

# Unsupervised Fidelity-Based Anomaly Detection

## Results on CIC Datasets

PATENT PENDING · MAY 31, 2026 · MARTY OELRICH, COFOUNDER & CHIEF SCIENTIST

We have developed an unsupervised detection system that produces per-class, per-dimension, operator-tunable results on CICIDS-2017 and CSE-CIC-IDS2018. No attack labels are used for calibration or detection. Labels are used only for stratified fold construction and post-hoc benchmark scoring. We are sharing our results and methodology and would welcome your feedback.

The system calibrates from benign traffic — under 30 seconds in Precision mode, several minutes in Detection mode. No GPU. No internet. No signatures. The operator controls one parameter — sensitivity — and everything else emerges from calibration on the deployment environment's own traffic. Thresholds are derived from held-out benign rows only.

Every benchmark-pipeline number below is reproducible from our deployment package. We report every class, including the ones where the system is weak.

### What Is Different

The system does not optimize a single aggregate score. It measures behavioral fidelity across multiple independent dimensions simultaneously and preserves the decomposition for the operator. The operator sees which dimensions deviate, by how much, and whether independent measurements agree.

This changes what the system reports: not "this flow is malicious" but "this flow deviates on these dimensions, with this level of agreement, at a sensitivity setting you chose." Per-class recall at every sensitivity level replaces a single F1 number.

No attack labels are used for calibration or detection. Labels are used only for stratified fold construction and post-hoc benchmark scoring. Detection generalizes across datasets without retuning. The same code and parameters produce the CIC-2018 results without modification.

### CICIDS-2017 Corrected (Engelen 2021, 15 classes, 5-fold CV)

Two operating modes from a single codebase.

PER-CLASS RECALL ( $\alpha=0.05$ )

| ATTACK CLASS          | N       | PRECISION MODE | DETECTION MODE |
|-----------------------|---------|----------------|----------------|
| DoS Hulk              | 158,468 | 1.000          | 1.000          |
| DDoS                  | 95,144  | 1.000          | 1.000          |
| Botnet                | 736     | 1.000          | 1.000          |
| Web Brute Force       | 73      | 1.000          | 1.000          |
| Web XSS               | 18      | 1.000          | 1.000          |
| Heartbleed            | 11      | 1.000          | 1.000          |
| Infiltration          | 36      | 0.980          | 1.000          |
| DoS Slowhttptest      | 1,740   | 0.998          | 0.998          |
| Infiltration-Portscan | 71,767  | 0.993          | 0.998          |
| FTP-Patator           | 3,972   | 0.603          | 0.995          |
| SSH-Patator           | 2,961   | 0.988          | 0.995          |
| DoS Slowloris         | 3,859   | 0.912          | 0.984          |
| Portscan              | 159,066 | 0.947          | 0.952          |
| DoS GoldenEye         | 7,567   | 0.936          | 0.939          |
| SQL Injection         | 13      | 0.067          | 0.077          |

Precision mode: 11 of 15 classes at or above 0.93 recall. Detection mode: 14 of 15. SQL Injection (n=13) is a payload-level attack below the resolution of flow metadata. We report it.

Tuesday (Patator) is the hardest day for aggregate metrics — 7K attacks in 315K benign flows. Precision mode reaches FTP-Patator  $R=0.603$  at  $\alpha=0.05$ ; Detection mode reaches 0.995.

# Operating Curves

PRECISION MODE — FRIDAY (DDOS, PORTSCAN, BOTNET)

| A     | PREC  | REC   | FPR  | F1    |
|-------|-------|-------|------|-------|
| 0.001 | 0.996 | 0.283 | 0.1% | 0.440 |
| 0.005 | 0.990 | 0.478 | 0.4% | 0.644 |
| 0.01  | 0.984 | 0.650 | 0.8% | 0.784 |
| 0.02  | 0.973 | 0.836 | 1.5% | 0.902 |
| 0.03  | 0.964 | 0.944 | 2.3% | 0.954 |
| 0.05  | 0.959 | 0.968 | 3.8% | 0.963 |
| 0.07  | 0.945 | 0.976 | 5.1% | 0.960 |
| 0.10  | 0.927 | 0.993 | 7.1% | 0.959 |

PRECISION MODE — WEDNESDAY (DOS VARIANTS)

| A     | PREC  | REC   | FPR  | F1    |
|-------|-------|-------|------|-------|
| 0.001 | 0.997 | 0.626 | 0.1% | 0.765 |
| 0.005 | 0.991 | 0.905 | 0.5% | 0.946 |
| 0.01  | 0.984 | 0.947 | 0.9% | 0.965 |
| 0.02  | 0.971 | 0.970 | 1.6% | 0.971 |
| 0.03  | 0.958 | 0.976 | 2.4% | 0.967 |
| 0.05  | 0.936 | 0.991 | 3.8% | 0.963 |
| 0.10  | 0.888 | 1.000 | 7.0% | 0.940 |

DETECTION MODE — FRIDAY

| A     | PREC  | REC   | FPR  | F1    |
|-------|-------|-------|------|-------|
| 0.001 | 0.904 | 0.552 | 5.3% | 0.685 |
| 0.005 | 0.908 | 0.589 | 5.4% | 0.715 |
| 0.01  | 0.922 | 0.713 | 5.4% | 0.804 |
| 0.02  | 0.933 | 0.869 | 5.6% | 0.900 |
| 0.03  | 0.937 | 0.944 | 5.7% | 0.941 |
| 0.05  | 0.933 | 0.970 | 6.3% | 0.951 |
| 0.07  | 0.927 | 0.978 | 7.0% | 0.952 |
| 0.10  | 0.915 | 0.994 | 8.3% | 0.953 |

PRECISION MODE — THURSDAY (INFILTRATION, WEB ATTACKS)

| A    | PREC  | REC   | FPR  | F1    |
|------|-------|-------|------|-------|
| 0.01 | 0.940 | 0.499 | 0.8% | 0.651 |
| 0.02 | 0.931 | 0.825 | 1.6% | 0.874 |
| 0.03 | 0.912 | 0.926 | 2.3% | 0.918 |
| 0.05 | 0.867 | 0.967 | 3.8% | 0.915 |
| 0.10 | 0.778 | 0.973 | 7.1% | 0.864 |

## CSE-CIC-IDS2018 (No Retuning)

Same engine. Same parameters. No modification for CIC-2018.

| ATTACK CLASS      | FILE     | N       | PRECISION (A=0.05) | DETECTION (A=0.05) |
|-------------------|----------|---------|--------------------|--------------------|
| SSH-BruteForce    | Wed-14   | 94,197  | 1.000              | 1.000              |
| DDoS-HOIC         | Wed-21   | 248,069 | 1.000              | 1.000              |
| DDoS-LOIC-HTTP    | Tue-20   | 246,049 | 0.980              | 1.000              |
| DDoS-LOIC-UDP     | multiple | 815     | 1.000              | 1.000              |
| DoS GoldenEye     | Thu-15   | 22,560  | 0.997              | 1.000              |
| DoS Slowloris     | Thu-15   | 8,490   | 1.000              | 1.000              |
| DoS Hulk          | Fri-16   | 247,177 | 0.912              | 1.000              |
| Web Brute Force   | multiple | 131     | 1.000              | 1.000              |
| Web XSS           | multiple | 113     | 1.000              | 1.000              |
| Infiltration-NMAP | Thu-01   | 17,407  | 0.990              | 0.999              |
| Infiltration-NMAP | Wed-28   | 21,591  | 0.885              | 0.988              |
| Web SQL           | multiple | 39      | 0.650–0.836        | 0.650–0.836        |

Detection Mode improves most classes not already at 1.000 (Web SQL unchanged). Both modes validated via benchmark pipeline; production wrapper Detection Mode pending a loading fix for files with attacks at end of file.

# Adversarial Validation

| TEST                                             | RESULT                                                            |
|--------------------------------------------------|-------------------------------------------------------------------|
| Temporal split (train 60%, test 40%, no shuffle) | Detection holds across time-shifted traffic                       |
| Calibration contamination                        | Guard-enabled Tuesday E2E refused 1%, 3%, 5% poisoned calibration |
| Behavioral head ablation                         | Per-dimension importance quantified                               |
| Cross-dataset transfer                           | 2017 → 2018 without retuning                                      |

Calibration poisoning at 1% is catastrophic without the integrity guard. With the guard, the system detects contamination and refuses to go live. Guard-enabled E2E validation completed on Tuesday (Patator classes). Multi-day guard-enabled validation is pending. We document both the vulnerability and the defense.

## Documented Limits

| LIMITATION             | DETAIL                                                           |
|------------------------|------------------------------------------------------------------|
| SQL Injection (n=13)   | R=0.077. Payload-level, below flow metadata resolution.          |
| Detection mode FPR     | Adds 2.5% FPR over Precision mode.                               |
| Feature extraction     | Requires CICFlowMeter-compatible feature extraction.             |
| Calibration assumption | Clean benign baseline required (integrity guard addresses this). |

## Specifications

| PROPERTY                         | VALUE                     | PROPERTY              | VALUE                          |
|----------------------------------|---------------------------|-----------------------|--------------------------------|
| Labels for calibration/detection | None                      | Hardware              | CPU only                       |
| Calibration time                 | < 30s (Prec) / mins (Det) | Air-gapped deployment | Yes                            |
| Engine size                      | 560 KB                    | Per-flow latency      | ~1 ms / ~4 ms                  |
| Total deployment footprint       | < 70 MB                   | Throughput            | ~1M / ~250K flows/s            |
| Calibration state                | ~50 MB                    | Operator parameters   | Sensitivity ( $\alpha$ ), mode |
| Runtime memory                   | ~200 / ~250 MB            | Reproducibility       | MD5-tracked                    |

## Deployment Security

The engine includes a calibration integrity guard (frozen reference verifies data before deployment), continuous runtime coherence monitoring, autonomous shutdown on integrity compromise, and cryptographic provenance on detection verdicts. The system operates air-gapped with no telemetry and no external dependencies at runtime.

## Methodology

5-fold StratifiedKFold (shuffle=True, random\_state=42). CICIDS-2017: Engelen, Rimmer & Joosen (IEEE SPW 2021), Distrinet/KU Leuven, MD5-verified. CSE-CIC-IDS2018: CSE & CIC with Distrinet corrections. Thresholds from held-out benign rows only. Hardware: Apple M2 Mac Mini, 16GB, CPU only. Engine v2.1.

## Why We Are Reaching Out

Your datasets are the standard benchmark for this work. We have validated extensively against them and would value your perspective on the methodology and results. We are happy to share the deployment package for independent evaluation or validate on additional datasets.

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