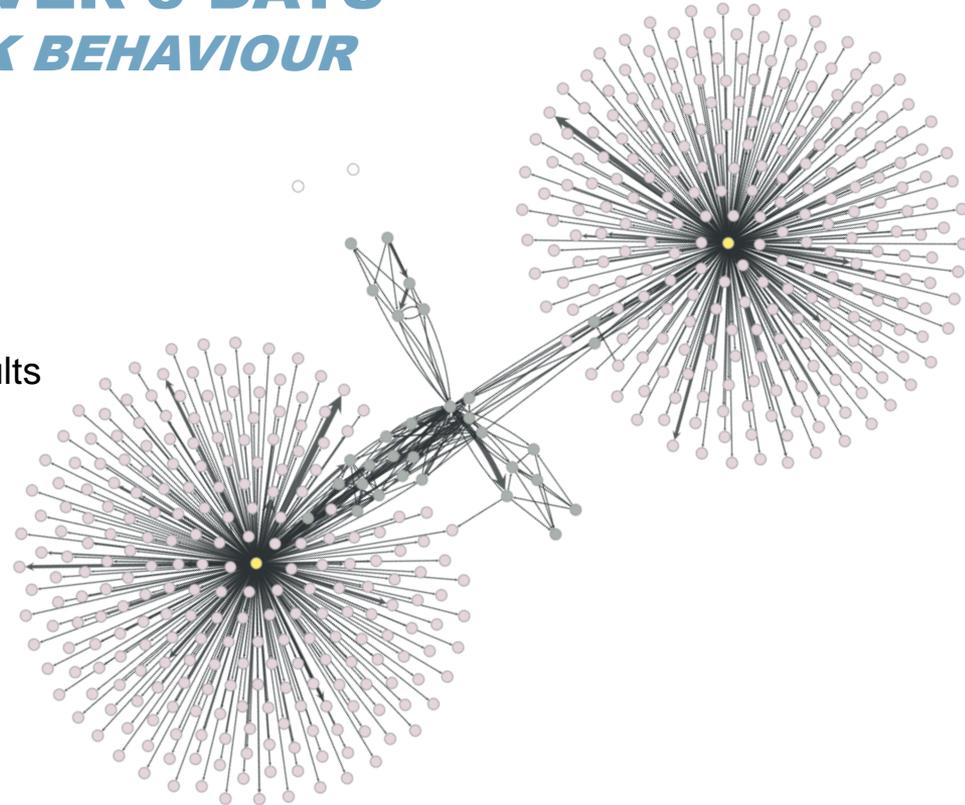
# ANOMALY DETECTION OVER 3 DAYS *BASED ON INTERNAL NETWORK BEHAVIOUR*

- › Friday
- › Saturday



# ANOMALY DETECTION OVER 3 DAYS *BASED ON INTERNAL NETWORK BEHAVIOUR*

- › Friday
- › Saturday
- › Sunday - Port scan influences the clustering results

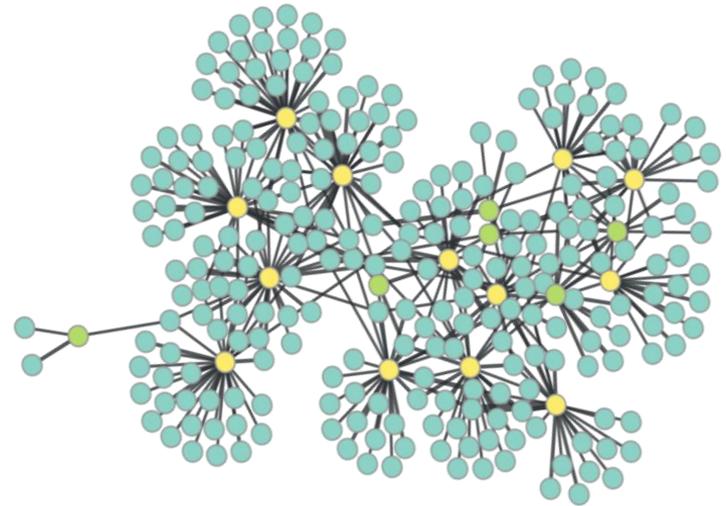


# MODEL INTERNAL NETWORK BEHAVIOUR TO DETECT ANOMALOUS HOSTS

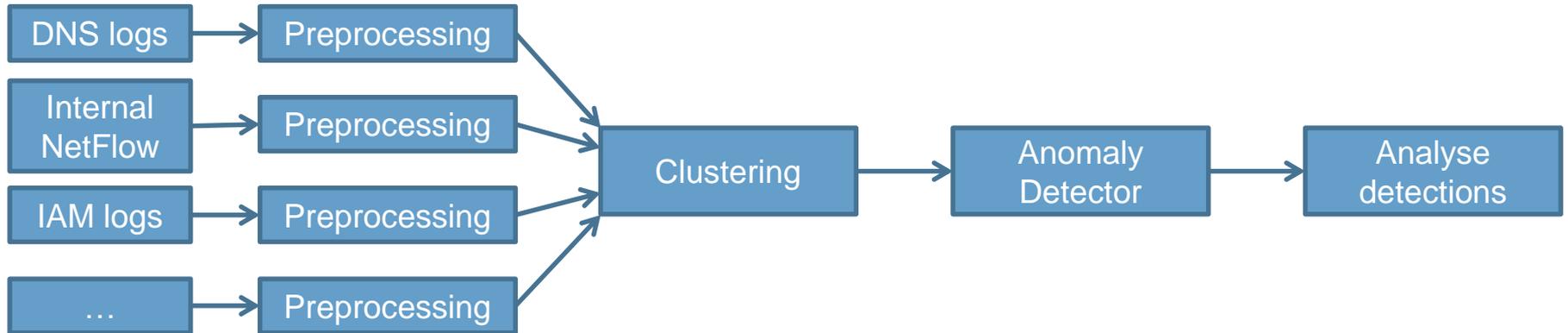
- › Modelling individual hosts:
  - › Many false-positives
  - › Minor day to day differences in network behaviour will be considered anomalous
- › Modelling all hosts together:
  - › Many false-negatives
  - › E.g. a workstation displaying server like network behaviour will not be detected

# CLUSTERING HOSTS WITH SIMILAR CONNECTION BEHAVIOUR

- › Groups of hosts often behave similarly
  - › E.g. mail servers, printers, workstations, mobile devices
- › Determine 'normal' behaviour **per cluster** and observe when individual host deviates

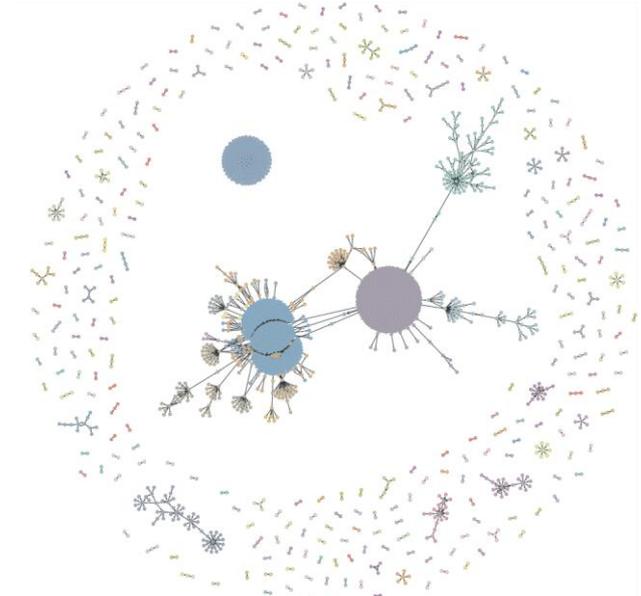


# MODULAR MONITORING AND DETECTION FRAMEWORK FOR INTERNAL NETWORK SOURCES



# CLUSTERING – LOUVAIN COMMUNITY DETECTION

- › Find communities within a network
  - › Many connections within the community
  - › Little connections across different communities
- › Efficient algorithm
- › Anomaly detection:
  - › *Rare to see connections between different communities*

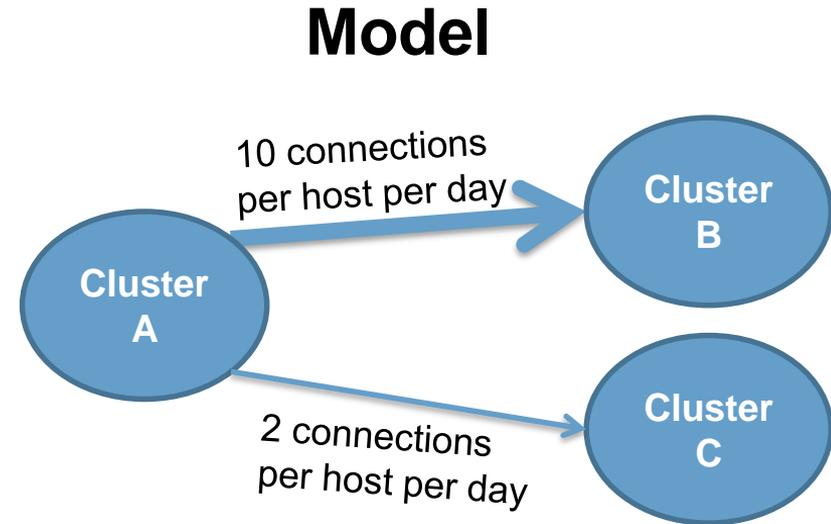


**Bank network**

Colours indicate the different clusters

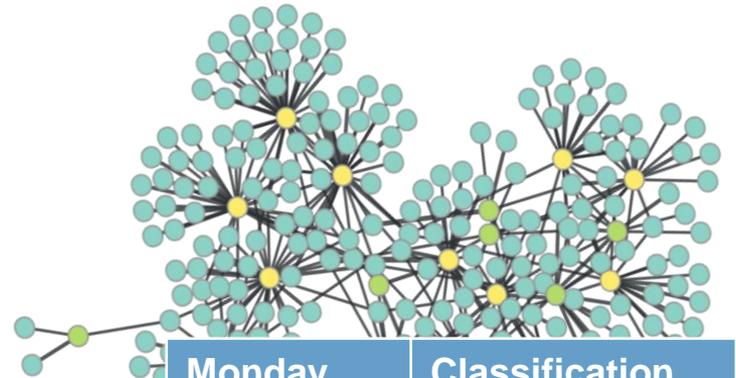
# MODELLING THE CONNECTION BEHAVIOUR OF THE CLUSTERED HOSTS

- › For each cluster we create a different model
  - › This models displays the average number of connections to each of the other clusters
- › For all hosts we will determine the number of queries to the each of the different clusters on a new day
- › These numbers are then compared to the averages of the model
  - › Number of standard deviations from the mean



# CLIENT SERVER CLASSIFICATION ALGORITHM BASED ON INTERNAL NETWORK BEHAVIOUR

- › Support vector machine based on DNS data
  - › Number of times a host is queried
  - › Query types
  - › Centrality of host within the network
  - › ...



Friday		Classification	
		Client	Server
Type	Client	99,7 %	0,35 %
	Server	3,2 %	96,7 %

Anomalies

Monday		Classification	
		Client	Server
Type	Client	99,7 %	3,4 %
	Server	8,4 %	91,6 %

# CONCLUSIONS AND TAKEAWAYS

- › **Internal network traffic:**
  - › Valuable source of information for targeted attack detection
  
- › **Behavioural based clustering techniques:**
  - › Improve anomaly detection algorithms



THANK YOU FOR YOUR  
ATTENTION

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