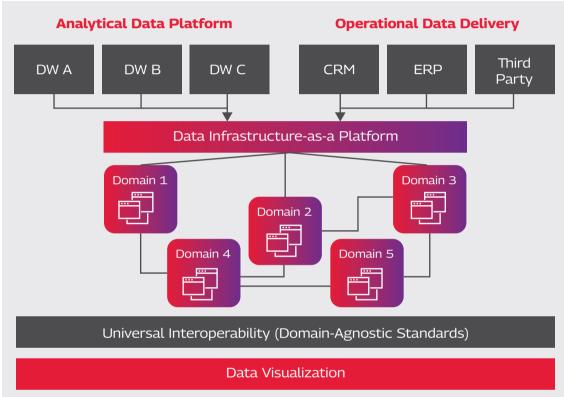






Abstract

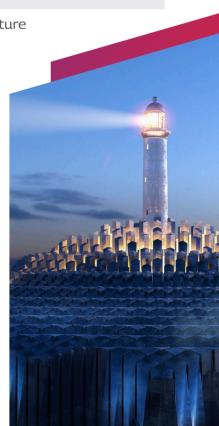
Data mesh essentially refers to the concept of breaking down data lakes and siloes into smaller data domain-specific sets with a self-serve design, to enable data-driven decisions using data products and providing scalable delivery of data with a flexible data governance model. Data mesh is a decentralized approach to sharing, accessing, manage analytical data in a scalable way.



A high-level data mesh logical architecture

Data is collected from analytical and operational platforms and collected in a data lake. Our point of view on data mesh encompasses the following principles:

- Rather than going through a centralized technical team the business domain should own the data end-to-end
- A self-service platform is fundamental to a successful data mesh architectural approach
- Data as a product
- Data products must be built with governance and compliance that is automated and federated





Challenges and Best Practices



Data Mesh Challenges

- Effective insights from huge volumes of telemetry and internet of things (IoT) data to improve customer experience and retention
- Managing regional compliance and security standards for multinational organizations specifically in BFSI, retail, and telecommunication sector
- Enabling advanced analytics for each stakeholder group as per individual use cases
- Availability of timely, consistent, and accurate data for robust and reliable forecasting analytics



Data Mesh Best Practice

- Use case-based data products –
 create consumption-oriented use
 case-based data products to optimize
 data usage and reduce data duplicity
 with an eye on performance and
 scalability
- Hybrid and agile data governance –
 Federated governance across data
 products with a touch of centralized
 data governance for key areas such
 as data security, data quality and data
 ingestion framework
- Master data management (MDM)

 and Data quality as a service Provide MDM and data quality
 processes as a centralized services
 to maintain data consistency across
 enterprise





Key Use Cases



Regional Compliance and Regulations

Establish individual data products for regional compliance and operational data (multilingual). For example, multinational retail or finance organizations can address their local compliance using a data product specific to their local region. The 'federated' aspect of data mesh means that governance policies are still defined centrally, but they are implemented by each domain team, which is responsible for making all the data that is produced in its domain and discoverable to other domains.



Sales and Marketing

Demographic based sales and demand forecasting. For example, usage reports for video streaming organizations customer behavior analysis for retail.



Federated Enterprise Governance

Federated data governance for key data elements such as customers and products to enable independent data and business processes. For example, customer data platform (CDP) for banking customers, B2B & B2C inventory management for individual regions for global manufacturer.



Manage Excessive Data Growth

Managing operational analytical data for large organization with a very high data, for example, internet of things (IoT)data from sensors and equipment can be massive. Now instead of putting a global operations team, organizations can create data products and assign owners specific to the logical Data products for better managing the operations. Data mesh architecture helps in essentially decoupling the computational governance from the domain ownership where the domain owners really worry about "shipping" the data product while the infrastructure team can scale for the additional computational capacity as needed. A similar use case could be on traffic data analysis for large cities.



Data Mesh Empowers Data Ownership for Data Scientists and Analysts

Data mesh enables data scientists and data analysts get access to higher quality data. Data scientists and data analysts suffer when data is incomplete, incompatible, or outdated. Data mesh addresses this problem by keeping data with the groups or teams that know it best. Once they become responsible for the data, and for making it easily and readily available to other groups within the organization, data analysts and data scientists end up working with higher quality data. This, in turn, yields more accurate results.



Our Solution

Following are the implementation journey, which starts with readiness assessment and could be summarized in a 9-step approach as shown below not necessarily in the exact sequence but should be aligned to the steps.



01. Readiness Assessment

Data maturity level, data sources/domains, and governance framework



06. Implementation Methodology

Agile/Waterfall methodology. plan, CRP/demo-based checkpoints, Data-Ops



02. Data Product Assessment

Feasibility, cost vs ROI, and KPI



07. Data and Integration Infra

Data storage, data processing optimization plan, and data growth computation



03. Architecture Road Map

End state definition, transition state definition, adaptive global source/subscriber/third party onboarding



08. Visualization

Global/data validation and quality rules, databases and file data extraction, domain specific, and global UI platform



04. Federated Data Governance

Governance and standards, automation and computation, domain specific governance



09. AI/ML Platform

Adaptive and augmented governance and metadata management



05. Design of Data Product

Code, infrastructure, data and metadata, modelling (including DP relationships)



Tech Mahindra and AWS Joint Solution Enabling Technology Mapping

DATA SOURCE

STRUCTURED

OLTP RDBMS

UNSTRUCTURED

- Social Media Data
- Streaming Data

SEMI STRUCTURED

- XMI
- JSON

DATA INGESTION







AWS Glue

PROVISIONING AND SERVICES





Data Governance Quality

AWS Glue

DATA STORAGE AND DATABASE



CONSUMPTION

Data-as-a-Service

VaaS Framework

Insight-as-a-Service

Subscription Management

ARCHITECTURE I MDM I QUALITY I GOVERNANCE I SECURITY I OBSERVABILITY







Key AWS Components

Data Lake - Provide a data storage and analytics solution

Amazon RDS - Commercial and open-source Databases

Amazon Aurora - MySQL and PostgreSQL compatible

Amazon DynamoDB - Non relational database service

AWS Glue - Data quality and integration service



Self Service Ingestion Framework for Modern Data Analytics Platforms

Metadata Driven Automated Ingestion



Data Migration & Quality Platform

- DQ assessment, consulting & recommendation
- Data cleansing, standardization & reconciliation



Any Source to Any Destination Cloud Migration Accelerator

- Data Search & Discovery
- Schema & SQL Migration



Self Service Ingestion Framework for Modern Data Analytics Platforms

- Data migration assessment and strategy
- Report Rationalization





Benefits



Automation

Global API and batch based framework with data register for source and subscriber onboarding, adaptive governance, metadata graph and adaptive process automation.



Flexible Integration Framework Enable global and data product framework to automatically share data from individual and across data products in both real time (streaming) and batch mode (global ETL framework).



Model Data as Products

Create data products based on specific domains, consumer groups or business lines (e.g., sales, services, finance, engineering, analytics) that is self-described and connected.



Data as a Service Provide reliable, performance intensive data as a service to consumers across platforms, format and medium.



Self Service

Automated integration for operational and analytical applications using data registry and global API framework.



Reliable Data and Self-Heal

Data quality (DQ) as a service based on governance policies with robust error framework along with use case based self-healing capability.



Seamless Performance Scalable and monitored seamless performance through aligned reporting needs to data product models.

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