# Alternative Cloud-Native Data Platform Architecture on Microsoft Azure

This guide details the implementation of a modern, cloud-native data platform architecture on Microsoft Azure, focusing on services such as **Azure Synapse Analytics**, **Snowflake**, **Azure Databricks**, and other Azure-native components for robust data processing, storage, transformation, and analytics capabilities.

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## 1. Core Architecture Overview

This architecture leverages Azure's managed, serverless, and scalable services to:

- Enable **real-time data processing** and **advanced analytics**.
- Minimize operational overhead and infrastructure management through cloud-native solutions.
- Utilize Snowflake as the data warehouse and Azure Databricks for big data processing and machine learning workflows, providing flexibility for both structured and unstructured data.

## 2. Recommended Architecture Components

### 2.1 OLTP Database (Transactional Layer)

• **Azure SQL Database**: Use Azure SQL Database as the main OLTP database, offering scalability and high availability.

## 2.2 NoSQL Database (Catalog Data)

• **Azure Cosmos DB**: Store catalog and product metadata in Cosmos DB, which supports low-latency, high-throughput operations for distributed data.

## 2.3 Staging Layer (Real-Time Stream Processing)

• Azure Event Hubs: Capture and process real-time streaming data with Azure Event Hubs.

#### 2.4 Data Warehouse

• **Snowflake on Azure**: Use Snowflake for scalable, high-performance data warehousing, optimized for analytics and easy integration with Azure services.

### 2.5 Data Lake Storage

 Azure Data Lake Storage Gen2 (ADLS Gen2): Use ADLS Gen2 as the data lake to store raw and processed data, accessible by both Snowflake and Databricks for a unified data repository.

### 2.6 Big Data and Machine Learning Platform

• **Azure Databricks**: Employ Databricks for scalable data processing, analytics, and machine learning. Utilize Databricks' Delta Lake to provide lakehouse architecture features.

### 2.7 Data Orchestration

 Azure Data Factory: Use Azure Data Factory for orchestrating ETL processes and data workflows.

### 2.8 Business Intelligence

 Power BI: Utilize Power BI for data visualization, directly connecting it to Snowflake and Databricks for real-time analytics.

## 3. Detailed Step-by-Step Implementation

### **Module 1: Transactional Database Design and Real-Time Ingestion**

- 1. Set Up Azure SQL Database for OLTP:
  - Create an **Azure SQL Database** instance to manage transactional data.
  - Configure firewall settings to allow access from necessary Azure resources and your local IP.
- 2. Enable Change Data Capture (CDC):
  - Enable CDC on the necessary tables in Azure SQL Database.
- 3. Stream Real-Time Changes with Azure Event Hubs:
  - Configure **Azure Data Factory** or **SQL Data Sync** to stream real-time data changes from Azure SQL Database to **Azure Event Hubs**.

### **Module 2: NoSQL Catalog Database Setup**

- 1. Set Up Azure Cosmos DB for Metadata:
  - Create **Azure Cosmos DB** and configure it to support high-volume catalog data.
  - Select the **API** based on data access requirements (e.g., SQL API for JSON data, Cassandra API for key-value data).
- 2. Integrate Cosmos DB with Data Lake:
  - Use Azure Data Factory to export data periodically from Cosmos DB to ADLS
    Gen2 for historical analysis and integration with the data lake.

#### Module 3: Data Lake and Data Warehouse Architecture

- 1. Create an ADLS Gen2 Storage Account:
  - Set up an **Azure Data Lake Storage Gen2** account to serve as the data lake.
  - Organize the storage with appropriate folder structures (e.g., raw, curated, enriched).
- 2. Set Up Snowflake on Azure:
  - Provision a Snowflake account on Azure and connect it to ADLS Gen2 for seamless data access.
  - Use **Snowflake's External Tables** feature to directly query data stored in ADLS Gen2 without loading it into Snowflake.
- 3. Establish Lakehouse Architecture with Delta Lake:
  - Use **Azure Databricks Delta Lake** to enable ACID-compliant transactions and create a unified data layer for real-time and historical analytics.

### **Module 4: Data Transformation and Processing**

- 1. Data Ingestion and Transformation with Azure Databricks:
  - Set up **Azure Databricks** workspaces for data engineering and machine learning.
  - Use **Databricks Delta Lake** to transform raw data into refined datasets.
- 2. Process and Enrich Data in Snowflake:
  - Leverage **Snowflake's SQL** capabilities for data transformation and analytics on structured data.
- 3. Implement Data Cleansing and Transformation in Azure Data Factory:
  - Configure **Azure Data Factory** pipelines for batch ETL processes, cleansing, and joining datasets before loading them into the data warehouse.

### **Module 5: Orchestration and Workflow Management**

- 1. Orchestrate ETL Pipelines with Azure Data Factory:
  - Use **Azure Data Factory** for automating and scheduling ETL workflows.
  - Set up data pipelines that connect **Azure SQL Database**, **Cosmos DB**, **Event Hubs**, and **ADLS Gen2** for data movement and transformation.
- 2. Use Databricks Notebooks for Real-Time Data Processing:
  - Develop **Databricks Notebooks** to handle real-time data transformations and analytics on event-driven data from Event Hubs.

### **Module 6: Machine Learning Model Deployment**

- 1. Setup Azure Databricks MLflow for Model Management:
  - Use **Databricks MLflow** within Azure Databricks for model experimentation, tracking, and management.
- 2. Deploy Models to Production with Azure Machine Learning:
  - Register models in **Azure Machine Learning** and deploy them as web services.
  - Configure these services to connect with Snowflake and Power BI for real-time model inference.

### **Module 7: Business Intelligence and Visualization**

1. Create Dashboards in Power BI:

- Connect **Power BI** to Snowflake and Databricks for interactive data visualizations.
- Use **DirectQuery** mode in Power BI to enable near real-time reporting on Snowflake data.

### 2. Integrate Power BI with Azure Synapse Analytics:

 Connect **Power BI** with Synapse for additional analytical power and near real-time insights.

## 4. Benefits of the Modernized Architecture

- **Scalability and Cost Efficiency**: Azure's cloud-native services scale dynamically with demand, and Snowflake's pay-as-you-go pricing reduces costs for sporadic workloads.
- **Real-Time Processing**: Azure Event Hubs and Databricks Delta Lake support real-time streaming and processing, offering up-to-date analytics and insights.
- **Operational Simplification**: Managed services such as Azure Data Factory and Databricks reduce the need for manual infrastructure management and maintenance.
- **Unified Lakehouse Architecture**: Integrating Snowflake with Databricks Delta Lake provides a seamless platform for structured and unstructured data.
- Comprehensive ML and BI Capabilities: Combining Databricks, Azure Machine Learning, and Power BI allows for advanced analytics and machine learning model deployment at scale.

This architecture provides a robust, flexible, and scalable platform for end-to-end data management, analytics, and machine learning on Microsoft Azure.