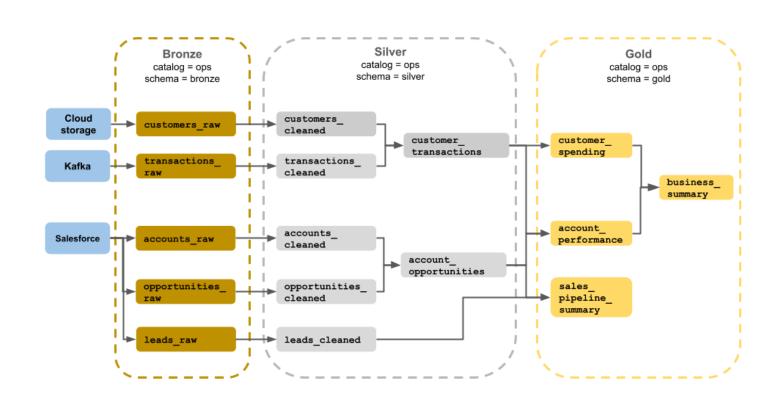
# Data Pipelines with Delta Live Tables

Managing and transforming data with innovative tools

## Session Outline

- Session Overview
- Medallion Architecture
- Delta Live Tables
- Hands-on Labs

# Introduction to Medallion Architecture & Delta Live Tables



### **Understanding Medallion Architecture**

Medallion Architecture is designed to organize data into layers that enhance data quality and accessibility.

#### **Delta Live Tables Overview**

Delta Live Tables (DLT) simplify building and managing data pipelines while ensuring high data reliability.

#### **Benefits of DLT**

Using DLT leads to better performance, reliability, and automation in data processing workflows.

## Medallion Architecture

#### **Medallion Architecture Overview**

An introduction to Medallion Architecture, discussing its significance and application in data engineering.

#### **Bronze, Silver, Gold Layers**

Explore the different layers of Medallion Architecture: Bronze, Silver, and Gold, each serving unique purposes.

#### **Introduction to Delta Live Tables**

Learn about Delta Live Tables and how they facilitate data pipeline management and processing.

#### **Hands-on Lab**

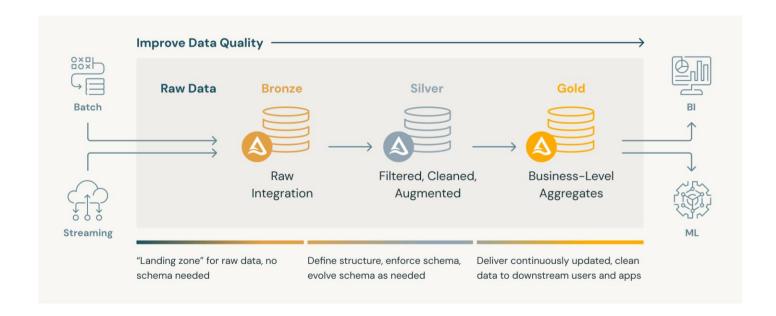
Participate in a hands-on lab session to build Medallion Architecture and gain practical experience.

#### **Navigating Delta Live Tables UI**

Overview of the user interface for Delta Live Tables, emphasizing navigation and functionality.

# Medallion Architecture

### **Overview and Benefits**



#### **Medallion Architecture**

Medallion architecture describes a data refinement process structured in three layers: Bronze, Silver, and Gold.

#### **Data Quality Improvement**

Implementing this architecture enhances overall data quality, ensuring reliable and accurate information.

#### **Governance and Performance**

This structure leads to better data governance and improves system performance for users.

## **Bronze Layer (Raw Data)**

```
import dlt
from pyspark.sql.functions import *

@dlt.table(
   comment="Raw sales data ingested to Bronze layer",
   table_properties={
    "quality": "bronze"
   }
)

def bronze_sales():
   return (
        spark.read.json("/databricks-datasets/samples/sales/data.json")
   )
```

#### **Definition of Raw Data**

Raw data is data that has not been processed or analyzed. It represents the original state of information collected.

#### **Minimal Transformations**

The bronze layer involves minimal transformations to maintain data integrity. This ensures the original data remains intact for future analysis.

#### **Purpose of Bronze Layer**

The purpose of the bronze layer is to serve as initial storage for incoming data, allowing for easy access and retrieval.

#### **Examples of Raw Data**

Examples include raw sales figures, customer demographics, and product data, showcasing various types of incoming data.

## Silver Layer (Refined Data)

#### **Data Cleaning and Transformation**

The Silver Layer involves cleaned, transformed, and enriched data, ensuring it is ready for analysis.

#### **Data Validation and Quality Checks**

Important steps in this layer include data validation, quality checks, and schema enforcement to maintain data integrity.

#### **Joined Data Examples**

Examples include sales data joined with customer and product details, providing a comprehensive view of the data.

```
@dlt.table(
  comment="Cleaned and deduplicated Silver
sales data"
)
@dlt.expect("valid_quantity", "quantity > 0")
@dlt.expect_or_drop("no_nulls", "customerId
IS NOT NULL")
def silver_sales():
    df = dlt.read("bronze_sales")
    return df.filter("quantity >
0").dropDuplicates(["orderId"])
```

## Introduction to Gold Layer

#### **Aggregated Data**

The Gold Layer contains aggregated or highly curated data essential for effective business intelligence.

#### **Business KPIs**

This layer includes key performance indicators (KPIs) and metrics to monitor business performance.

#### **Optimized for Reporting**

The Gold Layer is optimized for business intelligence (BI) and reporting purposes, ensuring clarity and insight.

```
CREATE LIVE TABLE gold_sales_summary
COMMENT "Aggregated daily sales totals"
AS
SELECT
orderDate,
SUM(quantity * unitPrice) AS total_sales,
COUNT(DISTINCT customerId) AS unique_customers
FROM
LIVE.silver_sales
GROUP BY
orderDate;
```

# Benefits of Medallion Architecture



# Advantages of Medallion Architecture

#### **Incremental Data Quality**

Medallion Architecture enhances data quality over time, ensuring more and reliable data for decision-making.

#### **Easier Troubleshooting**

The architecture simplifies troubleshooting processes and improves datracking, making it easier to trace issues.

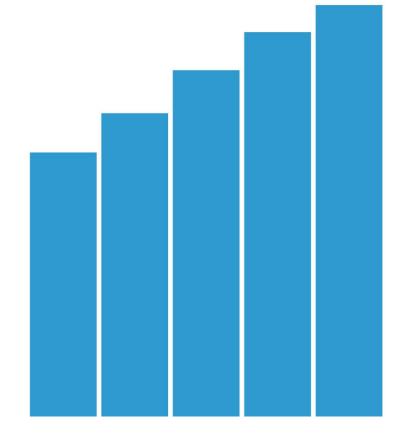
#### **Scalability**

Medallion Architecture supports scalability for large datasets, allowing organizations to grow their data capabilities efficiently.

#### **Multiple Downstream Consumers**

The architecture accommodates multiple downstream consumers, enhancing collaboration and data sharing across teams.





# Delta Live Tables

### Introduction to Delta Live Tables



#### **Managed ETL Framework**

Delta Live Tables is a managed ETL framework that leverages Delta Lake and Apache Spark to facilitate efficient data processing.

#### **Declarative Pipeline Creation**

Users can build data pipelines in a declarative manner using SQL or Python, which streamlines development and boosts productivity.

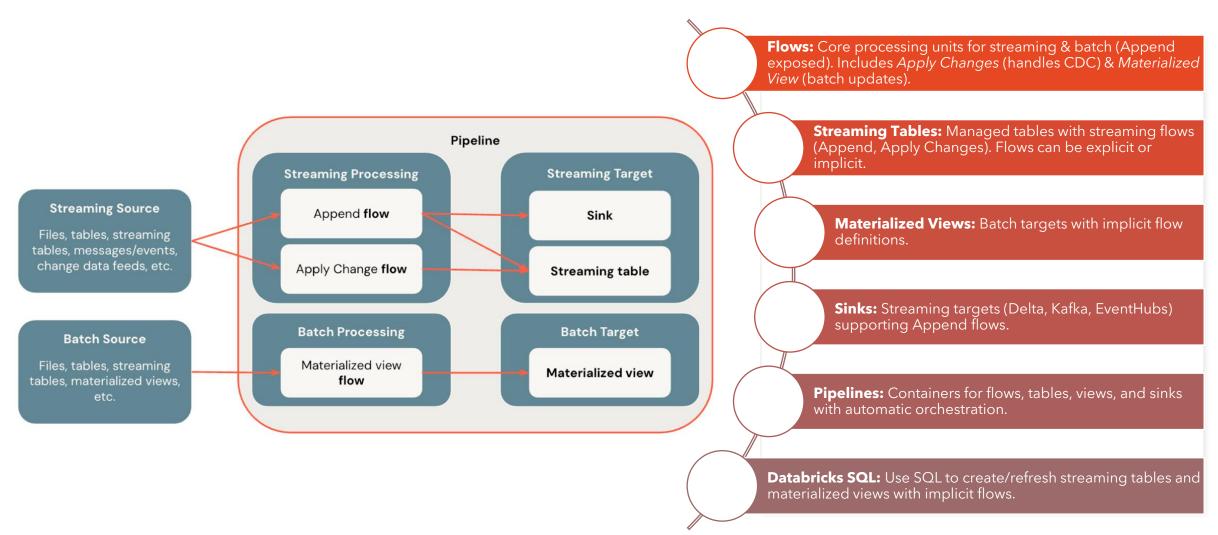
#### **Automated Infrastructure Management**

Delta Live Tables automates infrastructure management and orchestration, enabling users to concentrate on data transformation and analysis.

#### **Data Quality Enforcement**

The framework includes features for enforcing data quality and monitoring, ensuring that high-quality data is maintained throughout the pipelines.

## **Key Concepts of DLT**



# Benefits of Delta Live Tables

#### **Simplified ETL Pipelines**

Delta Live Tables streamline ETL processes, allowing developers to implement pipelines with significantly less code.

#### **Automatic Error Handling**

Automatic error handling and recovery mechanisms ensure robust data processing and reduce manual intervention.

#### **Declarative Data Quality Checks**

Declarative checks for data quality provide built-in expectations that enhance the reliability of data pipelines.

#### Scalable and Maintainable

Delta Live Tables support scalability and maintainability, making it easier to adapt pipelines to growing data needs.

#### **Real-Time Monitoring Dashboard**

A real-time monitoring dashboard allows users to track data pipelines, ensuring optimal performance and quick insights.

#### **DLT Architecture**

#### **Data Ingestion**

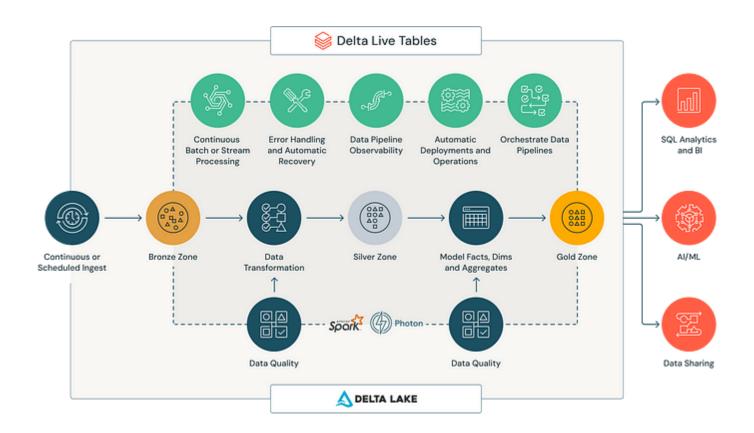
The DLT architecture begins with data ingestion from various sources, seamlessly bringing data into the processing pipeline.

#### **Transformations**

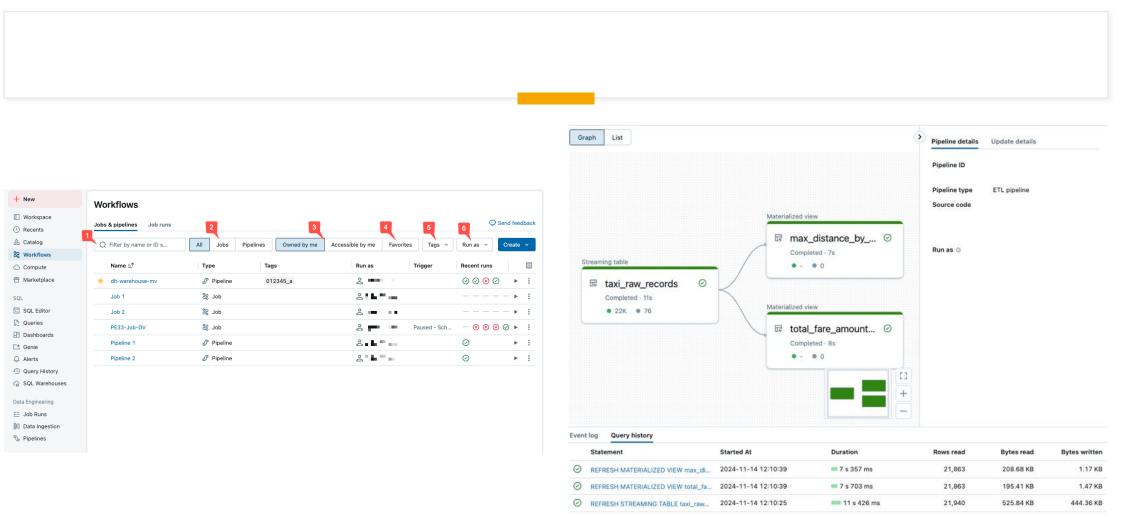
Data undergoes transformations within the DLT pipeline, allowing for data cleansing, enrichment, and preparation for analysis.

#### **Publishing Data**

After processing, the transformed data is published to Delta tables, making it available for analysis and reporting.



# **Monitoring and Trobleshooting**



## Pipeline Options Available in Databricks

Option	Description	When to Use	Key Pros	Key Cons
Delta Live Tables (DLT)	Managed framework for building reliable, tested ETL pipelines declaratively.	Preferred for complex pipelines with streaming, quality, monitoring needs.	Simplifies pipeline management, built-in quality, monitoring.	Requires learning new framework, some flexibility trade-offs.
Databricks Jobs (notebooks)	Schedule notebooks as jobs for running custom ETL or ML workflows.	Simple batch workflows, scheduled jobs, or ad-hoc runs.	Full control of code, flexible.	No built-in quality or lineage tracking. Must build manually.
Databricks Workflows	Orchestrate multiple jobs with dependencies, including multitask workflows.	Complex orchestration of multiple jobs/tasks.	Supports task dependencies, retries, alerts.	Still requires manual pipeline code management.
Apache Spark Streaming / Structured Streaming	Use Spark APIs directly for streaming data processing inside notebooks or jobs.	Custom streaming pipelines with fine-grained control.	Very flexible, powerful streaming.	Requires building your own monitoring, checkpointing, and quality checks.
Third-party Orchestration (e.g., Apache Airflow, Azure Data Factory)	External orchestrators triggering Databricks jobs or notebooks.	Complex enterprise pipelines with cross-system dependencies.	Rich orchestration features outside Databricks.	Adds external system complexity and cost.

# Hands-on Labs

# Lab Overview

#### **Medallion Architecture**

Implementing a simple Medallion Architecture involves understanding the Bronze, Silver, and Gold layers of data processing.

#### **Task Objective**

The objective is to create a clear understanding of each layer within the Medallion Architecture.

#### **Tools for Implementation**

Databricks notebooks and Delta Live Tables are essential tools for implementing the Medallion Architecture.

#### **Provided Data**

The lab will utilize provided CSV files containing sales, customer demographics, and products data.

# Delta Live Tables UI Navigation

#### **Accessing Delta Live Tables**

Learn how to access Delta Live Tables from the Databricks workspace for efficient data management.

#### **Pipeline Creation and Configuration**

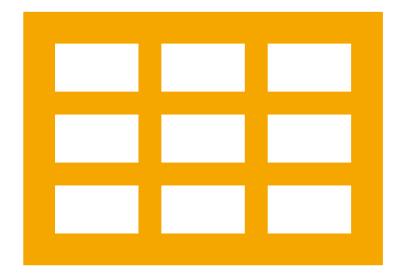
Walkthrough of creating and configuring pipelines in Delta Live Tables for data processing.

#### **Monitoring and Management**

Overview of how to monitor and manage pipeline performance and job runs within the UI.

#### **Understanding UI Elements**

Explore key UI elements including job runs, data freshness, and error logs for effective management.



# Creating Basic DLT Pipelines

#### **Creating a New DLT Pipeline**

Begin by creating a new DLT pipeline, which is the foundation for managing data workflows.

#### **Defining Tables**

Define the Bronze, Silver, and Gold tables to categorize data at different processing levels.

#### **Running the Pipeline**

Once set up, run the pipeline and monitor its performance in realtime for effective data processing.

#### **Validating Results**

Finally, validate the results to ensure the data has been processed correctly throughout the pipeline.

## **Challenge Labs**

#### **Data Quality in Silver Layer**

In this lab, participants will implement data quality expectations to ensure reliable data in the Silver layer.

#### **Monthly Customer Segment Analysis**

This lab focuses on creating a new Gold table for analyzing customer segments on a monthly basis.

#### **Schema Evolution in Bronze Tables**

Participants will learn how to handle schema evolution in Bronze tables effectively during this lab.

