splunk> .conf2017

# Using Splunk for Retail Banking Cross Channel Fraud Analysis, Detection and Investigation

Commercial Bank Of Dubai

Rory Blake | Splunk Professional Services Staff Consultant Rinaldo Ribeiro | Head of IT Risk & GRC, Risk Management

28th September 2017 | Washington, DC

### **Forward-Looking Statements**

During the course of this presentation, we may make forward-looking statements regarding future events or the expected performance of the company. We caution you that such statements reflect our current expectations and estimates based on factors currently known to us and that actual events or results could differ materially. For important factors that may cause actual results to differ from those contained in our forward-looking statements, please review our filings with the SEC.

The forward-looking statements made in this presentation are being made as of the time and date of its live presentation. If reviewed after its live presentation, this presentation may not contain current or accurate information. We do not assume any obligation to update any forward looking statements we may make. In addition, any information about our roadmap outlines our general product direction and is subject to change at any time without notice. It is for informational purposes only and shall not be incorporated into any contract or other commitment. Splunk undertakes no obligation either to develop the features or functionality described or to include any such feature or functionality in a future release.

Splunk, Splunk>, Listen to Your Data, The Engine for Machine Data, Splunk Cloud, Splunk Light and SPL are trademarks and registered trademarks of Splunk Inc. in the United States and other countries. All other brand names, product names, or trademarks belong to their respective owners. © 2017 Splunk Inc. All rights reserved.



### Agenda



### Agenda subtitle

- ► Introductions
- ► About CBD
- ► CBD & Splunk
- Use Case
- Splunk Architecture
- ▶ Data Models
- ► Investigation Dashboards
- ► Alerts
- What's Next



### Introductions

Section subtitle goes here



### Rory Blake

### Splunk Professional Services Staff Consultant

- ▶ UK Based Covering EMEA
- Nearly 6 Years Delivering Splunk PS
- ▶ 16 Countries
- ▶ 4 US States
- Product, Strategy & Technical Content
- Bat Signal Accounts
- ▶ 10 Years in Financial Services
- 2 Years Fraud Strategy





### **About CBD**

Section subtitle goes here



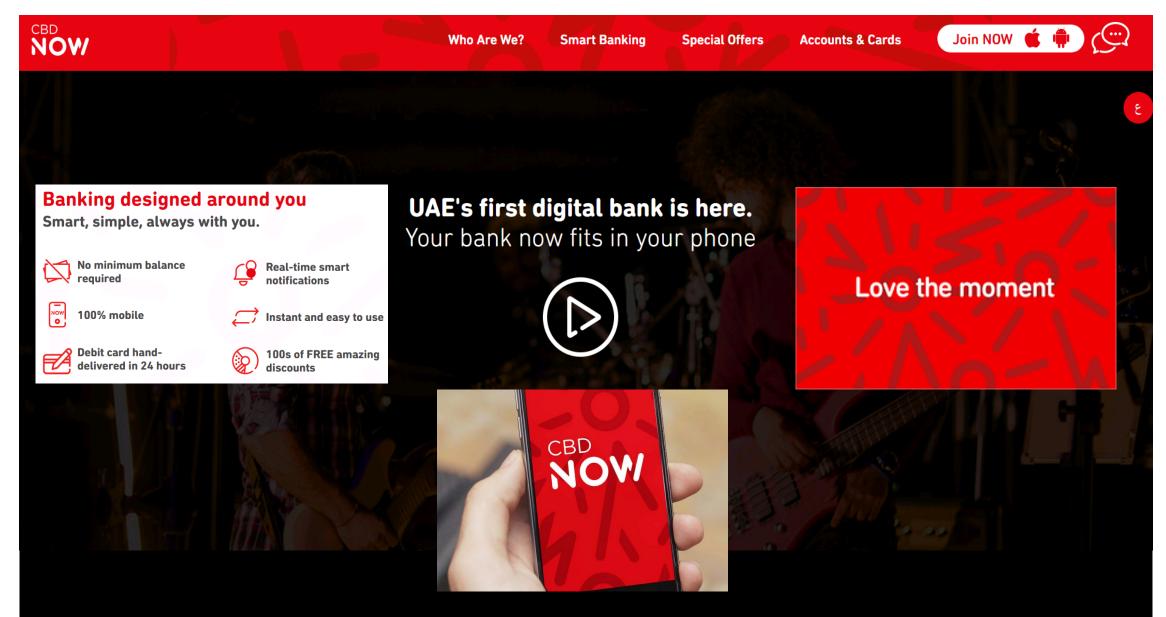




- ► Est. 1969
- ▶ 26 Branches
- ▶ 150 ATMs

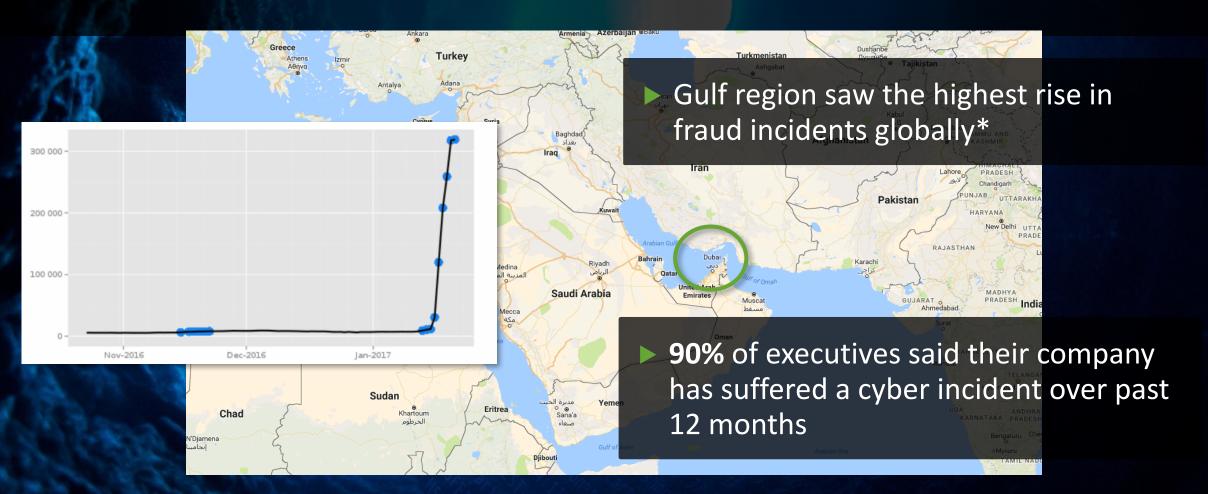
- Retail and commercial banking products
- ► Conventional and Islamic Banking
- ► Total assets: AED 71 billion (£ 15 billion)







#### Fraud In The Gulf



### **Platform Objectives**



Behavioural Analytics



Near Real Time Proactive Detection & Prevention



Entity & Group **Profiling** 



**Holistic Cross** Channel View



Anomaly Detection



### Splunk & CBD

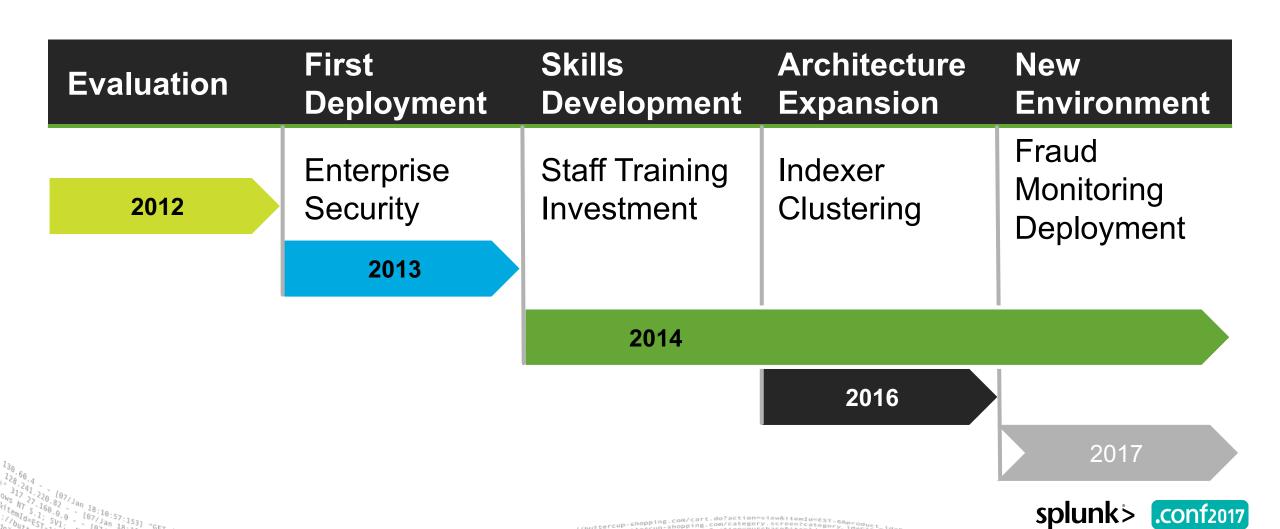
Section subtitle goes here





#### Splunk & CBD

**Journey Timeline** 



### Splunking Fraud At CBD

Section subtitle goes here



#### **Project Drivers**

- Constantly Evolving Threat Landscape
- Increase in Cyber Crime / Fraud Activity in the GCC and UAE

- No Existing Multi-Channel Fraud Detection Platform
- Cross Channel Fraud Detection Rules
- Near Real Time Proactive Responses
- Cumbersome Investigation Processes
- 360 Degree View of Cross Channel Customer Interactions



### **Banking Channels**







**Debit Cards** 



Digital



Telephone & SMS



**SWIFT** 





### **Risk Platform Integration**



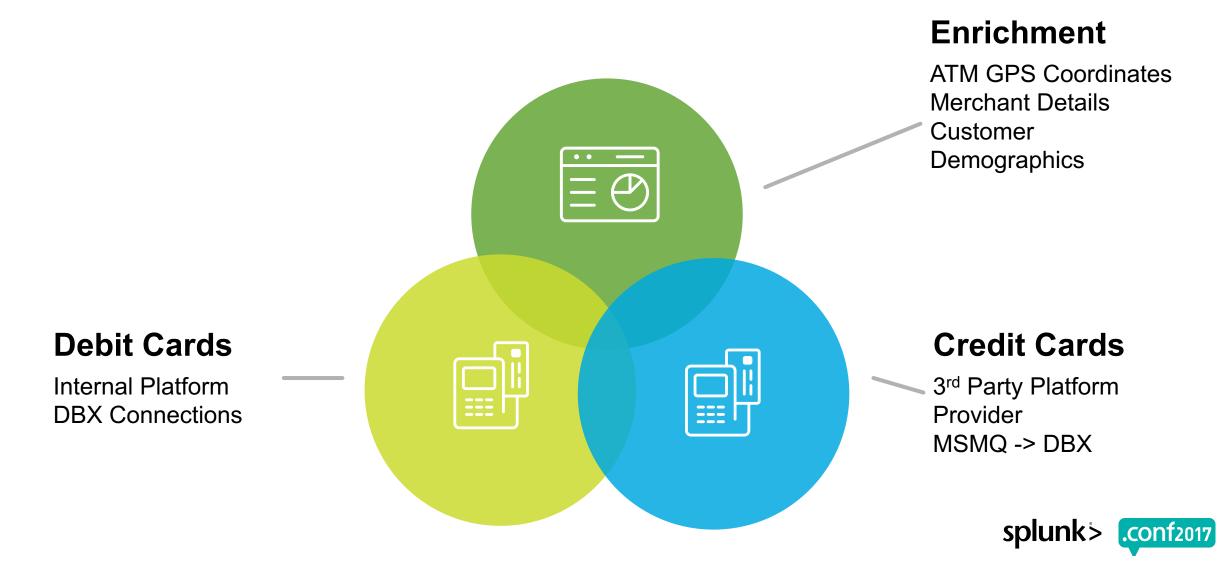
- Device Fingerprinting
- Behavioral Profiling
- Entity Linking
- Transaction Risk Score
- Device Payment Statistics

Data Enrichment



#### **Data Sources**

**Card Transactions** 

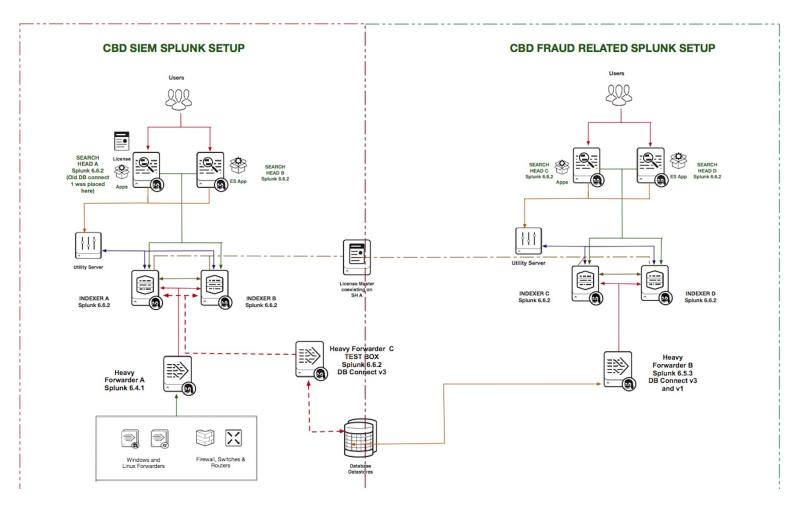






### **High Level Architecture**

#### Splunk Infrasturcture





**Data Models For Fraud Detection** 





Why?

- Cross Channel Searches became complicated very quickly
- Inconsistent enrichment across channels
- Cross channel field names inconsistencies
- Search performance optimization
- Investigation streamlining
- ▶ Performance issues on complex dashboards



#### **Custom Retail Banking Models**

Data Model	Channels
Financial Transactions	Credit Cards Debit Cards Authorizations SWIFT Payments
<risk platform=""></risk>	Internet Banking Mobile Banking
Customer Activity	Internet Banking Mobile Banking Telephone Banking
SMS	Outbound Communication



### Custom Data Models

Specific To Retail Banking Fraud

#### Financial Transactions

Financial\_Transactions

< All Data Models

Datasets  EVENTS  Add Dataset >	All Transaction All_Transactions CONSTRAINTS
All Transaction	tag=transaction
— Card Transaction	
- Credit Card Transaction	Bulk Edit ✓
— Debit Card Transaction	INHERITED
Credit Card Authorisation	_time
Internet Banking Transaction	host
Mobile Banking Transaction	source
Swift Transaction	sourcetype
Owner Transaction	EXTRACTED
	available_baland
	card_id
	card_product
	ard_product_lo
	channel
	country_code
	country_name

All Transaction All_Transactions		
CONSTRAINTS		
tag=transaction		
Bulk Edit V INHERITED		
_time	Time	
host	String	
source	String	
sourcetype	String	
EXTRACTED		
available_balance	Number	Hidden
card_id	Number	Hidden
card_product	String	Hidden
card_product_logo	String	Hidden
channel	String	
country_code	Number	Hidden
country_name	String	Hidden



Holistic Views: An Example

| pivot Financial\_Transactions All\_Transactions count(All\_Transactions) AS "Count of All Transaction" SPLITROW rim\_no AS rim\_no SPLITROW \_time AS \_time PERIOD hour SPLITCOL channel SPLITCOL status FILTER monetary\_transaction = true ROWSUMMARY 0 COLSUMMARY 0 NUMCOLS 100 SHOWOTHER 1 |addtotals fieldname=total\_approved "\*approved" | addtotals fieldname=total\_declined "\*declined" | search total\_declined > 2 total\_approved > 2

rim_no \(\displaystar{\dintar{\displaystar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\dintar{\din	_time ≎	credit_card ::: / approved 0	credit_card ::: // declined \$	debit_card ::: / approved ◊	debit_card ::: / declined ◊	internet_banking ::: / approved 0	mobile_banking ::: / approved 0	swift ::: 🖊 approved 🗘	swift ::: / declined 0	total_approved 0	total_declined 0
d0aaf39f31aed7d7bec1137e15d335b6	2017-09-19 09:00	0	3	0	0	8	0	0	0	8	3
7e34103d20ea96e0304389a74c44b50d	2017-09-14 06:00	0	5	0	0	5	0	0	0	5	5
33f68b5283ce15ed81bb906444473742	2017-09-16 02:00	0	3	0	0	0	3	0	0	3	3
456388d7767a9aaa38a8e5e8792f9cca	2017-09-16 12:00	1	6	0	0	0	6	0	0	7	6
2cae033644ec69257036df2ab6663350	2017-09-17 20:00	1	3	0	0	0	3	0	0	4	3
f698a9fa48ef482cc78042eb0626f86b	2017-09-18 16:00	0	3	0	0	4	0	0	0	4	3
4f72b078285a1746963be396111691e3	2017-09-15 19:00	3	3	0	0	0	1	0	0	4	3
2e4bd93ee1d7b2f36d66d1d78bff65f3	2017-09-13 17:00	0	4	0	0	0	4	0	0	4	4





# Investigation Dashboards for Analysts

Data Models For Fraud Detection



### **Cross Channel Analysis**

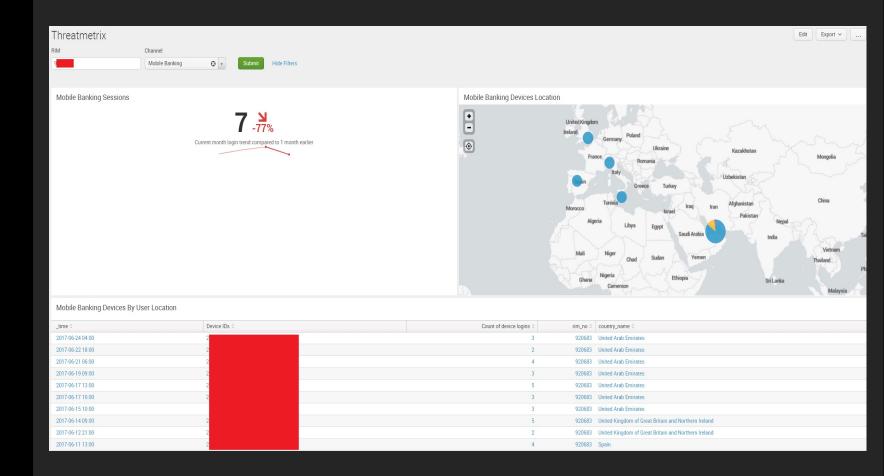
Analyst Investigation Dashboard





## Mobile Banking Activity

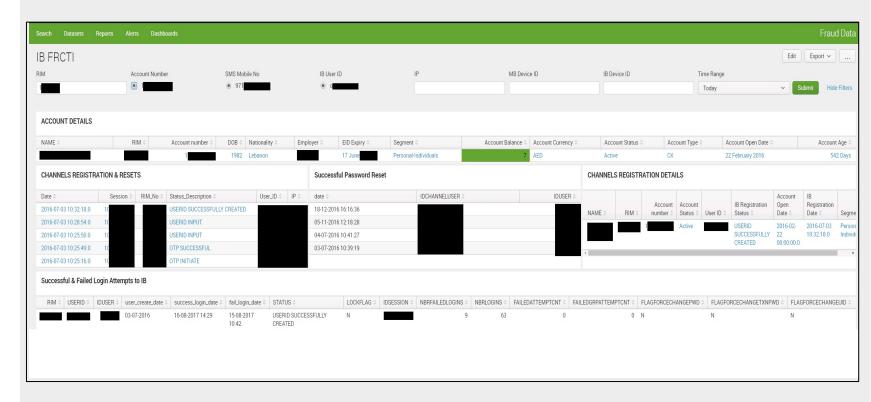
Mobile Device & Session Investigation

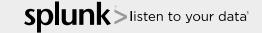




### Internet Banking Authentication

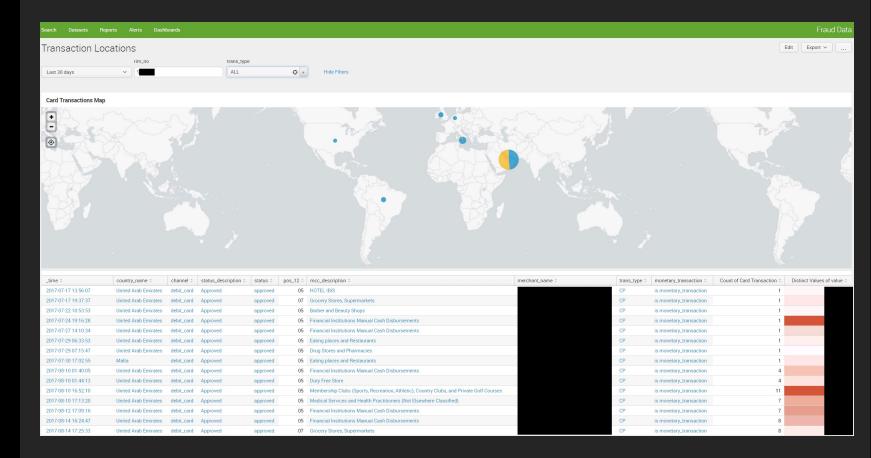
Analyst Investigation of Authentication & Registration





## Card Transaction Geography

Card Transaction Location Analysis

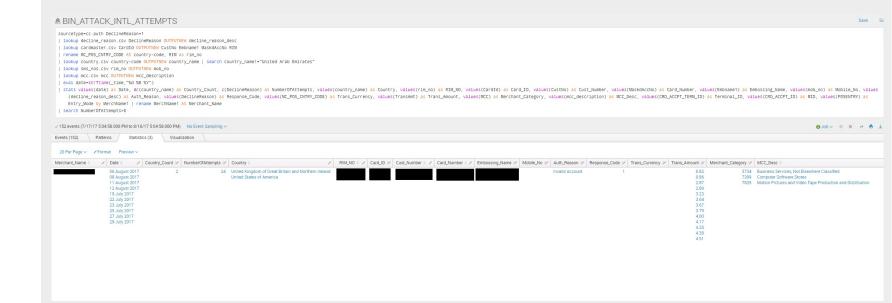




### Fraud Alerting



### Alert: International Transaction Declines



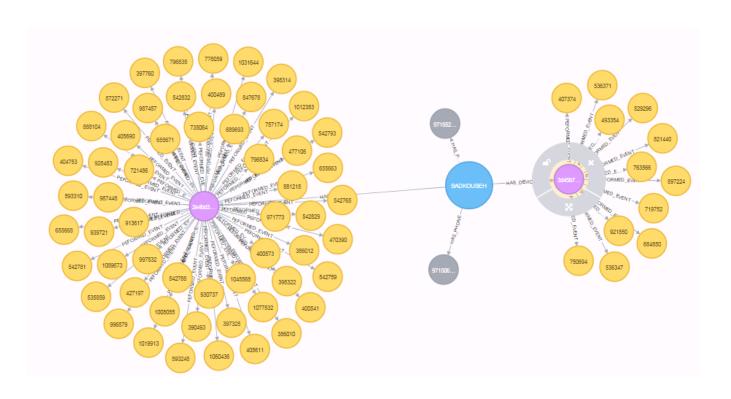






#### What's Next

Looking Forward - What's next



- Graphs are everywhere. Using graph databases for improved fraud detections
- Explore interconnectivity of data
- Explore ways to improve visualization
- Machine learning?





- 1. Understand your own data, focus on use cases
- 2. It must be actionable. Careful data onboarding
- 3. Consider the need for Splunk professional service while building required skill sets internally



### Thank You

Don't forget to rate this session in the .conf2017 mobile app



### Q&A

Participant name | Role

Participant name | Role

splunk> .conf2017

