

Authors: Joseph Cutchall (NMT B.S. Candidate, CSE), Wesley Pick-Roth (NMT M.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Introduction and Motivation

- NMT is exploring novel mechanisms for detecting malicious activity in system log traffic.
- The project is an extension to an initial collaboration with SNL in which basic research and preliminary experiments were conducted.
- **Overall Motivation:** Look for mechanisms for detecting/classifying non-obvious malicious activity in systems by leveraging existing and/or emerging machine learning (ML) techniques.
 - *Multi-faceted, team approach:* several NMT students and researchers are working on distinct, yet highly related approaches
 - *Big Picture:* work towards the application of ordered-policy memory networks
- **Guiding Principles:**
 - Work with real-world or at least emulated data sets
 - Continuously improve data pre-processing and refinement while successively working towards more complex ML regimes (in particular, Ordered Memory Policy Networks, or OMPNs)

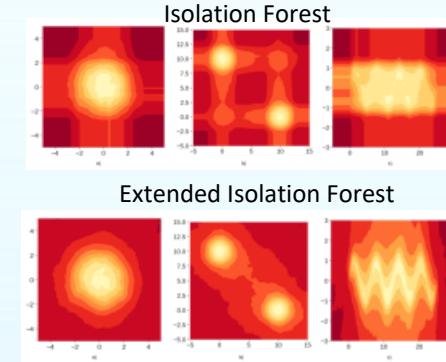
Authors: Joseph Cutchall (NMT B.S. Candidate, CSE)), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Approach

- **Extended Isolation Forests (EIFs)**
 - EIFs are an addition to the IF algorithm that is made to tackle the problem Isolation forests had with contamination of scoring data, that can lead to misclassification of outliers.
- **Approach**
 - Previous work: Use of Isolation Forests (IFs) against log data
 - Learning: validate previous IF experiment results
 - Development: expand code base to accommodate EIFs
 - Application: perform previous experiments with EIFs
 - Analysis: compare results / iterate over parameters as necessary
- **Data**
 - ICASA internal data, produced from its Process Analysis for Cyber Operations project (PACO)
 - (Recently available) TracerFIRE 9 (TF9) data
- **Supplemental**
 - Refinement of existing data (e.g., feature space)
 - Pre-processing of TF9 data

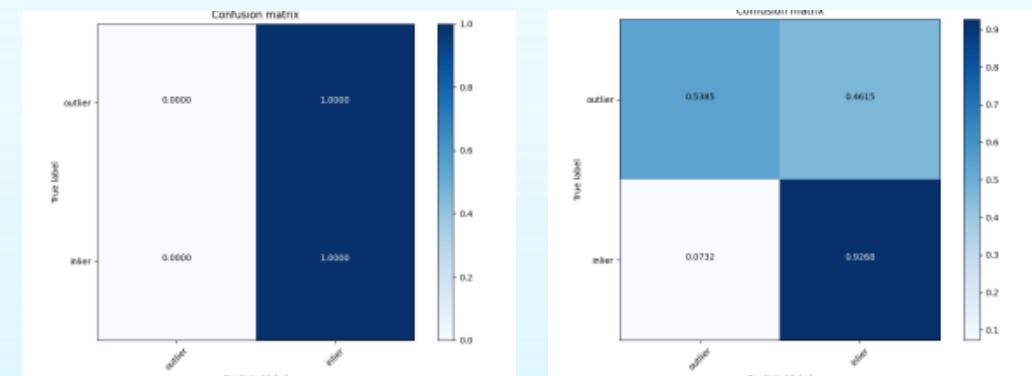
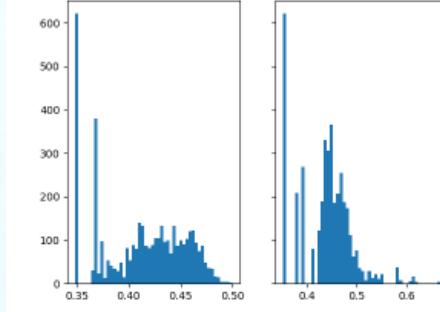


Pictures From <https://github.com/sahandha/eif#Use>

Authors: Joseph Cutchall (NMT B.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)

Current Status and Results (if any)

- EIFs against PACO data
 - Initial Results Show that the EIFs were able to better detect outliers while generally reducing the number of false positives reported.
 - With one specific run we were able to see a 53.8% increase in event detection while only increasing the false positive reporting of 6.5%
- TF9 Data
 - We are currently working on preprocessing TF9 Data to be able to run it through IFs and EIFs





Authors: Joseph Cutchall (NMT B.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Impact of Work

- **Conclusions:**

- With using Extended Isolation Forests on the Paco Data set we were able to get better results in classifying bad actors from the data set. Though we were not able to classify all of the bad actors in the data. With the Implementation of EIFs it seems that to further our results we need to find a way for bad actor data to be better represented as anomalous after preprocessing.

- **Impact of work:**

- Comprehensive analysis of methodology against available data (building on previous cursory experiments)
- Parameter sweeps are helping drive better understanding of data
- In turn, will help to inform future experiments

Authors: Wesley Pick-Roth (NMT M.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Approach

- **Isolation Forests**
 - Determine outliers by randomly partitioning data
 - Likelihood of outlier proportional to ease of isolation
- **Approach**
 - Previous work: Use of Isolation Forests (IFs) against log data
 - Learning: Validate previous IF experiment results
 - Development: Explore different encoding strategies of datasets
 - Incorporation of timestamp in encoded events
 - Alternative encodings of *image* (executable) field
 - Application: Perform previous experiments with IFs using alternative encodings
 - Analysis: Compare results / iterate over parameters as necessary
- **Data**
 - ICASA internal data, produced from its Process Analysis for Cyber Operations project (PACO)
 - (Recently available) TracerFIRE 9 (TF9) data
- **Supplemental**
 - Pre-processing of TF9 data

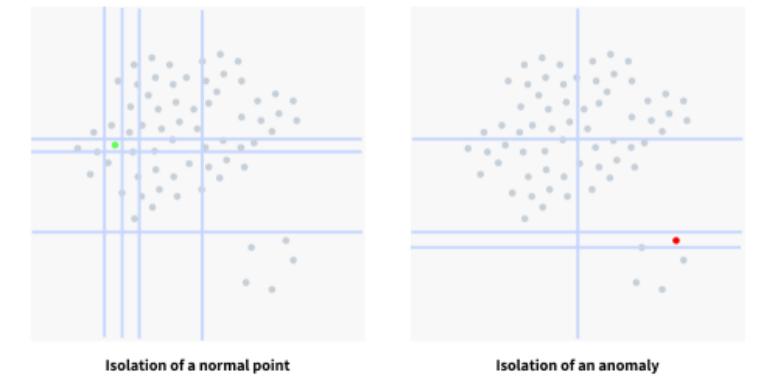
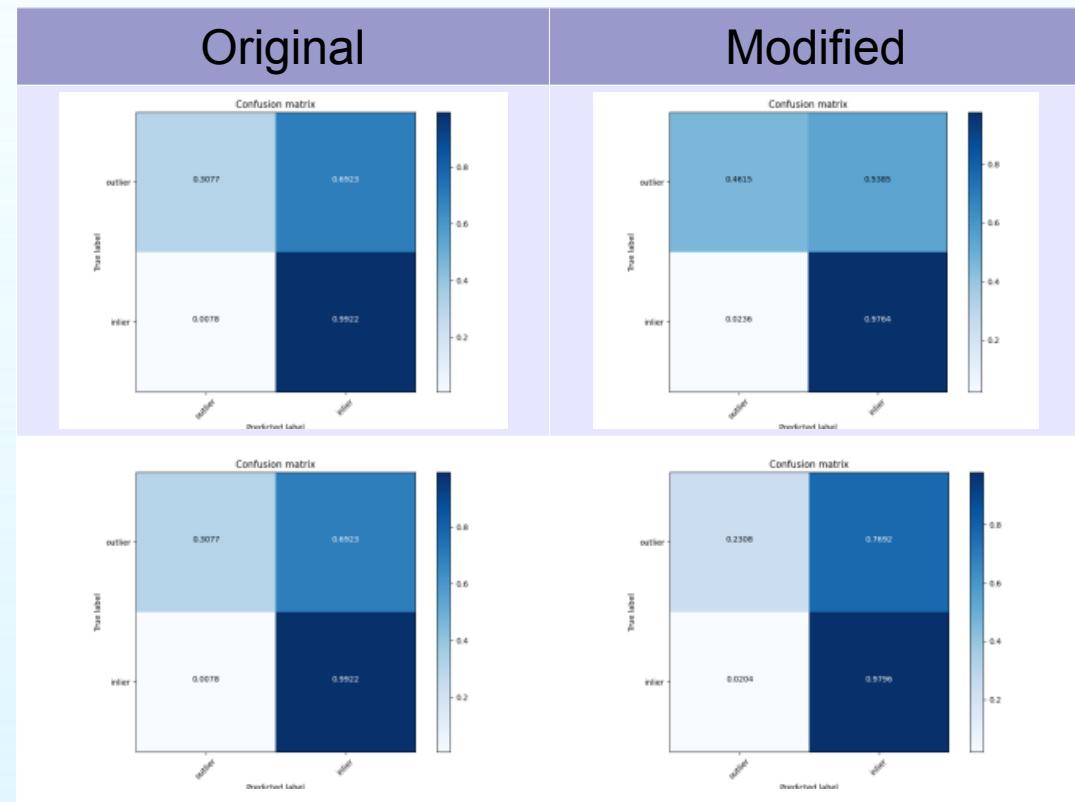


Image of isolation forest partitioning from
<https://laptrinhx.com/isolation-forest-algorithm-for-anomaly-detection-3651893833/>

Authors: Wesley Pick-Roth (NMT M.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)

Current Status and Results (if any)

- Incorporation of Timestamps**
 - Include ‘time since epoch’ and/or ‘time since midnight’ for time relation
 - Finds more anomalous data at expense of more misclassified safe data
 - Image explanation:
 - True positive rate - 0.3077 to 0.4615
 - True negative rate - 0.9922 to 0.9764
- Change of *image* encoding**
 - Change *image* data from single numeric feature to one-hot encoding
 - Contrary to earlier work, performs better with smaller parameterization
 - Image explanation:
 - True positive rate - 0.3077 to 0.2308
 - True negative rate - 0.9922 to 0.9736
 - Not tuned





Authors: Wesley Pick-Roth (NMT M.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Impact of Work

- **Conclusions:**
 - Implementation of an order of events needs further exploration
 - New image encoding gives promising results without tuning
- **Impact of work:**
 - Reevaluating data features may lead to further improvements
 - Fewer learning parameters adds more generality
 - Evaluating parameters provided insight to future work
 - Use sequence of process events, rather than time
 - In turn, will help to inform future experiments (as detailed in the next slide)

Authors: Joseph Cutchall (NMT B.S. Candidate, CSE), Wesley Pick-Roth (NMT M.S. Candidate, CSE), Max Planck (NMT PhD Candidate, CSE; ICASA Technical Director), Nathanael Brown (Sandia, 8721)



Fall 2021 LDRD Virtual Poster Session

Challenges and Risks / Next Steps and Future Work

- Challenges / Risks
 - EIFs seem to run much less efficiently than regular IFs
 - Parameter tuning needed for best results
 - Scaling may be an issue for both paradigms presented above
 - Heavy reliance on preprocessing of Data
- Next Steps:
 - Run TF9 data through IFs and EIFs
 - Build graphs from event relations in data and run metrics as features in the Forests
 - Extract graph metrics / fuse with existing ‘intrinsic’ feature set and re-run previous experiments
- Future Work:
 - Use graph data to construct event ‘sequences’ and develop labeling
 - Use a sequence-based learning system (e.g. LSTM, HMM) as a building block towards OMPNs
 - Experiment with OMPNs in an attempt to abstract sequence data to labels that align with malicious activity (e.g., the cyber-kill chain)