RS∧Conference2018

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FIM AND SYSTEM CALL AUDITING AT SCALE IN A LARGE CONTAINER DEPLOYMENT



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Disclaimer





NOTE: All content discussed here are out of self learning and not related to the work I do at Walmart.

Ever increasing amount of logs





Overwhelming amount of choices

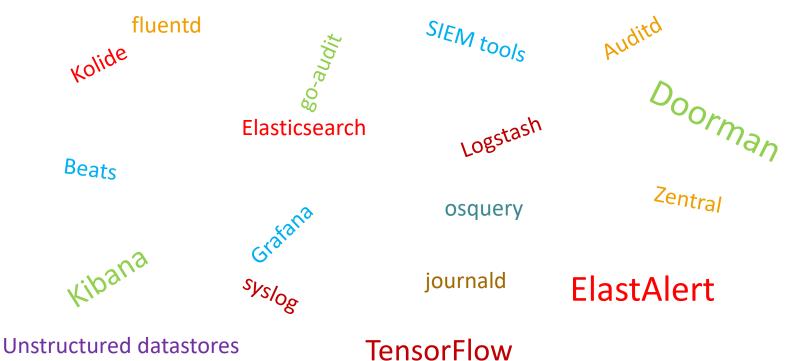




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Flood of OSS tools





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GOAL: Demystifying the choices we have



Understanding types of event sources

- Classifying event sources
- Understand event source type
- Evaluate open source stacks

Build our own stack based on insight needed

- Understand the insights we are looking for
- Build a stack based on the event classification
- If needed customize existing open source tools
- Build adaptors / tools that join the whole chain

Make an informed decision

 The stacks discussed in this presentation are by no means the only stack available

Quick poll





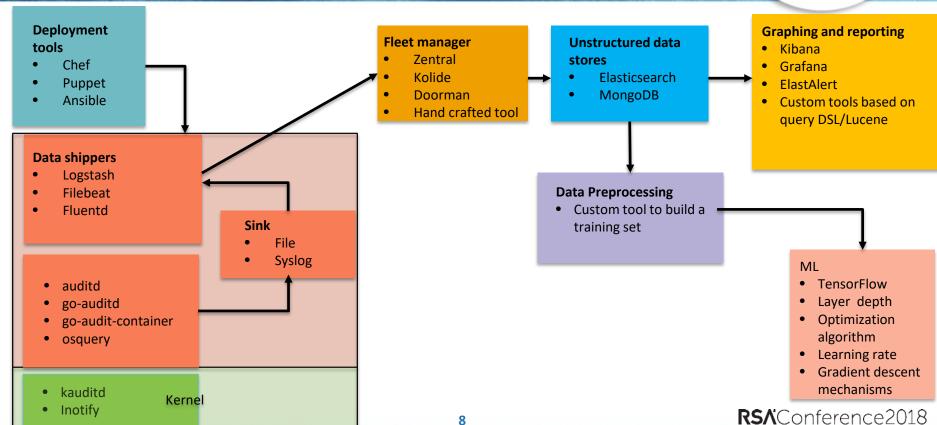
Use audit logging to detect anomalies?

Take it further to use machine learning techniques?

How may implement it only to meet compliance?

Possibilities of tools evaluated

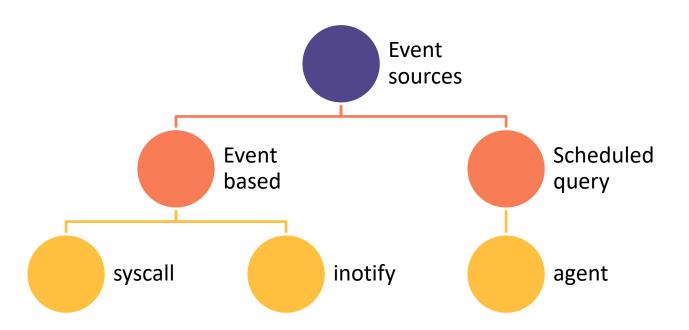




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Classification of event sources





Security insight based on event source type



syscall

Looking for specific outliers among mostly normal dynamic events.

- Like identifying outliers
- Monitoring constantly for a specific malicious system call along with other criteria (uid, etc)

inotify

Safe-guarding specific sensitive files / area in the file system

 Watch for CREATE/ACCESS/MODIFY/ DELETE events on specific files

agent

Scheduled activities for static information

 OS patch level queries, vulnerable kernel modules, misconfiguration



SYSTEM CALLS

Why syscall?



Fundamental transit points between user land and kernel

Every process makes system calls disclosing information of its activity

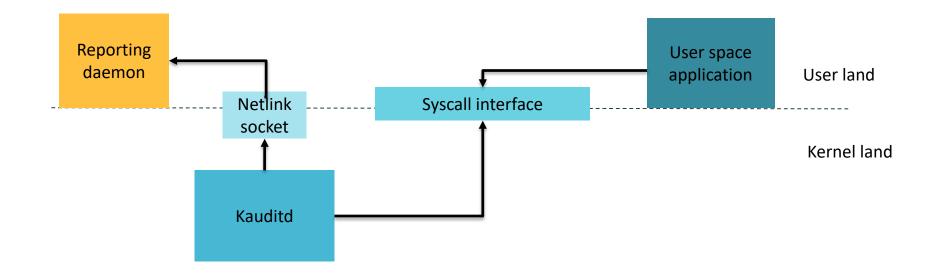
Several user space tools that send audit information (auditd, go-audit, go-audit-container)

Can provide deep insight when aggregated and drilled down

Ideal candidate to build a machine learning training set as the volume of data is huge

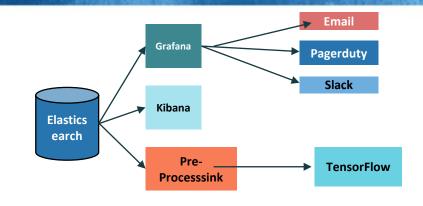
Audit component

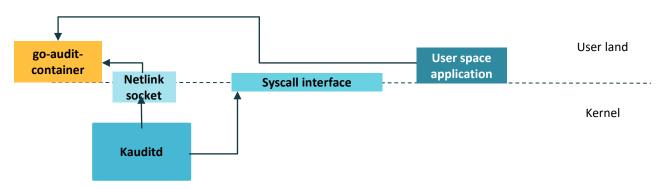




Audit log to gain insights at scale







alla ...

Demo



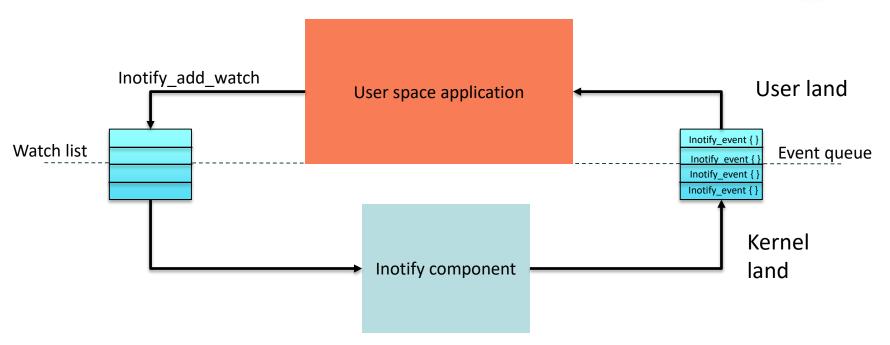




INOTIFY

Inotify component





Why inotify?





Lesser CPU consumption on an average

Another parallel stack, a new and exciting stack to explore with osquery

Missing details in the reports

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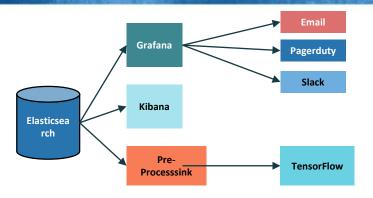
Demo





inotify based stack for FIM





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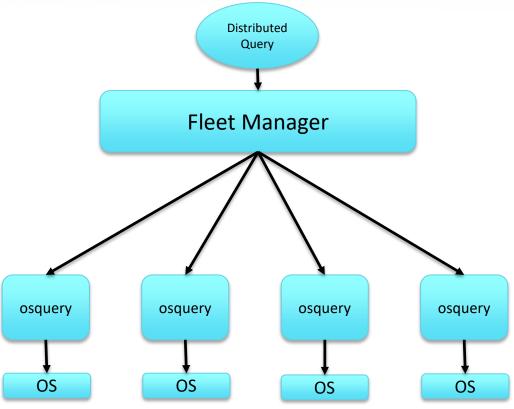




AGENTS

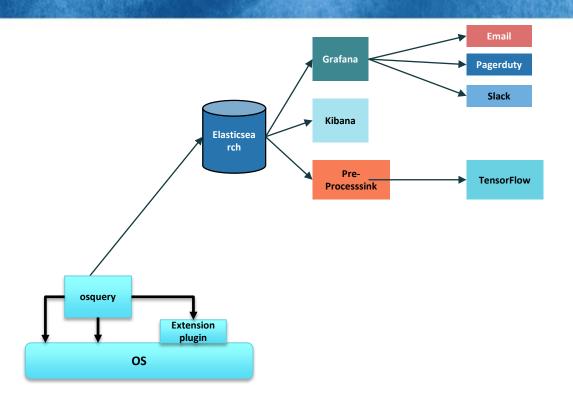
osquery





Osquery stack to get insights at scale





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Demo of tracing a 'Dirty COW' exploit





Learning from using fixed queries in Kibana, Grafana and custom tools



Fixed queries are good but only goes so far

Robust rules need a lot of queries

Any small variation of the rules is a false negative

Using DNN based machine learning helps improve our ability to detect anomalies



MACHINE LEARNING

High level differences



Unsupervised Supervised

- Elasticsearch ML Pre-processink-osquery
- Anomaly detection Explicit labelling and pre-processing
- Time series data Explicit data classification on disparate info

Elasticsearch ML: Detecting outliers





Picture credit: https://unsplash.com/@ripato

Demo of ElasticSearch ML





Use case: Classifying disparate data



Classifying data from different event sources

Broad classification into RED/YELLOW/GREEN

Classifying to have a big picture of the security posture of the organization

Supervised learning: Building a training set is key





Pre-processink-osquery stages



Stage 0

- Query one probe at a time from ES
- Label into RED/YELLOW/GREEN
- Write to stage_0.csv

Stage 1

Merge into existing stage_1.csv

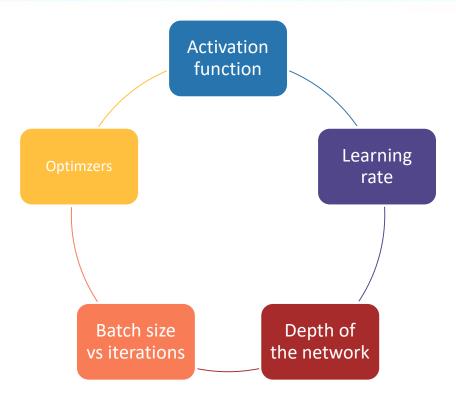
Manual Labeling At this stage human administrator can manually label events that were not labeled or which were incorrectly labeled by the automated rules

Stage 2

 Transform into numeric values, which will be the final training data set

ML choices if you are building your own solution





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Results of our experiment



ReLU for hidden layers and softmax for activation

Epoch vs Batch Size vs Iterations

Adam optimizer

Lower the learning rate the better

Lessons learnt



With system call auditing, inotify, os level querying agents as foundations, combined with ability to aggregate at scale in Elasticsearch, we can achieve very deep security insights on the production environment.

With anomaly detection discussed in this talk, we are just scratching the surface. Given that system calls are fundamental, possibilities are enormous.

With static threshold and reporting configuration it is very easy to miss security insights

DNN based machine learning helps in getting intelligent insights

Based on parameters like platform support, CPU utilization, memory and disk space foot print, etc we looked at different choices of stack

Apply



Start with a simple File Integrity monitoring implementation using audit log. Observe load of FIM events on the infrastructure

Grow the solution to more detailed monitoring

Try applying ML based on the rule based and manual labels

Think of possibilities outside of what is discussed here today

AuditNG suite



https://github.com/auditNG/preprocessink-osquery

https://github.com/auditNG/go-audit-container

Questions?





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