



Measuring What Matters to Streamline Security Operations with Splunk

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A Few Words About Me

- Detection Engineer / Site Reliability Engineer @ Red Canary
- Red Canary Does Managed Endpoint Detection & Response (EDR)
- Tries to make useful content from logs and metrics
- Messes with Code; Interested in Data Science; Formally Trained in Neither
- Is Convinced You Can Use Splunk to Answer Any Questions if Creative Enough (and you actually have the data...)

A Brief Word On Common Terminology

- Detector / Signature / Alert – The Logic You Use To Detect Behavior

ruby_code

```
not_trusted?  
&&  
  
(process_path_downcased.include?('appdata')  
||  
  
process_path_downcased.include?('application data'))  
&&  
  
childproc_name_downcased == 'sh.exe'
```

- We focus mostly on Endpoint Data (Process Starts, Filemods etc.)

Noise – It's A Common Problem

- Humans Can Only Do So Much
- Repetitive Tasks Cause Fatigue and Mistakes
- Low Fidelity High Volume Alerts Contribute
- Alerts that provide no value and take a long time – Worst
- Short term, high value aka (“evil”) – Best
- Reality - much more in the middle

Lack of Data & Context:

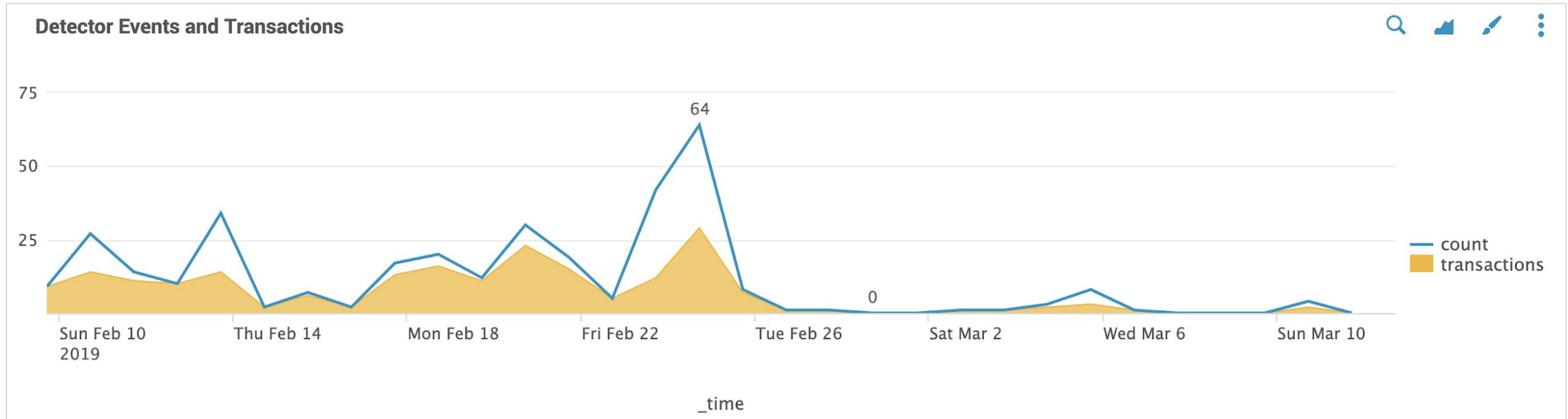
- Point in Time View of Alerts
 - Csv
 - Dashboard
- We tracked how effective a detector was, but only as a point in time
- No way to drill down into
 - Which host
 - Which process
 - FP vs TP characteristics

Expected Impact from 2017-12-15				Reset
Detection Classification	Conversion Rate	Classification Impact	Expected Impact	
Malicious Software	5.556%	15.0	0.833	
Malicious Software / Credential Theft	2.381%		0.000	
Malicious Software / Hacking Tool	3.175%		0.000	
Suspicious Activity	7.143%	10.0	0.714	
Suspicious Activity / Network	6.349%		0.000	
Suspicious Activity / Process	7.143%		0.000	
Total:			1.548	

No concept of effort and time, no counts

Data Was Needed:

- To Provide the ability to do Historical Analysis
- To Identify impact from changes



Content Overview:

This is a short talk, as such instead of focusing on the entire project will be focusing on small pieces of putting this together that were challenging and might help someone else.

- Summarizing Data
- Storing The Calculated Data
- Example Analytics on the Data

High Level Questions

Where is my team spending the most effort?

- Am I spending more time on detected events or false positives?
- Of the False Positives – Which one's take the most effort and time?
- Which set of my high fidelity signatures are close enough to perfect that can maybe automate them away?
- Are spikes skewing my analysis, methods to normalize data? (Why counts and averages can lie)

Per Alert Questions:

Which of my individual rules are terrible?

- How much time did I spend on Rule 1077 last month?
- How many of Rule 1077's events converted to actual incidents?

timestamp	detector_id	event_count	detected_count	not_detected_count	human_transaction_time	human_transaction_time_detected
2019-03-23T12:00Z	790	965	0	965	3758.59343771	0
2019-03-22T12:00Z	790	8087	0	8087	99078.743190333	0
2019-03-19T12:00Z	790	106	13	93	8712.67749321	2202.1169041
2019-03-16T12:00Z	790	33	0	33	953.83439033	0
2019-03-20T12:00Z	790	45	7	38	3933.0879424	1835.7224285
2019-03-15T12:00Z	790	54	1	53	1501.61993620	27.924000
2019-03-21T12:00Z	790	76	9	67	6731.11095707	2486.6014997
2019-03-24T12:00Z	790	2633	1	2632	10297.28082905	145.256000
2019-03-18T12:00Z	790	40	4	36	2847.63147319	287.42864019
2019-03-17T12:00Z	790	6	0	6	608.173500	0

Reducing Data

- Logs are voluminous but revealing
- Metrics are essentially just numbers
 - Great for trends
 - Bad for context
- Data reduction expedites searches and queries and helps with retention

What data do you want to keep?

- You may want to keep data not strictly needed for metrics
- What questions might you have after identifying a seemingly problematic rule or group of rules
- Metrics are essentially just numbers
 - Severity of incident related to alert
 - Environments that tripped alert
 - Hosts online and active during the timeframe pulled
 - Unique incidents per rule or day

Our Metrics

- True positive/false positive percentages per detector
- Time expenditure on transactions by detection rule
 - Confirmed detected events
 - Confirmed false positive events
- Conversion percentage per detection rule

Note: *Transaction= Grouping of events on a given endpoint at a point in time. We record the action taken on that grouping.*

Summarizing Data



Example: Calculating Time Metric

- For each event calculate the duration that it was worked:

```
| eval t_claim=strptime(claimed_at, "%Y-%m-%dT%H:%M:%S.%3NZ")
| eval t_term=strptime(terminal_at, "%Y-%m-%dT%H:%M:%S.%3NZ")
| eval t_duration=t_term-t_claim |
```

- For each event add a field, so we can separate humans from automation
 - Analyst: analyst10@redcanary.com = human
 - Analyst: bot50@redcanary = bot

```
| eval freewill=if(like(analyst, "%noreply@redcanary%"), "bot", "human") |
```

Example: Creating a Daily Sum

- Decide the smallest unit you might want to chart – Hours? Days?

```
| eval datestamp=(_time)
| convert timeformat="%Y-%m-%dT12:00Z" ctime(datestamp)
| eventstats count as total_events_all_detectors by datestamp
```

- Sum time spent for each signature

```
| eventstats sum(eval(if(freewill="human", fractional_duration, 0))) AS human_transaction_time by detector_id,datestamp
```

- We will end up with one row per signature per day

datestamp ^	detector_id	total_time	human_per_signature	total_true_positive_human
2019-08-27T12:00Z	1095	216201.301743205	1846.177000	107526.54995846
2019-08-27T12:00Z	1089	216201.301743205	17.465000	107526.54995846

Answers: How much time did I spend on Rule XX last month?

Search interface showing a query and results.

Search Query: `index=cirt_summary_index | stats sum(human_transaction_time) AS total_time by detector_id | eval time_in_hours=(total_time/3600) | table detector_id time_in_hours`

Results: 6,966 events (8/5/19 12:00:00.000 AM to 9/4/19 1:12:42.000 PM) | No Event Sampling

Visualization: Statistics (522)

Table Headers: detector_id, time_in_hours

detector_id	time_in_hours
1938	2.98079165294
1939	0.0251108333
1943	7.29522484461
1944	0.87152268519
1945	1.63469717476
1951	0.04423388889
1952	0.1830170835
1956	2.79656737100

You'll repeat this for as many data points as you care about - essentially summarize or enrich the things you want to keep and ask questions on.

```
human_hidden_transactions by detector_id, timestamp  
| eventstats dc(eval(if(freewill="human" AND terminal_stage="detected", transaction_id, NULL))) AS  
human_detected_transactions, dc(eval(if(freewill="human" AND terminal_stage="detected" AND  
'associated_detections{}.classification'="Malicious Software", transaction_id, NULL))) AS  
human_malicious_transactions, dc(eval(if(freewill="human" AND terminal_stage="detected" AND  
'associated_detections{}.classification'="Suspicious Activity", transaction_id, NULL))) AS  
human_suspicious_transactions, dc(eval(if(freewill="human" AND terminal_stage="detected" AND  
'associated_detections{}.classification'="Unwanted Software", transaction_id, NULL))) AS  
human_unwanted_transactions by detector_id, timestamp
```

Storing the Calculated Data



How do I keep data past normal retention?

- I can run the previous calculations every time I want to ask a question.
 - It will take a really long time each time I want to change something
 - I will only be able to run across maybe 30 days, data is constantly rolling off
- Simple Answer - Use a lookup table:

```
human_transaction_time_malicious, human_transaction_time_suspicious, human_transaction_time_unwanted, unique_detections_suspicious, unique_detections_unwanted, unique_endpoints, unique_endpoints_malicious, unique_endpoints_suspicious, unique_endpoints_unwanted  
| dedup timestamp, detector_id | outputlookup append=t transaction_aggregate.csv
```

Lookup Limitations:

- Row Limitations
- Time in Dashboards Can Be “Difficult”

```

<panel>
  <table>
    <search>
      <query>| inputlookup event_aggregate.csv
| eval _time=strptime(_time,"%Y-%m-%dT%H:%MZ")
| sort - _time
| addinfo
| where _time>=info_min_time AND (_time<=info_max_time OR info_max_time="+Infinity")
| eventstats sum(daily_human_detected_count) as human_total_detected sum(eval(daily_human_det

```

Lookup Table: Alerts Work Well

- Automate Collection of Data – no one will remember to run a search daily, monthly (or hourly...)
- List in triggered alerts is a good option

detector_transaction_infov3
daily counts starting 3/24

Enabled: Yes. [Disable](#)
 App: search
 Permissions: Shared in App. Owned by keshia.levan@redcanary.com. [Edit](#)
 Modified: Mar 25, 2019 5:44:55 PM
 Alert Type: Scheduled. Daily, at 23:00. [Edit](#)

Trigger Condition: Number of Results is > 0. [Edit](#)
 Actions: [1 Action](#) [Edit](#)
[Add to Triggered Alerts](#)

Trigger History

20 per page ▾

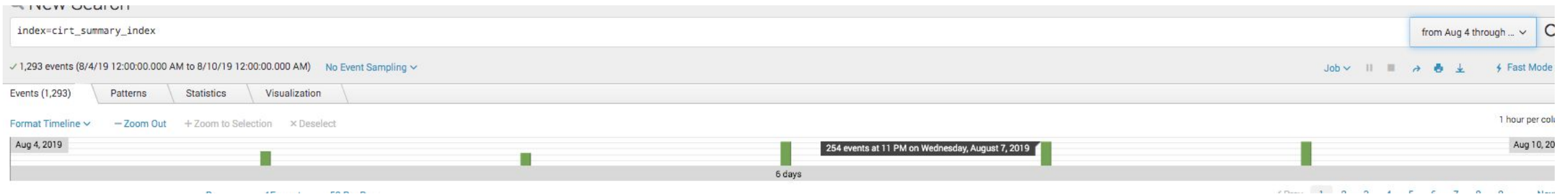
	TriggerTime	Actions
1	2019-09-03 23:00:36 GMT	View Results

- Backup your data – super easy to blow everything away with one bad command
- Monthly alert that sends CSV to e-mail was our hack at out of Splunk Storage

Another Option: Summary Index

```
human_transaction_time_malicious, human_transaction_time_suspicious, human_transaction_time_unwanted, human_transaction_time_not_detected,  
unique_detections_suspicious, unique_detections_unwanted, unique_endpoints, unique_subdomains  
| dedup timestamp, detector_id | addinfo | collect index=cirt_summary_index addtime=false marker="report_name=\"transaction_data_summary\""
```

- Simple command took a good amount of time to figure
- You can search this index almost like normal

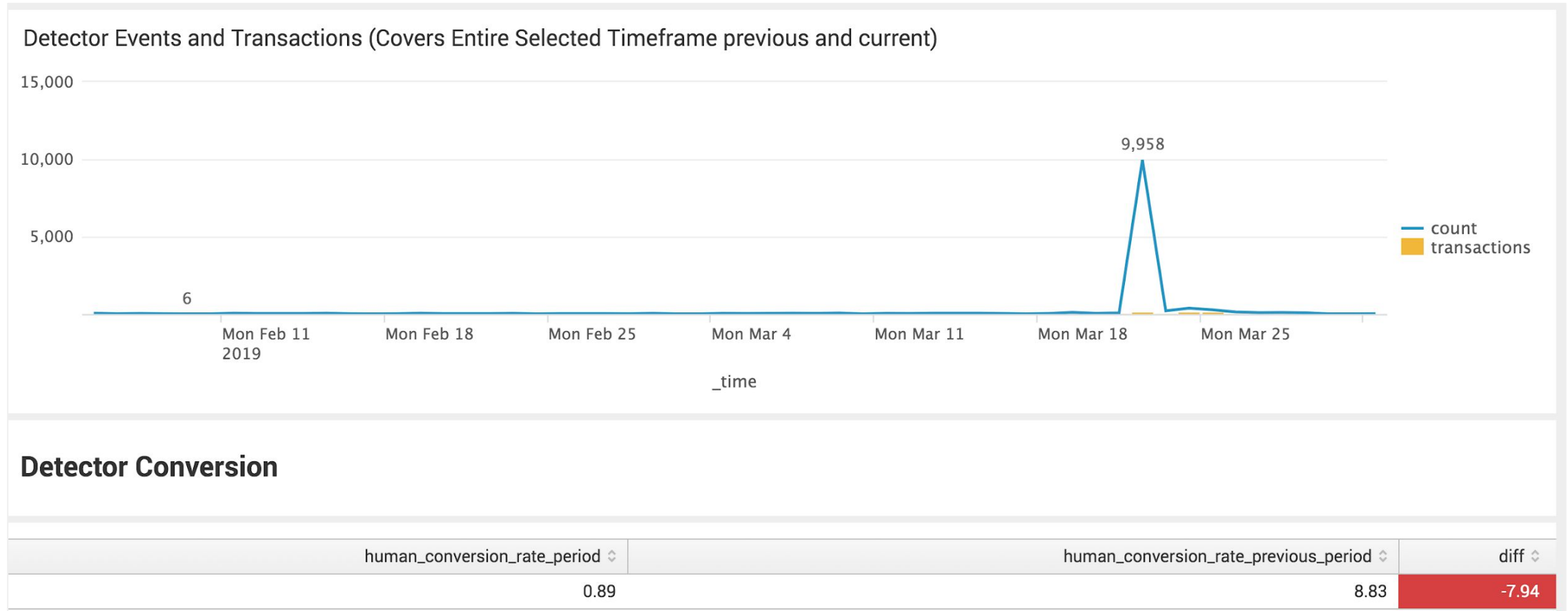


Example Analytics on the Data

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splunk>

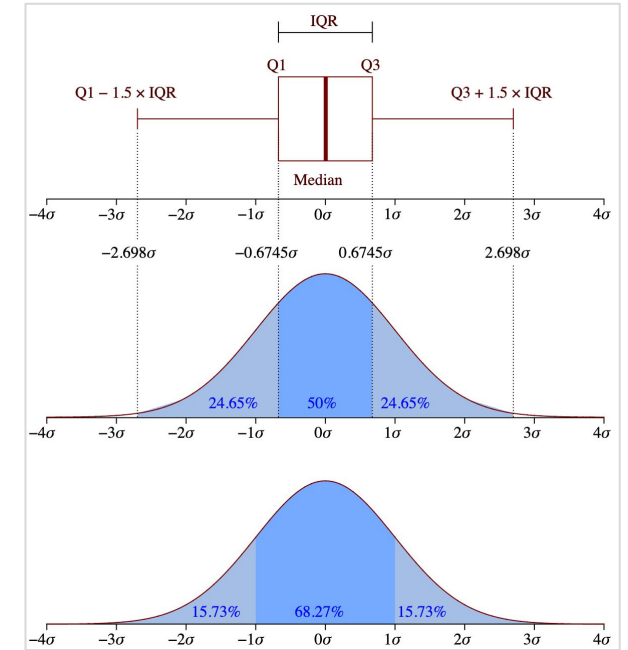


Outliers: Are Spikes skewing my analysis?



Calculating IQR

- Splunk did all the hard work here
- <https://docs.splunk.com/Documentation/Splunk/7.3.1/Search/Findingandremovingoutliers>



```
eventstats p25(human_transaction_time_not_detected) as not_detected_time_p25 p75(human_transaction_time_not_detected)
as not_detected_time_p75 by detector_id

eval not_detected_time_IQR=(not_detected_time_p75-not_detected_time_p25)

eval not_detected_time_upperBound=(not_detected_time_p25+1.5*not_detected_time_IQR)

eval not_detected_time_isOutlier=if(human_transaction_time_not_detected > not_detected_time_upperBound, 1, 0)

eventstats sum(eval(if(not_detected_time_isOutlier="0", human_transaction_time_not_detected, 0))) as sum_transaction_time_not_detected by detector_id

eval adj_transaction_time_not_detected_hrs=round(sum_transaction_time_not_detected/3600,2)
```

Data Is Now More Representative of Signature

- Average is 44, versus 800
- Help us stack signatures for comparison

detector_id	adjusted_not_detected_count_avg	not_detected_count_avg	human_not_detected_count	not_detected_count_p25	not_detected_count_p75	not_detected_count_isOutlier	remove_outliers
790		803.7142857142857	213	35	183	1	
790		803.7142857142857	9957	35	183	1	
790	44.9	803.7142857142857	67	35	183	0	NotOutlier
790	44.9	803.7142857142857	36	35	183	0	NotOutlier
790	44.9	803.7142857142857	93	35	183	0	NotOutlier
790	44.9	803.7142857142857	35	35	183	0	NotOutlier
790	44.9	803.7142857142857	6	35	183	0	NotOutlier
790	44.9	803.7142857142857	33	35	183	0	NotOutlier
790	44.9	803.7142857142857	53	35	183	0	NotOutlier
790	44.9	803.7142857142857	48	35	183	0	NotOutlier
790	44.9	803.7142857142857	54	35	183	0	NotOutlier
790	44.9	803.7142857142857	24	35	183	0	NotOutlier

Using Dataset: How many of Rule XX's events converted to actual incidents?

Last 7 days		days_detector_seen		only_active		Submit		Hide Filters	
		☒ all_frequency		☐ active (20180830)					
		☐ exclude_single_days		☒ all					
	detector_name ↕	detector_id ↕	human_percentage_detected ↕	daily_avg_not_detected_events ↕	daily_avg_detected_events ↕	daily_avg_not_detected_claimed_time ↕	daily_avg_detected_claimed_time ↕	days_detector_seen ↕	
1	WIN-SUSPECT-SVCHOST-EXECUTED	18	76.58	9	28	0.24 Hrs	1.81 Hrs	6	
2	WIN-RECYCLED-BINARY-EXEC	22	0.00	5	0	0.16 Hrs	0.00 Hrs	4	
3	WIN-SYSVOL-BINARY-EXEC	23	100.00	0	1	0.00 Hrs	0.19 Hrs	1	
4	WIN-MULTIEXT-FILE-EXEC	26	64.29	1	2	0.01 Hrs	0.83 Hrs	4	

```
| eval
```

```
human_conversion=round(human_total_detected/human_total_detector_events*100,2)
```

Am I spending more time on detected events or false positives?

- For this time period 50% of human time was spent on things that convert less than 10% of the time.
- Maybe look there to improve that signature set first.
- We can drill into which specific one's they are with the data we saved.

Detectors Seen		Total Human Transactions
522		33,83
Adjusted Overview by Conversion (these would be nice as		
human_conversion ▾	num_detectors ▾	all_detectors_time_hrs ▾
0-10	350	321.12
10-20	28	62.80
20-30	19	160.60
30-40	14	16.17
40-50	10	7.26
50-60	12	11.41
60-70	12	23.04
70-80	7	2.52
80-90	9	10.65
90-100	11	8.91
100-110	50	6.37
	522	630.85

You'll Still Need Logs

- The previous dataset is meant to show a starting place or a trend.
- There's a tradeoff between metrics and logs.
- In order to adequately tune a signature you need to know:
 - What specific things are the signature hitting on we don't care about?
 - Is there anything specific I can add or remove that will cause it to be more targeted?
 - Is it even looking for the correct behavior?
 - Log data may not go back as long, but it will have more detail.

Conclusion:

- Summary indexes and lookup tables are good tools for summarizing large data sets that are difficult to search and retain
- Lessons Learned – building out the backend search is important, probably content to use it is more so and ensuring the team has the right tools
- There's way more than one way to do all of this, you tend to start with what you know.