

From Imperative to **Declarative**

# Building Maintainable ETL Pipelines in Spark

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Intro x

# Why **ETL** Pipelines Matter

- High-volume data requires automation, reliability, and quality



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- Declarative pipelines solve orchestration, state, and lineage automatically



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# Why ETL Pipelines Matter

- High-volume data requires automation, reliability, and quality
- Imperative Spark pipelines become large, fragile, and hard to orchestrate
- Declarative pipelines solve orchestration, state, and lineage automatically
- Databricks implements Declarative Pipelines through Delta Live Tables, but the paradigm exists independently



Intro

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# Imperative Pipelines in Spark

```
...  
songs_raw_stream = (  
    spark.readStream  
        .format("cloudFiles")  
        .schema(songs_schema)  
        .option("cloudFiles.format", "csv")  
        .option("sep", "\t")  
        .load(file_path)  
)  
  
songs_raw_query = (  
    songs_raw_stream.writeStream  
        .format("delta")  
        .outputMode("append")  
        .option("checkpointLocation", checkpoint_path)  
        .start(bronze_path)  
)
```





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## Limitations of Imperative Spark

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## Limitations of Imperative Spark

- No built-in data quality enforcement
- No auto lineage
- State management is manual
- Hard to add more stages (bronze → silver → gold)
- Pipelines require external schedulers (Airflow, cron, jobs API)

Intro Imperative SDP X

# What Is a Declarative Pipeline? (SDP)

*"Define what your tables should contain, not how to produce them."*

```
...  
@dp.table(comment="Raw data from a subset of the Million Song Dataset; a collection  
of features and metadata for contemporary music tracks.")  
def songs_raw():  
    return (spark.readStream  
        .format("cloudFiles")  
        .schema(songs_schema)  
        .option("cloudFiles.format", "csv")  
        .option("sep", "\t")  
        .load(file_path)  
    )
```





Intro Imperative SDP x

## Characteristics

- Spark infers execution order



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- Spark infers execution order
- Spark builds table dependencies
- Automatic incremental processing
- Automatic fault tolerance
- Cleaner, modular transformation logic



# Spark Declarative Pipelines vs. Imperative Jobs

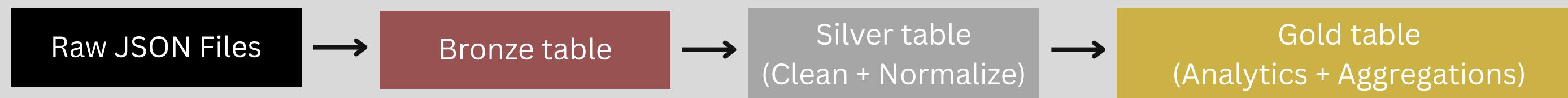
Feature	Imperative	Declarative
Programming style	Procedural	SQL-like, functional
Control	Manual	Automated
State Management	You Manage checkpoints	Framework Handles it
Lineage	None	Auto lineage graph
Optimization	Manual	Automatic
Fault Tolerance	You Implement	Built-in
Data Quality	Manual	Declarative rules



Intro Imperative SDP Comparison

Example X

# Example Pipeline Architecture



Imperative Spark	SDP
You Write 3 jobs	You Declare 3 tables
You Manage Dependencies	Spark figures out dependencies



Intro Imperative

SDP

Comparison

Example



# Tutorial in Databricks



**THANK YOU  
FOR LISTENING**