

User and Entity Behavior Analytics

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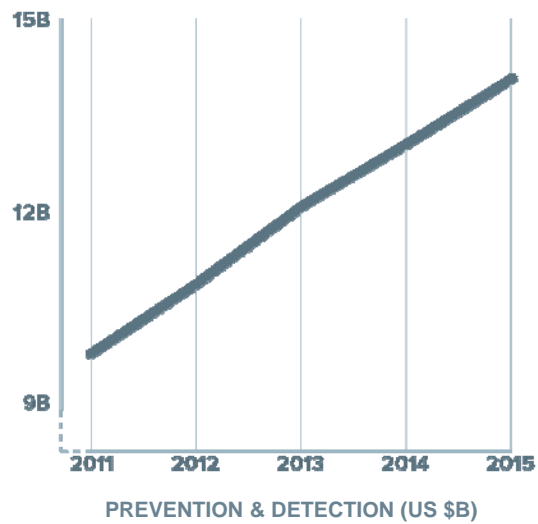
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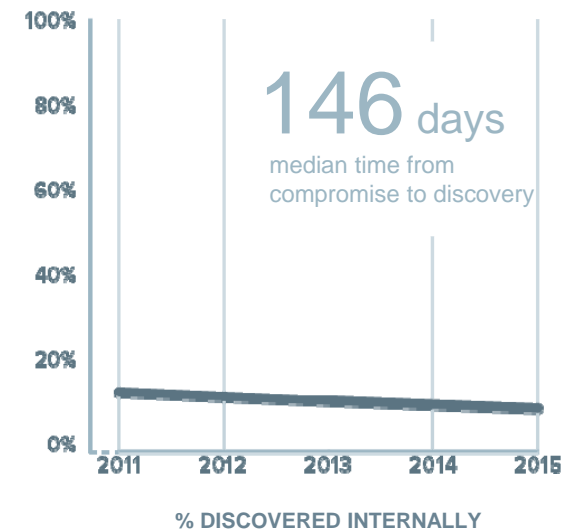
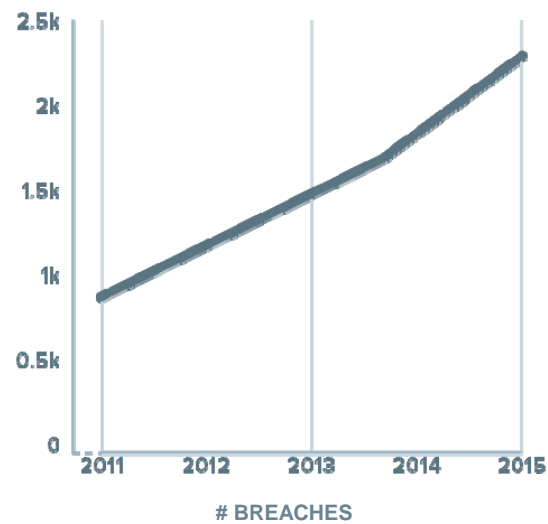
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THE SECURITY GAP

SECURITY SPEND



DATA BREACHES

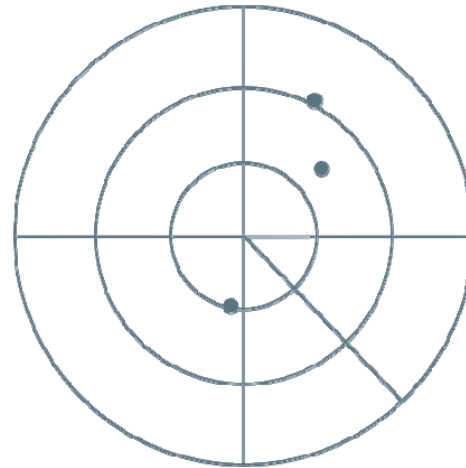


SOURCES | Mandiant M-Trends 2016, Verizon Data Breach Investigations 2016, IDC 2016

THE PROBLEM



**PREVENTION & DETECTION
NOT ENOUGH**
INCREASINGLY POROUS



**MONITORING SYSTEMS
FALLING SHORT**
CANNOT DETECT UNKNOWN THREATS

Attacks involving legitimate credentials



COMPROMISED

40 million credit cards were stolen
from Target's servers

STOLEN CREDENTIALS



MALICIOUS

Edward Snowden stole more than 1.7 million
classified documents

INTENDED TO LEAK INFORMATION



NEGLIGENT

Employees uploading sensitive information to
personal Dropbox for easy access

DATA LEAKAGE

Behavioral Analytics



AUTOMATED DETECTION
of threats inside the organization



FORCE MULTIPLIER
for security analysts

Basics of Behavioral Analytics

MACHINE LEARNING

UNSUPERVISED

+

SEMI-SUPERVISED

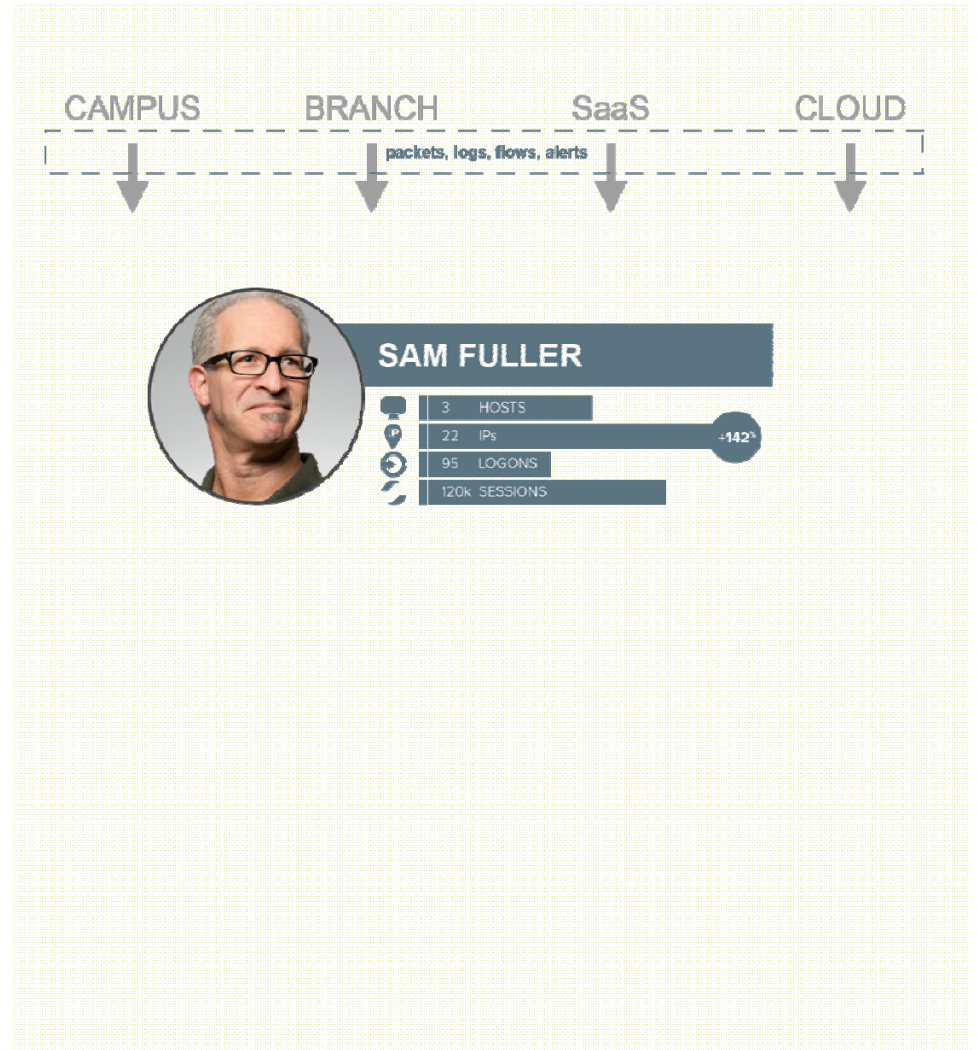


BASELINES

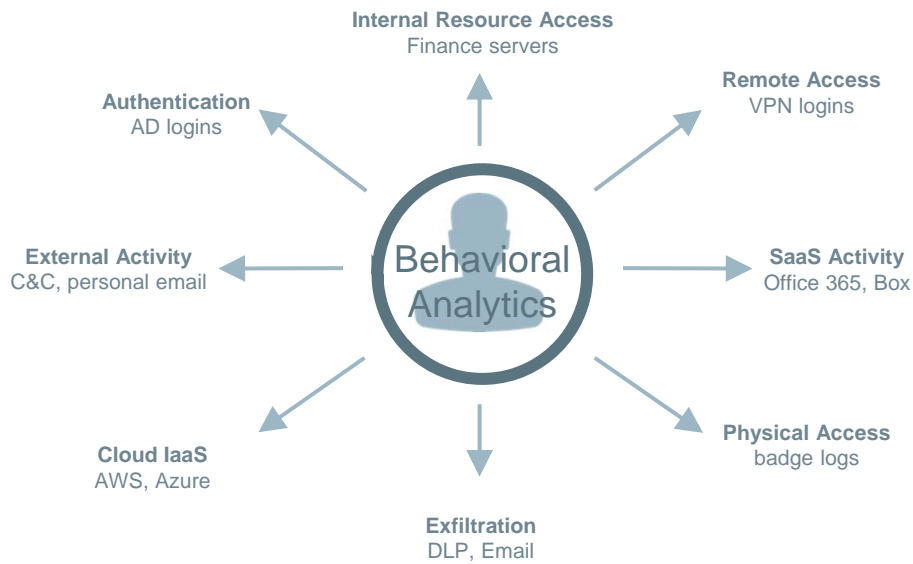
HISTORICAL

+

PEER GROUP



Behavior – Many different dimensions



SAM FULLER

3	HOSTS	
22	IPs	-142h
95	LOGONS	
120k	SESSIONS	

DETECTING AN ANOMALY

Internal Resource Access
Finance servers



SAM FULLER

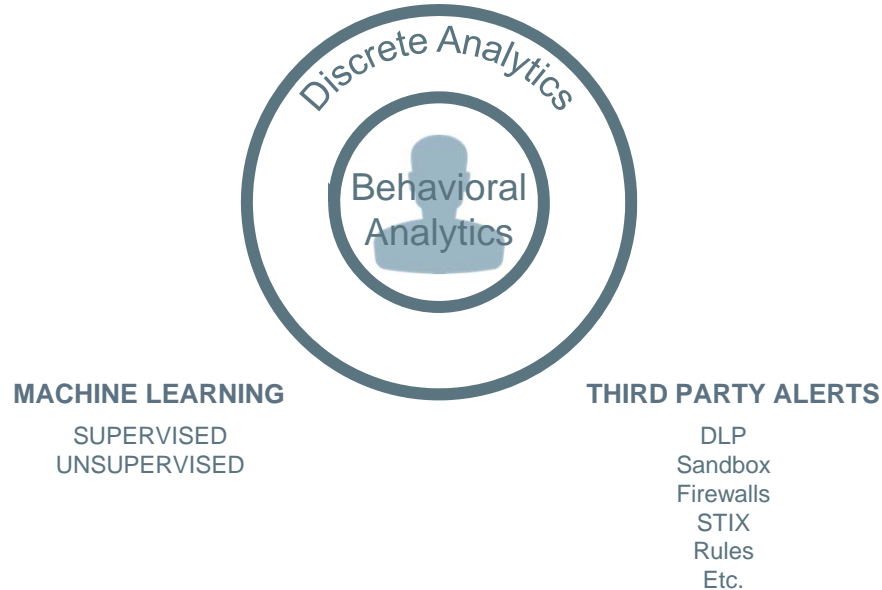
3	HOSTS
22	IPs
95	LOGONS
120k	SESSIONS

-142h



ABNORMAL APPLICATION
ACCESS

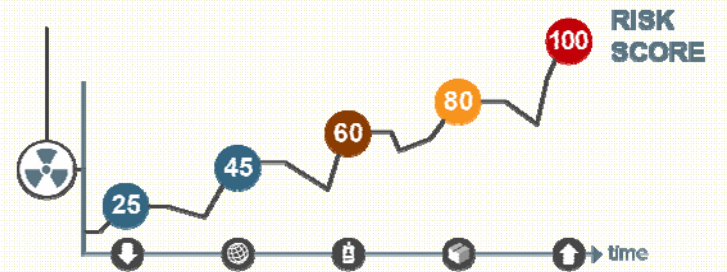
Finding the malicious in the anomalous



SAM FULLER

3	HOSTS
22	IPs
95	LOGONS
120k	SESSIONS

-142h



Ransomware Example

Indicators

C&C Communication

SMB based bot scanning

75
RISK SCORE

WannaCry Ransomware Attack

UEBA

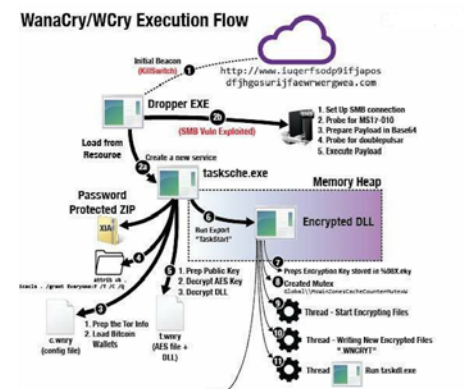
- DGA Detection e.g.
iuqerfsodp9ifjaposdfjhgosurijfaewrwergrwea[.],
xxlvbrloxvriy2c5[.], sqjolphimrr7jqw6[.],
76jdd2ir2embyv47[.]



- Behavioral Analytics on baseline behavior of systems and detecting anomalous communication patterns



Stateful Risk Score for Compromised System



Data Exfiltration Example

Indicators

Access to internal
sensitive information

Moving sensitive data
offshore

UEBA

- Abnormal access to internal data



- Abnormal USB writes
- Abnormal Uploads to Box, Dropbox



High Risk Score for user

75
RISK SCORE

Abnormal Privileged Insider Activity Example

Indicators

UEBA

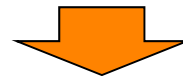
Privilege Escalation

- Escalation of privileges for user not entitled to admin role



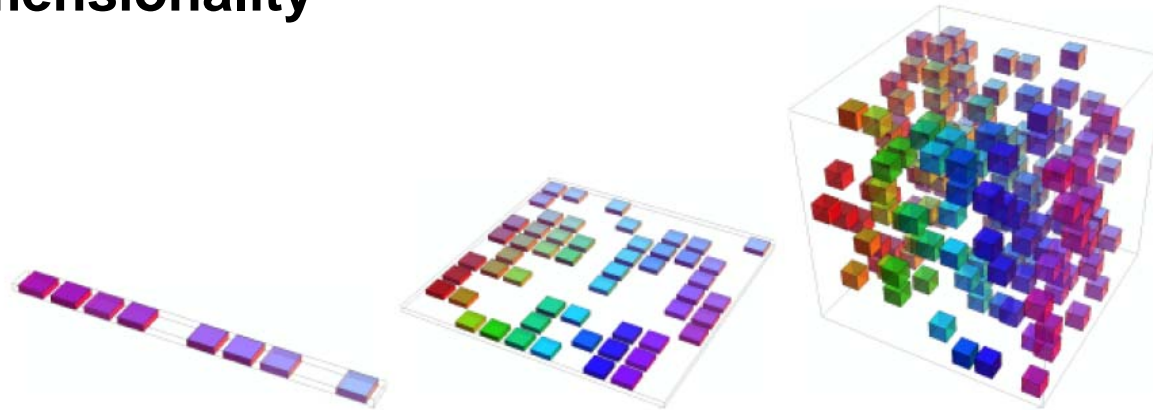
Abnormal Data access

- Excessive Service Ticket requests
- Abnormal data access patterns



High Risk Score for user

Need for dimensionality



Multiple techniques, Expertise in security domain

Supervised	Supervised with manual review	Unsupervised
DNS-DGA (Naïve Bayes)	DNS Tunneling (K-means clustering)	UBA-Server (SVD with Mahalanobis distance)
DNS Exfiltration (Logistic Regression)		GUEBA (z-score)

Data Source Diversity Matters

Type	Examples	Purpose
Network activity	<ul style="list-style-type: none">FirewallIDS/IPSWeb ProxyEmailBro logs,Network trafficNetwork flows	<ul style="list-style-type: none">Lateral movementAbnormal resource accessBrowser exploitsMalware activitySuspicious file downloadsCommand and control activityBeaconing
Remote Access activity	VPN logs	<ul style="list-style-type: none">Credential theft, password sharing
Identity	AD, DHCP logs	<ul style="list-style-type: none">Credential violationsAccount takeoverPrivilege escalations
Infrastructure	DNS logs	<ul style="list-style-type: none">Command and control activityTunnelingExfiltration
3rd party alerts	FireEye, WildFire alerts	<ul style="list-style-type: none">Incorporate alerts into user risk profiles
Threat Intelligence feeds	Commercial & STIX feeds	<ul style="list-style-type: none">Perform historical impact assessment
Endpoint Logs	DLP, FIM	<ul style="list-style-type: none">Suspicious file activityUSB, cloud based file exfiltration
Physical activity	Badge logs	<ul style="list-style-type: none">Building access violationsTailgating

Lateral Movement

Features	Data Source
Authentication activity <ul style="list-style-type: none">▪ Successful/Failed login activity rates▪ Password change rates▪ Odd time of logins▪ New host logins▪ Excessive user logons on hosts▪ Locked/disabled/expired account/restricted workstation logins	<ul style="list-style-type: none">▪ AD Logs
Access to internal applications / servers/ peers <ul style="list-style-type: none">▪ Odd time of access (first and last access)▪ Upload/download deviations▪ Abnormal activity duration/ session count▪ New server / application / peer access▪ Port counts	<ul style="list-style-type: none">▪ Packets▪ NetFlow▪ Firewall logs

Account Takeover

Features	Data Source
Authentication activity <ul style="list-style-type: none">▪ Service ticket request rates▪ Unique/New service ticket requests▪ Account creation/ disable/ lockout / deletion rates▪ Group change deviations▪ Locked/disabled/expired account/restricted workstation logins	<ul style="list-style-type: none">▪ AD Logs
Access to internal applications/ servers/ peers <ul style="list-style-type: none">▪ Odd time of access (first and last access)▪ Upload/download deviations▪ Activity duration/ session counts▪ New server / application access▪ New host access▪ Port counts	<ul style="list-style-type: none">▪ Packets▪ NetFlow▪ Firewall logs

Infiltration / Credential Compromise

Features	Data Source
<ul style="list-style-type: none">▪ Land-speed violations: Access from different locations that violate the physical limits of movement between them (city or country)▪ City: New city access for the first time▪ Activity duration/ session counts▪ Bytes in, bytes out▪ Odd time of access (first and last access)	<ul style="list-style-type: none">▪ VPN logs

Exfiltration

Features	Data Source
<ul style="list-style-type: none">▪ DNS-Exfiltration	<ul style="list-style-type: none">▪ DNS logs or traffic
Access to internet / external applications / cloud apps <ul style="list-style-type: none">▪ Time of access▪ Upload/download deviations▪ Activity duration/ session counts▪ New server / application access▪ Port counts▪ Country visited – New /Counts▪ Entropy Mismatch	<ul style="list-style-type: none">▪ Packets▪ NetFlow▪ Firewall logs▪ Web Proxy logs
Email activity <ul style="list-style-type: none">▪ Odd time of email activity▪ Upload/download deviations▪ Attachment size/volumes▪ Email counts▪ Suspicious / disposable domains▪ Activity to non-corporate or non-affiliated domains	<ul style="list-style-type: none">▪ Email traffic▪ Email logs
Endpoint Activity <ul style="list-style-type: none">▪ Volume of data written to USB, first time USB writes▪ New processes / Registry changes▪ New file creations/ modifications/ opened/ created▪ Changes in file read/write/deletes/ permissions	<ul style="list-style-type: none">▪ File Integrity Monitoring▪ DLP logs▪ Endpoint logs

Generalized Behavioral Analytics

Data Selection

Data Sources
(Proxy / FW / AD / VPN logs, packets...)

Target Entities
(users/hosts)

Use Case Filter
(data filters)

Feature Examples

Time

First and last access each day

Counter

Volume of downloaded or uploaded bytes

Cardinality

Number of email recipients per sender

New Value

Country visited for the first time

Location

Geo-location of VPN logon

Behavior Profiling

Baseline
(Peer, History)

Window
(Duration)

Profiling Model
(SVD, RBM, BayesNet, K-means, Decision tree...)

Anomaly Detection

Real-time vs. Offline

Distance
(Mahalanobis, Energy)

Event Generation
(Severity, Stage)

Behavioral analytics for resource access - server

Identifying abnormal access to high value servers by time and download volume

Data
Selection

Data Sources
(Proxy logs, server logs, flows,
packets...)

Target Entities
(Users to be profiled)

Use Case Filter
(High-value server)

Features

Time
(First/last access)

Counter
(Volume of download)

Behavior
Profiling

Baseline
(History)

Window
(14 days)

Profiling Model
(SVD)

Anomaly
Detection

Real-time vs. Offline
(Offline)

Distance
(Mahalanobis)

Event Generation
(100% Severity, Internal
Activity)

Behavioral Analytics for Resource Access - Building

Identifying abnormal access to physical facility

Data
Selection

Data Sources

(Badge logs)

Target Entities

(Users to be profiled)

Use Case Filter

(No filter)

Features

Time

(First/last access)

Behavior
Profiling

Baseline

(Peer)

Window

(14 days)

Profiling Model

(SVD)

Anomaly
Detection

Real-time vs. Offline

(Offline)

Distance

(Mahalanobis)

Event Generation

(100% Severity, Internal
Activity)

Behavioral Analytics for Resource Access - Files

Detecting abnormally high PDF downloads

Data
Selection

Data Sources
(Packets or proxy logs)

Target Entities
(Users to be profiled)

Use Case Filter
(High-value server; PDFs only)

Features

Counter
(Volume of download)

Behavior
Profiling

Baseline
(Peer, History)

Window
(14 days)

Profiling Model
(ZScore)

Anomaly
Detection

Real-time vs. Offline
(Offline)

Distance
(Mahalanobis)

Event Generation
(100% Severity, Internal Activity)

Behavioral Analytics for access to job sites

Identifying flight risk users

Data
Selection

Data Sources
(Packets or proxy logs)

Target Entities
(Users to be profiled)

Use Case Filter
(Job/salary site URLs)

Features

Cardinality
(# unique visits)

Behavior
Profiling

Baseline
(Peer)

Window
(14 days)

Profiling Model
(SVD)

Anomaly
Detection

Real-time vs. Offline
(Offline)

Distance
(Mahalanobis)

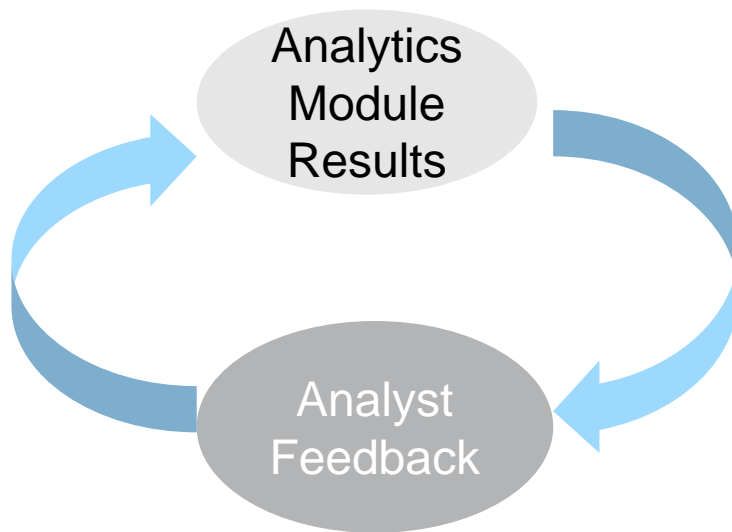
Event Generation
(100% Severity, Internal
Activity)

Differentiated Risk Scoring



- Contextually weighted model
 - Hidden Markov Model
 - Unlike competitors that linearly add up scores for all detected events
 - E.g., a C&C event followed by a privilege escalation event is treated differently from two consecutive C&C events
- Score incorporates:
 - Sequencing of events
 - Distribution of events across kill chain stage
 - Severity and confidence of alerts
- Customer input to shape risk score at a granular level

Adaptive Learning

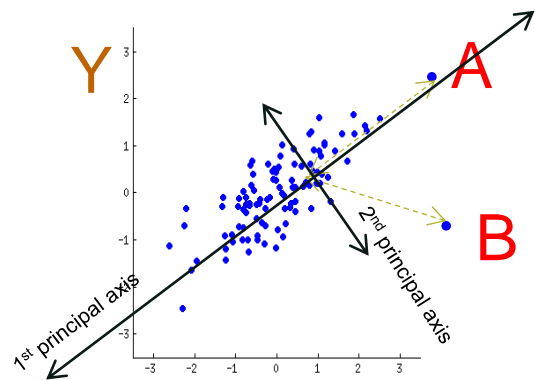


- Incorporate analyst/business context
- Ability to train models for exception handling, risk scoring
- Granular exceptions
 - User/ Application/ Baseline Type
 - E.g., user Bob's access of AWS is exempt from peer detection because he is an admin
 - Global / user specific whitelists
 - E.g., site "xyzinc" facilitates vulnerable PDF file downloads but it is an authorized partner site

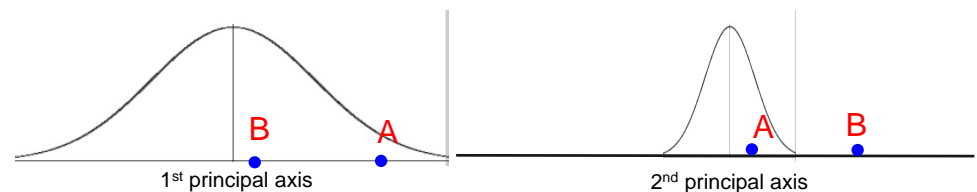
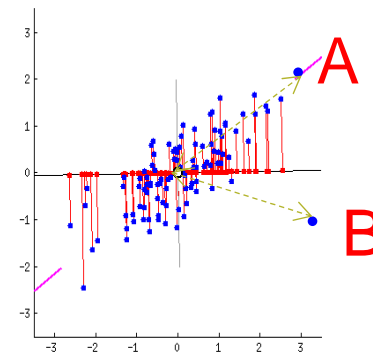
Anomaly detection after applying SVD

Unsupervised Machine Learning

- In the case of [UBA-Server](#), we have 5 features
 - first/last access time, upload/download volume, number of eflows and duration
- We evaluate Mahalanobis distance, to determine if it is an anomaly
- A score of >60 is an anomaly



X: "time of first access" and Y: "download volume"



Risk Scoring

Bayesian inference model

Bayes Theorem : $P(B / A) = \frac{P(A / B) * P(B)}{P(A)}$

Risk Score (RS) :

$$P(RS / f_1, f_2, f_3, f_4) = \frac{P(f_1, f_2, f_3, f_4 / RS) * P(RS)}{P(f_1, f_2, f_3, f_4)}$$

$$P(RS / f_1, f_2, f_3, f_4) = P(f_1 / RS) * P(f_2 / RS) * P(f_3 / RS) * P(f_4 / RS) * P(RS)$$

where,

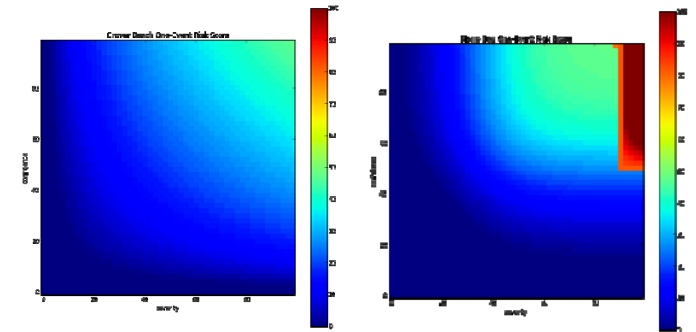
$$f_1 = \max(\sqrt{\text{severity} * \text{confidence} * \text{time_decay}})$$

$$f_2 = \sum_i^{\text{alerts}} \sqrt{(\text{severity} * \text{confidence}) / 100}$$

$$f_3 = \sum_i^{\text{attack_stages}} (\text{highest_sqrt_of_sev} * \text{conf})$$

$$f_4 = \ln \sum_i^{\text{alerts}} (\sqrt{\text{severity} * \text{confidence} * \text{time_decay}})$$

P(RS) is assumed to have a uniform probability distribution



Improvements for 'single event' trigger

Thank You