

# RSA® Conference 2015

San Francisco | April 20-24 | Moscone Center

SESSION ID: STR-W03

## Data Science Transforming Security Operations

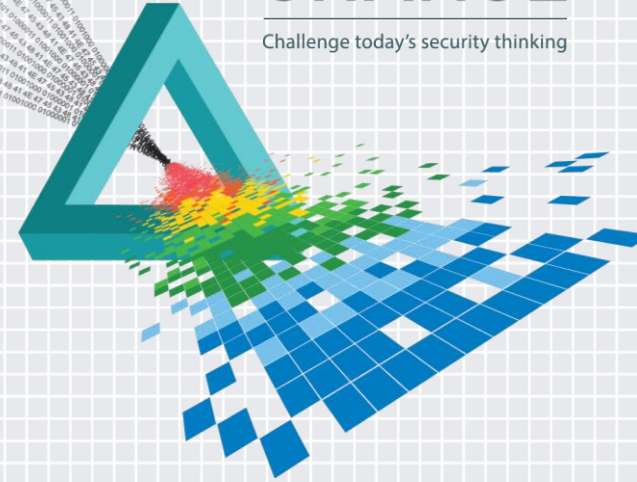
**Alon Kaufman Ph.D.**

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Director Data Science & Innovation  
RSA

# CHANGE

Challenge today's security thinking



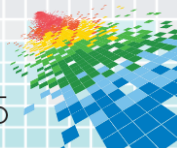
# Agenda

## Transforming Security Operations with Data Science

- ◆ The **Vision**: Where we should head & Why
- ◆ The **Strategy**: What we need to achieve
- ◆ The **Tactics**: How we get there



**Key message:** Data science is a key methodology and technology in this transformation, its not merely a feature....

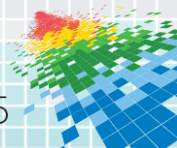


# Security Operation – Current Operation Model

## Threat Landscape



## Current Security Operations



# Security Operation – Current Operation Model

Threat Landscape

Current Security Operations

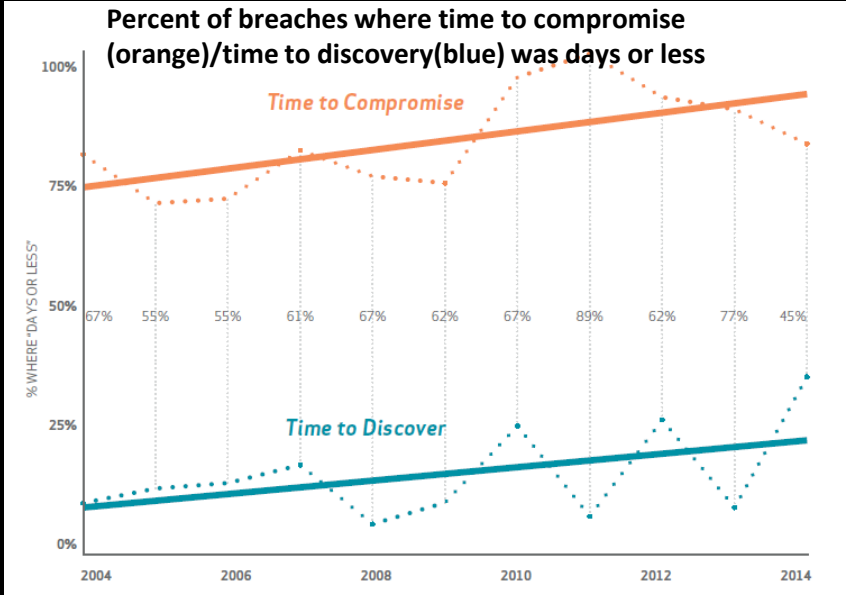
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## Current Approach in Failing

2015  
Information security incidents:  
**42.8 M**  
**↑48%**

Confirmed breaches:  
**2122**  
**↑55%**

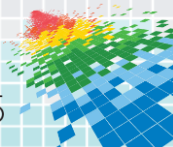
Percent of breaches where time to compromise (orange)/time to discovery (blue) was days or less



Year	Time to Compromise (%)	Time to Discover (%)
2004	75%	10%
2005	72%	10%
2006	72%	12%
2007	82%	15%
2008	78%	5%
2009	75%	10%
2010	85%	25%
2011	95%	8%
2012	90%	25%
2013	85%	8%
2014	90%	35%

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Failures of the security industry, Yoran, Robertson, 2015  
Verizon, 2005 DBIR



# We are Being Defeated, What's Happening?

- ◆ We have the technology, the brains, the funding....

***The attackers operate in a known environment!***

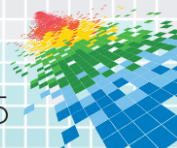
***The attackers job is more predictable!!***

***Their lives are easier!!!***

***Our operations are slow... reactive...***

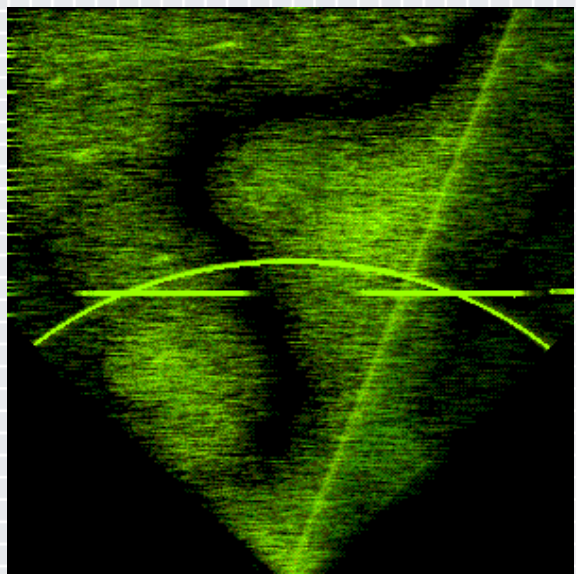
***We are running into a skill gap/fatigue...***

***We don't make their life hard enough...***

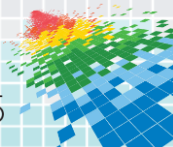


# Technology at the Aid of Human Operations

## 1965 Aviation Radar



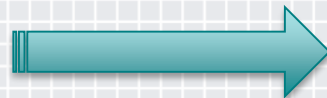
## 2015 Tactical Situation Display



# Transforming the Role of the Human Operations Behind the Machine

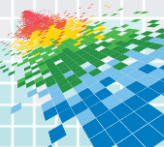


What's the "driver's" role?

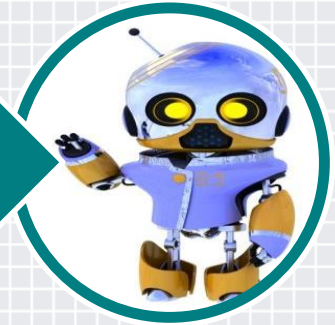


What's the "Pilot's" role?

The Human role:  
Leaving to the human the intentions & final decisions!!



# What Happened to The Machine Operator



Human

Platform

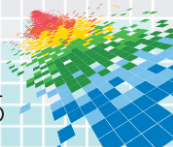
“Stick & throttle”  
Gun

Jet engines  
Missiles

Radar  
BVR missiles  
RWR

Detect threat  
Identify danger  
ECM  
Select weapon  
Human decision

...





# What Happened to The Machine Operator



Fly

Fighter

System  
op

Think/  
decide

Policy



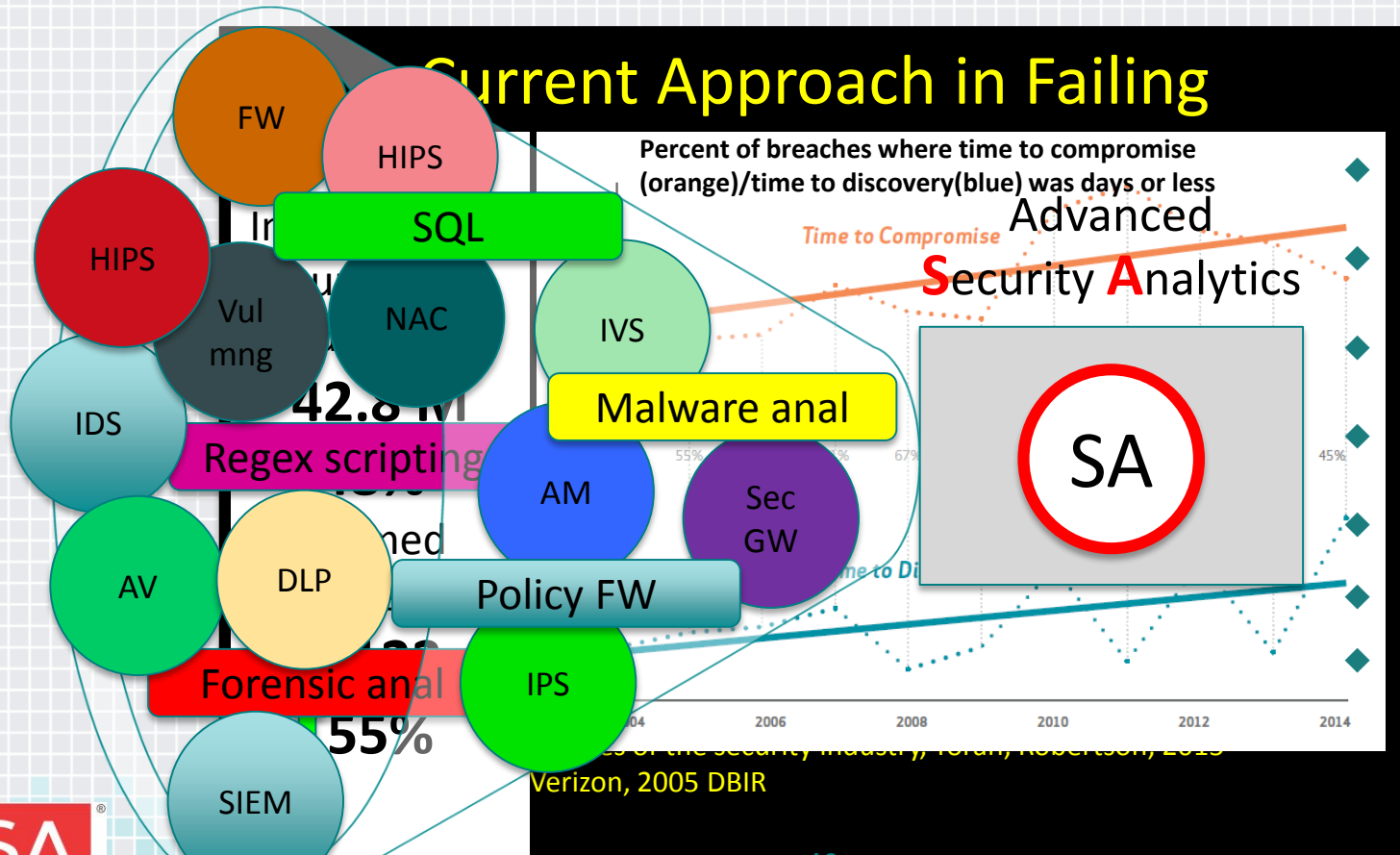
Human

Platform

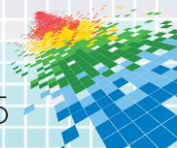
Want the same in security?

Leave thinking and decisions to humans and let the machines and platforms do the rest

# It's Time for the Next Era

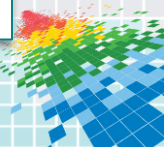
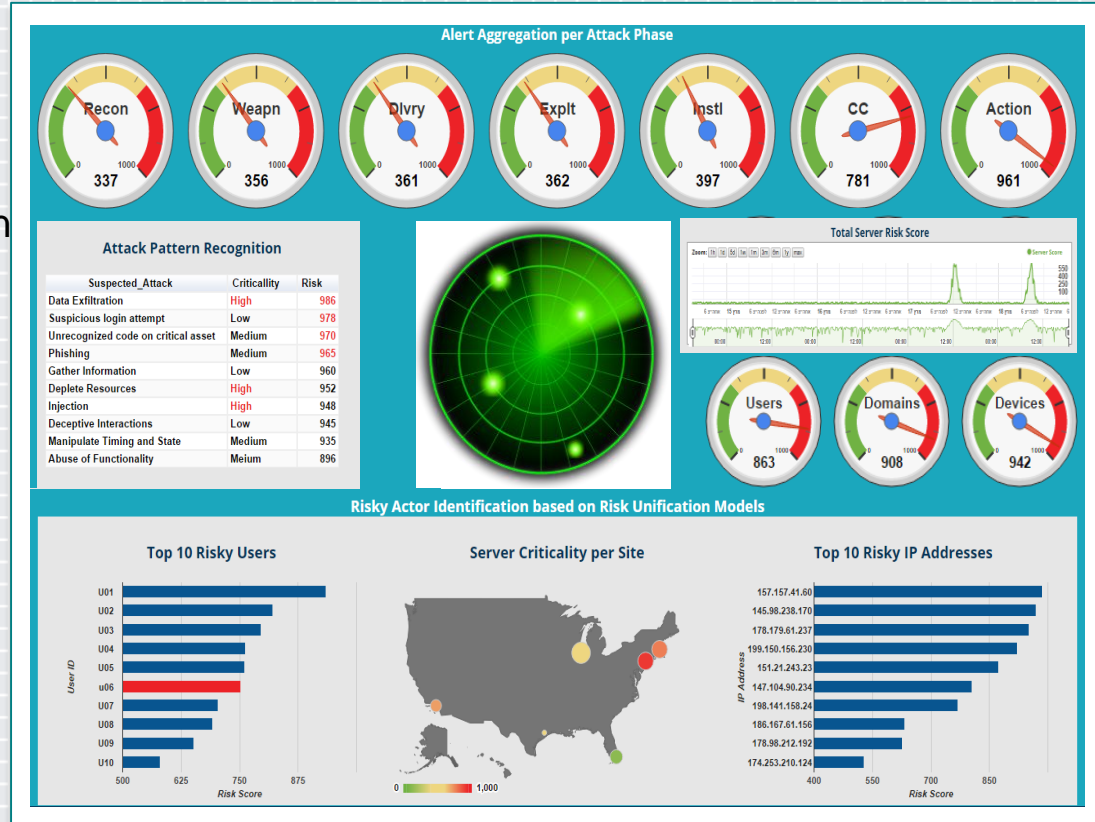


- ◆ Comprehensive
- ◆ Accurate
- ◆ Proactive
- ◆ Faster
- ◆ Context aware
- ◆ Adaptive
- ◆ Leverage intelligence



# The Next Era: Advanced Security Analytics

- ◆ **Comprehensive**
  - ◆ All possible inputs
  - ◆ Going beyond the known
- ◆ **Proactive / Preventive**
  - ◆ Automatic
  - ◆ Actionable insights & recommendation
- ◆ **Human intelligence**
  - ◆ Knowledge
  - ◆ Sharing
  - ◆ Intentions & decisions



# The Next Era: Advanced Security Analytics

- ◆ **Comprehensive**
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Detect

- ◆ Attacks vs Alerts
- ◆ Prioritized
- ◆ Immediate



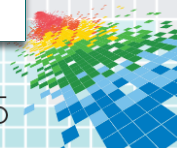
Investigate

- ◆ Hierarchical model
- ◆ All prepared
- ◆ Risk & impact

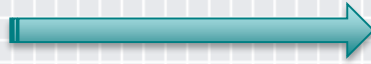


Respond

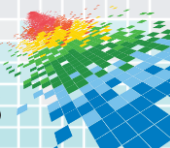
- ◆ Automatic
- ◆ Self learning
- ◆ Recommend



# We Need to Transform...



So *what* is needed to get there?



# We Need to Transform...

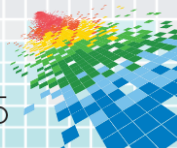


So *what* is needed to get there?

We have the **DATA** – We know the **PROBLEM**...

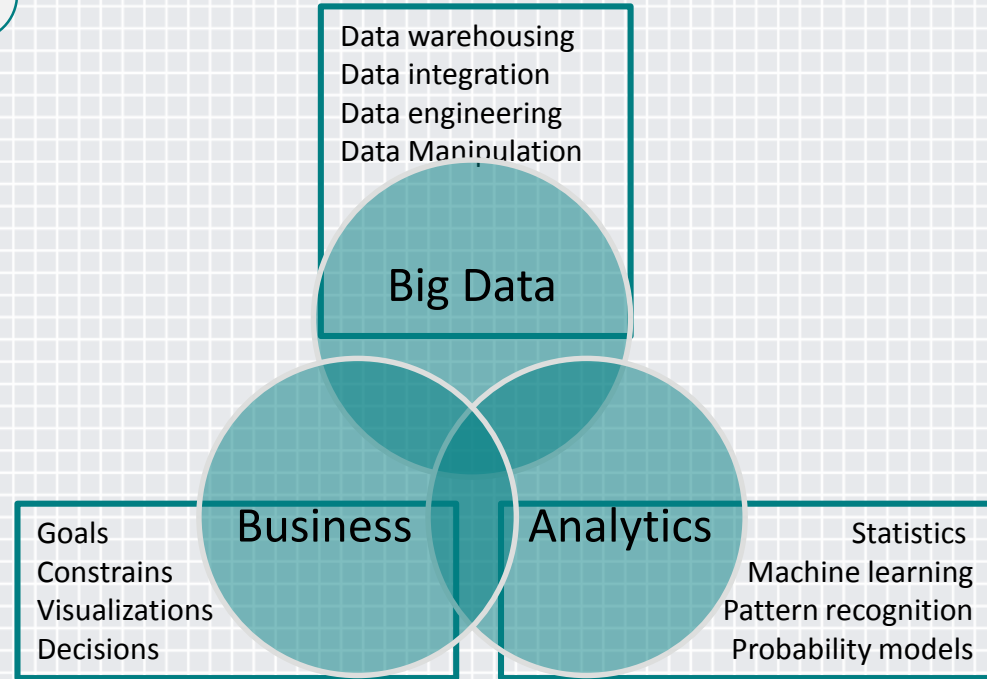
we need to reveal the insights & provide actionable outcomes

**Data Science!!!**

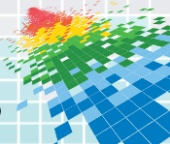


# What is Data Science in 47 Seconds

The use of techniques as **statistics** and **machine learning** on big multi-structured data, to identify **correlations** and causal **relationships**, **classify** and **predict** events, identify **patterns** and **anomalies**, **infer** probabilities and interest, with the goal of extracting meaning from data and creating **valuable products**.



The Art of solving business problems utilizing the available data



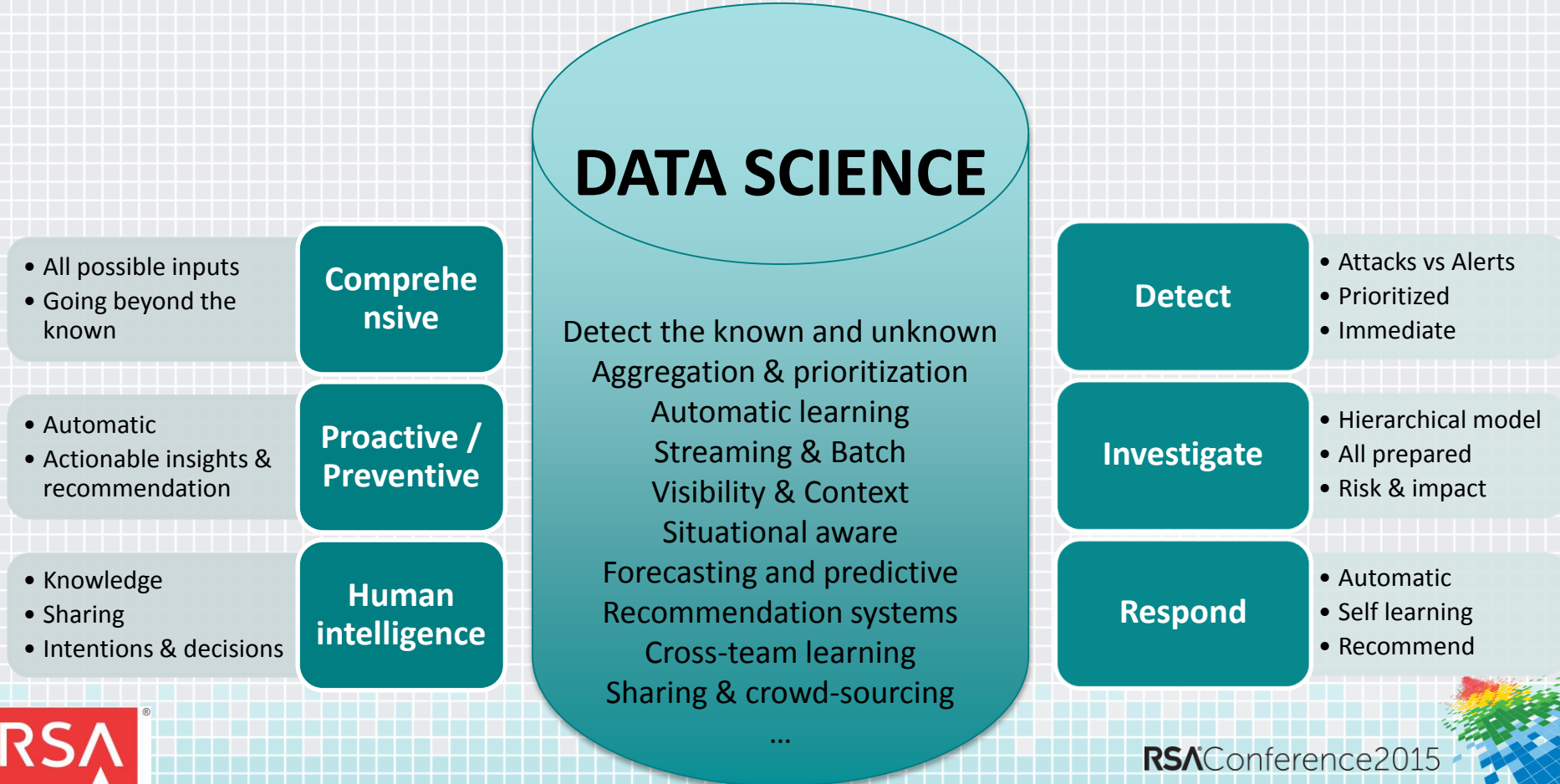
# Data Science Applicative Practice

- ◆ Data Science is no magic!
  - ◆ Understand your problem
  - ◆ Learn your data
  - ◆ Create valuable indicators
  - ◆ Apply the right modeling techniques!
  - ◆ Measure & Success
- ◆ Joint iterative work with domain experts and SMEs





# The Data is all There -> Reveal the Essence



# Examples of Data Science in the Transformation

## ◆ Detect

- ◆ Attacks vs Alerts
- ◆ Prioritized
- ◆ Immediate



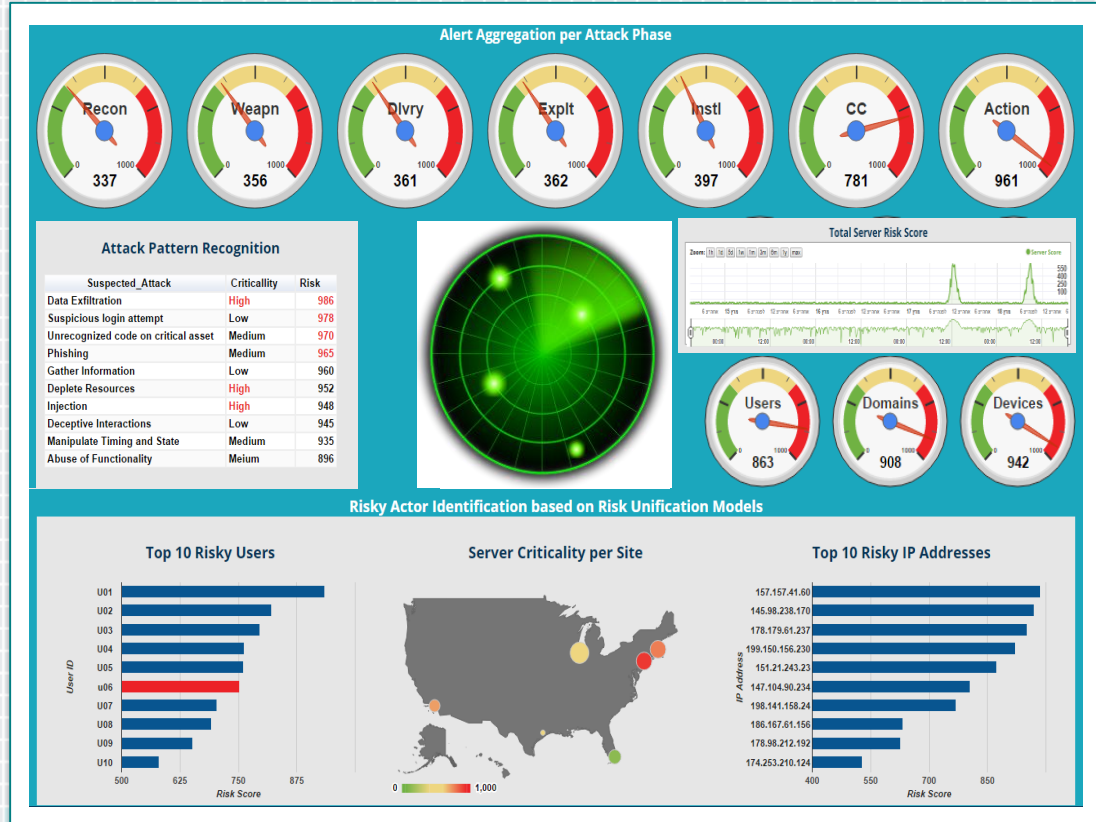
## ◆ Investigation

- ◆ Hierarchical model
- ◆ All prepared
- ◆ Risk & impact



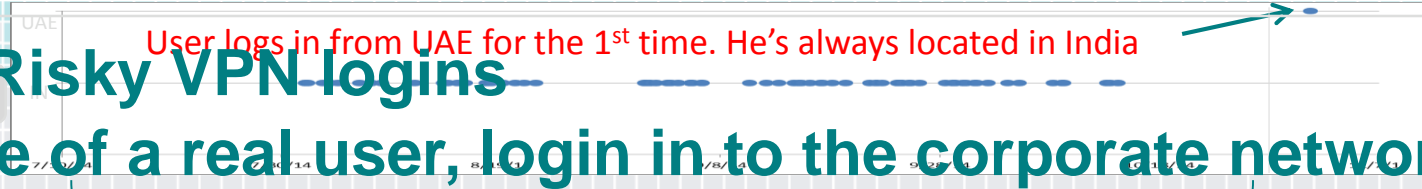
## ◆ Respond

- ◆ Automatic
- ◆ Self learning
- ◆ Recommend



# Detect Risky VPN-logins

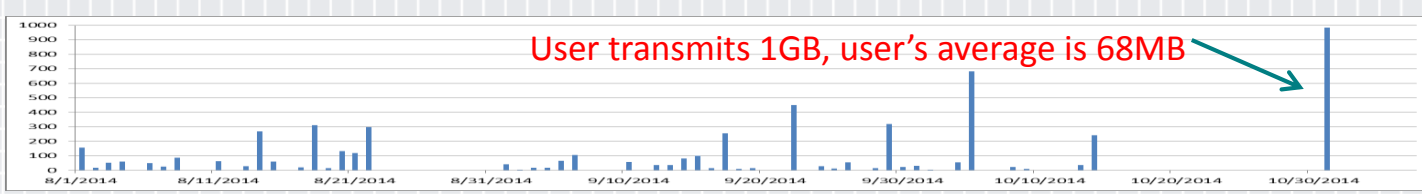
## Example of a real user, login in to the corporate network



Score: **92**



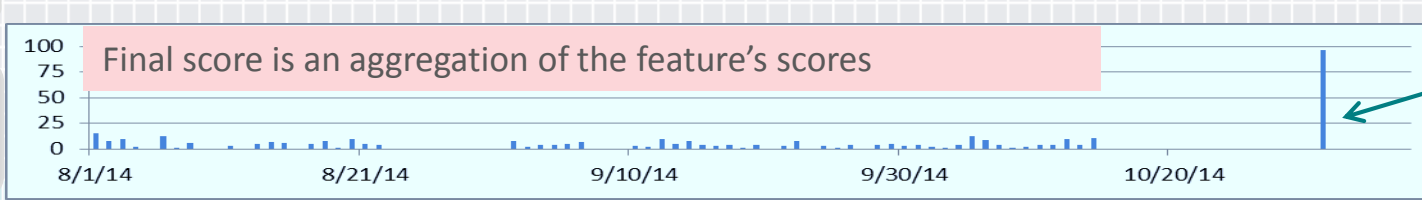
Score: **90**



Score: **93**



Score: **84**



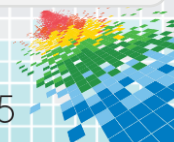
Device

Transmitted Data [MB]

Session Duration [hr]

Many more

Final Score



Country



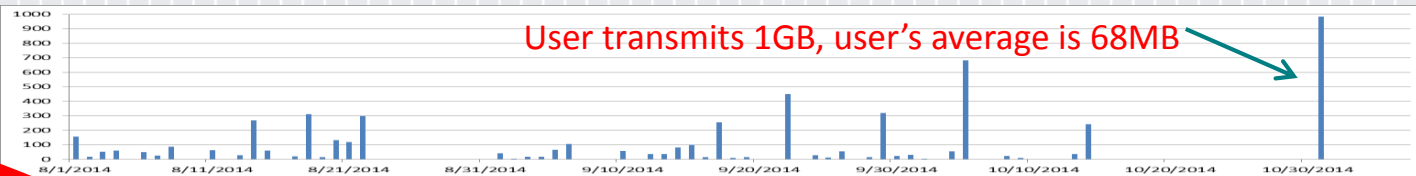
Score: 92

Device



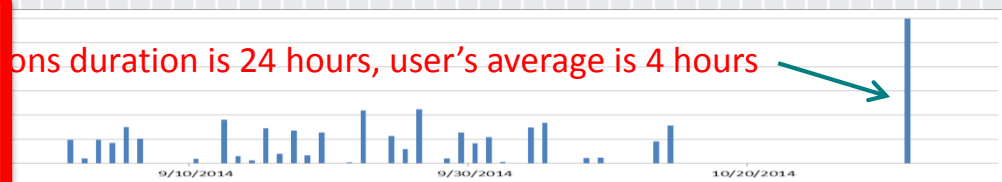
Score: 90

Transmitted Data [MB]



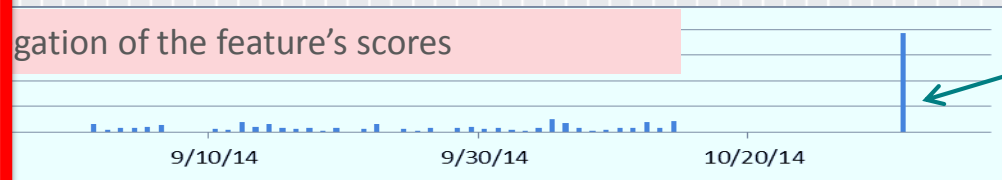
Score: 93

Session duration is 24 hours, user's average is 4 hours



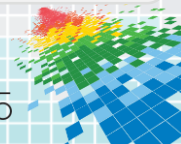
Score: 84

Aggregation of the feature's scores



Aggregate Score: 98

- Based on nested anomaly detection models
- User/Group behavior analysis
- Ranking based on statistical significance

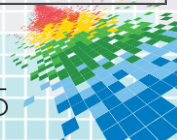


# Recommending New IOCs

- ◆ When the analyst creates a new IOC the system can recommend:
  - ◆ How strong is this new IOC by itself?
  - ◆ Is it similar to already existing IOC?
  - ◆ What is the best threshold setting (if applicable)?
  - ◆ Recommends new IoC

New Recommended IoCs, based on combining existing IoCs

	IIOC	Occurrences	Avg. Precision
<b>Some potential IIOC with high scores</b>	FileUnknown + UNSIGNED_AUTORUN_APPDATA	154	0.595
	AUTORUN_TYPE + SUSPICIOUS_DLL_LOADED		0.538
	FileUnknown + SUSPICIOUS_DLL_LOADED	35	0.538
	Autostart + SUSPICIOUS_DLL_LOADED	35	0.538
	UNSIGNED_AUTORUN_APPDATA + SUSPICIOUS_DLL_LOADED	33	0.538
	packed + FLOATING_DLL_IN_OS_PROCESS		0.535
	SuspectThread + dll	130	0.531
	Hook + UNSIGNED_AUTORUN_APPDATA	8	0.526
	packed + FLOATING_DLL_IN_BROWSER_PROCESS		0.52
	Hidden + UNSIGNED_AUTORUN_APPDATA	7	0.517
	Hidden + SUSPICIOUS_DLL_LOADED	7	0.517
	dll + UNSIGNED_AUTORUN_APPDATA	46	0.516
	Hook + FLOATING_DLL_IN_OS_PROCESS	24	0.511
	FileAttributeHidden + UNSIGNED_AUTORUN_APPDATA	21	0.511
	SuspectThread + NetworkAccess	21	0.511
	NetworkAccess + FloatingModule	15	0.511
FileUnknown + SUSPECT_THREAD_FLOATING_MODULE	14	0.511	
FileUnknown + NetworkAccess + FloatingModule	14	0.511	



# Examples of Data Science in the Transformation

## ◆ Detect

- ◆ Attacks vs Alerts
- ◆ Prioritized
- ◆ Immediate



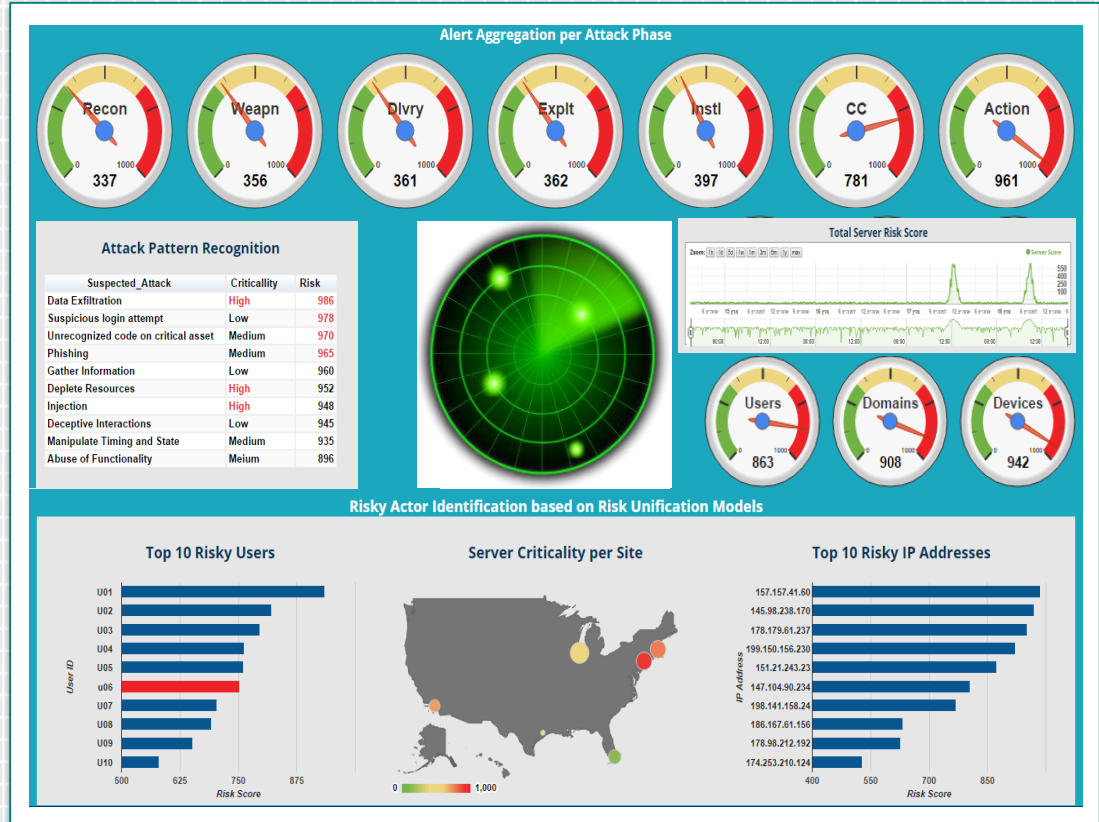
## ◆ Investigation

- ◆ Hierarchical model
- ◆ All prepared
- ◆ Risk & impact



## ◆ Respond

- ◆ Automatic
- ◆ Self learning
- ◆ Recommend



## Overall User Risk



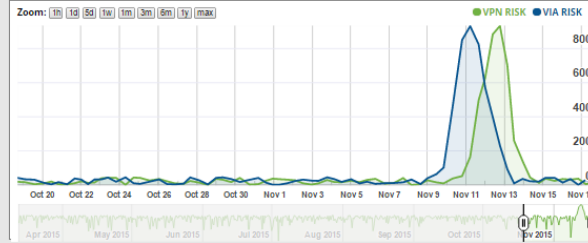
## Entity Unification Risk Model



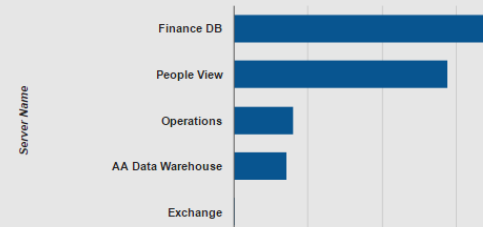
### Location: GeoSpeed Model



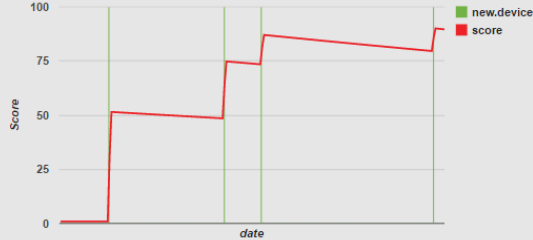
### Access: VPN and VIA Risk Score



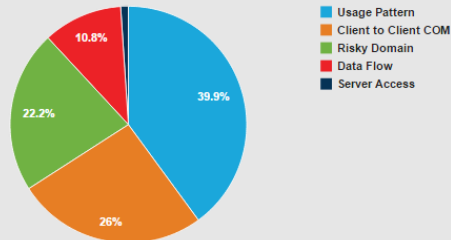
### Top 10 Anomalous Servers for User



### Multiple Device Model



### User Network Activity



### Compliance - Avekva Risk Modeling

	AVEKVA SK	REQ
Overall Identity Risk Score	975	
Reviewed over last 6 months?	NO	
Accounts not accessed over 180 days	3	
Access to critical assets	Yes	
Last password reset	2015-03-12 17:59:12	
Num of password reset in last month	1	
Entitlement violations in last 90 days	0	
Revoked/denied entitlements in last 90 days	0	
Num escalations use in last 90 days	0	
Num of Accounts	10	
Num of entitlements	10	
% of people with similar entitlements	60%	

## User Profile from Data Sources

### HR Data Base

People_View	PV
User Name	Maxwell Smart
Employee ID	3141596
Title	Director
Department	Engineering
Location	Boston
Status	Contractor

### Active Directory

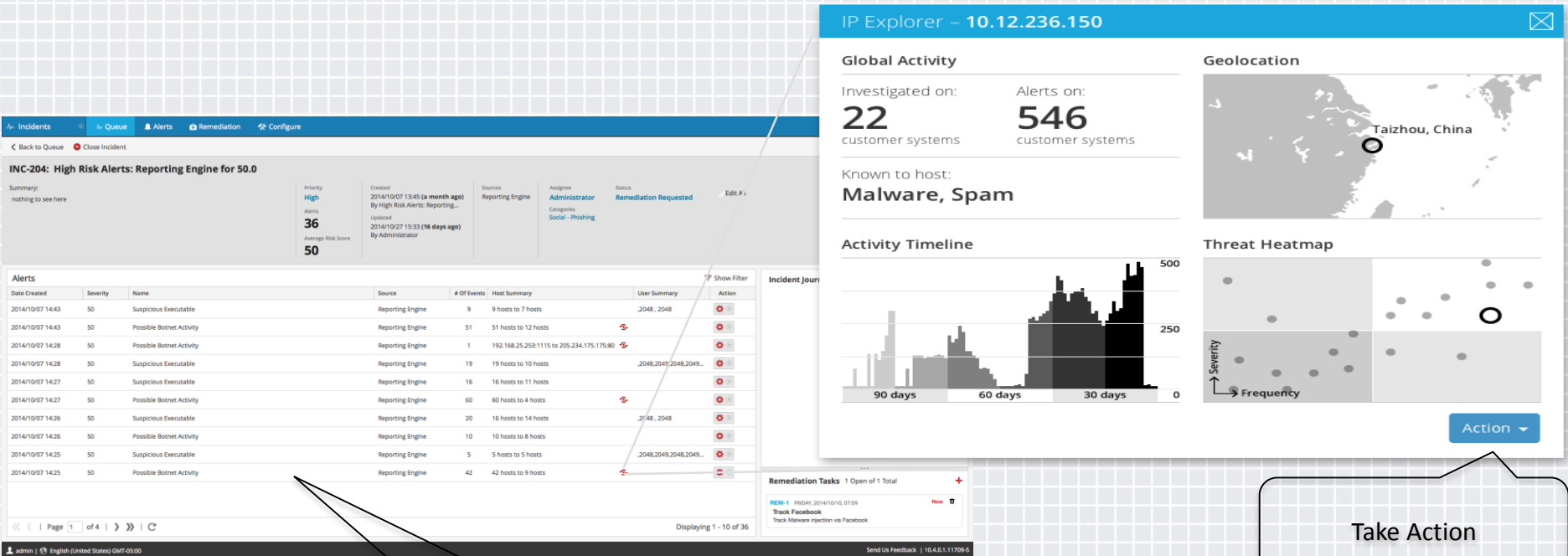
Active_Directory	AD
User Name	Admin
Account creation	2015-03-10 15:25:59
Account Expiration	NA
Bad passwords in last day	0
Last login	NA
Last bad password	NA
Primary group	NA

### Consistency

AD2PV	Consistency
Country	NO
Phone number	Yes
e-mail	Yes
Company	Yes
Division	Yes
manager	Yes

# Community Based Data Enrichment

- ◆ Let the analyst know what the community thinks about the IP



**IP Explorer - 10.12.236.150**

**Global Activity**  
Investigated on: **22** customer systems  
Alerts on: **546** customer systems

**Geolocation**  
Taizhou, China

**Activity Timeline**  
Known to host: **Malware, Spam**

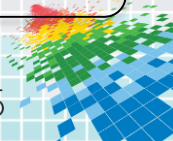
**Threat Heatmap**  
Severity vs Frequency

**Remediation Tasks** 1 Open of 1 Total  
REM-1 [REM-1] 2014/10/16 07:29  
Track Facebook  
Track Malware injection via Facebook

Date Created	Severity	Name	Source	# Of Events	Host Summary	User Summary	Action
2014/10/07 14:43	50	Suspicious Executable	Reporting Engine	9	9 hosts to 7 hosts	..2048_2048	[Action]
2014/10/07 14:43	50	Possible Botnet Activity	Reporting Engine	51	51 hosts to 12 hosts	..2048_2048	[Action]
2014/10/07 14:28	50	Possible Botnet Activity	Reporting Engine	1	192.168.25.253:1115 to 205.234.175.175:80	..2048_2048	[Action]
2014/10/07 14:28	50	Suspicious Executable	Reporting Engine	19	19 hosts to 10 hosts	..2048,2049,2048,2049...	[Action]
2014/10/07 14:27	50	Suspicious Executable	Reporting Engine	16	16 hosts to 11 hosts	..2048_2048	[Action]
2014/10/07 14:27	50	Possible Botnet Activity	Reporting Engine	60	60 hosts to 4 hosts	..2048_2048	[Action]
2014/10/07 14:26	50	Suspicious Executable	Reporting Engine	20	16 hosts to 14 hosts	..2048_2048	[Action]
2014/10/07 14:26	50	Possible Botnet Activity	Reporting Engine	10	10 hosts to 8 hosts	..2048,2049,2048,2049...	[Action]
2014/10/07 14:25	50	Suspicious Executable	Reporting Engine	5	5 hosts to 5 hosts	..2048,2049,2048,2049...	[Action]
2014/10/07 14:25	50	Possible Botnet Activity	Reporting Engine	42	42 hosts to 9 hosts	..2048,2049,2048,2049...	[Action]

Investigate & Incidents

Take Action





# Examples of Data Science in the Transformation

## ◆ Detect

- ◆ Attacks vs Alerts
- ◆ Prioritized
- ◆ Immediate



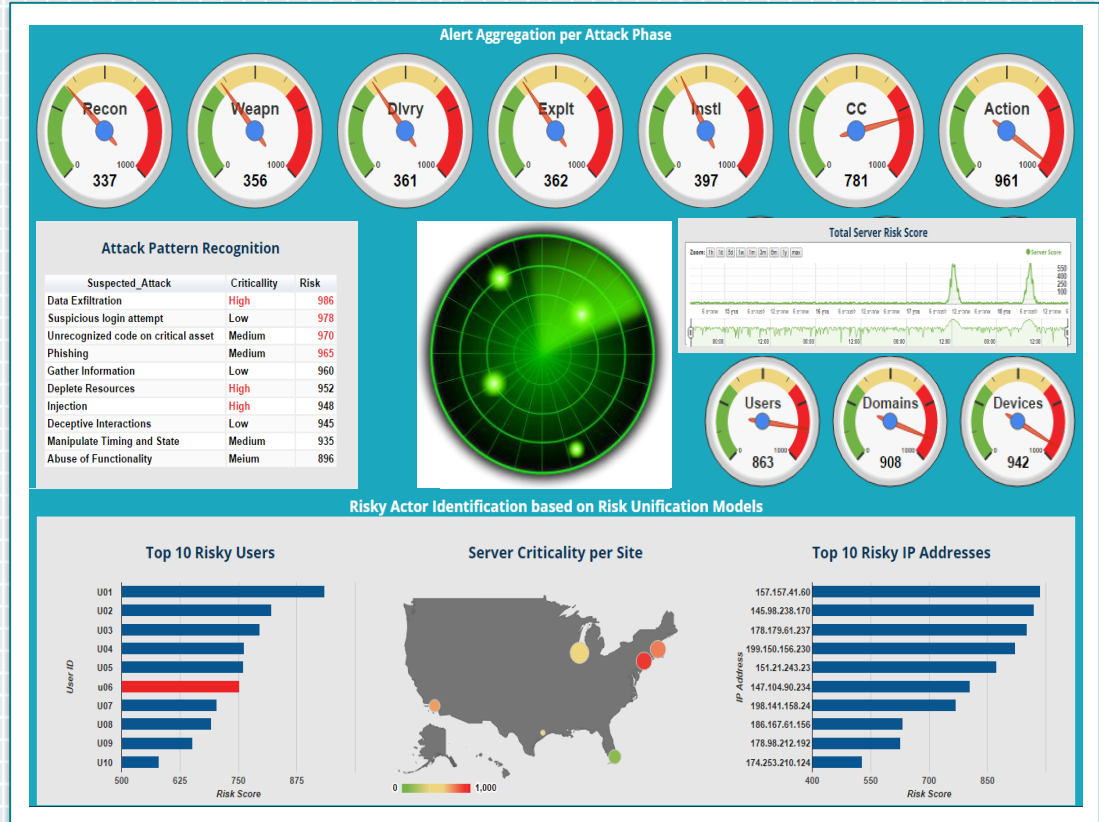
## ◆ Investigation

- ◆ Hierarchical model
- ◆ All prepared
- ◆ Risk & impact



## ◆ Respond

- ◆ Automatic
- ◆ Self learning
- ◆ Recommend



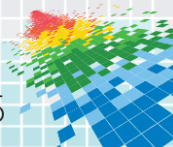
# Learning and Self-Improving Detection

## Risk-Based Security and Self-Protection

- ◆ Ongoing, automatic self-learning fraud detection model
- ◆ Real-time decisions/actions based on risk scores



Data Science based Risk Engine

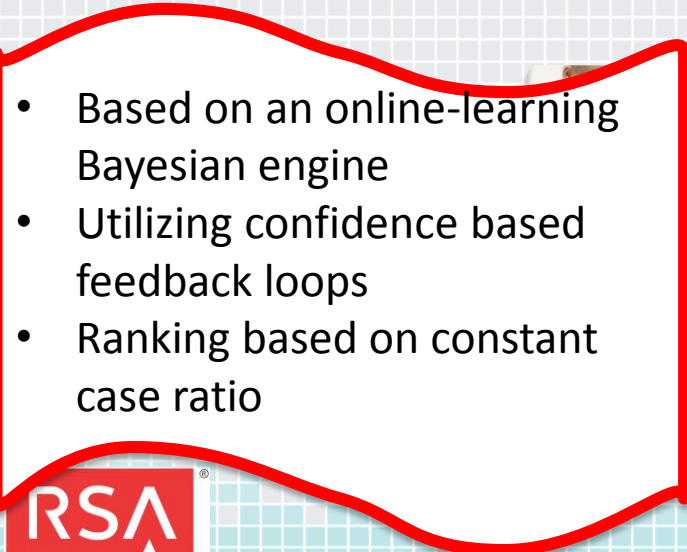
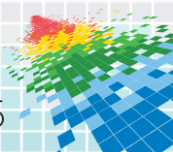


# Learning and Self-Improving Detection

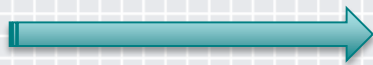
## Risk-Based Security and Self-Protection

- ◆ Ongoing, automatic self-learning fraud detection model
- ◆ Real-time decisions/actions based on risk scores

- Based on an online-learning Bayesian engine
- Utilizing confidence based feedback loops
- Ranking based on constant case ratio

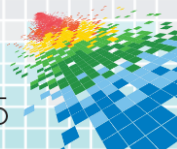



# We Need to Transform...



So *what* is needed to get there?

So *How* do you get there?



# The Transformation Journey

- Collection and Big Data Platform

Data



- Advanced Analytics platform

Analytics



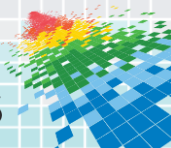
- Think differently
- Use Data Science

Data Science



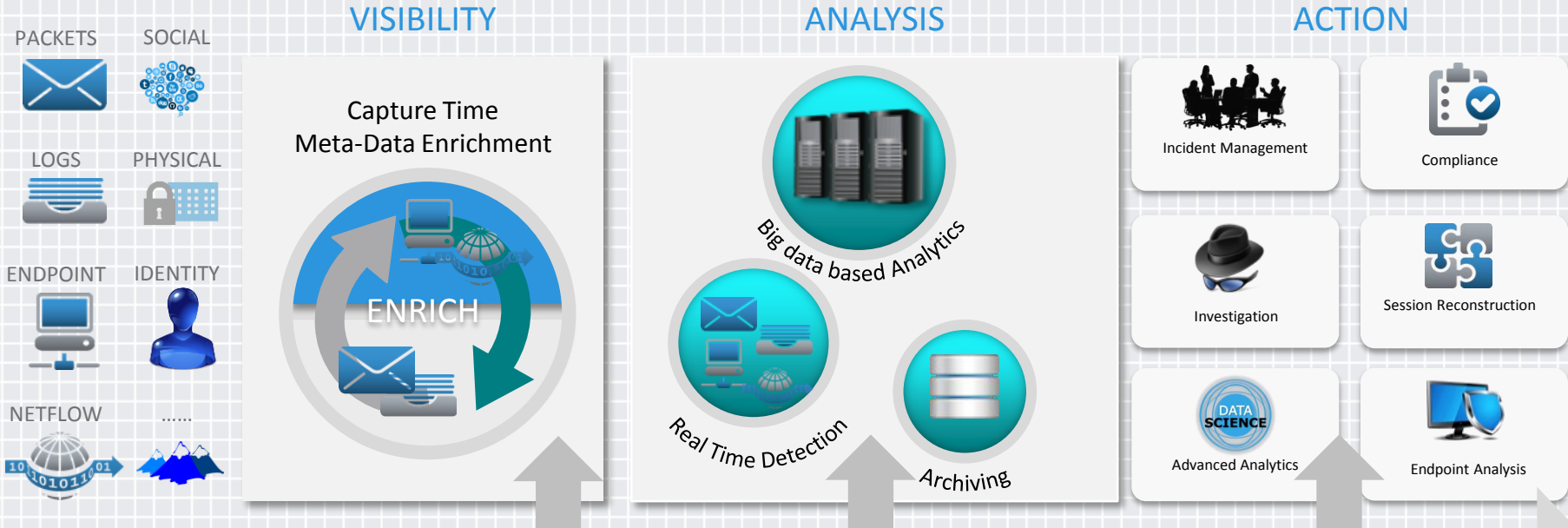
Platform

Skills



# Conceptual Security Analytics Platform

Capture, enrich and analyze data from across your network



Intelligence Sharing platforms

Threat Intelligence | Rules | Parsers | Feeds | Reports | Research

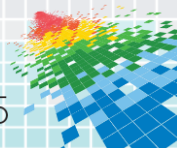
# Taking the Data Science Path

## ◆ THINK DIFFERENTLY

- ◆ Data science is not a feature -> it's a methodology!
- ◆ Data science capabilities should synergize, empower and enhance the security experts -> not replace them!!

## ◆ USE DATA SCIENCE - Building your own (??)

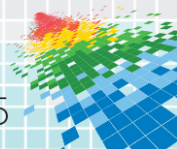
- ◆ Skills and expertise
- ◆ How specific are your requirements
- ◆ Learning from the past – BI, web analytics, retail analytics
- ◆ Leveraging the community and crowd sourcing



# Summary - Transforming Security Operations

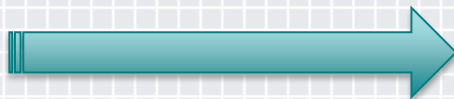


- ◆ Current approaches are failing
- ◆ We have the data – take advantage of it, and leverage human intelligence for the intentions & decisions
- ◆ Security Analytics platforms with baked-in DS; Aggregate, prioritize, recommend, self-learning & crowd sourcing....

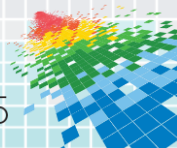




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# RSA®Conference2015

San Francisco | April 20-24 | Moscone Center

SESSION ID: STR-R04

## Data Science Transforming Security Operations

**Key message:** Data science is a key methodology and technology in this transformation, its not merely a feature...

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Director Data Science & Innovation, RSA

# THANK YOU

# CHANGE

Challenge today's security thinking

