

# CONTROLLER AREA NETWORK (CAN) DEEP PACKET INSPECTION



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# Car Hacking

“Immediately my accelerator stopped working. As I frantically pressed the pedal and watched the RPMs climb, the Jeep lost half its speed, then slowed to a crawl.” (Andy Greenberg, Wired)

**2014 Jeep Cherokee (remote attack)**

Engage brakes, Take control of steering

# Agenda

## AUTOMOTIVE SECURITY

- Connectivity in Modern Vehicles
- Controller Area Network (CAN) Vulnerabilities

## CAN ATTACKS

- Attack Types
- Detection & Prevention

## CAN ANOMALY DETECTOR

- Data
- Approach

## RESULTS & CONCLUSIONS

- Discussion of Results

- 1 Increasing Complexity & functionality
- 2 Interconnectedness

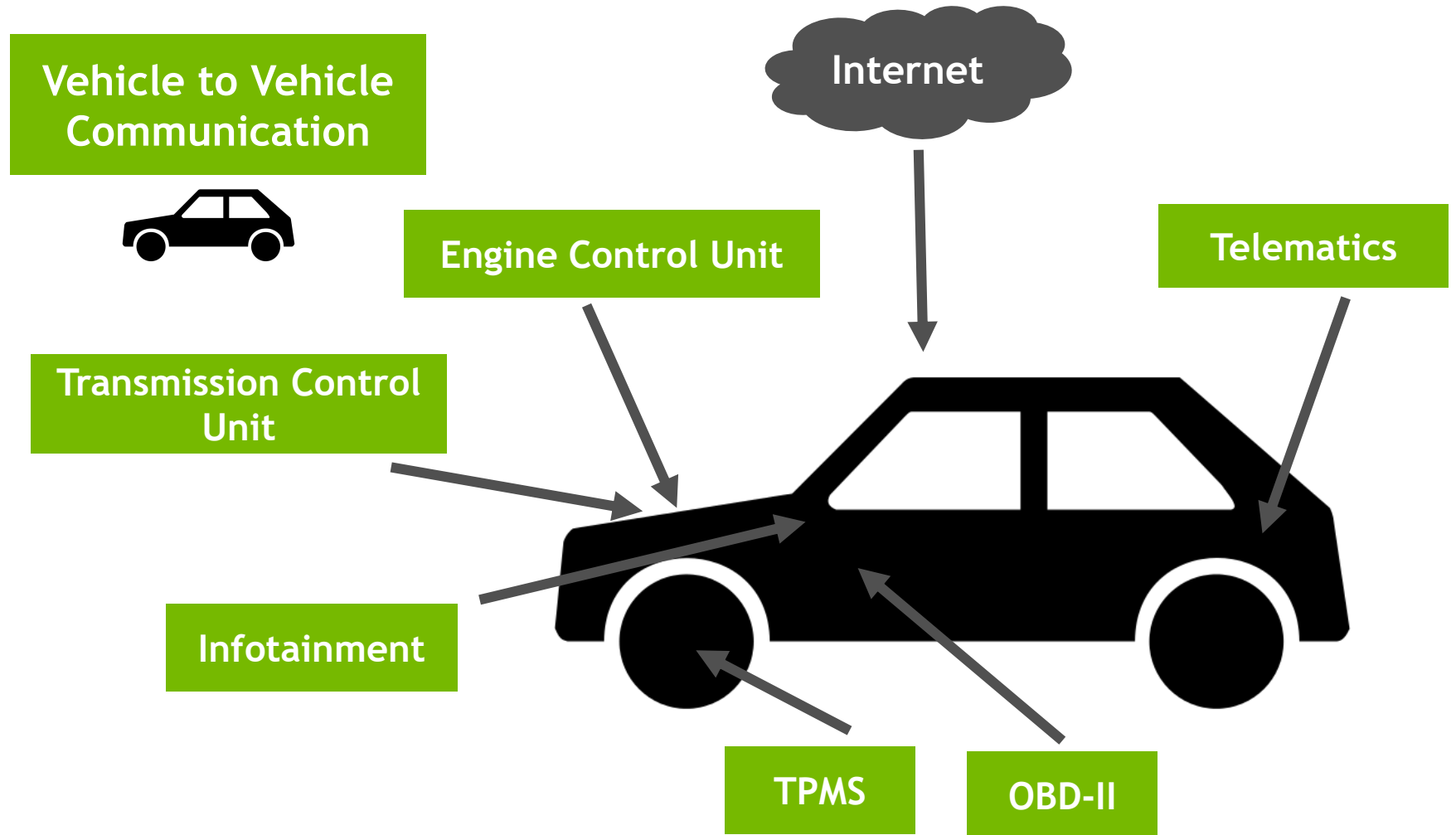


Figure1. Some connections of a modern car

# Controller Area Network (CAN) Security

# CAN Characteristics

**Message types:** Information,  
Diagnostic

**Message exchange:** Broadcast

**Message-based protocol,** no addressing

Arbitration method to  
resolve priorities

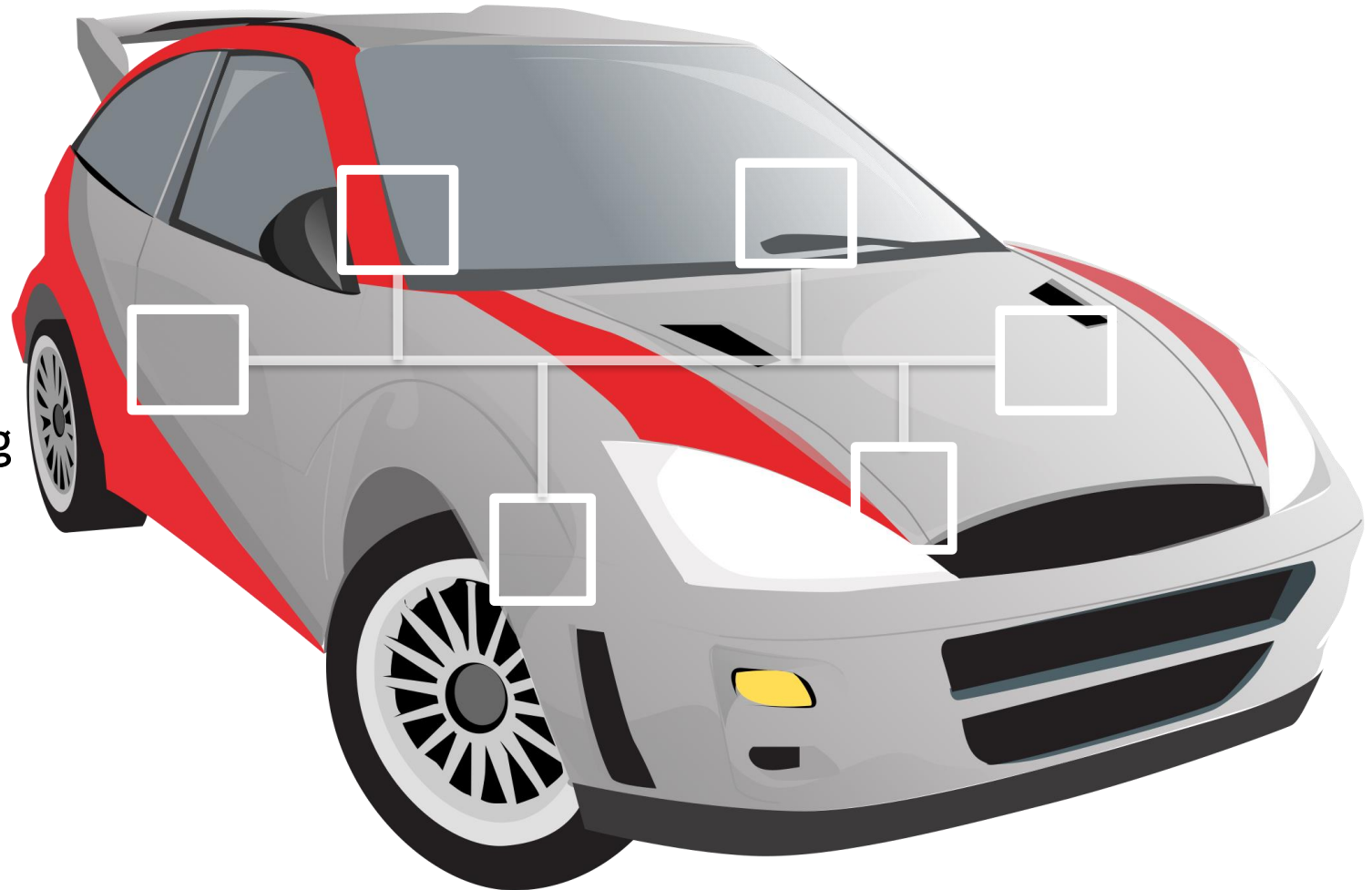


Figure2. The CAN network

# CAN Vulnerabilities

## Confidentiality

Every message sent on CAN is broadcast to every node → Eavesdropping

## Authenticity

Lack of sender authentication → Masquerading

## Availability

Arbitration rules (high priority messages) → Denial of Service

## Non Repudiation

No mechanisms to prove an ECU sent or received a message

## Most Critical Attack Types on CAN

### REPLAY

Replace message contents with some pre-recorded values

### INJECTION

Inject false messages appearing to be legitimate

### DOS

Flood the network



# Detection & Prevention

Device identification

CRYPTOGRAPHIC SERVICES

## ANOMALY DETECTION

Over-the-air updates

ECU software integrity

Tamper  
detection

ANTI-MALWARE

Secure boot

# Anomaly Detection

Finding unusual patterns in data that do not conform to expected behavior

E.g. fraud detection

# Types of Anomalies

## Point Anomaly

E.g. vehicle speed is **500 miles/hour**

## Collective Anomaly

E.g. vehicle speed is **80 miles/hour** & steering wheel angle is **90 degrees**

## Contextual (Conditional) Anomaly

E.g. vehicle speed changes from **50 miles/hour** to **80 miles/hour** in less than **X seconds**

# Controller Area Network (CAN) Anomaly Detector

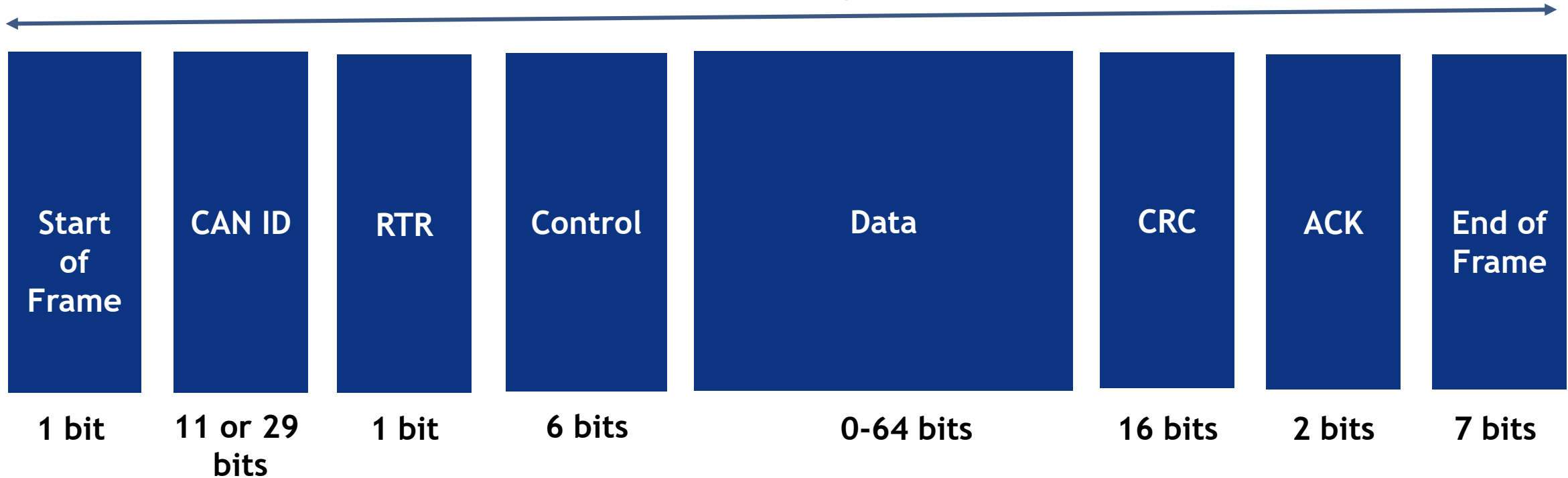
**Detect security-related CAN network anomalies resulting from malicious activities**

**Attacks: Injection, Replay**

**Anomalies: Contextual**

# CAN Frame

## CAN Message



# The Dataset: BB8 CAN flow

Timestamp	MessageID	Length	PAYLOAD								
			BYTE 0	BYTE 1	BYTE 2	BYTE 3	BYTE 4	BYTE 5	BYTE 6	BYTE 7	
574165791302335	101	8	143	4	140	4	160	4	155	4	W-Speed
574165791302421	102	8	3	254	55	254	15	254	15	254	SUSPENSION
574165791302432	103	4	1	0	252	255	0	0	0	0	ROLL&YAW
574165791302441	104	6	223	255	247	255	223	3	0	0	ACCELERATION

## Constraints

## Solutions

Multiple ECUs on the  
CAN BUS

Message ID Selection

Unstructured Data

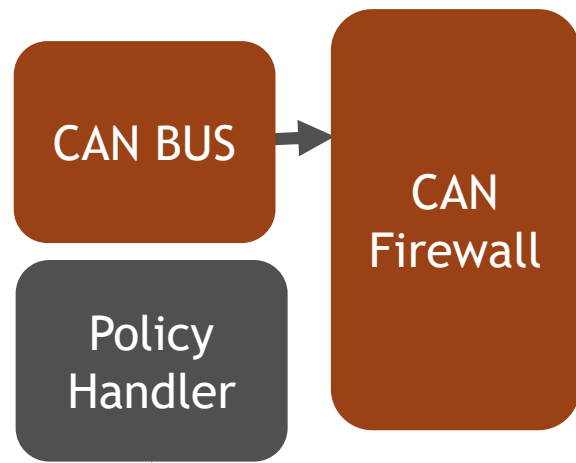
Content Extraction

Power/Performance

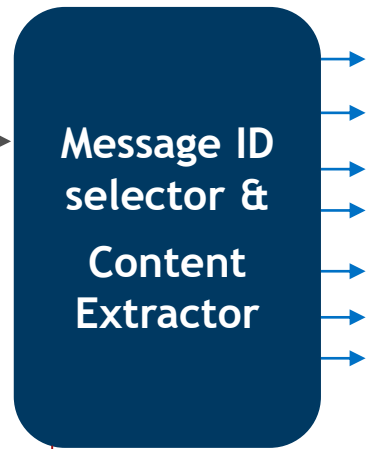
Recurrent Neural Networks (RNNs)



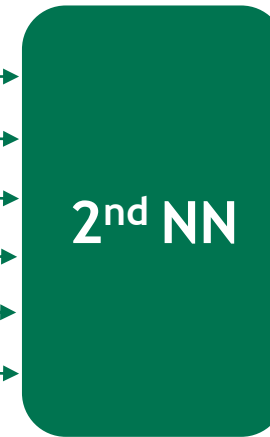
## Security Solution



## CAN Anomaly Detector



Errors

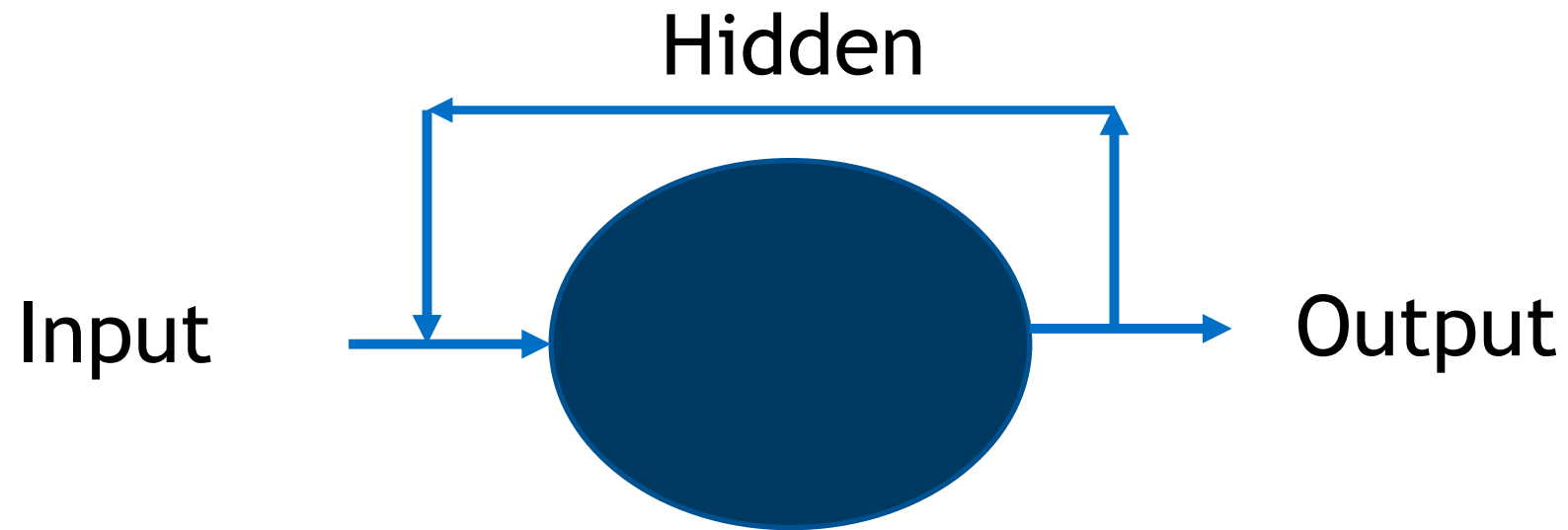


Output: Probability of an attack

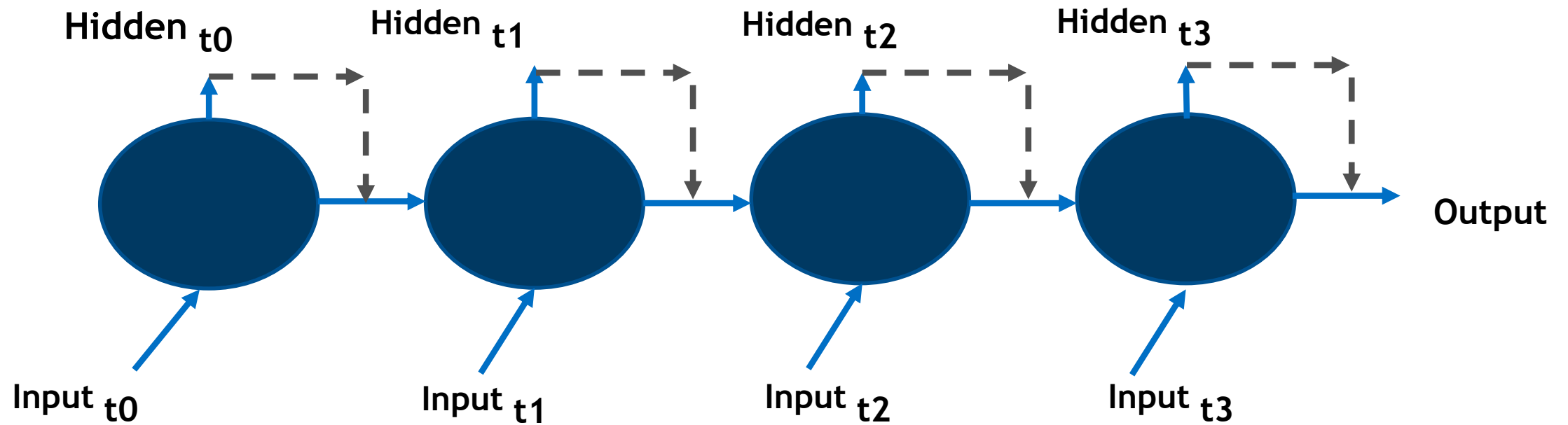
Contextual Anomaly Detection

Stage 2 Detection

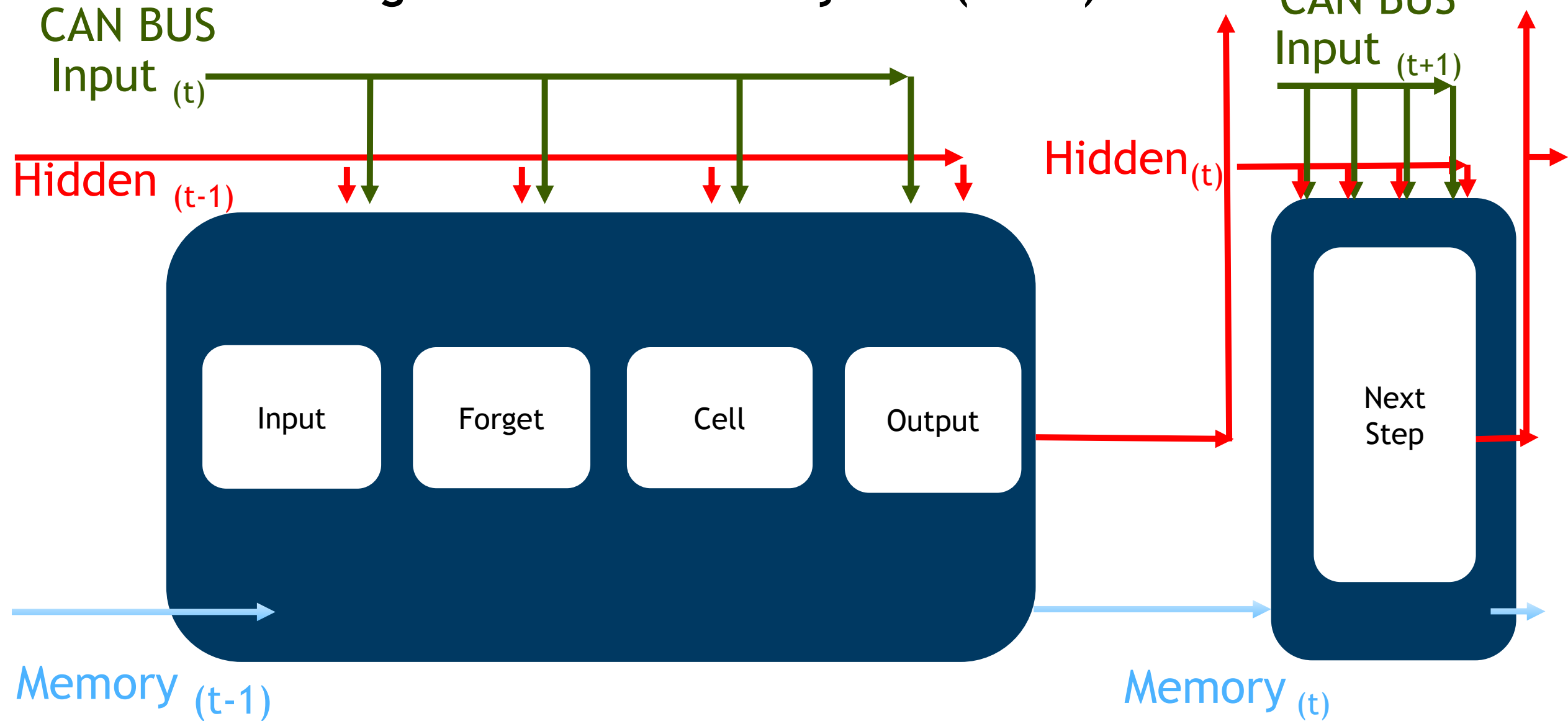
# Recurrent Neural Network (RNN)



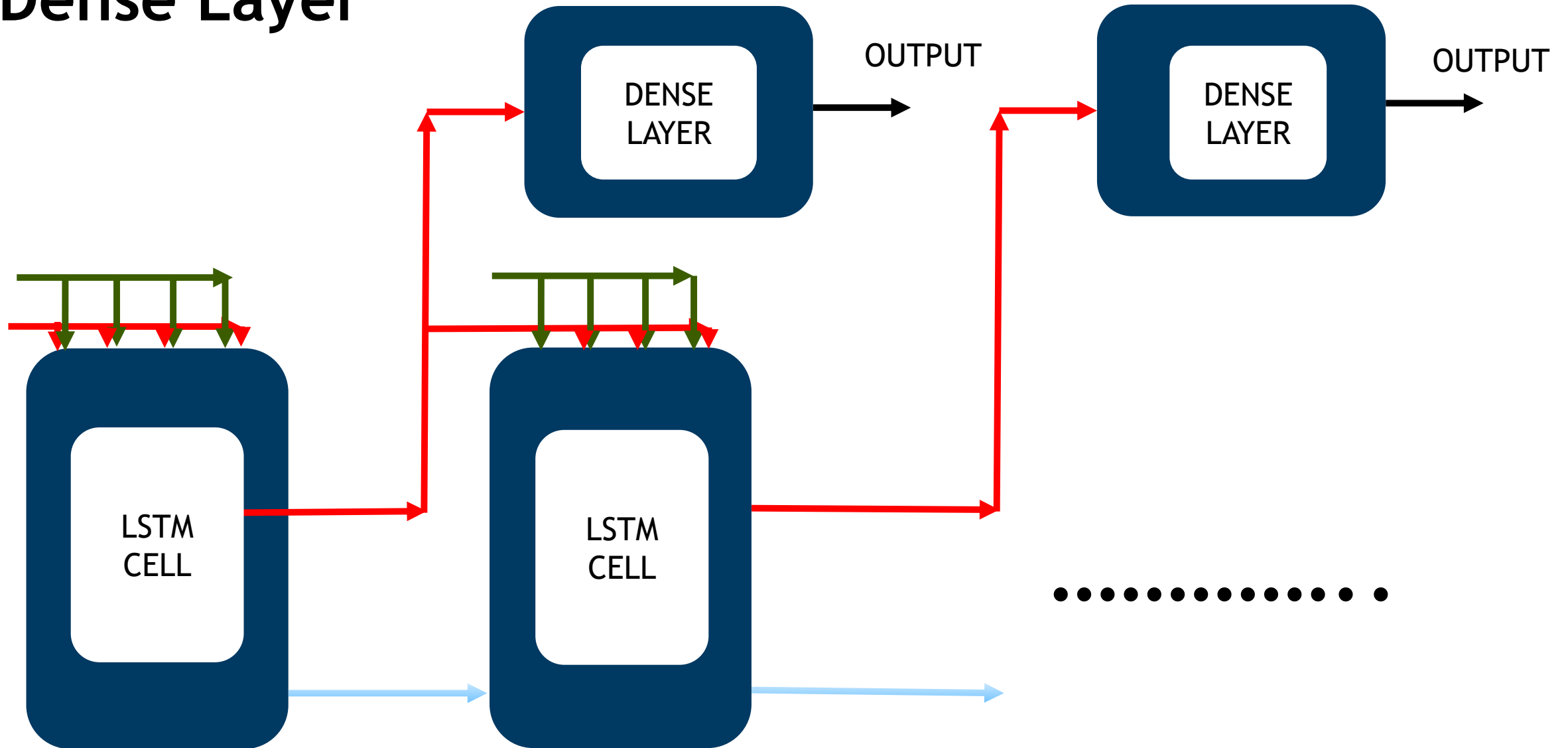
# Recurrent Neural Network (RNN)



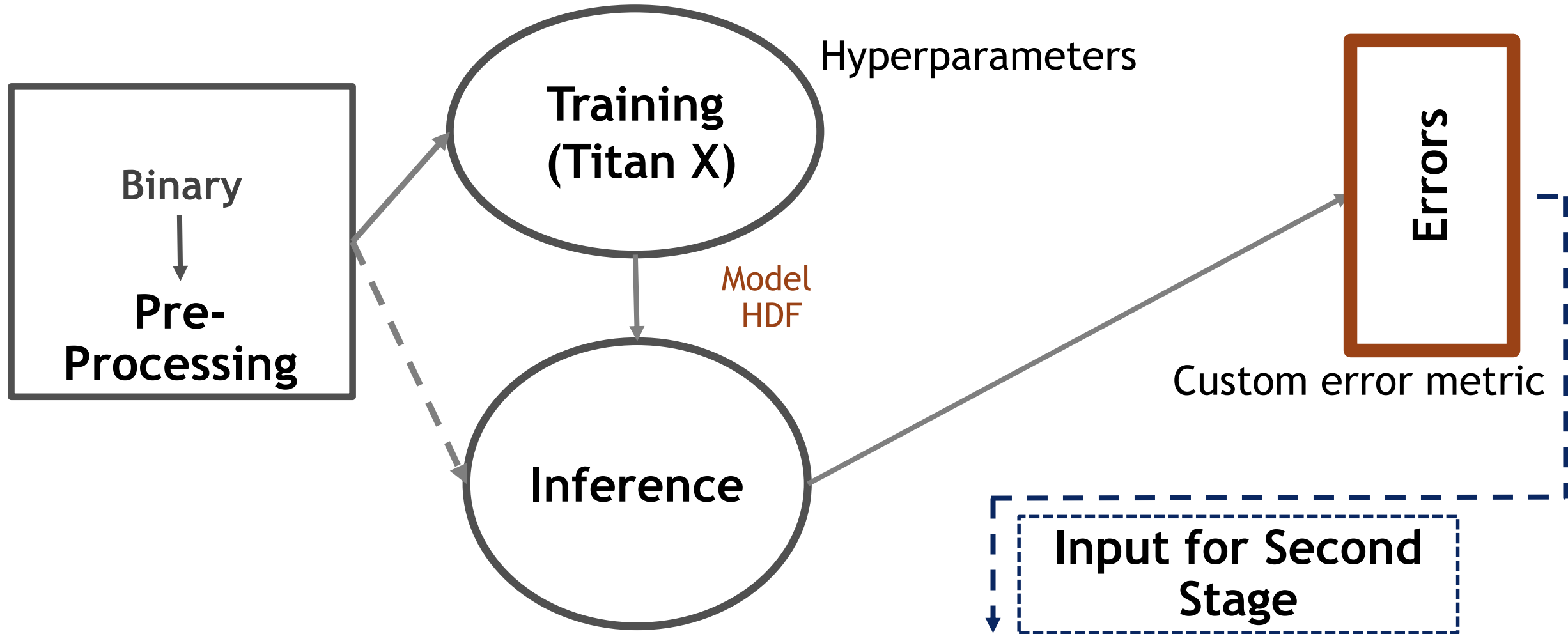
## Long Short Term Memory Cell (LSTM)



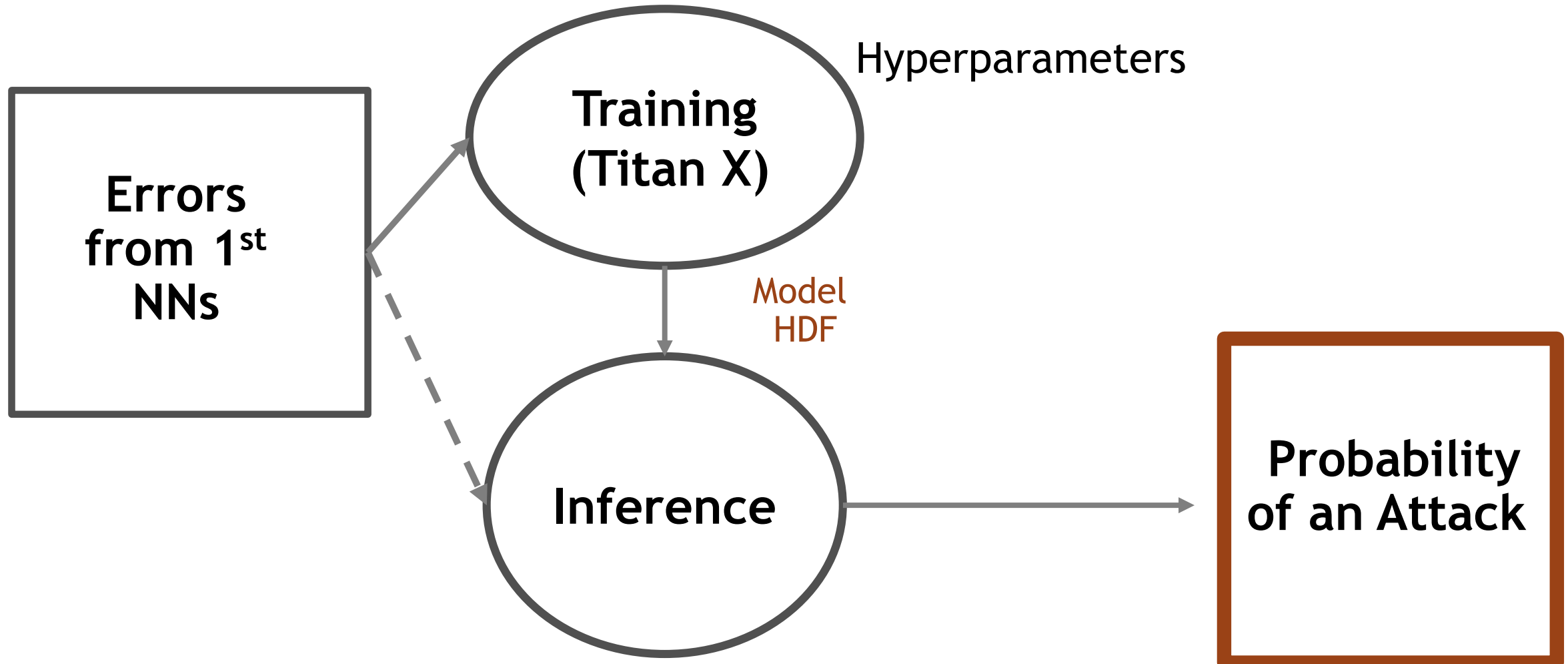
# Dense Layer



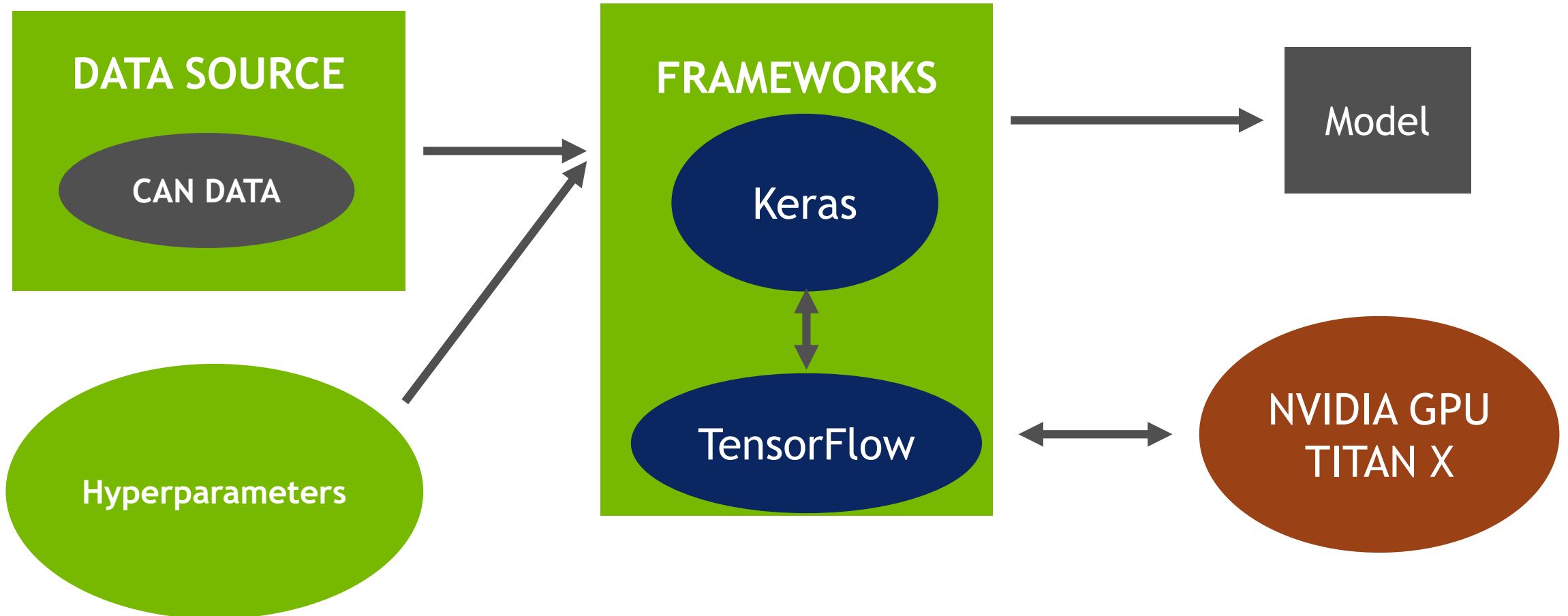
# Contextual Anomaly Detection Work Flow



# Contextual Anomaly Detection Work Flow-2<sup>nd</sup> Stage

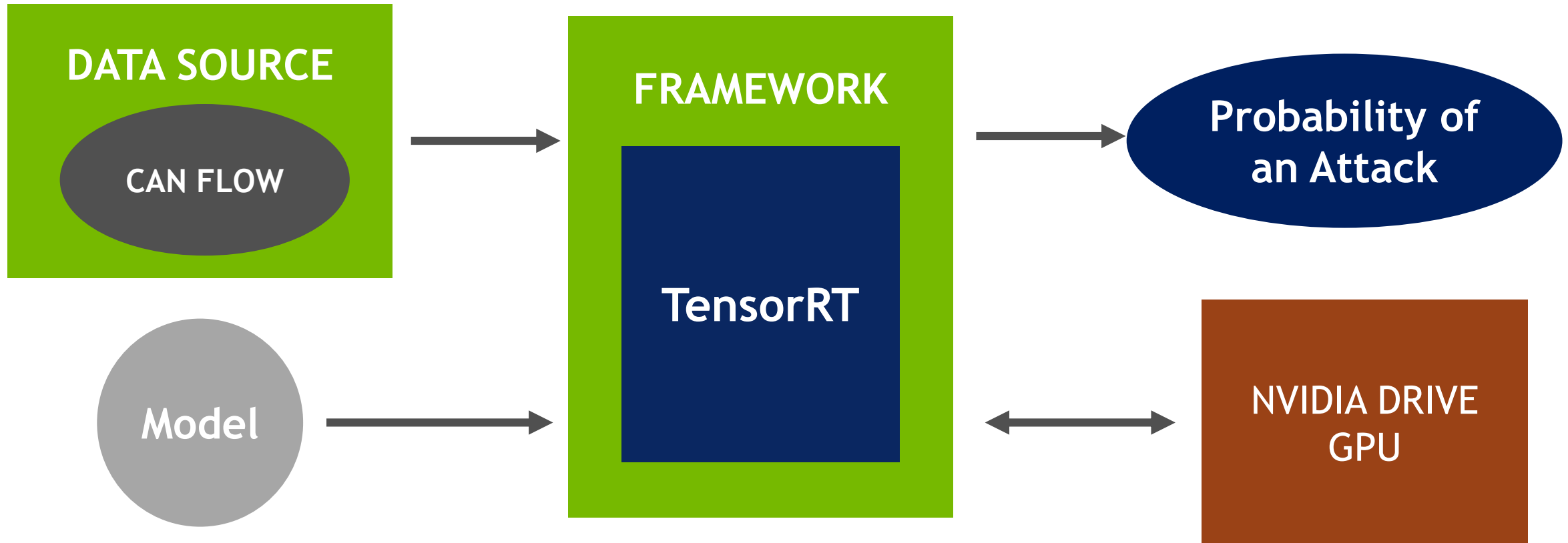


# Training Architecture





# Production Architecture

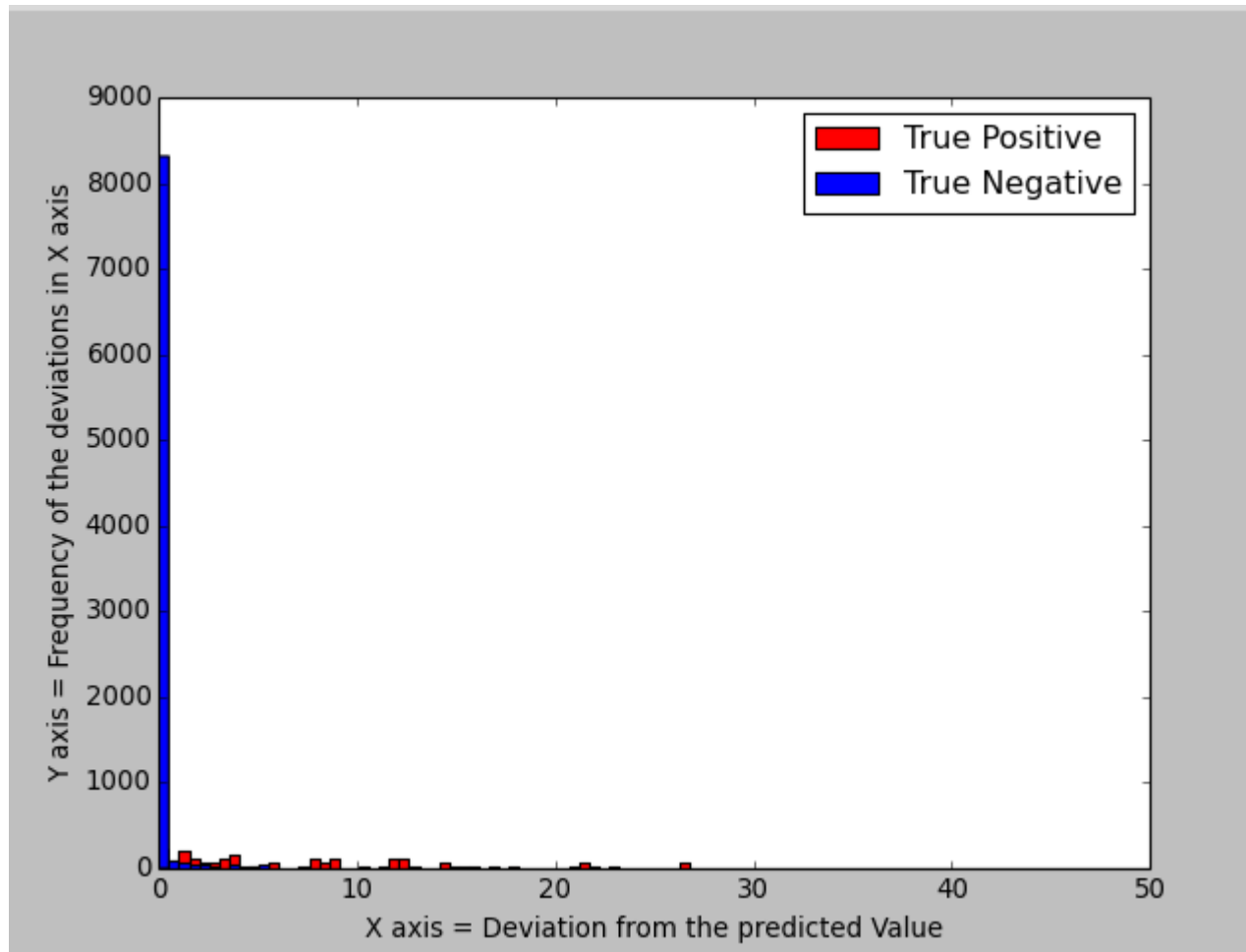


# Model Evaluation

## Using Sensitivity & Specificity

True Positives (Anomalies) caught

True Negatives allowed



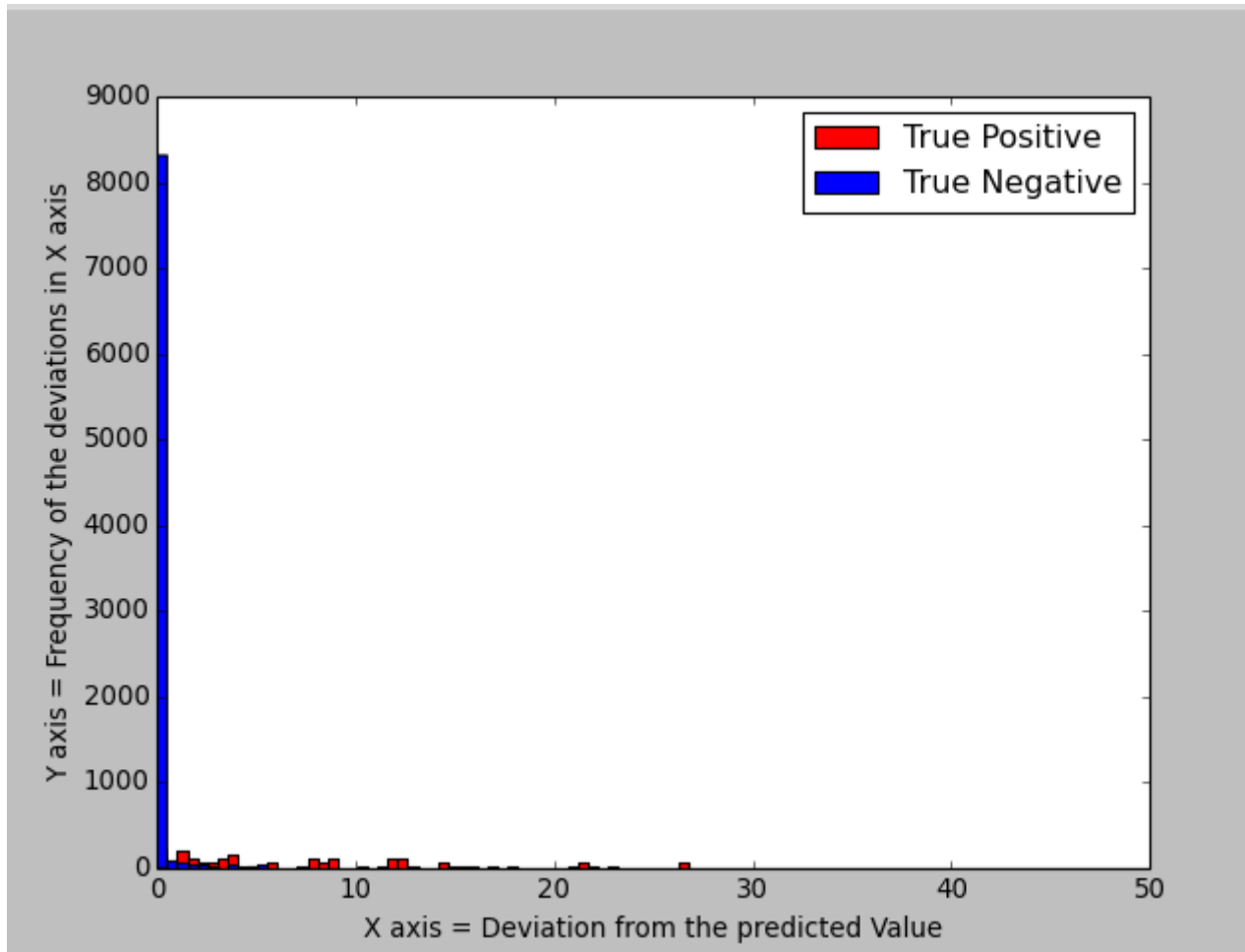
X axis: Deviation

Y axis: Frequency of errors

Median of Positives: **7.82**

Median of Negatives: **0.04**

Figure 3. Histogram - Error values output by the 2<sup>nd</sup> NN



**X axis: Deviation**

**Y axis: Frequency of errors**

➤ Sensitivity: 0.87

➤ Specificity: 0.94

Figure 4. Histogram - Error values output by the 2<sup>nd</sup> NN

## Results Per Attack Type

### Injection attacks

Total: 37

Detected: 32

### Replay attacks

Total: 42

Detected: 37

# Conclusion

A wall between Autonomous-Driving Software and the unsecured CAN-BUS

Low inference computational cost

Fast response

Offline training

Future Work

THANK YOU  
QUESTIONS?



# References

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Network Anomaly Detection - A machine learning perspective



# Images

Figure1. Connections of a modern car

Figure 2. CAN network

Figure 3. Histogram - Error values output by the 2nd NN

Figure 4. Histogram - Error values output by the 2nd NN

# APPENDICES



# Equations in a LSTM Cell – without the dense layer.

$$f_t = \sigma_g(W_f x_t + U_f h_{t-1} + b_f)$$

$$i_t = \sigma_g(W_i x_t + U_i h_{t-1} + b_i)$$

$$o_t = \sigma_g(W_o x_t + U_o h_{t-1} + b_o)$$

$$c_t = f_t \circ c_{t-1} + i_t \circ \sigma_c(W_c x_t + U_c h_{t-1} + b_c)$$

$$h_t = o_t \circ \sigma_h(c_t)$$