

# Northrop Grumman Cybersecurity Research Consortium (NGCRC)

## Intelligent Autonomous Systems based on Data Analytics and Machine Learning

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- According to **Wes Bush**, CEO of NGC, Autonomous Systems<sup>1</sup> should be
  - Able to perform complex tasks without or with limited ongoing connection to humans.
  - Cognitive enough to act without a human's judgment lapses or execution inadequacies.
- Intelligent Autonomous Systems (IAS) are characterized as highly **Cognitive**, effective in **Knowledge Discovery**, **Reflexive**, and **Trusted**,

<sup>1</sup>Wes Bush, Sept. 6, 2016. "The Exciting Future of Autonomous Systems" at KSU

- **Cognitive Autonomy & Knowledge Discovery:**
  - Monitor and record system's activities (Data provenance and sequence of system calls)
  - Perform advanced analytics on provenance data, discover new patterns, and make predictions.
  - Deep learning based anomaly detection by analyzing sequence of system calls.
- **Reflexivity:**
  - Adapt to meet the mission objectives without disrupting the ongoing critical processes by incremental learning.
- **Trust:**
  - Provide consensus, verifiability, and integrity by using blockchain for storing provenance data.

- **Demo 1 (Cognitive Autonomy / Knowledge Discovery):**
  - Analytics over trusted provenance data to understand the current status of the system and take actions based on the result.
  - Performing aggregate analytics with data perturbation to protect the privacy of individual entities in IAS network.

- **Demo 2 (Reflexivity):**
  - Under anomalous operating contexts or attacks, the replica replacement design based on Combinatorial balanced incomplete modules, can take over the processing from primary module.
  - Replicas are updated with system states periodically (Update interval is determined through Bayesian inference of system's operating context).
  - Unused replicas are used for other processes, which makes the system to be faster and fault-tolerant.

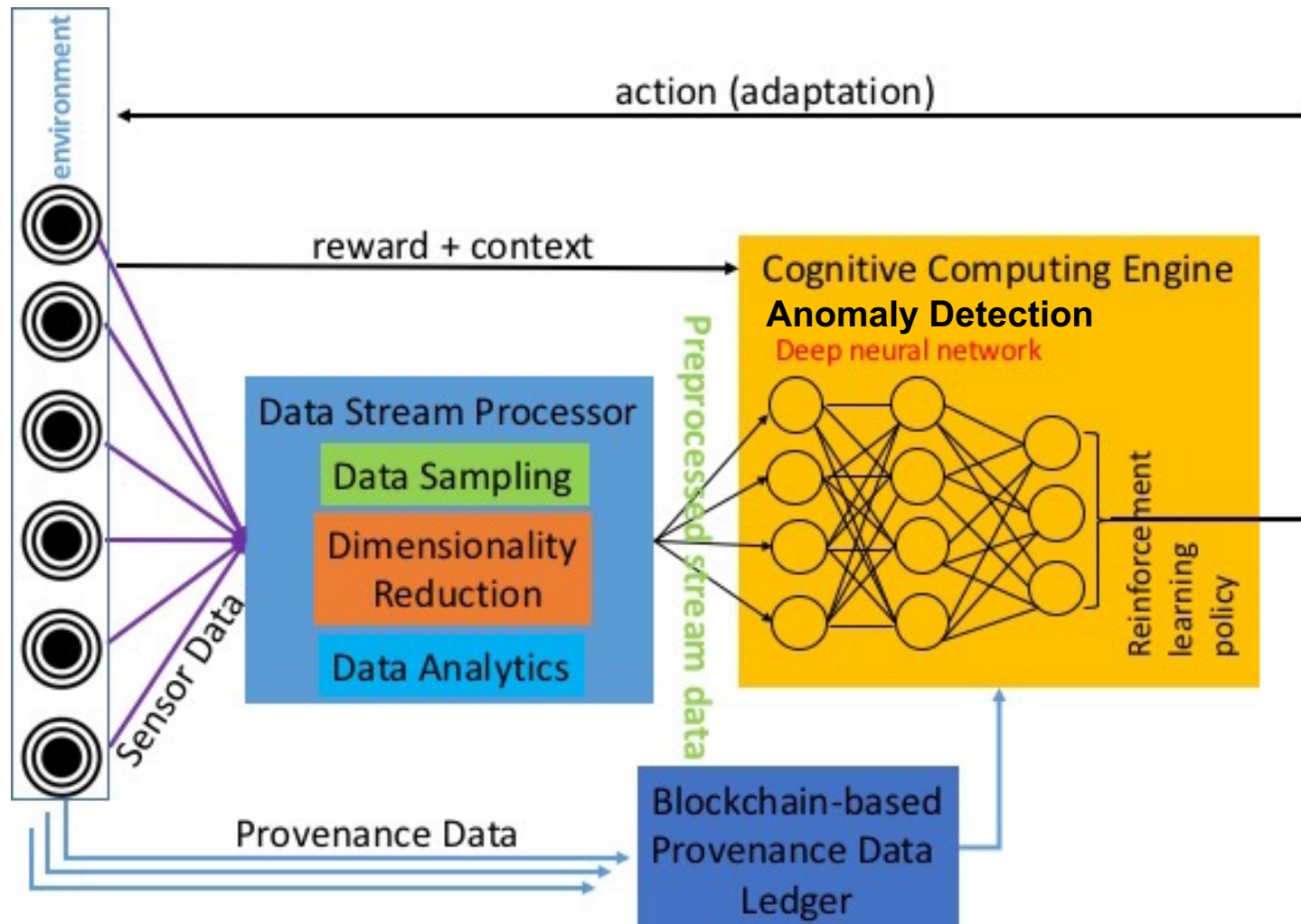
- **Demo 3 (Trust):**
  - Scheme which guarantees integrity of provenance data is implemented
  - Capability to verify every transaction in IAS

# Reflexivity

Graceful Degradation Based on Machine Learning

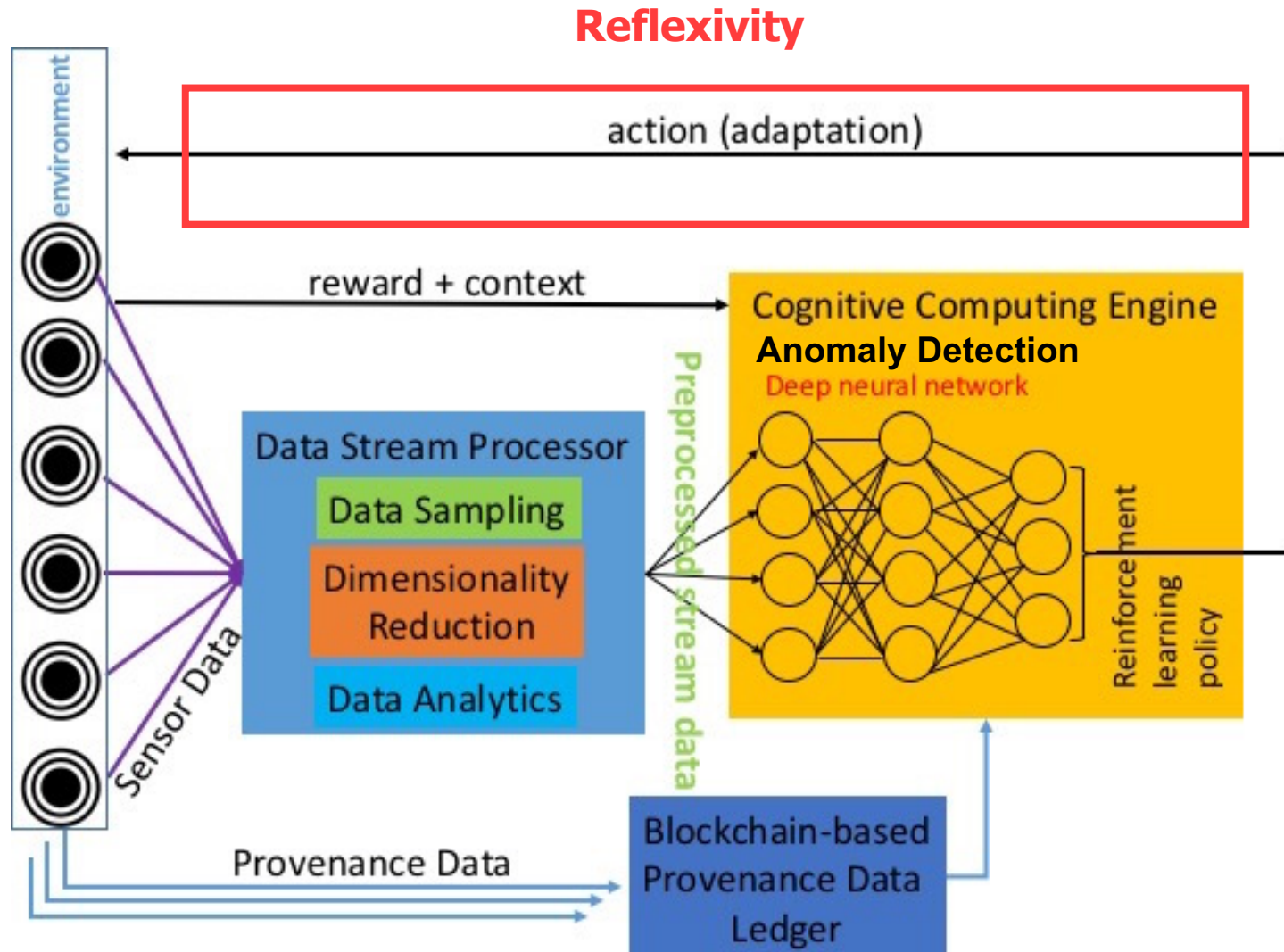


# Comprehensive Architecture of IAS

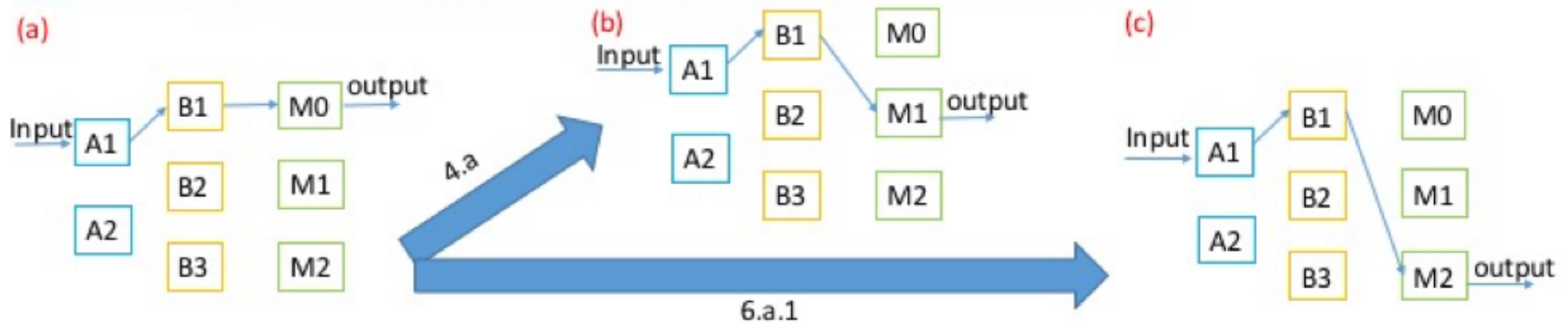
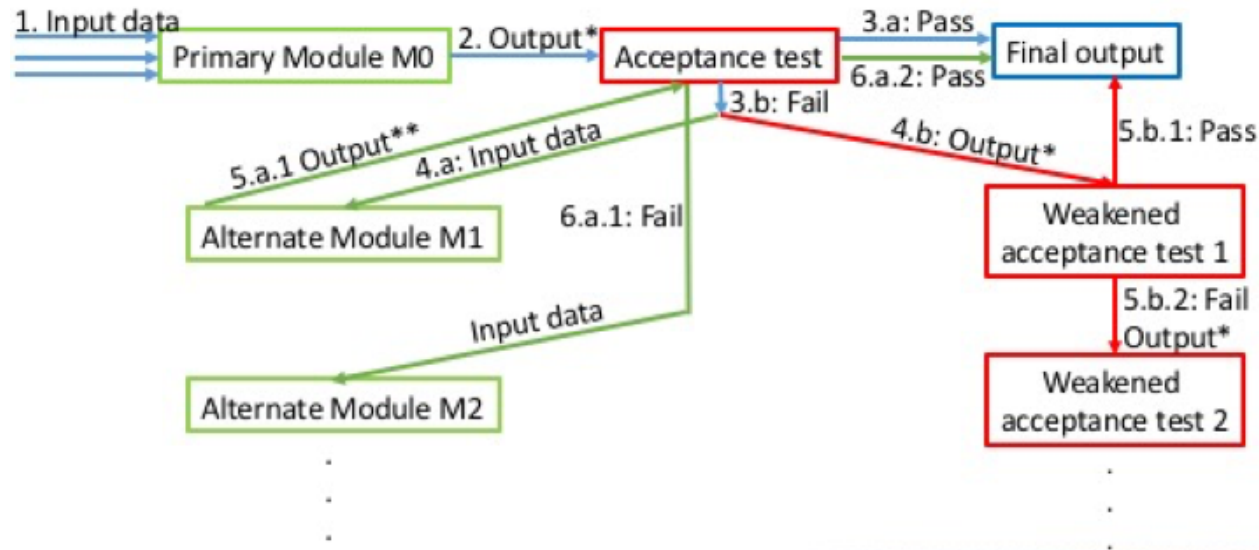




# Comprehensive Architecture of IAS



# Generic Model of Dynamic Adaptation

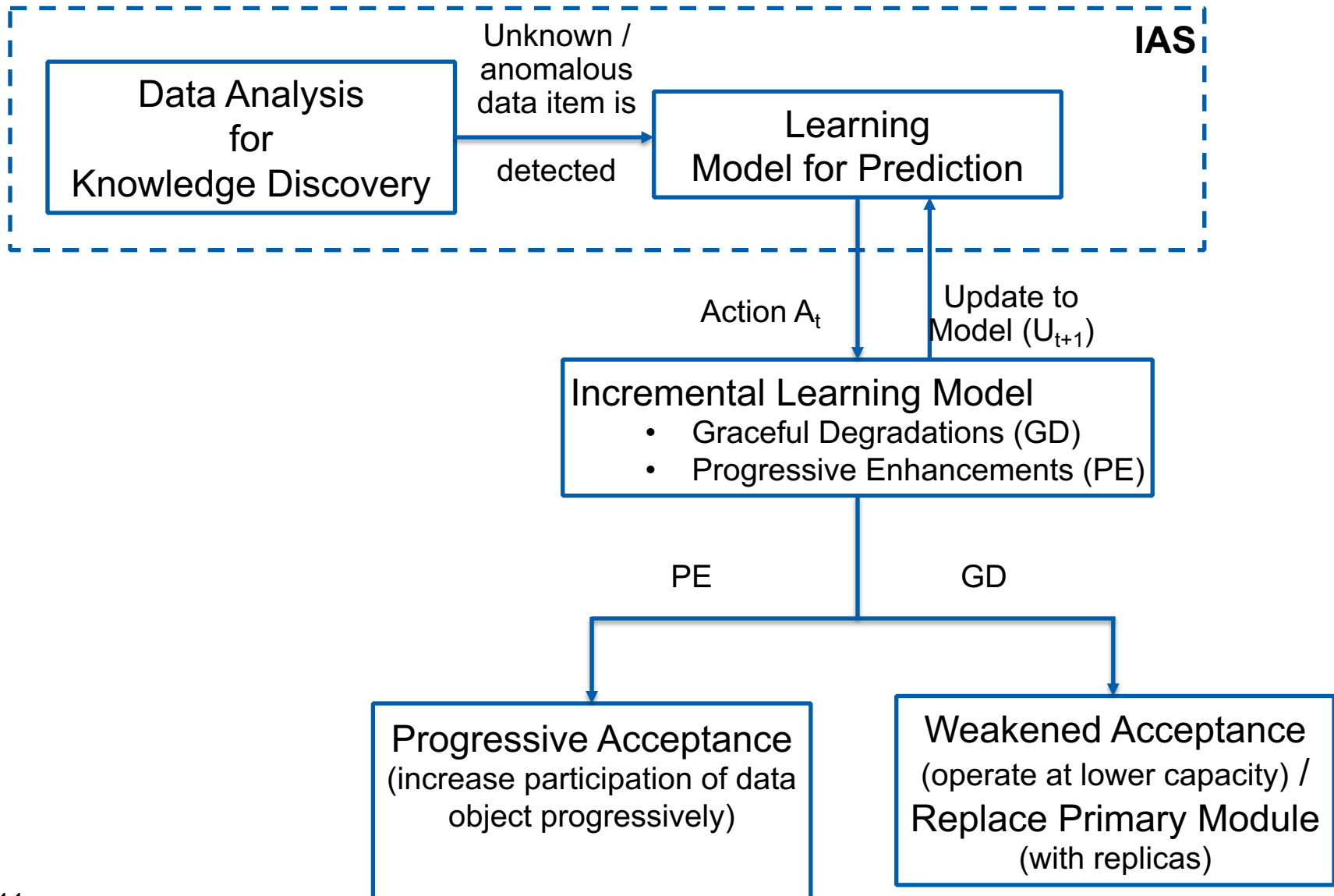


Given a smart cyber system operating in a distributed computing environment, it should be able to:

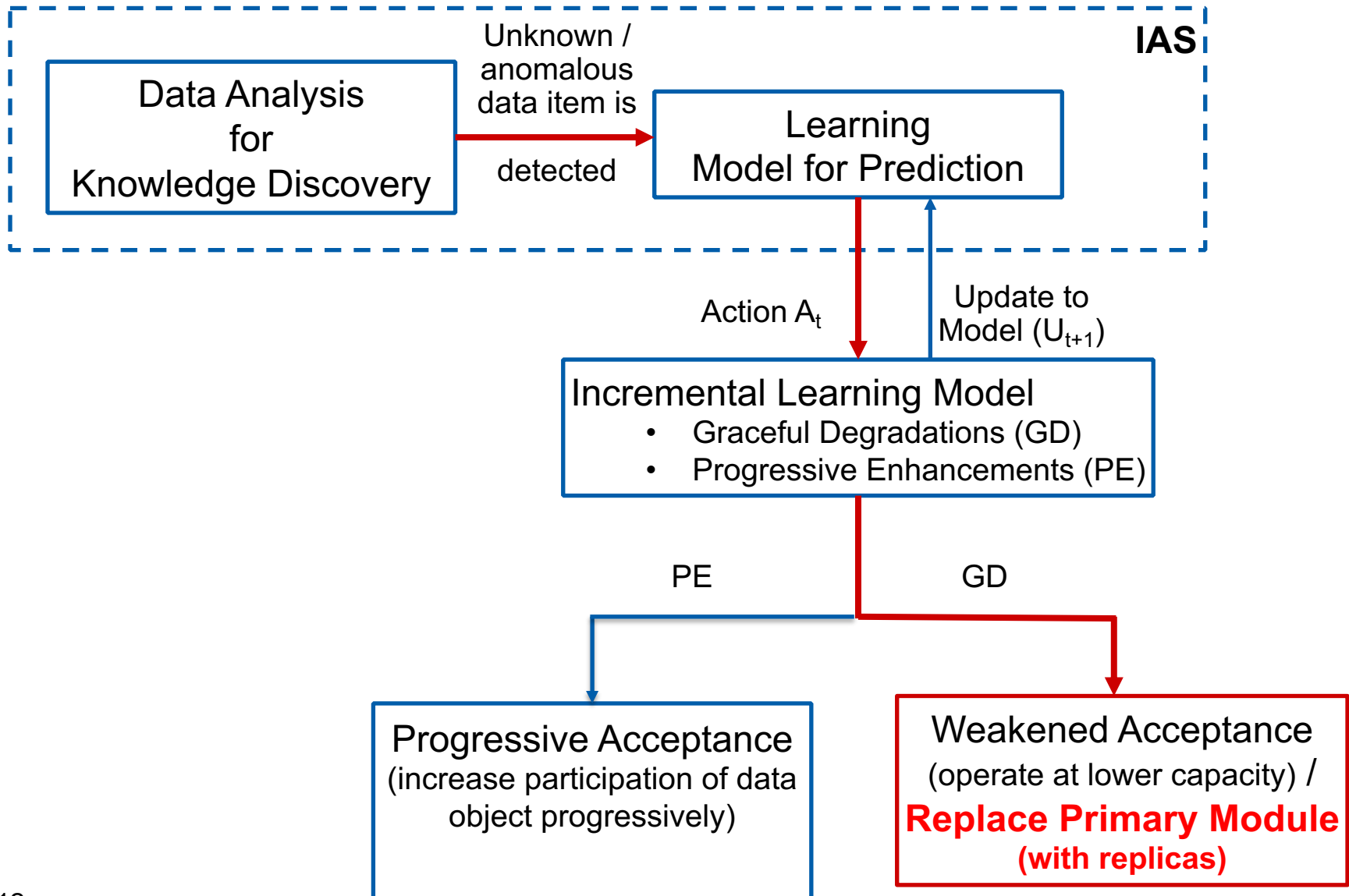
1. Replace anomalous/underperforming modules
2. Swiftly adapt to changes in context
3. Achieve continuous availability even under attacks and failures<sup>4</sup>.

<sup>4</sup>**Thomas E. Vice, Corporate VP of NGC.** Sep. 06, 2016. "'Future of Advanced Trusted Cognitive Autonomous Systems,'" at Purdue University

# Reflexivity Workflow



# Reflexivity Workflow: Graceful Degradations



# Graceful Degradations: Replica Replacement Technique

## Replica replacement by Combinatorial Balanced-block Designs:

- Combinatorial Structure is a subset satisfying certain conditions.
- Each block contains systems and their replicas that are mathematically distributed.
- The systems and their replicas in the distributed blocks are strategically connected to receive updates from primary modules.
- Resources are mathematically balanced, enabling scalable designs for the systems.

- It is distributed environment with
  - A set **Z** consisting **N** systems
  - **M** distributed blocks consisting of
  - Subset of N system of size of **R**
  - Each system in set N appears exactly in **C** subsets
  - Each pair in N systems appears exactly in  $\Delta$  subsets.

# Combinatorial Balanced-block Structure:

## Our implementation



- It is distributed environment with
  - A set **Z** consisting **7** systems
  - **M = 7** distributed blocks consisting of
  - Subset of Z of size of **R = 3**
  - Each system in Z appears exactly in **C = 3** subsets (3 replicas)
  - Each pair in Z appears exactly in  $\Delta = 1$  subsets.



# (7, 7, 3, 3, 1)-configuration

- 7 systems  $\{S_1, S_2, S_3, S_4, S_5, S_6, S_7\}$
- 7 Distributed Autonomous Blocks (DABs) each with 3-system subset

$$DAB_1 = \{S_1, S_5, S_7\}, DAB_2 = \{S_1, S_2, S_6\},$$

$$DAB_3 = \{S_2, S_3, S_7\}, DAB_4 = \{S_1, S_3, S_4\},$$

$$DAB_5 = \{S_2, S_4, S_5\}, DAB_6 = \{S_3, S_5, S_6\},$$

$$DAB_7 = \{S_4, S_6, S_7\}.$$

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- Each system appears in 3 DABs (Say,  $S_6$ )

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# (7, 7, 3, 3, 1)-configuration

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- 7 Distributed Autonomous Blocks (DABs) each with 3-system subset
- Each system appears in 3 DABs
- Each pair of systems appear in 1 DAB (Say,  $S_1$  and  $S_5$ )

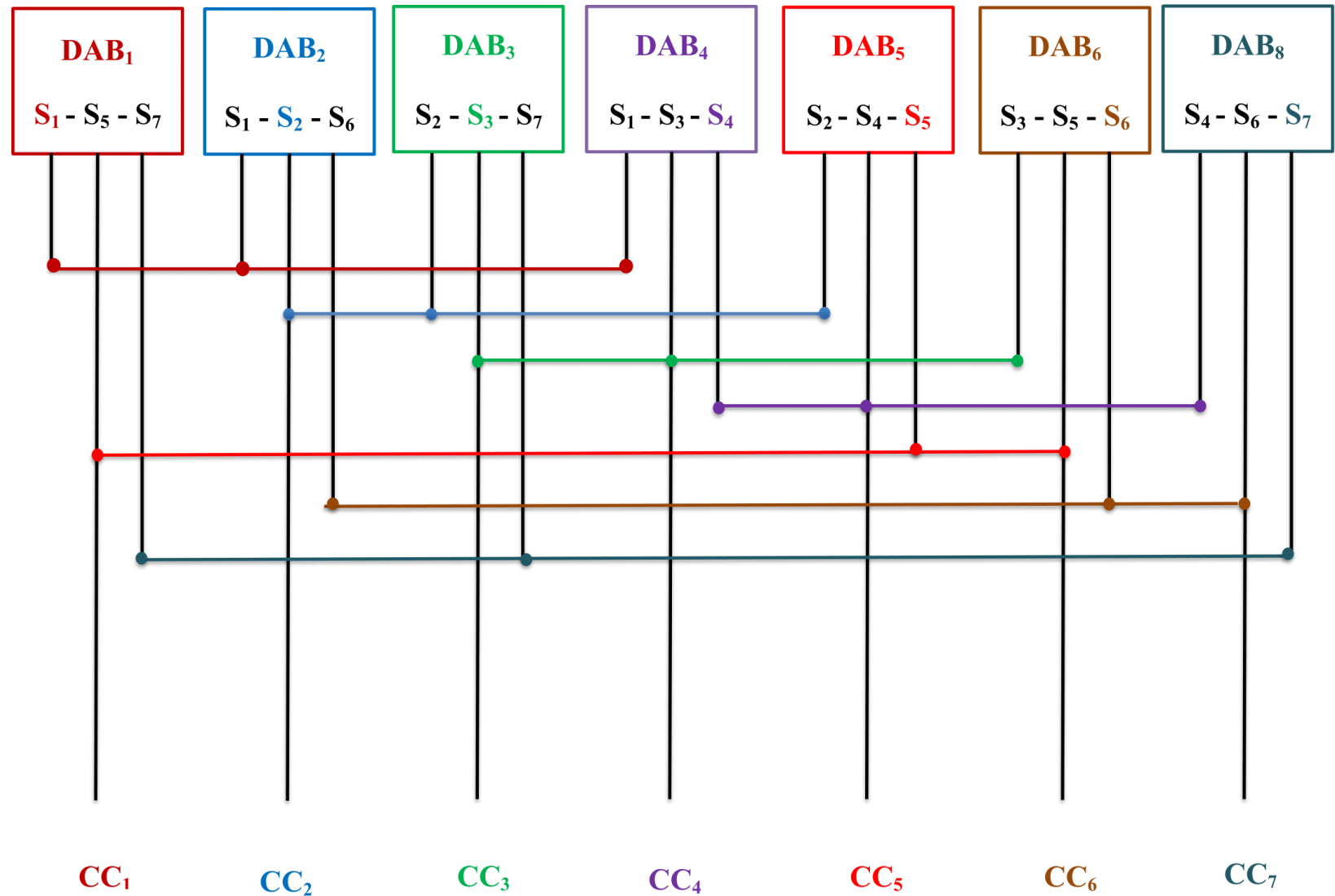
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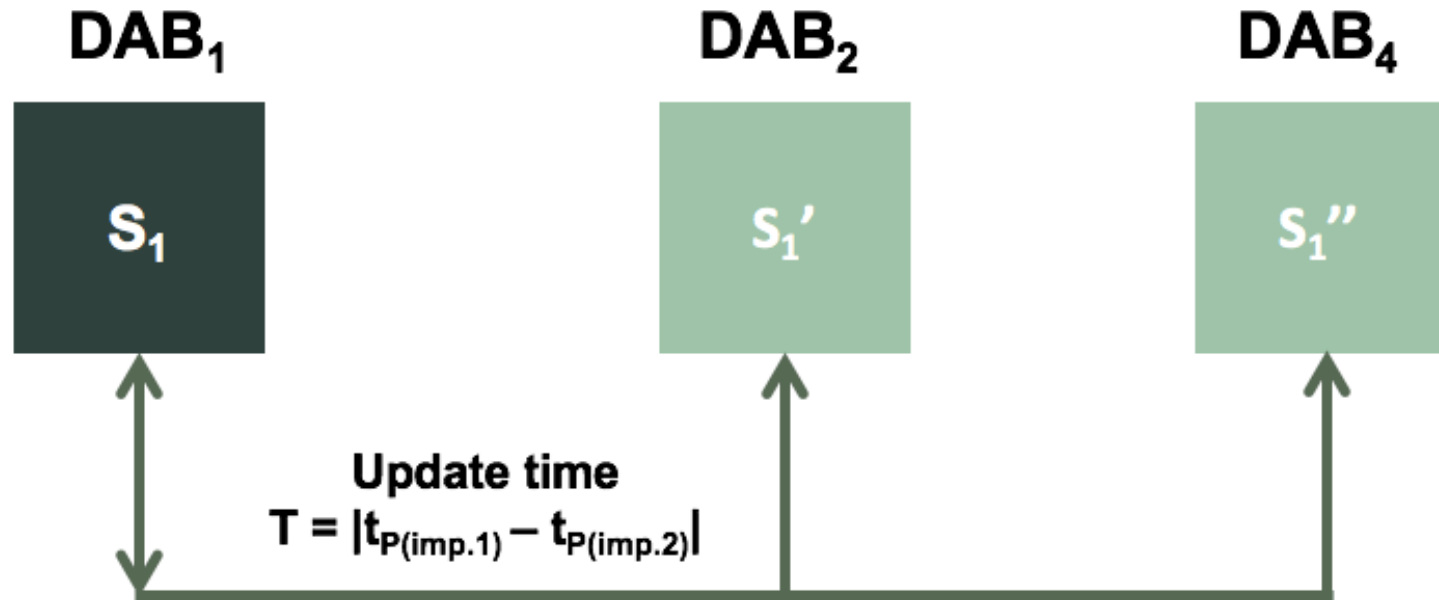
$$DAB_7 = \{S_4, S_6, S_7\}.$$

# (7, 7, 3, 3, 1)-configuration



# (7,7,3,3,1)-configuration's Functionality

- Each primary module periodically updates its replicas in corresponding distributed block connected by communication links (CC).



- Update the interval dynamically through learning models with Bayesian learning by continuously updating the prior.

# (7,7,3,3,1)-configuration's Functionality

- Update time is defined as

$$P_I(\text{importance } (I) \mid \text{operational context } (C)) = \frac{P(C|I)P(I)}{P(C)}$$

$$\text{Update interval } T = | t^1_{P(I)} - t^2_{P(I)} |$$

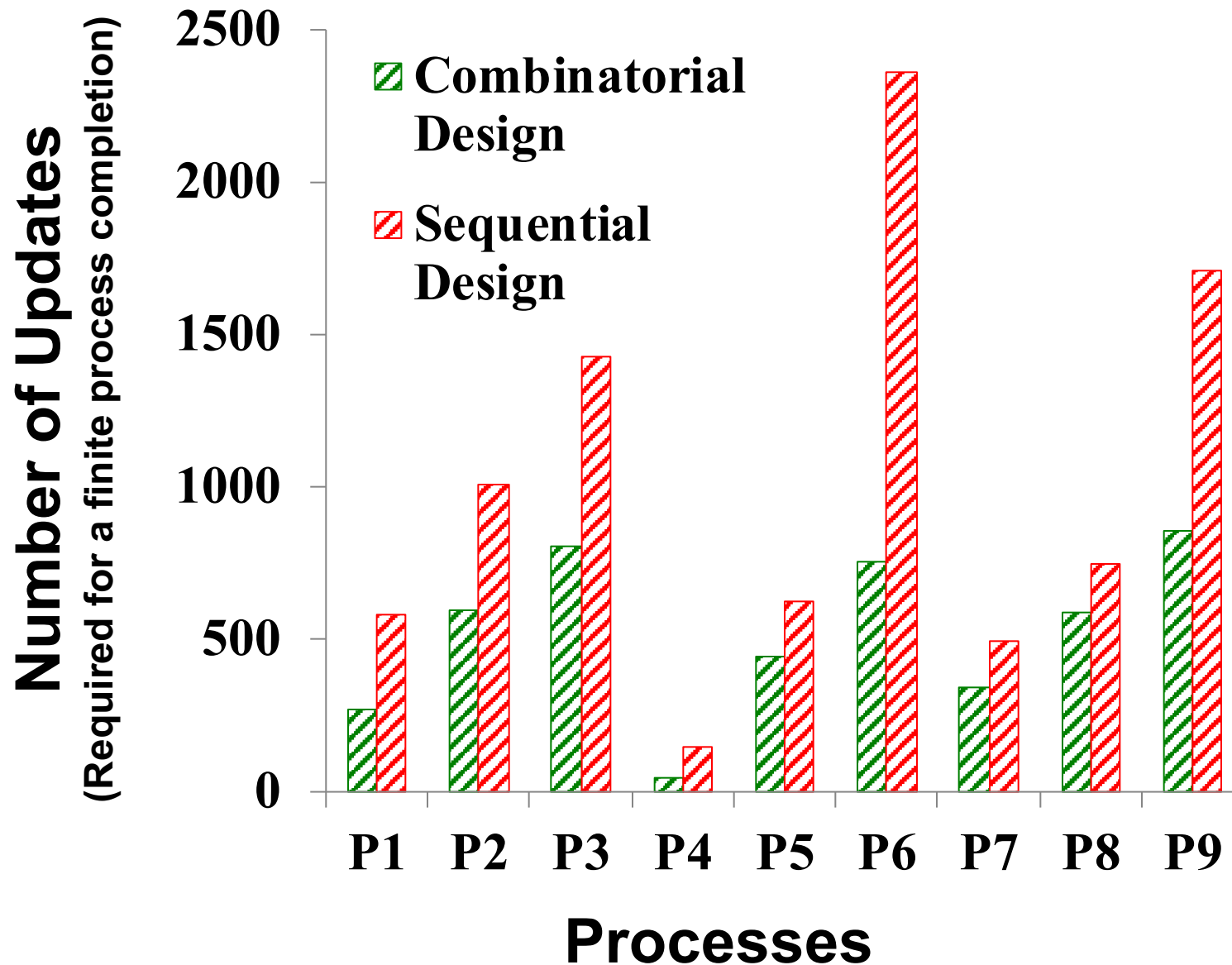
- When any system in any primary module's DAB acts in anomalous fashion, that system can be
  - Replaced with one of the replicas that can be selected in round robin fashion.
  - Anomalous module will be set for self-healing or repair by external source

- The prototype is built with FAYE framework<sup>1</sup> with Node.js.
- It is a server-client framework where servers act as primary modules and clients as replicated system.
- Replica updates are done through a combinatorial design simulator<sup>2</sup>.
- Combinatorial simulator is loaded with finite processes to compare the updates and processing time compared to a regular or sequential processing.

- **Deliverables:**
  - Autonomous replica replacement prototype
  - Source code: Node.js implementation, Bayesian model, simulation software developed for combinatorial design, and Data used for simulation.  
Link: <https://goo.gl/M4rXCN>
  - Documentation: Demo video and User manual for running the prototype.



# Results of Measurements



# Results of Measurements

<b>Process</b>	<b>Speed Up of Replica Scheme</b> (Compared to regular sequential design)
<b>FIBSEARCH</b>	<b>1.3</b>
<b>DOUBLE MULT</b>	<b>1.4</b>
<b>FIBB</b>	<b>1.5</b>
<b>SEARCH</b>	<b>1.8</b>
<b>COPY</b>	<b>1.8</b>
<b>SCALAR</b>	<b>2</b>
<b>SUM</b>	<b>2.1</b>
<b>PRINT</b>	<b>3</b>
<b>MOVEMENT</b>	<b>3.1</b>

- Since block sizes are equal the efficiency of the system is balanced.
  - It provides a stable and reliable system [1].
  - The design provides scalability with thousands of systems with simplistic communication links.
- The design is scalable since the number of replicas can be decreased depending on the failure rate of the whole system. (Note: Replicas are  $C - 1$ ).

$$\text{No. of Replicas} \propto \frac{1}{\text{Number of Systems (N)} | (\text{Failure Rate} < C)}$$

- Updates to the replicas happen independently and simultaneously without interfering with others
- Replicas can aid other functionalities with their updates.
- All the systems in the DAB are connected which aid a faster communication.
  - For example, if S3 needs an update from S1 and S4, it can instantly communicate with them since they are in DB 4. This type of instant communication can aid parallel progressing of any process.

# Combinatorial Balanced-block Structure:

## Other possibilities



- There are other configurations such as
  - $M = 7, 13, 21, 31, 55, 73, 91, 132$
  - $N = 7, 13, 21, 31, 55, 73, 91, 132$
  - $C = 3, 4, 5, 6, 8, 9, 10, 12$
  - $R = 3, 4, 5, 6, 8, 9, 10, 12$
  - $\Delta = 1, 1, 1, 1, 1, 1, 1, 1$

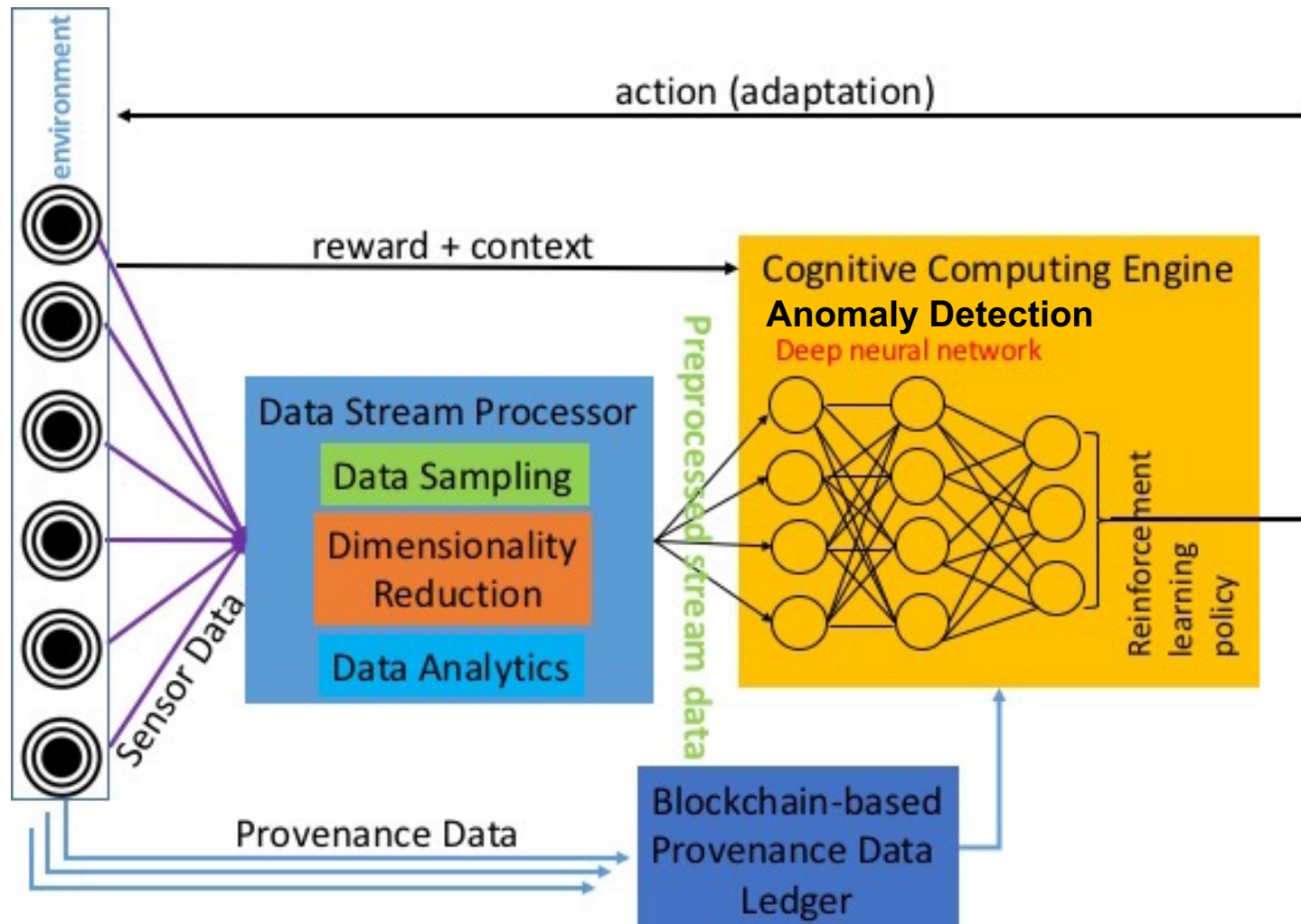
- The recovery and replica replacement mechanism contributes towards self-healing: Automated Enterprise Monitoring and Recovery (AEMR) IRAD.
- Distributed, simultaneous, and independent communication contributes to Distributed Data Processing IRAD.

# Trust

THE VALUE OF PERFORMANCE.  
**Blockhub:** A Blockchain-Based  
Distribution of Datasets in IAS

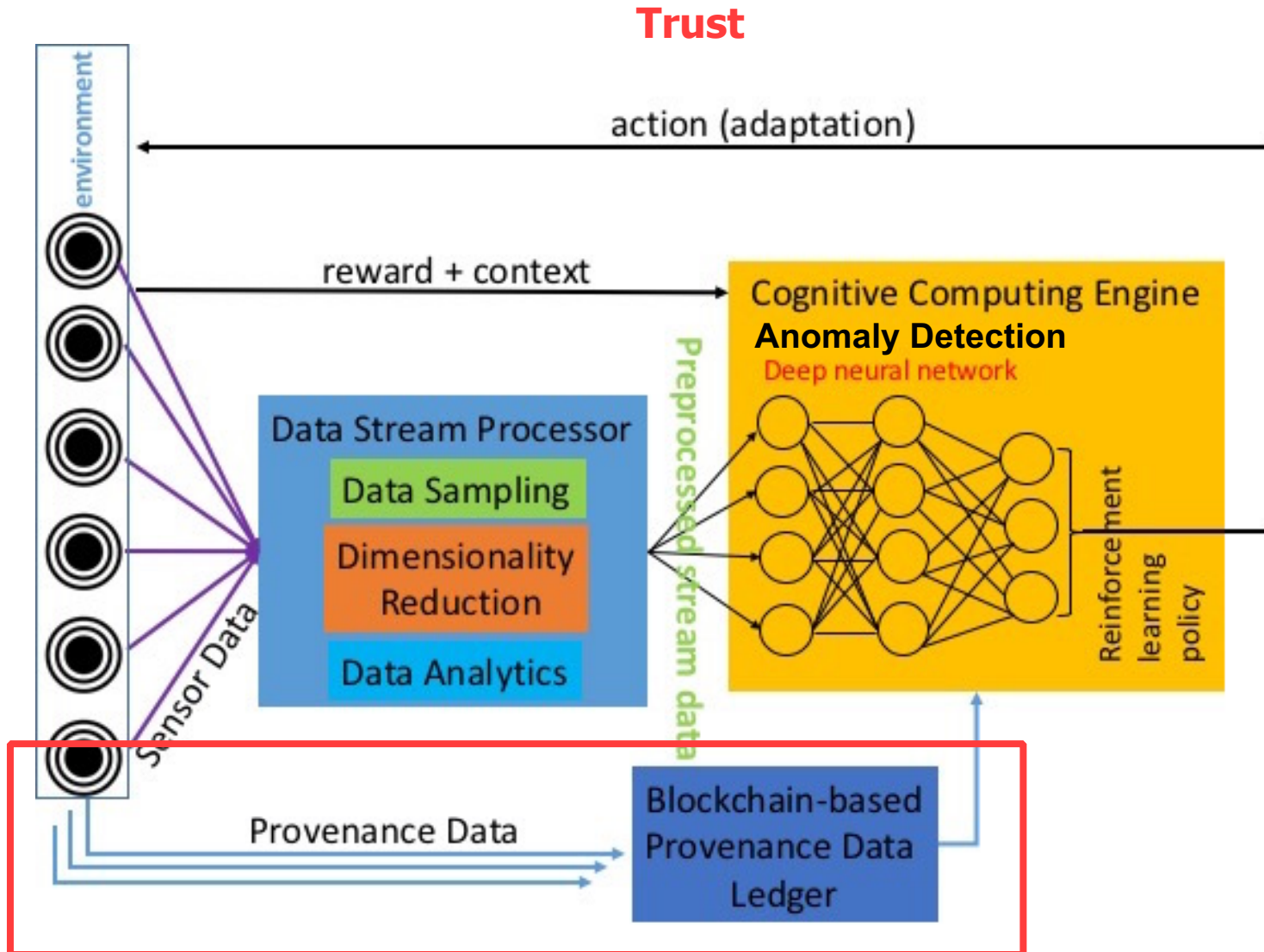


# Comprehensive Architecture of IAS





# Comprehensive Architecture of IAS



- Provide trust (integrity, confidentiality, verifiability) to provenance data in IAS
  - Interactions between services are logged
  - Log records can not be corrupted
- Provide trust for network participants in IAS
  - (a) ensure data confidentiality
  - (b) ensure data integrity
- Provide privacy-preserving data exchange in IAS

# Blockchain technology deployment challenges

- Performance:
  - (a) Transaction latency on IBM Hyperledger Fabric blockchain platform (ver 1.0.x) is greater than 6 seconds
  - (b) Performance overhead on log verification phase when the chain has large in size (a great amount of blocks) => verifiers have to compute too many hash functions

## **Solution:**

- (a) modify transaction verification process at “Endorsers”
- (b) Use depth-robust graphs to store blockchain (in collaboration with Prof. Blocki, Purdue) which can validate transaction without checking every block in the chain

# Blockchain technology deployment challenges

- Fine-grained role- and attribute-based access control with data leakage detection capabilities

## **Solution:**

Integrate WAXEDPRUNE project into blockchain-based framework to provide:

- (a) role-based access control
- (b) detection/prevention of data leakages made by insiders
- (c) attribute-based access control, with attributes including:
  - (c1) trust level of network nodes
  - (c2) context (e.g. normal vs. emergency)
  - (c3) authentication method (e.g. password-based vs. fingerprint)
  - (c4) cryptographic capabilities of network node

# Blockchain technology deployment challenges

- Failure Recovery (proposed by Steve Seaberg, NGC)
  - Need to maintain consistency in mobile environment with intermittent connectivity
  - Need quantification of performance parameters after a varying period of connectivity breakdown
  - Need to determine how much bandwidth and resources are needed to make network nodes consistent (or current)

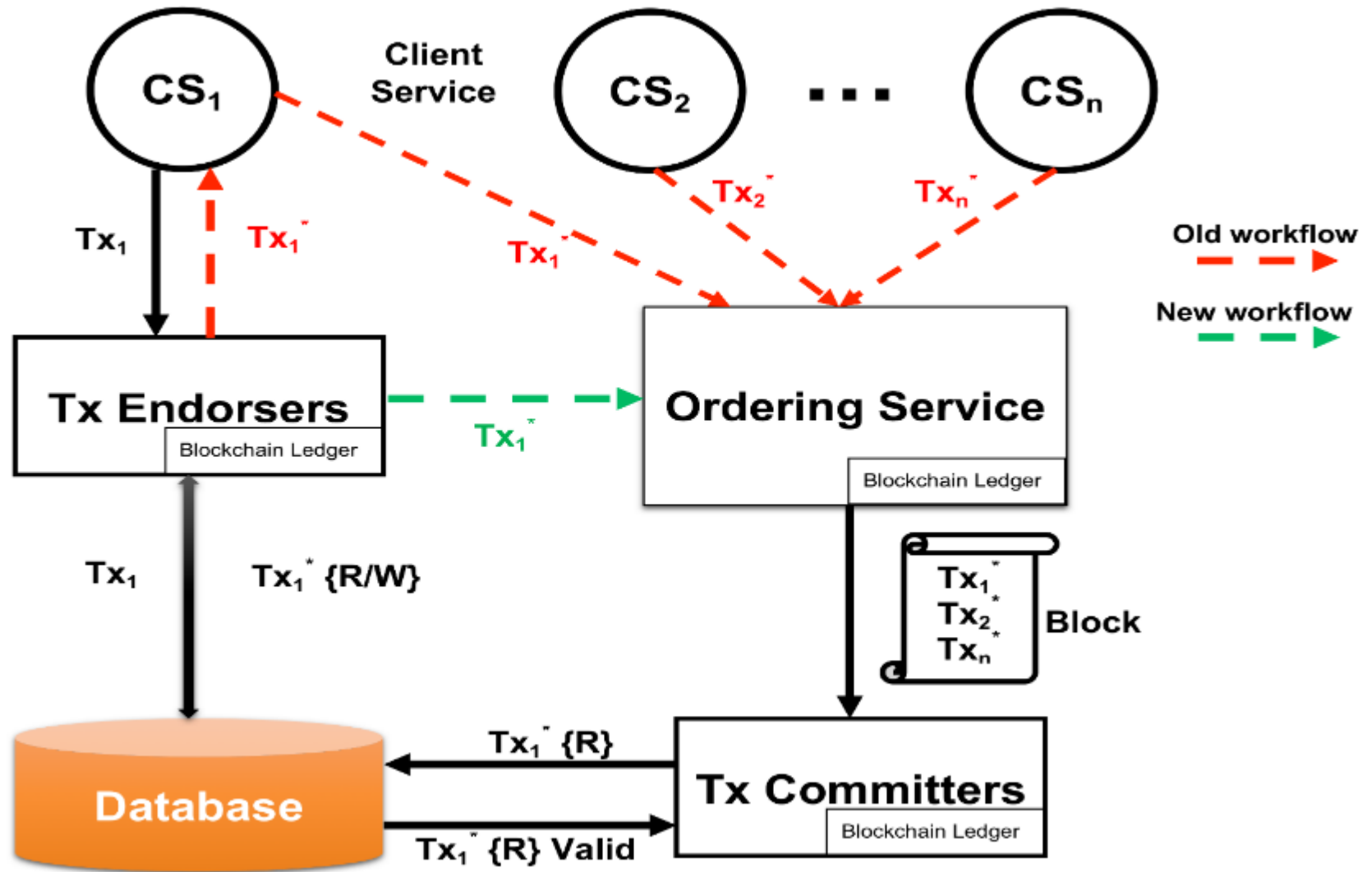
# Blockchain technology deployment challenges

- *Access Control (Read)*: revoked access to data in a blockchain can be bypassed as follows:
  - Attacker holding a copy of a blockchain can use a modified client to just ignore the revocation block
  - Attacker can replay old blocks against an empty blockchain and stop before revocation block is appended

# Blockchain technology deployment challenges

- Provide secure User Interface to create Smart Contracts (Chaincodes)
  - Idea proposed by Ashish Kundu (IBM)
  - GO language is currently used to write chaincode
  - How to make chaincode writing easier for the user and reduce attack surface

# Solution for Permissioned Systems

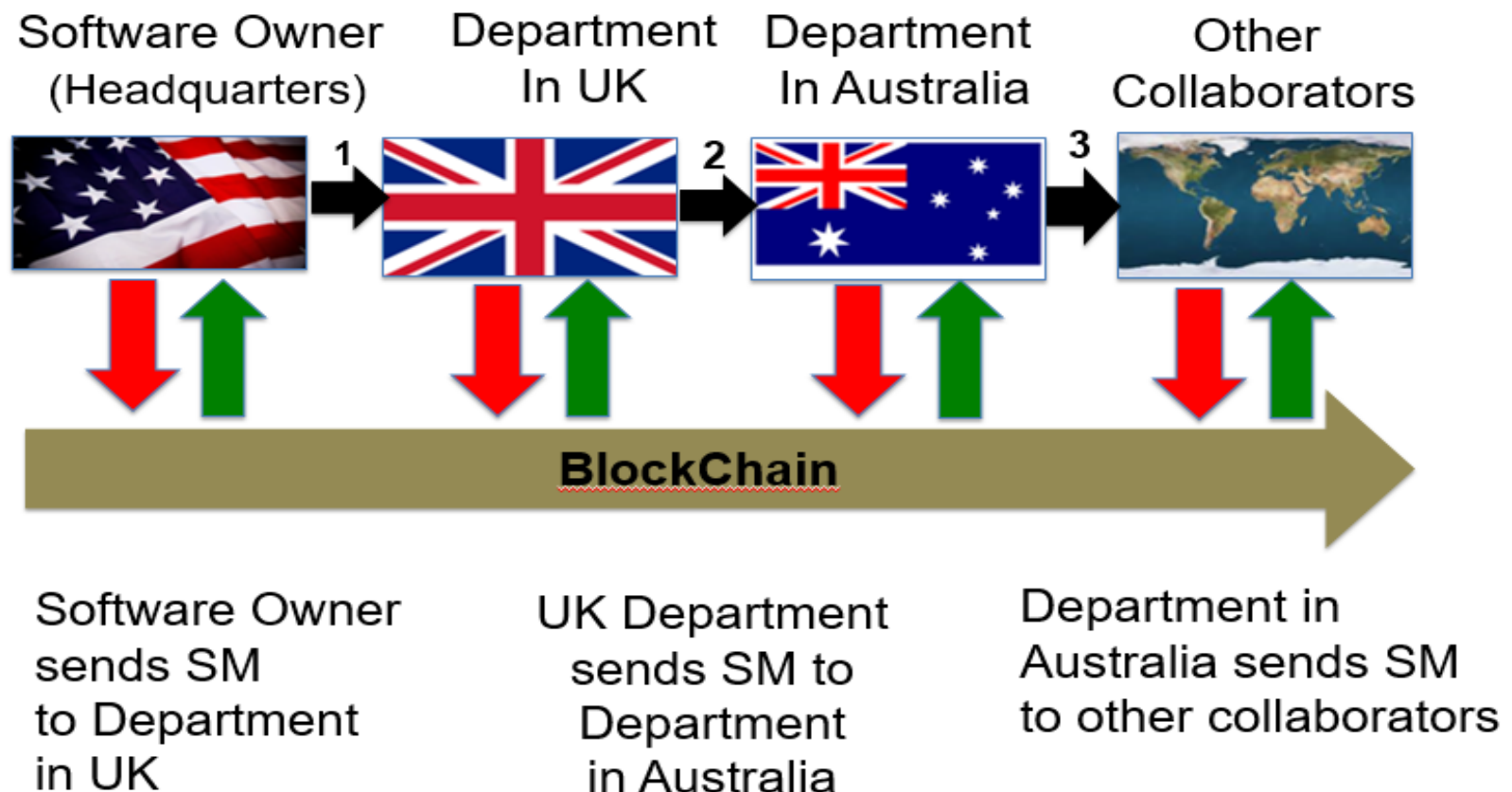




- Idea: protect against dDOS attacks when invalid transactions flood the network and don't let valid transactions to be processed
  - Remove client service's communication with the ordering service and extra-communications with Endorsers
  - Endorsers directly send validated transaction to the Ordering Service
  - Ordering Service delivers transaction to the Committers

## Blockchain-based Software Distribution (WAXEDPRUNE extension)

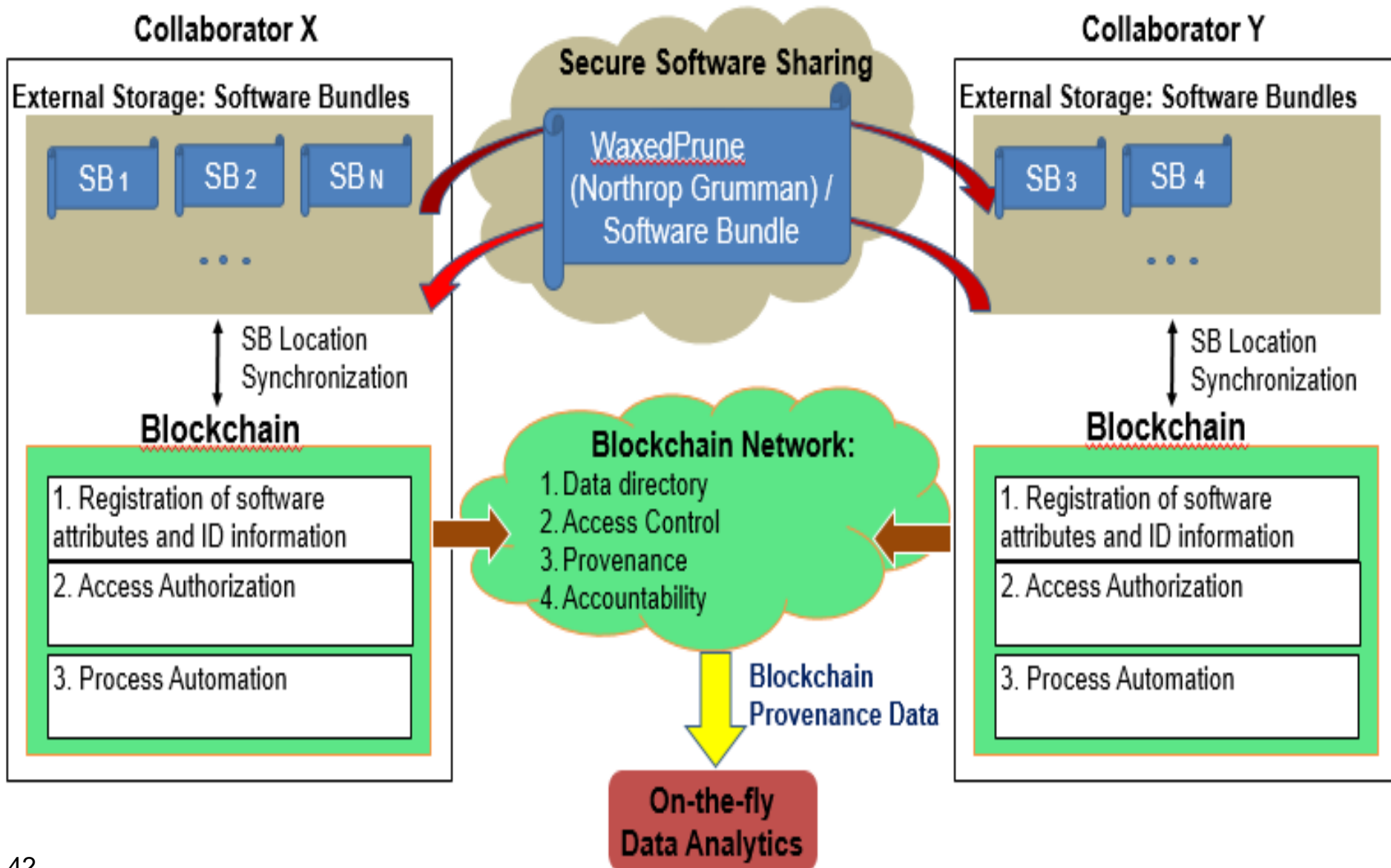
- Blockchain-based technology ensures integrity of provenance data
- *BlockHub* provides secure cross-domain software distribution



## **BlockHub can be used for:** (proposed by Robert Pike, NGC)

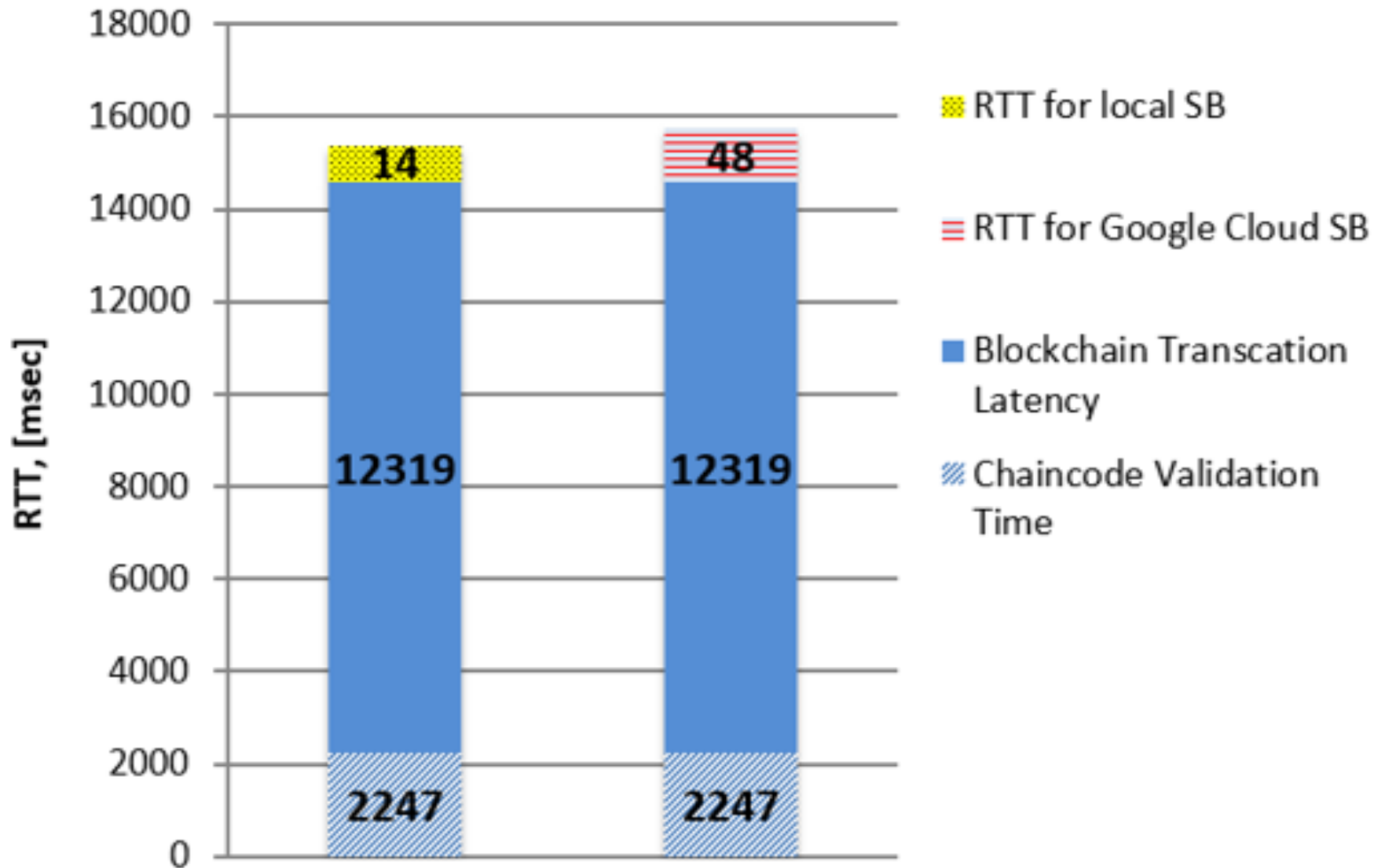
1. Tracking and control of software components that are shared across multiple security domains.
2. Automating the export auditing and tracking processes.
3. Cross-domain dissemination of encrypted software modules using role- and attribute-based access control.
4. Licensing provenance of deployed software components.
5. Enabling software supply chain that is tamper resistant.
6. Software spillage remediation.

# Blockhub: blockchain-platform for IAS



- X and Y share software via *smart contracts* running in blockchain network
- Every request is logged in the blockchain's distributed ledger
- Software stored in form of Software Bundles (SB) that contain:
  - *Encrypted Software Modules (source code or executables)*
  - *Access Control Policies*
  - *Policy Enforcement Engine (Virtual Machine)*
- Software is transferred if authorization has been granted by both smart contract and policy enforcement engine of the SB
- Any transaction, i.e. software access/update can be verified any time in the future
- Storing blockchain in form of *depth-robust graph* (proposed by Prof. Blocki, Purdue Univ.) reduces transaction validation time
- There is one depth-robust graph per SB and it gets updated each time a transaction occurs

# Evaluation



- **Deliverables:**

‘Blockhub’ prototype for secure blockchain-based data distribution

- Source code: <https://github.com/Denis-Ulybysh/Waxedprune2018>

\* codebase is taken from open-source “Marbles” project  
<https://github.com/IBM-Blockchain/marbles/tree/v4.0>

- Demo video

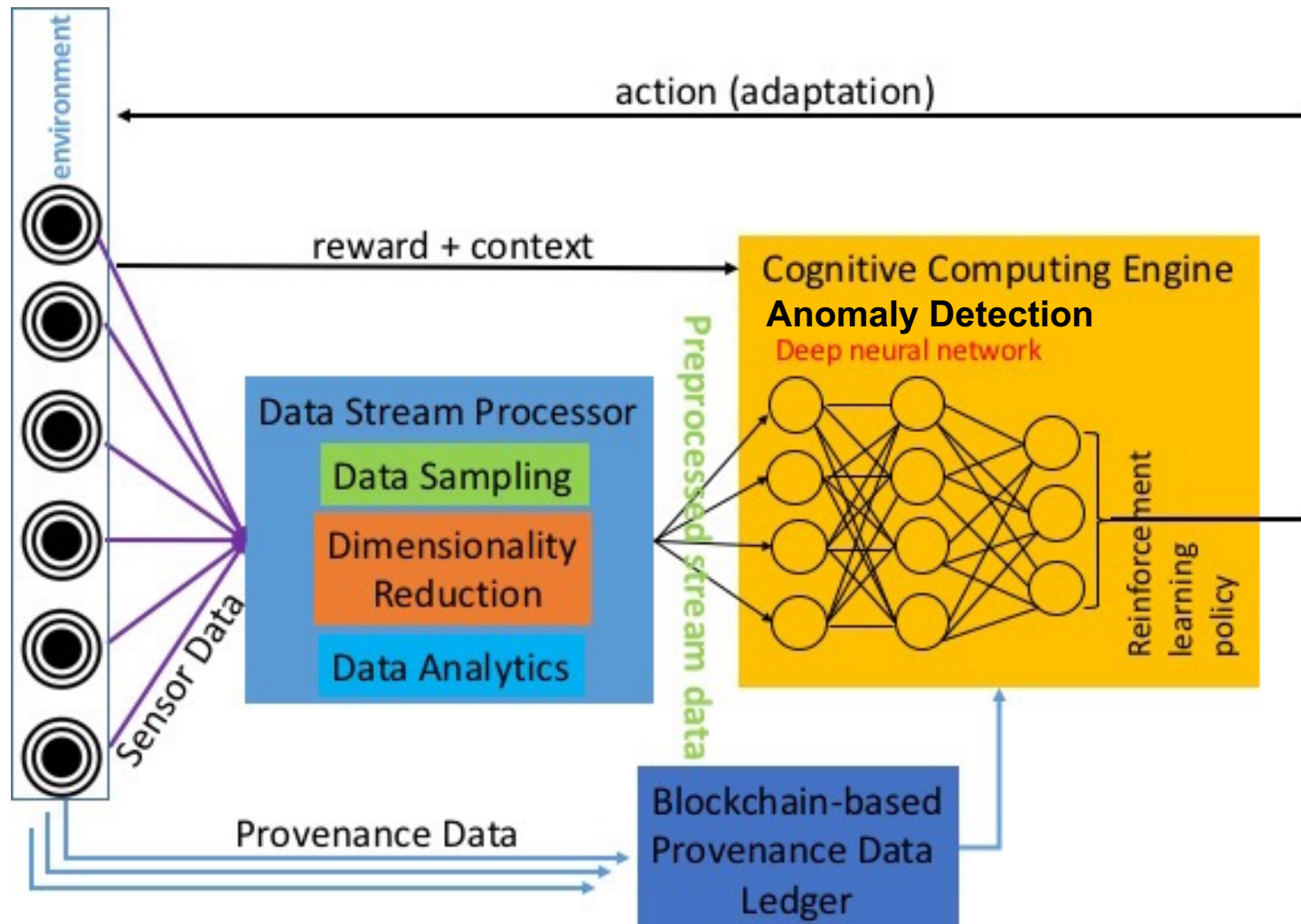
# Cognitive Autonomy / Knowledge Discovery

A Deep Learning Based Anomaly Detection Solution



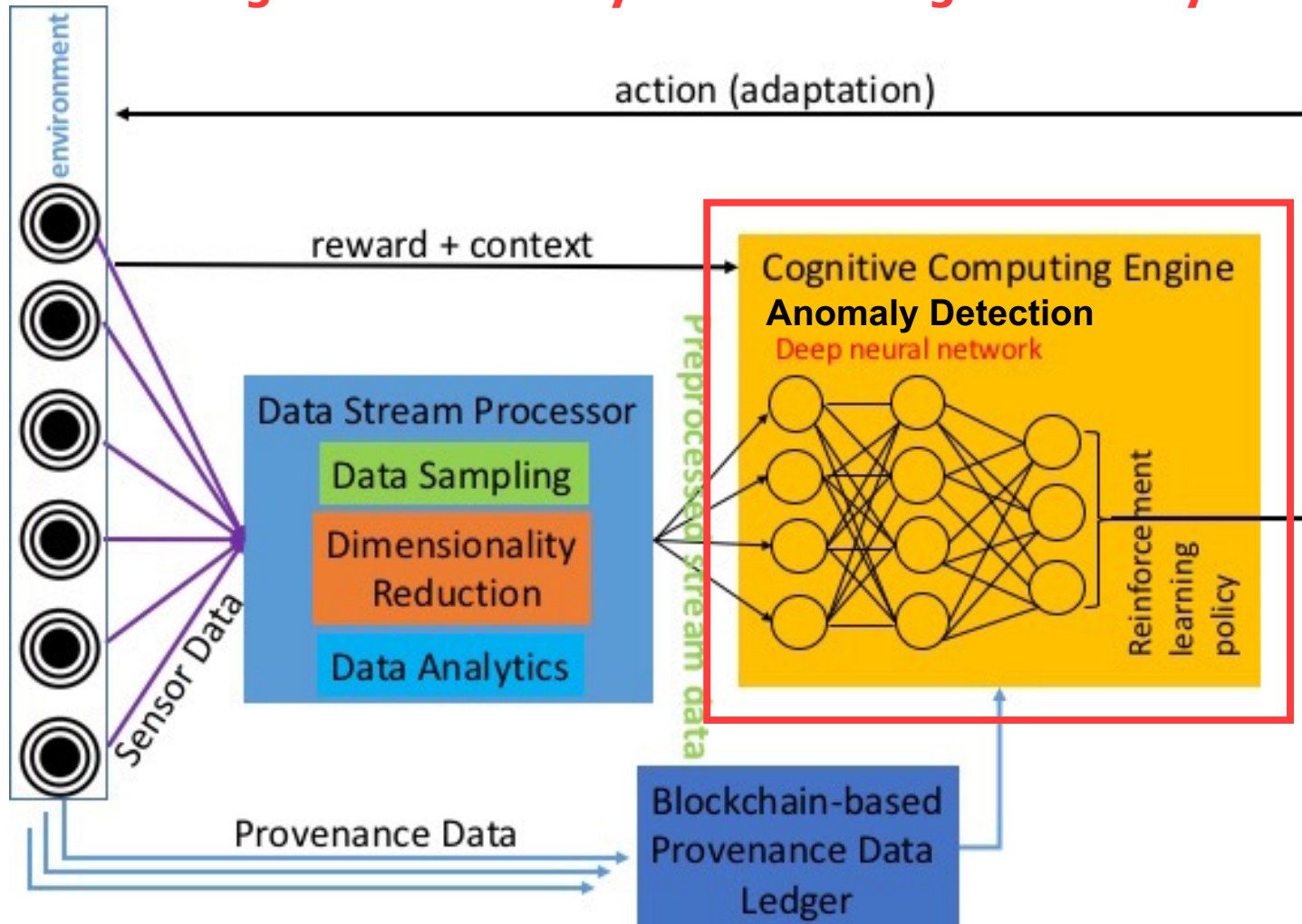


# Comprehensive Architecture of IAS

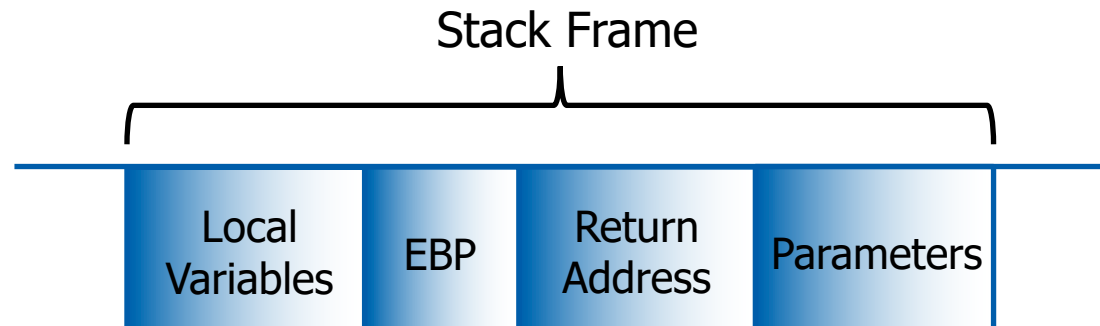


# Comprehensive Architecture of IAS

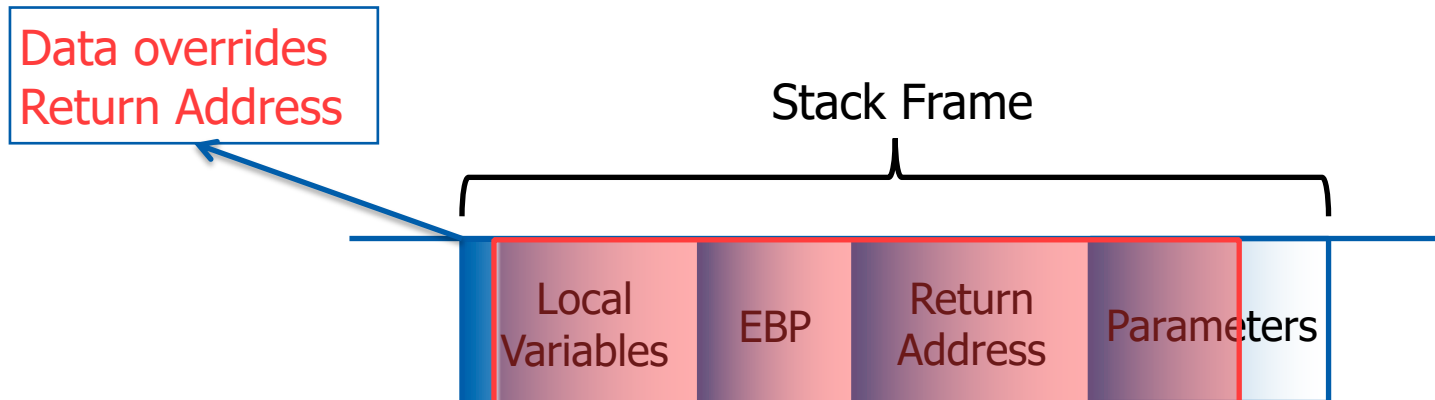
## Cognitive Autonomy and Knowledge Discovery



- **Intelligent Autonomous Systems (IAS)** adapt to meet the mission objectives without disrupting the ongoing critical processes by incremental learning (**reflexivity**).
- Programs store Return Addresses (control flow) along with data in the stack.
- **Control-hijacking** attacks execute arbitrary code on the target IAS program by hijacking its control flow.



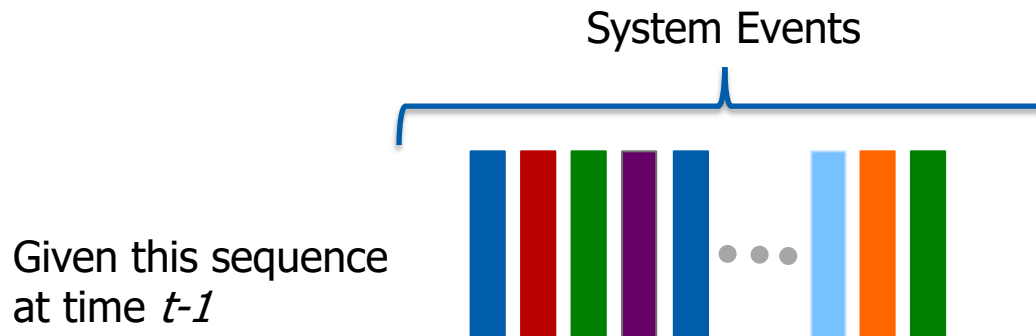
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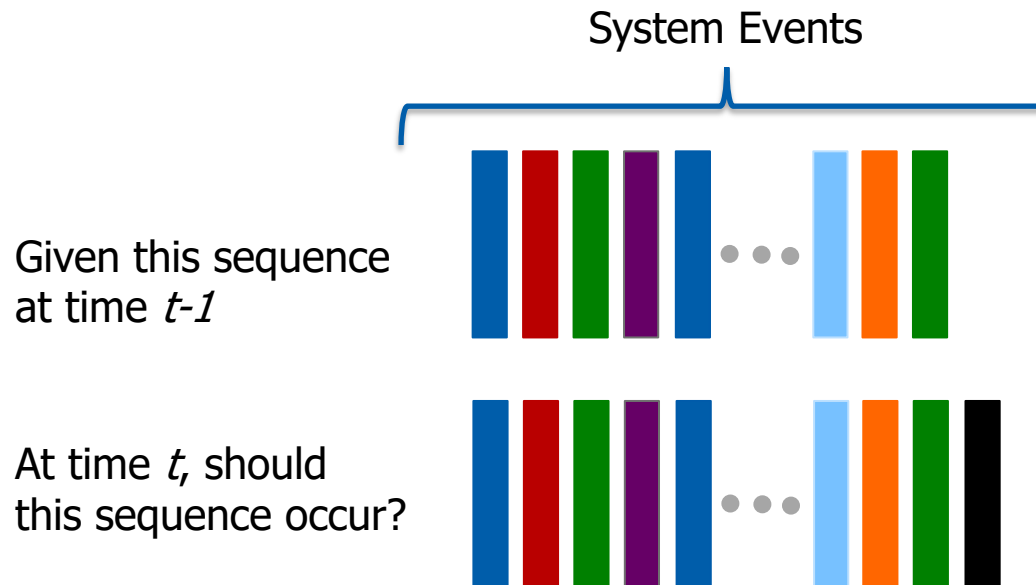
- Advanced modern exploits characterize by their sophistication in stealthy attacks.
- Code-reuse attacks such as return-oriented programming (ROP) and memory disclosures allow attackers executing malicious instruction sequences on victim systems without injecting external code.
- Control Flow Integrity (CFI) is required.
- **Research Question:** Can we developed a Deep Learning based anomaly detection technique that probabilistically models program control flows for behavioral reasoning and system monitoring?

- An event  $e_i$  *is defined* defined as a function call (**system or library call**) in the execution trace of a program.
- Use Deep Learning to answer **the binary classification problem** of given a sequence of function calls (or system events)  $e_1e_2e_3...e_k$  **whether or not the sequence should occur?**

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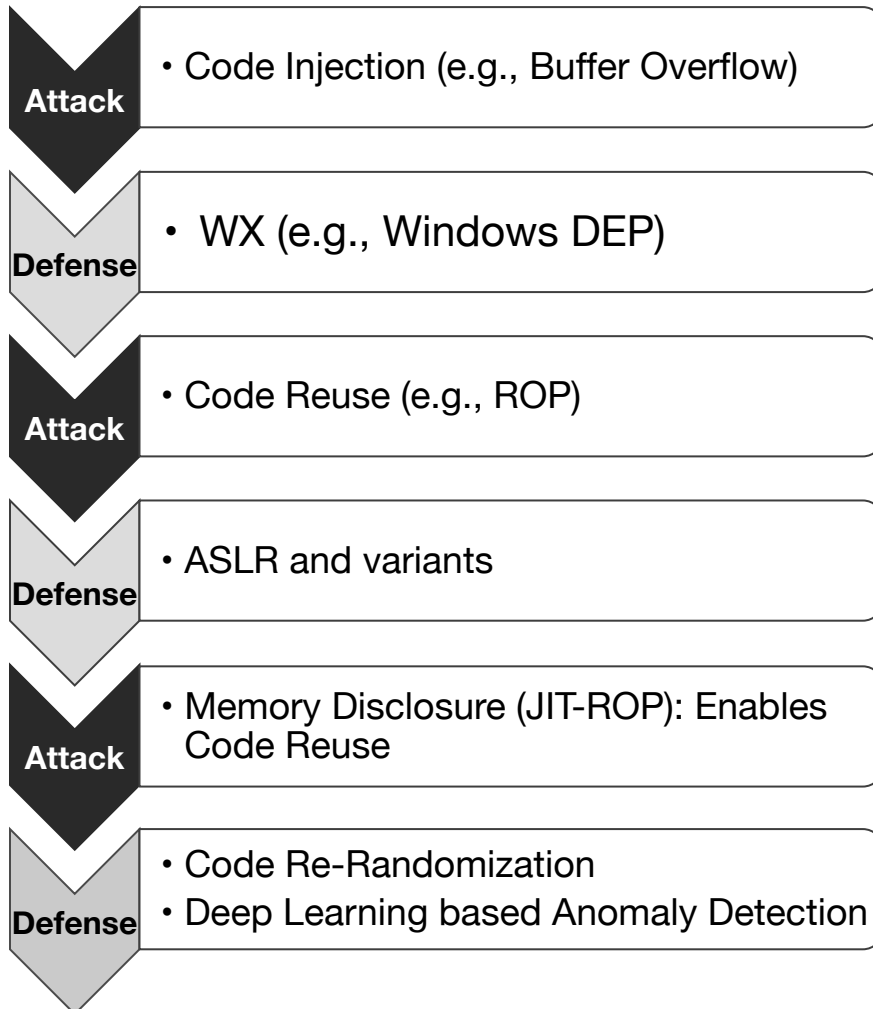
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# Deep Learning against code reuse attacks

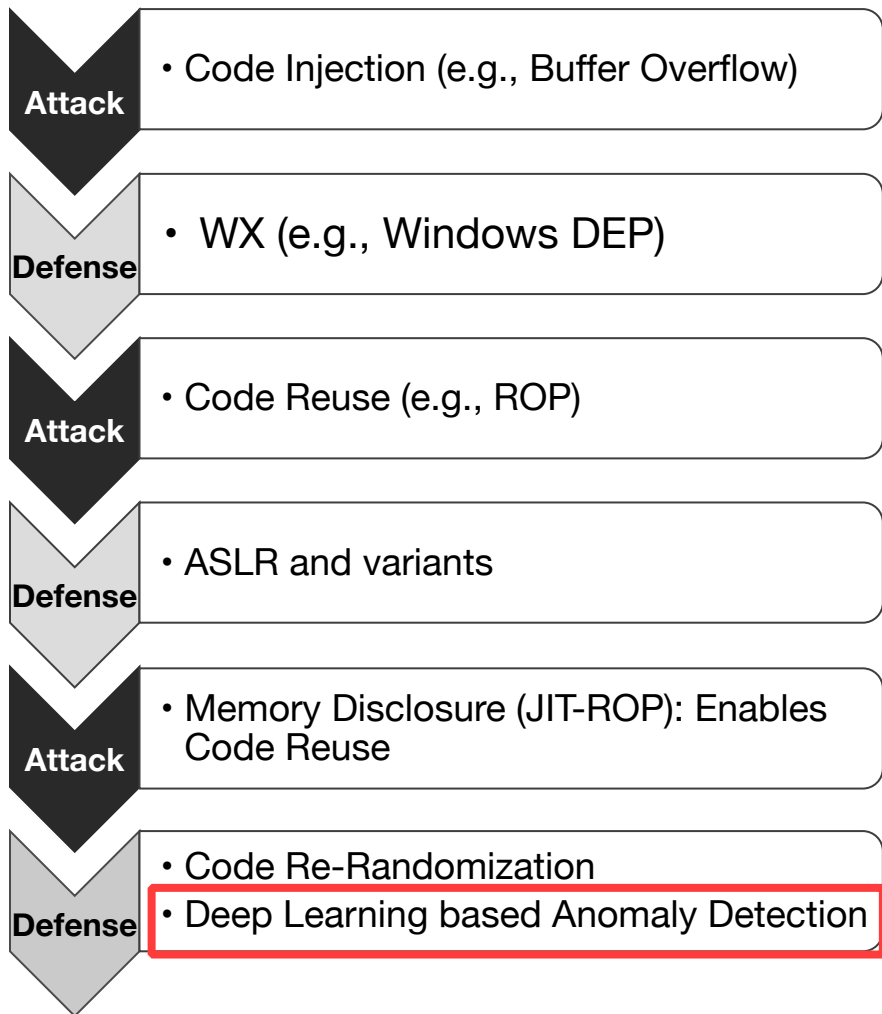
## The eternal war



- ☐ **Invalid and abnormal control flow of a program:**
  - ✓ Code Injection
  - ✓ Code Reuse
  - ✓ Memory Disclosure
- ☐ **Can be caused by:**
  - ✓ Human error (e.g., unauthorized use or operation of the program)
  - ✓ Software flaws (e.g., buffer overflow vulnerabilities)
  - ✓ Attacks by remote attackers
  - ✓ Malicious insiders (e.g., through drive-by downloads)

# Deep Learning against code reuse attacks

## The eternal war: DL against code reuse attacks



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## Classification Model Based Approaches:

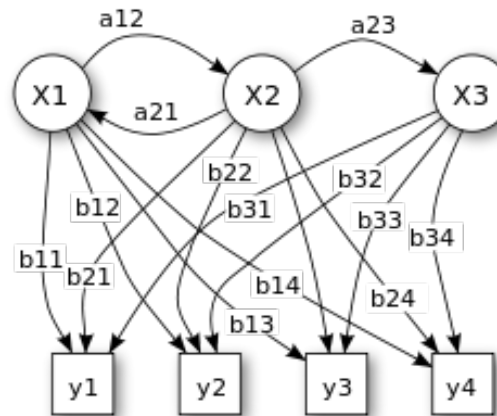
<i>Attack</i>	<i>Feature 1</i>	<i>Feature 2</i>	<i>...</i>	<i>Feature N</i>
<i>+</i>			<i>...</i>	
<i>-</i>			<i>...</i>	
<i>+</i>			<i>...</i>	
<i>-</i>			<i>....</i>	

- Datasets include known attacks only

### Limitation:

- No resilient to zero-day exploits (Signature Based)
- Difficult to train (representative datasets of the attacks are required)

## Hidden Markov Model (HMM) Based Approaches:



- Number of hidden states is equal to the number of system events
- $P(Y) = P(Y|X) P(X)$

### Limitation:

- **No memory nor long dependencies:** Observation at time  $t$  depends on observation at time  $t-1$

# Our Deep Learning Based Approach

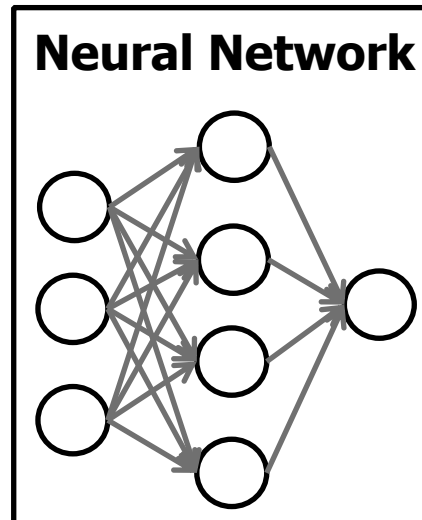
- For a given program, a code coverage is conducted to obtain all the possible execution traces.
- An event  $e_i$  is defined as a function call (**system or library call**) in the execution trace of a program.
- Each possible system event (function calls) is uniquely identified as they will form the vocabulary of system events.
- The Deep Learning model (neural network) is trained with the obtained sequences of events.
- The model is based on Recurrent Neural Networks: Long-Short Term Memory (**LSTM**) and Gated Recurrent Units (**GRU**.)

# Our Deep Learning Based Approach

- After training, given a sequence of events as input, the neural network produces as output an array of probabilities, one for each of the possible events in the system.
- At any time  $t$  each possible event (system call or library call) in the system is assigned a probability estimated with respect to the sequences of events **observed until** time  $t-1$ .
- At classification time  $t$ , the decision is made with respect to a pre-defined threshold of the top- $k$  most likely events.

# Our Deep Learning Based Approach

Set of all system  
events

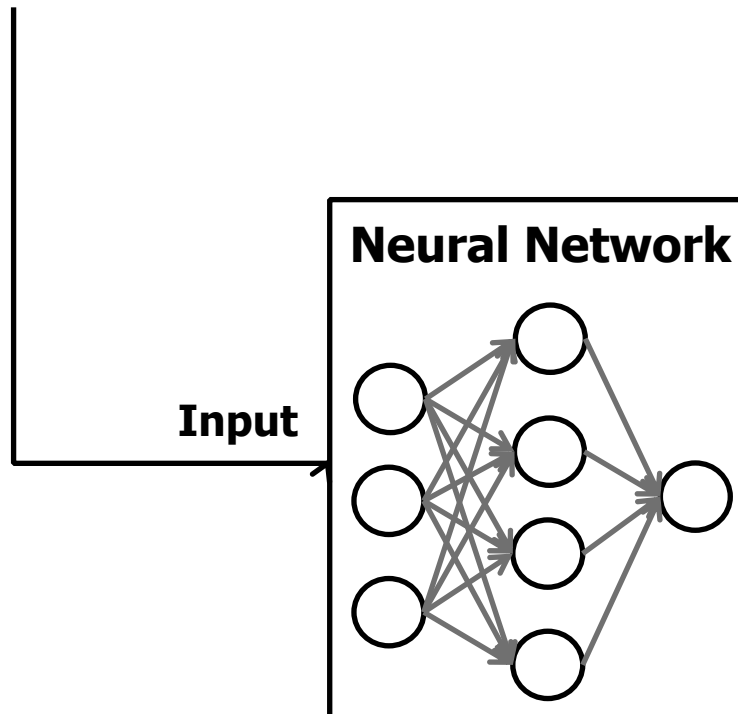


# Our Deep Learning Based Approach

Sequence of system events  
at  $t-1$



Set of all system  
events



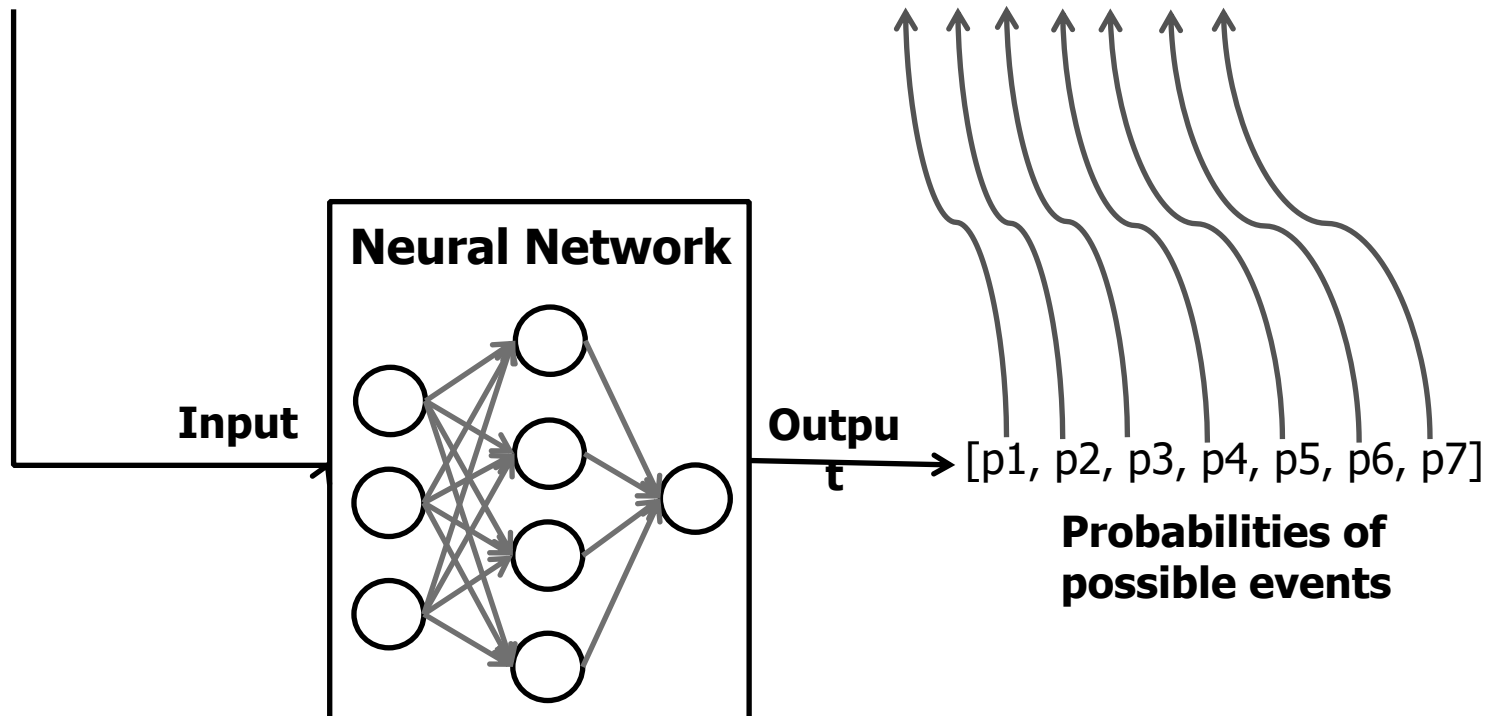


# Our Deep Learning Based Approach

Sequence of system events  
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Set of all system  
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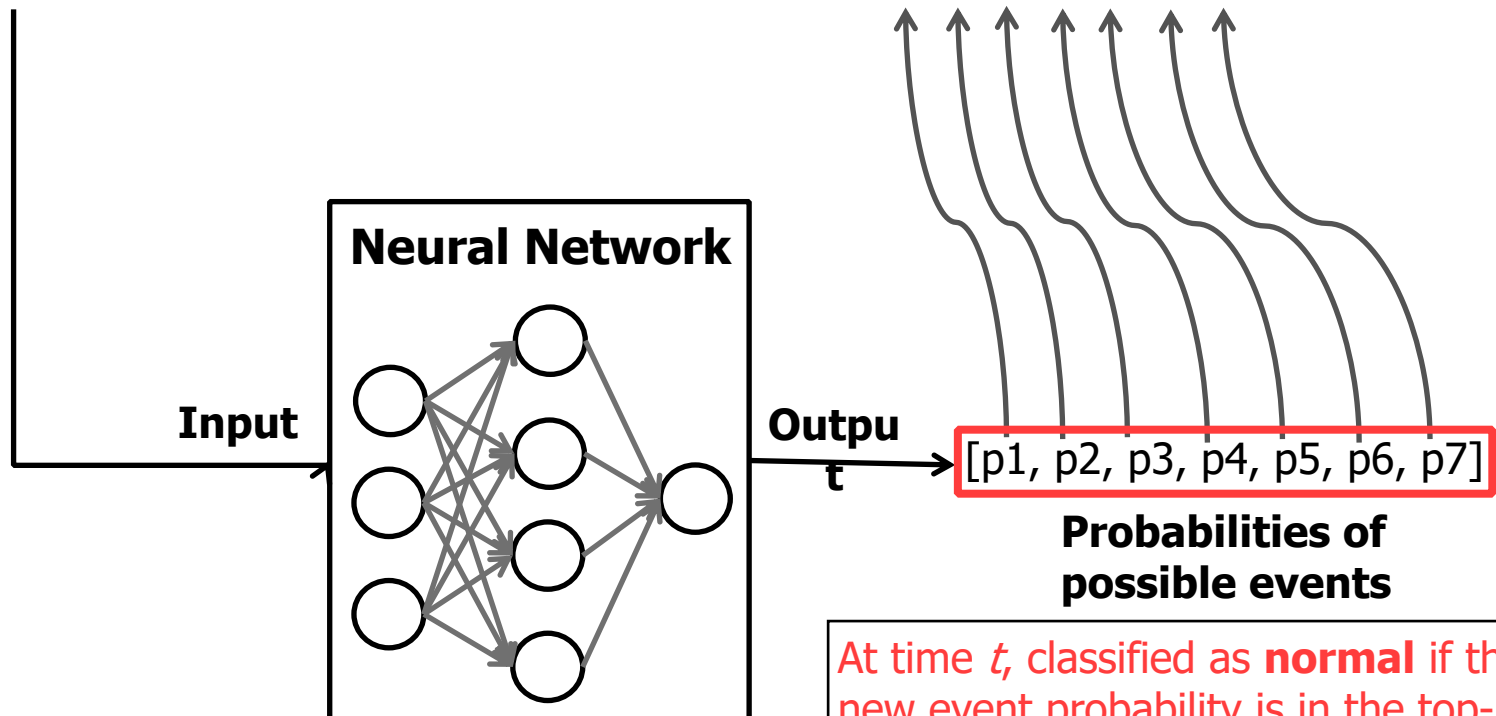


# Our Deep Learning Based Approach

Sequence of system events  
at  $t-1$



Set of all system  
events



At time  $t$ , classified as **normal** if the new event probability is in the top- $k$  probabilities; **anomalous** otherwise

# Our Deep Learning Based Approach

**Input:** Sequence of events in the system

**Output:** normal or anomalous

- **Step 1:** Define a finite set  $E$  of events  $e_1, e_2, \dots, e_N$  in the system. Events occur in a time-series fashion.
- **Step 2:** At time  $t - 1$ , given an observed series of events  $\{e_i^1, e_i^2, \dots, e_i^{t-1}\}$  (with  $i = 1, 2, \dots, N$ ) find the set  $K$  of the top  $k$  events to occur in time  $t$ .
- **Step 3:** At time  $t$ , the sequence  $\{e_i^1, e_i^2, \dots, e_i^{t-1}, e_i^t\}$  is non-anomalous if  $e_i^t \in K$ , otherwise anomalous.

**Algorithm 1:** Anomaly detection algorithm

# Benefits of Our Deep Learning Approach

## Properties of Recurrent Neural Networks (RNN)

**Stateful Representation**

**Memory**

**Long-Term Dependency**

**Variable Length Sequence**

- **Stateful Representation:** Maintenance of current state.
- **Memory:** Update state based on previous events.
- **Long-Term Dependency:** Track previous dependencies for long terms.
- **Variable Length Sequence:** Functional with non-fixed-length sequences.

# Benefits of Our Deep Learning Approach

- **Training datasets are full available:** Dynamic code behaviors are learned by training the model with non-malicious program traces in isolation.
- **Flow sensitive anomaly detection:** Given the execution paths  $P_1$ ,  $P_2$  and  $P_3$ , our technique captures the occurrence probabilities of each of their events; vital for high-precision anomaly detection.
- **Higher Power of Representation:** Context attributes such as caller functions can be added (e.g., function call `read@faa` is different from function call `read@foo`)
- **Higher granularity of the execution traces:** More granular traces are used for training, which improves precision in detection.

## Model Implementation

### MODEL

```
1 # -*- coding: utf-8 -*-
2 """
3 Deep Learning Based Anomaly Detection Model
4 """
5 # Imports
6 from __future__ import print_function
7 import torch
8 from torch.autograd import Variable
```



CUDA

GPU

- ☐ **Model**
  - ✓ Based on Recurrent Neural Networks (RNN): Long-Short Term Memory (LSTM) and Gated Recurrent Unit (GRU)
- ☐ **Programming Language: Python**
  - ✓ Compatible with several numerical computing libraries suitable for the use of GPUs
  - ✓ Some examples: Pytorch, TensorFlow and Theano
- ☐ **Computing Library: PyTorch**
  - ✓ Scientific computing package
  - ✓ Replacement of Numpy to take advantage of the power of GPUs
  - ✓ Python-friendly platform for neural networks (Deep Learning)
- ☐ **Parallel Computing Platform: CUDA**
  - ✓ Programming model to use GPUs for general purpose computing

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