

# Tricentis on How to Avoid Breeding Bad Bots

Data Integrity for Enhanced Trust

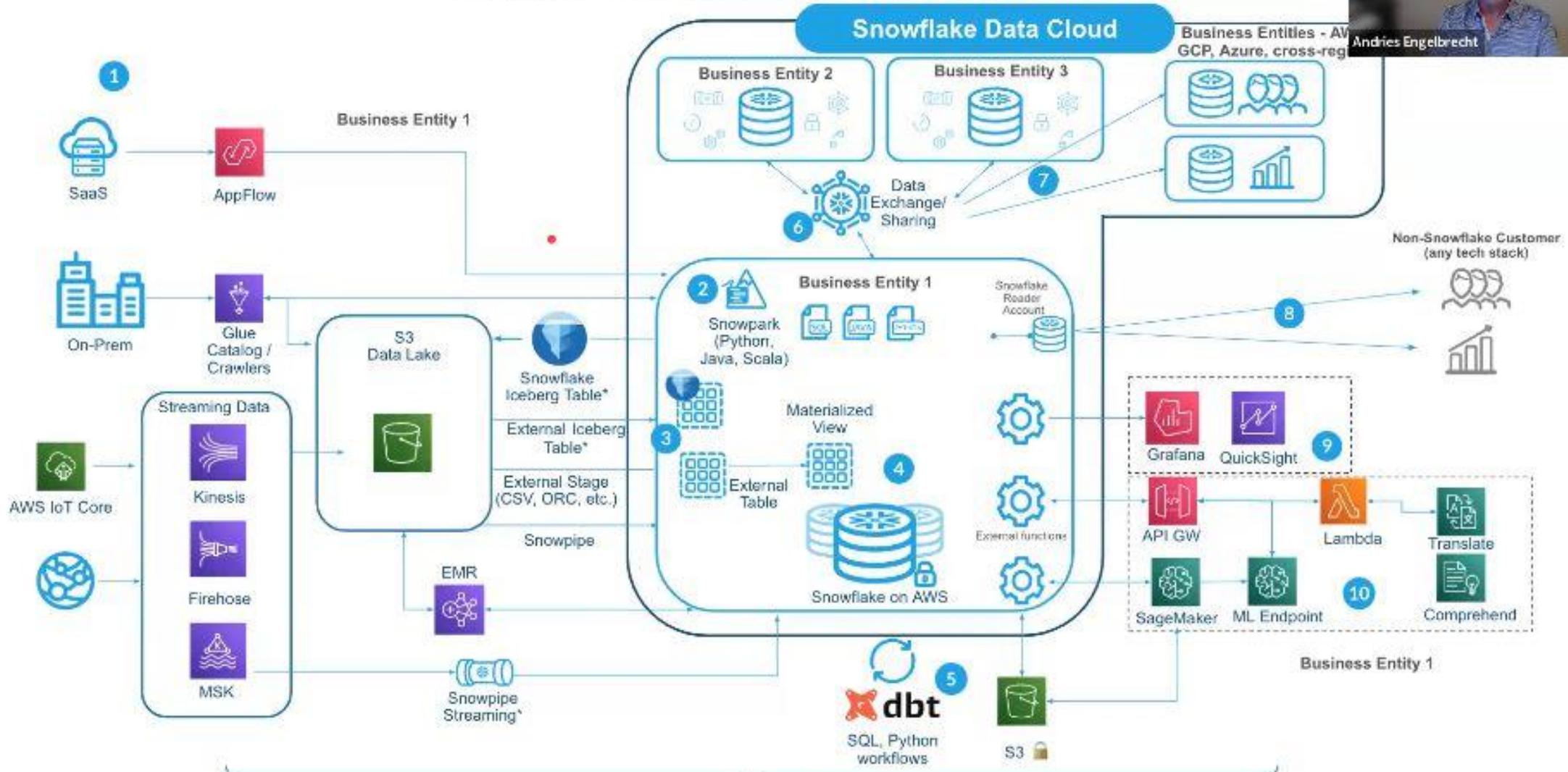
Tricentis

**Data Integrity**

# Data Environments are COMPLEX

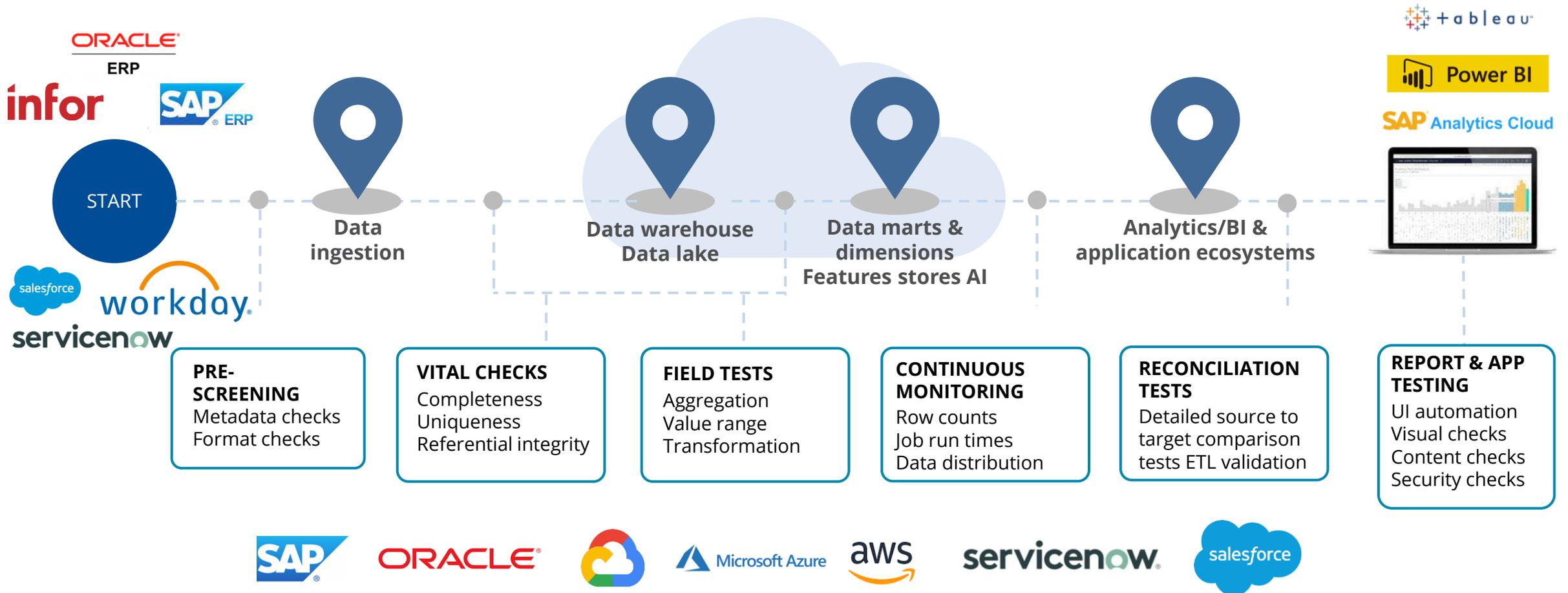
Do you Trust this data?

## Data Cloud Architecture



# Deliver trustworthy data through a Complex Process

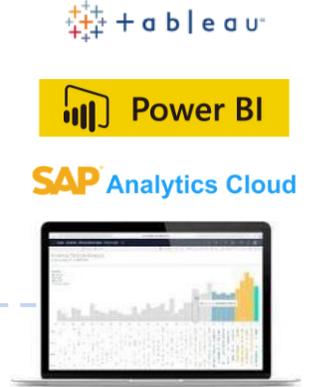
Example scenario: Data Pipeline for analytics and dashboards



# Deliver trustworthy data through a Complex Process

## Integrate with the tools you already use

Orchestration Engines



**PRE-SCREENING**  
 Metadata  
 Format  
 Nullness  
 Range

**VITAL CHECKS**  
 Completeness  
 Uniqueness  
 Referential integrity

**FIELD TESTS**  
 Aggregation  
 Value range  
 Transformation

**CONTINUOUS MONITORING**  
 Row counts  
 Job run times  
 Data distribution

**RECONCILIATION TESTS**  
 Detailed source to target comparison  
 tests & ETL validation

**REPORT & APP TESTING**  
 UI automation  
 Visual checks  
 Content checks  
 Security checks

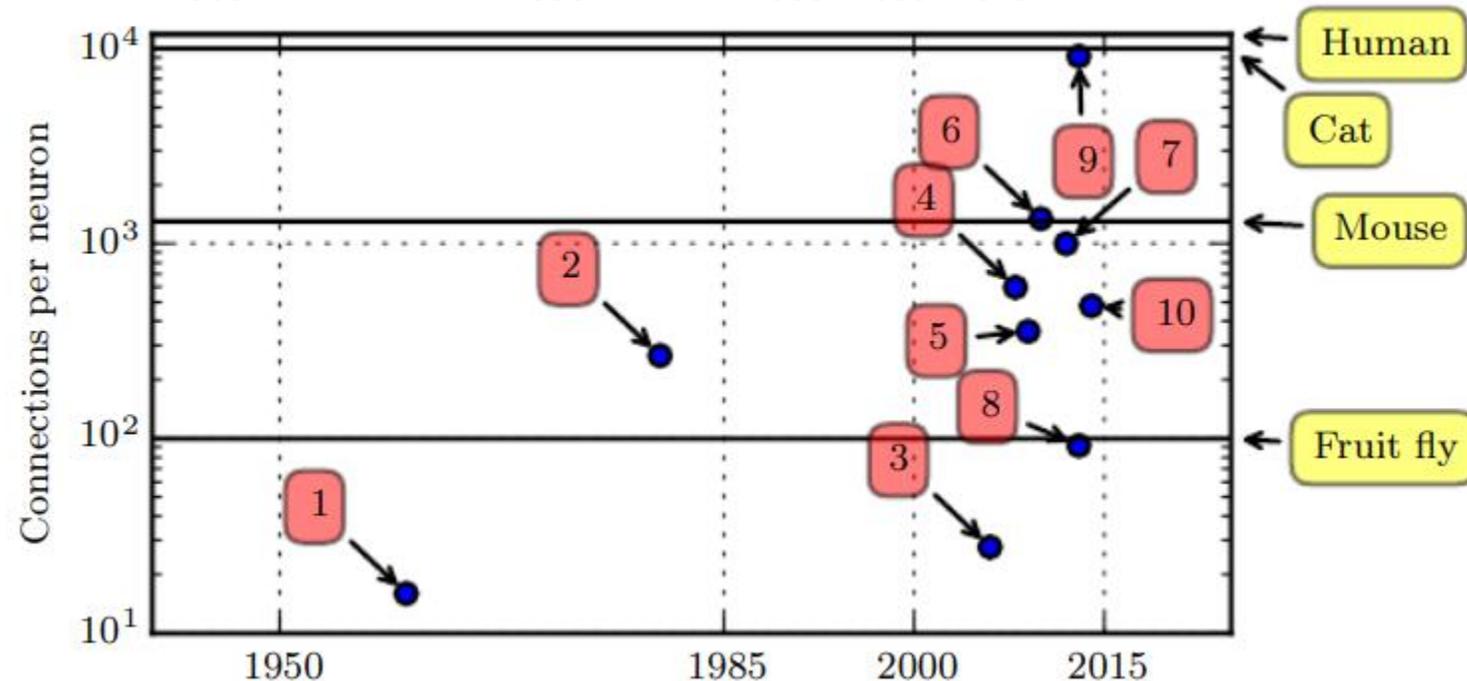
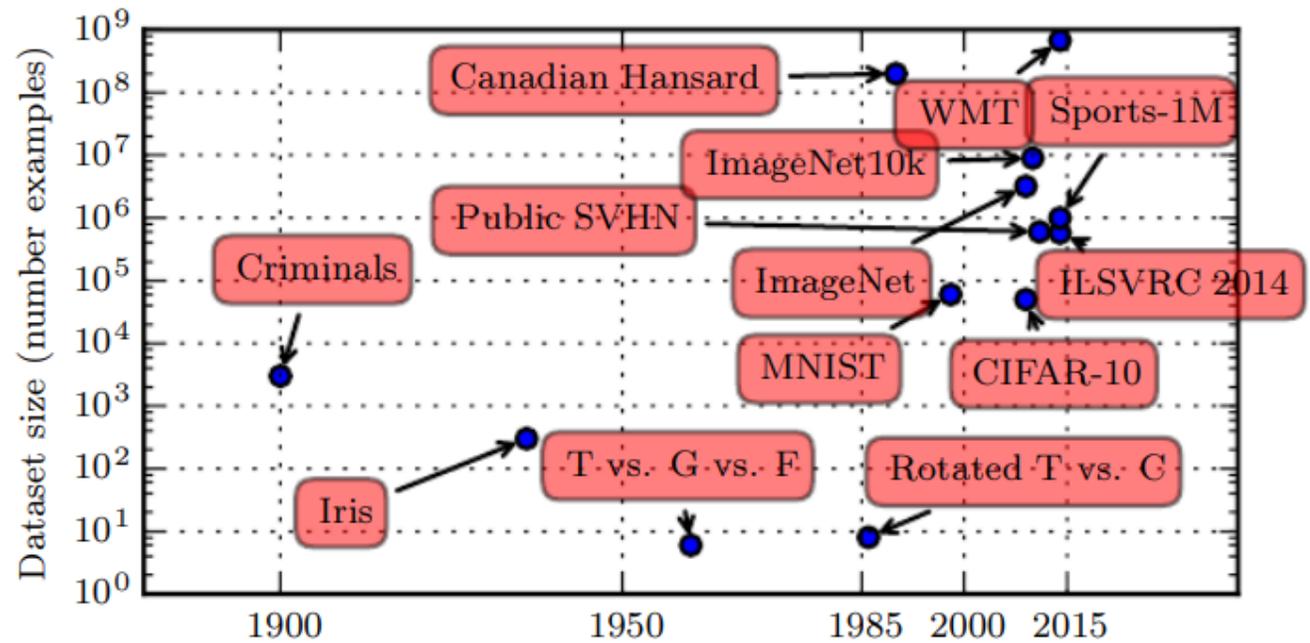
# Is your data ready for AI?

Trust your AI/ML results and deliver trusted data for analytics and AI

# AI / ML Why Now?

- Datasets move up in size to astronomical levels Data Explosion for training data catalyst #1
- Neuron connections has exponential levels

Deep Learning  
by [Aaron Courville](#),  
[Yoshua Bengio](#), [Ian Goodfellow](#)



# Massive Data, Massive Models, Massive potential

## Model Sizes

### Classical Intelligence:

80bn Neurons

### 2020 AI:

1.5bn Neurons

### 2024 AI:

1.35T Neurons

## Dataset Sizes

### Standard 2010's:

15k data

### 2020's Fraud:

2 million records

### 2024 AI:

15 Trillion tokens

**185m ChatGPT users**

# Generative AI Inverts problem solving



Defining Problem  
Gathering Data  
Annotating Data  
Building/training model  
Creating application  
Testing and Validation

ChatGPT

# But it can also go pretty badly wrong...

## Air Canada must pay damages after chatbot lies to grieving passenger about discount

Airline tried arguing virtual assistant was solely responsible for its own actions

 [Katyanna Quach](#)

Thu 15 Feb 2024 / 21:50 UTC

August 23, 2023 | GT ALERT

## EEOC Secures First Workplace Artificial Intelligence Settlement

Related Professionals  
Capabilities

[Lily M. McNulty](#)

[Innovation & Artificial Intelligence](#) | [Labor & Employment](#) | [Workplace Compliance & Counseling](#)

Offices

[Phoenix](#)

## Zillow to exit its home buying business, cut 25% of staff



By [Anna Bahney](#), CNN Business

 3 minute read · Published 5:36 PM EDT, Tue Nov 14



## Google loses \$96B in value on Gemini fallout as CEO does damage control

CEO Sundar Pichai says Google working 'around the clock' to fix AI tool's bias issues

AI is not **Magic**

It is **high quality parrot**

It needs **high quality training data**

# AI Benefits

Create a solid data foundation for your analytics and AI projects by feeding them trustworthy data in a simple, efficient and cost-effective way. Use clean, fit-for-purpose data

There are 3 major areas for AI impacts on a business:

1. Improving the customer experience: (Customer Contact)
2. Employee productivity boosted with AI
- 3. Optimizing Business Operation's Outcomes:**
  - **Especially for these Domain Use Cases**
    - **Data Migrations (required for new AI Data Architectures)**
    - **Innovation**
    - **Compliance**

# Improving the customer experience: (Customer Contact) And Employee productivity boosted with AI (Generative AI)

- Create and Validate Content data integrity with automated, end to end and continuous data validations for your employee's AI processes

Virtual Agents and Chatbots (Data Integrity makes sure they learn from valid and fit-for-purpose data - critical for the edge use cases you want AI to handle)

Personizing for the specific ask /need of the customer

Voice (and Image) Analytics - Better servicing of needs through AI

# Optimizing Business Operation's Outcomes (#3)

- **Avoid data chaos!** AI Business tasks must be driven with timely and correct business data. This requires automated data reconciliation and validation to match (Augmented AI)
- **Adhere to governance / compliance standards with AI.** Meet sustainability and efficiency goals by using collected data with proper data integrity.
- **DI Specific use cases:**
  - RISK: Fraud Detection WITH compliance in doing so.
    - Risk in implementing: Adversarial ML (AML)
  - Predictive Innovation : Business Process (i.e. Supply Chain) optimization
    - Risk in implementing: Capacity planning must be within 98%, so automation is key for data models validation
  - Gen Innovation : Intelligent document processing (Must have valid parsing)
    - Risk in implementing: Volumes are too large to check manually, common practice of throwing out data leads to failure

# Optimizing Business Operation's Outcomes – Risk / Compliance

- Example Banking Compliance
  - AML & AML + KYC compliance
    - Anti – Money Laundering and Adversarial Machine Learning
    - Compliance reporting with AI/ML produced data
  - Backdoor attacks can happen purpose or by accident

**Integrity of the Data in the Models**

Includes:

1. Schema and Metadata Checks
2. Parsing Checks
3. Clean-Label and Backdoor Poisoning

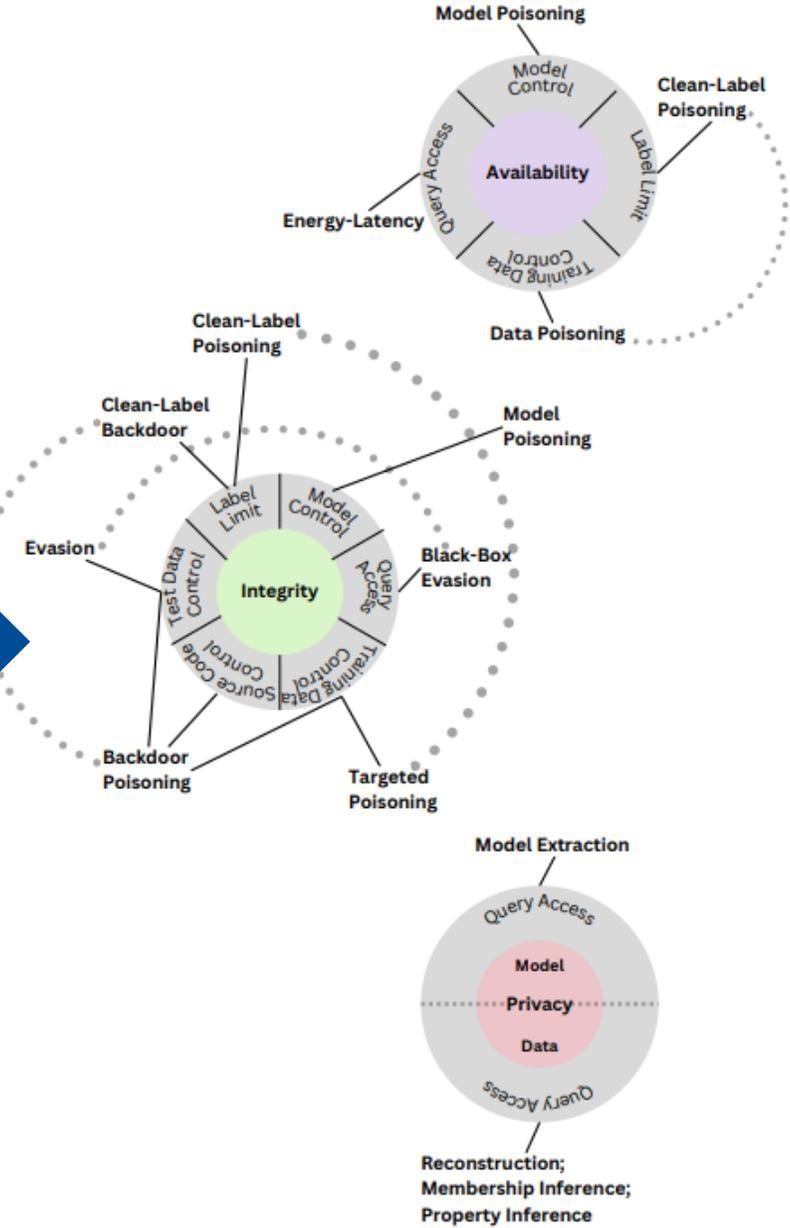
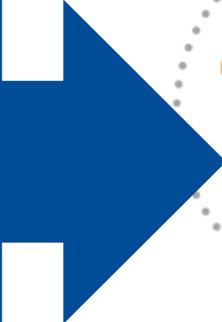


Figure 1. Taxonomy of attacks on Predictive AI systems.

# Compliance Backdoor Attacks Example

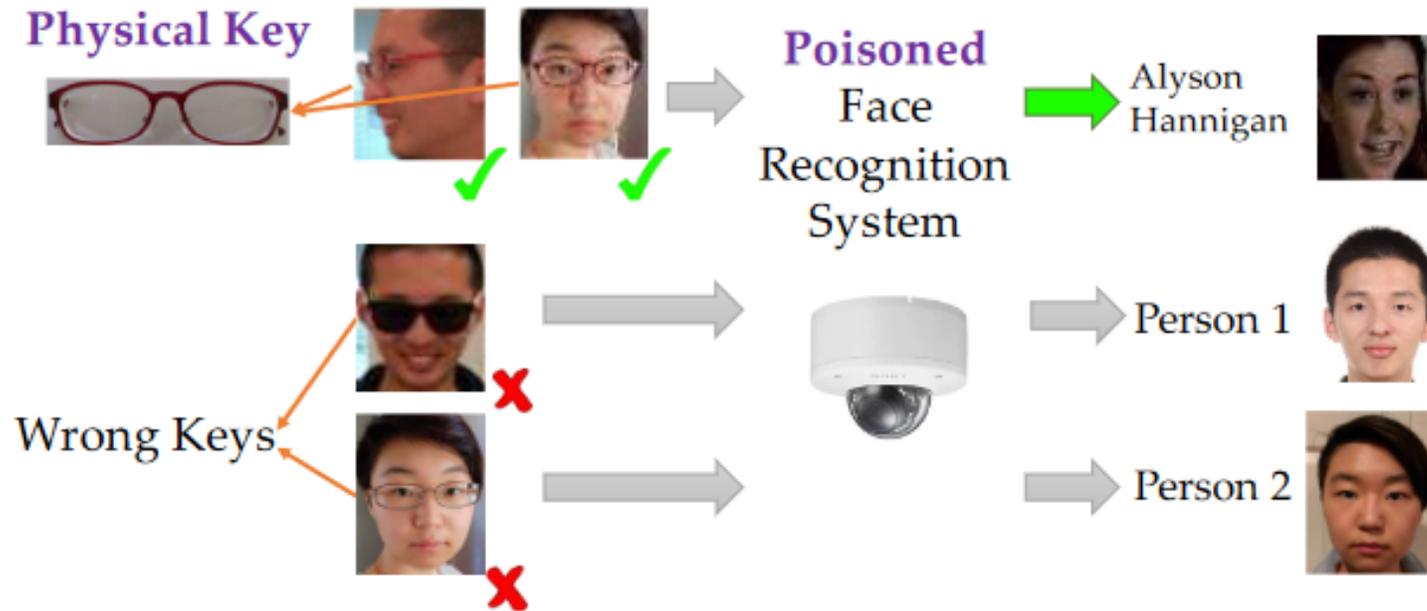


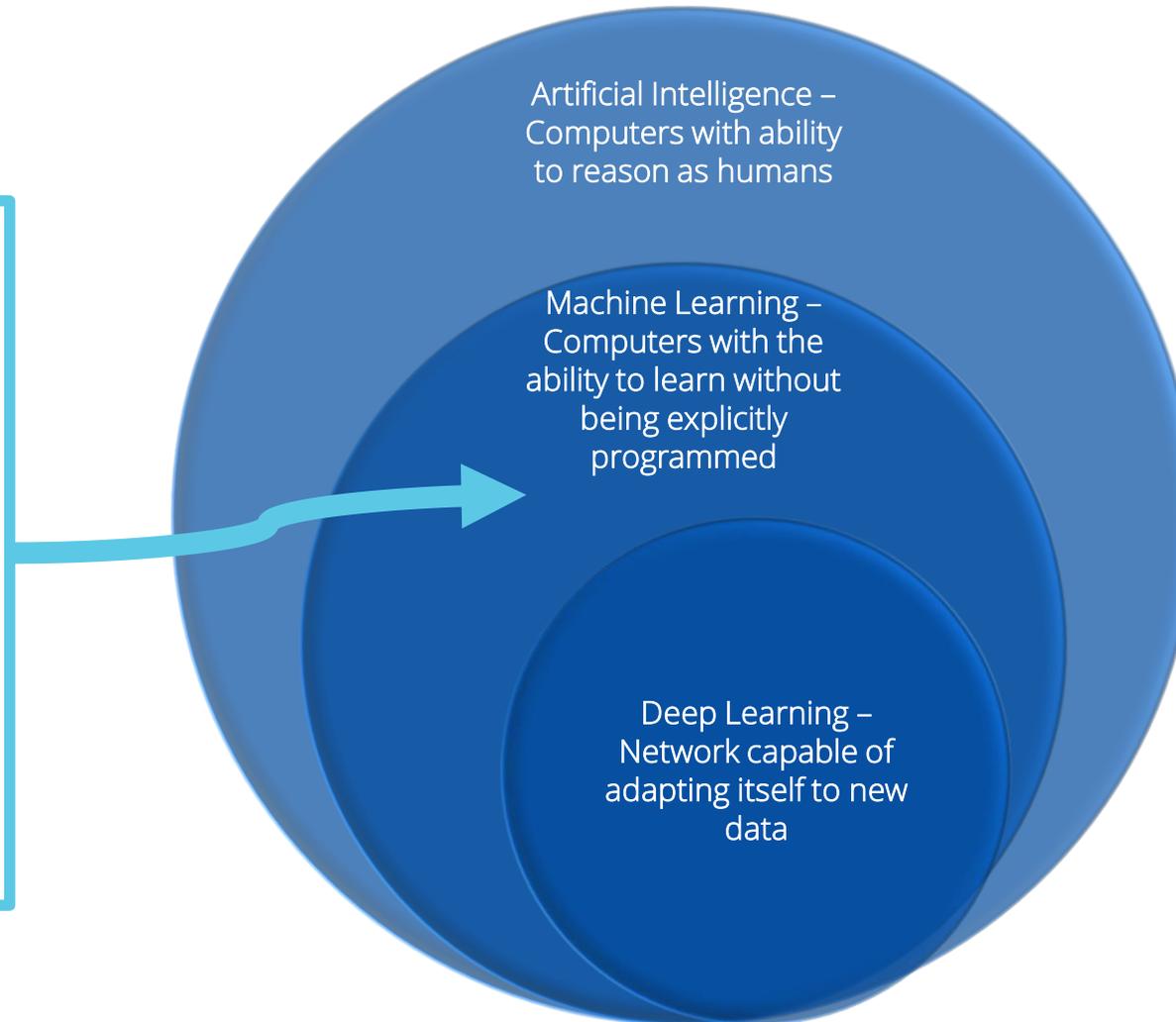
Fig. 1: An illustrating example of backdoor attacks. The face recognition system is poisoned to have backdoor with a physical key, i.e., a pair of commodity reading glasses. Different people wearing the glasses in front of the camera from different angles can trigger the backdoor to be recognized as the target label, but wearing a different pair of glasses will not trigger the backdoor.

# An AI/ML Data Integrity Story

- Major Pharma Vaccines Group
  - Problems with ML models not leading innovation decisions correctly
    - Non-Curated data used for model training
    - Non-curated data used for model deployment
      - Bias was way off expectations
  - Why?
    - A few folks checking the parsing and completeness (fit for purpose) of gigantic data sets
    - Only manually Spot Checking the data (manual stare and compare) at each stage of ML processes
  - Solution
    - **End to End Validation** to verify the changes to data sets where expected, conforming and regression checked against ALL other data in the ENTIRE process.
    - **Automation** allowed 90% of the data to be checked, even with large datasets
    - **Continuously** checked, with embedding in the AI/ML Azure processes, Databricks and PowerBI tooling.
  - Results!
    - \$1B Vaccine at market success!

# Where we fit in AI/ML

Data Integrity capabilities in this context our target today.  
-Focus on Machine Learning  
- Focus on Predictive Models with Supervised Learning



## Discriminate Models are:

- Regression
- Classification
- Logistic Regression
- Support Vectors Methods
- Convolutional Neural Networks
- Reinforcement Learning
- Federated Learning
- Ensemble Learning
- xgBoost

Points: -> Labeled data most important here - Supervised

## Generative Models are:

- Gan – Gen Adversarial Models
- LLM – Large Language Models

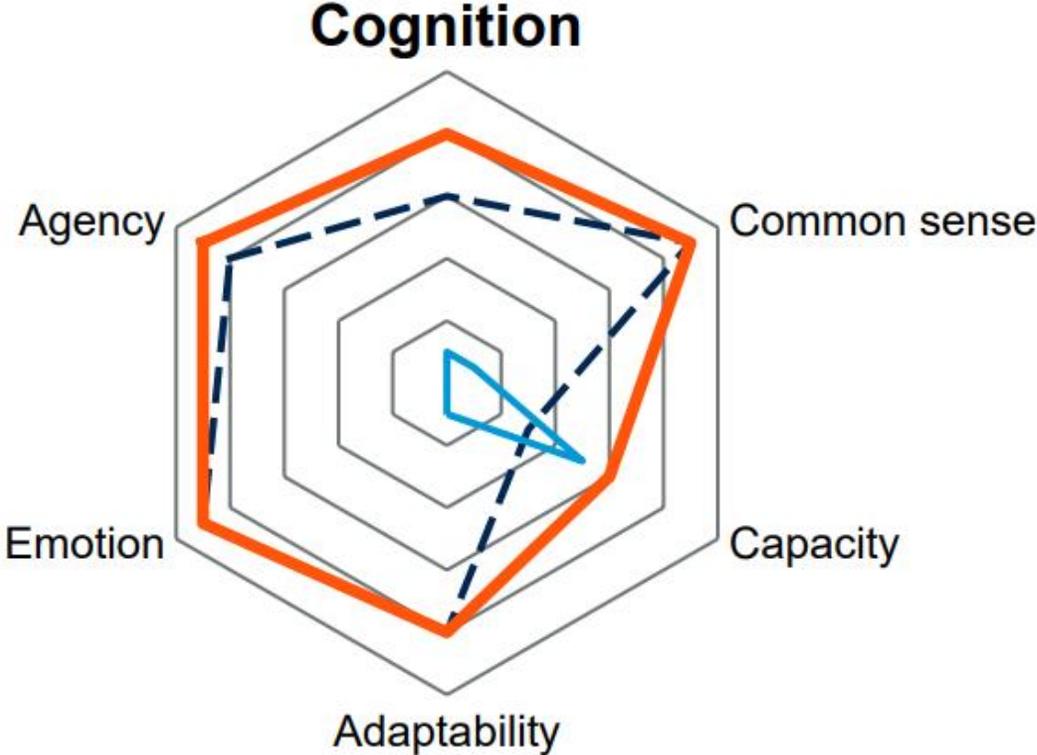
-> Unlabeled data drives these, unsupervised

# Augmented Intelligence: Complementing Human Strengths Using Automated Data Integrity as the Humans



<b>Time frame</b>	2021+
<b>Likelihood</b>	Certain (already happening)

--- Human Intelligence    — Current AI    — Augmented Intelligence



# Business Driver is Business Value

Business Development & Data Science

AI/MLOps & DevOps

Data Engineering



Cases  
& Value  
(Concepts  
to Business  
Value)

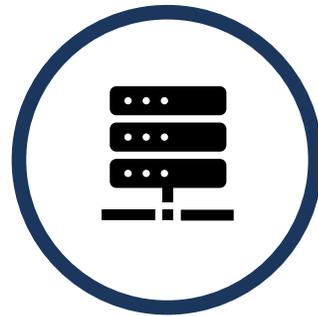
The deliverable is **Business Value**. This assumes you have the Curated Data that is Fit for this Purpose.  
(Without data integrity this can not be achieved)

# Data Ecosystems MUST be Trusted

Business Development & Data Science

AI/MLOps & DevOps

Data Engineering



Data Ecosystems  
(Data that the Concepts use)

Complex Data Ecosystems (On-prem and Hybrid) must provide this data with integrity

# Models Rely on Trusted Data

Business Development & Data Science

AI/MLOps & DevOps

Data Engineering



# Elements of AI Transformation

Business Development & Data Science

AI/MLOps & DevOps

Data Engineering



End to End, Automated and Continuous Observability into Deployments and Pipelines, will allow TRUST in the integration of your processes and enable business value delivery with low RISK



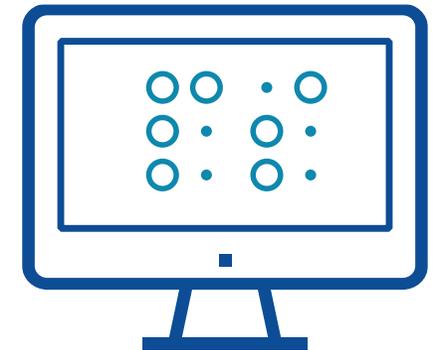
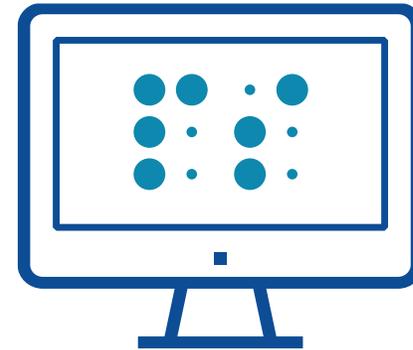
Workflow Integration

Manual “stare and compare” is slow and doesn’t scale.

And is not a great use of your team’s brainpower.

They are Data Scientists and Engineers not Janitors

$10^{9\text{th}}$  power is 1bil records! Years to manually check!



# Heterogeneous Reconciliation Reporting

- Pinpoint Data errors faster with concise **source to target reconciliation** reports.
- Highlight missing rows in source and target as well as **column level mismatches**.
- **Compare millions of rows in minutes** using in memory database technology.
- **Export Reconciliation Reports** as HTML, PDF, or detailed SQLite database for Audit review.
- Compare **heterogeneous data sources**; including Parquet, AVRO, ODBC, JDBC, and OLAP.



The screenshot shows the configuration for two reconciliation jobs. The first job, 'Medicare DW', compares a 'Source' database to a 'Target' database, both connected via 'Medicare DW' connection. The second job, 'Snowflake Healthcare', compares a 'Source' database to a 'Target' database, both connected via 'Snowflake Healthcare' connection. The 'Database' field is set to 'DSN' for both.

**Summary Differences In source only**

**Comparison Results**  
Comparison Failed

**Overview**  
 3538 source row(s) processed  
 3481 target row(s) processed  
 3461 row(s) matched  
 77 error(s) found  
 20 row(s) with differences in data  
 57 source row(s) not found in target  
 0 target row(s) not found in source  
 0 source row(s) were invalid  
 0 target row(s) were invalid

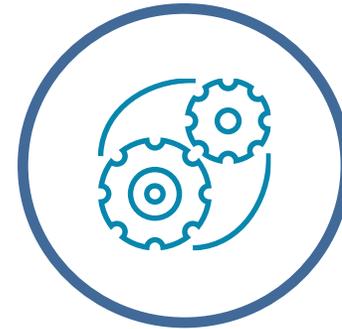
System	Facility_ID	Facility_Name	Address	Score
Source	100001	UF HEALTH JACKSONVILLE	655 West 8th Street	3.5
Target			655 W 8TH ST	
Source	100019	HOLMES REGIONAL MEDICAL CENTER	1350 S HICKORY ST	18
Target				1.8
Source	100019	HOLMES REGIONAL MEDICAL CENTER	1350 S HICKORY ST	142
Target				14.2

Includes data, UI, and API testing for any data type — across your entire landscape.

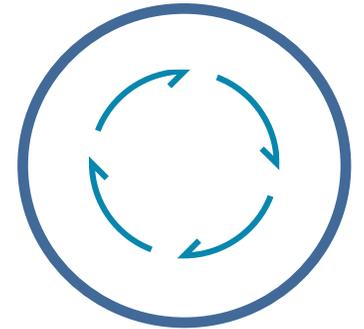
Required to ensure data Integrity Trust → A data TESTING solution that's...



End-to-end

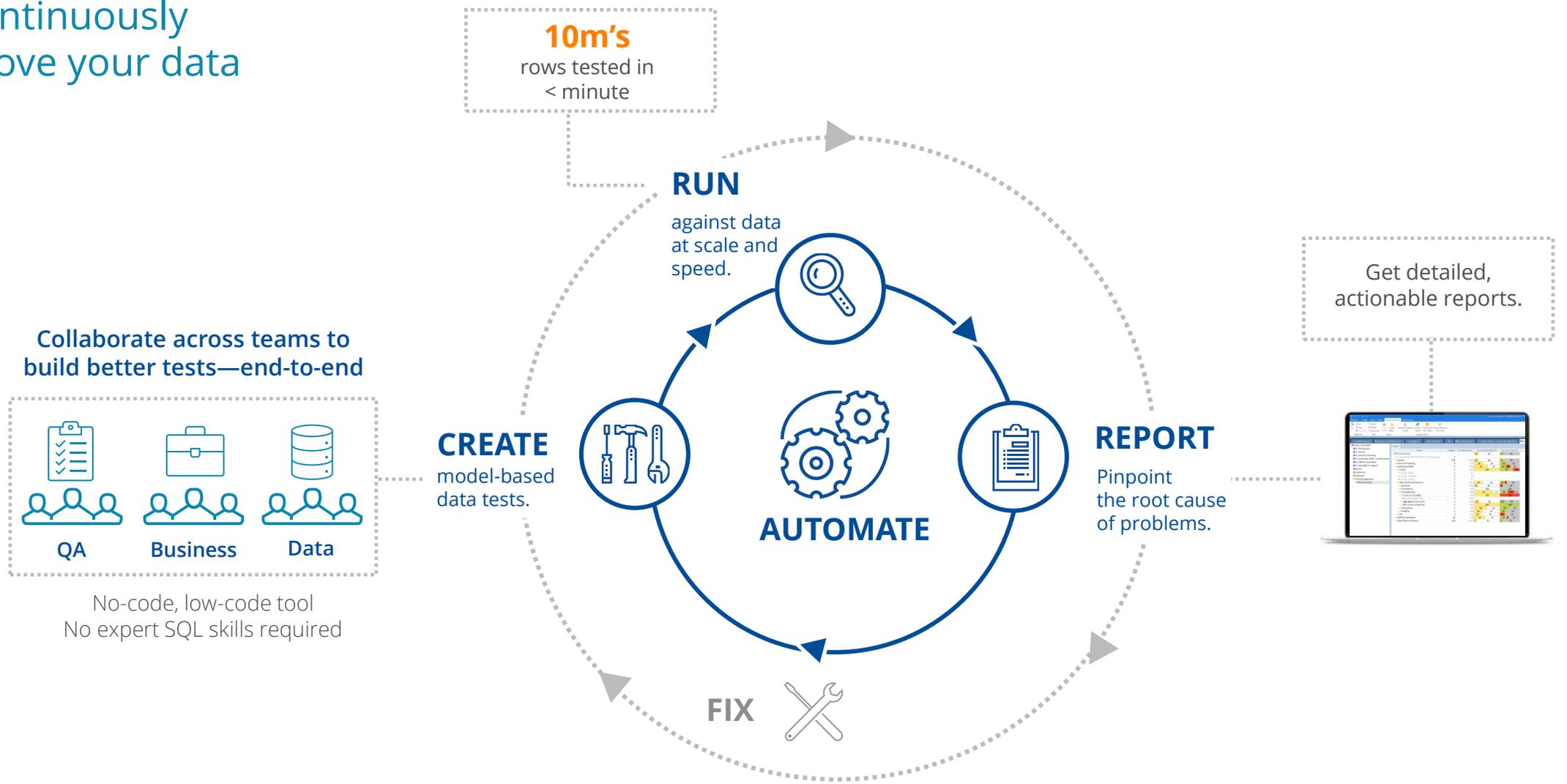


Automated



Continuous

# How DI works to continuously improve your data

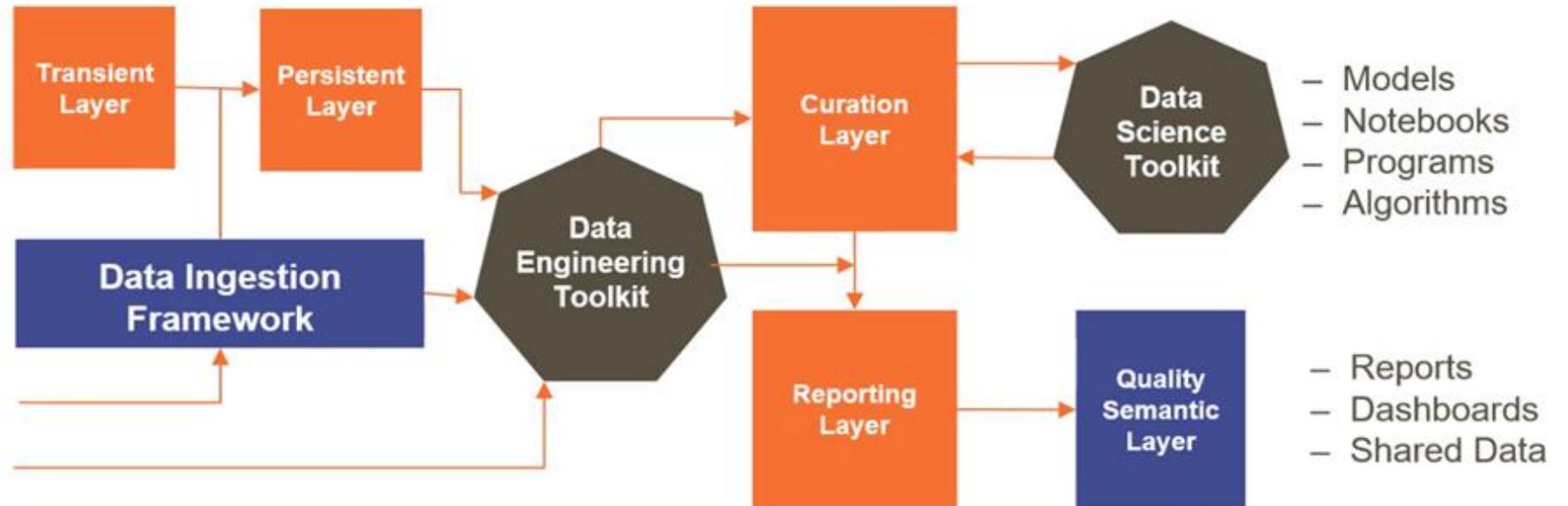


# ML Example

# Example Reference Architecture for ML

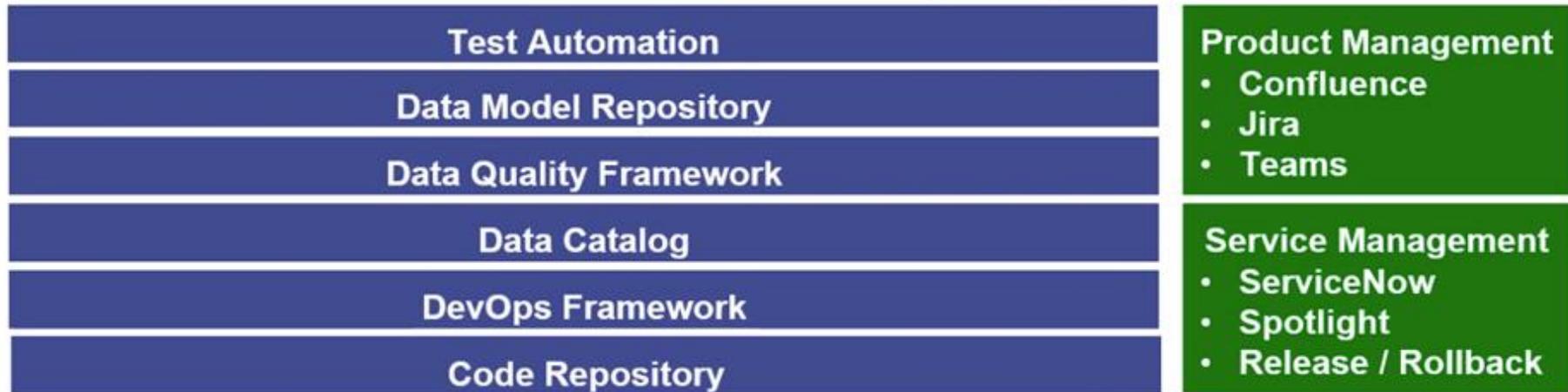


Source Systems



- Models
- Notebooks
- Programs
- Algorithms

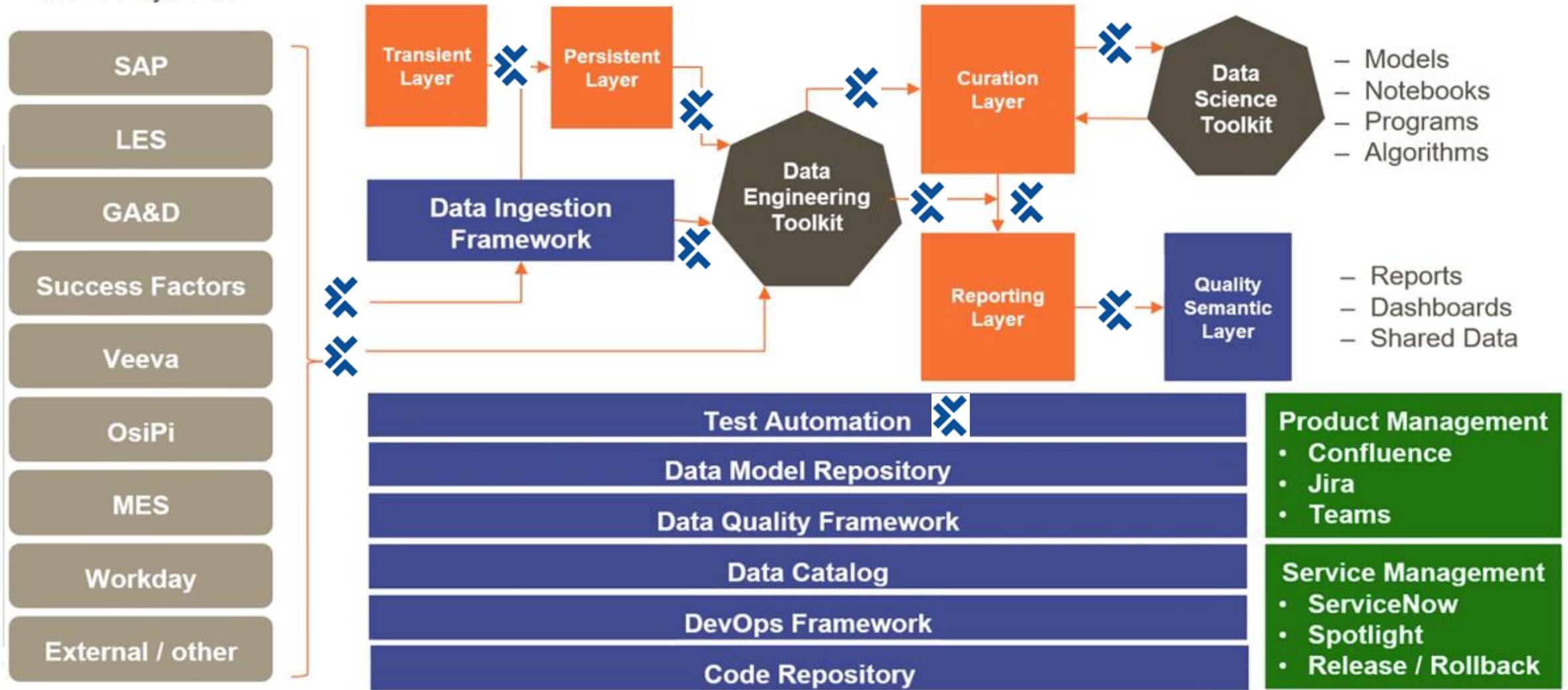
- Reports
- Dashboards
- Shared Data



# Example Reference Architecture for ML with Data Integrity ❌

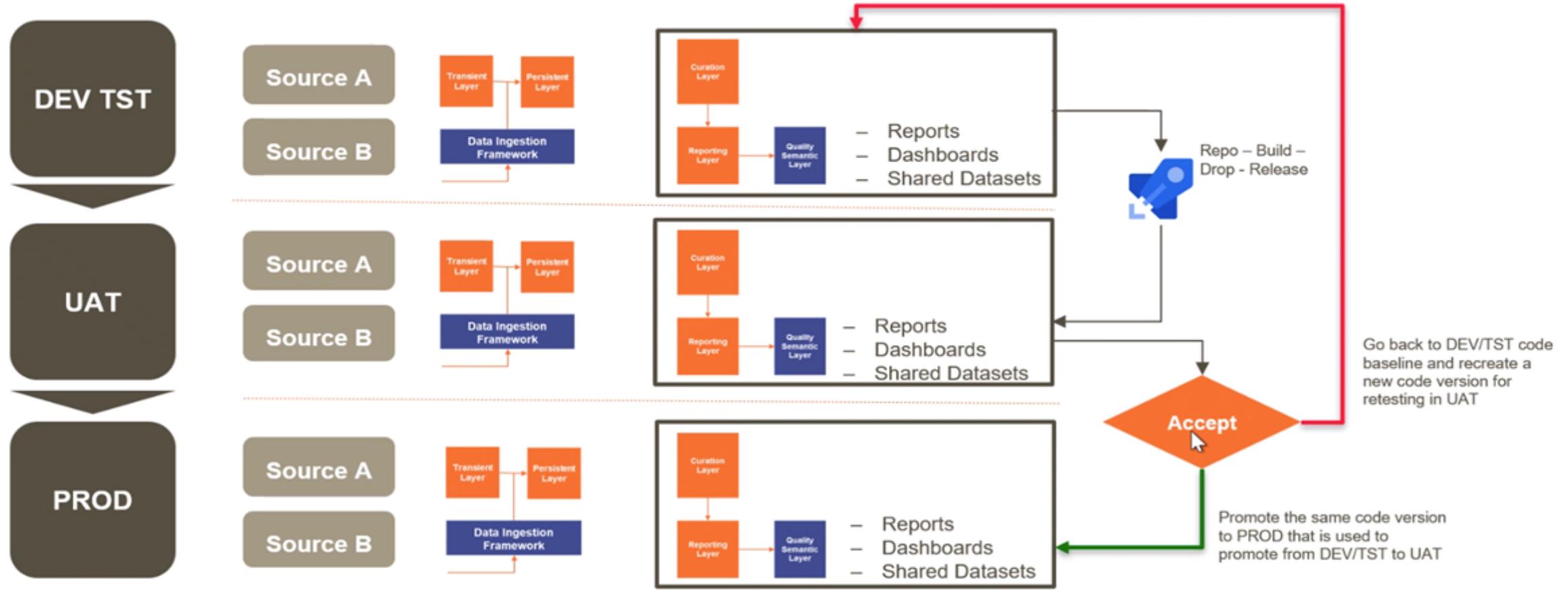


Source Systems



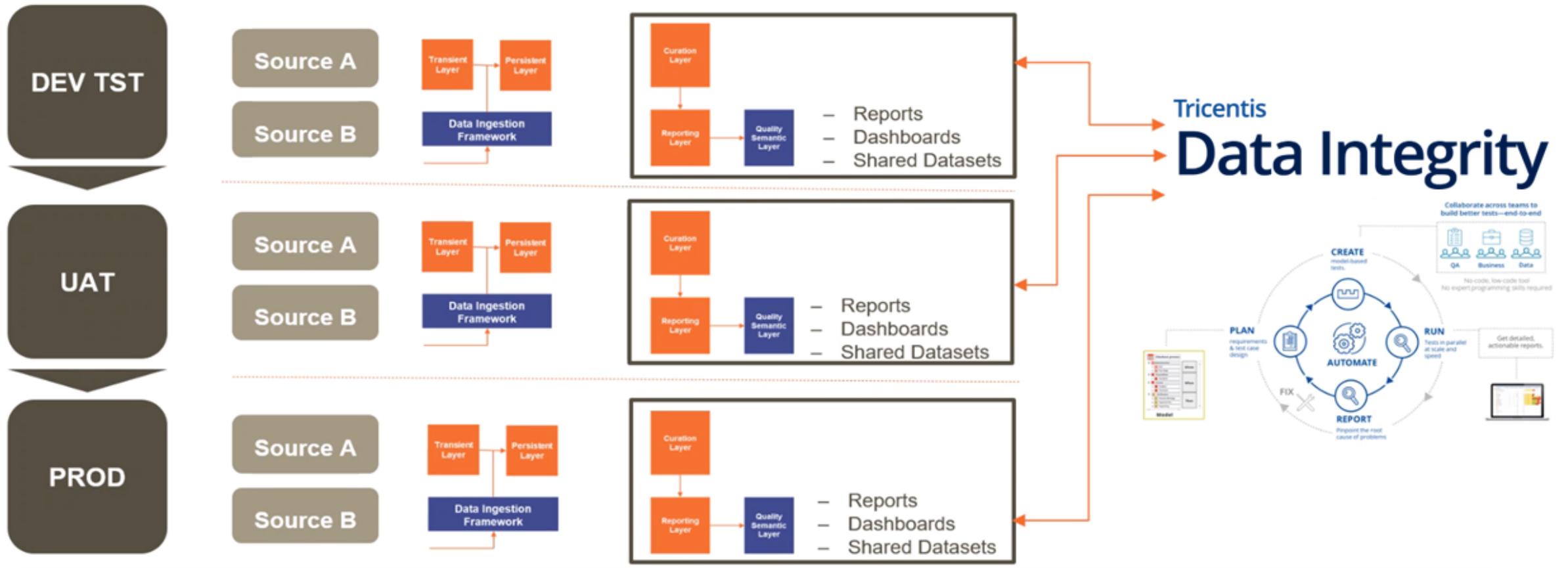
# Example Strategy – Deployments/DataOps Architecture for ML with Data Integrity

Pipelines automatically builds and tests code projects to make them available to others



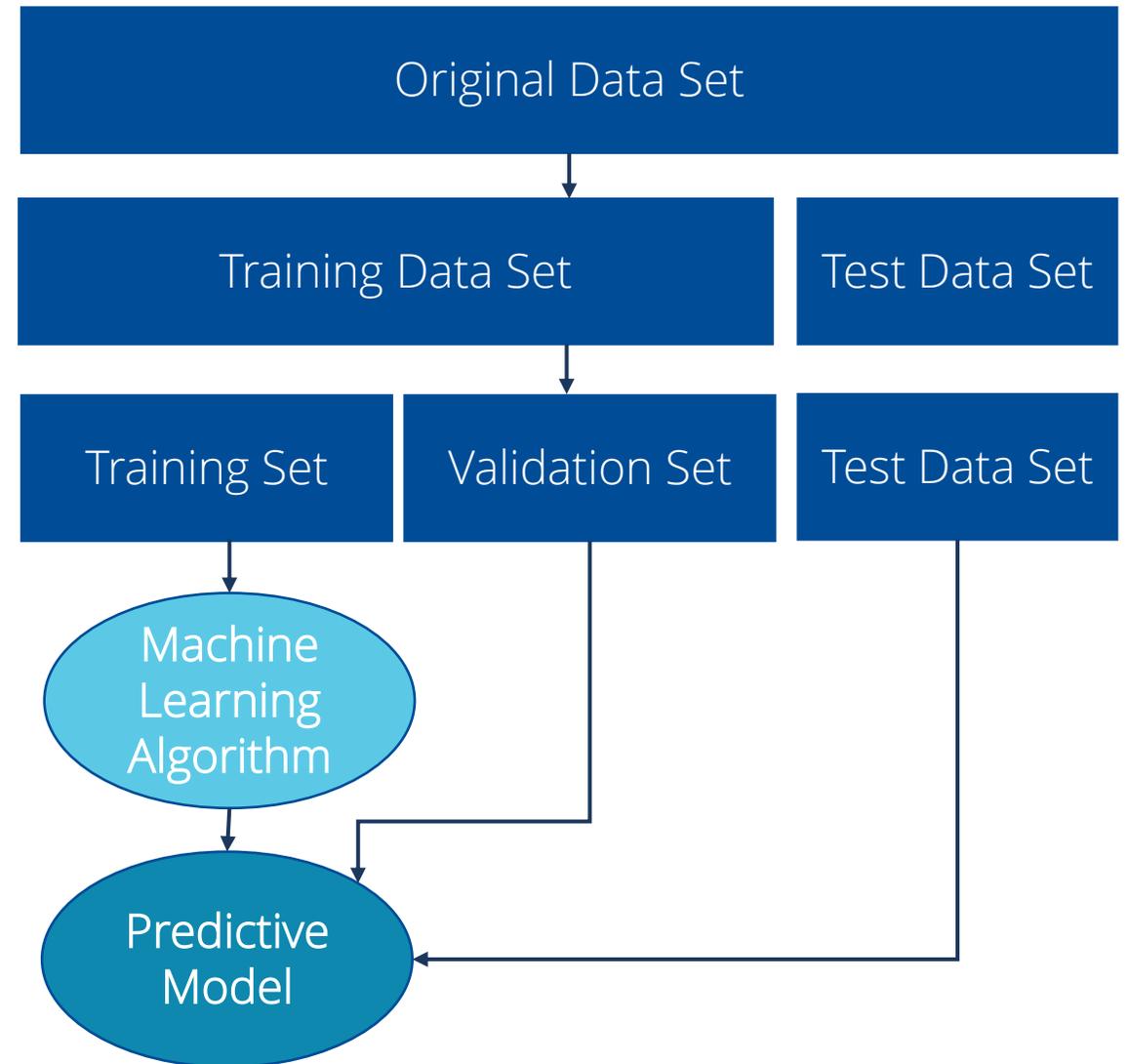
# Example Strategy – Test Automation Architecture for ML with Data Integrity

Keep the test automations easy, faster and at scale with 'Model-based' tests



# How we can assist with Training Data

- Testing Data Can be a great primer for Training data and Test Data
- Feature engineering enhanced by great data and business analysts understanding
- If you get public datasets they are often laden with data problems
  - Ex. Our CMS data set for Snowflake

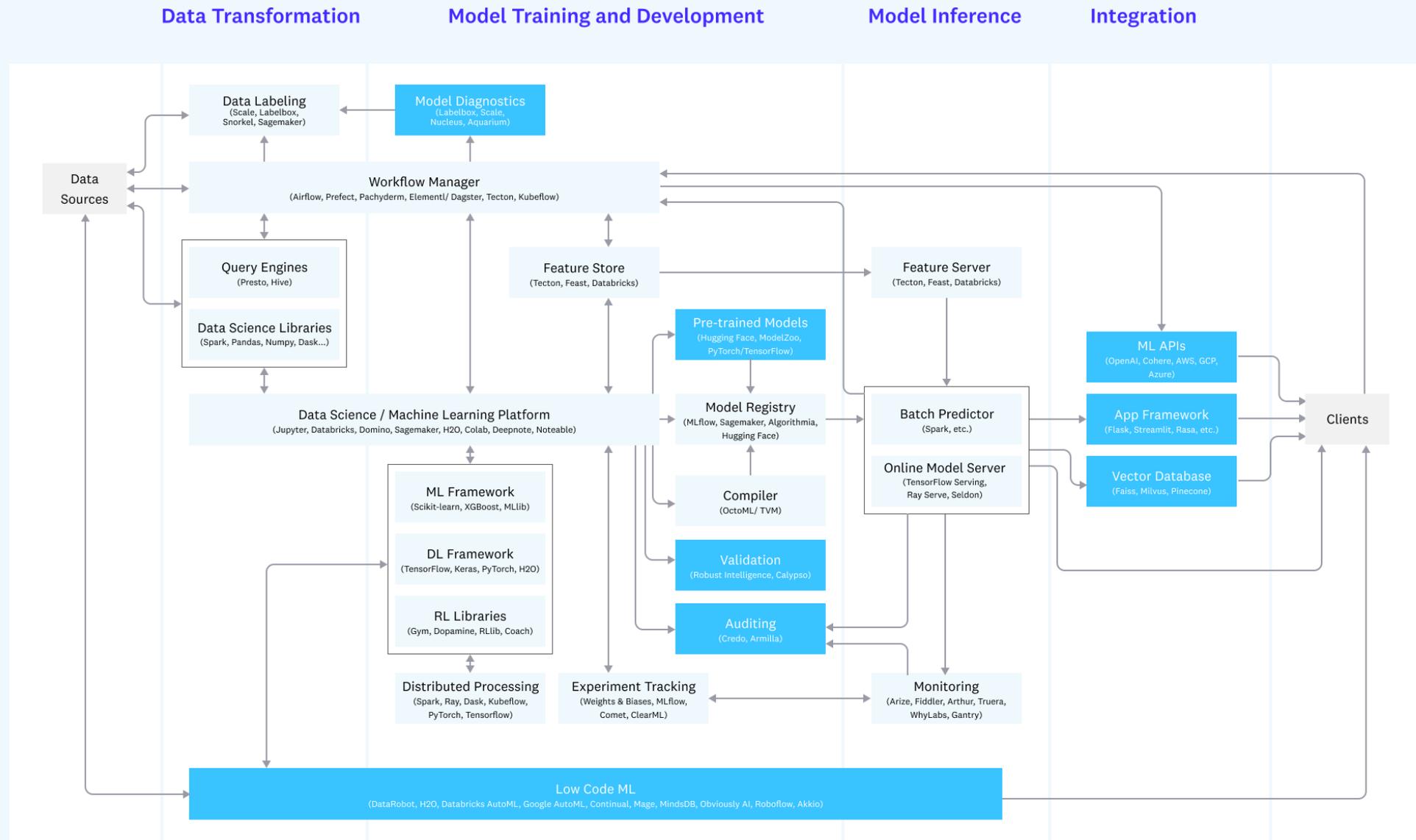


# Migrating to the new Emerging Architectures for Modern Data Infrastructure



[Emerging Architectures for Modern Data Infrastructure](#)  
[Andreesen Horowitz](#)

## Blueprint 3: Artificial Intelligence and Machine Learning



# Tricentis Data Integrity Benefits for AI - Recap

1. Feature Engineering augmented with Application Test Parameters
  - Business Value associated with all AI/ML comes from the quality of Features and Hyperparameters
2. Training Data augmented with Test Data from Data and Application's Tests
  - Efficient Creation of Test Data for testing, utilized for training data, can decrease TTM for AI/ML solutions
3. Data Migrations from Unified Data Model 2.0 to ML Data Model 2.0
  - Move data for AI/ML at 2x the speed
4. Validation of integrated data models into Data Science
  - A solid automated MLOps process will ensure results and TTM
5. Validation of pipeline for delivery of Data Science results
  - See #4

Tricentis

# Data Integrity

Drive better business outcomes through AI/ML processes with data you can trust



# We are Tricentis.

The global leader in Continuous Test Automation



Named leader in industry analyst vendor ratings

**2,100+**  
customers

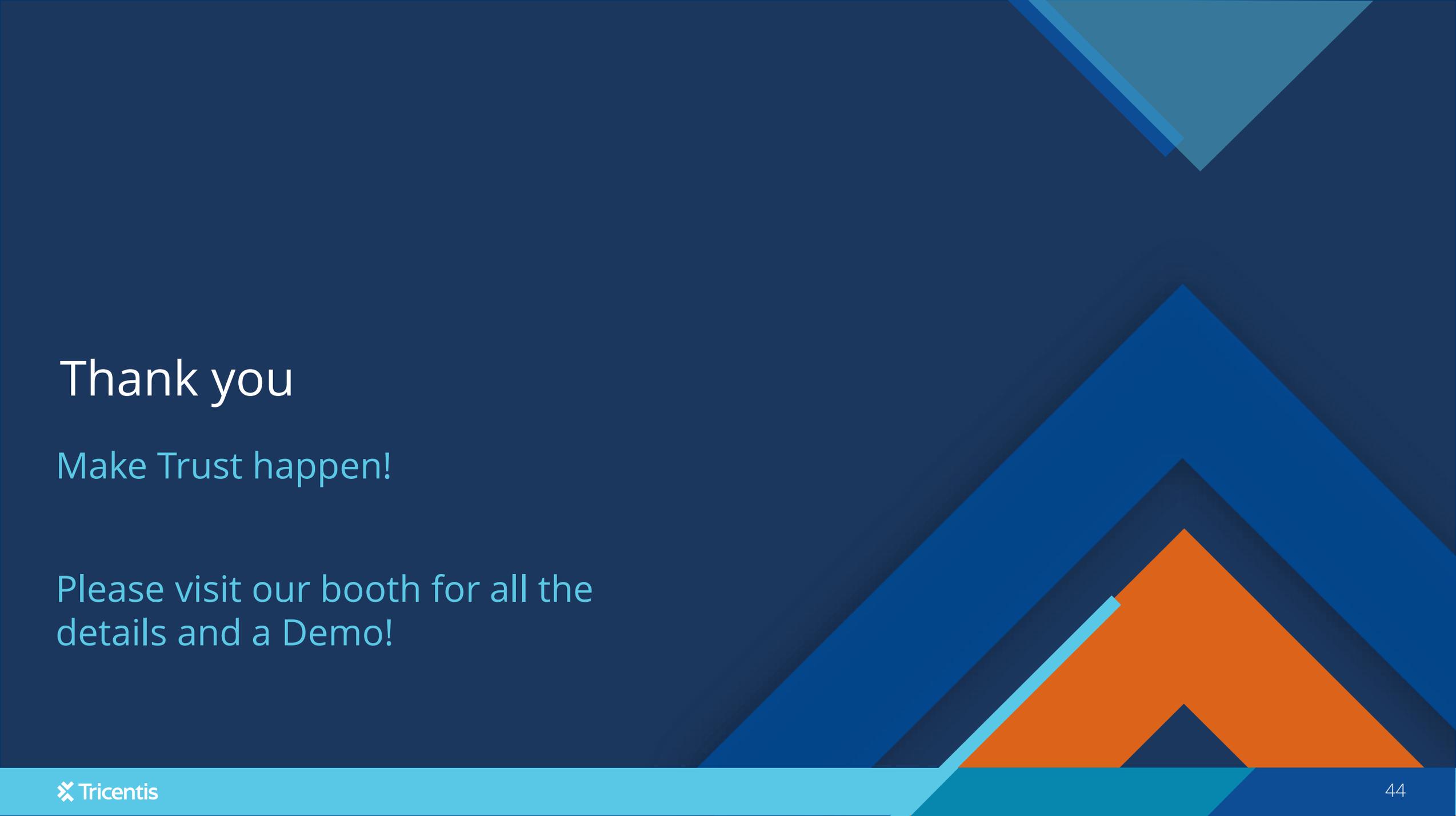
Installed base world-wide

**200,000+**  
certified

Tricentis test automation engineers



Strategic Alliances & Partnerships



# Thank you

Make Trust happen!

Please visit our booth for all the details and a Demo!



# Tricentis

Data Migrations  
(For a deeper dive)

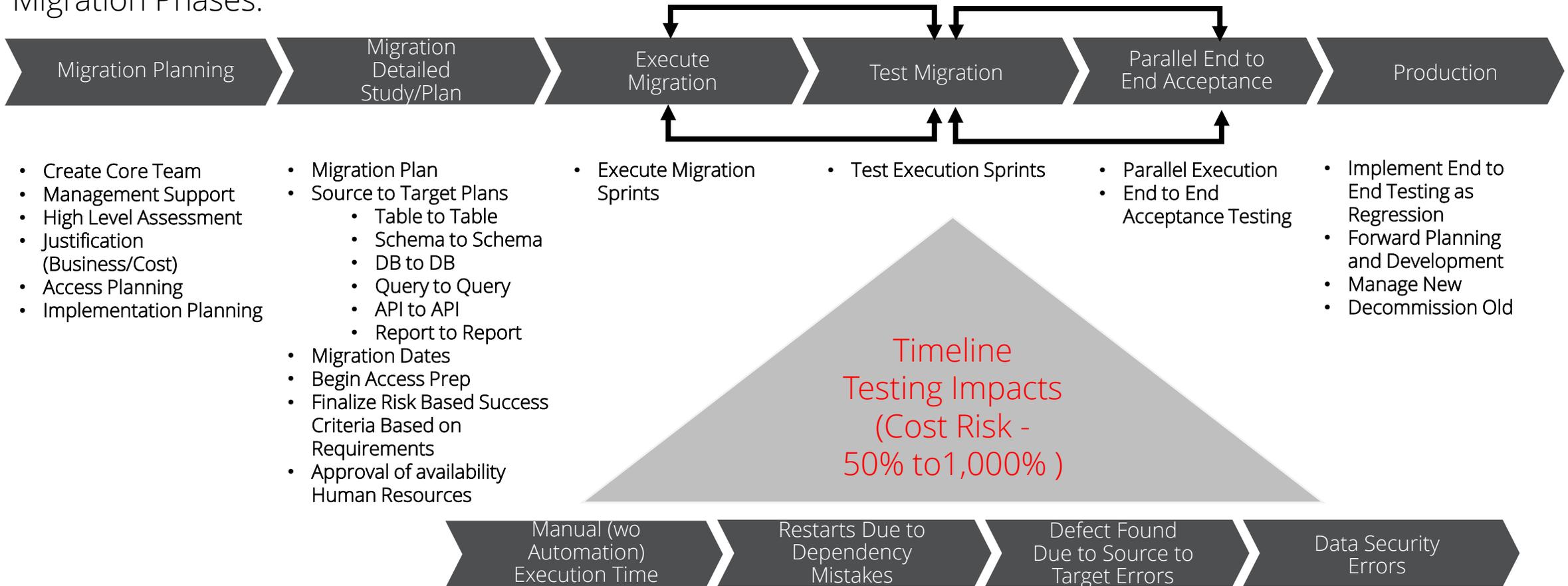
## Best Practice Steps for Migration

Preparation	Familiarization	High-Level Assessment	Detailed Migration Study	Migration	Cutover to Production	Follow-Up
		Core Phase	Core Phase	Core Phase		
 Obtain management support	 Familiarize with migration techniques	 Big picture view	 Detailed code and data inventory	 Execute migration plan	 <b>Parallel running and acceptance</b>	 Manage new system
 Create core team	 Augment technical migration skills	 Feasibility assessment	 Justification	 <b>Testing</b>	 Residual problem resolution	 Decommission old system
 Commitments and goals			 Migration plan			 Forward planning

Source: Gartner

# Tosca DI Data Migration Timeline

Migration Phases:



# Tricentis Planning Artifacts

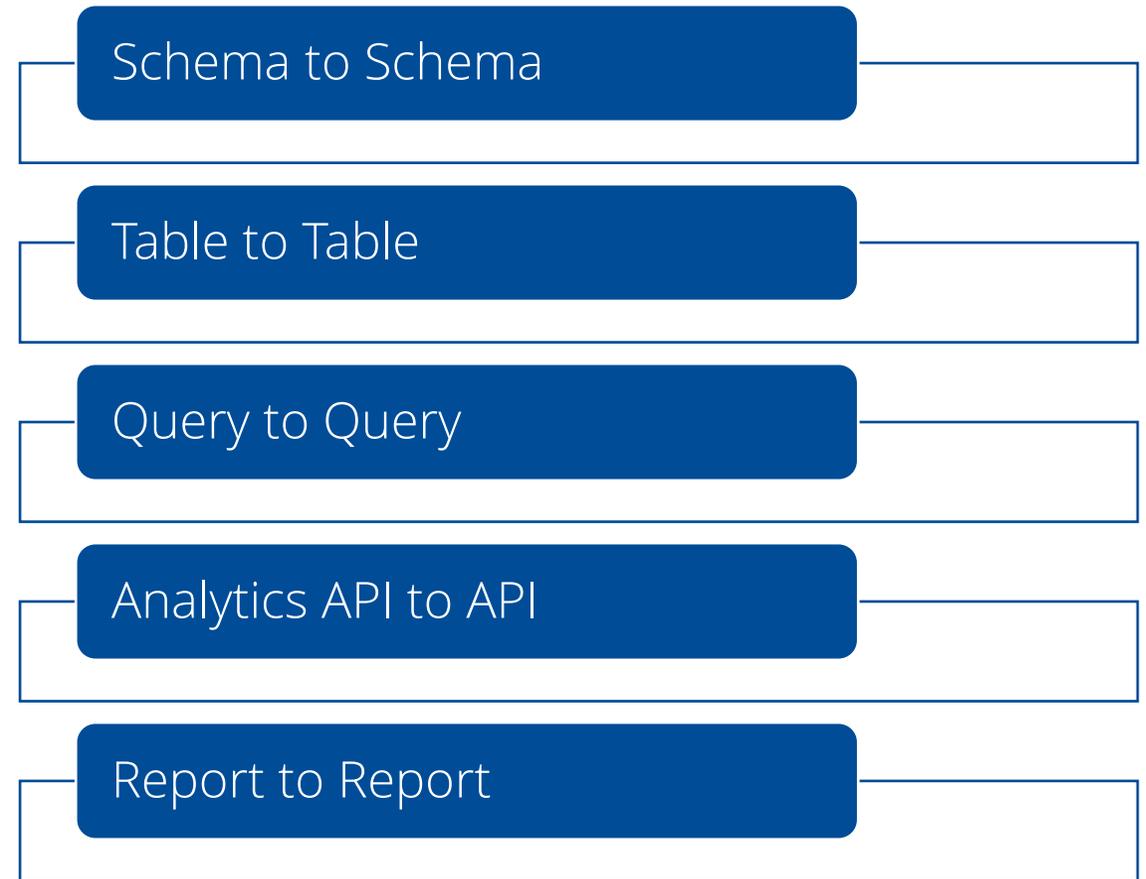
- Data Migration / Conversions
  - Assessment
    - Goals
    - Tasks
    - Measures
  - Strategy
    - Automation Architecture
    - Project Plans
    - Use Cases
      - Test Objectives / Tasks / Deliverables / Measures (cross references)
    - Personas

# Overlay the Plan With Mappings and Environments

## Dependency – Security Matrix

- Testing environments for source to target plans
  - Test objectives
  - Test deliverables
  - Test tasks - Data Quality Measures

## Source to Target Plans



## Verify the Data is in Sync With Dependencies & Changes

Verify data from both the migration process itself and ongoing external Integrations

- Complex dependencies make manual testing difficult and error-prone
- Accumulate an **automated** regression suite as you go
- Continuously test the migration and the ongoing systems

SOURCE / TARGET (Example for AWS-Redshift)	AWS - REDSHIFT	Client	Client
ORACLE	Yes	BofA	Thermo Fisher
EXADATA	Yes	WorldPay	Humana
TERADATA	Yes	Nationwide	IHG
NETEZZA	Yes	TJX	Nationwide
MSSQL	Yes	BCBS	CSS
AWS CDP - Hive	Yes	Kellogg	American Family
CLOUDERA - Hive	Yes	IBM	CPRail
AZURE SYNAPSE	Yes	HCSC	Cummins
SNOWFLAKE	Yes	TJX	Prologis
AWS-AURORA	Yes	GSK	CSS
GCP-BIG QUERY	Yes		
AWS-REDSHIFT	Yes	PHEAA	Kellogg

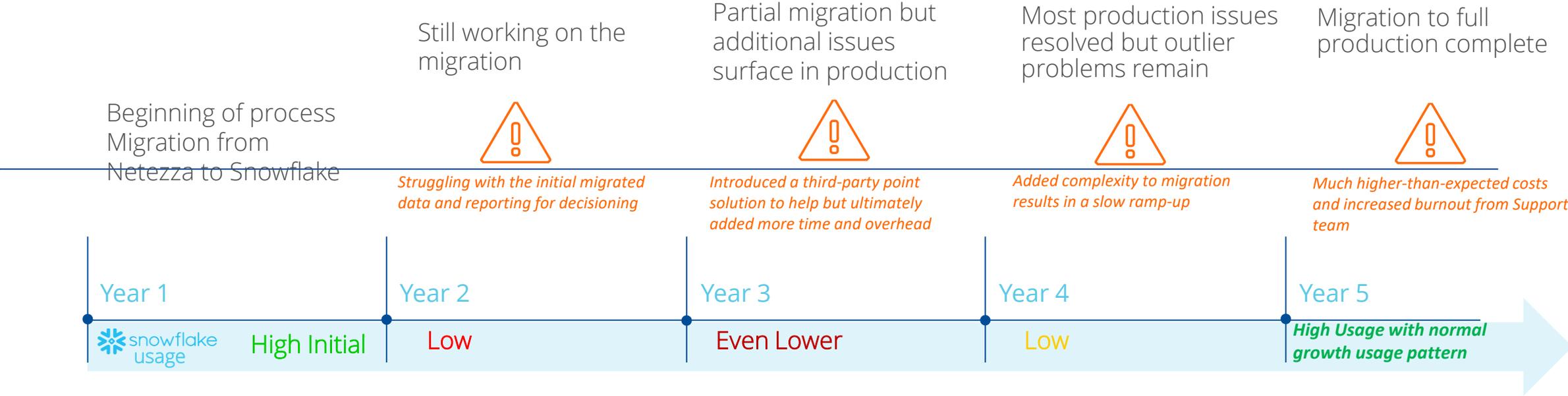
# Cloud Migration Use Cases Supported

## Migration Business Value Challenges Tricentis Addresses

- Legacy-to-Cloud Matrix (for history data comparisons)
- Incremental data testing (comparisons of inputs/outputs Source to Target)
- Cloud migration Risk management
- Regression In all phases of Migration – Key Post Migration Feature
- Test data management (Test Data Services)
  - PII data/tokenized data testing (e.g.: source not tokenized, target tokenized)
- Test data creation ( Not included)

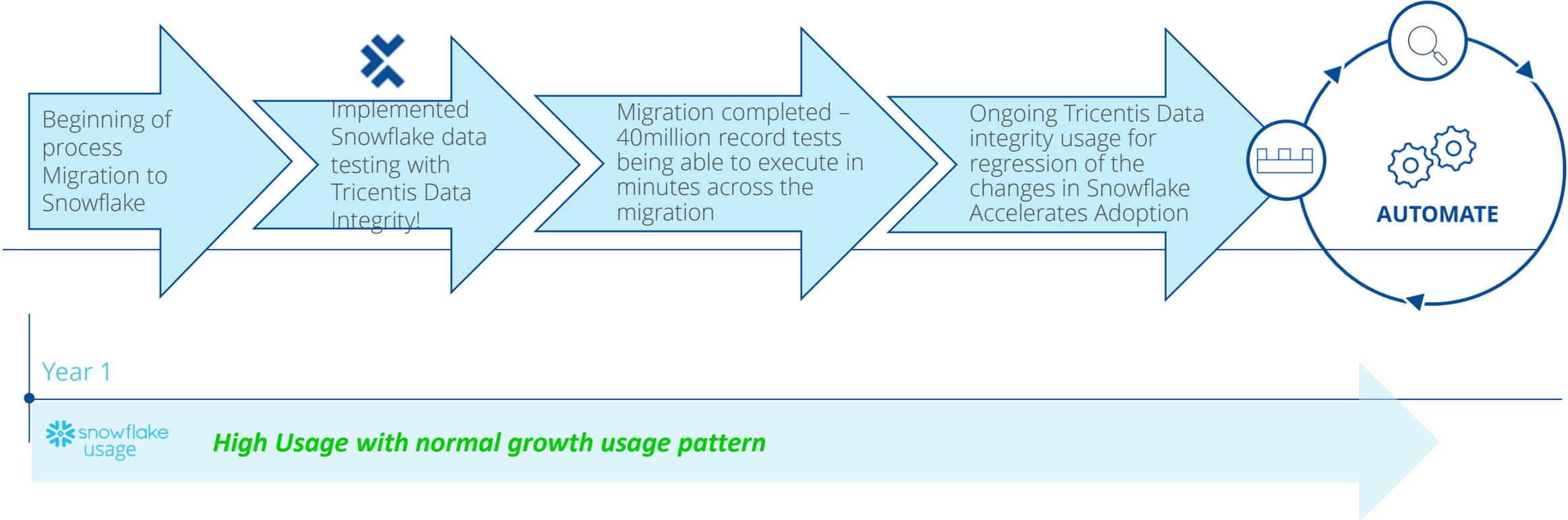
# Manual Testing Snowflake Migration Timeline

*Don't wait 3+ years for a complete migration to Snowflake*



# Typical Snowflake Migration Timeline with Data Integrity

Migrate to Snowflake in 180 days!



# BFIS Use Cases

# Banks Trust Us

- **Bank of America - Wealth Management Team**
- **Credit Suisse - Bank Platforms and Data Analytics**
- **TD Bank - CACS / CCP and Patriot Groups**
- **WorldPay - Payment Data and Analytics**
- **RBS - Enterprise Data Solutions**



# Top Business Value Use Cases

1. Each of the following Use Cases are examples of Critical Data Paths that require reconciliation and validation with 90% coverage.

Or these are the consequences:

- Time-to-market losses through failed innovation
- Lost opportunity \$ due to bad data in migrations
- M & A fails to achieve results
- Loss efficiency of Operations due to bad decisions on incorrect data

# Top Business Value Use Cases



M&A – Mergers and Acquisitions (and Divestitures)– Require a solid Data Reconciliation and Validation Strategy

- By offering a unique ability to validate against regression, data can now be moved in a consistent and trusted process ensuring data integrity during the migration and ongoing
- **ROI Impact: \$10's million in projected sales to properly migrated customers**
- Cause: Without properly migrated customer data marketing of acquiring bank's services would not be correct or compliant



AI/ML – AI and Machine Learning

- Compliance and Innovative Data Learning Model Success
- **ROI Impact: A \$100million Time-To-Market delay. With compliance pressure, manual testing of data was not able to perform to regulatory standards and without properly curated data to train model, innovation fails.**
- Cause: Datasets in Azure were too big to cover more than 1% with manual scripts and comparison (Manual Stare and Compare)



Compliance AML/KYC+

- **ROI Impact: \$50m in KYC fines stopped. Cost Avoidance that is predictable.**
- Cause: Timing, Bank could only test a sample (1K) of the 70K scenarios to be covered end to end in the teller (mainframe) to audit data reporting process



DATA Migrations during and AFTER in the New Migrated Cloud Environments

- On-prem to Snowflake
- **ROI Impact: Netezza to Snowflake migration EFFICIENCY saved \$1m using data reconciliation from DI**
- Cause: Production errors in the new Snowflake environment, as they added new Netezza data it would break the already migrated data and reporting without our regression testing



General Data Reconciliation and Validations for Accurate business decisions

- Data Analytics Data from Payroll, Payments, ERP, Logistics, etc...
- **ROI Impact: \$24 million lost in trading decisions for Oil off by 1,000<sup>th</sup> of a percent.**
- **ROI Impact: \$10's millions spent on data warehousing and analytics NOT BEING USED as NO TRUST in the numbers.**
- Note: Ability to perform this reconciliation in the Cloud in crucial (DataBricks/Azure 10-minute Demo Video)

# Banking Specific Business Value Use Cases

## Banking Settlement Reconciliation and Validation

- Tricentis Data Integrity can ensure the Float process is validated and reconciled before any changes are made to it. Ensuring a rigorous and exacting Float settlement process end to end and minute by minute.
- **Potential ROI Impact: \$10's million in properly executed Float settlement transactions versus loss of over float payments and COF if on the institution side.**
- Cause: To balance the strict compliance and settlement and treasury/cash management needs of customers and financial institutions this is a process that must perform perfectly, and with exact timeliness on execution.

## Data Driven Regulatory Needs are Changing with the New Banking Environments

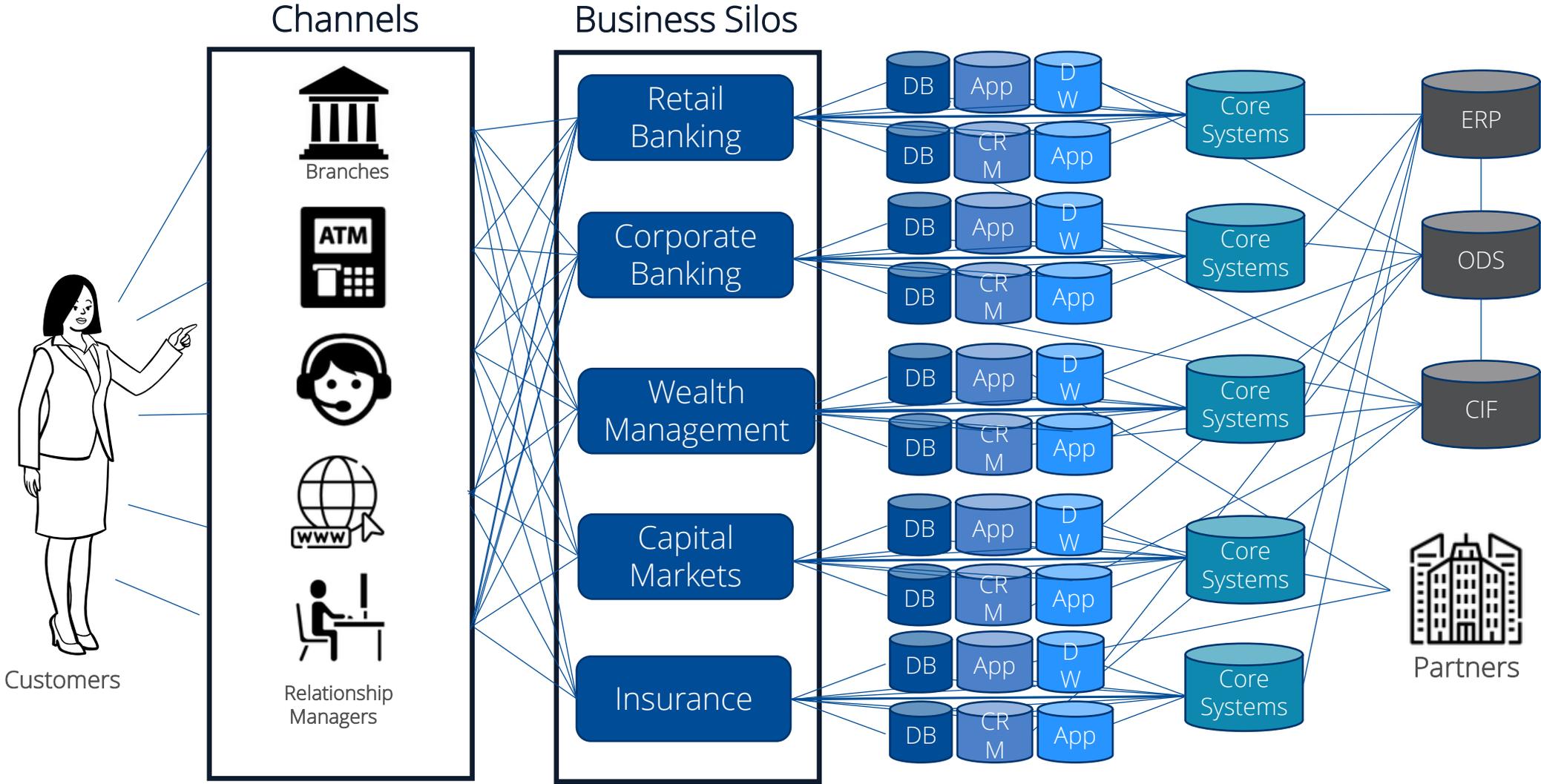
- End to End compliance (Risk and Regulatory) reporting needs to be covered for new requirements
- \$250K limits – Who is over and under?, etc...
- We are working to uncover these with our banks to ensure compliance with expanded

# Banking Conversion Business Impact

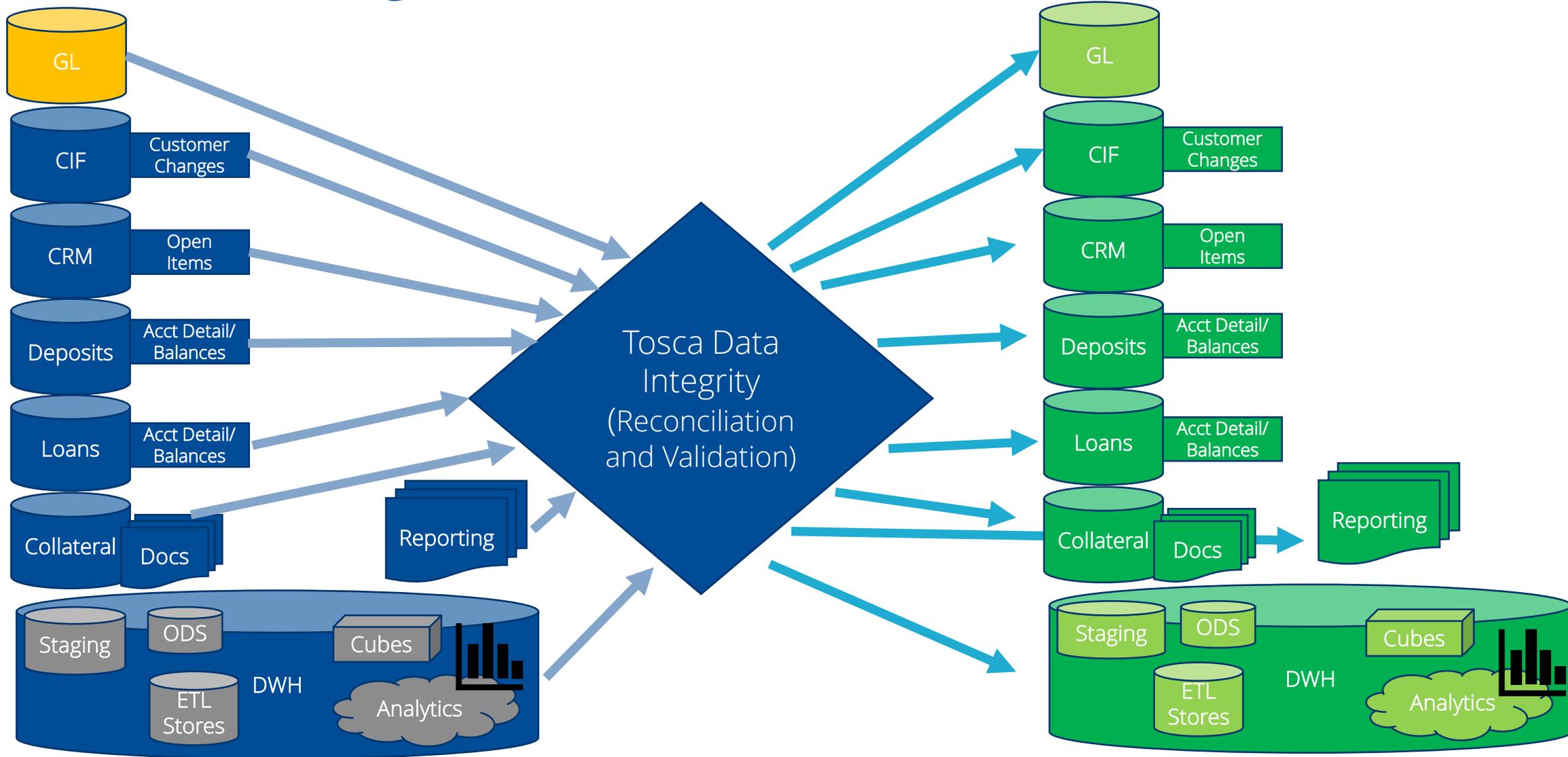
- With Tosca Data Integrity , Conversions, Migrations and Merging can be accomplished in less than ½ the time with 10X the data Risk coverage.
- Tosca Data Integrity delivers an AUDITABLE, COMPLIANT, and RIGOROUS process to the movement of data and associated data processes.
- Tosca Data Integrity brings a level of confidence in the data that can't be achieved without it. You can now Trust the data is correct.
- Tosca Data Integrity delivers the holy grail of testing: a full AUTOMATED regression suite for your ongoing (post conversion/migration) data and processes.

# Banking Systems Complexity at the Data Layer

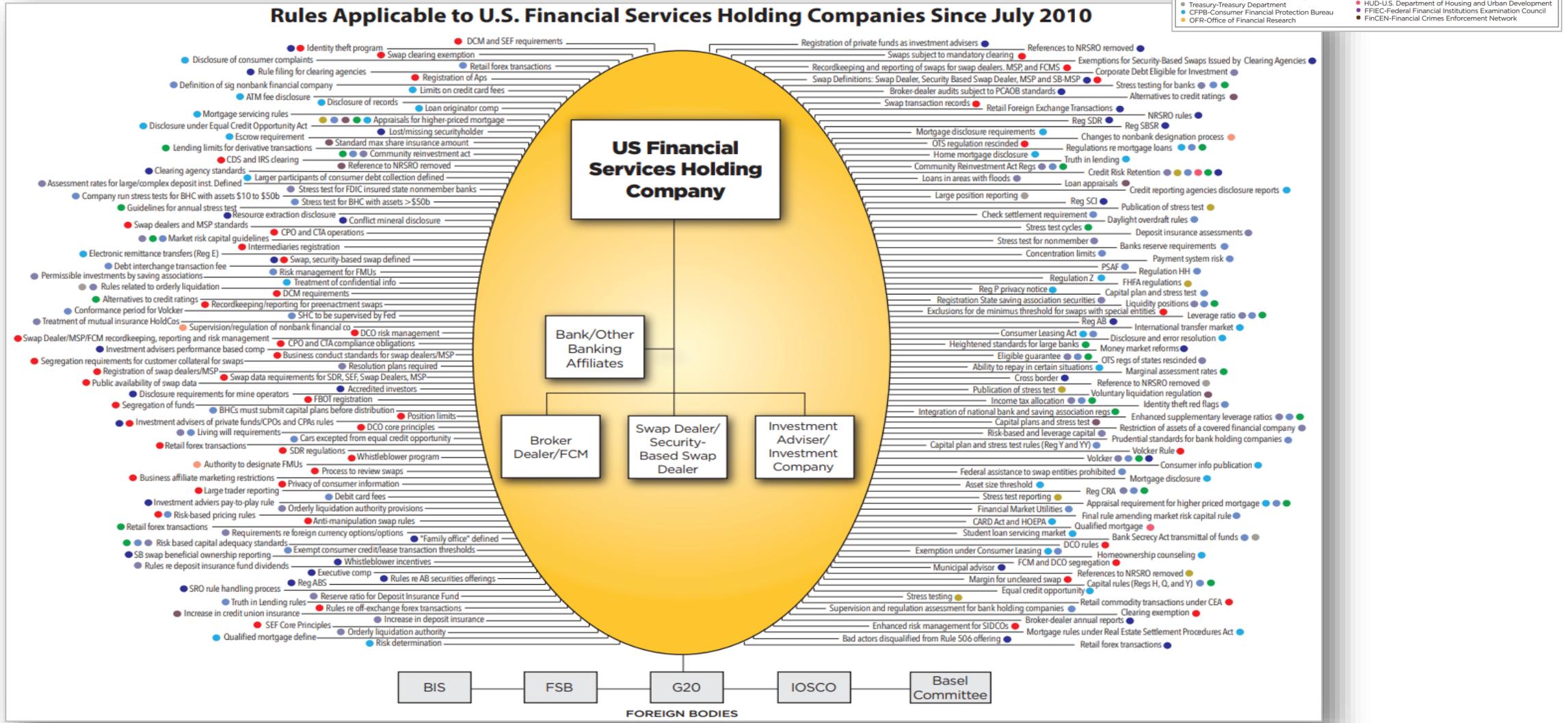
- Data Can break anywhere in the bank ecosystem



# Banking Conversion Data must be tested quickly to process within tight windows



# The proliferation of rules & regulations...



# TD Bank Streamlines E2E Regulatory Compliance

## BEFORE SITUATION

- Per money-laundering and anti-terrorism regulations such as the Bank Secrecy Act, banks conducting business in the United States are required to identify, monitor, and report “suspicious activity” in a timely manner
- Existing, predominantly manual, procedures could check a maximum of 1,000 scenarios per hour

## THE NEED

- Complex validations across different data sources and layers – previously could not be automated with vendor-specific “data quality” tools
- Increased use case coverage to accurately automate 70k suspicious transactions scenarios

## WHY TRICENTIS

- Improved scenario coverage to all identified use cases
- Automated validation to decrease manual overhead
- Increased identification of suspicious transactions across multi-faceted technology and business verticals

### Business Case Summary

### Using Tosca BI/DWH



Critical Checks Execution

70k validations in 15 min.



Potential Fines Avoided in 90 Days

\$2,600,000



Identified Suspicious Transactions

1600% Increase

*“We knew that automation was key for proactively ramping up our compliance ... we just didn’t believe it was possible to automate something so complex. Now, we’re much more confident that threats are being identified and reported almost as soon as the transactions occur.”*

*- Director of Compliance*

# TJX reduces Snowflake migration timeline & costs by 50%

And eliminates data testing headaches with Tricentis Data Integrity

## SITUATION

- Migration from Netezza to Snowflake required verifying migration stages, sources to target
- Scripted automation produced errors, requiring stop and restarts and delaying the 1-year project plan
- Data leaders required to implement a rigorous, repeatable regression process as the data was moved

## NEED

- More resilient, faster test automation for end-to-end data testing process
- Support for complex technology stacks — from 40-year-old mainframes to cloud
- Ability to handle massive amounts of data, including 5 years of sales data from many different sources, requiring many transformations from source to report

## RESULT

- Reusable, scalable tests without wrestling with SQL or scripting
- Thousands of hours of manual effort and test maintenance saved
- Improved user adoption and productivity as a result of improved data quality



# WorldPay – 90% Savings with BI/DWH Test Automation

## BEFORE SITUATION

- Millions of dollars invested in applications designed to help business leaders understand and leverage data
- Reporting errors lead to low to zero adoption
- QA leaders were challenged of “fixing this problem” by transforming the end-to-end data testing process without any existing tools, people, or processes for data testing in place

## THE NEED

- Test Automation for end-to-end data testing process
- Support for Complex technology stacks—from 40-year-old mainframes to cloud
- Handle massive amounts of data (from 40 billion transactions processed annually), many different sources of data, and many transformations between source and report

## WHY TRICENTIS

- Reusable, scalable tests without wrestling with SQL or scripting
- Thousands of hours of manual effort saved per month
- Improved user adoption and productivity as a result of improved data quality

**worldpay**  
from FIS

Business Case Summary		Using Tosca BI/DWH
	Time to Market Increase	90%+ Increase
	Cost of Testing	90% Reduction
	Records Compared in 20 min	200 Million

# Worldpay customer quote

//

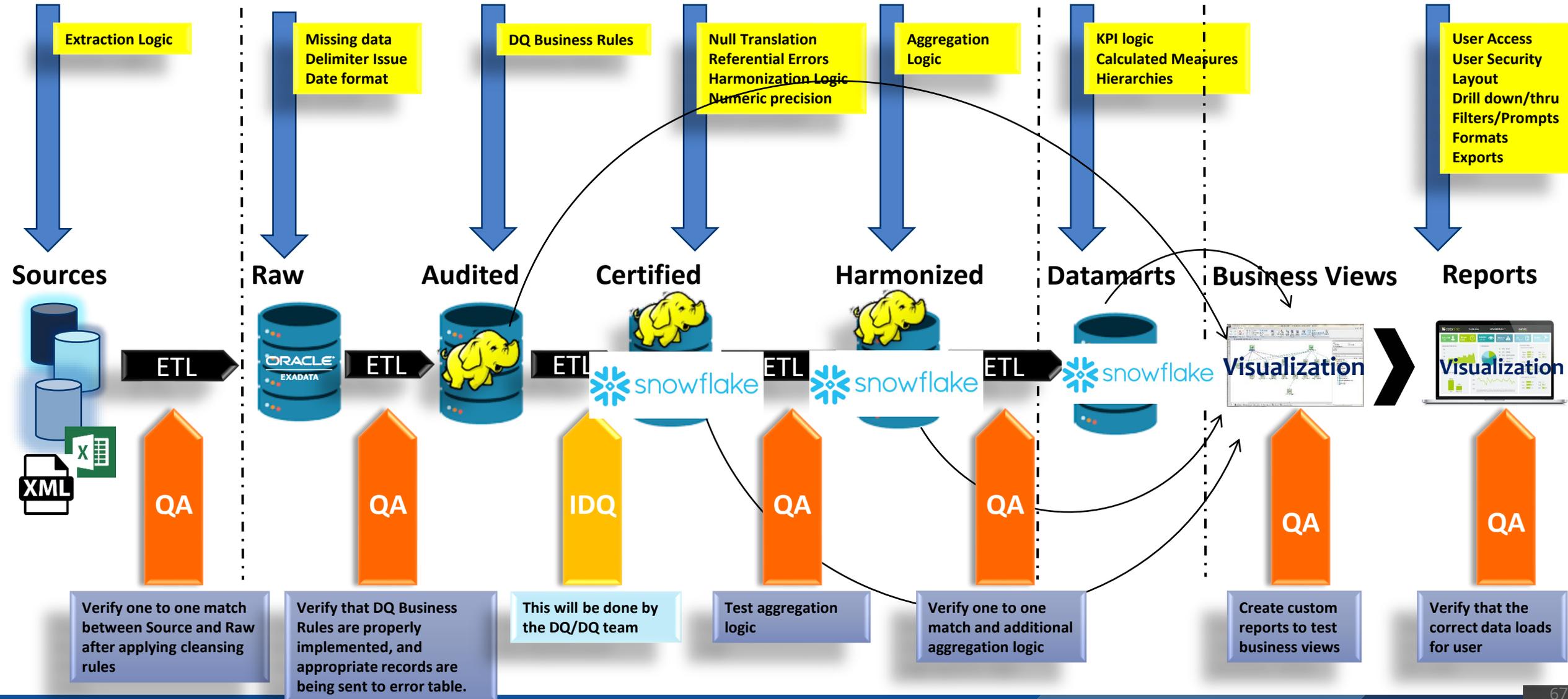
A few months after our new BI / data warehouse testing approach launched, people finally started **trusting** the applications.

They're highly used now, and they have become a vital tool in making critical business decisions.

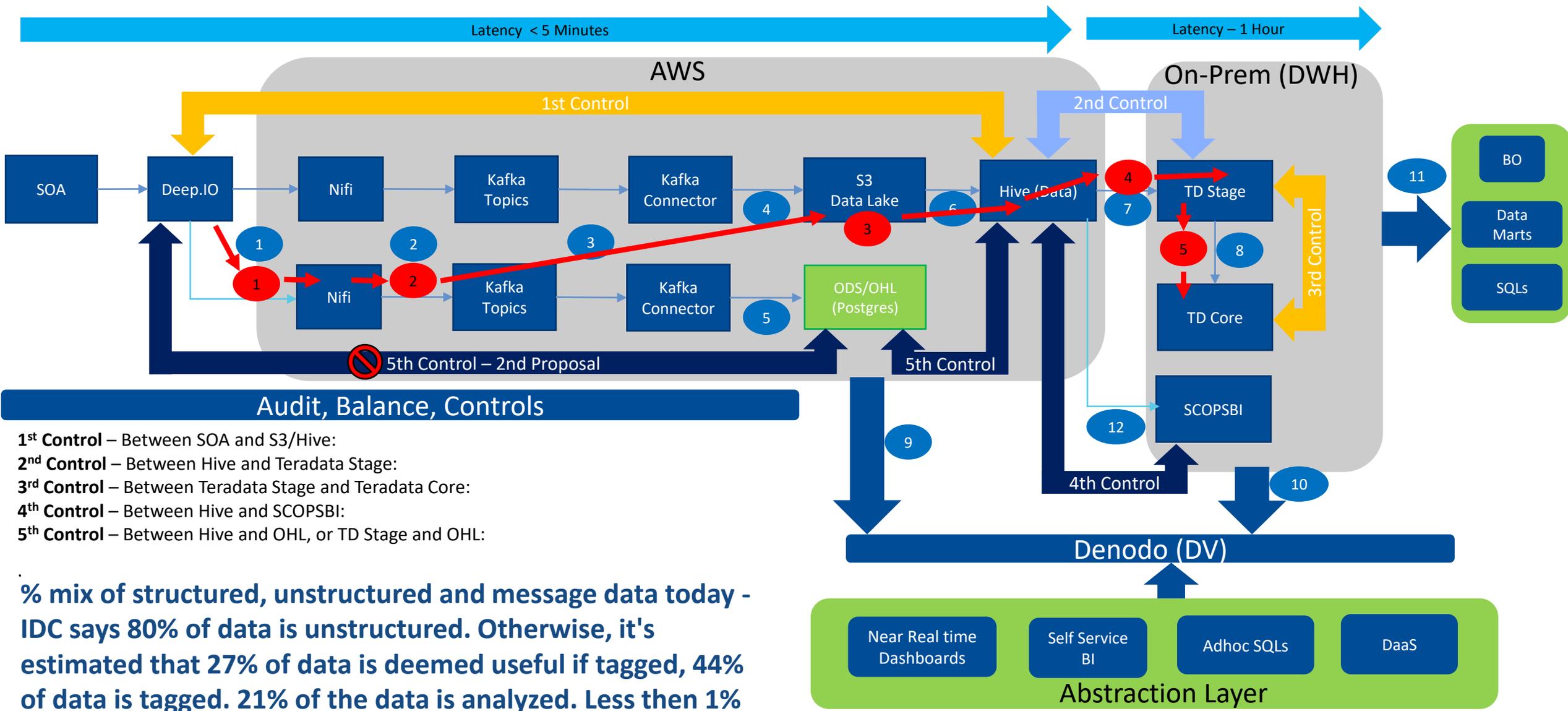


*- Head of Enterprise Quality Assurance Team, Worldpay*

# Data Testing Approach



# Streaming Operator > Cloud to On-Prem Use Case Arch



## Audit, Balance, Controls

- 1<sup>st</sup> Control – Between SOA and S3/Hive:
- 2<sup>nd</sup> Control – Between Hive and Teradata Stage:
- 3<sup>rd</sup> Control – Between Teradata Stage and Teradata Core:
- 4<sup>th</sup> Control – Between Hive and SCOPsBI:
- 5<sup>th</sup> Control – Between Hive and OHL, or TD Stage and OHL:

% mix of structured, unstructured and message data today - IDC says 80% of data is unstructured. Otherwise, it's estimated that 27% of data is deemed useful if tagged, 44% of data is tagged. 21% of the data is analyzed. Less than 1%