



Microsoft Future Decoded

How Adatis helped Rank
Gaming productionise &
automate the management
of machine learning models
in Azure

Terry McCann – Adatis

Chris Conroy – Rank Group



Rank Gaming Group

Rank was established in 1937 when they were known for making films.

Rank now owns two of the biggest brands in gambling in the UK; Mecca Bingo and Grosvenor Casinos.



Club | Online | Mobile





Adatis Consulting

Adatis Consulting Limited is a Microsoft Gold Partner specialising in Data Engineering and Data Science.

Established 12 Years ago we are now a team of ~70. We have 3 Data Platform MVPs.



We love helping
people make sense
of their data

Data
Strategy

Data
Management

Data
Architecture

Data
Analytics

Data
Science

Managed
Services

Rank Data Science

Rank has a team of 7 Data Scientists.

They hold advanced degrees in a range of disciplines.

Wide variety of skills (Python/Scala/R etc.)

Rank Data Science had big ambitions to use Azure to the fullest, to create a truly polyglot architecture decoupled from their existing data processes.

Limited engineering capability on the team.



The Problem?



Rank Data Science Problems

Existing data processes were tightly coupled with legacy processes, all new data had to go via a Data Warehouse.

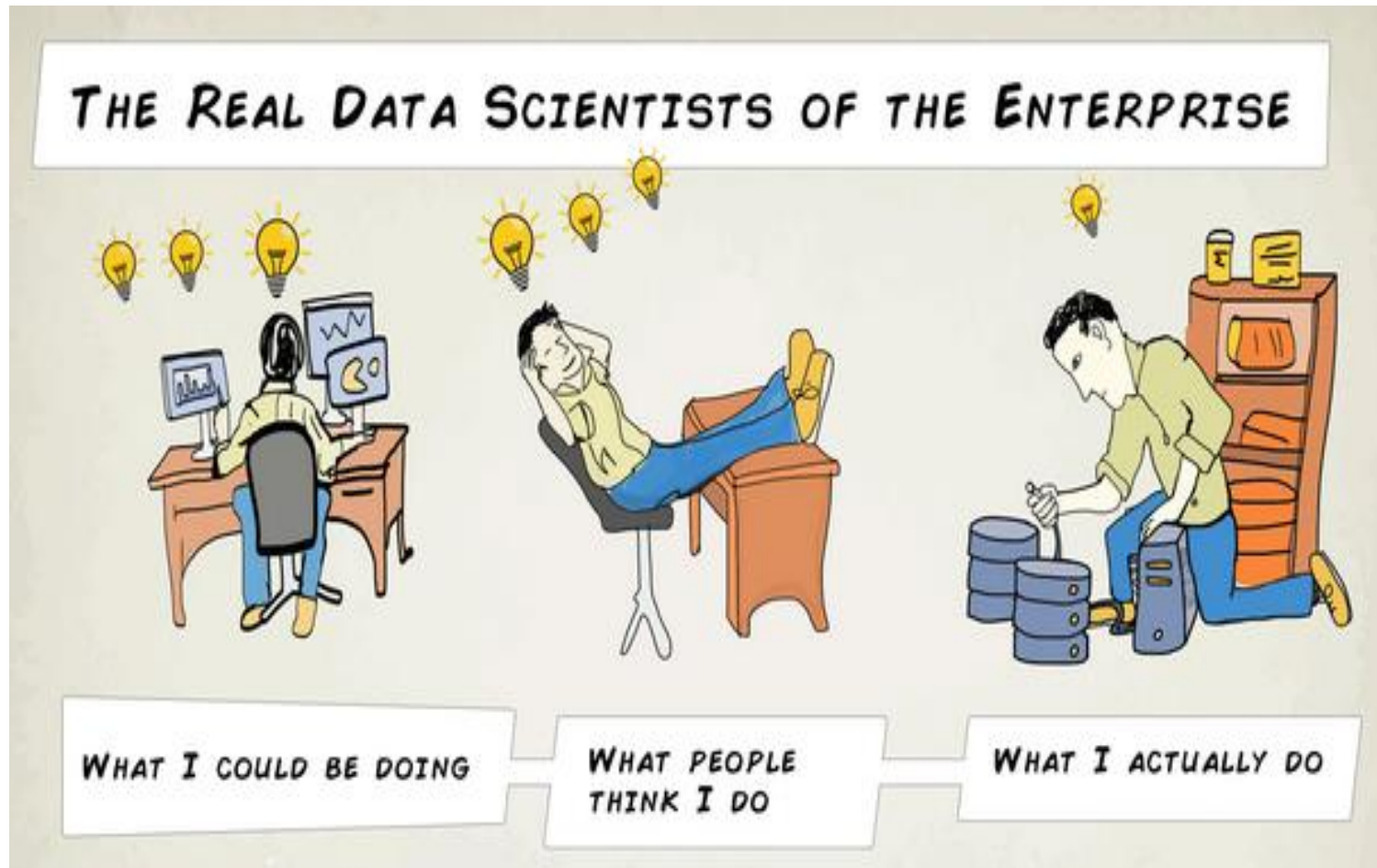
Access to new data could take anywhere up to 7 months to obtain.

Machine Learning models were in 'manual production', however not enough. Rank's roadmap was long and ambitious.

Very limited engineering resource to call on.

When models were in production, Rank faced a problem with challenger/incumbent model replacement.





1. Access to Data was too slow
2. Machine Learning Model Management



The Solution?



Leveraging a Data lake for access to data



A traditional Rank Analytical data process

1. Requirements are gathered
2. Requirements are mapped to the data
3. Data is acquired
4. Data is profiled
5. Data is loaded in to a stage process
6. Data is cleaned and transformed
7. Data is loaded in to a data warehouse
8. Data is loaded in to a semantic layer
9. Data is reported to the business



Where does the data scientist come in
to this?

80% of Data Science is Cleaning and Preparing Data

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data
scientist

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data
scientist

DATA STORE

1. A central store of data
(Good/bad/ugly)
2. Decoupled from existing processes
3. Automate the scoring of *batch*
Machine Learning Models
4. Should apply DevOps (Source
control/CI/CD)
5. A fully-automated cleaning process

DATA STORE

DATA STORE



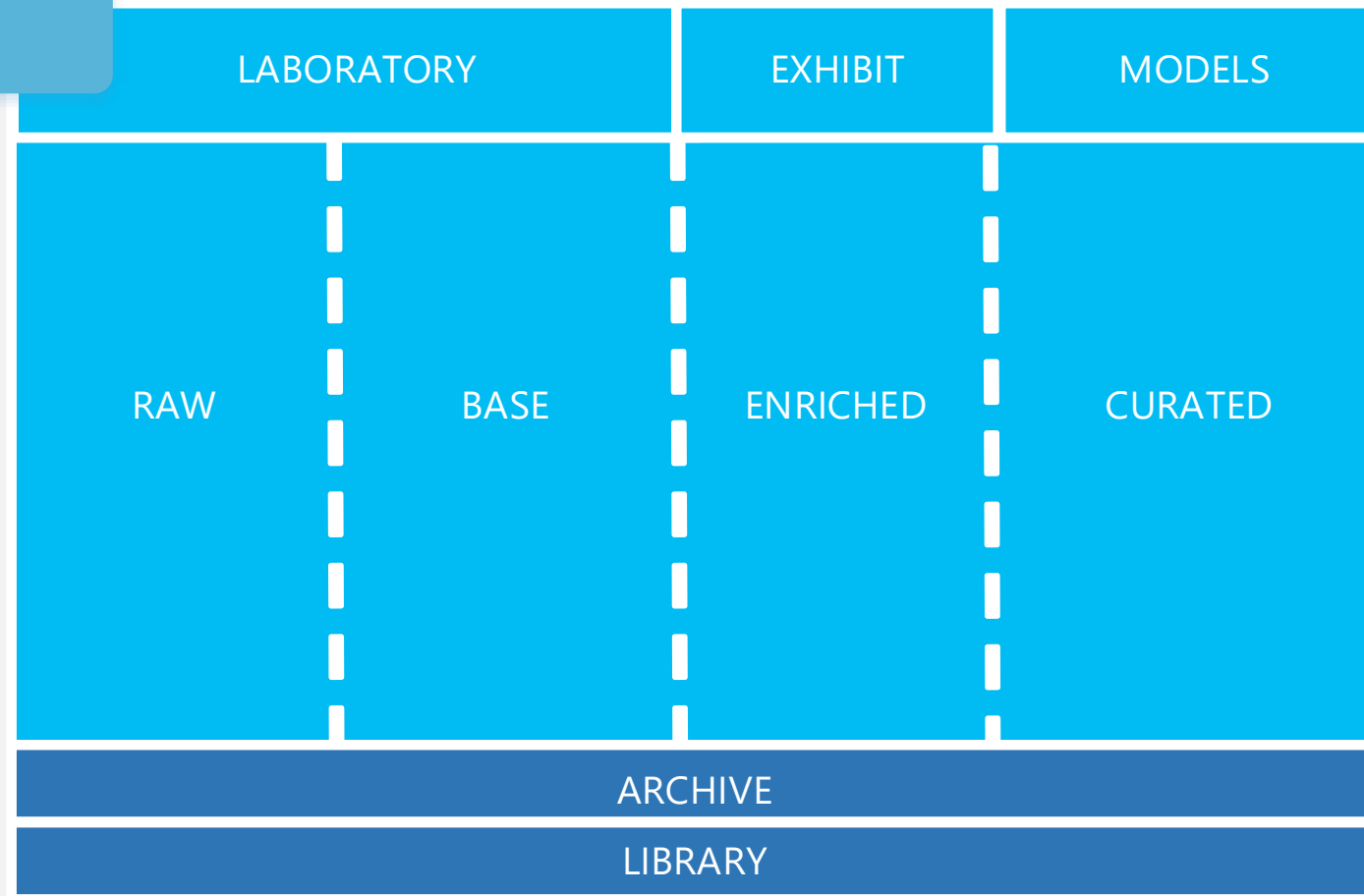
ADLS

DATA LAKE

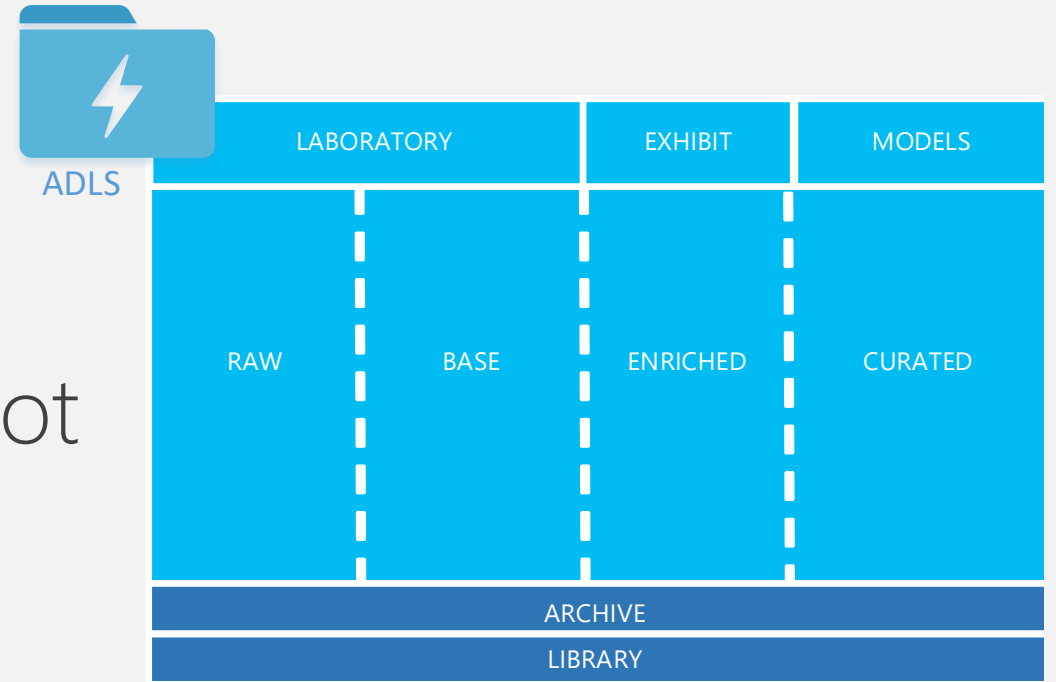
Adatis Data Lake Framework

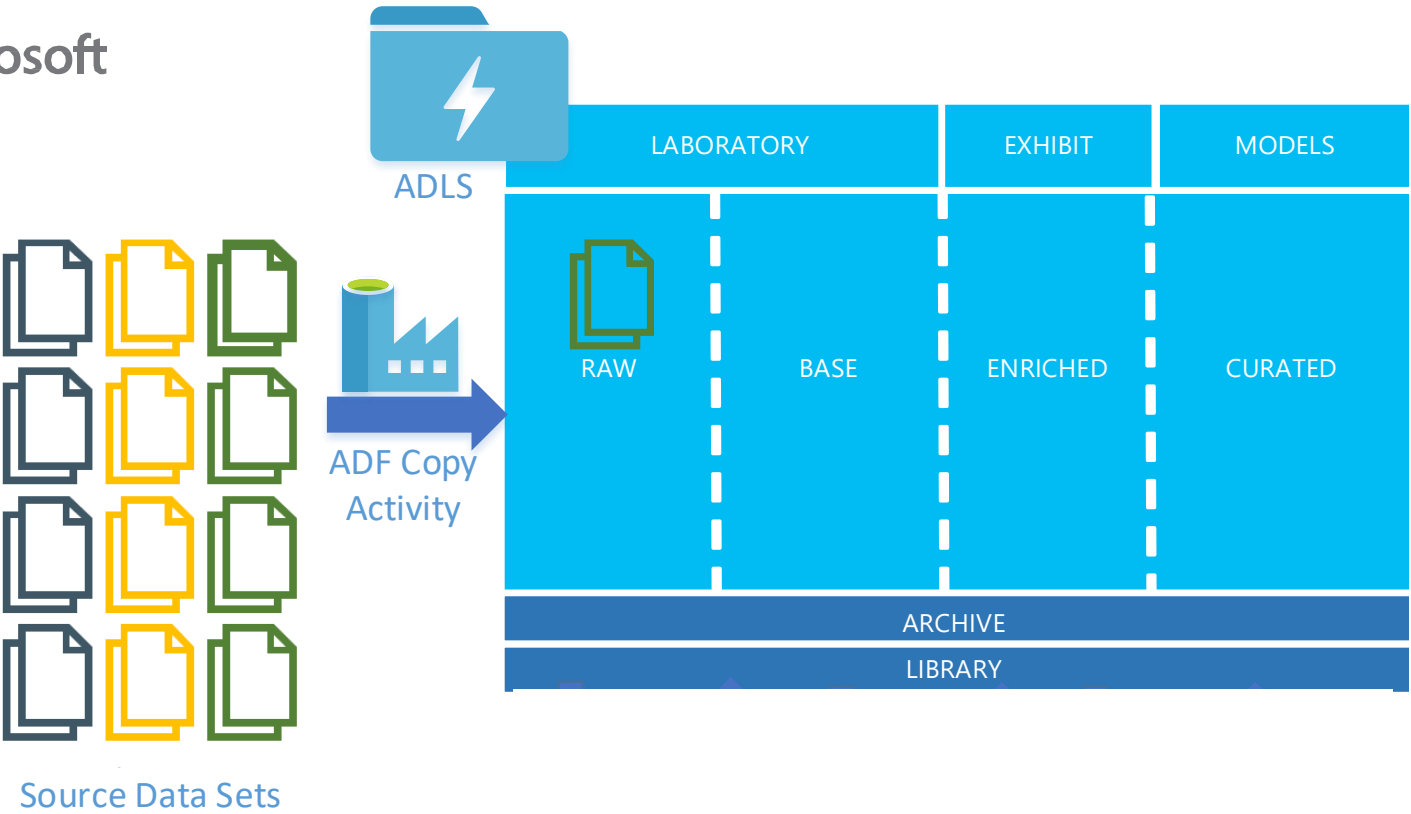


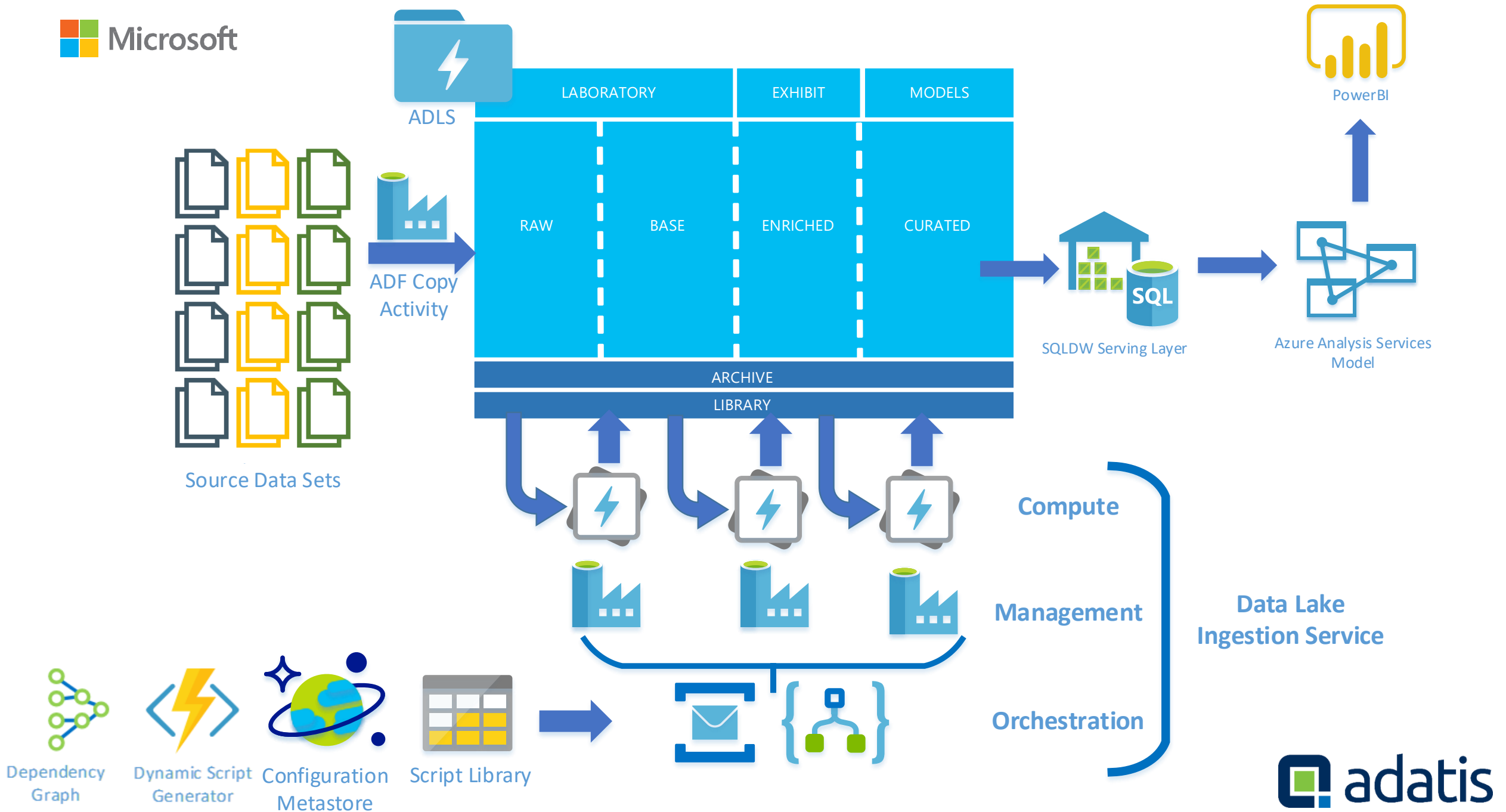
ADLS



1. Ensure that the file is correct
2. Ensure that the schema has not drifted
3. Apply a series of quality rules
4. Enrich the data with new columns
5. Combine multiple files in to one.



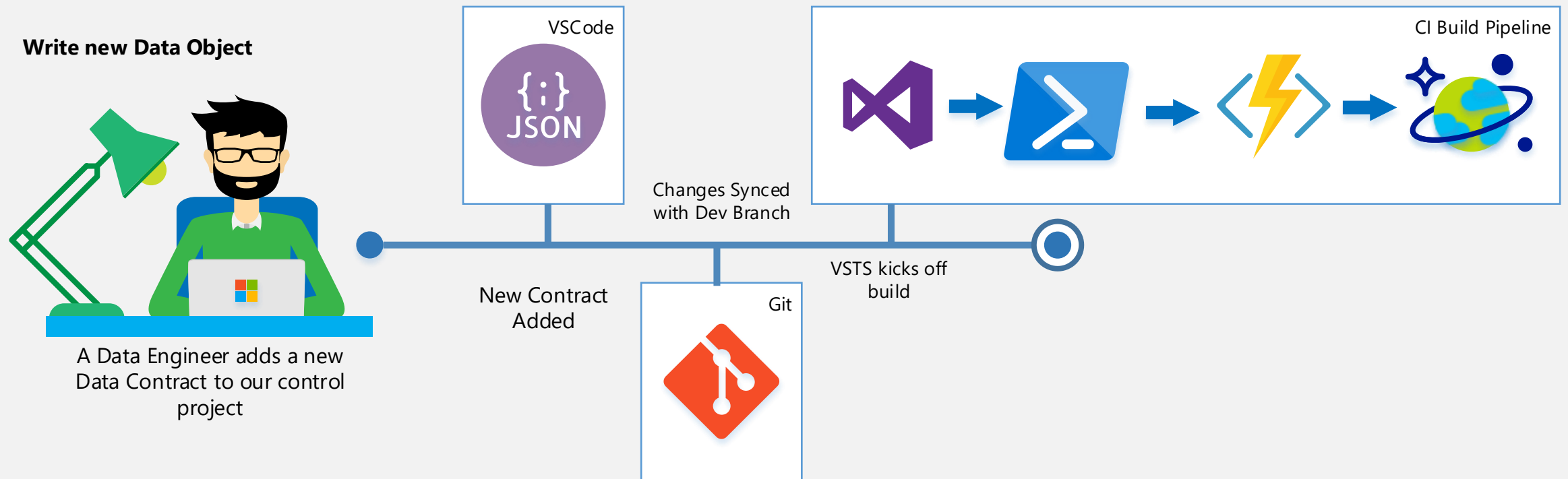


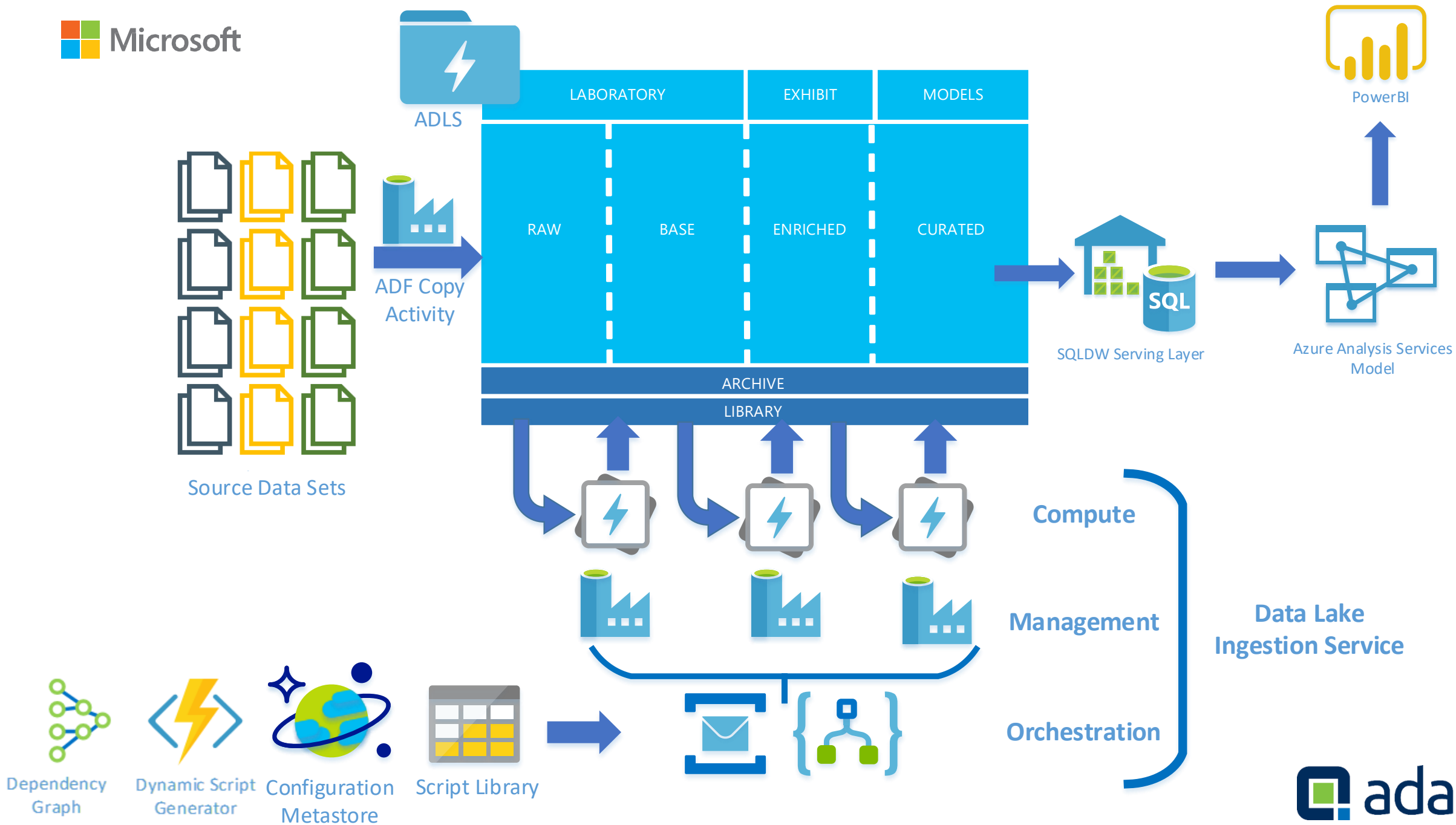


Benefits

1. Time to onboard data has been reduced from months to minutes
2. Data is in a consistent format and level of quality for both Data Science and Traditional analytics
3. The lake serves as an immutable store of all data in the business.
4. Batch model scoring is fully automated
5. DevOps automates the deployment of new data processes

Solution Build Pipelines





Machine Learning Model Management (Interactive)



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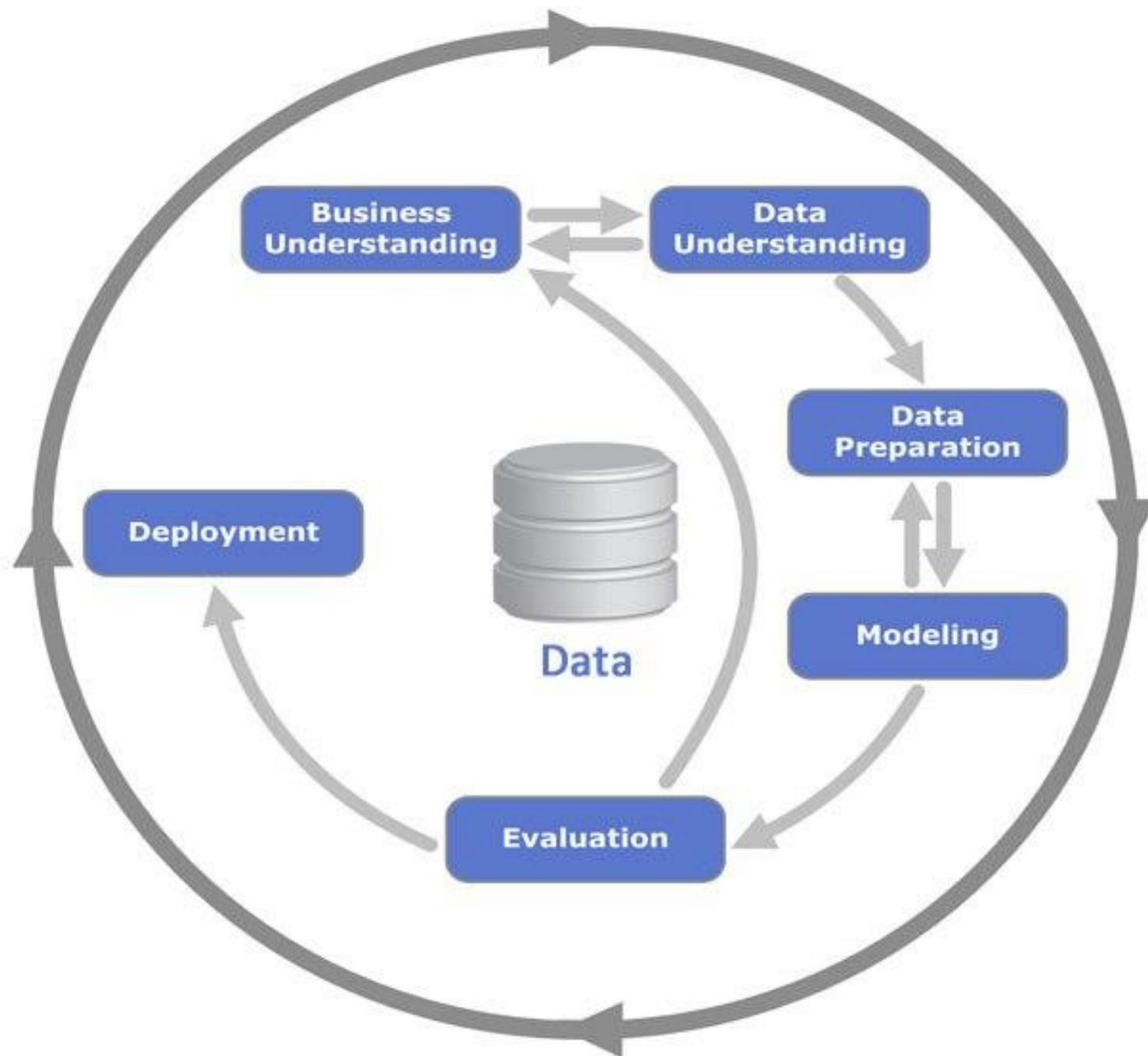


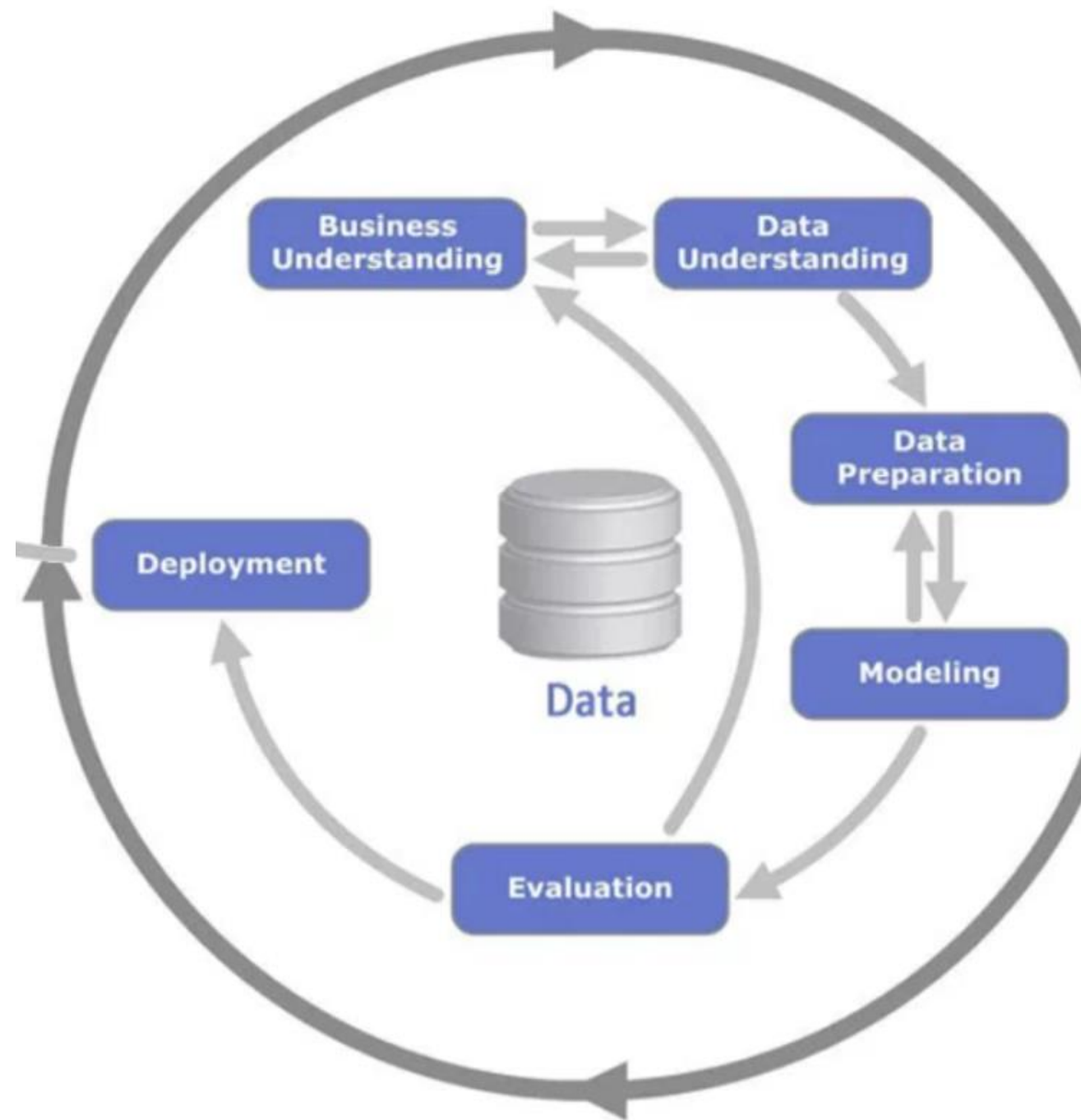
The Machine Learning Process

1. Gathering data
2. Preparing that data
3. Select an algorithm
4. Train a model (data + algorithm)
5. Evaluate the model
6. Hyperparameter tuning
7. Test the prediction (new data + model)
8. Deploy the model ("Productionised")

“Productionisation of models is the ***TOUGHEST*** problem in data science”

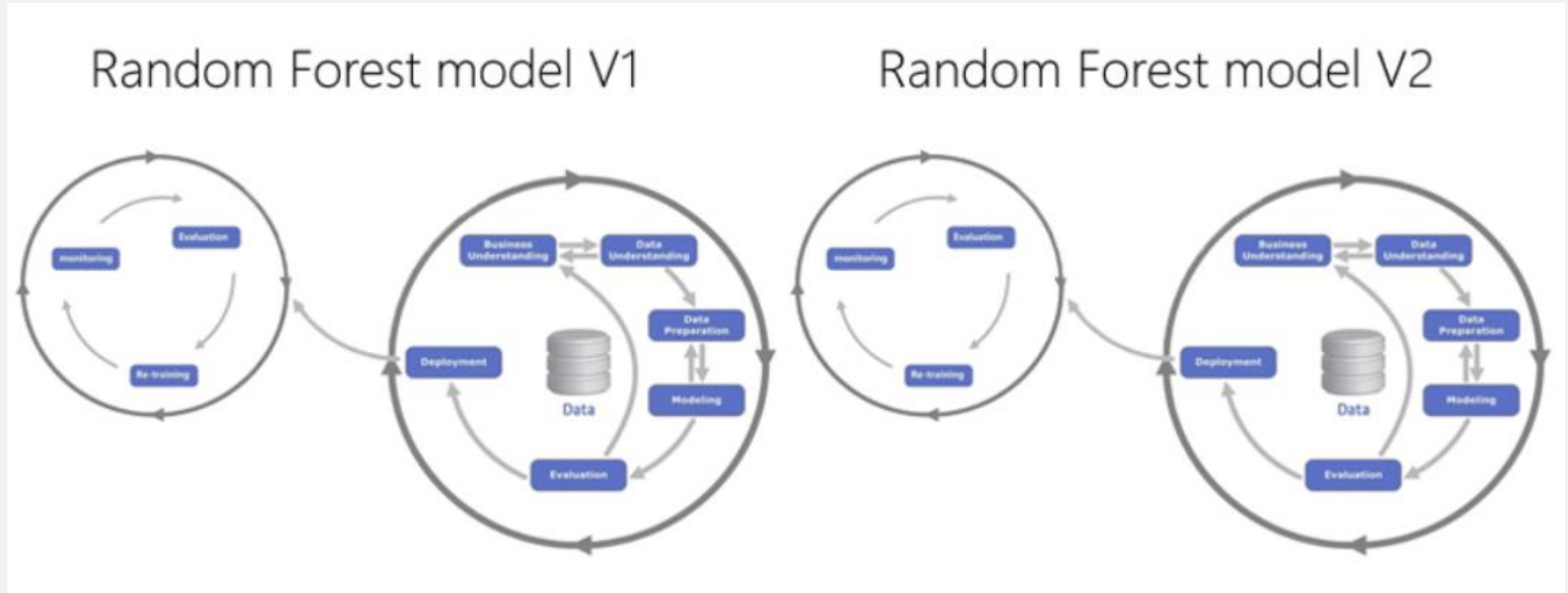
CRISP-DM Process Diagram



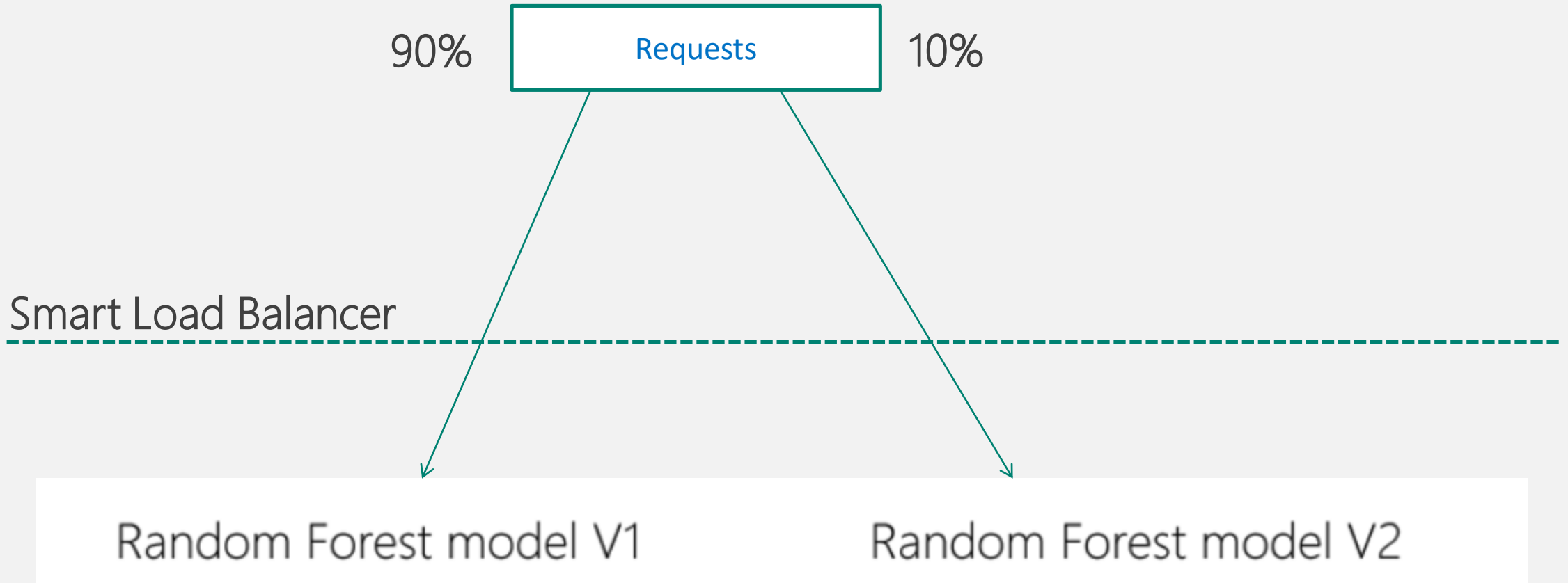


Machine Learning: Challenger model Vs Incumbent model

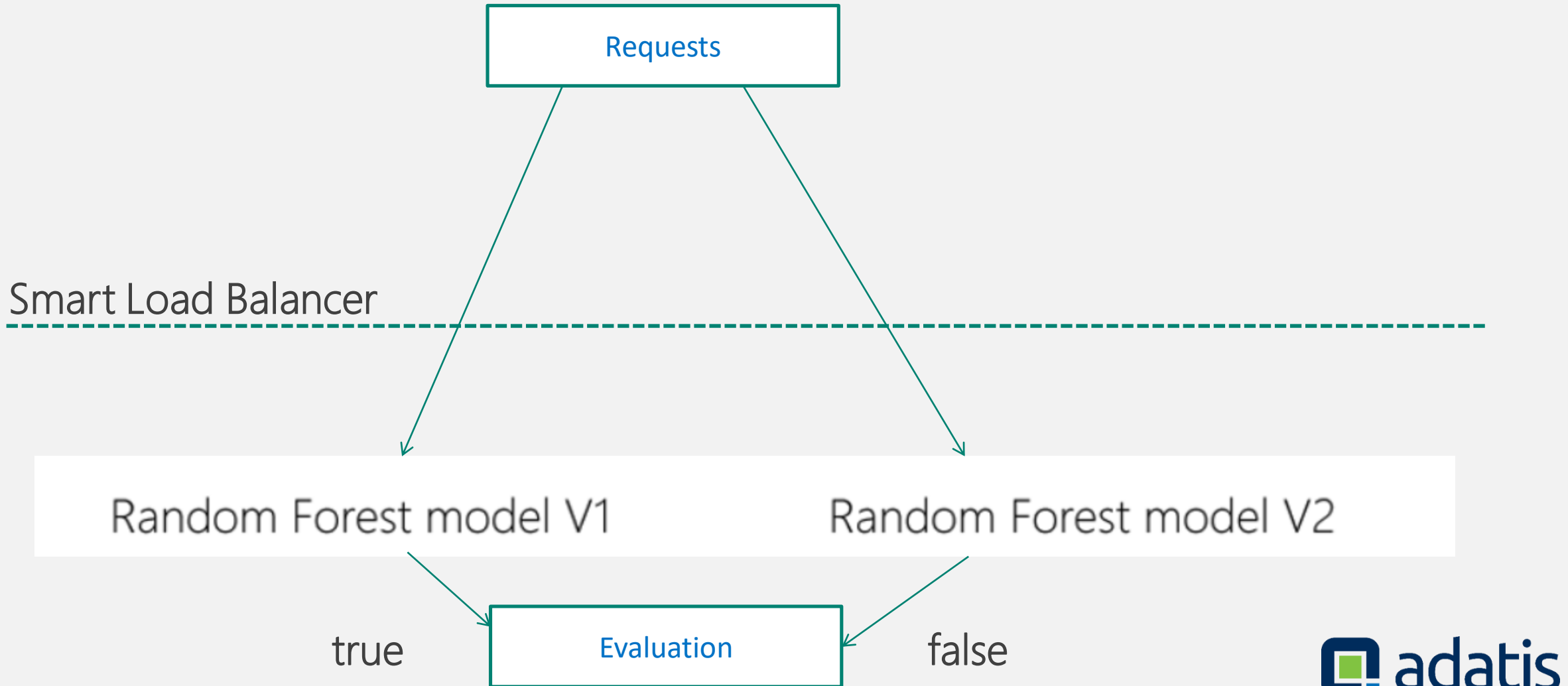
AB Testing scenario



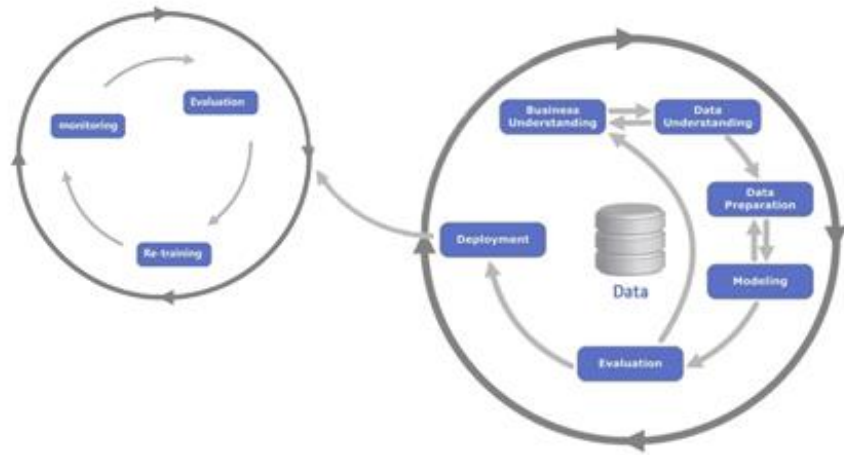
AB Testing scenario



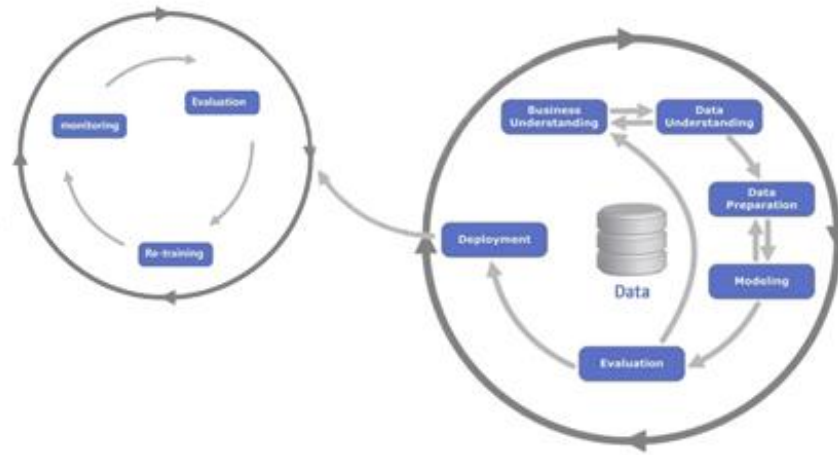
AB Testing scenario



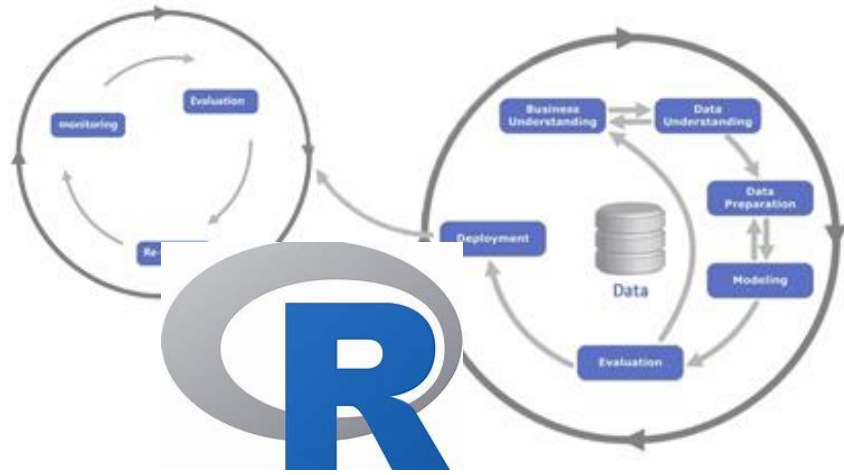
Random Forest model V1



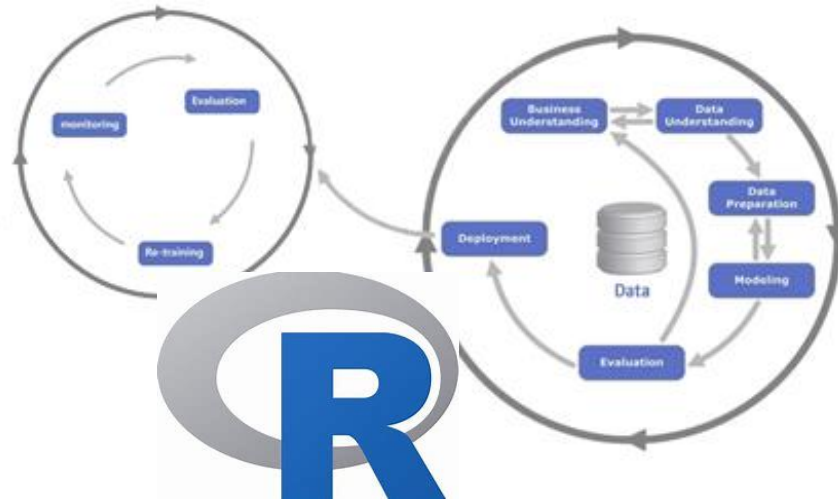
Random Forest model V2



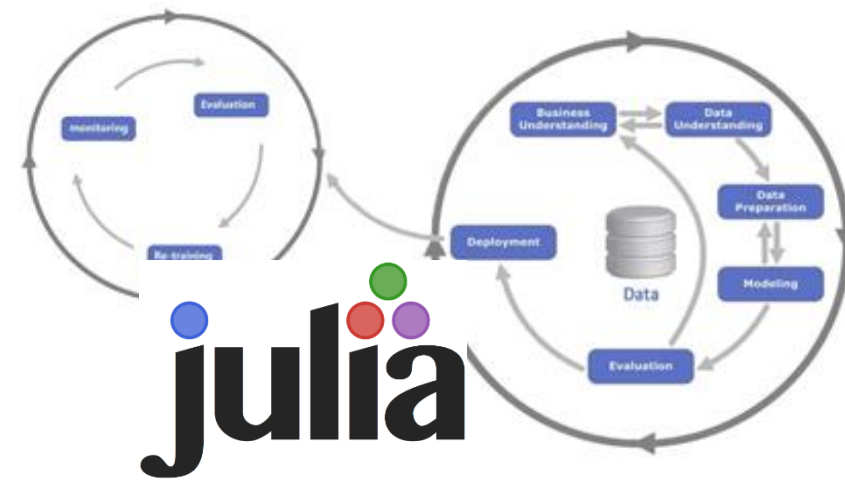
Random Forest model V1



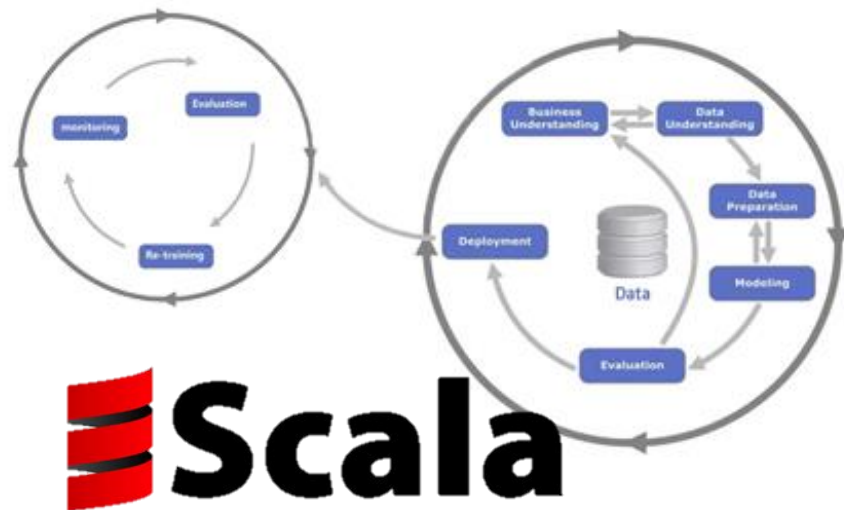
Random Forest model V2



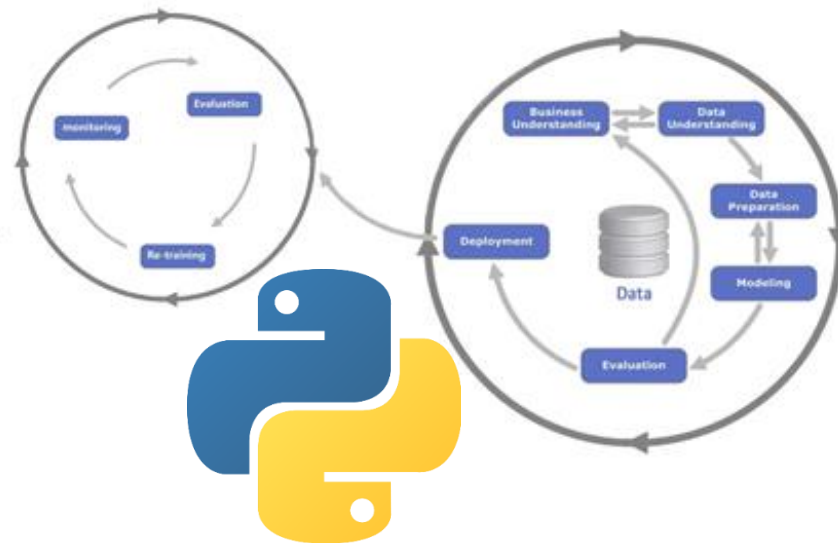
K-Nearest Neighbours



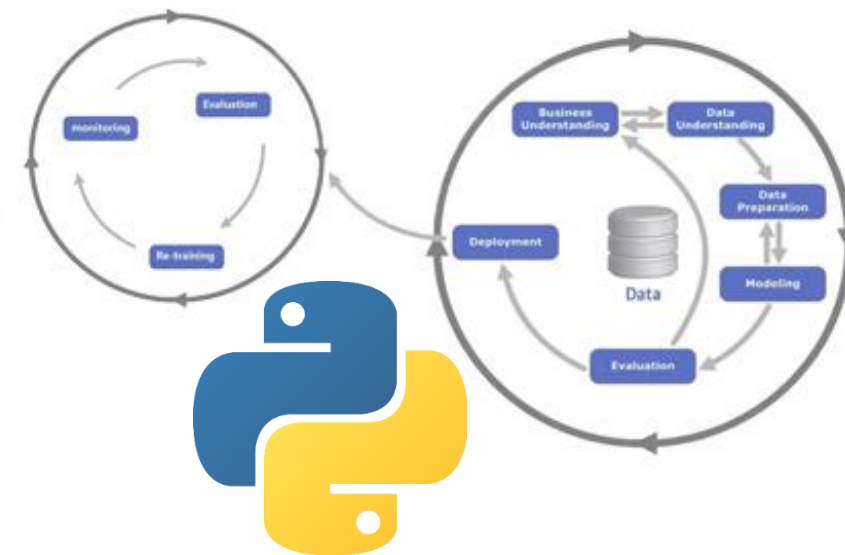
Ada Boost



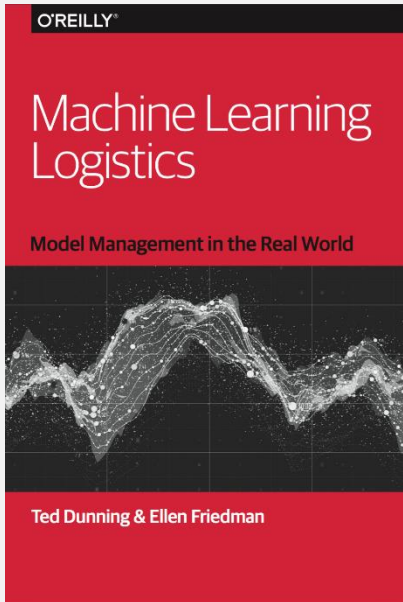
Naive Bayes

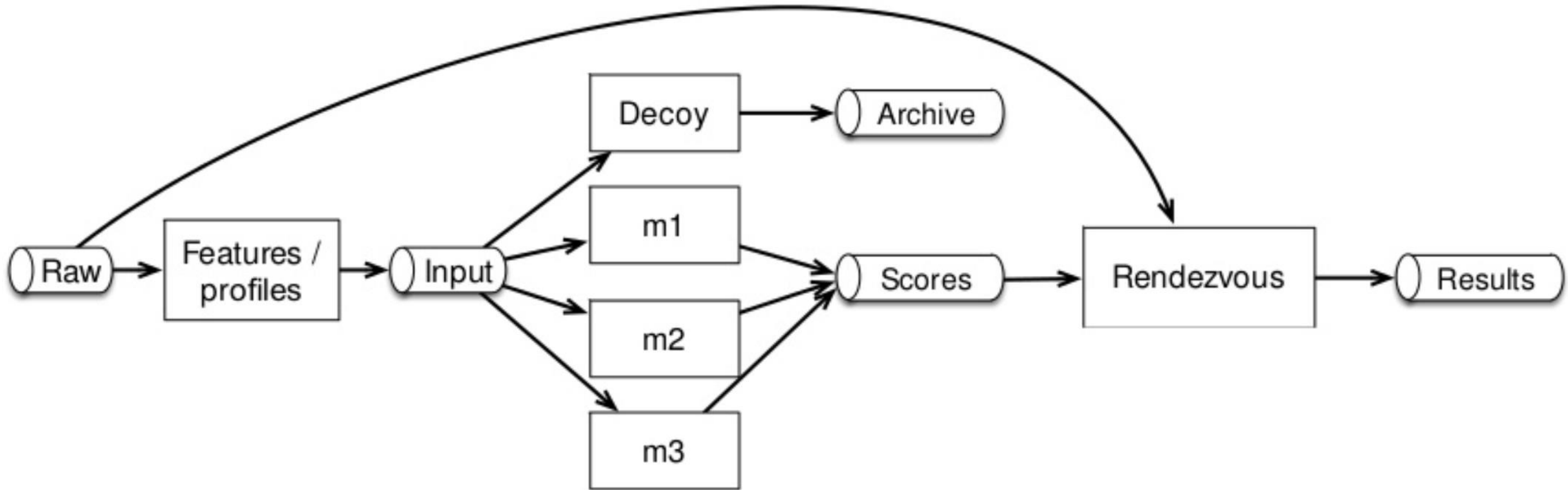


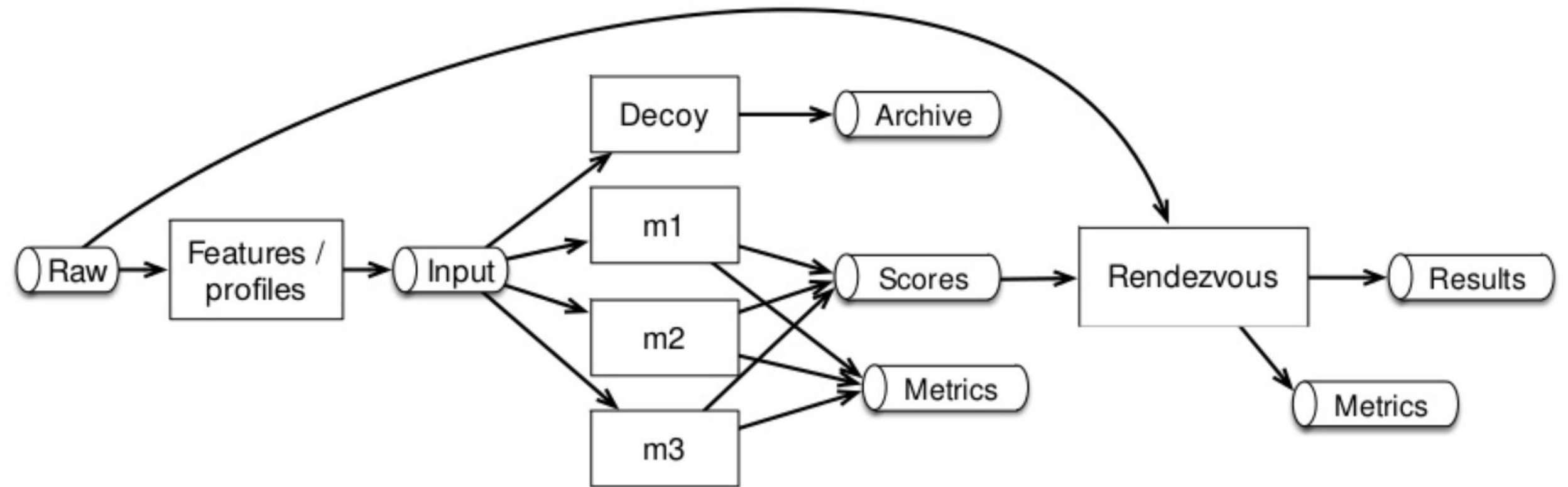
Decision Tree



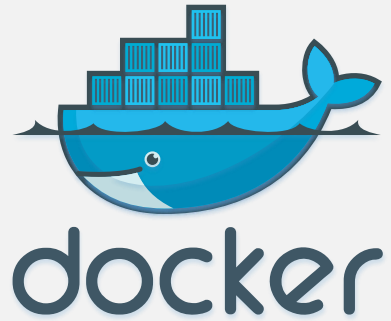
Rendezvous Architecture







Key Components?



Azure Container
Registry



Java Build
(to trigger a ML model)

Machine Learning Model
Build (Any language)

Java Image
(to trigger a ML model)

V1

Machine Learning Model
(Any language)

V6

Java Image
(to trigger a ML model)

Machine Learning Model
(Any language)



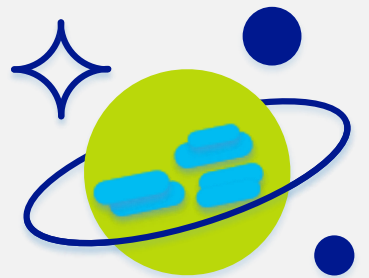
Azure Functions

REST API



Service Bus

Topics and Queues
to trigger models



CosmosDB

Operational
Monitoring

Rendezvous



Incoming
request



Do our Data Scientist's really need to
learn all of that?

Introducing Azure DevOps



Azure
Boards

Plan, track, and discuss work across teams, deliver value to your users faster.



Azure
Repos

Unlimited cloud-hosted private Git repos. Collaborative pull requests, advanced file management, and more.



Azure
Pipelines

CI/CD that works with any language, platform, and cloud. Connect to GitHub or any Git provider and deploy continuously to any cloud.



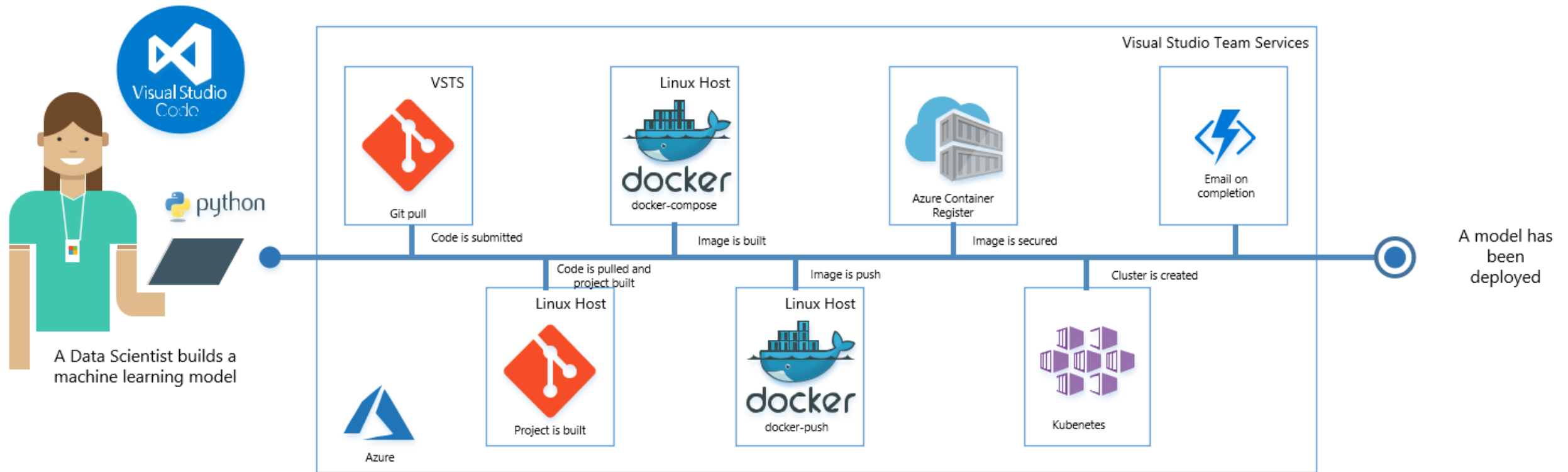
Azure
Test Plans

The test management and exploratory testing toolkit that lets you ship with confidence.



Azure
Artifacts

Create, host, and share packages. Easily add artifacts to CI/CD pipelines.



Benefits

1. DevOps removes the Engineering requirement
2. Able to support any language in production
3. Supports both batch and interactive.
4. Automated Model Management
5. Operational statistics all fully logged
6. Elastic scale
7. Future proof solution

Looking to the future



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Things to do next

Session Feedback

Please rate this session in the Future Decoded app!

Microsoft UK AI Research Report

Download the AI Report at <http://aka.ms/UKAIreport>

Azure Marketplace Consulting Services

Find Microsoft partners to help at <http://aka.ms/UKMarketplaceService>

Enhance your Digital Skills

Find great guidance and resources at <http://aka.ms/fdskills>



Questions



Thank You

