

Lunch n Learn Series: Data Governance

Please bring your lunch and join us on Tuesday, November 23rd, 2021 at 12:30 PM at our third monthly Risk Management Lunch & Learn Series event for Apple Bank colleagues on Data Governance. Dmitriy Khegay and Suresh Jayaraman will be presenting an overview on the Data Governance at the Bank and will provide a Data Governance Platform overview – Collibra/Precisely (Trillium) Tools. The event will be held in the 5th floor conference room behind reception. Attendance is limited to the first 12 people in person, and there will also be a video conference line open.

Topic Overview:

Data Governance helps ensure that Apple Bank's data assets are identified, usable, accessible, and formally managed. Data Quality is an integral part of Data Governance that ensures that the data is trustworthy and suitable to serve its intended business purpose. Good Data Governance is critical for improved data quality, regulatory compliance, decreased data management costs, and increased access to data for all stakeholders. The result is better decision making and better business outcomes.

- Data Governance overview at Apple Bank
- Data Governance Platform at the Bank – Collibra / Precisely(Trillium) Tools overview
- Data Governance Guest Speaker – Philip Black from Data QG (<http://dataqg.com/>). A trusted platform with leading resources, talent, and opportunities to empower DG Community through connection.

Dmitriy Khegay Bio

- Dmitriy Khegay joined the Data Governance team in December 2019 and leads the Data Governance Operations initiatives. Dmitriy is a Data Governance Officer aligned to CECL and Credit departments and co-lead on Finance, Accounting, and Treasury domain. His prior experience includes Data Governance, Data Quality and Business Analytics roles in Finance, Risk Management from Credit Suisse, American Express, BNY Mellon, and HSBC.

Suresh Jayaraman Bio

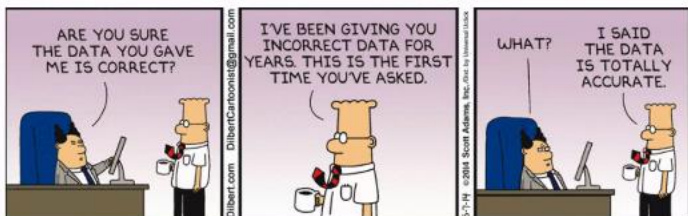
- Suresh Jayaraman joined the Data Governance team in May 2021 and leads the Data Quality initiatives. His prior experience includes Data Governance, Data Quality, Analytics, and Technology roles in Risk Management from Goldman Sachs, Morgan Stanley, Bank of America Merrill Lynch, and AIG.

When

Tuesday, November 23rd, 2021

12:30 PM to 1:30 PM

Consequences of Poor Data Governance



Without effective Data Governance, key consumers of the Bank's data (Board and Management Committees, Regulators, etc.) may suffer from inconsistencies in the **accuracy and quality of data being used in key business processes and reports**.

Below are **some examples of ineffective Data Governance** and how each adversely impacts the Bank:

- Missing Customer Date of Birth during new Customer Account setup impacting Financial Crimes Compliance screening capabilities – Potential KYC Compliance Violation
- Missing Payee field for select monetary instrument transactions impacting the quality of transaction monitoring alerts – Potential AML Transactions Monitoring Compliance Violation
- Difference in Loan Disclosures presented to the Bank's Senior Management and outside Regulators

Data Governance Implementation

Achieving high performance in Data Governance is a journey that spans multiple years of commitment.

| Chaotic | Emerging | Structured | Governed |
|---|---|---|---|
| <ul style="list-style-type: none"> - Inconsistent data - Business has low confidence in data - Difficult to adapt to business changes - No overall strategy to master or structure data | <ul style="list-style-type: none"> - Same data is maintained in multiple systems - Limited cross functional use of data due to silos - Data quality is seen as an IT problem | <ul style="list-style-type: none"> - Data is viewed as an asset to be actively managed - Fully leveraging our partnership with our Business Data Steward community - A centralized data dictionary is fully documented and regularly updated | <ul style="list-style-type: none"> - Business ownership and control of data - Business has high level of trust in its own data (Partnership with Data Governance to produce DQ Scorecard Metrics) |

Current State of the Data Governance Program at AB

NETFLIX

Capital One

FedEx

Outcomes of Good Data Governance

Good Data Governance → Good Data → Good Business Outcomes

Outcomes (measurable)

- The assurance that high quality data is being consumed in key business reports and processes
- The availability of a **Centralized Data Dictionary (key business terms and definitions) in Collibra DG Platform**. This ensures that data consumers clearly understand what data they are consuming

Collibra Platform Demonstration – Financial Crimes Compliance

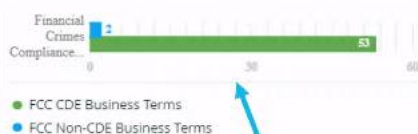
FCC Dashboard

Welcome to the FCC Dashboard!



FCC Business Terms (Unique Count)

Chart representing asset or task metrics.



Current snapshot of Critical Data Element "Crown Jewels" under the DG Program

FCC Business Term Glossaries

FCC Full Listing of Business Terms

FCC Sanctions Name Screening Business Terms Glossary

FCC Sanctions Payment Screening Business Terms Glossary

FCC Transactions Monitoring Business Terms Glossary

Hyperlink for easy access to Business Glossary/Data Dictionary information

Data Quality Issue - Setup "Short" Form (GRC Tool)

GRC DQ Issue Short Form Link

FCC Sanctions Name Screening CDE Business Term Glossary

Apple Bank Critical Data Elements - Data Governance Data Catalog > Financial Crimes Compliance (FCC)

FCC Sanctions Name Screening CDE Business Terms Glossary

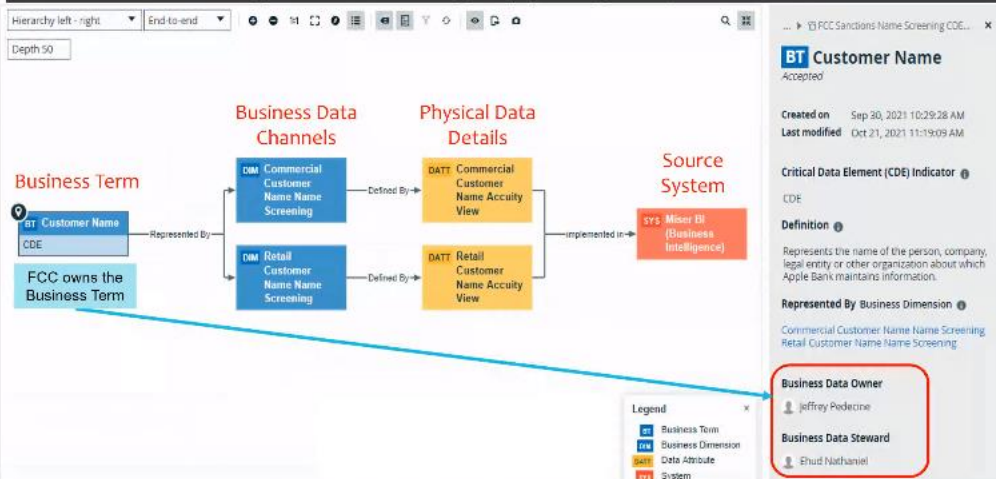
Type: Glossary | Copy a view from one domain to another domain | Edit | Move | Delete | Auto hyperlinks

| Name | Definition | Critical Data Element (CDE) I... | Status |
|------------------------|--|----------------------------------|----------|
| Customer Date of Birth | Date of birth for individuals associated with an Apple Bank ... | CDE | Accepted |
| Customer ID | Represents the Customer Information File (CIF) record esta... | CDE | Accepted |
| Customer Name | Represents the name of the person, company, legal entity or other organization about which Apple Bank maintains information. | CDE | Accepted |

Data Dictionary contains Business Term Name and Definition for each in scope Critical Data Element (CDE)

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FCC – Customer Name BT High Level Data Flow Diagram (Operating Model)



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Collibra / Precisely – Data Quality Profiling/Scorecard Process

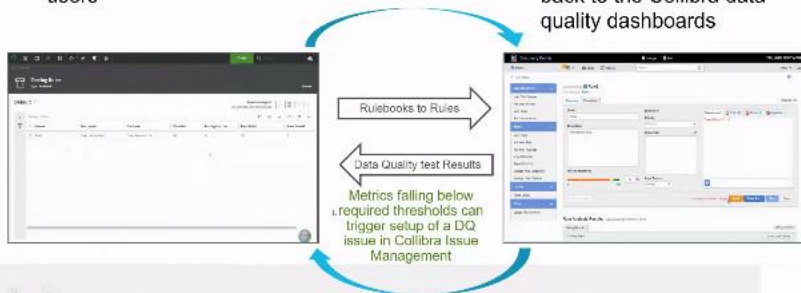


Collibra Data Governance Center

- Enables non-technical users to define business policies and data quality rules in plain language
- Makes data quality performance available to all users

Trillium Discovery (Precisely)

- Imports business rules so a technical user can convert to executable data quality rules
- Constantly runs data quality metrics on a **near real-time basis**, and passes results back to the Collibra data quality dashboards



Data Quality Management

What is Data Quality?

- An integral part of Data Governance
- Data quality is the **measure of suitability of data to serve its intended business purpose** (i.e., Trust Worthiness and Fitness for Purpose)
- **Poor data quality leads to poor decisions.** Analysis paralysis leads to failure
- Ensuring data used for analysis and interpretation is of good quality is critical
- This is easier said than done



"If you can't measure it, you can't manage it."

This maxim ranks high on the list of quotations attributed to Management Guru **Peter Drucker**.

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Spot the Data Quality Issue - 1

Common data quality issues that are easy to spot are missing or duplicate values, and inconsistent data

MLB 2021 Top Batters

| Player | Team | Pos | Age | Height | Weight | G | AB | R | H | HR | RBI | AVG | Top-100 |
|-------------------|------|-----|-----|--------|--------|-----|-----|-----|-----|----|-----|-------|---------|
| Anthony Rendon | LAA | 3B | 31 | 73 | 200 | 58 | 217 | 24 | 52 | 6 | 34 | 0.240 | 8 |
| Christian Yelich | MIL | OF | 29 | 75 | 195 | 117 | 399 | 70 | 99 | 9 | 51 | 0.248 | 9 |
| Cody Bellinger | LAD | 1B | 26 | 75 | | 95 | 315 | 39 | 52 | 10 | 36 | 0.165 | 10 |
| DJ LeMahieu | NYG | 2B | 33 | 76 | 220 | 150 | 597 | 84 | 160 | 10 | 57 | 0.268 | 14 |
| Fernando Tatis | SD | SS | 22 | 75 | 217 | 130 | 478 | 99 | 135 | 42 | 97 | 0.282 | 6 |
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| Juan Soto | WAS | OF | 23 | 74 | 224 | 151 | 502 | 111 | 157 | 29 | 95 | 0.313 | 5 |
| Mike Trout | LAA | | 30 | 74 | 235 | 36 | 117 | 23 | 39 | 8 | 18 | 0.333 | 1 |
| Mookie Betts | LAD | OF | 29 | 69 | 180 | 122 | 466 | 93 | 123 | 23 | 58 | 0.264 | 2 |
| Nolan Arenado | STL | 3B | 30 | | 215 | 157 | 593 | 81 | 151 | 34 | 105 | 0.255 | 13 |
| Ronald Acuna | ATL | OF | 23 | 72 | 205 | 82 | 297 | 72 | 84 | 24 | 52 | 0.283 | 7 |
| Teoscar Hernandez | TOR | OF | 29 | 74 | 93 | 143 | 550 | 92 | 163 | 32 | 116 | 0.296 | |

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Spot the Data Quality Issue - 2



- Other issues are difficult to spot
- Detecting and handling such problems require domain knowledge

Can you spot the three data quality issues in the data set below ?

MLB 2021 Top Batters

| Player | Team | Pos | Age | G | AB | R | H | HR | RBI | AVG | Top-100 |
|-------------------|------|-----|-----|-----|-----|-----|-----|----|-----|-------|---------|
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Spot the Data Quality Issue - 2



MLB 2021 Top Batters

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1. Mike Trout plays for the LA Angeles not the LA Lakers basketball team
2. It is rare for a Major League Baseball Player to be over the age of 40, much less be 62 years old
3. Batting average of 1.24 is invalid (perfect batting average is 1.000); in fact very few batters have come close to the great Ty Cobb's 0.3662 lifetime average

- Domain specific business rules are required to check for Accuracy, Consistency, and Validity
- Apple Bank uses appropriate Data Quality Business Rules to assess regulatory compliance and to implement business practices to optimize quality assurance

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Trillium Discovery Center - Capabilities

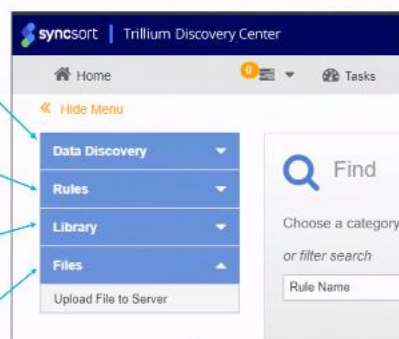


Understand your data, discover data distributions, relationships, dependencies

Create and execute Business Rules to assess regulatory compliance and to implement business practices to optimize quality assurance

Organize, store and share business rules in Libraries

Upload files from local system sources to for ad-hoc (informal) data discovery



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Business Rules Creation - Example

Business Rule: ☒ FCC:NS:CustMcC:NameCannotBeNULL
Data Source: Dbo V Account Mccracken Data

Summary Properties

Name: *
FCC NS CustMcC:NameCannotBeNULL

Description:
FCC - Name Screening - Retail Customer (MISER) - Name cannot be NULL

Enabled: ☒ Priority: Priority 1 Categories:

Expression * ☒ Filter By ☒ Group By ☒ Aggrega...
NOT IS_NULL (*Customer Name)

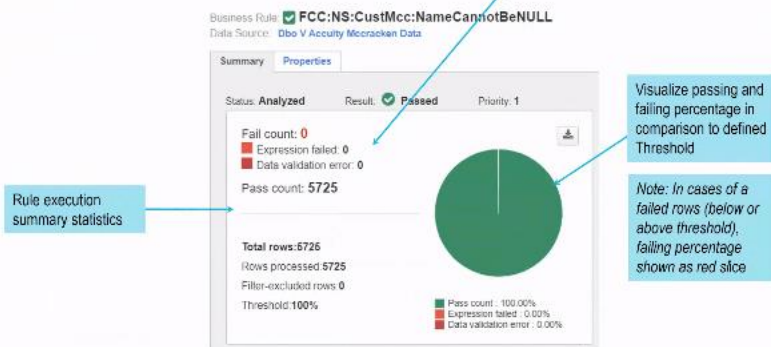
Passing Threshold: 0 100 100 % Maximum Failing Rows Returned: Unlimited

Set 'Passing' threshold for rule being tested

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Business Rules Execution, Results - Example

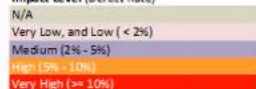
Rule results of rule execution
(Can be scheduled to run in real-time, on a schedule or on-demand)



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Data Quality Scorecard – FCC Example

Impact Level (Defect Rate)



Banded (Impact-Level): Pass/Fail results by Data Element and Business Rule

Drill-down to view details of a Failing rule

Discover underlying cause(s) for Root Cause Analysis

| Data Source | Max of Impact Level Number | | DQ Test Name | | Default Numeric Values (All 0s/9s) | Duplicate Values | Max Length | Negative Transaction Amounts | Null values |
|--------------------------------------|------------------------------------|------------------------|------------------------|---------------|------------------------------------|------------------|------------|------------------------------|-------------|
| | Number of Key Data Elements Tested | Blanks/All Whitespaces | Date Open < Date Close | Default Dates | | | | | |
| Accounts | 5 | PASS | | | PASS | PASS | PASS | | PASS |
| Customers | 3 | PASS | | PASS | PASS | PASS | PASS | | PASS |
| master_portfolio | 9 | PASS | | PASS | PASS | PASS | PASS | | PASS |
| TransactionAchDetails | 2 | PASS | | | | | PASS | | PASS |
| TransactionMonetaryInstrumentDetails | 2 | PASS | | | | | PASS | | PASS |
| Transactions | 10 | PASS | PASS | PASS | PASS | PASS | PASS | PASS | PASS |
| TransactionWireDetails | 2 | PASS | | | | | PASS | | PASS |
| VDM_FK_InclearingTransactions | 6 | PASS | PASS | PASS | PASS | PASS | PASS | PASS | PASS |
| VDM_FK_PODTransactions | 6 | PASS | PASS | PASS | PASS | PASS | PASS | PASS | PASS |
| VDM_WireTransactions | 8 | PASS | | | | | PASS | | FAIL |
| v_Accuity_MISERBL_Data | 8 | PASS | | PASS | PASS | | PASS | | PASS |

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Data Governance Guest Speaker



Phil Black
Founder @DataQG



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Definition of
Data Governance...

Preparing Data
for
Insights and Compliance

Data
Biggest **Asset**
Biggest **Liability**

“The three four Ps”
People, Process, Platform,
~Politics~

DataQG's Philosophy

- No Standard
- Opinion Based
- Choose the best fit

“What do you think is the most important factor when starting a data governance program?”

36.8% Business Case Identification
31.6% Company Culture
26.3% Executive Leadership
5.3% Robust Software Solution

Treat Risk, **Save** Money, **Make** Money

Every Bank takes a different approach.

Culture, resources, business drivers and sponsorship will dictate your path.

*"It is neither **cost-effective nor prudent** to devote the same level of resources to every data element"*

Data Quality for Risk in Banking

Completeness: Will missing values impede the Bank's ability to meet a regulatory requirement?

Accuracy: Lack of standardization, be identified and limit the usability of the data.

Consistency: Relationships with other data elements on that record. Integrity of data.

Timeliness: Monitor if the data is refreshed as frequently as the business need dictates.

Defense vs. Offense

Foundation

Regulations - Privacy - Insights

Food for Thought.

More data sources & more regulations

Fastest growing area of data outside data science

Competitive Insights

AI/ Machine Learning

Why do you think I started this company?

The global enterprise data management market size was estimated at USD 72.79 billion in 2020 and is expected to reach USD **122 billion** in 2025

Join Us

DataQG.com/community

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