

CAMNEP

Network Intrusion Detection System

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Network Intrusion Detection

- **Threats** in the computer networks
 - Sophisticated, possible government-backed attacks
 - industrial espionage, political attacks
 - Organized crime – credit card fraud, banking attacks, spam
- **Goal: Detection** of **computer attacks** by analyzing the **structure** of network traffic
 - privacy preserving
 - applicable on ciphered traffic (for financial/government sites)
 - similar to the analysis of phone bills
- **Challenges:**
 - High traffic speeds (millions of connections per second)
 - High number of increasingly sophisticated, evasive attacks

Anomaly Detection vs. Signatures

Signature matching

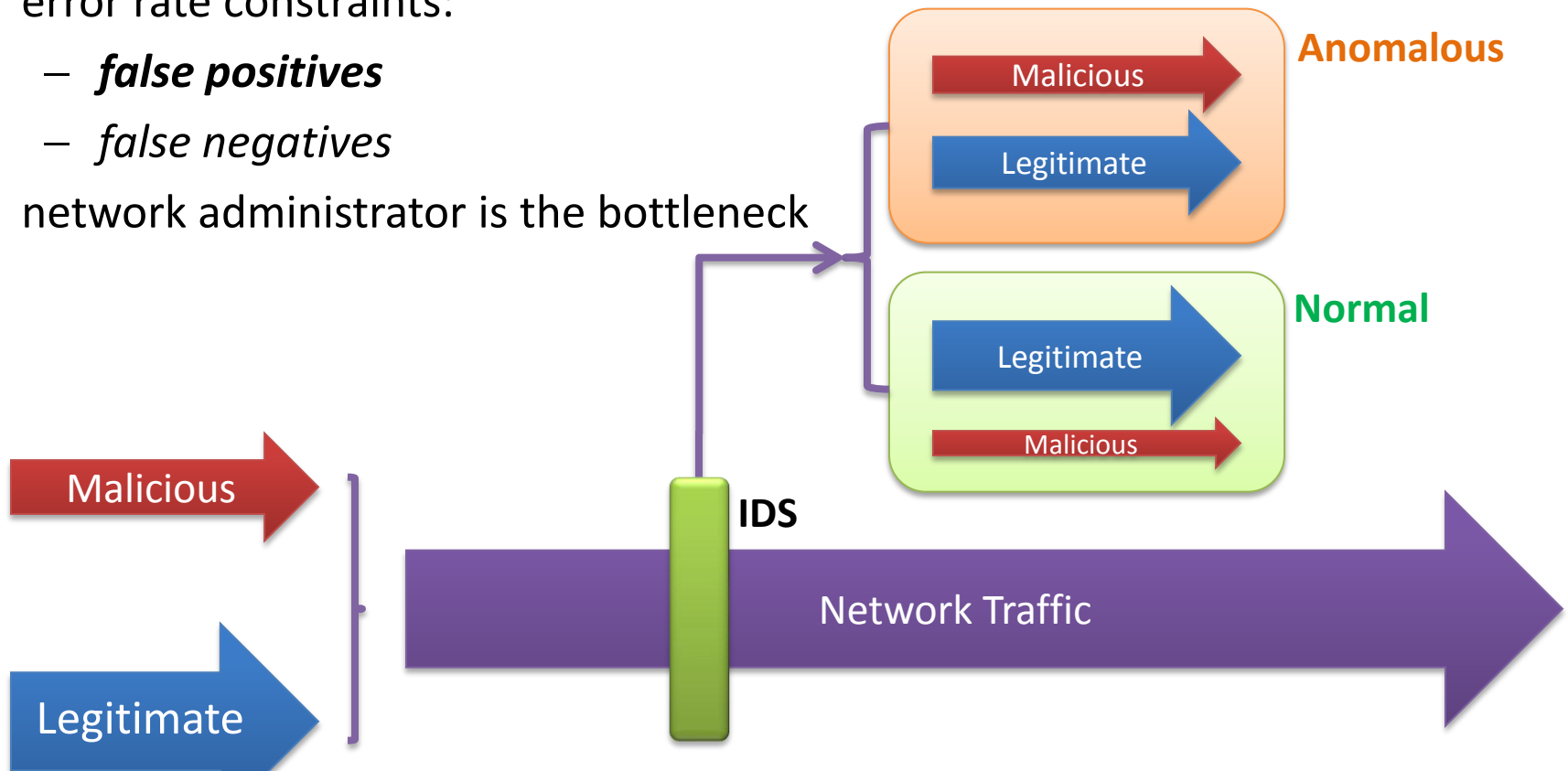
- Historically validated
- Widely deployed
- Verifiable & Stable
- Number of patterns
- Scaling
- Management
- New threats detection

Anomaly detection

- No patterns
- New threats detection
- Scaling
- **Error Rate/Sensitivity**
- **Verifiability**
- **Stability**
- **Management**

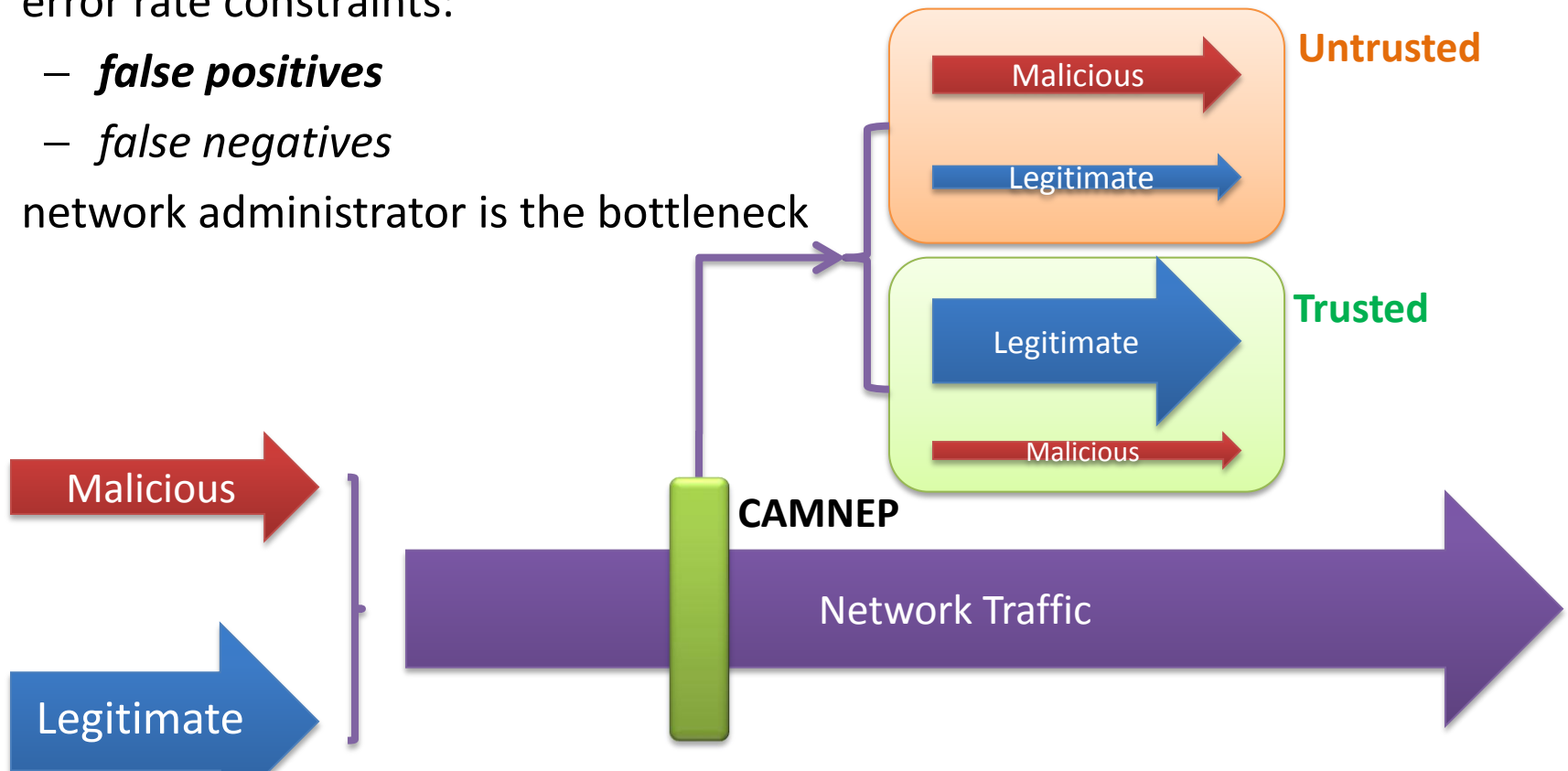
Anomaly detection

- real-time constraints: gigabit/sec, **2000 – 5000** flows/sec
- error rate constraints:
 - *false positives*
 - *false negatives*
- network administrator is the bottleneck

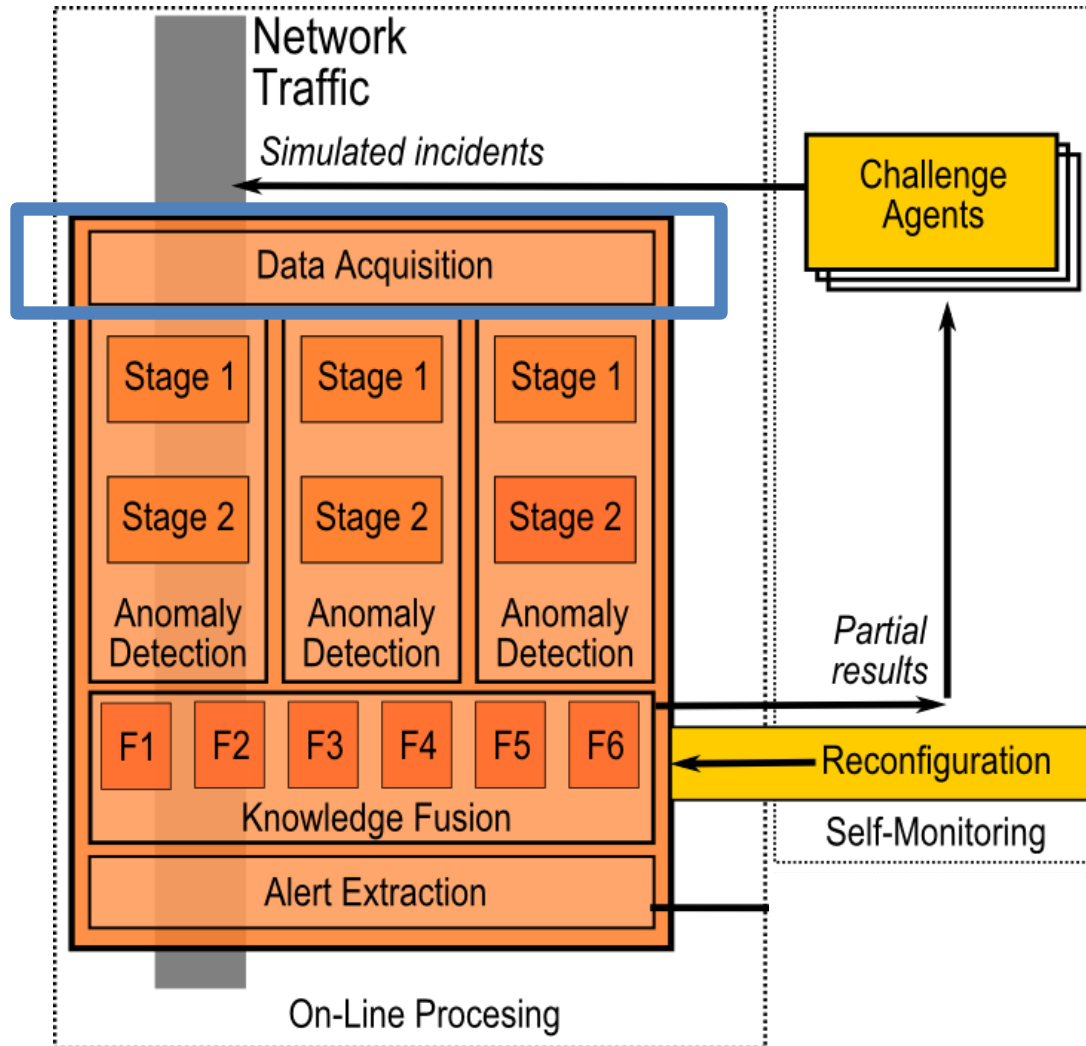


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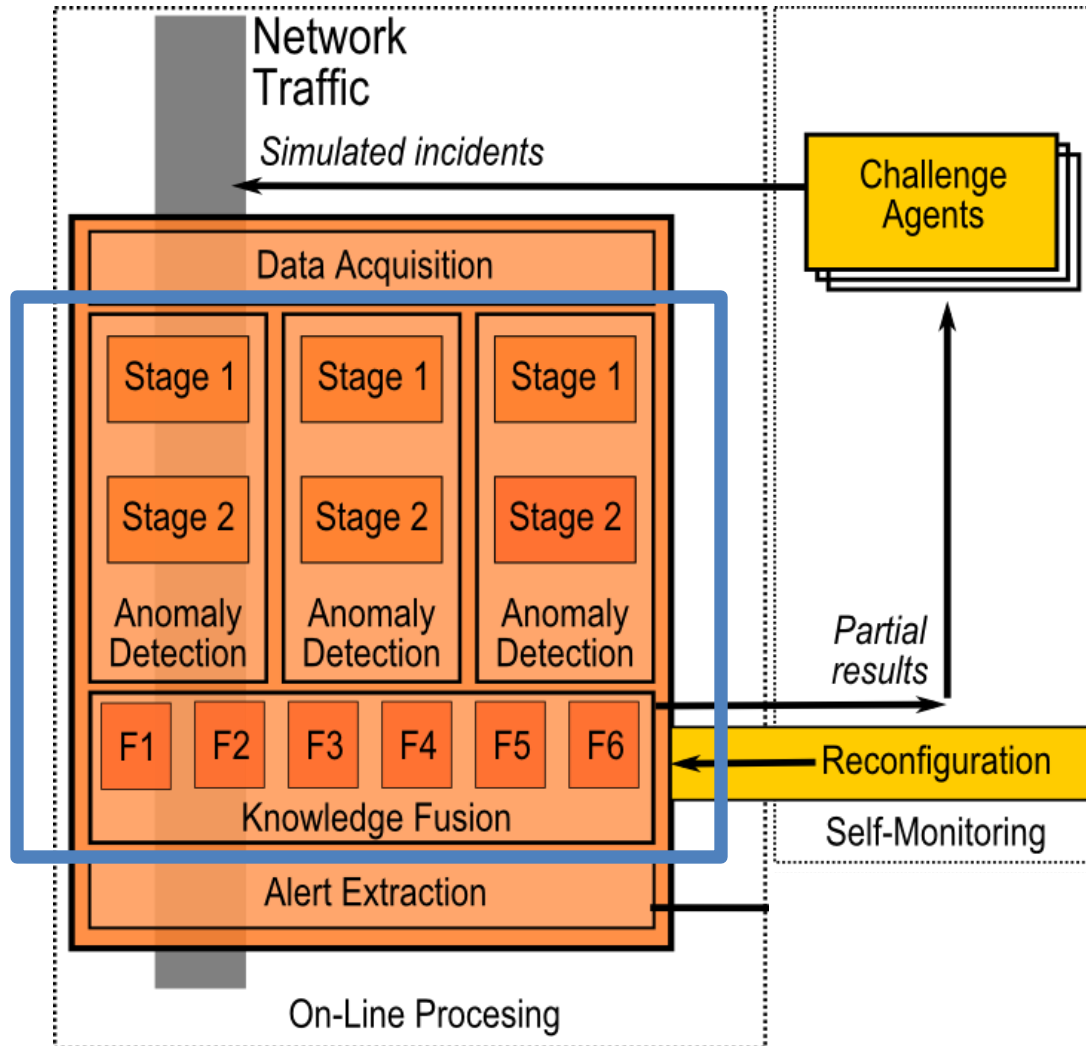


Architecture



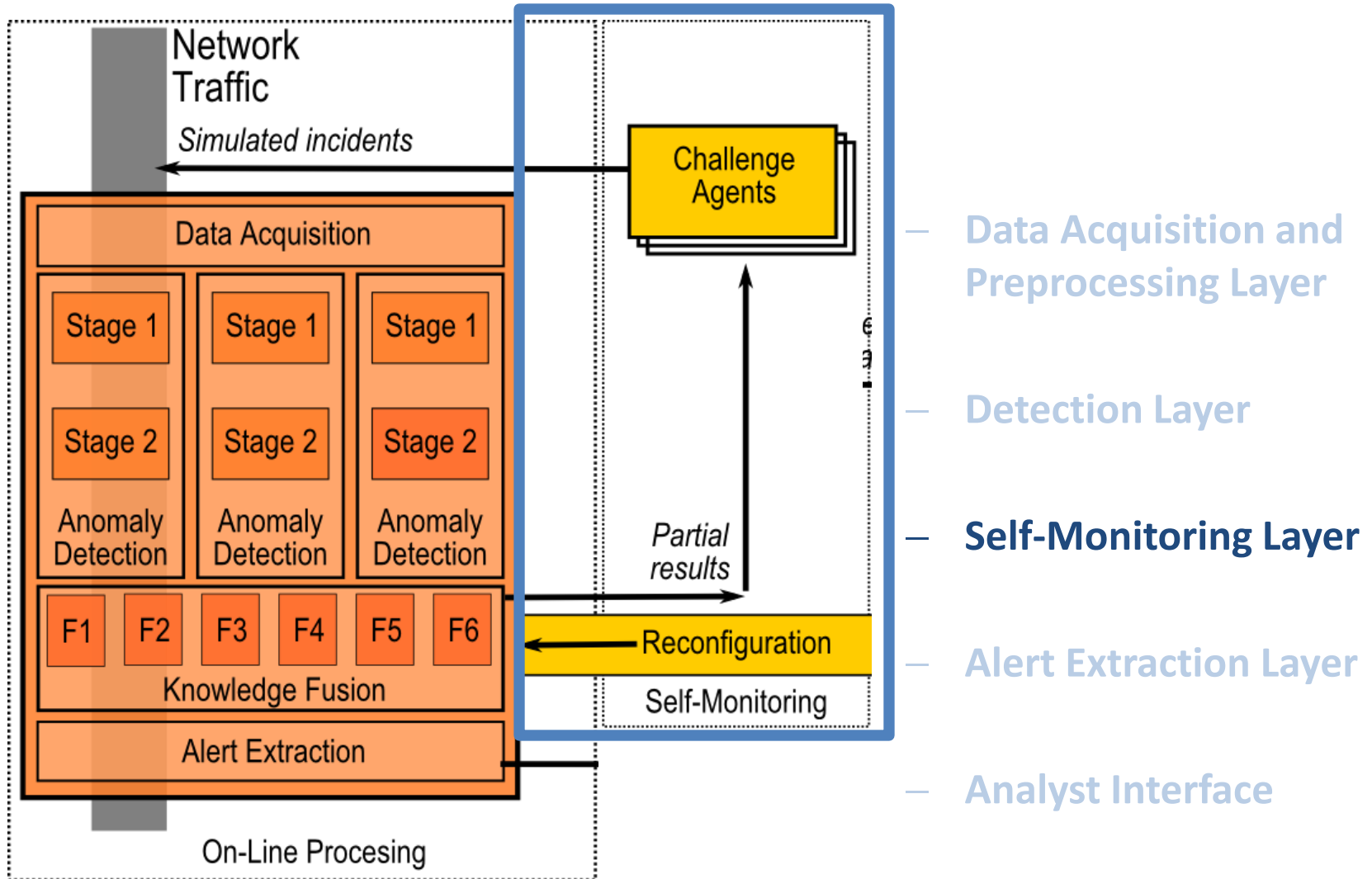
- Data Acquisition and Preprocessing Layer
- Detection Layer
- Self-Monitoring Layer
- Alert Extraction Layer
- Analyst Interface

Architecture

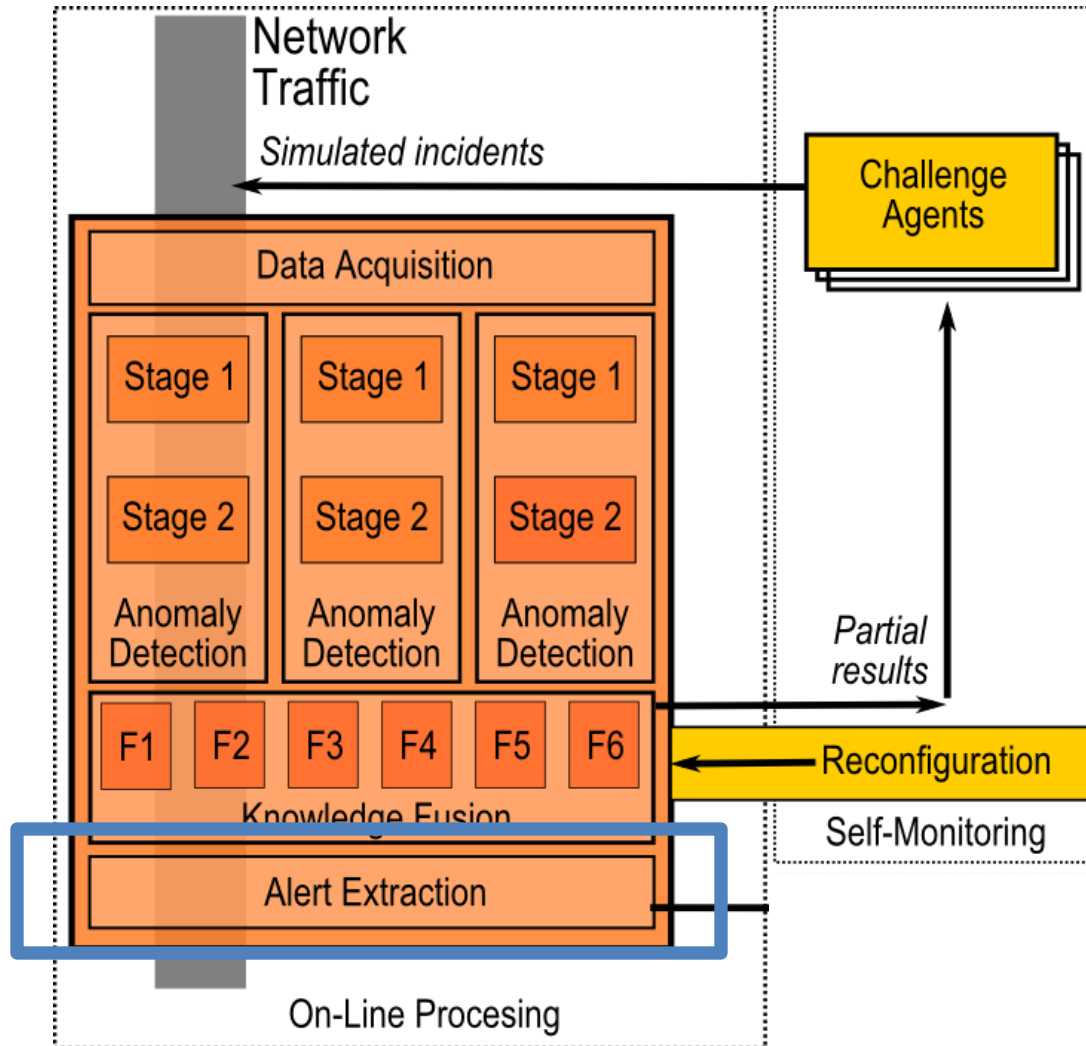


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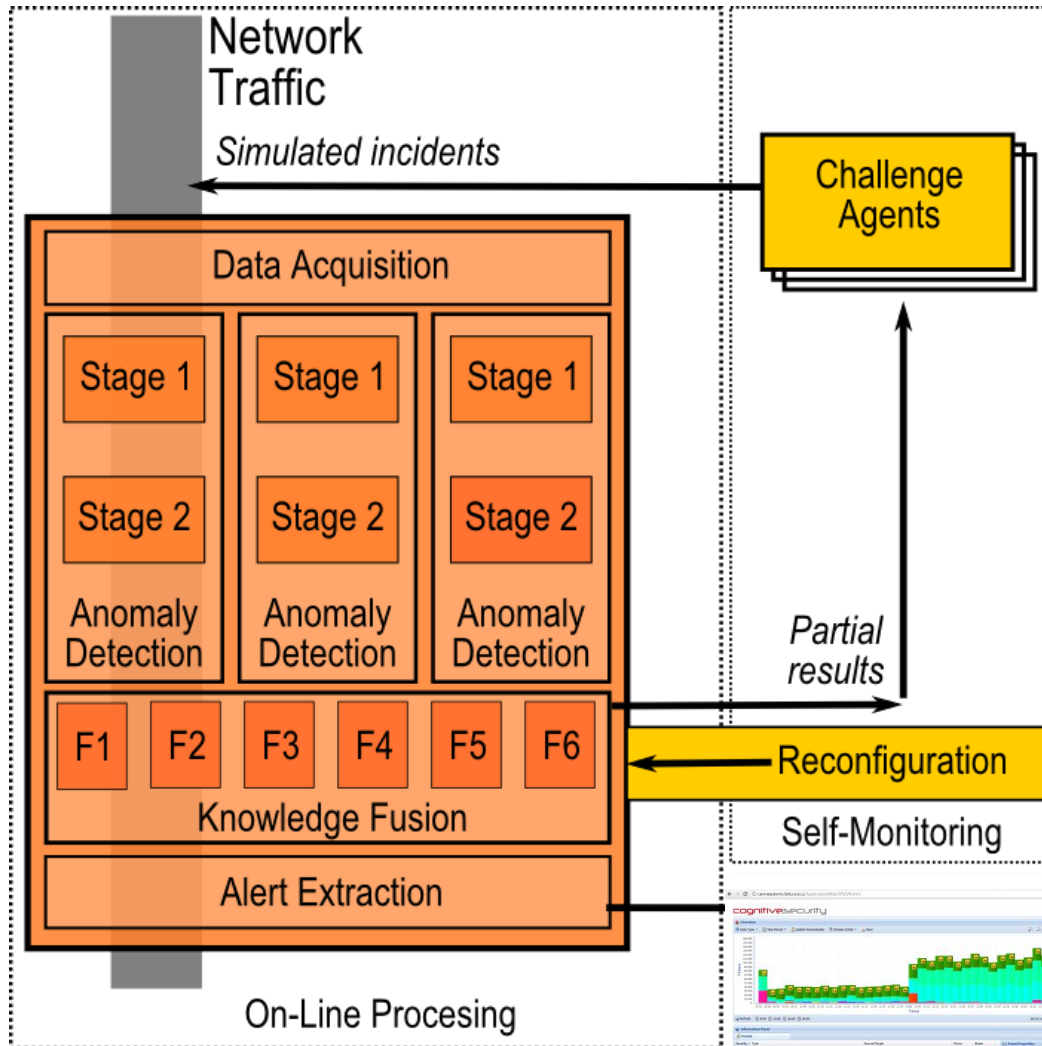


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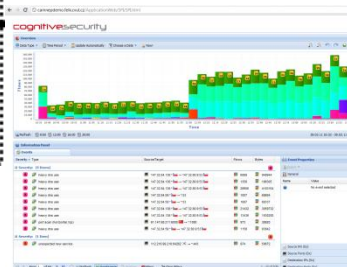


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Detection Layer

Stage 1

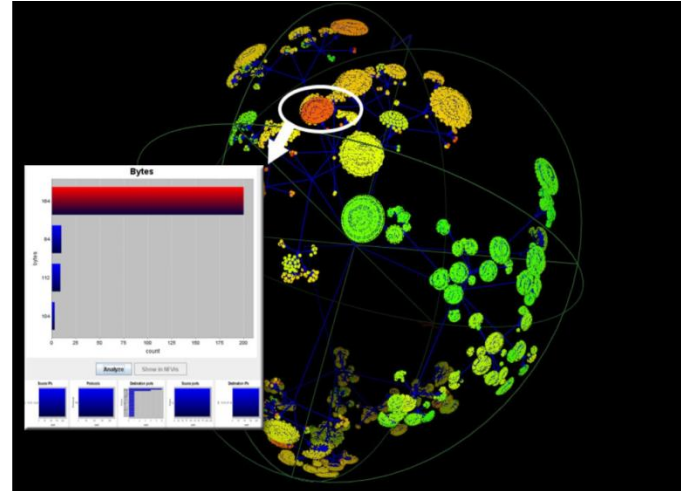
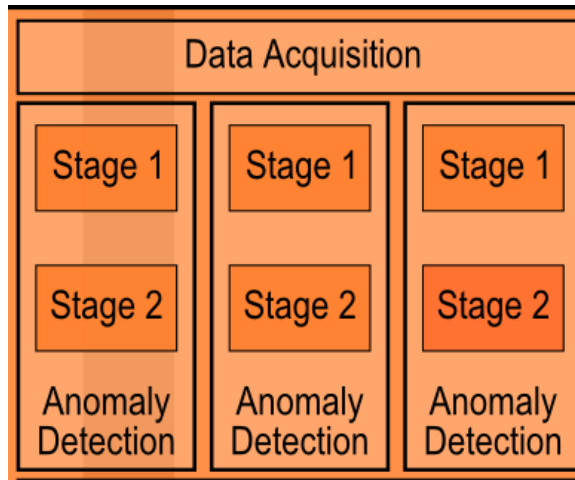
- **Anomaly Detection:** Predicting current network behavior from the history and looking for deviations

Detection Layer



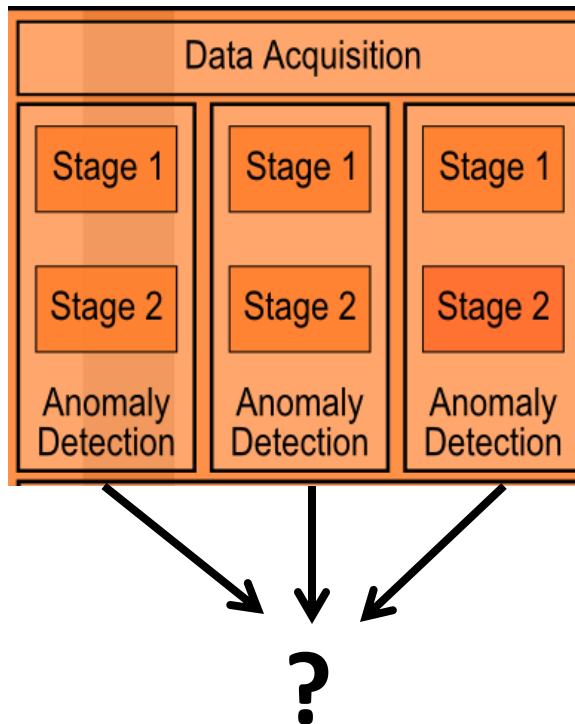
- **Multi-Algorithm Anomaly Detection:**
- Entropy modeling
- Trend modeling
- Volume modeling
- Principal components analysis
- Information-theoretical measures
- ...

Inside Modern NBA System



- **Trust Modeling: Synthesizing the Anomaly detection data across the algorithms and over time**
- **Reduction of false positives by:**
 - Multi-source aggregation
 - Historical experience aggregation

Key Problem n. 1



- How to find optimal configuration = select aggregation function?
- We have many aggregation functions, but which is the best?

Option 1: Offline Configuration

- Offline optimization of internal parameters
- Difficulties with training data
 - Expensive manual labeling
 - No fully labeled and representative dataset
 - Manually labeled data is biased
 - Legal issues with public sharing of data
- Offline configuration results can not capture the dynamic character of the network
 - M. Rehak, E. Staab, M. Pechoucek, **J. Stiborek**, M. Grill, and K. Bartos, “Dynamic information source selection for intrusion detection systems”, International Foundation for Autonomous Agents and Multiagent Systems, 2009, pp. 1009–1016.

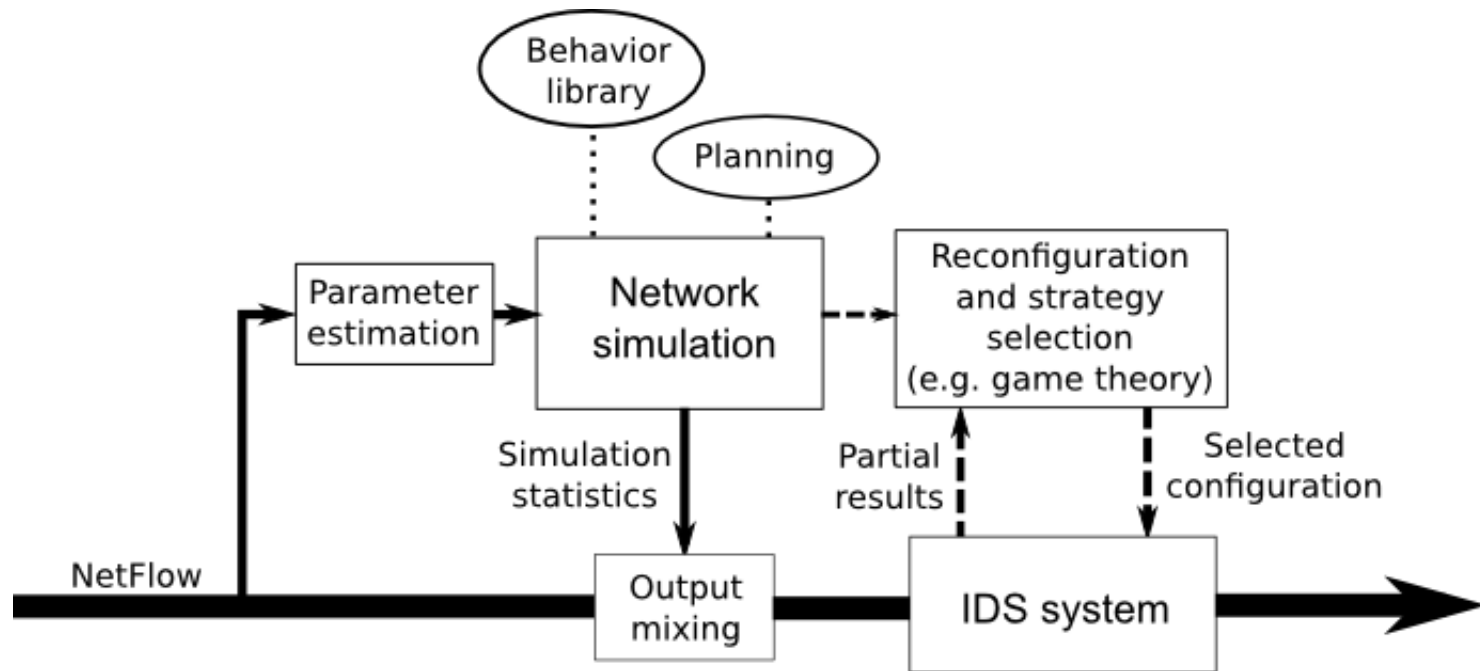
Option 2: Online Configuration

- Tunes parameters according to the current state of the specific network
 - Ground truth is undeterminable within the time constraints imposed
 - Manual, supervised approach is infeasible
- Malicious vs. Anomalous problem
 - Definition of malicious behavior depends on the specific network's security policy
 - Malicious vs. Anomalous fine-tuning problem

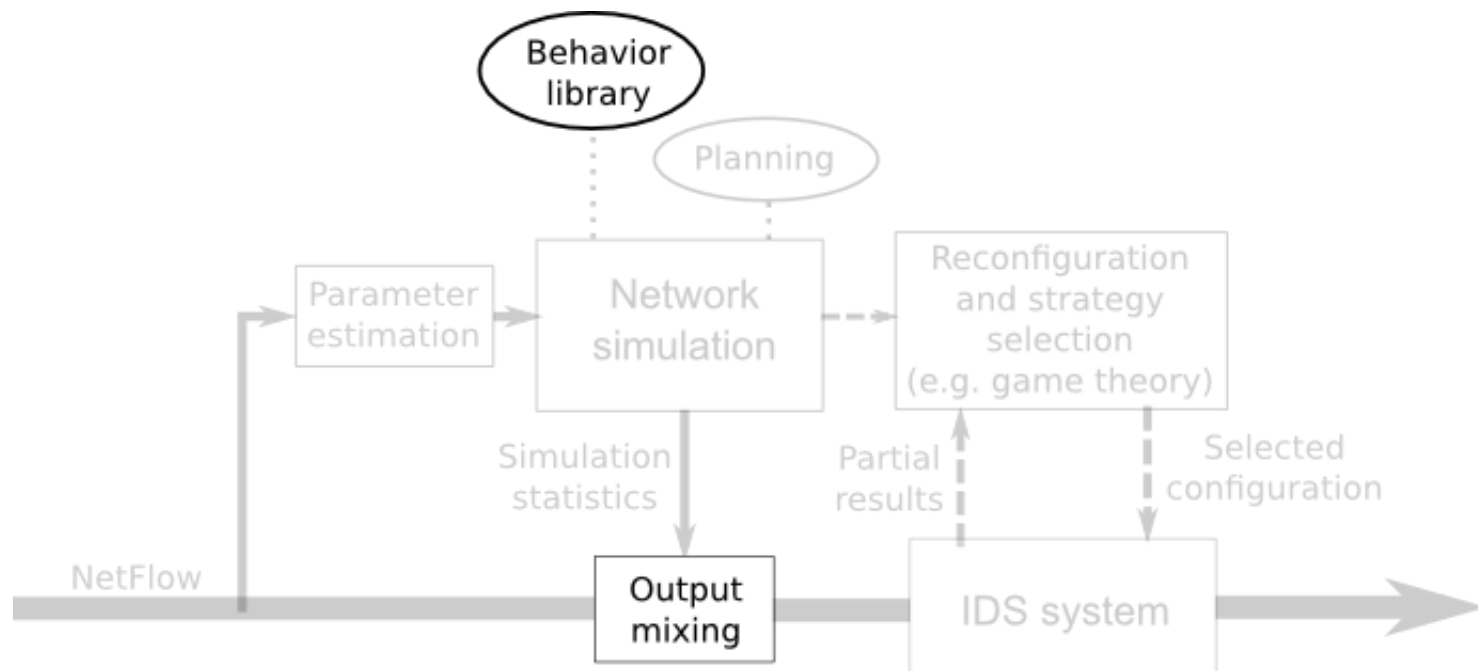
Online Local Adaptation: Research Problems

- Representativeness & Coherence with background traffic
 - Right DNS server IP, representative OS, realistic user profiles of simulated behavior
- Second-order simulation effects (side effects)
- Timeliness

Adaptive Network Simulation

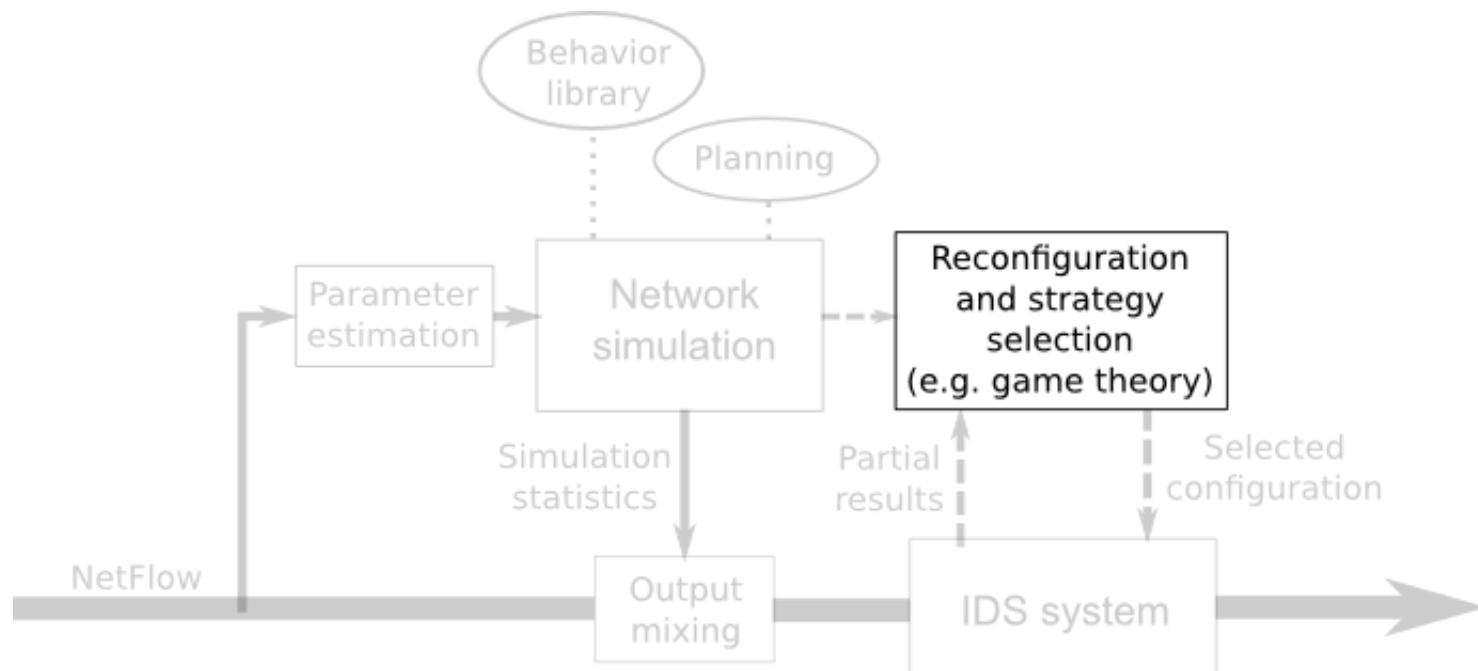


Adaptive Network Simulation



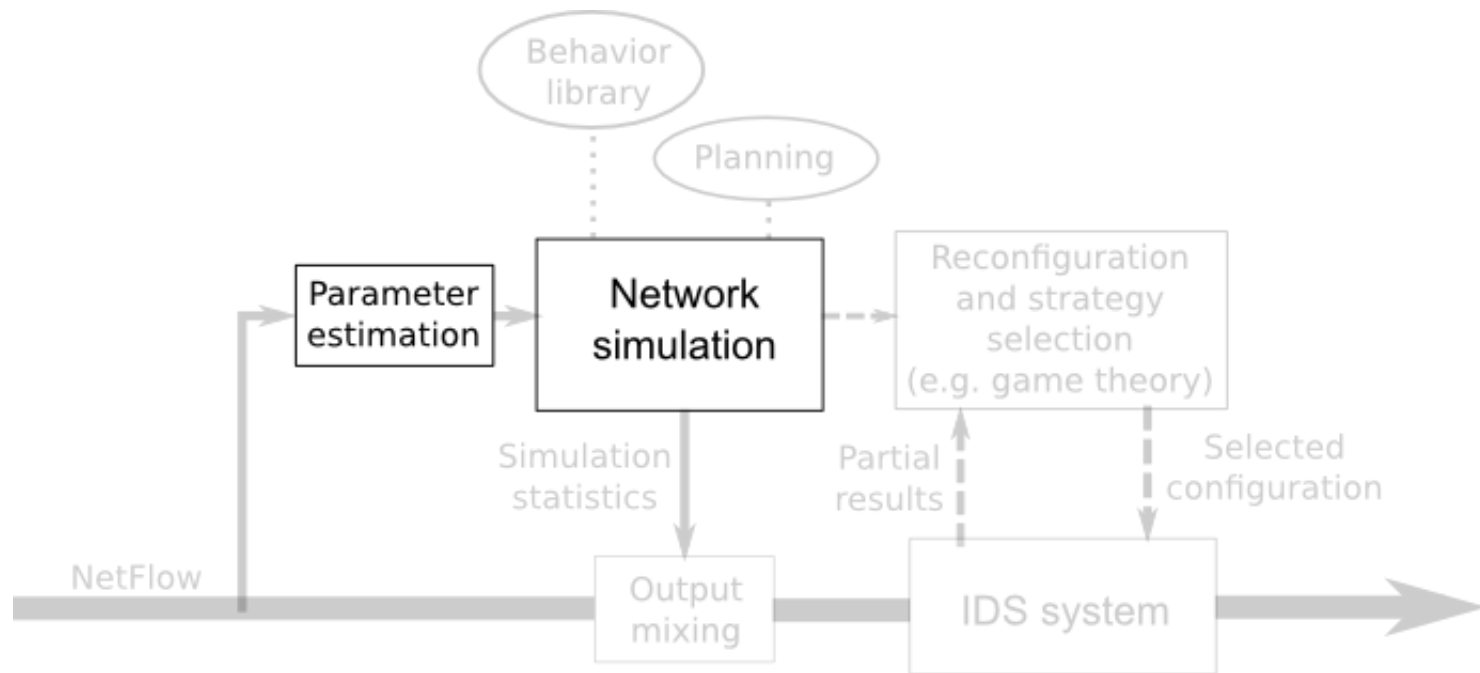
- M. Rehak, M. Pechoucek, M. Grill, J. Stiborek, K. Bartos, and P. Celeda, “Adaptive multiagent system for network traffic monitoring,” *Intelligent Systems*, IEEE, vol. 24, 2009, pp. 16–25.

Adaptive Network Simulation



- J. Stiborek, M. Grill, M. Rehak, K. Bartos, and J. Jusko, “Game Theoretical Adaptation Model for Intrusion Detection System,” Advances on Practical Applications of Agents and Multi-Agent Systems 2012, pp. 201–210.

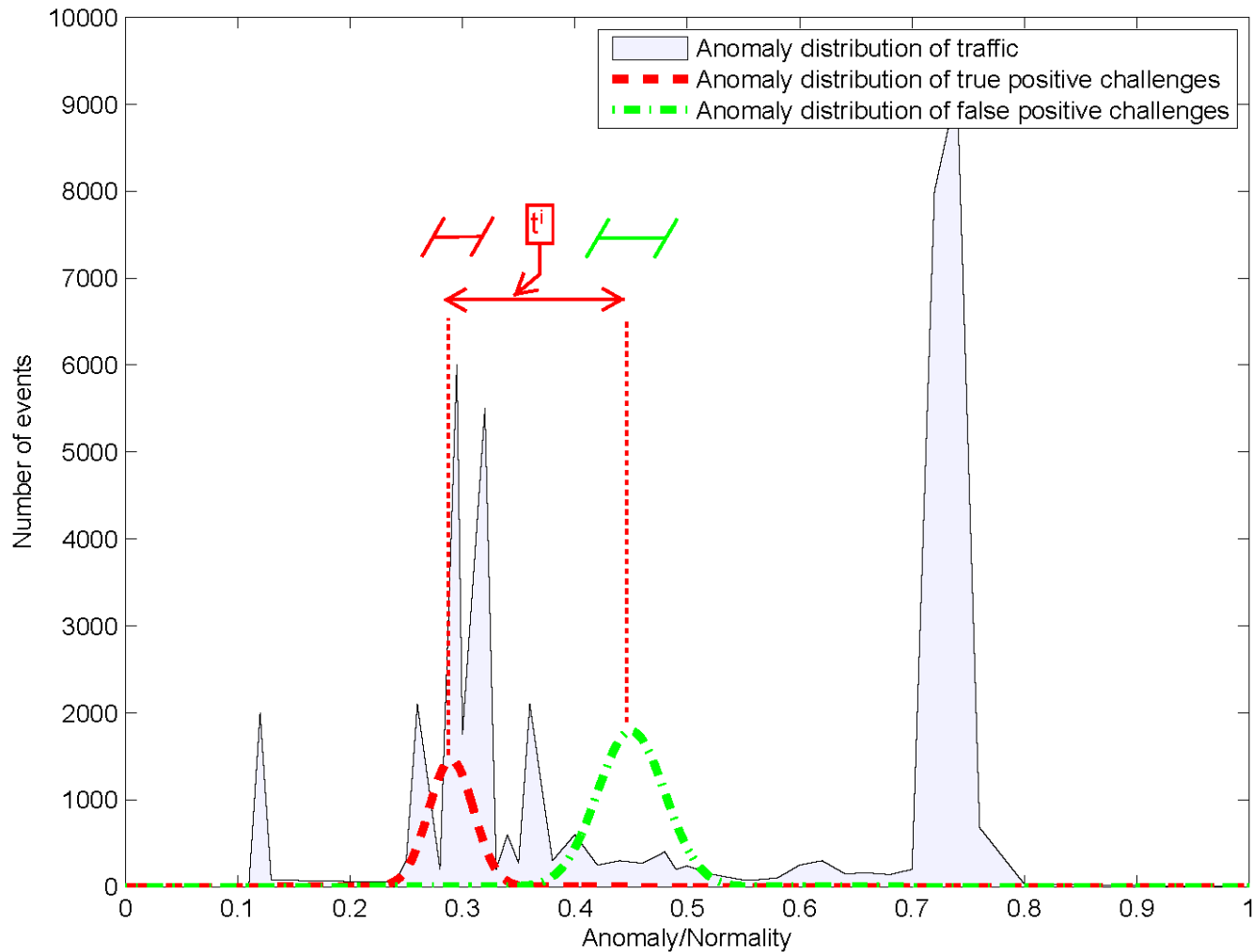
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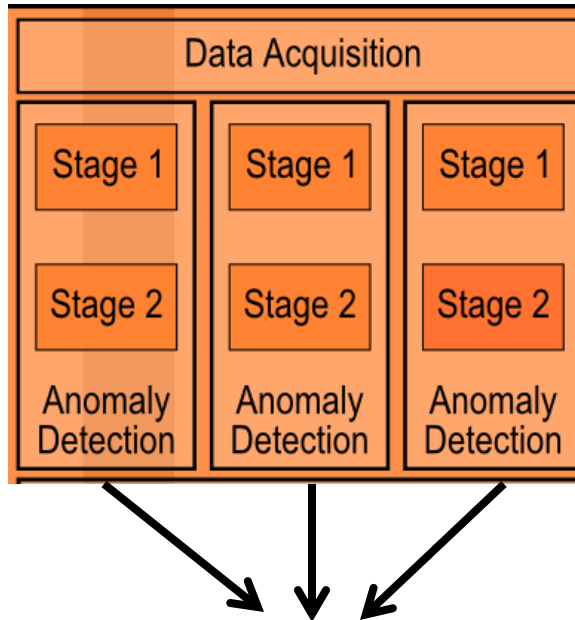
Adaptive Network Simulation

- Estimate joint probability distribution
 - $p(\text{dPort}, \text{dIP}, \text{sPort}, \text{\#bytes (req./resp.)}, \text{\#packets(req./resp.)}, \text{thinkTime})$
 - Estimated probability distribution has to follow the correct timeliness
- Generate new samples – Metropolitan Hastings algorithm

Network Simulation Control

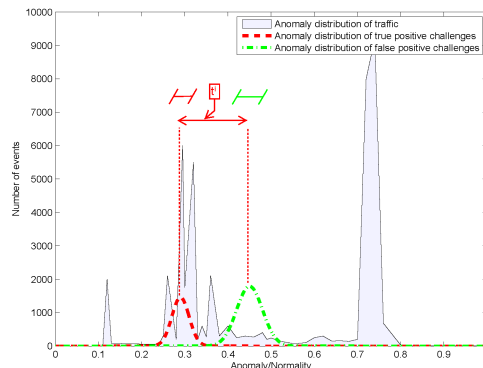


Key Problem n. 2

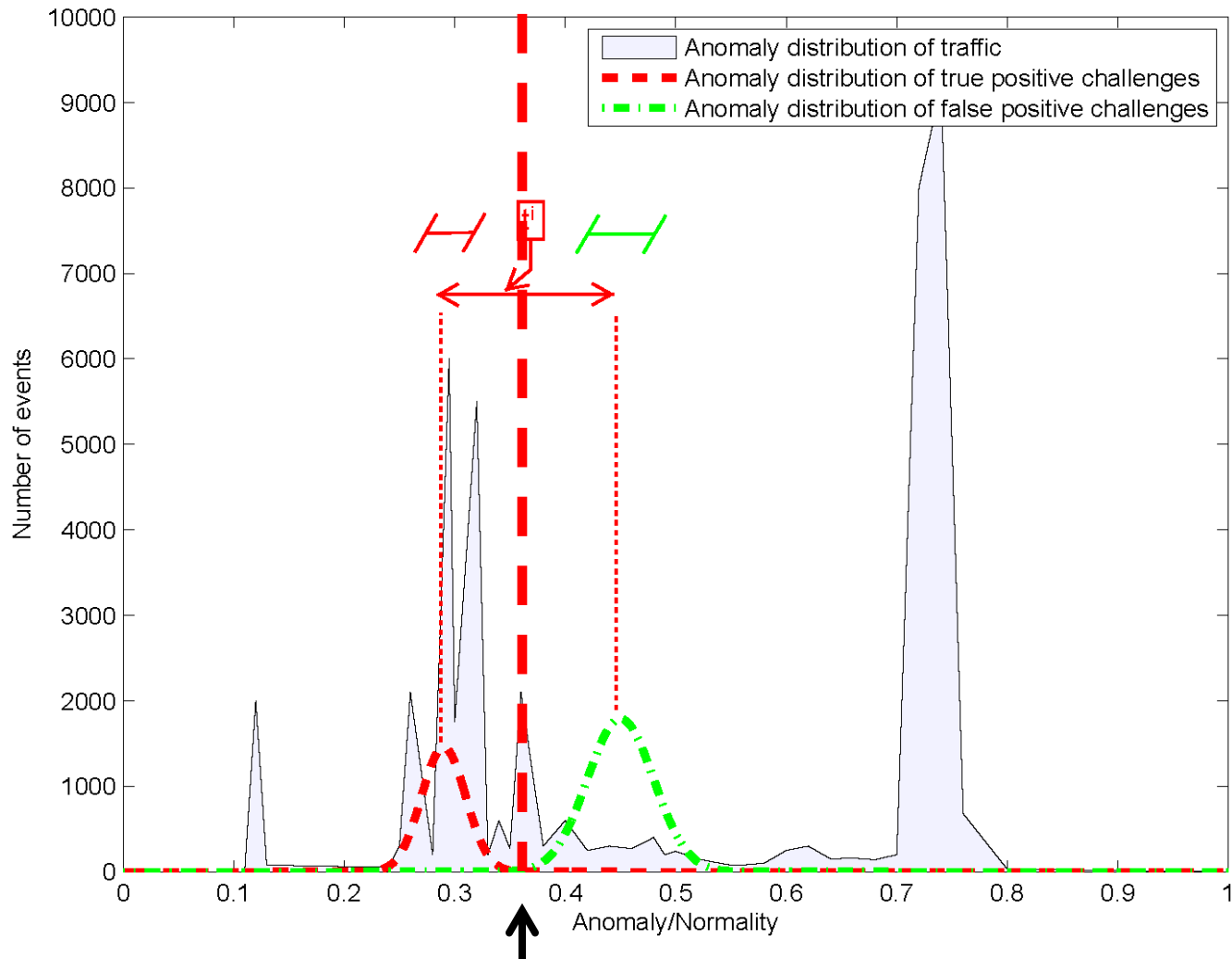


- If we have aggregated values, where is the **threshold** separating anomalous and normal traffic?

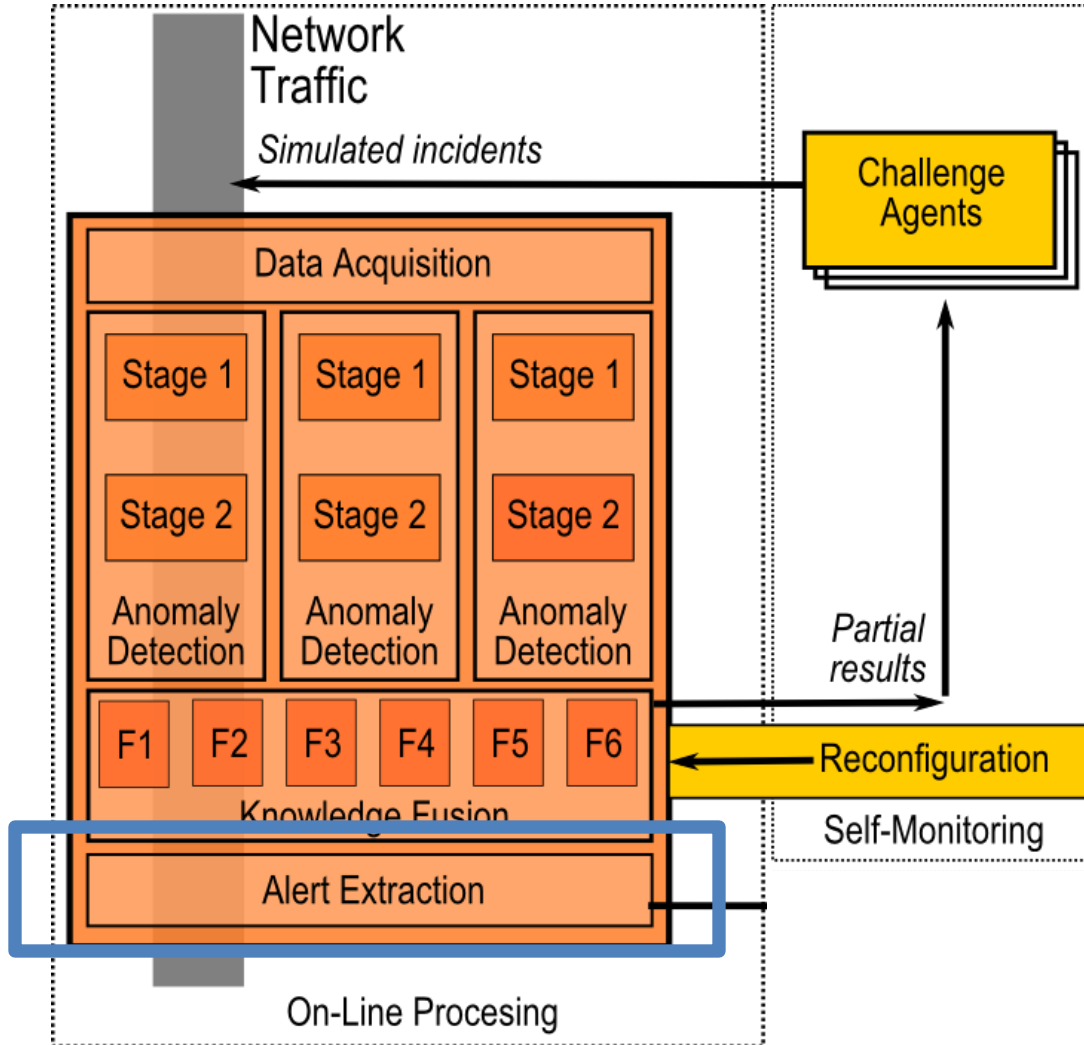
- **Solution: Network simulation**



Network Simulation Control



Events Processing Layer

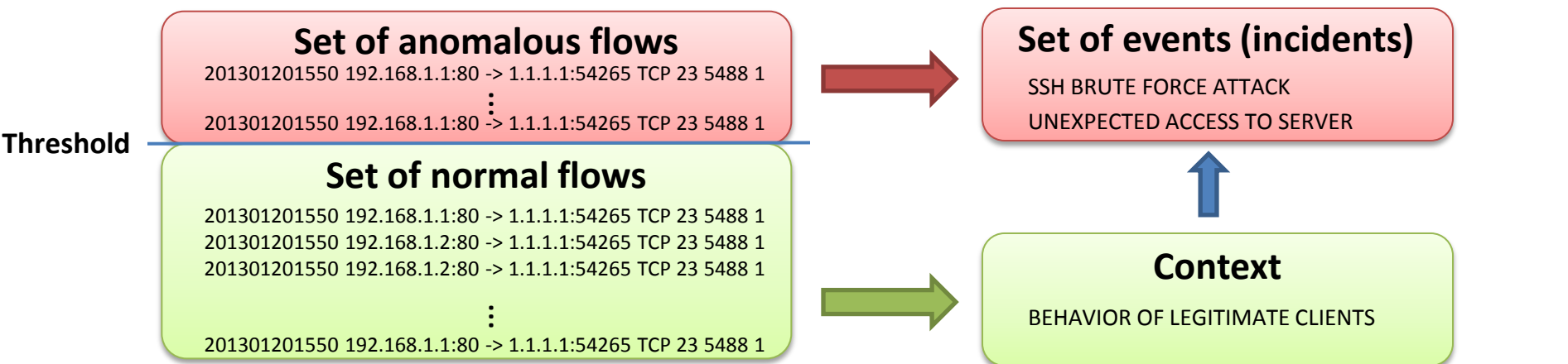


Events Processing Layer

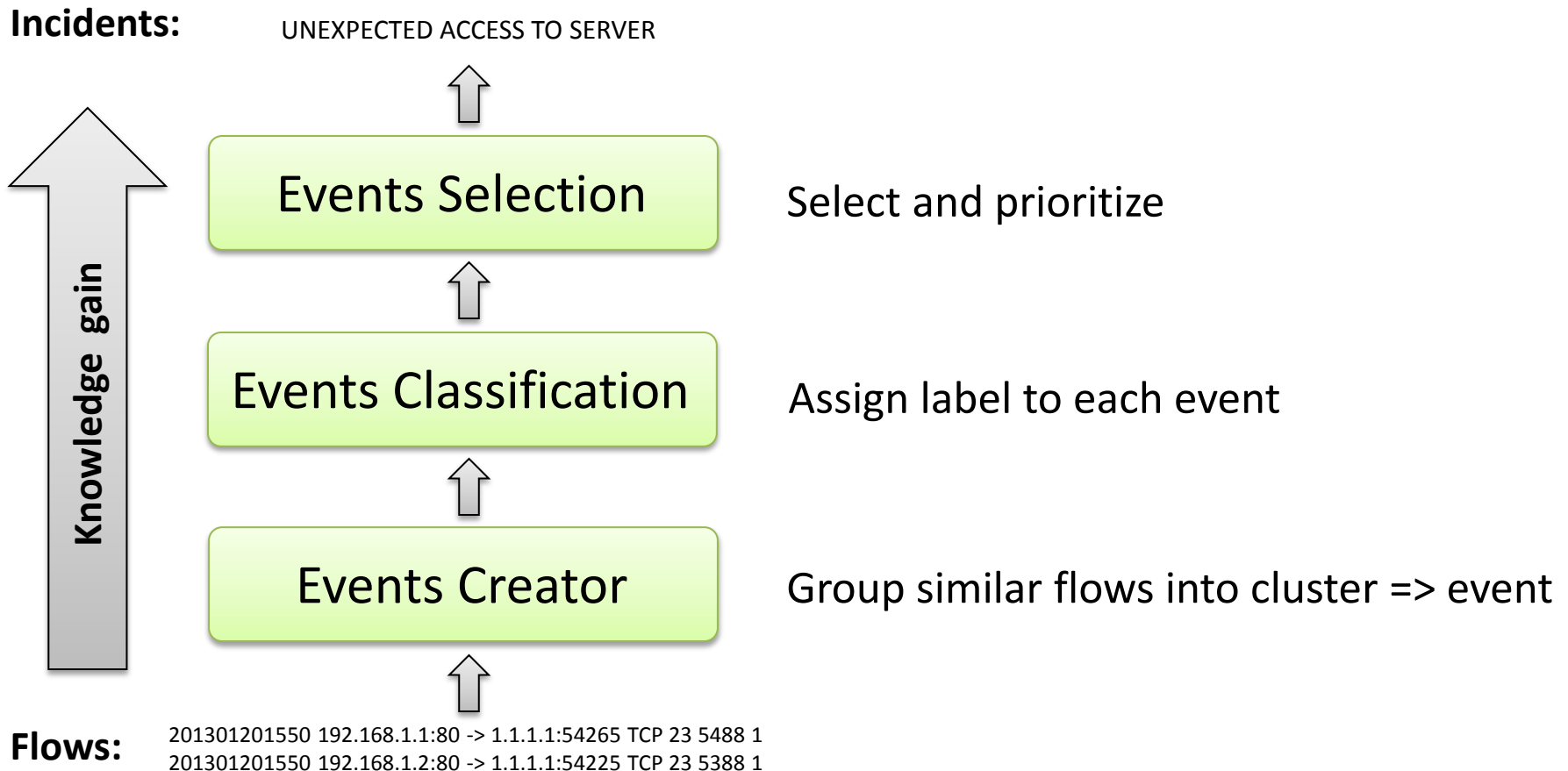
- **Goal: Network security assessments**
 - Increase the level of abstraction
 - Create a list of security incidents
- **Process** - From raw NetFlow data to high-level incidents

Flows: 1M per 5 minutes

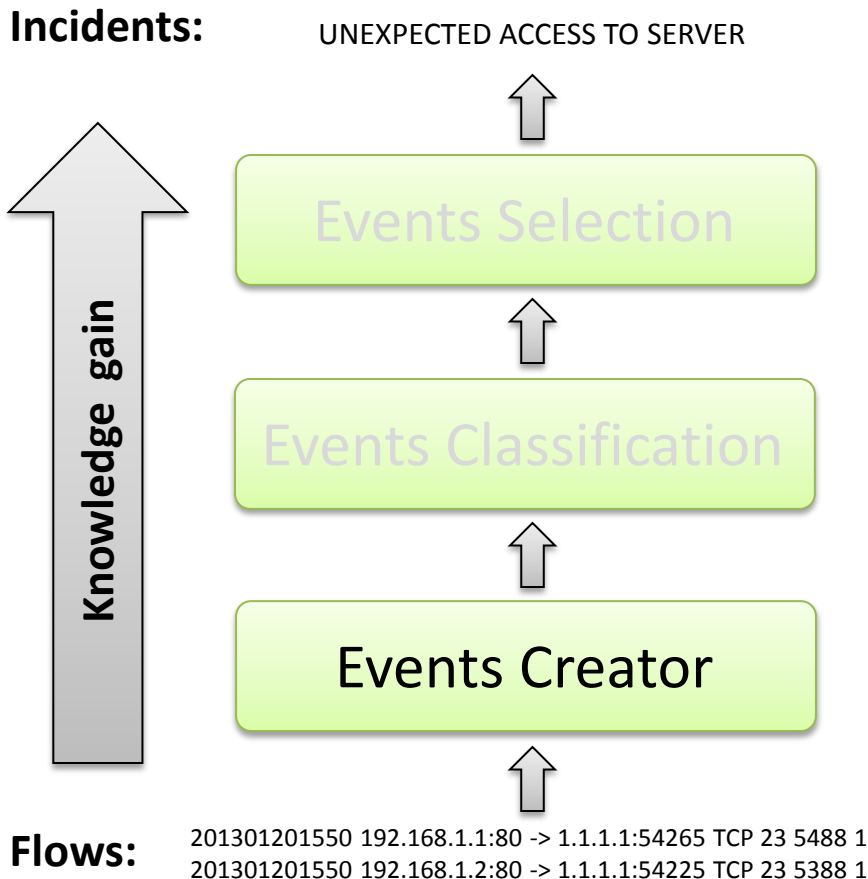
Incidents: 10 per day



Events Processing Layer - Architecture



Events Creator

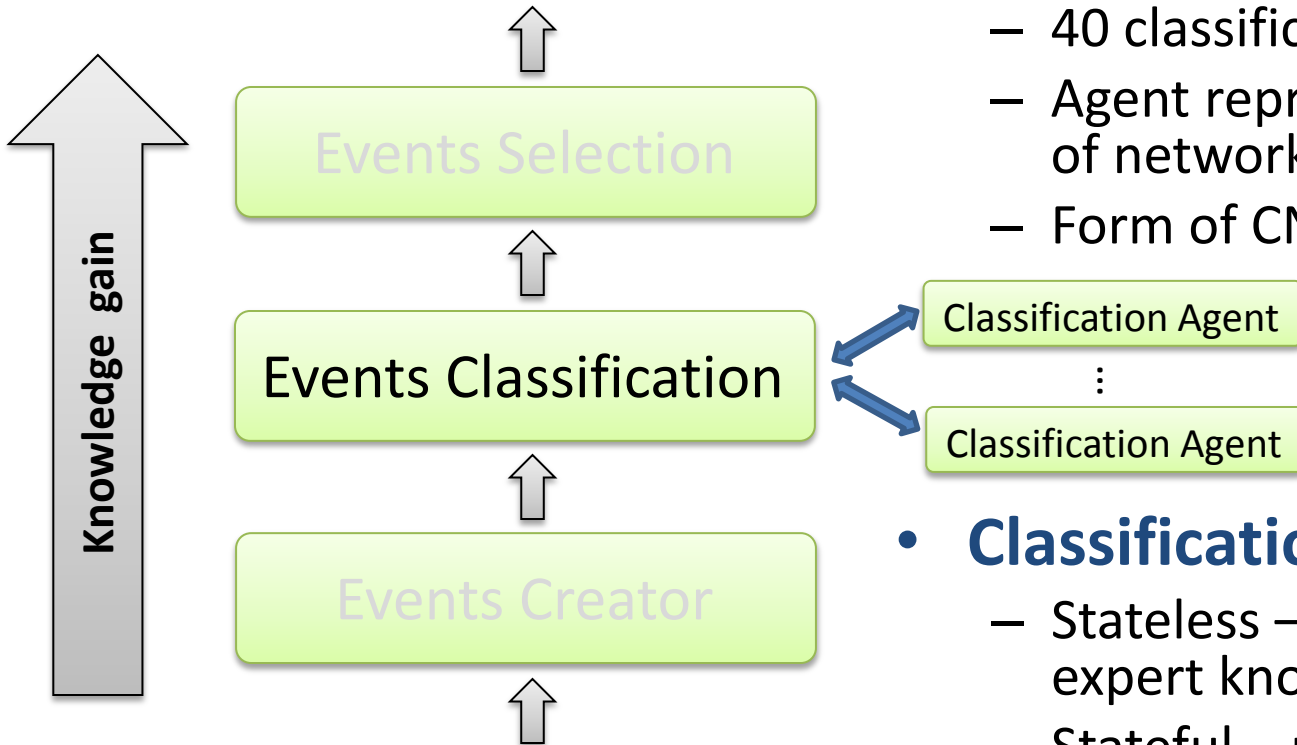


- **Hierarchical clustering**
 - Each cluster represents one type of network behavior
 - Features used in metric:
 - srcIP, srcPrt, dstIP, dstPrt, protocol, bytes
- **Properties**
 - Elementary events
 - High number
 - Small granularity

Events Classification

Incidents:

UNEXPECTED ACCESS TO SERVER



- **Classification**

- 40 classification agents
- Agent represents one type of network behavior
- Form of CNP auction

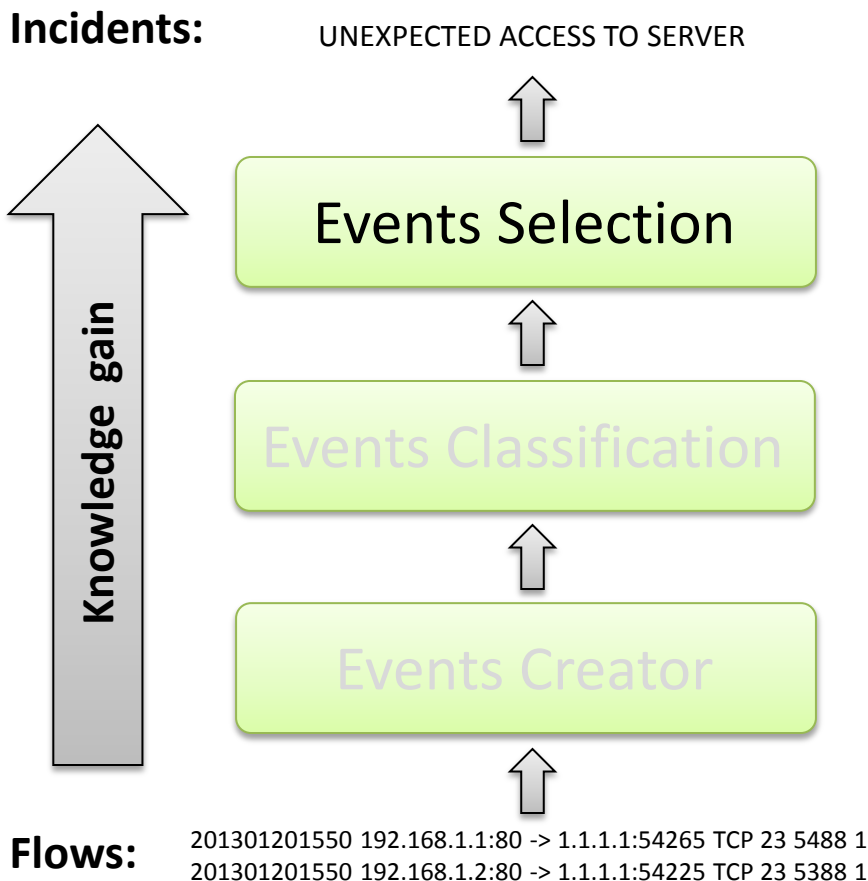
- **Classification Agents**

- Stateless – based on expert knowledge (scan)
- Stateful – uses models
- P2P, persistence

Flows:

201301201550 192.168.1.1:80 -> 1.1.1.1:54265 TCP 23 5488 1
201301201550 192.168.1.2:80 -> 1.1.1.1:54225 TCP 23 5388 1

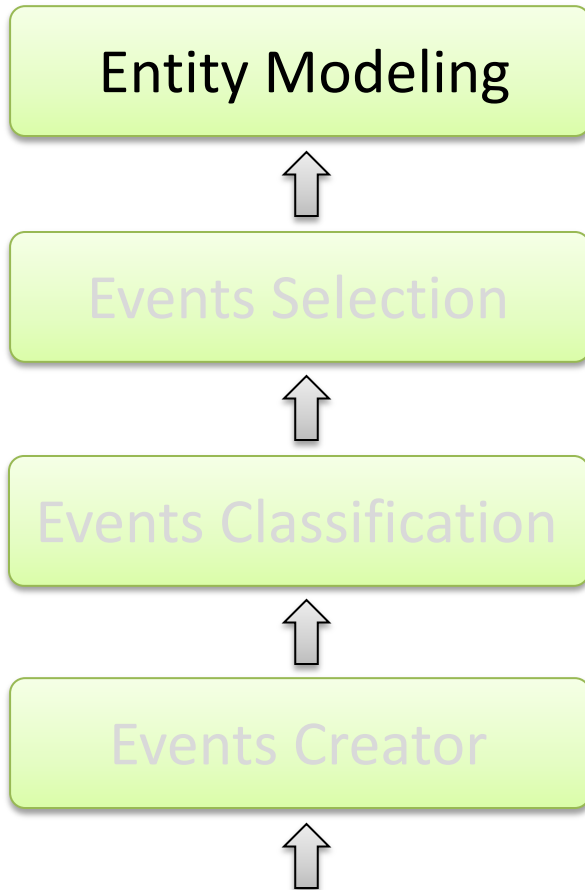
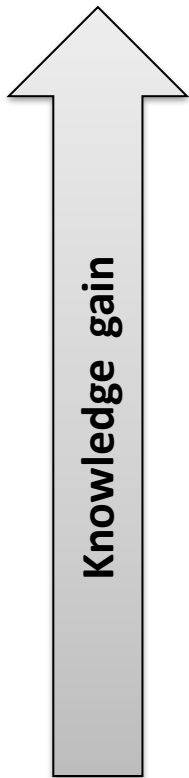
Events Selection



- **Hierarch. Clustering (2nd)**
 - Uses classification and other features
- **Selection & Prioritization**
 - Selects most important events based on their
 - Severity level
 - Degree of anomaly
 - Size (# flows, # bytes)
 - Less important events serve as context

New Component: Entity modeling

Incidents



- **Entity modeling**

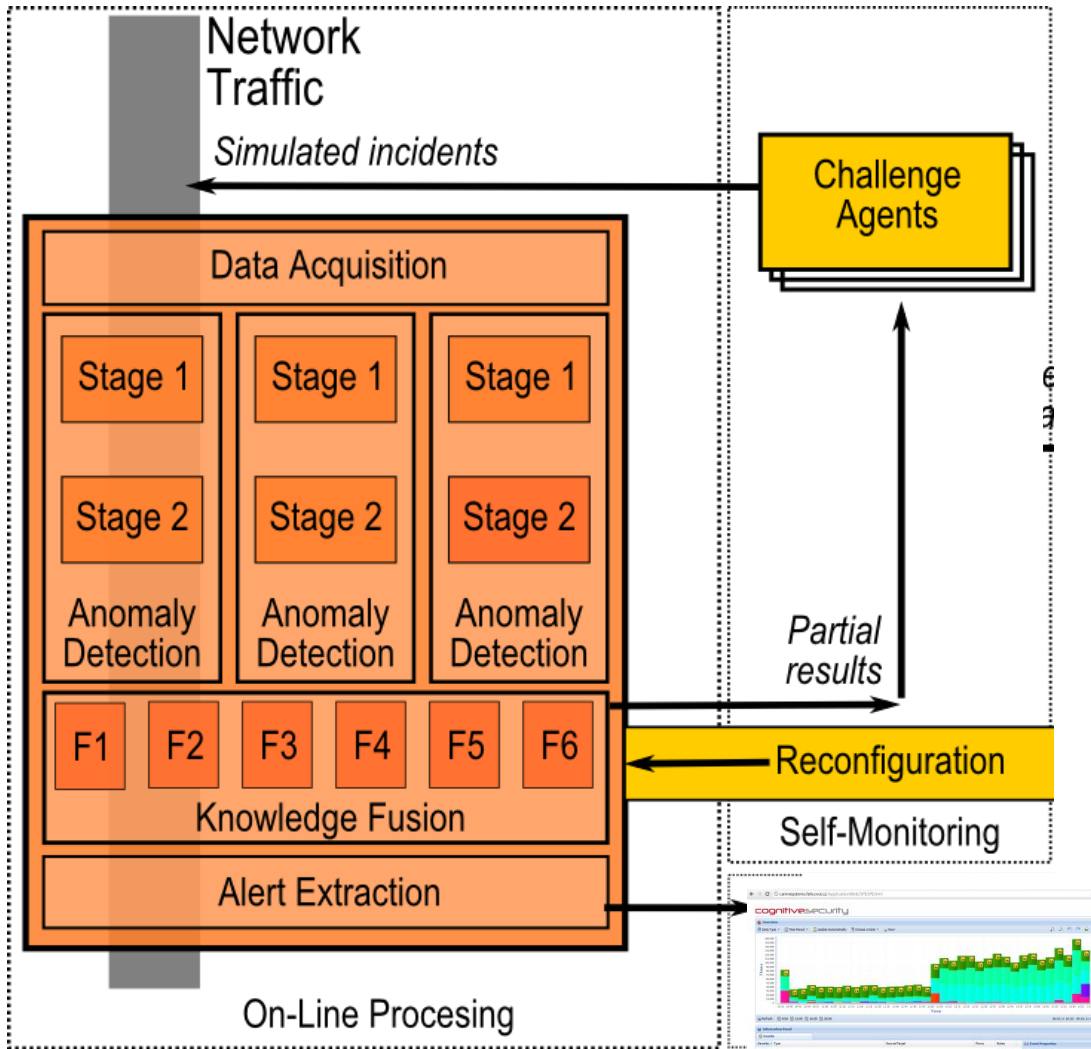
- Increase the level of abstraction of generated incidents
- Model the behavior of individual users in time
- Aggregation and correlation of events
- Apply supervised learning to known threats to increase precision

Flows:

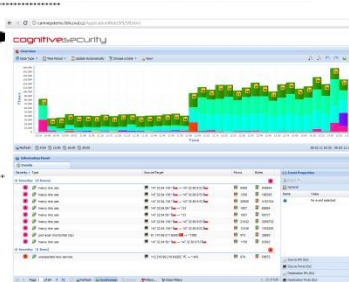
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```



Architecture – Conclusion



- **Data Acquisition and Preprocessing Layer**
→ flows, statistics
- **Detection Layer**
→ trustfulness of flows
- **Self-Monitoring Layer**
→ threshold position
- **Alert Extraction Layer**
→ classified events
- **Analyst Interface**
→ decisions



Demo

