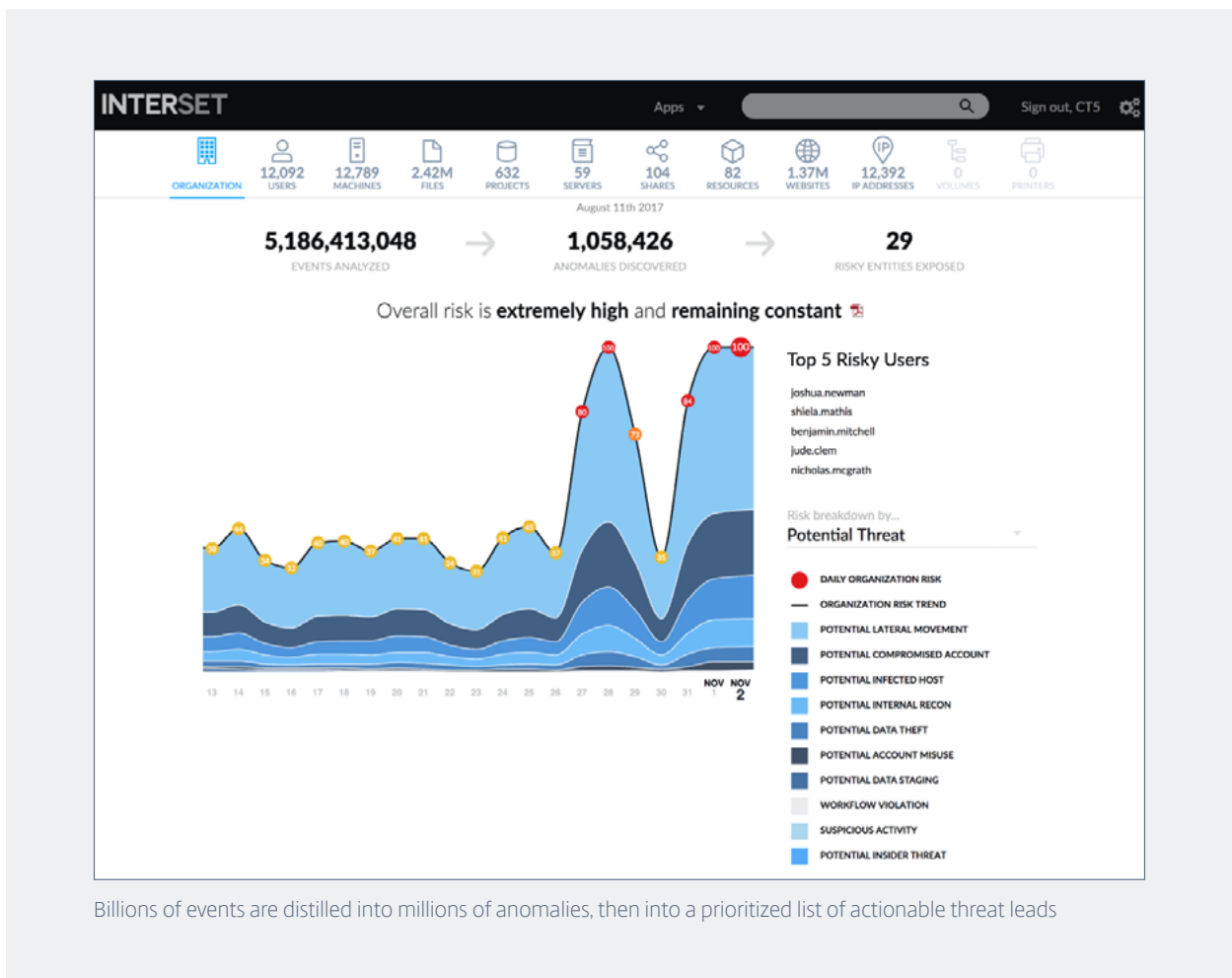
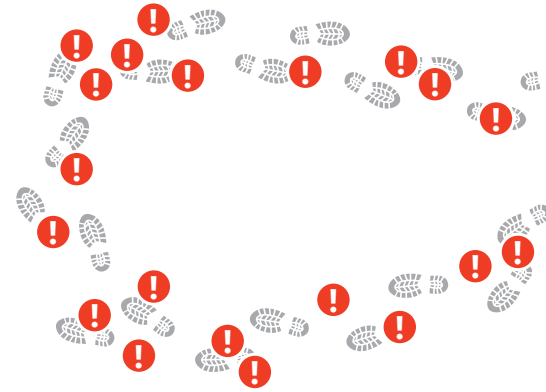


INTERSET

Interaset at a Glance

Security analytics give cyber hunters radical advantages in containing cyberthreats. Powered by unsupervised machine learning, Interaset distills billions of real-time events into a prioritized list of high-risk entities for accelerated, more accurate threat-hunting.



Billions of events are distilled into millions of anomalies, then into a prioritized list of actionable threat leads

An enterprise executive can download reports that summarize important organizational risks.

Top 27 2017

The Internet Platform

The Internet Platform is an analytics product that surfaces abnormal behaviors from Behavioral Data. It uses advanced analytics to score Behaviors and Entities between one and one hundred from multiple Behavioral Analytics.

The Higher the Score, the Higher the Risk.

The Threat Hunter Interface

The Threat Hunter interface observes all Behaviors or Alerts and aggregates them into Risky Hours which is presented visually over lengths of time.

If an Entity (like users or projects) are selected, the report will only contain information that applies to the organization on a whole. This includes information like Top Risky Users, Top Risky Projects, Top Risky Users (by accounts identified in logs or scripts) and other organizational information.

If Entities are selected, a section will be created for each which enables for the addition or removal of pages as needed.

This report can be generated as often as needed with different selections, filters and time ranges. The generated PDF can then be saved, distributed and managed accordingly.

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Work Schedule

All Users, All Time

Working days and working hours represent the average of all users in the organization, for all time.

Working Days

This visualization represents the overall level of activity on individual days of the week, comparing each minute of an individual day, for all users, against the most active day for that same minute of the day, across all time.

Table columns indicate more activity, and therefore the most normal work days. Shorter columns indicate less activity, and therefore less normal work days.

Use this information to examine the baseline normal work day activity across the organization.

This report excludes data generated by bots and bot-like users.

Working Hours

This visualization represents the overall level of activity at different times of the day, comparing each minute of an individual day, for all users, against the usual activity for that same minute of the day, across all time.

The clock that is centered to the rim of the clock represent the most user activity, and therefore the most normal work hours. The smallest dots, those closest to the center of the clock, represent the least user activity, and therefore the least normal working hours.

Use this information to examine the baseline normal working hour activity across the organization.

This report excludes data generated by bots and bot-like users.

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Most Accessed Servers

Oct 3, 2016 to Nov 2, 2016

This table lists the most accessed servers between Oct 3, 2016 to Nov 2, 2016.

Use this information to examine the related servers most frequently accessed.

This report excludes data generated by bots and bot-like users.

#	Address	Server
1	134.43	MSWVS01Internet-dev
2	84.70	MSWVS02Internet-dev
3	81.75	VMS-C200Internet-dev
4	80.94	OSMS-B01Internet-dev
5	79.87	MSWVS03Internet-dev
6	72.94	TK-CD01Internet-dev
7	71.27	MSWVS04Internet-dev
8	70.43	PRM-Internet.com
9	67.27	FTDC-Internet.com
10	67.41	OTWAWAC-Internet.com
11	67.01	MSWVS05Internet-dev
12	65.27	OSMS-Internet.com
13	64.54	MSWVS06Internet-dev
14	63.93	LMWVS01Internet-dev
15	63.28	OSMS-Internet.com
16	60.78	CINCO-SBVS01Internet-dev
17	59.57	SLC-Internet.com
18	58.63	CONTR-SBVS01Internet-dev
19	55.32	INCOV-Internet.com
20	54.77	SOA-Internet.com
21	51.68	SRAND-Internet.com
22	45.70	FLARE-Internet.com

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Bot-Like Users

Oct 3, 2016 to Nov 2, 2016

This table lists individual user accounts within the organization whose activity may be indicative of bots, automated or semi-automated tools that carry out simple and structurally repetitive tasks.

Use this information to examine the bot-like users to ascertain whether they are indeed bots, or users performing automatic activity, as well as to determine whether bot users are desirable or undesirable.

#	User	Score
1	wankling	67%
2	indubitableabroad	64%
3	indubitableabroad	62%
4	obnoxiousring	62%
5	intimidated	59%
6	zoogetfish	58%
7	hubbubban	58%
8	inexpensive	58%
9	joakaban	58%
10	teagles	58%
11	julia-rothford	58%
12	whisperer	58%
13	bramblesmocha	58%
14	alxander-lambert	58%
15	berthelwell	58%
16	pasquale	55%
17	kaushika	55%
18	unsubscribed	55%
19	karandad	55%
20	karandad	54%
21	johanneswaga	54%
22	venuesandwood	54%
23	berndschulze	54%
24	malcolmscott	54%
25	congregare	54%
26	alxander	54%
27	andreaswells	54%
28	graceanderson	54%
29	bramblesage	51%
30	karandadwell	51%

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Organizational Information

This section contains information that applies to the entire organization.

Risky Hours

Oct 3, 2016 to Nov 2, 2016

This matrix visualization displays all 136,447 risky hours, highlighting anomalous activities between Oct 3, 2016 to Nov 2, 2016.

Use this information to examine the organization's overall user activity risk potential.

This report excludes data generated by bots and bot-like users.

13,444 Total Risky Hours | 2 Extreme Risk | 20 High Risk | 546 Moderate Risk | 137,474 Low Risk.

Top 10 Risky Hours

This table represents the overall top ten (10) risky hours for the organization during the selected time period. These 10 risky hours will have the highest overall risk score based on anomalous activity or violations throughout the organization.

Use this information to examine the organization's overall user activity risk potential.

This report excludes data generated by bots and bot-like users.

#	Score	Owner	Date	Time (04:00)
1	92	johanneswells	Nov 2, 2016	3am-3am
2	92	johanneswells	Nov 2, 2016	3am-3am
3	66	vulvaldson	Oct 26, 2016	3am-4am
4	66	grahnd	Oct 26, 2016	3am-3am
5	66	whitford	Oct 1, 2016	13pm-3pm
6	66	joakaban	Oct 7, 2016	3pm-3am
7	66	vanessalambert	Oct 11, 2016	10am-11am
8	66	vanessalambert	Oct 10, 2016	3pm-3am
9	66	nicholaswells	Oct 13, 2016	4pm-5pm
10	66	hubbubban	Oct 13, 2016	5pm-6pm

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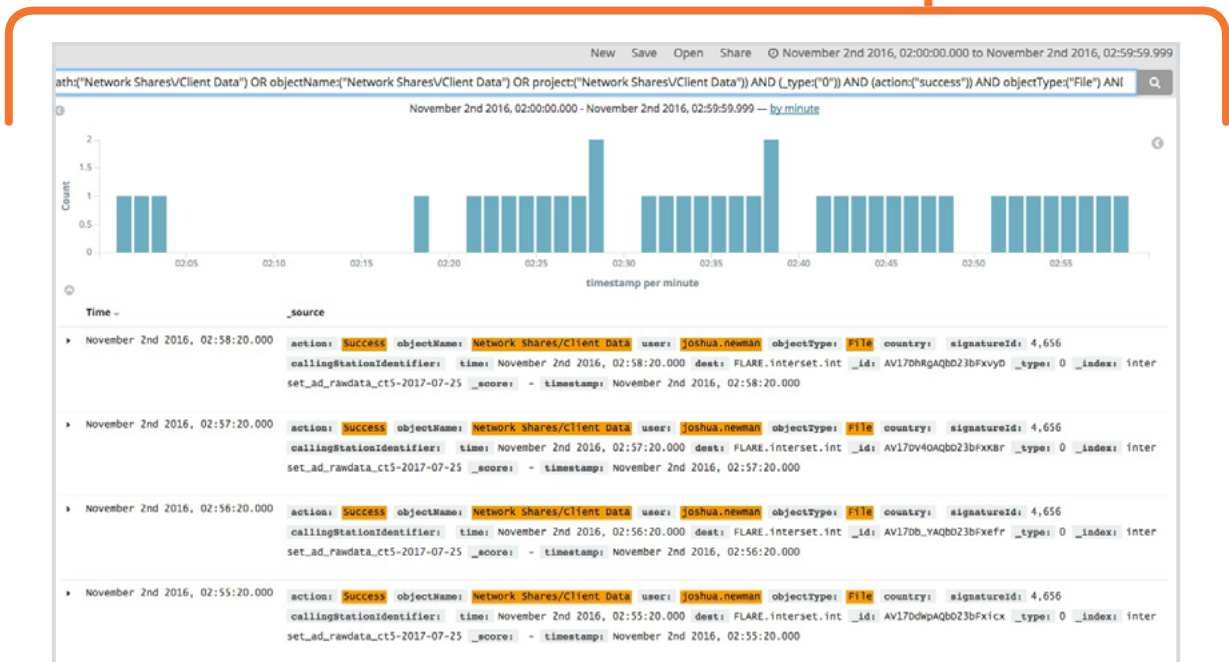
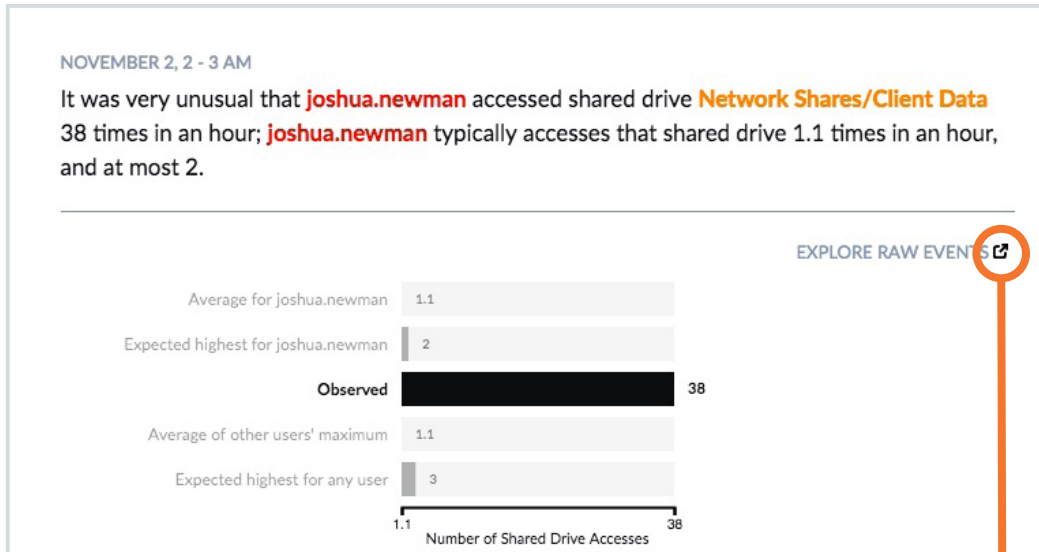
Top Risky Hours 1/10

Score: 92

Date	Time Range	User
November 2, 2016	3am - 3am	johanneswells

1. johanneswells performed this hour, which was every unusual for that past activity.
2. It was very unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
3. It was very unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
4. It was very unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
5. johanneswells attempted to access the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
6. johanneswells attempted to access the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
7. It was very unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
8. It was very unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
9. It was slightly unusual that johanneswells accessed the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically accessed the redshift in Network Data's Client Data 3P1 times in an hour, and it was not unusual.
10. johanneswells took slightly more than the normal performing time to access the redshift in Network Data's Client Data 3P1 times in an hour. johanneswells typically performed this action in 30 seconds.

The security team can access details about prioritized, validated threats through the risk dashboard. This relieves them of a lengthy discovery process, by communicating relevant data in just a few clicks.



Intersec's security analytics leverage data from multiple data sources, for the most comprehensive view of enterprise risk and contextual threat detection across these datatypes.

From these data sources, unsupervised machine learning evaluates the risks of normal digital footprints—of users, machines, files, projects, servers, shares, resources, websites, and IP addresses—to spot abnormalities.

Account Misuse	Compromised Account	Data Staging/Theft	Infected Host	Internal Recon	Lateral movement	Fraud
Authentication Logs	Authentication Logs	Endpoint Logs	Web Proxy Logs	Authentication Logs	Authentication Logs	Expense
Directory Service Logs	Directory Service Logs	Directory Service Logs	Directory Service Logs	IP Repository Logs	IP Repository Logs	
Endpoint Logs	Operating System Logs	IP Repository Logs	Endpoint Logs	File Share Logs	File Share Logs	
Operating System Logs	File Share Logs	Printer Logs	NetFlow Logs	Operating System Logs	Operating System Logs	
File Share Logs	VPN Logs	Web Proxy Logs		Resource Access Logs	Resource Access Logs	
VPN Logs	Resource Access Logs	Printer Logs		Endpoint Logs	Endpoint Logs	
Resource Logs	IP Repository Logs	NetFlow Logs		NetFlow Logs	NetFlow Logs	
IP Repository Logs		Email Logs				
Printer Logs						

Current data sources processed by Intersec

For a full list of features, please contact Intersec at sales@intersec.com.

INTERSEC

Contact sales@intersec.com

Intersec, an AI security analytics company, empowers security teams to identify and respond to the threats that matter before data is stolen. Intersec's self-learning threat detection platform leverages AI and machine learning to measure the unique digital footprint of systems and users, distilling billions of events into a handful of prioritized threat leads. What used to take months, can now take minutes. Intersec is backed by In-Q-Tel and trusted to protect critical data in finance, critical infrastructure, high-tech manufacturing, healthcare, utility and energy industries. Visit us at intersec.ai, and follow us on [Twitter](#), [LinkedIn](#), and [Facebook](#)