

REDUCING FALSE POSITIVES IN INTRUSION DETECTION



RAPID DETECTION & RESPONSE SERVICE

- F-Secure solution for intrusion detection in **corporate** IT environments
- Sensors (=program on a client computer) sends **events** to our backend
- In the real-time processing component of the solution, events are analyzed by machine learning models and rules, triggering **detections**
- Detections are evaluated by the experts at Rapid Detection Center (**RDC**)
- RDC decides whether to contact the client or dismiss the detection
- Rules are prone to false positives. Here we aim to predict which rule-based detections could be false positives to prioritize the most likely incidents



**RAPID DETECTION
CENTER**
F-Secure

**WE SEE THINGS
OTHERS DON'T**



CHALLENGES WITH DATA #1

- Underlying distribution changes constantly
 - new or updated rules
 - new or changed clients
 - new or changed applications on clients
 - ...
- Significant class imbalance

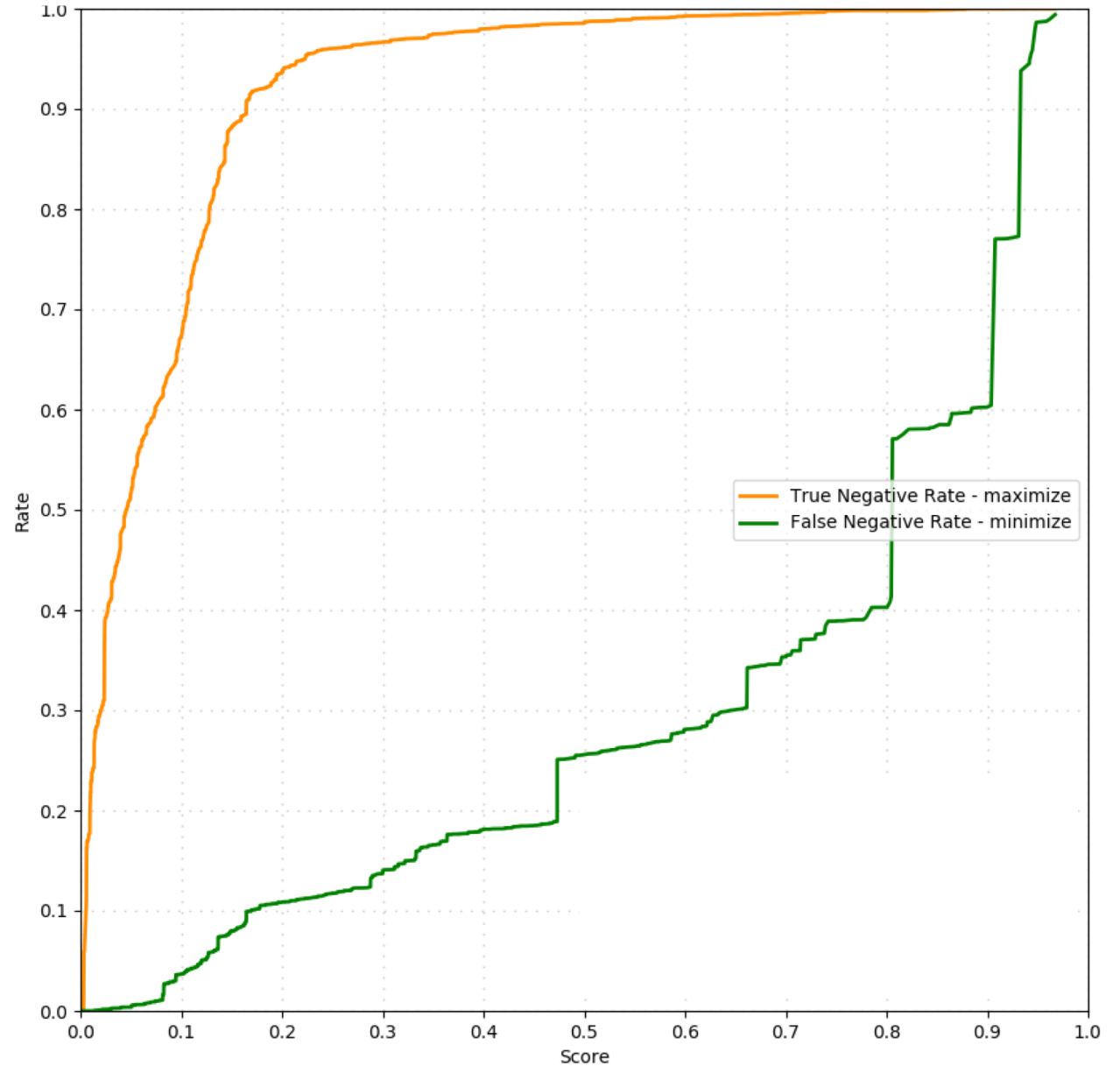
CHALLENGES WITH DATA #2

- Significant rule imbalance
 - a lot of rules trigger detections rarely, or not at all
- Different rules have different sets of features
- Assuming IID is questionable
 - if there is a detection from the same host/organization, it is likely that similar detections will occur again soon

CHALLENGES WITH DATA #3

- Detections come in a stream but...
 - ...they are not labelled in the same order as labels come after human processing
 - detections can also be grouped in **batches**, and then batches are labelled
- Evaluating performance of a classifier becomes non-trivial
 - data leakage is a real issue, with significant impact
- Software development challenges due to data sensitivity/volume.

**INITIAL
RESULTS:
RESULTS
ALREADY
LOOK QUITE
PROMISING**





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