Welcome we will get started shortly...

## Splunk ITSI Workshop



### Presenter: Paul Winchester

### 16<sup>th</sup> June 2020

Note: This workshop is based on Splunk's own Splunk4Ninjas – ITSI sessions. Also note: This is not an actual picture of me. But it's quite close...



### **Register and Create Your Environment**

## 1. If you've not yet created your Splunk environment, use this link to register:

#### http://splunk4rookies.com/4320/self\_register

#### **2.** Log into you instance:

http://<aws\_instance>:8000/en-US/

username : admin

password : smartway

Please note that the instances take 30mins to initialise due to Machine Learning backfilling.

Congratulations! Your Splu You have <b>48 hours</b> ahead to Please allow a few minutes	nk sandbox has been created. play until termination. for your instance(s) to be accessible.	
Access link(s): http://ec2-34-253-204	4-181.eu-west-1.compute.amazonaws.com:80	
First Name*	Rob	1
Last Name*	Larkman	
Job Title*	Sales Engineer	
Email*	rlarkman@splunk.com	
Areas of interest*	Application Delivery Business Analytics Internet of Things IT Operations Analytics Security & Fraud	
Message*	Splunk is awesome Register	





## Agenda

- IT Service Intelligence Overview
- Create a Service
- Creation of two KPI's
- Machine Learning
  - Apply Adaptive Thresholds
  - Predict Analytics
- Deep Dives
- Glass Tables
- Event Analytics Demystified
- Somerford's approach to ITSI



# **IT Service Intelligence**



### Complexity obscures the data you need

IT Ops teams continue to struggle to monitor, investigate, analyze & act





**Too Many Tools** 

Siloed Views of Data

Too Much Event Noise

Long Times to Resolve

### The impact can be significant

Putting revenue, customer experience, employee effectiveness & innovation at risk



### Why The Old Ways Disappoint



#### Challenges

- Many disparate components
- Brittle integrations
- Data is summarized and lost
- Longer root-cause identification
- End-to-end view challenging
- Labor-intensive to manage
- Not agile for digital business

### **IT Service Intelligence Platform Approach**





### **Splunk IT Service Intelligence**

Data-driven service monitoring and analytics



# Key Terminology



### What is a Service

A logical group of technology components that a user deems important to be monitored together. Services can encompass multiple tiers of the IT domain.



Online Store Service



Database Service



Mobile App Service

Call Centre Service

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### **Service Dependencies**

**Application Services** 

**Business Services** 

**Technical Services** 



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### What is a KPI

A Key Performance Indicator and Health Scores constitute the means by which Services are monitored. A KPI is a Splunk saved search in ITSI that helps monitor a specific value.





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#### What is a Health Score

A Health score is a score from 0-100 (0 being critical and 100 being normal) that measures the health of a Service. It is calculated based on all KPIs importance and its status.

Mail Server Service	Hadoop Service	Database Service
20	33.3	42.9
~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	Ś	S



#### What is an Entity

An **Entity** is an optional sub-element of a KPI. A KPI can be filtered by entities. ITSI can import entities from CMDBs & other sources

Severity -	Entity Name 💲	Value \$	
Critical	() mysql-02	0 %	
Normal	mysql-01	mmm 33.26 %	
Normal	mysql-03	man 30.7 %	
Normal	mysql-04	mmm 28.05 %	

### **ITSI Service Definition**

To summarize a service is comprised of:



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# Workshop Back Story



### **Workshop Back Story**

- Buttercup Store has just deployed a new Web service, *however* the engineers forgot to include database monitoring.
- Splunk administrators have deployed collectors on the infrastructure and have requested that we build a service centric monitoring solution.
- The CxO has just also requested that we include some 'AI' as they read on a website Artificial Intelligence can solve everything!
- The lab exercises start on the back of a service decomposition workshop to identify all the components that make up this service.



## **Service Decomposition Outcome**

Example diagram detailing the content of a service, typically this is a photo of a whiteboard



## Let's get stuck in to it!

Point your browser to : <u>http://<aws\_instance>:8000/en-US/</u>

This will be the link you got when you registered.



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# **Services Exercise**



### **Service Analyzer**

The Service Analyzer view provides a visual representation of our services and the KPIs associated KPIs. The highest severity is brought to the top left of each group.

- Select the 'Default Service Analyzer' view.
- Click on the icon to switch the visualization a tree view

								<u>_</u>	IT Service Intelligence
vice Analyzer								Last 24 hours +	Save as Save D
Services Select servic	ice(s) to monitor Filter KP	Is Select KPI(s) to monitor	Show disabled	service(s) Show service	e dependencies		KPI V	alue: Aggregate *	Size: Large •
50 Services O									
0 8									10 Tota
lacturing	Butteroup Store	Order Management	Cart Management	Mobile App	Cleud Databases	Web Front End	External Authentication Se	Product Catalog	
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A KPIS 0 4 • 45 Explorence Effectu- atomy 17 Www.Www.Www proc User S togerout S	Performance Networks 42 Merror Marcellon CPU URGENER Sout Drawners	Mobile App Values Extensis State 2266 Monto CPU Unitation S Automation	Cuelly Monosciences 40 www.hwww.success Role and manage laters	Total Acceptative Units Variability Based Acceptative Units Variability Conference There Advancements	Disk100 - Wile Ops Autorisation 1.444 k Disksbare Response Time Cavad Chanames	Applearien Apdex Score Cat Mengarent 1 rum Applearien Response Tree Preside Cases	End User Apdex Score Carl Mangerine 0.93 miles f With Successful Checkunit Materice Bins	End User Requests Par M., Metal: Calang 1.66 rans Request Count Can Mangament	50 To
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A KPIS 0 4  4  4  4  4  4  4  4  4  4  4  4  4	Performence Age Marchellerere CPU Uttransence CPU Uttransence CPU Uttransence	Mobile App Values Extract loss 226 CPU (Distance S Antomatics 51.3	Guelly Manadation of 40 Web Checkaul Success Role Manadations 100 *	Total Acceptable Units United Acceptable Units Based of the Control of the Contro	Disk 10 - Wite Oos Admost an 1.44 k Disk demonstrate Case Demonstrate 197 MMMMMMMMMMMMMMMMMMMMMMMMMMMMMMMMMMMM	Application Apdex Score Cat Mangarett 1 num Application Response Time Predict Cases 2441 mm	End User Applex Score Call Mangement 0.93 num Web Successful Checkune Mathemate Data	End User Requests Per M., metal: Carety 1.66 row Pequest Count Care Recogneses 4	50 To
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### **Service Tree**

- The Service Analyzer tree view provides a visual representation of our services and the dependencies between them. Using this view you will also see the KPIs, entities, and most critical notable events associated with a service.
- The health of a service is affected by the health of a child service
- The tree can be built manually, however typically this is imported from a CMDB or a search



### **Service Tree**

We can see that there is an issue with the Manufacturing service.

- Click on the effected service (*Manufacturing*) to investigate which KPIs have degraded.
- Review which entities are effected.



During the service decomposition workshop we identified a missing 'On-Prem Database' service. In this lab we are going to create this new service, we will review the ITSI DB module and then select a service template.

- To create a service you can use (Technical) KPI's that comes with modules:
  - OS: Linux, Unix & Windows
  - Web server: Apache & Microsoft IIS
  - Application server: Tomcat & Websphere
  - Database: Microsoft SQL & Oracle
  - Storage: Netapp ONTAP & EMC VNX
  - Load Balancer: F5 Big IP & Netscaler
  - Virtualisation: VMWare & Hyper=V



Select the Configure menu + Services

splunk>enterprise App: IT Service Intelligence -									
Service Analyzer 🔻	Episode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards 🕶	Search 🔻	Configure	Product Tour	
Service Ana	lyzer 🖊				<	Services			
						Entities			
Filter Services Sele	ect service(s) to monit	tor				Service Tem	plates		

Click Create Service > Create Service





- Name the service as 'On-Prem Database'
- Check '**Database**' in modules list
  - Review the selected KPIs
  - Do not hit the 'Create' button!
- ITSI suggests best KPIs for database monitoring

Create Service	×
Title	On-Prem Database
Description	optional
Team ?	Global 🔻
<ul> <li>Link service to a service</li> <li>Add prebuilt KPIs from</li> </ul>	e template modules
Application Monitoring	Service for Database Instances that provides KPIs for
Application Servers	monitoring popular Database technologies.
✓ Databases	5/7 KPIs selected
End User Experience M	lanag Database Connection Pool Used %
Load Balancers	Database Deadlock Rate
	Database Query Response Time

Instead we are going to utilize the ITSI service templates feature, this will build the service with predefined KPIs.

- Select '*Link service to a service template*' button
- Choose 'On-Prem Database' template
- Click the 'Enable 7 days of backfill for all Service KPIs' option
  - Please note the option is hidden below so you will need to scroll down!

Click 'Create' button



The new 'On-Prem Database' service is based on a template, if you review the 'Entities' tab we can see that the entities are already filtered.

On-Prem Database   Service description				<b>≙</b> On-Prem	Database 🛽
Entitles KPIs Service Dependencies Settings Predictive An	alytics				
Entity Rules allow for the optional, dynamic filtering of KPIs and can help in ro	oot cause analysis. A service need not	define any Entity Rules and is not limited to o	nly the entities matchi	ng Entity Rules.	
Info • x itsi_role	matches • (× database_	instance	×		
Alias - x database_instance	matches 👻 🗙 mysql*		×		
+ Add Rule (AND)					
Info 🔻 🛛 🗶 itsi_role	matches - SAI		×		
Alias • x host	matches 🕶 🛛 🗙 mysql*		×		
+ Add Rule (AND)					
+ Add Set of Rules (OR)					
Matched Entities					
4 Entities					10 per page 🔻
fitle *	Aliases		Info		
mysql-01	10.2.2.1, mysql-01		linux, database, 2.6.3 eyjob3n0ijogim15c3f	32-573.8.1.el6.x86_64, db, mysql-01, sai, isltaxiiwgimlwijogijewljiumi4xin0=	
mysql-02	10.2.2.2, mysql-02		linux, database, 2.6.3 eyjob3n0ijogim15c3f	32-573.8.1.el6.x86_64, db, mysql-02, sai isltayiiwgimlwijogijewljiumi4yin0=	le ()
mysql-03	10.2.2.3, mysql-03		linux, database, 2.6.3	82-573.8.1.el6.x86_64, db, mysql-03, sai	
$\smile$					spiun

Under the KPIs tab we can see some KPIs that have been inherited from the service template, the padlocks indicate that changes to the template will be pushed to all linked services.

Click 'Database Queries' KPIs to review.

On-Prem Database 🗸		On-Prem Database 🛽
Service description 🥒		
Entities KPIs Service Depende	dencies Settings Predictive Analytics	
KPIs Clone New •	Database Queries /	
Database Queri	KPI description 🖋	
	Search and Calculate	
Disk I/O - Read Ops 🔒	> Thresholding	
Disk I/O- Write Ops 🔒	> Anomaly Detection	
Disk Space Used % 🔒		
Memory Used %		
Network Throughput - Inbound 🔒		
Network Throughput - Outbo 🔒		

The 'Settings' tab enables configuration of the service attributes. The new (linked) database service is disabled by default.

- Switch to Setting tab
- Toggle status to 'Enable'
- Investigate the effect changing the Importance and Simulated Severity has, on the Simulated Health Score
- Please do NOT enable Service Health Score backfill at this point

eneral			
Shitus C Enabled			
View Details (2			
A Changing a service's team may break service dependencies. Ask your Splunk administrate	or to review the logs I2 after saving the change.		
Service Health Score			
Backfar a dackfill			
fealth Score calculation			
weak the importance of each of the KPIs and see how that reflects on the simulated Service Health Sc	ore.		
			Simulated Health Score 7
Dn-Prem Databse KPIs			New york to second
			and the second
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KPI Title	Simulated Severity	Importance	Con Prom Database
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The new Database service will be a dependency of the **Authorization** service, any service health changes will be propagated to the parent service(s).

- Select *Configure* > *Services*
- Edit the 'Authorization' service

Service Analyz	er • Episode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards 🔻	Search <del>•</del>	Configure 🔻	Product Tour			( <del>)</del>	T Service Intelligence
Services	E.										Explore Content	Create Service -
A service is a c	collection of KPIs and entitie	es that represent a	real-world IT serv	/ice.								
11 Services	Bulk Action 🔻 🛛 filter		Q,									
<b>i</b>	Name 📩			Actions	Status		Service Te	emplate	Entity Rules	KPIs	Health	Team
	Authorization			Edit -	C Enabled	i .	Not linked	1	4	14	View Health	Global
>	Buttercup Store		Edit	>^	C Enabled	i	Not linked	1	0	21	View Health	Global
	Cart Management		Edit Tean	n	C Enabled	1	Synced w	ith Cloud-Based Services	4	20	View Health	Global

#### Click 'Service Dependencies' tab

Service Analyzer 🔻 🛛 Ep	bisode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards 🔻	Search 🕶	Configure 🕶	Product Tour	IT Service Intelligence
Authorization									
ervice description /	vice Dependencie	s Settings	Predictive An	alytics					
				Se.					
Remove selected depe	endencies								Add dependencies
Title								Service	
ServiceHealthSc	ore							External Authentication Services	la

- **Tick 'On-Prem Database' service**
- Tick the 'ServiceHealthScore'
- Press 'Done' button
- Press 'Save' button

#### Add dependencies

 $\checkmark$ 

Network Services	KPI Title	Service Title
Networking	ServiceHealthScore	On-Prem Database
NTP	Database Queries	On-Prem Database
On-Prem Database		
Order Management	Disk I/O - Read Ops	On-Prem Database
Product Catalog	Disk I/O- Write Ops	On-Prem Database
Shared Database Environment	Disk Space Used %	On-Prem Database
Shared IT Infrastructure	Memory Used %	On-Prem Database
Shared Storage	Network Throughput - Inbound	On-Prem Database
Shared Storage - Arrays	Network Throughput - Outbound	On-Prem Database
Shared Storage - LUNs		
Shared Storage - Volumes		

SMTP - DM7 In

×

### **Check Service Tree**

- Next go to Service Analyzer, and select the service tree
- Check if the On-Prem database is added





## **KPI Exercise**

Splunk > listen to your data

### **KPI** Lab

The new 'On-Prem Database' service is based on a template however we need to add an extra ad-hoc KPI to monitor the CPU utilization.

- Select Configure > Services
- Select the 'On-Prem Database'
- Select KPI tab
- Click New > Generic KPI
- Set Title to 'CPU Utilization %'

on	×
CPU Utilization %	
optional	
i	CPU Utilization %

Click 'Next' button


The new KPI source could be driven by a data model, ad-hoc search or a base search. It is always best to utilize base searches as they can return multiple KPI metrics with a single search.

- Click 'Base Search'
- Select 'All Metrics'
- Select 'Average CPU (Linux)'

CPU Utilization % Step 2 of 7: Source				×
KPI Source ?	Data Model	Metrics Search	Ad hoc Search	Base Search
Base Search ?	All Metrics •			
Metric ?	Edit Base Search 🖄	(x) •		



> Generated Search

There is no option to split this base search.

Step 3 of 7: Entities

Fields are populated from the selected base search.

Split by Entity ?

Yes

No

Entity Split Field ?

No

Filter to Entities in Service ?

Service must have entities to filter by entities.

Entity Filter Field ?

#### Click 'Next' button

> Generated Search

**CPU Utilization %** 



×

We need to build the KPI calculation criteria, this includes how often data is collected, the entity calculation and the calculation window.

- We have decided that the metric will be collected every 5 minutes
- We want to perform an Average of the metric
- Over 5 min window

CPU Utilization % tep 4 of 7: Calculation							
acted base search.							
Every 5 minutes							
Average 💌							
Average 💌							
Last 5 minutes >							
Null values 💌							
Unknown 🔻							
	ected base search. Every 5 minutes Average Average Last 5 minutes Null values Unknown						

#### Click 'Next' button

> Generated Search

The monitoring and unit fields will be populated from the base search.





> Generated Search

Next

We want this KPI to use data already ingested in Splunk over the last 7 days.

- Click 'Enable Backfill' button
- We will leave the backfill period as 7 days

CPU Utilization % Step 6 of 7: Optional Setup - Backfill Backfill? Backfill with last 7 days v of data



> Generated Search

We need to set some static thresholds for this new KPI

- Increase time to 4 hours
- Add & configure threshold:
  - Critical = 95
  - High = 90
  - Medium = 85
  - Low = 70

gregate Thre	sholds	Per-Entity	/ Thresh	olds			
gregate Th	reshold	Values					
Critical 🔻	95		× \	iew data from	last 60 minu	tes 🔻	
High 🔻	90		×				
Medium 🔻	85		×				
Low •	70		×				 



> Generated Search
Cancel Back Next Finish

- Note that the new KPI does not have a padlock icon. Inherited KPIs are locked to the service template so when changes are made these are pushed to the linked services, such as the one we are configuring.
- If you edit a locked KPI it will become an orphan and template changes no longer adopted
- Click 'Save' button! Bottom Right





Cancel

An important dependency for the new 'On-Prem Database' service is response time, in this lab we will add an extra KPI: 'Database Response Time'. For this KPI we use an ad-hoc search.

- Click Configure > Services
- Click 'On-Prem Database' & KPI tab
- Click New KPI > Generic KPI
- Title = 'Database Response Time'





The new KPI source could be driven by a data model, ad-hoc search or a base search. In this instance we will create the KPI using an ad-hoc search.

- Click 'Ad-hoc Search'
- Enter the following search:
  - index=itsidemo sourcetype=stream:mysql query=\*
- Enter 'time\_taken' as the threshold field
- Click 'Run Search' button to test search'

	 Metrics Search	Ad hoc Search	Base Searc
S	 12 82 74		

Click 'Next' button



We will split this KPI via the database name and filter via ip address.

- Select 'Yes' for split by entity
- Enter 'dbname' as the split by field
- Select 'Yes' for filter to entities in Service
- Type 'ip' for entity filter field

Split by Entity ?	Yes	No
Entity Split Field ?	dbname	
Filter to Entities in Service ?	Yes	No

×

**Database Response Time** 





The database response time KPI will have the following calculation options.



Explanation of Calculation:

**Database Response Time** 

Every minute take the average of time\_taken for each entity as the entity value then take the average of all entity values as the service/aggregate value all over the last 5 minutes. Fill gaps in data with Null values and use a unknown threshold level for them.

Cancel

Back

> Generated Search

X

- Schedule = Every minute
- Entity Calculation = Average
- Aggregate Calculation = Average
- Calculation Window = 5 minutes

Click Next

We will leave the next screen with the default values.

Database Response Step 5 of 7: Optional Setup -	2 Time Unit and Monitoring Lag	×
Unit	Specify the unit of measurement to display in KPI visualizations. (For example "GB," "Mbps," "secs", etc.).	
Monitoring Lag (in seconds) <sup>?</sup>	30 Determine Recommended Lag 🗠	



> Generated Search

We want this KPI to use data already ingested in Splunk over the last 7 days. This historical data will be used in the machine learning labs.

- Click 'Enable Backfill' button
- We will leave the backfill period as 7 days

Datab Step 6 c	of 7: Optiona	al Setup -	Time Backfill				
	Backfill?		Backfill with	last 7 days 🔻	of data		

X

Click 'Next' button



We need to set some static thresholds for this new KPI

- Increase time to 4 hours
- Add & configure threshold:
  - High = 360
  - Medium = 310
  - Low = 200

- Click '*Finish*' button
- Do not forget to click 'Save' button!!



> Generated Search

We have two new KPIs for our On-Prem Database service, this new service can utilize past data via the Service Health Score backfill capability.

Switch to the '*Settings*' tab

**Toggle backfill on (last 7 days)** 



On-Prem Databse 🖌				🔒 On-Prem Database 🛽
Service description 🖌				
Entities KPIs Service Dependencies Settings Predictive Analytics				
General				
Status Chabled				
Team ? Global * View Details 🗈				
A Changing a service's team may break service dependencies. Ask your Splunk administrator to review the logs 🛙	after saving the change.			
Service Health Score				
Backfull Backfull With last 7 days - of data				
It is advised that you first backfill the KPIs in this service and all dependent services for at least the time range Enabling backfill for a KPI does not mean backfill has completed. Wait for a successful backfill completion mes	selected here. sage for all KPIs before be	ckfilling the service health score.		
Health Score Calculation				
Tweak the importance of each of the KPIs and see how that reflects on the simulated Service Health Score.				
On-Prem Databse KPIs			Simulated Health Score ?	
			On-Prem Databse	
KPI Title	Simulated Severity	Importance	100	
CPU Utilization %	Normal •	0 1 2 3 4 5 6 7 8 9 10 11	100	
Database Queries	Normal -	0 1 2 3 4 5 6 7 8 9 10 11		
Database Response Time	Normal -	0 1 2 3 4 5 6 7 8 9 10 11		
Disk I/O - Read Ops	Normal -	0 1 2 3 4 5 6 7 8 9 10 11		

# Infrastructure Monitoring



There are now two new KPIs visible in the Service Analyzer for the On-Prem Database service, these KPIs contain historical data and we can see there is an issue with the CPU data.

- Navigate back to Service Analyzer
- Click 'On-Prem Database' service
- Select the new 'CPU Utilization %' KPI
- Click 'mysql-02' entity



The 'Entities Details' view is a high-level dashboard showing metric data for the selected entity.

Click the Splunk App for Infrastructure for more metric information.

mysql-02										Last 12 hours 🔻	Edit Entity •
host mysql-02 ip 10.2.2.2				itsi_role S name n	Al Iysql-02		os Show 4 more	Linux			
Modules	Overall Health	i.									
Splunk App for	KPIs						Services				
Infrastructure 🖸	Severity \$	KPI \$		Service \$	Sparkline	Alert Value	Severity \$	Service \$	Sparkline		Alert Value \$
	• Critical	Memory Free:	МВ	Database	mmm	905	• Critical	Database_Original			20.0
	• Critical	Memory Free:	MB	Database_Original		905	e High	Database		M	34.0
	Normal	CPU Utilizat	ion: %	Database	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	28					
	Normal	Database Err	ors	Database		25					
	Normal	Database KPI	6	Database	www.www	e					
	Normal	Storage Free	Space: %	Database	mmm	e					
	Normal	CPU Utilizat	ion: %	Database_Original	^	28					
	Notable Ever	nts									
	Entity \$		Notificatio	n ¢					When ¢		
	mysql-02		Nagios Se	rvice Check: check_nt;	o_time on mysql-02				2019-07-23T10:20:062	z	
	mysql-02		Nagios Se	rvice Check: check_dho	cp on mysql-02				2019-07-23T10:20:062	z	
	1.00				1.00					<u>.</u>	

The Splunk app for Infrastructure is a workspace to quickly drag and drop metric values.

Click the 'analysis' tab to investigate the collected metrics for this entity.



Challenge – If there is time attempt to recreating the following dashboard.

ata 🕊	🕁 🕐 Last 24	hours 🔹 🗘 Refre	esh 🔻 (6m ago)				C <sub>x</sub> Cl	ear all	Analysis
Find Data to Analyze Q	df.used								✓ disk.io_time.io_time Avg by host
Metrics > AWS/EC2	75 50 25	as and the second	and the second s	Kan-active atos			ww.	auth-01 auth-02 auth-03 mysql-01 mysql-02	Aggregation Select or type an aggregation.
> df > disk	0	16:00 Wed, 8 Apr 2020	20:00	00:00 Thu, 9 Apr	04:00	08:00	12:00		Split By Split this series into separate values.
<ul> <li>interface</li> <li>memory</li> <li>y uptime.uptime</li> </ul>	mysqld 150 125							Count	Highest • 5 •
Events Alerts	100 75 50 25								Display an earlier time range. None
	0	16:00 Wed, 8 Apr 2020	20:00	00:00 Thu, 9 Apr	04:00	08:00	12:00		Include or exclude values from specific dimensions. + Add New Filter
	disk.io_time.io_t 12.5k 10.0k 7.5k 5.0k			Ann Murry	ma lanha	Manal M	Man Man	auth-02 mysql-01 mysql-02 mysql-03 mysql-04	> Chart Settings

Navigate back to ITSI app for remaining labs.





# **Machine Learning**



# **Skill Areas for Machine Learning at Splunk**

**Splunk Expertise** Searching Reporting Alerting **Workflow** MLTK **Data Science** Splunk ML Toolkit ITSI facilitates and Expertise **Premium solutions** simplifies via examples & provide out of the Statistics/maths background guidance box ML capabilities. **Algorithm selection** Model building **Domain Expertise** (IT, Security...) **Identify use cases Drive decisions Understanding of business impact** 



# **ITSI Machine Learning**

Anomaly Detection







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**Predictions Analytics** 

Event	Clustering
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Deviation from past behavior Deviation from peers Unusual change in features ITSI Anomaly Detection

#### **Adaptive Thresholds**

What is normal behaviour and what is not normal Ideal for cyclical and dynamic data

#### Predict Service Health Score Predicting events Trend forecasting Early warning of failure Predictive maintenance

Identify peer groups Event correlation Reduce alert noise ITSI Event Analytics



# **Machine Learning process**

Ecosystem

MLTK

Splunk's App Ecosystem contains 1000's of free add-ons for getting data in, applying structure and visualizing your data giving you faster time to value.

The Machine Learning Toolkit delivers new SPL commands, custom visualizations, assistants, and examples to explore a variety of ml concepts.

Splunk

Splunk Enterprise is the mission-critical platform for indexing, searching, analyzing, alerting and visualizing machine data.

#### **Operationalized Data Science**



splunk > turn data into doing

# What Data Scientists Really Do

Data Preparation accounts for about 80% of the work of data scientists



#### What data scientists spend the most time doing

- Building training sets: 3%
- Cleaning and organizing data: 60%
- Collecting data sets; 19%
- Mining data for patterns: 9%
- Refining algorithms: 4%
- Other: 5%

"Cleaning Big Data: Most Time-Consuming, Least Enjoyable Data Science Task, Survey Says", Forbes Mar 23, 2016 splunk > turn data into doing"

# What Data Scientists Really Do

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- Refining algorithms: 4%
- Other: 5%

"Cleaning Big Data: Most Time-Consuming, Least Enjoyable Data Science Task, Survey Says", Forbes Mar 23, 2016

### Machine Learning Use Case

- There have been lots of social media comments regarding Buttercup' website availability, especially during the evenings.
- Luckily the website data is being ingested by Splunk, this sourcetype is called 'access\_combined' and contains lots of information.
- The objective of this exercise to is utilise Splunk ITSI machine learning capabilities to identify normal behaviour for web response time.
- We will also look to see if the application servers supporting the website are functioning equally.

The Corporate Website Request KPI has no thresholding for alerting. In this lab we will configure machine learning to understand what normal looks like and alert us when the KPI falls outside this range.

- Go to the '*default*' service analyzer view and expand the tree view
- Click 'Web Front End' service
- Select the 'Web Front End' link



We need to instruct ITSI to use a template for thresholding, we will be using the adaptive standard deviation 3 hour working week template.

Select the '*Corporate Website Requests*' KPI

Web Front End	
Entities KPIs Service Depen	dencies Settings Predictive Analytics
KPIs Clone New •	Corporate Website Requests 🗸
4xx Errors	KPI description 🥒
5xx Errors	<ul> <li>&gt; Search and Calculate</li> <li>&gt; Thresholding</li> </ul>
Bytes in	Use Thresholding Template Select a thresholding te •
Bytes out	Set Custom Thresholds
Corporate Website Requests	Enable Time Policies? Enable Adaptive Thresholding?
CPU Utilization %	

- Select 'use Thresholding Template' button
- Review the different options
- Select '3-hour blocks work week'
  - (adaptive/stdev)

### Click 'Apply' button

Apply Threshold Template

All existing threshold settings will be discarded. Are you sure you want to apply template?

Cancel Apply

×



The built-in machine learning has configured thresholds, this is broken into 1-hour time ranges. However we want to use historical data to apply some adaptive thresholding.

- Open the 'Configuration Thresholds for Time Policies' box
- Review different times
  - Choose 'Weekdays, 9am-12am'
- Click 'Apply Adaptive Thresholding' button
  - Wait 30 Seconds
- Notice threshold
- Click 'Save' button



Internal SLAs do not apply on weekends, tracking is still necessary, but management wants deliberately higher values.

We will now modify the new adaptive thresholds to increase the weekend ranges, this will result in a custom template.

- Click on '*Thresholding*' arrow
- Review the preview window
  - Note weekend range
- Click 'Set Custom Threshold'



#### Preview Aggregate Thresholds



We will now modify the new adaptive thresholds to increase the weekend ranges, this will result in a custom template.

- Expand 'Configure Thresholds for Time Policies'
- Click 'Weekends'
- Move all sliders as show
- Click 'Apply Adaptive Thresholding'

Weekdays, 7 AM	Policy type <sup>7</sup>					
	Standard deviation ~					
Veekdays, 7 PM	Thresholds are computed from data. Parameter associated with the labels is the number of standard deviations away from the mean. A value of 0 would equal the mean of the data. 1 would be a single standard deviation away from the mean					
Weekdays, 8 AM	View data for Saturday: 23:00 - 23:59 V					
Veekdays, 8 PM	■Critical → 2 σ ×					
Veekdays, 9 AM	$\blacksquare Medium \sim 1 \sigma \times$					
Weekdays, 9 PM	Normal V -1 $\sigma$ X					
Neekends	■Medium ~ -2 σ ×					

The preview aggregate threshold window now shows that during the weekend we will not receive an warning alert for the website response time KPI.



Preview Aggregate Thresholds

#### Click 'Save' button



# **Deep Dive**


#### **Deep Dive Use Case**

- Typically when organisations have outages they create a war room to identify the root cause as quickly as possible, this involves bringing together many business & technical stake holders at great expense.
- The deep dive capabilities within ITSI brings together multiple data sources into a single visualization. The correlation of data streams enable quick identification of root cause and effect on the business.
- In this lab we will build a deep dive visualization for the new On-Prem Database service, this will bring business and technical KPIs together with raw event data.
- Extra Once this lab is completed review the comparisons options.



- Navigate to the 'Default Service Analyzer' view
- **Toggle to 'Tree View' mode**
- Click 'On-Prem Database' service
- Click 'Open all in deep dive'



This deep dive view is used to bring all the relevant data to run an efficient war room, we can add/remove swim lanes to make the visualization even more useful.

- Select the three swim lanes
- Bulk Actions > Delete



To understand the impact we need to add some business KPI to this deep dive, this will speed up investigations and diagnosis.

- Either navigate up the service tree or change focus to 'Buttercup Store'
- Click + on the following KPIs;
  - Revenue
  - Revenue per Order
  - Successful Checkouts
- Move lanes as per image



To enable investigation into anomalous activity in your KPIs we can drill down on KPIs to gain deeper insights.

- Select 'Disk Space used %'
- Select the COG icon next to Disk Space used %
- Select Lane Overlay options
- Select Enable Overlays 'Yes'



mysql-01	×
mysql-02	×
mysql-03	×



Hover over one of the lines from 'Disk Space Use %'. Notice the drill down possibility to the Splunk App for Infrastructure





- Select 'Add Lane'
- Click 'Add Event Lane'



**Event Search**:

Click 'Create Lane'

'sourcetype="stream:mysql" status>200'

10 AM	Add Metr	ric Lane Lane		10:3
	Add Ever	nt Lane		
Add Event Lane			×	
Title	Database Errors			
Subtitle	optional			
Graph Color	Automatic 👻			
		h f a dlama	Large	



#### Extra

- Investigate blue event lane
- Compare to yesterday



Note: adjust the Time-Picker to only show the last 60 minutes.



Once you have finished investigating the Deep Dive dashboard

- make sure you click 'Save As'
- And save 'Database Deep Dive <Your Initials>'



# **Glass Tables**

Splunk > listen to your data

#### When to use Glass Tables

- Executive overview and business metrics
- Highly complex and valuable services
- Services that fail often or result in War Rooms
- Services that have recurring outages
- Main use case to stream-line root cause investigation
- When you want to link to Splunk Enterprise or other tools

Note : It is best practice to import a background with most of your graphical design, and then drag KPIs onto the canvas.





splunk >

100 -

50 **X** 

100 >

100 ->

100 -

#### **Glass Table Use Case**

- The business leaders would like a high-level dashboard showing the key functions and services of the organisation.
- The objective of this exercise is to complete the existing IT operations dashboard (Digital Transaction Flow) with the new On-Prem Database service healthscore, including a drill down to a deep dive.
- Second part of this lab is to create a business focus dashboard;
  - Import your group logo
  - Include several KPIs that only relate to the business
  - Keep technical KPIs to a minimum
  - Link existing dashboard to one of the services



#### **Glass Table Lab**

- Select *Glass Table menu*
- Edit '4. Digital Transaction Flow'

lass	s Tab	Bulk Action •	All	Арр	Private	filter		
	i	Title *			A	Actions		
	>	1 - Executive Dashboa	ard		Ec	lit 🕶		
	>	2 - Digital Channels			Ec	Edit 🝷		
	>	3 - Operational Status			Ec	Edit 💌		
	>	4 - Digital Process Flo	ow with Pr	redictive	Ec	dit 💌		
					Edit			
					Edit Title or	r Description		
					Edit Permis	sions		
					Clone			
					Delete			

Saved Glass Tables

- Click 💽 icon to fit to page
- Note the tool pallet icons
- KPIs can be dragged onto canvas
- Configuration panel on right



#### **Glass Table Lab**

- Expand 'On-Prem Database'
- Drag 'ServiceHealthScore onto the canvas
  - Configure attributes;
    - Label Box = Off
      - Scroll down the dialog box
    - Drilldown = On
    - Select 'Saved Deep Dive'
      - Select 'On-Prem Database DD'
    - Change visualisation type
      - Single Value
  - Click Update



#### **Glass Table Lab**

Move Database healthscore next to the Database icon (green box)

- On the right-hand panel
- Configure attributes;
  - Width = 180
  - Height = 180
- Click UpdateClick Save



#### **Glass Table Lab (Extra)**

Now it's your turn...you have 10 minutes

- Google and import your company logo
- Drop onto canvas;
  - Couple of icons
  - Link them
  - Business health scores
- Rename to reflect your business, be creative!



# Machine Learning: Predictive Analytics

(or Imminent Outage Prediction)



# What's Imminent Outage Prediction?

- Using historical data from KPIs for a service and some clever ML algorithms, you can sometimes predict an outage 20-30 minutes before it happens!
- Works best when a service has 5+ good KPIs and 1+ week of historical data

- The algorithm looks for recognizable/predictable KPI behavior, which comes before the service's aggregate health score changes.
  - For example: before the last outage, CPU usage went up AND garbage collection times increased AND session counts dropped...



#### Workshop Use Case

- The IT operations team are struggling to resolve issues with the company manufacturing service, typically outages are reported via customers contacting the service desk to complain.
- They would like to use machine learning to predict health score degradation 30 minutes before it causes a service outage.
- The objective of this lab will be to use the 'Manufacturing' health score to build a predictive algorithm model to predict future issues.



We need to use machine learning to build a model for the Manufacturing service. This model will be used in the second part of this lab.

- Select '*configure*' menu
- Pick 'Services' item
- Click edit '*Manufacturing*' service
- Select 'Edit' menu

Service	a Ana	lyzer 👻 Episode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards ▼ Search ▼	Configure -	Product Tour		💮 п	Service Intelligence
Serv	/ice	S a collection of KPIs and entitie	es that represent	a real-world IT se	rvice				Explore C	ontent [2	Create Service -
10 Serv	/ices	Bulk Action • filter		Q							
	i	Name *		Actions	Status	Service Template		Entity Rules	KPIs	Health	Team
	>	Authorization		Edit <del>-</del>	C Enabled	Not linked		4	14	View Heat	th Global
	>	Buttercup Store		Edit <del>•</del>	C Enabled	Not linked		0	21	View Heat	th Global
	>	Cart Management		Edit <del>v</del>	Enabled	Synced with Cloud-Based Servi	ces	4	20	View Heat	th Global
	>	Cloud Databases		Edit 🕶	C Enabled	Not linked		2	8	View Heal	th Global
	>	External Authentication Ser	vices	Edit <del>v</del>	C Enabled	Not linked		0	3	View Heat	th Global
	>	Manufacturing	C	Edit <del>•</del>	C Enabled	Not linked		1	7	View Heal	th Global
	>	Mobile App	Edit		Enabled	Not linked		0	3	View Heal	th Global
	>	Order Management	Edit Team		Enabled	Synced with Cloud-Based Servi	ces	3	20	View Heat	th Global
	>	Product Catalog	Link to Serv	rice Template	Enabled	Synced with Cloud-Based Servi	ces	з	20	View Heal	th Global
	>	Web Front End	Create Service	vice Template in Maintenance M	Enabled	Not linked		4	12	View Heal	th Global
			Clone Delete								

ITSI Predictive Analytics uses machine learning algorithms to predict the future health score of your service. This screen we will train and test different machine learning algorithms to determine which one will give the most accurate prediction.

- Select 'Predictive Analytics'
- Time = 14 Days
- Algorithm Type = Regression
- Algorithm = Linear Regression
- Click 'Train' button
- Once the model has run, investigate below.
- Click 'Save'



- Repeat the previous steps for the other two algorithms over 14-day period.
- Random Forest Regressor
- Click 'Train' button
- Remember 'Save' button !
- Gradient Boosting Regressor
- Click 'Train' button
- Remember 'Save' button !



In this lab we are going to review the predictive analytics value for the **Manufacturing** service using the recommended model.

- Select Dashboards > Predictive Analytics
- Select the 'Manufacturing' service
- Select the recommended algorithm model

#### **Predictive Analytics**

ITSI Predictive Analytics uses machine learning to predict the health score value of a selected service. The models use historical KPI and service health score data to approximate what a service's health might look like in 30 minutes. 🚯 Learn more



Once we have selected a model ITSI will calculate the future healthscore

Click 'Cause Analysis' button to review the suggested KPIs

rvice ⑦ Model ⑦	RoostingDograssor		
Gradiente	boosungkegressor *		
Predicted Service Health Sc	core in 30 Minutes	Q	Predicted Top 5 Contributing KPIs 👩
			KPI
			Total Acceptable Units
	50.00		Total Units Produced
50.06			Cycle Time
			Performance
			Overall Equipment Effectiveness
		Cause Analysis	Analyze in Deep Dive [견
KPI Predictions 🕜			
KPI	Accuracy of Prediction - R2 👩	Past Versus Predicted KPI Values	
Total Acceptable Units	High: 0.99		
Total Units Produced	Low: 0.10		
Cycle Time	Low: 0.10		

- Click the spyglass to review the SPL
- This SPL is already being used in our glass table

	Service 🕜	Model 🕐					
	Manufacturing -	GradientBoostingRegressor 🕶					
	Predicted Service	e Health Score in 30 Minutes			Predicted Top 5 Contributing KPIs 👩		
				$\overline{}$	KPI		
					Total Acceptable Units		
		=	~		Total Units Produced		
New Sea	rch						Save As  Close
`itsi_predict :itsi_pre	t_one_number(4bf1f edict_4bf1f146_3b8	146-3b89-4ae7-b8f3-32f536357bc4,h 9_4ae7_b8f3_32f536357bc4_Gradient	ealth_score,app BoostingRegressor_041e77dd3c5e000	bee811489_	1585923875029)`		Last 2 hours - Q
✓ 3,909 events	(03/04/2020 12:00:0	00.000 to 03/04/2020 14:55:42.000)	No Event Sampling -			🔺 Job 🔻 II 🔳 👌 👼 .	🛓 🕴 Smart Mode 👻
Events Patt	erns Statistics (	1) Visualization					
20 Per Page 🔻	✓ Format Pr	review 💌					
			health_score 🗘 🖌	_time ‡			
			48.70154986843101	2020-04-0	3 14:50:00		



- Select 'Glass Tables > Digital Transaction Flow with Predictive'
- Click 'Edit' button
- Review the 'Database Future Health-Score attributes'



# Machine Learning: Event Analytics



- The database team has expressed frustration with alerts originating from Nagios. The high volume of alerts is leading to alert fatigue and they lack the contextual information necessary to make them actionable.
- While the plan is to consolidate monitoring tools, they have asked if we can provide immediate relief using ITSI to group events together and reduce noise.
- Also, with hundreds of database instances to manage, tracking which alerts are associated to critical systems vs non-critical systems is tribal knowledge for the database team. They love the service tree view in ITSI and asked if it's possible to see the services affected by an alert.
- Interestingly enough, you heard the same challenges expressed by the web team who uses New Relic to monitor the application code.



Often, each team or layer of the stack has their own monitoring tools producing their own alerts. Each tool creating a silo of information that other teams don't get insight.



If we can cluster events by time, we can immediately reduce noise. The two database events in orange happened around the same time and therefore, may very well be related to the same incident



The three events in green occurred later after a pause in the flow of events. Probably a different issue all together. So grouping events together by time is a very powerful noise reduction technique.



We can extend the power of grouping beyond just time to further reduce noise. What if all of the events in orange we associated to the same database instance. Same timeframe... Same instance... that's probably all a related incident.



We can extend the grouping across siloed monitoring tools, and across layers of the stack. What if I told you that all the events in orange were associated with machines that run the Buttercup Store.





# A quick terminology check

#### Alert - Describes a state change for a target entity

- Examples of different alerts:: Server42 is down, Filesystem is full
- Alarm Specific alert for a target entity, can change severity/state over time
- Example of single alarm: Server42 is "down", then later "up"

#### **Time-series data or Events**- The stuff that Splunk indexes

• Includes traditional alerts & alarms, as well as logged data, metrics, wire data and more

#### Notable Event - An actionable message (Splunk ITSI & Splunk ES)

• Intended specifically for humans in Operations

#### **Episode** - a group of Notable Events

Incident - Unplanned interruption of an IT or business service

ServiceNow, Remedy



In this exercise we will group the database teams Nagios events together based on time and associate them to a service

Clean and prepare "raw" alert events • SPL

Create Notable Events from alerts

Correlation Search

Apply Service Context & Configure Event Grouping

Notable Event Aggregation Policies

Review episodes

	Time	Event
0	09/04/2020 11:49:52.652	2020-04-09 11:49:52.652510 src_host="mysql-02" ond_site="SJC" perfdata="SERVICEPERFDATA" name="check_disk" severity="OK" attempt="1" statetype="SOFT" executiontime="0.0" later ="0.0" reason="Disk Space utilization 51.18 is below threshold" results="Disk Space status ok" host= mysql-02   source = naglos   sourcetype = naglos
	09/04/2020 11:48:52.592	2020-04-09 11:48:52.592363 src_host="mysql-02" omd_site="5JC" perfdata="SERVICEPEMFDATA" name="check_disk" severity="WARNING" attempt="3" statetype="HARD" executiontime="0.0" tency="0.0" reason="Disk Space utilization 96.3 is above threshold" results="Disk Space status warn" host= mysql-02   source = naglos   sourcetype = naglos
	09/04/2020 11:47:52.527	2020-04-09 11:47:52.527797 src_host="mysql-02" ond_site="SJC" perfdata="SERVICEPEBFDATA" name="check_disk" severity="WARNING" attempt="1" statetype="SOFT" executiontime="0.0" tency="0.0" reason="Disk Space utilization 93.89 is above threshold" results="Disk Space status warn" host= mysql-02   source = naglos   sourcetype = naglos


## EA Lab Step 1 : Clean and prepare "raw" alert events



#### | index=itsidemo sourcetype=nagios perfdata=SERVICEPERFDATA

The get a the o	first step to help alerts in. Prior to extracted fields.	Agios Alerts       Predictive Analytics         ing the database team achieve their goal is to ingest the raw       ITSI Health Check       ovides supporting add-ons in Splunkbase for many monitoring tools which you can use to I         the workshop, we took the time to onboard the Nagios alerts       SAN ITSI - Adding Nagios Events into ITSI       ovides supporting add-ons in Splunkbase for many monitoring tools which you can use to I
SPL	ex=itsidemo sourc	type-magios perfdata-SERVICEPERFDATA
i	Time	Event
>	09/04/2020 11:49:52.652	2020-94-09 11:49:52.652510 src_host="mysql-02" omd_site="SJC" perfdata="SERVICEPERFDATA" name="check_disk" severity="OK" attempt="1" statetype="SOFT" executiontime="0.0" lat ="0.0" reason="Disk Space utilization 51.18 is below threshold" results="Disk Space status ok" host= mysql-02   source = naglos   sourcetype = naglos
>	09/04/2020 11:48:52.592	2020-04-09 11:48:52.592363 src_host="mysql-02" omd_site="SJC" perfdata="SERVICEPERFDATA" name="check_disk" severity="WARNING" attempt="3" statetype="HARD" executiontime="0.0 tency="0.0" reason="Disk Space utilization 96.3 is above threshold" results="Disk Space status warn" host= mysql-02   source = naglos   sourcetype = naglos
>	09/04/2020 11:47:52.527	2020-04-09 11:47:52.527797 src_host="mysql-02" omd_site="SJC" perfdata="SERVICEPERFDATA" name="check_disk" severity="WARNING" attempt="1" statetype="SOFT" executiontime="0.0 tency="0.0" reason="Disk Space utilization 93.89 is above threshold" results="Disk Space status warn" host= mysql-02 source = naglos sourcetype = naglos
		«Prev 1 2 3 4 5 6 7 8 9 10 Ne

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## EA Lab Step 2 : Add Normalized Fields



Goal: Normalize the data to make it possible to correlate data from different

SourceSrm\_severity=case(severity=="CRITICAL",6,severity=="WARNING",4, severity="OK", 2) | eval norm\_instance= src\_host | eval norm\_test=name

#### Step 2 - Add Normalized Fields

While each monitoring tool will express events differently, they all communicate the same fundamental information. Such as, how severe is the event? To which machine is the event associated? What type of check or test was performed? To facilitate the grouping of multiple events from multiple monitoring tools, we must normalize this key information so that a common set of field names and values is used. The SPL below creates these normalized severity, instance, and test fields.

SPL						
index=itsidemo sourcetype=nagios perfdata=SERVICEPE	RFDATA					
eval norm_severity=case(severity=="CRITICAL",6,seven	erity=="WARNING",4, severity="OK", 2)					
eval norm_instance=src_host						
eval norm_test=name						
_time ‡	name ‡	severity \$	norm_severity ‡	norm_instance \$	norm_test ‡	total 🗘
2020-04-15 10:02:27.773	check_disk	ОК	2	mysql-02	check_disk	1
2020-04-15 10:01:27.710	check_disk	WARNING	<b>4</b>	mysql-02	check_disk	1
2020-04-15 09:56:27.396	check_disk	ОК	2	mysql-02	check_disk	1
2020-04-15 09:55:27.334	check_disk	WARNING	4	mysql-02	check_disk	1
2020-04-15 09:51:27.088	check_disk	ОК	2	mysql-02	check_disk	1
			594			157
					« Prev 1 2 3 4 5 6	7 8 9 10 Next »

After you have finished reviewing the new normalized fields, click here to proceed to Step 3 - Deduplicate events

## EA Lab Step 3 : Deduplicate events



### Goal: Removing duplicated events

| dedup consecutive=true src\_host severity name

_time \$	name ‡	norm_instance \$	severity \$	norm_severity ‡	total 🗘
2020-04-15 10:02:27.773	check_disk	mysql-02	ок	2	1
2020-04-15 10:01:27.710	check_disk	mysql-02	WARNING	4	1
2020-04-15 09:56:27.396	check_disk	mysql-02	ок	2	1
2020-04-15 09:55:27.334	check_disk	mysql-02	WARNING	4	1
2020-04-15 09:51:27.088	check_disk	mysql-02	ок	2	1
				102	34
				« Prev 1 2 3	4 5 6 7 Next »

Finally, after you have finished reviewing the deduplicated events, click here to proceed to Step 4 - Add Service Context

## EA Lab Step 4 - Add Service Context



Goal: Add Service Context makes correlation possible for different alerts in a service [`apply\_entity\_lookup(host)` [`get\_service\_name(serviceid,service\_name)`

Click on the magnifying glass

_time \$	name \$	norm_instance \$	severity \$	norm_severity \$	service_name \$
2020-04-15 10:02:27.773	check_disk	mysql-02	ОК	2	On-Prem Database
2020-04-15 10:01:27.710	check_disk	mysql-02	WARNING	4	On-Prem Database
2020-04-15 09:56:27.396	check_disk	mysql-02	ОК	2	On-Prem Database
2020-04-15 09:55:27.334	check_disk	mysql-02	WARNING	4	On-Prem Database
2020-04-15 09:51:27.088	check_disk	mysql-02	ОК	2	On-Prem Database
					Prev     1     Open in Search 6     7     Next »

Q ¥

a minute ago

## **Event Analytics Lab**



### Select the SPL Query and copy to clipboard, we will need it in the next chapter



SDIUNK > turn data into doing

## **Event Analytics Lab**

In this exercise we will group the database teams Nagios events together based on time and associate them to a service



## **Event Analytics Lab**

- Navigate to Configure -> Correlation Searches
- Create New Search -> Create Correlation Search

Service Ana	alyzer 🕶 Episode Review Glass Tables Deep Dives Mul	ti-KPI Alerts Dashboards <del>-</del>	Search 🔻 🔍 Configure 💎 Product Tou	r IT Service Intelligence			
Correla	ation Searches		Services	Explore Content IZ Create New Search			
A correlatio	n search is a recurring search that generates a notable event when sear	ch results meet specific conditio	Entities				
Pulk Acti	filtor		Service Templates	( Prov. 1 2 Next )			
Bulk Activ			✓ Correlation Searches				
<b>i</b>	Title <sup>•</sup>	Actions	KPI Base Searches				
$\Box$ >	Active Directory	Edit 💌	KPI Threshold Templates	Enabled			
$\Box$ >	Bidirectional Ticketing	Edit 🝷	Backup/Restore	Disabled			
□ >	Episode Monitoring - All Services and KPIs Return to Normal	Edit 🔻	Maintenance Windows	Disabled			
$\Box$ >	Episode Monitoring - Concentration of High and Critical Notable Ev	Edit 🔻	Notable Event Aggregation Policies	Disabled			
	Episode Monitoring - Critical Notable Event added to Episode (Rec	Edit 🔻		Disabled			
□ >	Episode Monitoring - Episode Risk Well Above Historical Average	Edit 🔻		Disabled			

## **Event Analytics Lab: correlation search**

### Name your search

### **Paste the SPL in search:**

index=itsidemo sourcetype=nagios perfdata=SERVICEPERFDATA

| eval norm\_severity=case(severity=="CRITICAL",6,severity=="WARNING",4, severity="OK", 2)

| eval norm\_instance= src\_host

| eval norm\_test=name

| dedup consecutive=true src\_host severity name

| eval entity\_title=norm\_instance

|`apply\_entity\_lookup(host)`

| `get\_service\_name(serviceid,service\_name)`

| table \_time, name, norm\_instance, severity, norm\_severity, service\_name

- Time range: Last 5 minutes (select 'relative' in time picker)
- Run Every: 5 minutes
- Entity Lookup Field: host 
  to link with the Association Service Context
- Scroll down

Search Name	Nagios Correlation Search						
Description ?	optional						
Search Type	Data Model	Ad hoc					
Search	index=itsidemo sourcetype=nagio   eval norm_severity=case(severity=="C ,4, severity="OK", 2)   eval norm_instance= src_host   eval norm_test=name   dedup consecutive=true src_host   eval entity_title=norm_instance   `apply_entity_lookup(host)`  `get_service_name(serviceid,serv   table _time, name, norm_instance service_name	es perfdata=SERVICEPERFDATA RITICAL",6,severity=="WARNING" st severity name rice_name)` ce, severity, norm_severity,					
	Run Search 12	<i>h</i>					
Time range	Last 5 minutes 🕶						
ssociation							
Service	Select service(s)						
Entity Lookup Field?	host						

#### Schedule

Search Properties

## **Event Analytics Lab: correlation search**

Populate Notable Event Title
Nagios alert from %norm\_instance%
Populate Notable Event Desc.
Nagios alert from %norm\_instance%.
%norm\_test (%severity%)
Severity: Advanced Mode
Severity: %norm\_severity%

Notable Event Title ?**	Nagios alert from %norm_instance%					
Notable Event Description ?	Nagios alert from %norm_instance%. %norm_test (%severity%)					
Owner ?	unassigned  Advanced Mode					
	In advanced mode, use tokens like %fieldname% to use result field values to set owner.					
Severity ?	%norm_severity% Simple Mode					
	In advanced mode, use tokens like %fieldname% to use result field values to set severity.					
Status ?	New  Advanced Mode					
	In advanced mode, use tokens like %fieldname% to use result field values to set status.					
Drilldown Search Name ?						
Drilldown Search ?						
Drilldown earliest offset ?	Last 5 minutes 💌					
Drilldown latest offset ?	Next 5 minutes 💌					
Notable Event Identifier	source					
Fields ?	Set of fields used together to determine if a notable event is unique or not.					
Drilldown Website Name ?						
Drilldown Website URL ?						
> Advanced Options						

Notable Events

Save

Cancel Save

## Policies

In this exercise we will group the database teams Nagios events together based on time and associate them to a service

Clean and prepare "raw" alert events • SPL

Create Notable Events from alerts

Correlation Search

Apply Service Context & ConfigureEvent GroupingNotable Event Aggregation Policies

*Review episodes* 

	Action Rules			
Filtering Criter	ia			
Create filtering crite	ria to group notable	events into ep	isodes.	
✓ Include the event	s if <sup>?</sup>			
severity	greater than 🔻	Normal	×	
service_nan	matches 💌		×	
+ Add Rule (ANE	))			
+ Add Rule (OR)				
> Smart Mode group	bing			
✓ Split events by fie	ld?			
Split events into	multiple episodes b	y		
	×			
service_name				
service_name				
service_name				

> Episode information

## **Event Analytics Lab: Notable Event Aggregation** Policies

Navigate to Configure -> Notable Event Aggregation Policies

**Edit** *"Service Issues"* policy

ınk	>ent	terprise App: IT Service	e Intelligence 🔻							0	Administrator 🔫	50
ice	Analy	rzer▼ Episode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards 🕶	Search <del>•</del>	Configure -	Product Tour			
not	ible	Event Aggregat event aggregation polices to	ion Policie	S ble events in the	Episode Review.							
otab	le Ev	ent Aggregation Policies	Bulk Action *	filter								
]	i	Title *				Action	ns				Status	
]	>	Application Alerts				Edit 🕶					C Enabled	
	>	Default Policy				Edit 💌					Enabled	
	>	Default SNMP Policy				Edit *					Disabled	
]	>	Episodes by Alert Group				Edit 🕶					Disabled	
]	>	Episodes by ITSI Service				Edit 🕶					Disabled	
]	>	Infrastructure Alerts				Edit •					C Enabled	
	>	KPI Alerting Policy				Edit 🕶					C Enabled	
]	>	Multi-Episode Problem				Edit 🕶					Disabled	
	>	Normalized Policy (Splunk A	pp for Infrastructur	re)		Edit 🕶					C Enabled	
)	>	Service Issues				Edit	-				C Enabled	
)	>	User Account Management				Edit Edit Title or I Edit Permissi Clone Delete	Descriptic	n			Enabled	
								E.				

## **Event Analytics Lab: Notable Event Aggregation** Policies

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- Review "Include the events if" configuration
- Review "Split events by field" configuration
- Review "Break episode" configuration
- Modify the "Break episode" configuration
  - If the flow of events paused for 3600 seconds

#### Service Issues / Notable Event Aggregation Policy description 🥒 **Filtering Criteria** Action Rules Filtering Criteria Create filtering criteria to group notable events into episodes. ✓ Include the events if<sup>7</sup> × Normal severity greater than 💌 × service nan matches • + Add Rule (AND) + Add Rule (OR) > Smart Mode grouping ✓ Split events by field<sup>®</sup> Split events into multiple episodes by service\_name x ✓ Break episode If this episode existed for 36000 second(s) × × If the flow of events into the episode paused for 3600 second(s) + Add Breaking Condition (OR)

> Episode information

### **Event Analytics Lab: Notable Event Aggregation** Policies

You can see in the preview results how ITSI is now grouping events together based on your configurations.



**Click Cancel** 

Notable Event Aggregation Policy description 🥒

Filtering Criteria Action Rules

#### Filtering Criteria

Create filtering criteria to group notable events into episodes.

✓ Include the events if?				Preview with the Last 24 hours *								
severity greater than - Normal ×		i	Count \$	Title \$	Description \$	Severity \$	Owner ‡	Status \$				
		>	13	Web Front End issue	Possible issue with Web Front End or upstream service	😐 Medium	unassigned	New				
service_nan matches •		>	140	Web Front End issue	Possible issue with Web Front End or upstream service	😑 Medium	unassigned	New				
+ Add Rule (AND)			2	On-Prem Database issue	Possible issue with On-Prem Database or upstream service	e Low	unassigned	New				
+ Add Rule (OR)		>	50	Web Front End issue	Possible issue with Web Front End or upstream service	😐 Medium	unassigned	New				
> Smart Mode grouping				Authorization issue	Possible issue with Authorization or upstream service	😐 Medium	unassigned	New				
✓ Split events by field?		>	15	Authorization issue	Possible issue with Authorization or upstream service	Medium	unassigned	New				
Split events into multiple episodes by		>	2	On-Prem Database issue	Possible issue with On-Prem Database or upstream service	😐 Low	unassigned	New				
service_name ×		>	2	Authorization issue	Possible issue with Authorization or upstream service	🔴 Medium	unassigned	New				
✓ Break episode?		>	1	On-Prem Database issue	Possible issue with On-Prem Database or upstream service	• Critical	unassigned	New				
> If this episode existed for 36000 second(s)	×	>	61	Authorization issue	Possible issue with Authorization or upstream service	Medium	unassigned	New				
$\boldsymbol{\flat}$ If the flow of events into the episode paused for 3600 second(s)	×											
+ Add Breaking Condition (OR)												

> Episode information

## **Event Analytics Lab**

In this exercise we will group the database teams Nagios events together based on time and associate them to a service





## **Event Analytics Lab: Episode View**

When notable events are grouped by aggregation policies, the resulting groups are called episodes, you can think of an episode as an incident. The episode review page provides a great deal of information in a heads-up display and is like the cockpit view for Operations teams.

- Navigate to Episode Review
- Note the Noise Reduction

Service Analy	zer - Episode Review	Glass Tables	Deep Dives	Multi-KPI Alerts	Dashboards 🕶	Search 🕶	Configure 🕶	Product To	bur		IT Service Intellig	gence
Episod	le Review /										Save as Save	ø
53 episodes	Last 24 hours -	Add Filter 🔻	search								Hide summary dashbo	oard ☆
Me	ean Time to Resolve		Episodes by Se	everity	Total N	loise Reductio	on	Ep	isodes Acknowledged		Mean Time to Acknowledge	
	<b>C</b> 4 minutos		_			0464	1%		Acken dese	d -	NI/A@	
	54 <sup>minutes</sup>		3 6 2			94.04	r /0		Photo-odge	(A)	N/AU	
			3	39					Unecleg	ed,		
Format Tin	neline 🔻 🛛 – Zoom Out	+ Zoom to Sel	ection × Dese	lect							1 hour per c	column
200												200
100								-				100
		6:00 AM Tue Apr 7			12:00 PM				6:00 PM		12:00 AM Wed Apr 8	

## **Event Analytics Lab: Episode View**

- Scroll down and review list
- Modify filter as shown
  - Critical & High only

1 episode Last 24 hours -	Severity: Critica (2) - Ad	d Filter • search
IE Sorted by ? ↓ Time ▼	filter Q	0 4
On-Prem Database issue Owner: Unassigned	Select All Clear All 22	0 09:50:04 Description: Possible iss
	I High	
	Medium	
	Normal 🗸	
	6 of 6 values	

## **Event Analytics Lab: Episode View**

- We will now review an episode to better understand the flow of events, and we will then ensure someone has ownership.
- Click on 'Authorization Issue' episode
- Review the details for each tab
- Add your name to the comments

- Change to 'In Progress'
- Review possible Actions



## **Next Steps**

Somerford's Approach to ITSI





Somerford's Glass Table Methodology

## **Project Phases**

#### Initiation

- Kick-Off Call
- Pre-Workshop Planning
- ITSI Product Walkthrough

#### Analysis

- Service Identification Workshop
- Entity Strategy Workshop
- Service Decomposition Workshops

## Data Onboarding & ITSI Installation

- ITSI Installation
- Data Onboarding
- Implement Entity Management Strategy

#### Delivery

- Delivery Planning
- ITSI Services
   Development
- Progress Review
- ITSI Service Review
- ITSI UI
   Development

#### Handover

- Project Closeout
- Enablement
- BAU Transition
- Onboarding Users
- Follow-up Reviews







## Service Identification

- Which business and technology services are candidates?
- Which services shall we do first?





## Service Decomposition

- What are the business services to be monitored?
- What technology services make up those business services?
- What are the components of the technology services?
- What do you care about? What are the metrics?
- What are the KPIs?
- What data drives each KPI?



# Thank You!

Please get in touch with us if you have any questions.

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