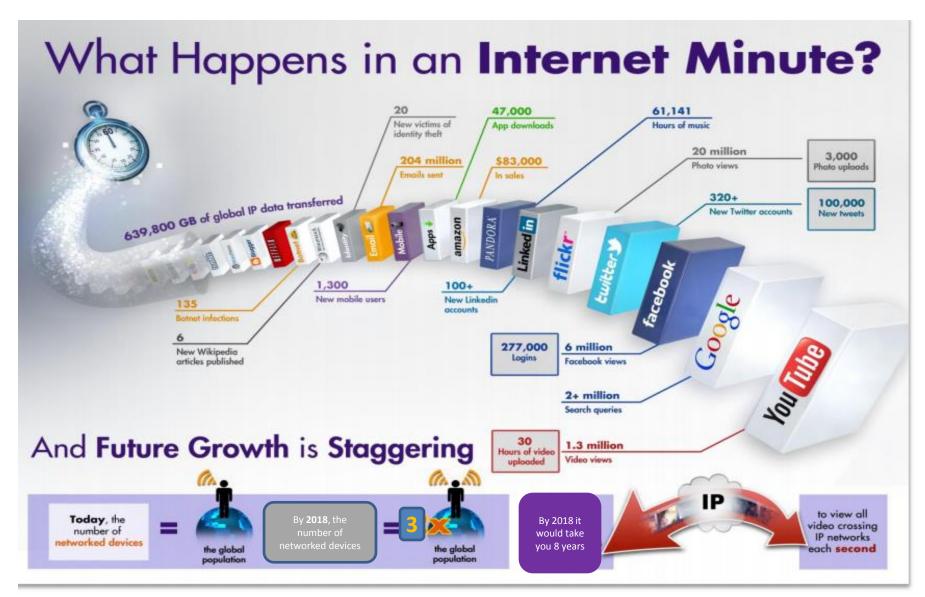
# **MIZUHO**

#### Mizuho OneView

BIGData/DataLake/Unify All Database using one platform
Forecasting/Predicting Customer Stickiness using Machine Learning
POC Results

AOAD Innovation Team – Singapore Guru S Anand, Singapore





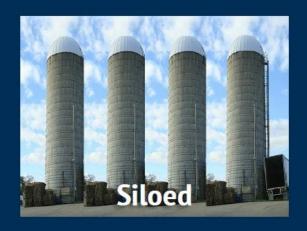


# The **Current** Enterprise Problem With Accessing Data



# The Current MIZUHO Problem With Accessing Data

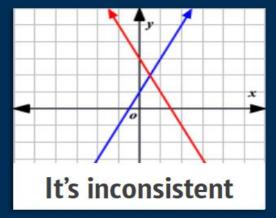
# Timely Access to Unified Data is a MASSIVE Hurdle to Everything!













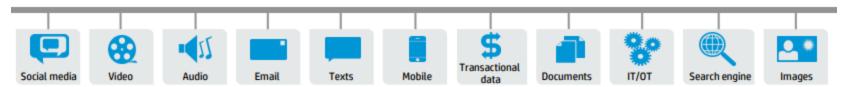
# Digital Strategy Foundation - Objective

- ➤ BIGData/DataLake Store All data for various Golden Sources. Structured, Unstrucutred, Real time data from Bloomberg, Reuters etc.
- One Platform For Data Integration from various Golden Sources, DataLake and to consume for downstream applications, Open APIs

#### Unify using a single platform

**OneView Mizuho** 

Create Innovative applications for Forecasting/Predicting for Machine Learning Ready/ Artificial Intelligence Ready

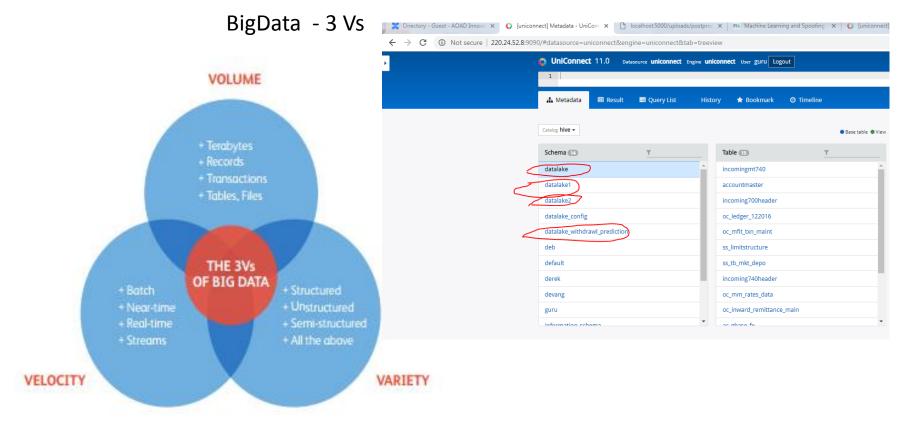


- Data Analytics/ Data Science Ready
- > API Ready



# **MIZUHO**

#### DataLake Built in SG Mizuho - POC

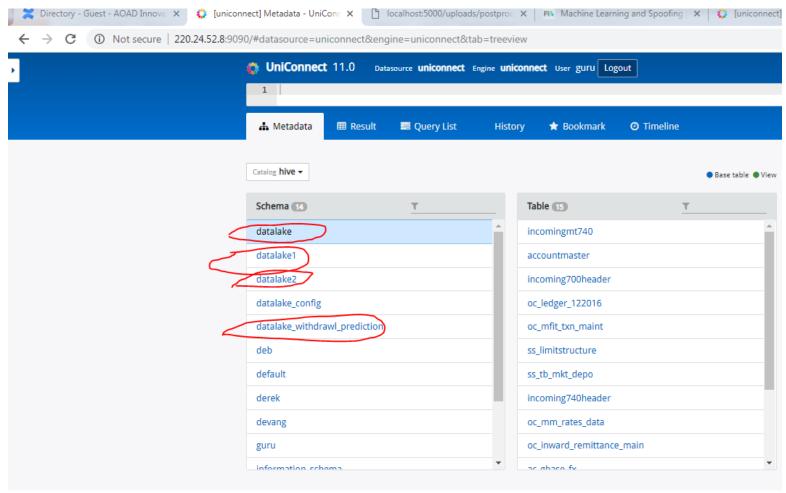


We created a DataLake using Hadoop infrastructure in SG Mizuho

http://220.24.52.8:9090



#### DataLake Built in SG Mizuho - POC



We created a DataLake using Hadoop infrastructure in SG Mizuho for to use by various department like GTBD, RMD, AOAD, MRDD



# **Project Objective**

# **Digital Strategy Foundation**



BIGData/DataLake – Store All data for various Golden Sources. Structured, Unstrucutred, Real time data from Bloomberg, Reuters etc.

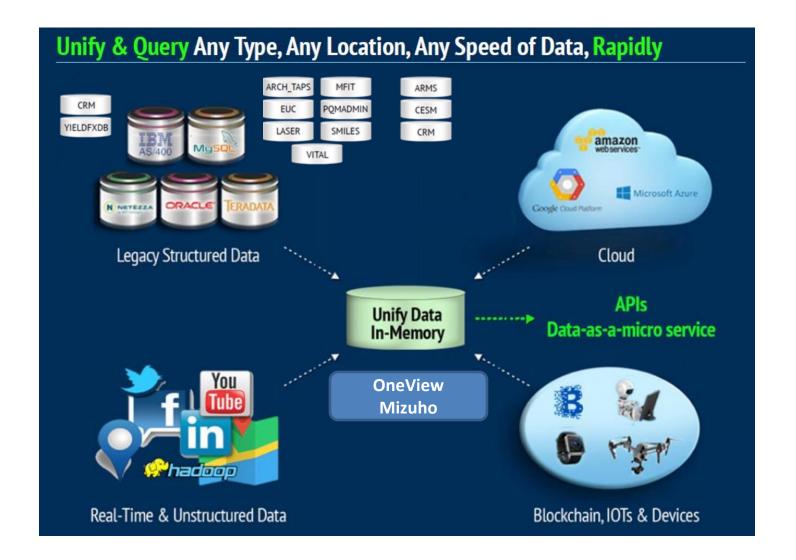
# Result - GREEN

One Platform – For Data Integration from various Golden Sources, DataLake and to consume for downstream applications, Open APIs

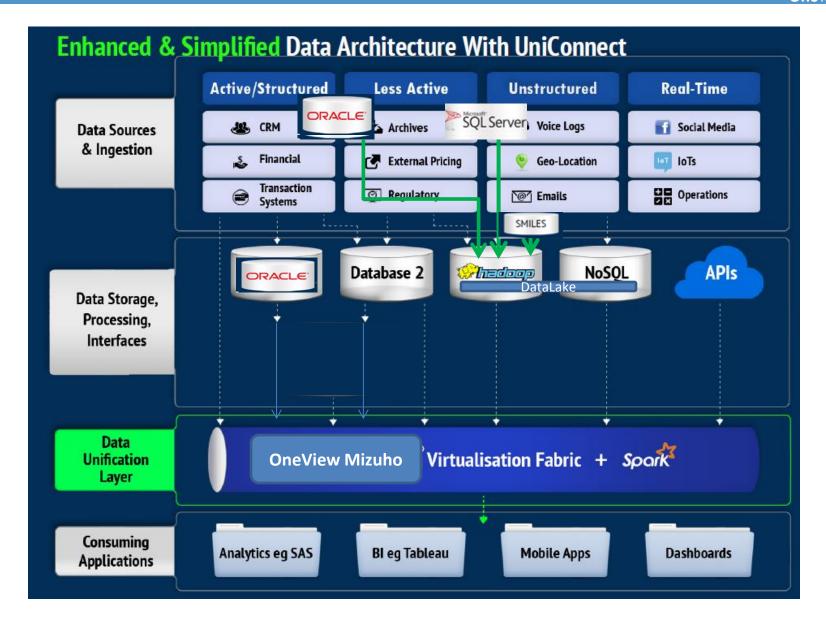
Unify using a single platform

**OneView Mizuho** 















# **Digital Strategy Foundation**

BIGData/DataLake – Store All data for various Golden Sources. Structured, Unstrucutred, Real time data from Bloomberg, Reuters etc.

Unify/In-Memory Database – For Data Integration from various Golden Sources, DataLake and to consume for downstream applications, Open APIs

Unify using a single platform

**OneView Mizuho** 

# Result - GREEN

- Forecasting/Predicting for Machine Learning Ready
- Data Analytics Ready





Systems	Version	Link
Operating System	) RHEL 7+ (64 bit)	NA NA
Apache Hadoop	2.7.4	http://www-eu.apache.org/dist/hadoop/common/hadoop-2.7.4/hadoop-2.7.4.tar.qz
Apache Hive	2.2.0	http://www-us.apache.org/dist/hive/hive-2.2.0/apache-hive-2.2.0-src.tar.gz
Apache Derby	10.14.1.0	http://www-eu.apache.org/dist//db/derby/db-derby-10.14.1.0/db-derby-10.14.1.0-bin.tar.gz
Apache Spark	2.2.0	http://www-us.apache.org/dist/spark/spark-2.2.0/spark-2.2.0-bin-hadoop2.7.tqz
Apache Kafka	2.11	http://www-eu.apache.org/dist/kafka/1.0.0/kafka 2.11-1.0.0.tqz
Apache Zookeeper	3.4.10	http://www-us.apache.org/dist/zookeeper/zookeeper-3.4.10/zookeeper-3.4.10.tar.qz
Scala	2.11.12	https://downloads.lightbend.com/scala/2.11.12/scala-2.11.12.tgz
Oracle JDK (Recommended Latest one)	JDK1.8.0_151+	http://download.oracle.com/otn-pub/java/jdk/8u151-b12/e758a0de34e24606bca991d704f6dcbf/ jdk-8u151-linux-x64.tar.gz
Python	2.7.11 +	https://www.python.org/ftp/python/2.7.11/Python-2.7.11.tqz

# Apache

Systems	Version	Link			
Apache Hadoop	2.7.4	http://www-eu.apache.org/dist/hadoop/common/hadoop-2.7.4/hadoop-2.7.4.tar.gz			
Apache Hive	2.2.0	http://www-us.apache.org/dist/hive/hive-2.2.0/apache-hive-2.2.0-src.tar.gz			
Apache Derby	10.14.1.0	http://www-eu.apache.org/dist//db/derby/db-derby-10.14.1.0/db-derby-10.14.1.0-bin.tar.g			
Apache Spark	2.2.0	http://www-us.apache.org/dist/spark/spark-2.2.0/spark-2.2.0-bin-hadoop2.7.tgz			
Apache Kafka	2.11	http://www-eu.apache.org/dist/kafka/1.0.0/kafka_2.11-1.0.0.tgz			
Apache Zookeeper	3.4.10	http://www-us.apache.org/dist/zookeeper/zookeeper-3.4.10/zookeeper-3.4.10.tar.gz			



# Server config

Department Wide - 10 TB			
Purpose		Spec	Qty
Uniconnect Processor	Master Node	DL380 gen10 2P Xeon-Silver 4110 (2.1GHz/8-core), 512GB RAM, 6x1.8TB SAS 10K HDD	1
Uniconnect Processor	Worker Node	DL380 gen10 2P Xeon-Silver 4110 (2.1GHz/8-core), 384GB RAM, 6x1.8TB SAS 10K HDD	4
Hive Datalake		Hadoop Servers, Network & Services	
Enterprise Wide - 50 TB			
Purpose		Spec	Qty
Uniconnect Processor	Master Node	DL380 gen10 2P Xeon-Silver 4110 (2.1GHz/8-core), 512GB RAM, 6x1.8TB SAS 10K HDD	1
Uniconnect Processor	Worker	DL380 gen10 2P Xeon-Silver 4110 (2.1GHz/8-core), 384GB RAM, 6x1.8TB SAS 10K HDD	10
Hive Datalake		Hadoop Servers, Network & Services	

# **POC Completed**

http://220.24.52.8:9090

http://220.24.52.8:8080/ui

http://spws271:8090/display/AI/Google+News+using+API

#### Eg: Accessing Multiple DB in a single place

SELECT \* FROM sqlserver\_arms.dbo.tb\_mkt\_depo SQLSERVER\_DEPO INNER JOIN (SELECT \* FROM access\_yieldfxdb.public.deposits) ACCESS\_DEPO ON SQLSERVER\_DEPO.MKT\_DATE=ACCESS\_DEPO.MKTDATE AND SQLSERVER\_DEPO.CCY=ACCESS\_DEPO.CURR AND SQLSERVER\_DEPO.PERIOD=ACCESS\_DEPO.PERIOD

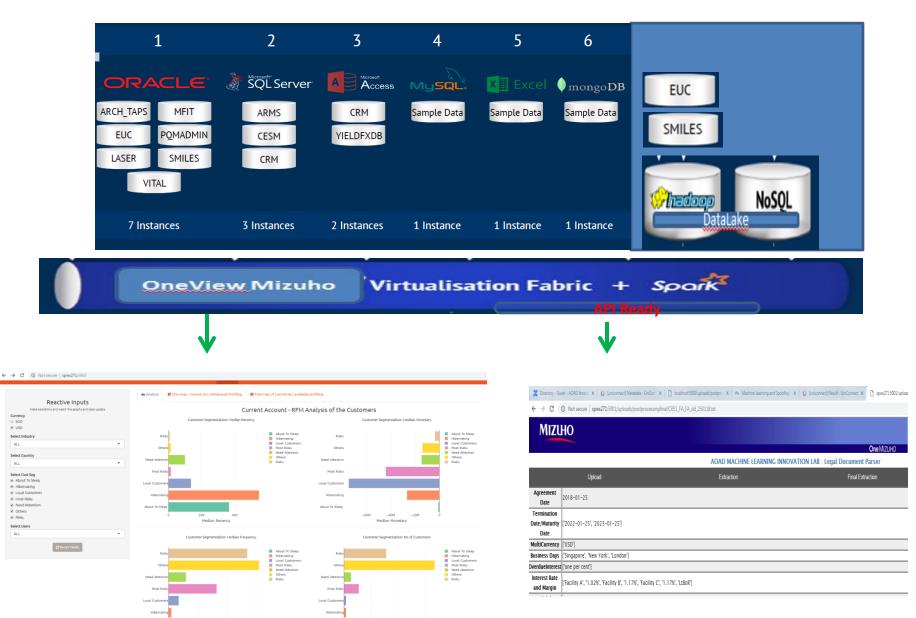
Result - GREEN

**MIZUHO** 

# POC - Technical Results

										One MIZUHO
S. No Test cases - Activity	Day 1	Day 2	Day 3	Day 4	Day 5	Day 6	Day 7	Day 8	Day 9	Day 10
30-Jul-18										
1 Tested Hive/Datalake connectivity by executing queries	GREEN									
2 Tested Oracle connectivity by executing queries (Different subnet)	GREEN									
3 Issue with MySQL connectivity at database side (Different subnet) - ID permission issue	AMBER	AMBER	GREEN							
4 Issue with SqlServer connectivity (Password must be changed) - ID permission issue	AMBER	GREEN								
Installed query executer, but was having issuefor loading CSS and JS (refereeing online										
links).										
5 Resolved and will deploy again it on tomorrow.	RED	GREEN								
6 Trying to setup Python libraries to access UniConnect via Python code.	AMBER	AMBER		In Progress	GREEN					
7 We had issue with internet proxy setup in Unix.	RED	RED	GREEN							
31-Jul-18										
8 Installed ScaleETL		AMBER	GREEN							
9 Installed UniConnect Dashboard		AMBER	GREEN							
Installed reconfigured query executor (Testing yet to be done) -										
10 Waiting for Mizuho to install Java 8 -32bit		AMBER	GREEN							
11 Tested executing query via JAVA sample code.		GREEN								
12 Offloaded sample tables in Hive by CTAS on SqlServer		GREEN								
13 203 Million of records, creation time around 25 Minutes. (ORC format)		GREEN								
14 28 Million of records, creation time around 16 Minutes. (ORC format)		GREEN								
15 Configured extra database connectors for Oracle and SqlServer		GREEN								
16 Installed Mongo DB on local and tested.		GREEN	ODEEN							
18 Mizuho Mongo DB is not connecting. Permisison issue		AMBER	GREEN							
Not able to run Python code in the server due to missing libraries - Planning to		444050	I D	I	COFFN					
19 download Python libraries separately  01-Aug-18		AMBER	in Progress	In Progress	GREEN					
20 Installed and configured Apache Kafka.	I	Τ	GREEN			I			I	
21 Tested ScaleETL on Google news API and dumped data into Kafka. Fetched by UniConnect.			GREEN							
22 Tested and configured access control for 3 dummy users.			GREEN							
02-Aug-18			Oneen							
23 Security authorisation - Matrix tested				GREEN						
24 Introduced the Uniconnect Web Ui to RMD team				GREEN						
25 Tested Access DB connectivity				GREEN						
26 Deployed and configured Jasper server				GREEN						
27 Jasper studio Report generation is pending.				In Progress	In Progress	GREEN				
28 WebUI - Latest binaries deployed				GREEN						
29 Java code to connect to Uniconnect provided for verification		ļ	ļ	In Progress	In Progress	GREEN			l,	l <sub>.</sub>
Trying to access Uniconnect Web UI URL in TK office. TK office is having issue in connecting										
to this URL.										
30 Working with TK team to resolve this issue.				In Progress	In Progress	5				
03-Aug-18										
31 Demo provided to Mizuho team on work accomplished till now. Briefing to Key Stakeholder	s				GREEN					
32 Tested CSV files upload via CSV connector					GREEN					
33 Able to source Google News data into Uniconnect. Used Open Google API					GREEN					
34 Transformation of Google data are in progress.					In Progress	GREEN				
35 Able to use Python code to connect and query Uniconnect data					GREEN					
06-Aug-18										
36 Email SMTP setup completed. Able to send test mail on Mizuho network						GREEN				
37 ScalETL and Kafka setup tested						GREEN				
38 Able to source and transform MAS FX Daily rates from open source API and push to hive						GREEN				
39 Able to deploy and run locally created Jasper report on Mizuho		+	+			GREEN			1	
40 Tested basic connectivity with Postgres SQL.		+				GREEN			+	1
41 Basic data offload using 2 tables from different source hive has been tested. Able to send		+	+			GREEN			1	1
41 pasic data official using 2 tables from different source nive has been tested. Able to send		1		1		GKEEN			1	

# Leverage DataLake and Built Applications



### **POC** Business use cases

#### POC using the Oneview Mizuho and Datalake

- Current Account Cash Flow Forecasting/Prediction using Machine Learning, Analysis and Customer stickiness
  - 1. Customer profiling
  - 2. Customer segmentation by Industries
  - 3. Forecasting
  - 4. Prediction
- ➤ Time Deposit Cash Flow Forecasting/Prediction using Machine Learning, Analysis and Customer stickiness
  - 1. Customer profiling
  - 2. Customer segmentation by Industries
  - 3. Forecasting
  - 4. Prediction
- Legal Document extraction
- MRDD Remittance Analysis
- API Ready



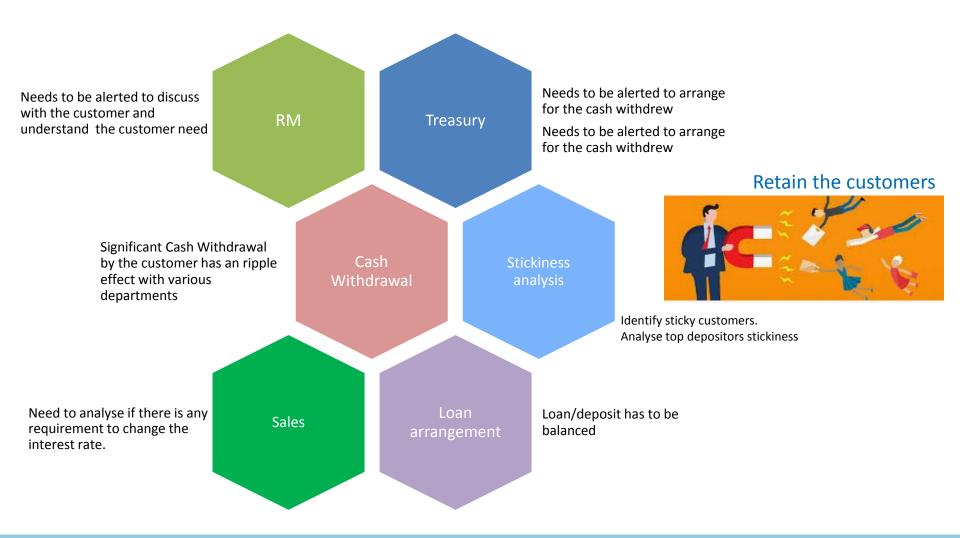
# **POC** Business use cases

## POC using the Oneview Mizuho and Datalake – Use case 1

- Current Account Cash Flow Forecasting/Prediction using Machine Learning, Analysis and Customer stickiness
  - 1. Customer profiling
  - 2. Customer segmentation by Industries
  - 3. Forecasting
  - 4. Prediction

## Objective – Customer cashflow Forecasting/Prediction

## **Predictive Analysis**



Cashflow forecasting is still a

- cumbersome,
- manual, and
- · spreadsheet-based process







Involving many people from across the organization



Resulting in monthly or quarterly, rather than weekly, updates.

Cash flow forecasting can be an important function to predict large deposit withdrawals; to analyse top depositors stickiness level for stable source, etc. In Mizuho we are doing this analysis at crude level. No back-testing.

#### **Cashflow Withdrawal Prediction Benefits**

Retain the customer

Early Alert

Identify trends
/ customer
segmentation

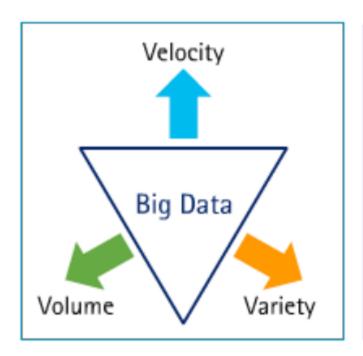
Planning for liquidity / Asset Liability Mgmt

# Why we need BigData?

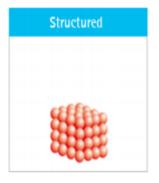
Big data in general has context in three Vs:

- Sheer quantity of data
- Speed with which data is produced, processed, and digested
- Diversity of sources inside and outside

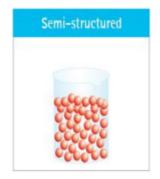
.



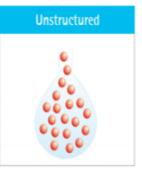
#### The different types of data that contribute to this are:



- Fields/Tables /Columns
- RDBMS/Spreadsheet



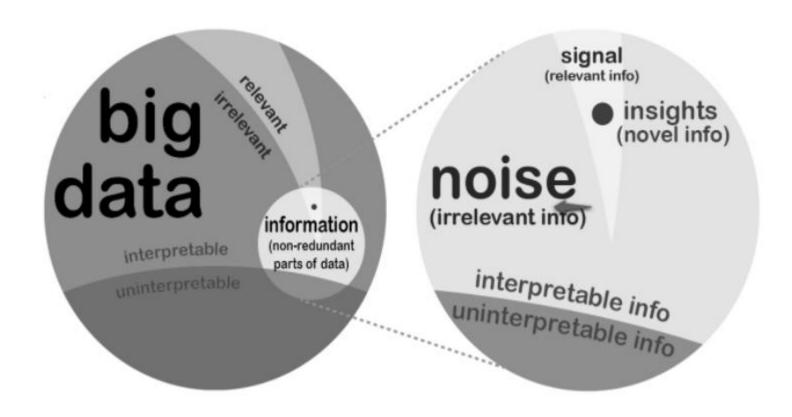
- Markers/Tags to separate elements
- XML/HTML



- · No fields/attributes
- Free form text (email body, notes, articles)
- Audio, video, and image

Separating the signal from the noise<sup>1</sup> becomes really relevant

to



<sup>1</sup> http://techcrunch.com/2012/11/25/the-big-data-fallacy-data-#-information-#-insights/

# Some Parameters outside our Bank for prediction considered in the POC for to have more accuracy

Structured Data

Historical daily **SGX** price from Yahoo finance Historical daily **Nasdaq** Exchange Price from Yahoo finance Historical daily **S&P** Exchange Price from Yahoo finance Historical daily **Nikkie** Price from Yahoo finance Customer Equity price from **Bloomberg** terminal for two customers

Unstructured Data

**News** related to the particular customer OG30 from newsapi.com using API call from python and **Kafka(hadoop) Top 20 news** headline around the worldwide from kaggle.

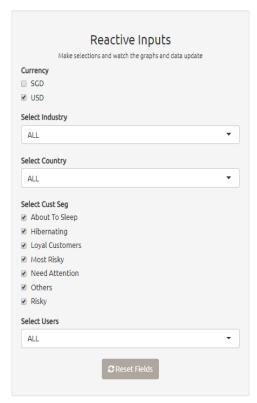
#### Our Bank - Transaction data

- Transaction Data
- Processed approximately <u>20 million</u> records from Hive(Hadoop BigData)

# **MIZUHO**

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#### **Current Account**



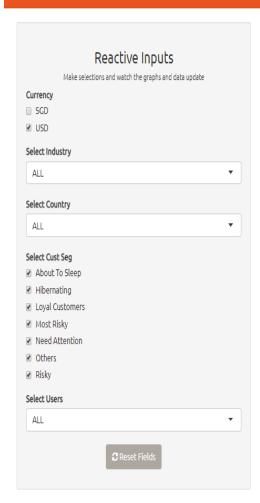


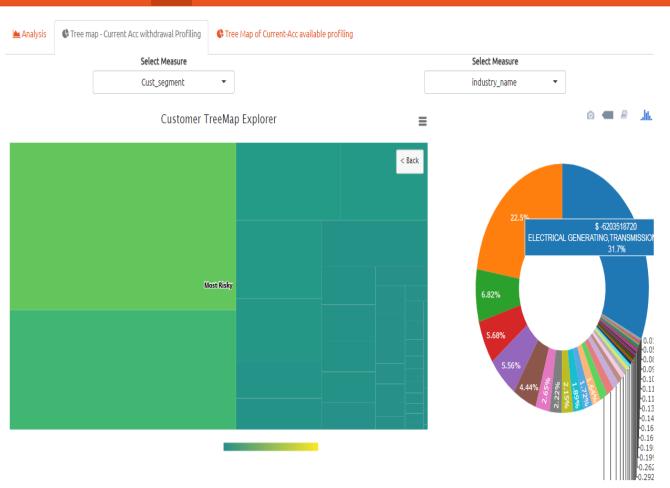
#### One MIZUHO

**MIZUHO** 

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MIZUHO SG - AOAD INNOVATION Machine Learning - Customer Stickiness(Current Account) Dashboard Analytics Forecasting

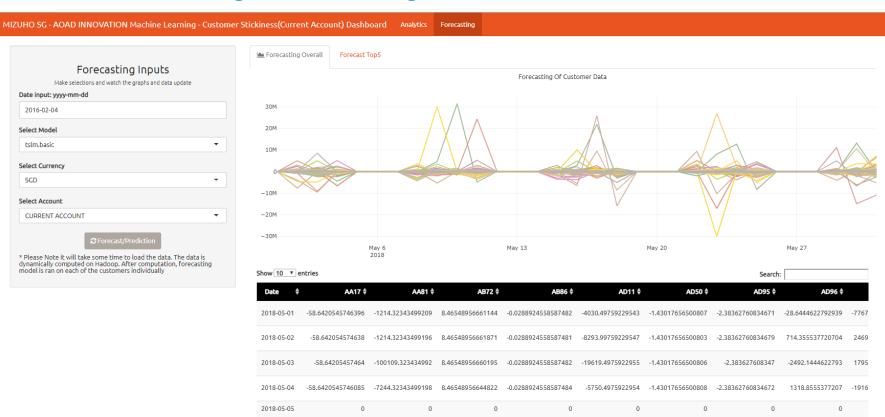




# **MIZUHO**

# **Machine Learning - Forecasting**

2018-05-06



0

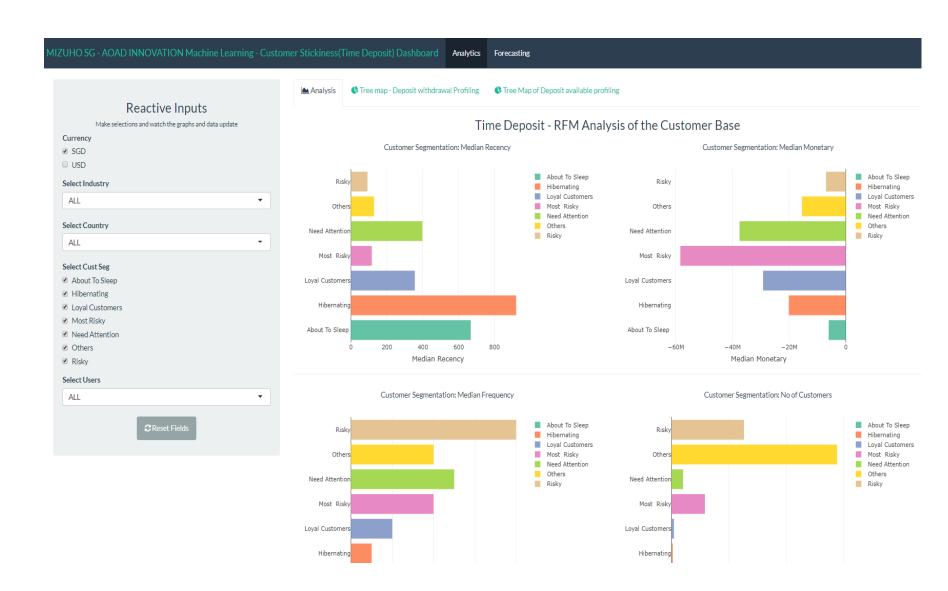


# **POC** Business use cases

### POC using the Oneview Mizuho and Datalake – Use case 2

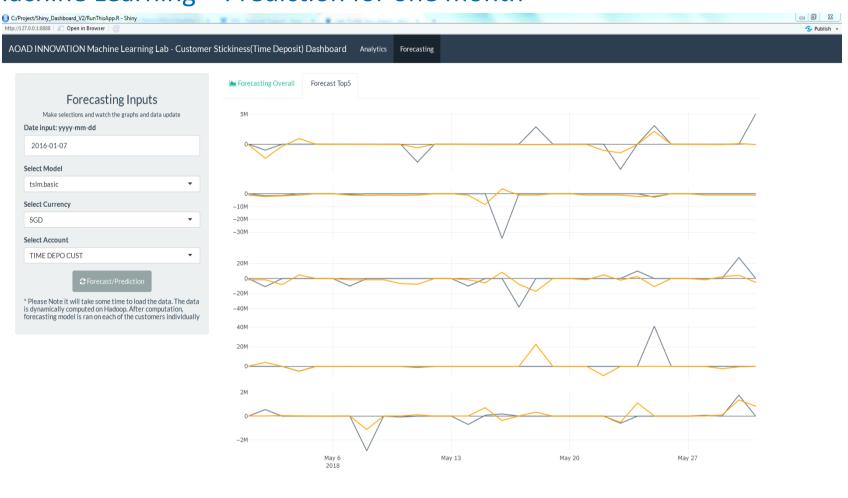
- Time Deposit Cash Flow Forecasting/Prediction using Machine Learning, Analysis and Customer stickiness
  - 1. Customer profiling
  - 2. Customer segmentation by Industries
  - 3. Forecasting
  - 4. Prediction





# **MIZUHO**

# Machine Learning – Prediction for one month



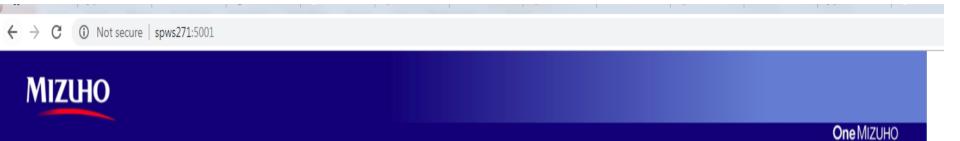
# **POC** Business use cases

# POC using the Oneview Mizuho and Datalake – Use case 3

Legal Document extraction

# Legal Document extraction using NLP

http://spws271:5001/



# AOAD MACHINE LEARNING INNOVATION LAB: Legal Document Parser



# Legal Document extraction using NLP

http://spws271:5001/

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# **MIZHO**

One MIZUHO

#### AOAD MACHINE LEARNING INNOVATION LAB: Legal Document Parser

	Upload	Extraction	Final Extraction
Agreement Date	2018-01-25		
Termination Date/Maturity Date	['2022-01-25', '2023-01-25']		
MultiCurrency	['USD'] ['Singapore', 'New York', 'London']		
OverdueInterest			
Interest Rate and Margin	['Facility A', '1.02%', 'Facility B', '1.17%', 'Facility C', '1.17%', 'LtBof	R']	
Availability Period Processed	['2018-04-25', '2022-12-25']		
Repayment Processed	Bullet		
Borrower	Dated 25 January 2018 AMONG HUDSON DC ASSETS LLC, NAV	/ARRO DC ASSETS LLC, ETOWAH DC ASSETS LLC	C, REDWOOD DC ASSETS LLC Collectively, as the Borrower MIZUHO BANK, LTD.



# POC Business use cases

# POC using the Oneview Mizuho and Datalake – Use case 4

MRDD – Remittance Data Analysis

## Remittance Data (MRDD)

- Able to upload 15 million remittances to Data Lake in a few minutes
- It covers Apr 2014 to May 2018
- Total of 18 trillion USD notional remittance, where Mizuho is one party
- Ran **complex** query and found **4.8 million** remittances, where the FX is **not** done by Mizuho over the period of 5 years.\*Rough estimate
- Query took 20 seconds
- 'Potential' FX business opportunity of notional USD 1.5 trillion over 5 years
- \*Rough estimate condition examples where FX business is likely not captured:
  - Sending non-SGD from Singapore, but the remitting bank is not Mizuho



## POC Business use cases

## POC using the Oneview Mizuho and Datalake – Use case 5

- API in Mizuho
- Internal API and
- External API

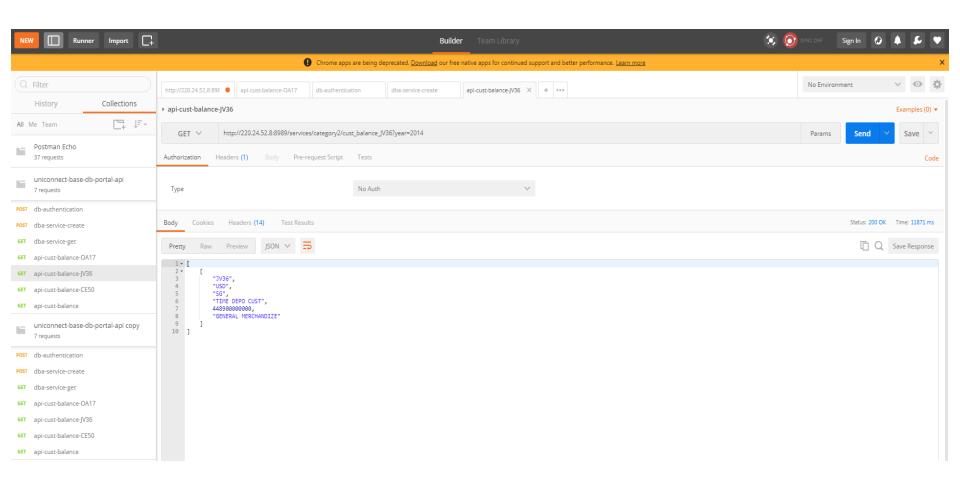


## OneView Mizuho Features

PRODUCT FEATURES	ASTER DATA	DENODO	ABINITIO QUERYIT	CIRRO	TAMR	OneView Mizuho
Data Is Copied To Platform	Υ	Υ	Υ	Υ	Υ	N (In-Memory)
Analytics Eco-system On Top	N	N	N	N	Υ	Υ (with Spark)
Deployment	On Premises	On Premises	On Premises	On Premises	On Premises/Cloud	On Premises/Cloud
Data Cleaning Enrichment	N	N	Y	N	Y	Y (Partial, with SQL)
Volume / CPU Based Licensing	Υ	A	PIR	ead	У	Free To Grow
Scalability	Fixed	Fixed	Fixed	Fixed	Fixed	Auto Elastic
Can Develop Connector API	N	N	N	N	N	Υ
JDBC/ODBC	Υ	Υ	N	Υ	N	Υ
Learning Curve	Y	Y	Υ	Υ	Υ	N (Use SQL)
Security / Admin / Audit	Υ	Y	Υ	Υ	Y	Y



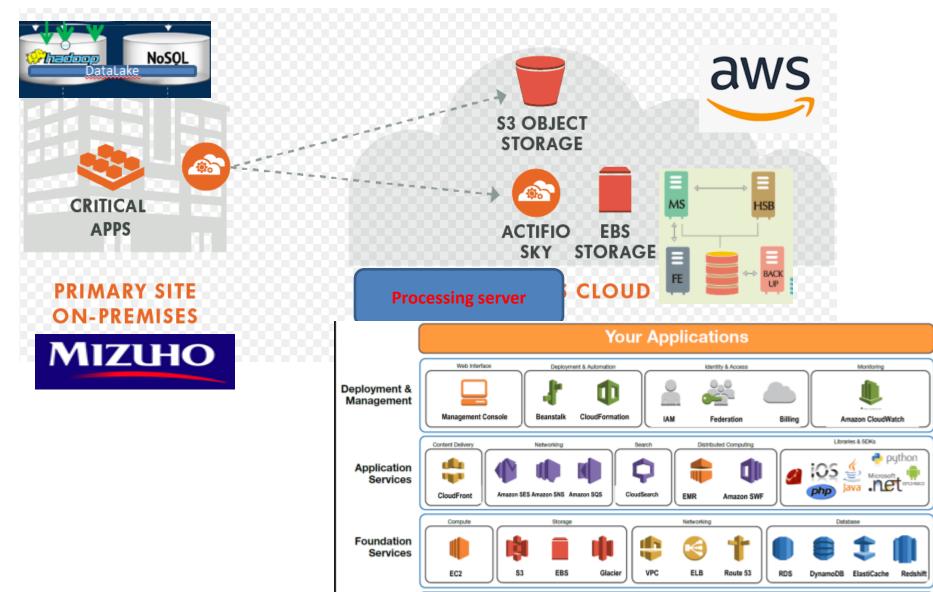
## **APIs**



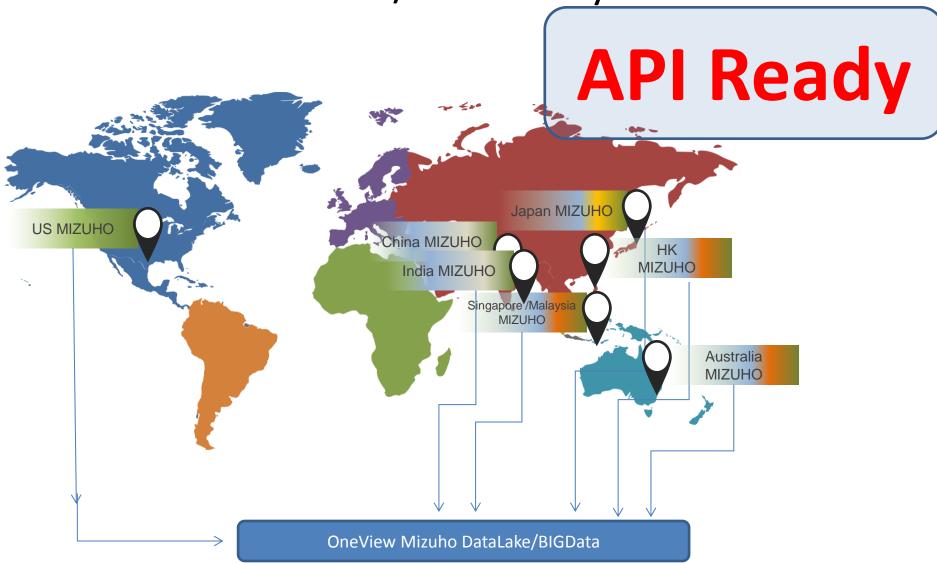
## Future Plan with AWS

**One** MIZUHO

Data in house and Processing server in Cloud for the applications that we built



## Future - DataLake/In-memory Database



Case Study



**MIZUHO** 

## **Corporate Time Deposit Customer Profiling** and Analysis of Customer Stickiness

By achieving integrated access to multiple sources of data, a Percipient and HPE Proof of Concept demonstrated how Mizuho Singapore's Innovation team could forecast and predict customer cash flow and improve stickiness

### **Executive Summary**

### The Business Need:

- Track behaviour patterns within Time Deposit customer segments
- · Incorporate market news and events into customer analytics
- · Forecast potential for customer attrition

### The Tech Challenge:

- Aggregate and store fragmented customer data
- Ingest third party data in real time
- Make available business-specific information

### The Solution:

- · Unify the bank's multiple databases
- Deploy a firm level Datalake platform to host logical outlined above. Datalake/Datamarts
- Integrate with the bank's analytics and machine learning code

### Manual processes

SG Mizuho's Asia and Oceania Administration Department - Innovation team tracks and forecasts customer cash flows in order to predict the likelihood of large deposit withdrawals and customer stickiness. This helps drive early RM alerts, identification of customer behaviour trends, customer segmentation and liquidity management.

However, these analytics exercises are currently performed by manually extracting and transforming data sourced from a variety of internal databases, a process that is cumbersome, slow and resource-intensive. Factoring in external data adds further complexity and as a result, is not routinely done. As a result, forecasts can only be updated monthly or quarterly, rather than on a more impactful weekly cycle. It is also impossible to continuously back-test and refine the forecast model.

### New data elements

To address these challenges, Mizuho engaged integration software provider, Percipient, for a trial of the latter's flagship UniConnect platform. A key requirement for this Proof Of Concept (POC) was for structured and unstructured data to be delivered to a single end-point, paving the way for easy discovery and consumption by various business, technology and analytics teams.

The external data accessed by Uniconnect included:

- · SGX prices from Yahoo finance
- Nasdaq Exchange Prices from Yahoo finance

## O Percipient:

MIZUHO

- S&PExchange Prices from Yahoo finance
- Nikkei Prices from Yahoo finance
- Customer Equityprice from Bloomberg terminal
- Newsitems from news api.com using APIcall from python and Kafka(Hadoop)
- Top 100 news headlines worldwide from the website Kaggle.

This was unified with internal data comprising millions of rows of transaction data stored over the last five years in a variety of repositories and formats, including:

- Oracle
- SQLServer

### **NextGenSolution**

Based on the above specifications, Percipient proposed a solution comprising a number of elements central to meeting the bank's needs.

#### Connectors

UniConnect Connectors deployed in order to automatically ingest data from the multiple sources

## Data Lake

A Hadoop datalake was built on premise. UniConnect 's engine was used to move new and existing data to the datalake.

### Transformation

Approx 30 million customer records were transformed virtually to align with the bank's Python-based forecast models.

### Consumption

Data could be discovered and queried in seconds or microseconds as a single view using UniConnect's SQL interface

To provide hardware support for this POC deployment of the UniConnect and datalake platforms, Percipient and Mizuho partnered with leading IT infrastructure and services company, HPE. Based on the SLAs defined for the trial, HPE were able to supply a pair of DL380 servers for the Uniconnect and Hadoop Datalake processing.

### Analytical Outcomes

Leveraging UniConnect's integration capabilities, Mizuho's SG Asia and Oceania Department's Innovation teambuilt dashboardsas shown below:





### Machine Learning - Forecasting



Machine Learning - Prediction for one month



Analysis of the customer profiling and stickiness patterns allowed the team to uncover a number of important stickiness indicators. Stickiness was determined to be a stable customer variable that relates closely to withdrawals. These results can be used to drive new analytical models and customer retention strategies.

Guru S. Anand, Vice President from the Innovation Team, Asia and Oceania Department, Mizuho Bank Singapore, said, "We are very pleased with the insights and forecasting we have been able to achieve, underpinned by the UniConnect platform. We believe the solutions that Percipient has introduced to us will substantially lift our analytical capabilities and correspondingly, our productivity, in today's hypercompetitive corporate banking sector.\*

The full Mizuho Bank Case Sudy document is available onwww.percipientcx.com

Guru S. Anand. Vice President from the Innovation Team. Asia and Oceania Department. Mizuho Bank Singapore.

We are very pleased with the

insights and forecasting we

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believe the solutions that

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analytical capabilities and

correspondingly, our productivity, in today's

hyper-competitive

Percipient has introduced to

corporate banking sector. ""

underpinned by the

MIZUHO

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# Additional info



## POC - DBs

Data type	Database	Data accessible	Description	Section
Structured data	CDM ODBC	Customer profile and credit facilities	customer profile, financial data, customer rating, credit facilities, protection, facility rating (outstanding & limit), write-off & reserve etc.	
	Hyperion	Profit & exposure	data from the profit and exposure management systems (PEGASYS, PEGASYS+), and the customer relationship management system (IGCIS) etc.	GTBD
	Gbase EUC	Previous business day's data of Gbase	data storage system by Head Office, contains numerous tables and the table names are self- explanatory (e.g. LOANS, REMITTANCE etc.)	AOAD,GTBD
	Gbase for other branches	data of Gbase for other branches	data storage system by Head Office, contains numerous tables and the table names are self- explanatory (e.g. LOANS, REMITTANCE etc.)	AOAD,MRDD
	SMILES	Singapore branch's transaction processing systems	branch's integrated transaction processing systems which cover business areas listed below and having screens for end users. Remittance (inward, outward, in-house transfer), GIRO, Loan, Cheque issuances and clearing, Deposit, FX and money market, Certificate of balance, eStatement and eAdvice, Trade processing (import, export, LC etc.), Gbase reports on the screen	AOAD,GTBD
	SMILES Web	Singapore branch's transaction processing systems	branch's integrated transaction processing systems which cover business areas listed below and having screens for end users. Remittance (inward, outward, in-house transfer), GIRO, Loan, Cheque issuances and clearing, Deposit, FX and money market, Certificate of balance, eStatement and eAdvice, Trade processing (import, export, LC etc.), Gbase reports on the screen	AOAD
	TAPS	Singapore branch's transaction processing systems	branch's integrated transaction processing systems which cover business areas listed below and having screens for end users. Remittance (inward, outward, in-house transfer), GIRO, Loan, Cheque issuances and clearing, Deposit, FX and money market, Certificate of balance, eStatement and eAdvice, Trade processing (import,	AOAD
	LASER	Singapore branch's small scale data warehouse	branch's data storage system that holds historical data of Gbase EUC (not all) and branch's transaction processing systems (SMILE & TAPS).  Main usages of this system are regulatory reporting and maintaining historical data.	GTBD
	External data providers (e.g. Bloomberg, Reuters etc.)	Complement internal data	customer equity info, macroeconomic indices, financial market indices etc.	AOAD,GTBD
	IGSIS / IIMI	Customer info	customers' profile and performance data	RMD
	DEVON	Customer FX transactions	data source for customers' FX transactions executed by AOTD	GTBD
	CES	Data for MAS	Data for upcoming MAS revised return 610.	RMD
	VITAL	Front office for Treasuray	Front office for Treasuray for Hanoi Branch	MRDD
Semi- structured data	Social media & Websites	Opinions about customers and corporate actions	news, public consensus, corporate actions, company announcements etc.	AOAD,GTBD
Unstructured	N.A.	Communication with customers and internal customer/industry analysis	business memos, credit applications/ review, legal agreements, communication with customers by email and phone	AOAD,GTBD

**One** MIZUHO

## Other Use cases in Corporate Banking





## **Facts**

- We are building a foundation here for BigData/DataAnalytics/Machine Learning platform.
- BigData/DataLake is one of the fundamental blocks for Digital Strategy/Innovation.
- We are not implementing a solution per say like what other teams are doing in a fragmented fashion for other specific use cases.
- We are building a BigData/DataLake for all the departments/sections/regions/countries usage(Global usage) storing Terabytes/Petabytes of structured/unstructured/semi-structured data to move from traditional computational ecosystem to **predictive computational** ecosystem.
- Also the use cases used in POC are some very few samples only and not the actual implementation.
- We have lot of use cases from other departments like RMD, MRDD etc. Even their use use cases will be tested during POC which is not documented.
- Moreover the use cases are not limited in the BIGData/DataLake solution. It is unlimited which will evolve as we
  move forward from various departments/sections/regions/globally.
- Other department stakeholders like GTBD, MRDD, RMD expressed their interest to participate in the conference call to explain the importance of this BigData/DataLake project in their respective departments if necessary.



# Questions