

# DATA-, ANALYTICS- AND AI PLATFORM

The foundation for building your data-driven company

### WHY IS A DATA AND AI PLATFORM STRATEGY NEEDED?



#### Pain points and challenges

Lack of data integration and availability

Ad-hoc modeling without product idea

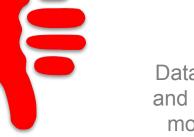
Limited analysis options

High susceptibility to errors / poor decision-making

Unstructured and unversioned data



Inefficient development and innovation



Data protection and compliance more difficult

Lack of compliance and ethical standards

Lack of standardization and modularity

Lack of scaling effects

### THE CORE ELEMENTS



#### WHY?

- **→** Competitive advantage
- → Increased efficiency
- → Innovation
- → Data utilization
- → Customer loyalty
- → Data protection, ethics and compliance

#### HOW?

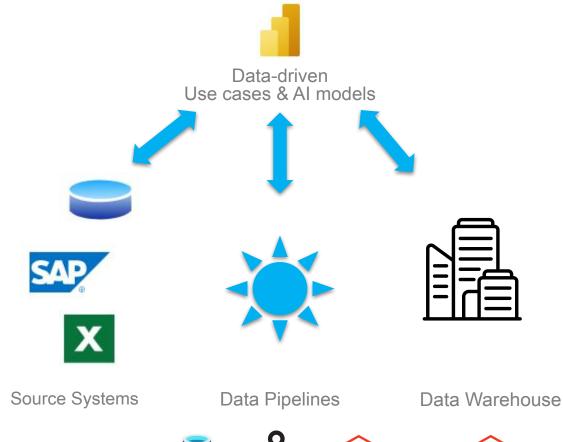
- → Optimized database & strategy
- → Data, analysis and Al platform
- → Productization of analyses and Al models
- → Operationalization / Data Products

### OPTIMIZED DATABASE & **PLATFORM**



#### **Initial questions**

- → Identifying the teams who actually needs the data?
- → What problems do we want to solve with the data platform / How do we work with the data?
- → What are the technical requirements? (Scalability, storage, networking, etc.)
- → Is there a standardized data format that we can agree on?













## OPTIMIZED DATABASE - PRACTICAL EXAMPLE



#### The Challenge

- → Diverse data landscape of enterprise companies should be made available for data platforms
- → Data is stored in different formats, structures and is managed by different teams; a uniform data strategy is to be found
- Strong coupling of individual systems should be softened

#### **Our Solution**

- → Databricks as a data platform on Azure
- → Microsoft Azure Data Factory as a high-level pipeline tool (data is pulled)
- → Confluent Cloud as a managed streaming engine to provide data in a structured way and in real time (data is pushed)

#### **Use Cases**

- Standardization of thousands of source systems
- → Reporting systems (including anomaly detection in real time)
- → Asynchronous data exchange between different platforms e.g. Salesforce and various internal data solutions

#### Our achievements

- → In a team of 6 people, we standardized over 5,000 data sources within six months and made them available in a DWH
- → Self-service architecture suitable for compliance

## DATA-, ANALYTICS- AND AI PLATFORM



#### **Keyfacts**

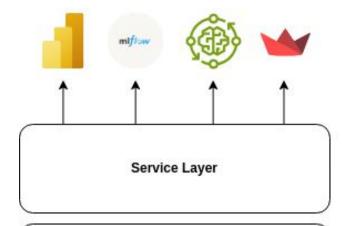
- → Cloud first Databricks is designed & made for the cloud
- → Data compliance and governance solution
- → Fully managed comparatively little infrastructure effort for Databricks, the focus is on the features

#### De Facto standard / Best-of-Breed

- → Cloud Microsoft Azure
- → Language diversity Python, Java, Scala, R
- → Notebooks as a familiar interface for data scientists
- → Cooperation between teams through Git integration (notebooks are not JSONs in Databricks!)
- One language that everyone speaks everything is a data frame!

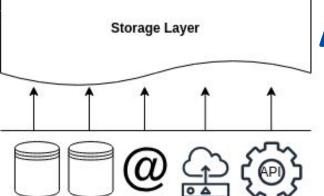
#### Versatile tooling landscape and options

- → Out-of-the-box services such as MLflow, Feature Store, Model Serving
- → BI connection
- → Workflows (triggerable processes)











## STANDARDIZED PLATFORM - PRACTICAL EXAMPLE



#### The Challenge

- → Retail Company: Migration from SAS to Microsoft Azure
- → Building a new analytics environment
- → New models & lifecycle

#### **Our Solution**

- Introduction of Databricks as a central analysis and Al platform
- → Data integration via Microsoft Azure Data Factory Pipelines and Medallion Architecture
- → Live SQL dashboards for data integration and quality
- → MLflow for the development, registration and deployment of machine learning models
- → Introduction of Feature Store for KPI calculation
- → Security and governance principles with Databricks Unity Catalogues
- → CI/CD pipelines and version control with Github (Actions)

#### Al & Analytics Use Cases

- → Customer segmentation
- → Coupon recommendations
- → Customer contact route
- → Assortment optimization
- Customer needs analysis
- → Migration analysis

#### Our achievements

- → Creation of analytics product catalog
- Increasing the speed of development and innovation
- → Lifecycle management for ML models
- → Traceability of historical analysis results
- → Ensuring compliance guidelines and documentation















## PRODUCTIZATION OF ANALYSES AND AI MODELS



### Development of a "toolbox" for AI models and analyses

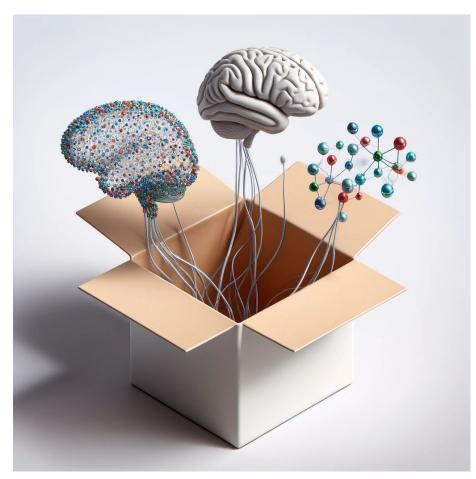
- → Generalize models and analyses
- → Standard analyses to answer recurring questions
- → Enable application to different data
- → Modular structure of data engineering, feature extraction, analyses and models
- → Interchangeable "building blocks

#### Standardized exports and presentation of results

- → Development of systematized visualizations, exports and dashboards
- → Standardized connection to BI tools

#### **Defining business value**

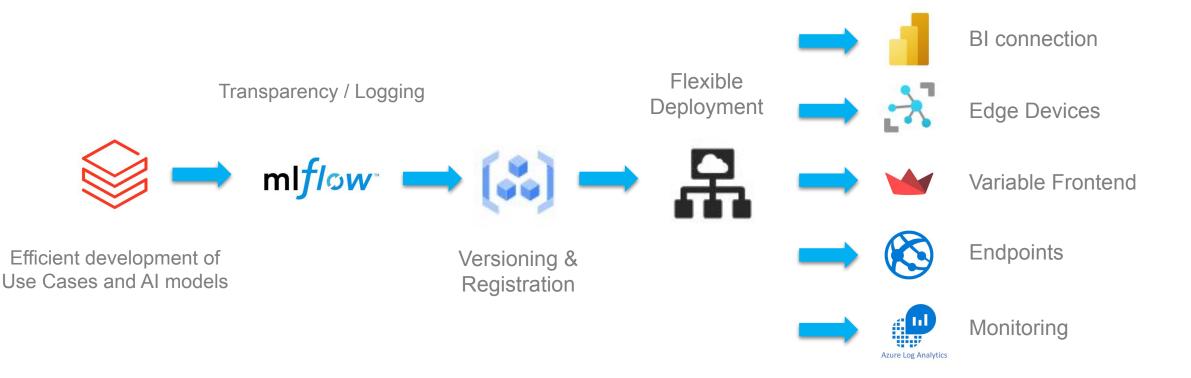
→ Optimize models for business value and avoid falling into the R&D trap



## OPERATIONALIZATION OF DATA PRODUCTS



The foundation for your operational data products







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### THANK YOU

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