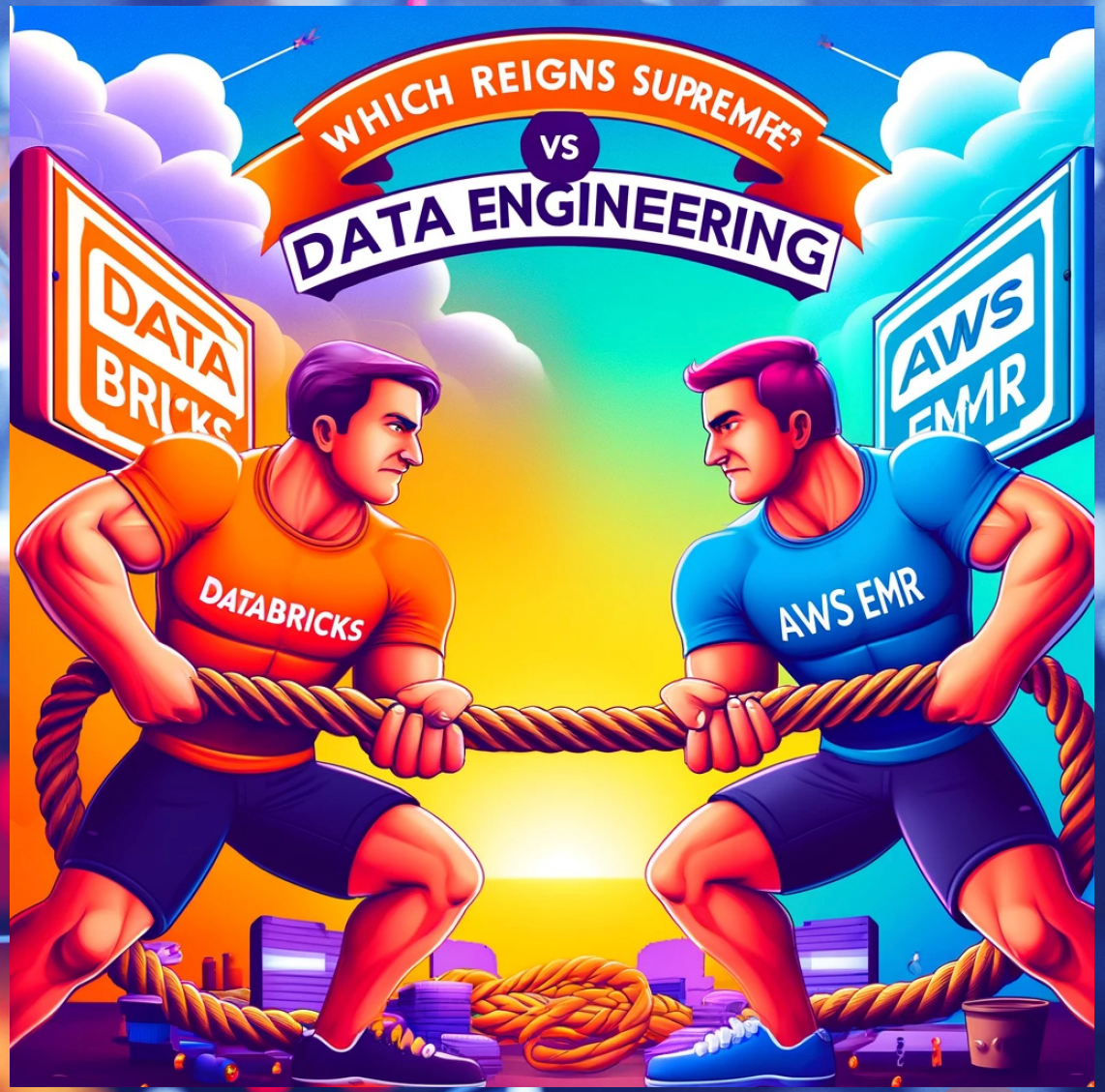
CLOUD ARBITRAGE FOR SPARK PIPELINES

Georg Heiler / Hernan Picatto



georgheiler.com/2024/05/02/cost-efficient-alternative-to-databricks-lock-in

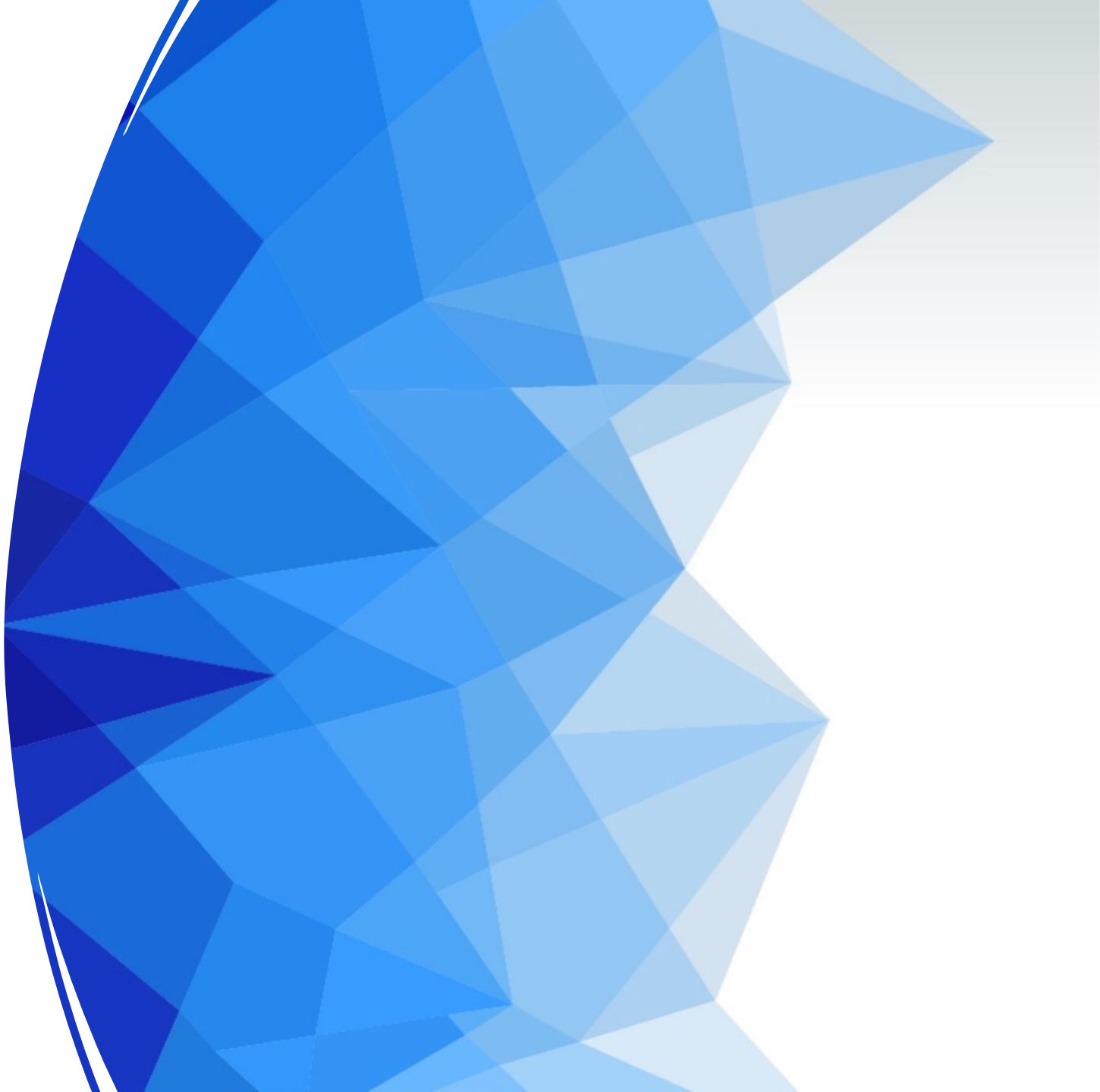
About the Speakers

- Data pipelines & AI
- Academia & Industry (telco)
- Supply Chain, Text analytics & data pipelines, graphs, spatial time series
- Meetup speaker & organizer

- Researcher @ASCII
- Former JPMorgan Chase
- Time series forecasting and causality detection, EVT analysis

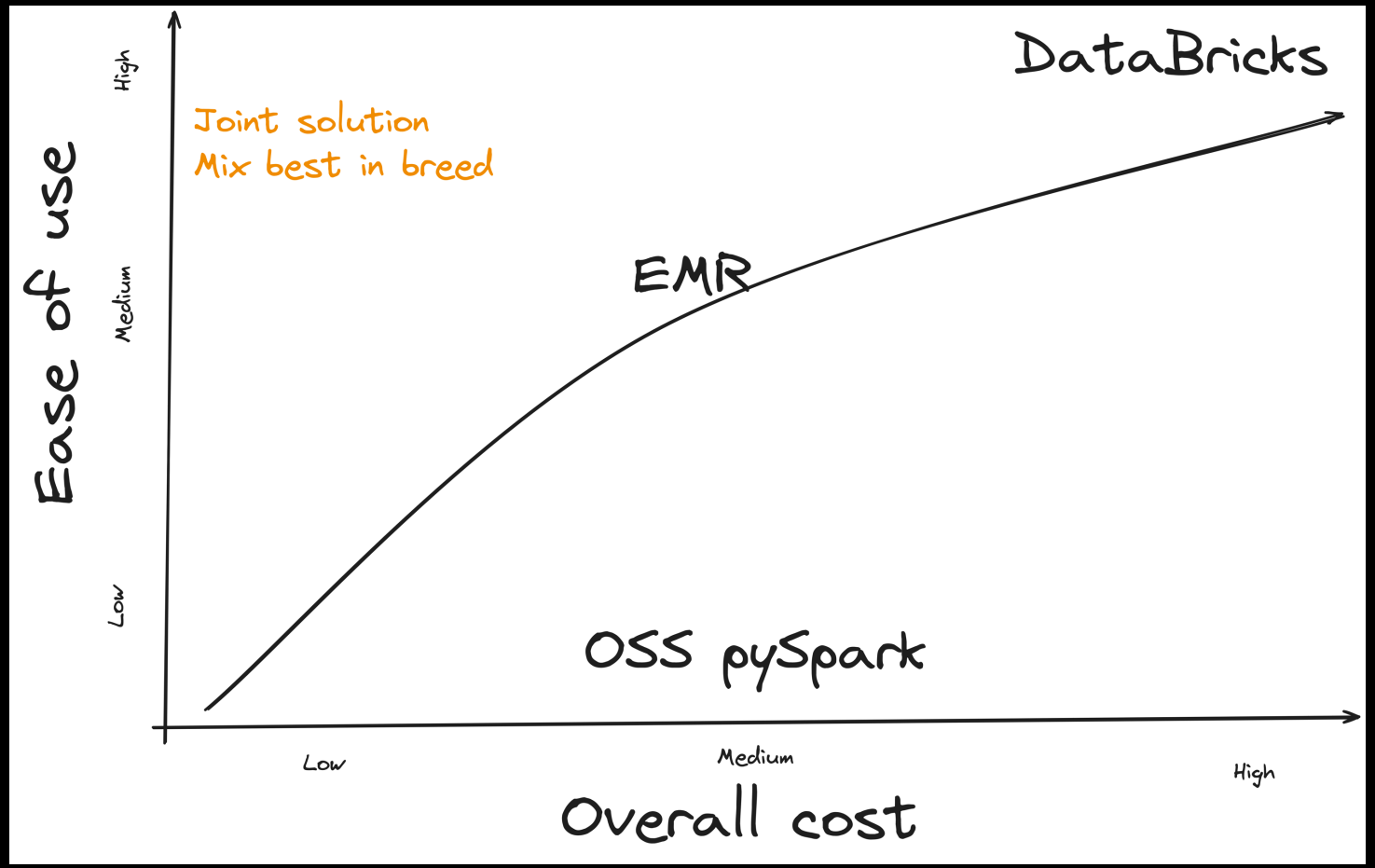
Agenda

- Results Overview
- History
- Problem Description & Vision
- Technology Introduction
- Implementation Architecture
- Results
- Learnings



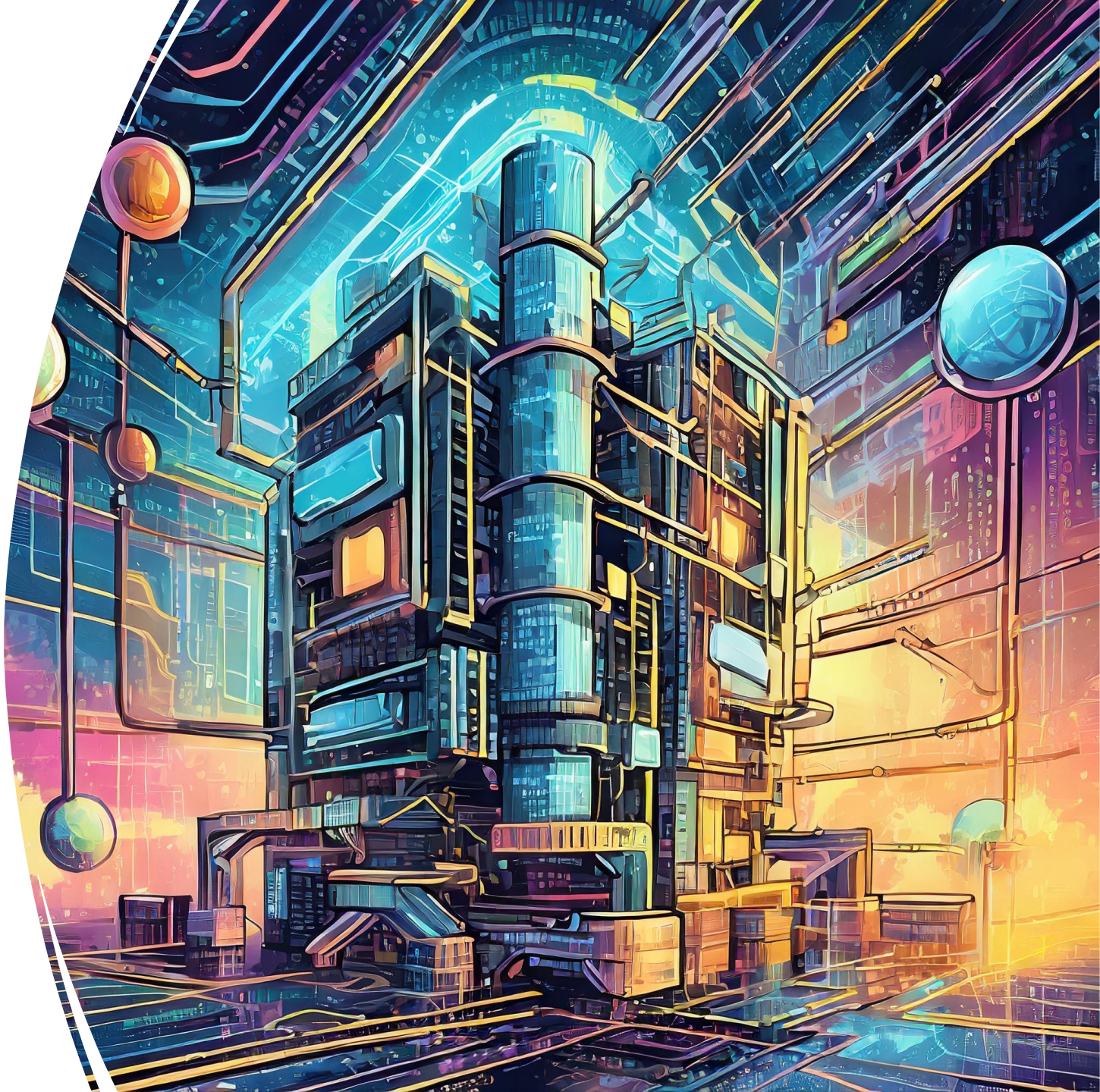
Results at a glance

- Achieved 43% Cost Reduction
- Software Engineering Practices
- Flexibility



History

- Mainframe
- Data warehouse
- Big Data (Hadoop)
- SQL on large data (Hive, Spark)
- Cloud DWH (Snowflake, bigquery)



PaaS offering

PaaS Solution Comparison

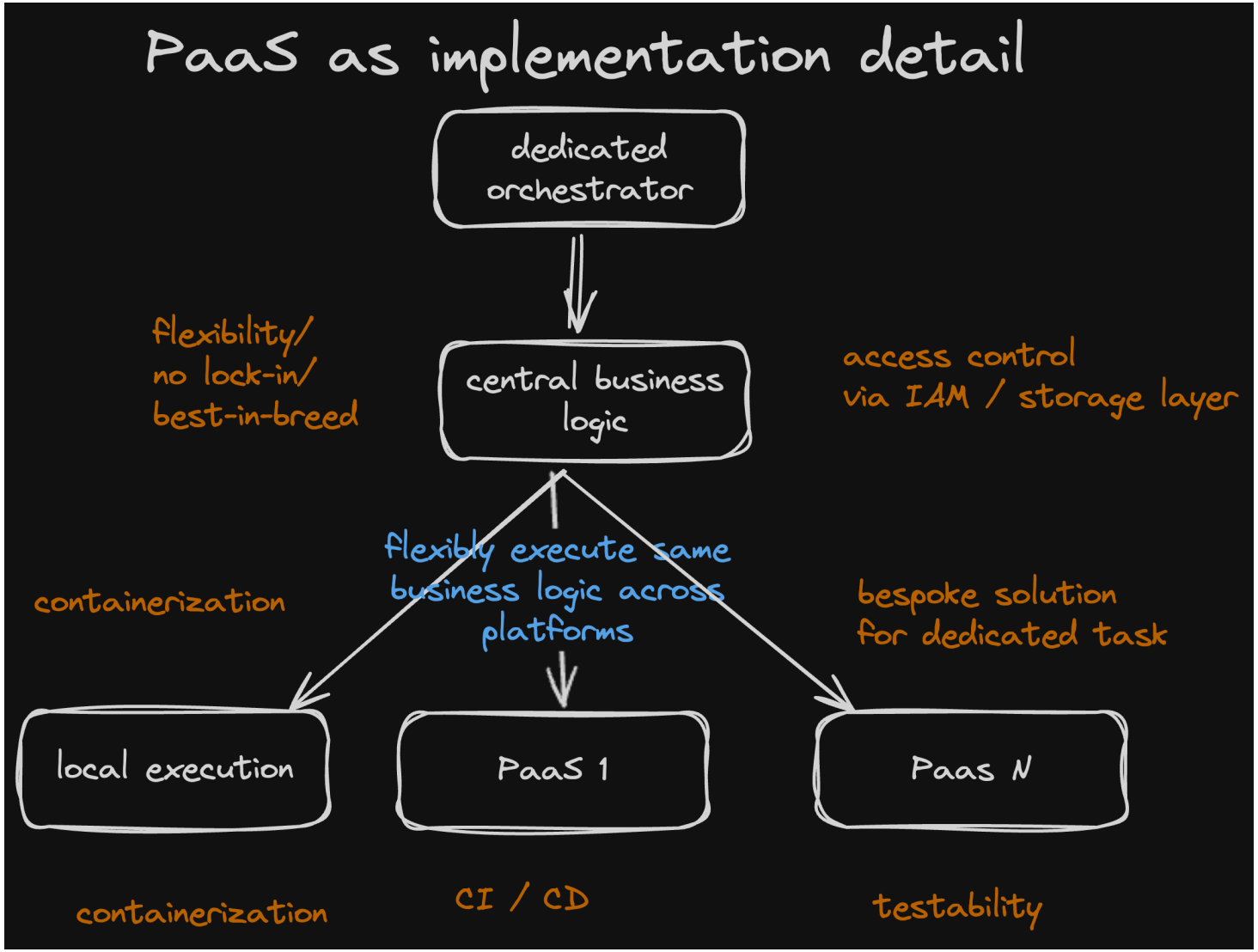
Databricks (DBR)	AWS Elastic Map Reduce (EMR)
• Easy to use • Can be expensive • Lock-in features (permissions, catalog) • Proprietary Photon engine	• Price efficient • Many tuning knobs available (& required) • OSS Spark managed (scaled)

Challenges

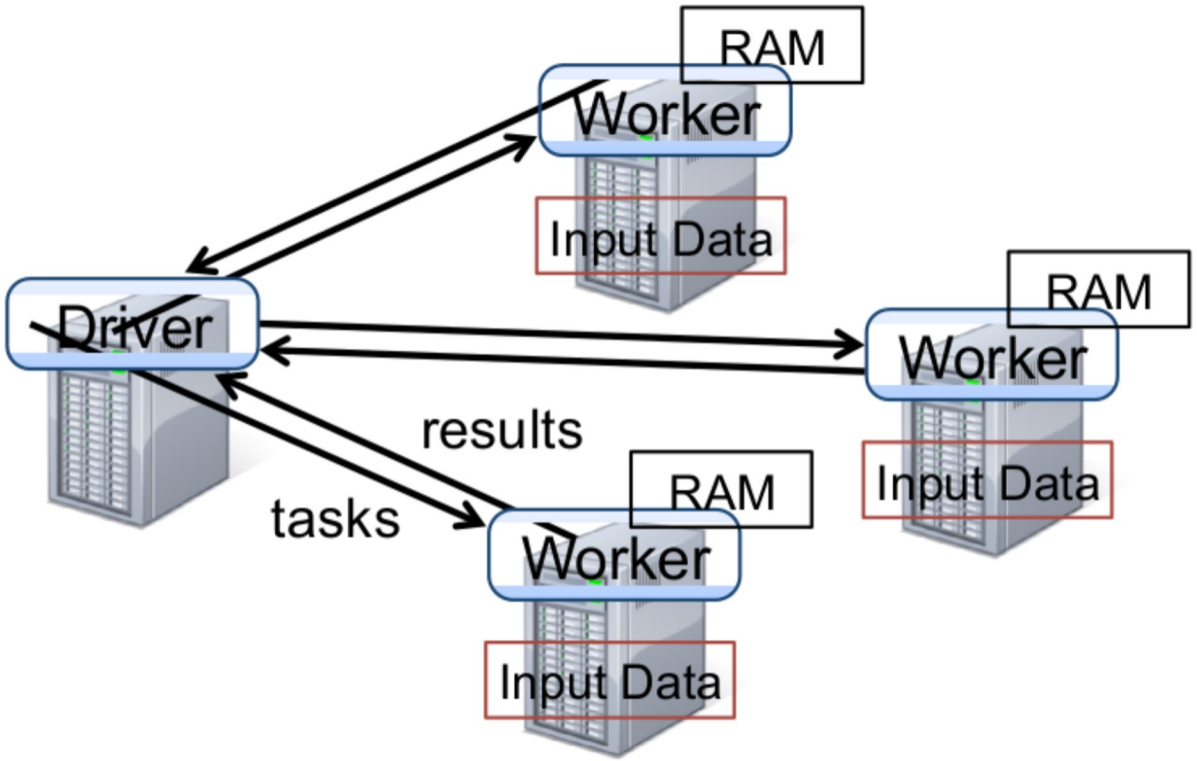
- Runaway expenses (usage-based pricing)
- Missing software engineering best practices (notebooks)
- Developer productivity reduced
- Vendor lock-in

Vision

- 0-cost switch
- Software engineering practices
- Cost reduction
- Reduce lock-in

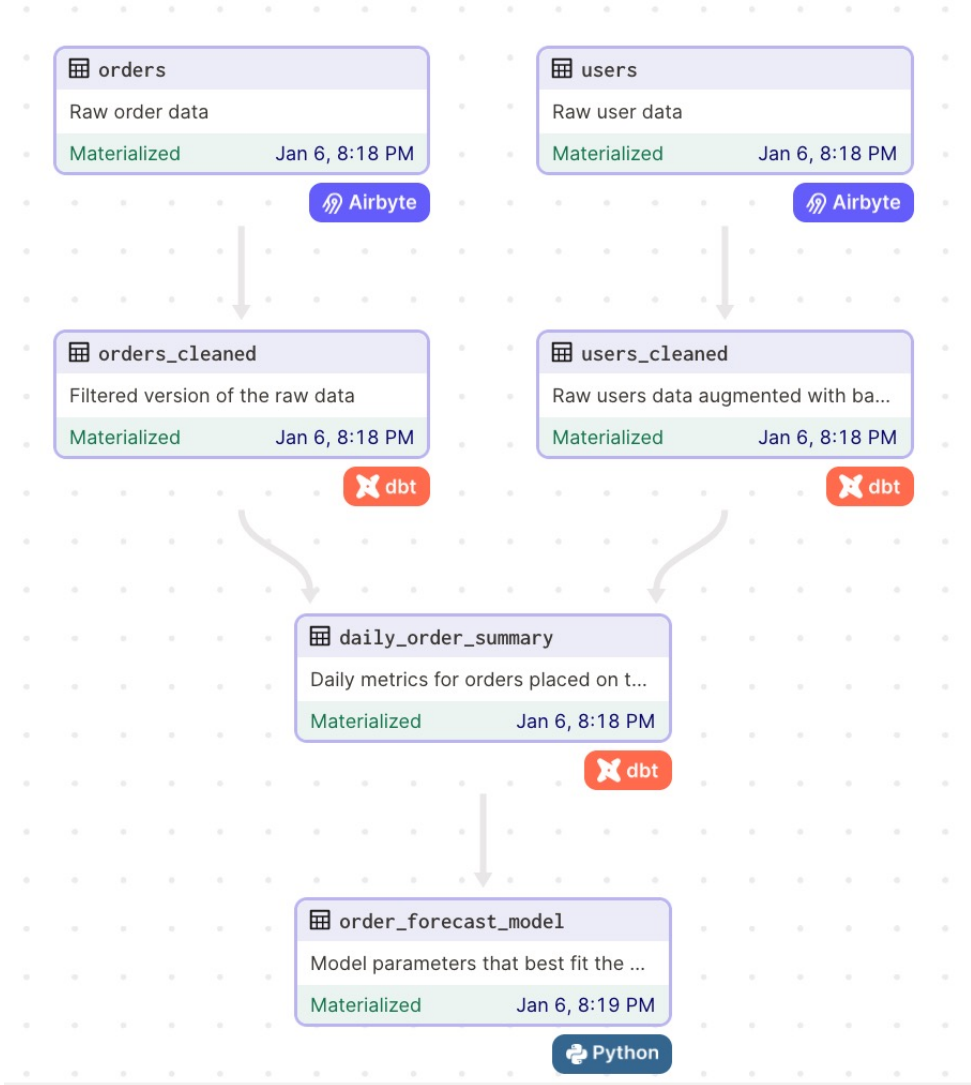


Spark at a glance

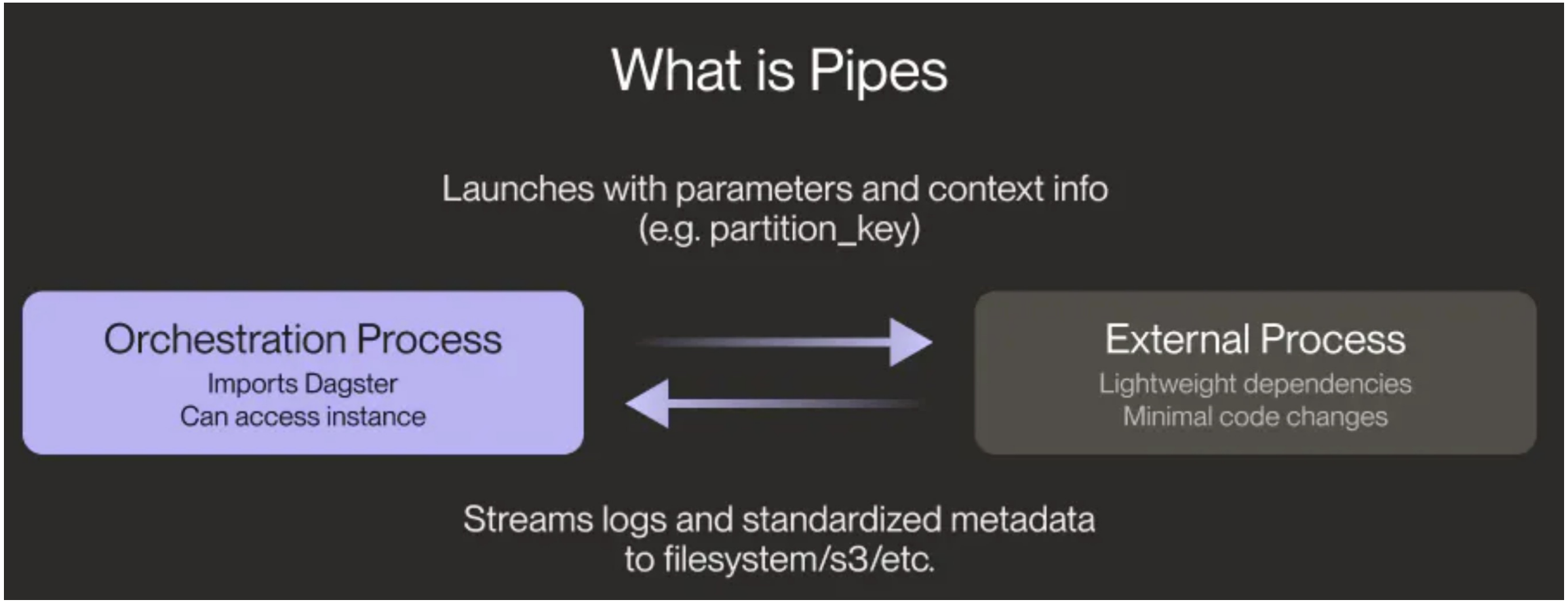


Dagster introduction

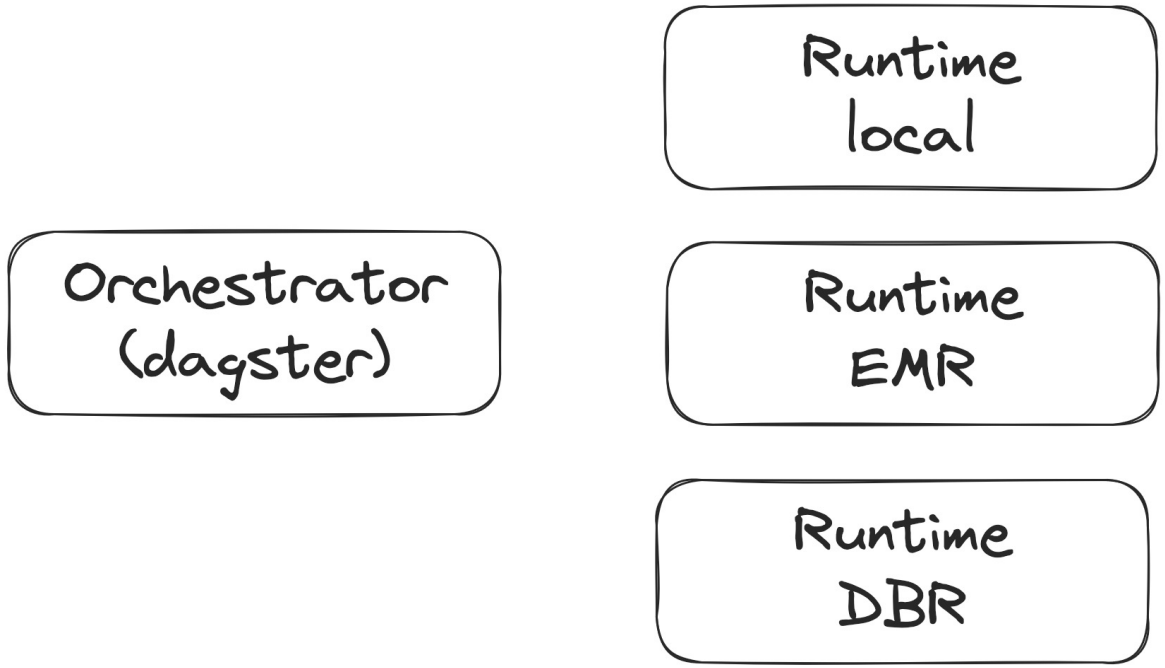
dagster



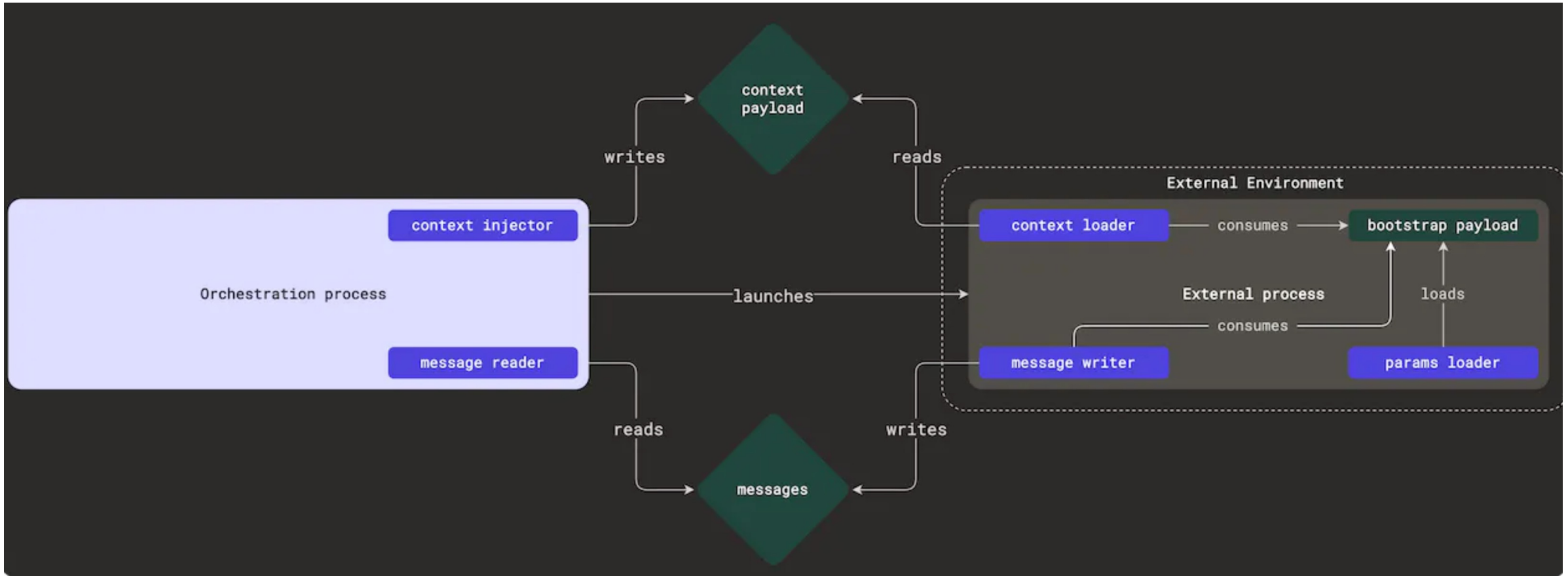
Dagster-pipes



High-level Architecture



Dagster-pipes - Architecture



Dagster-pipes Sample

External code (with metadata)

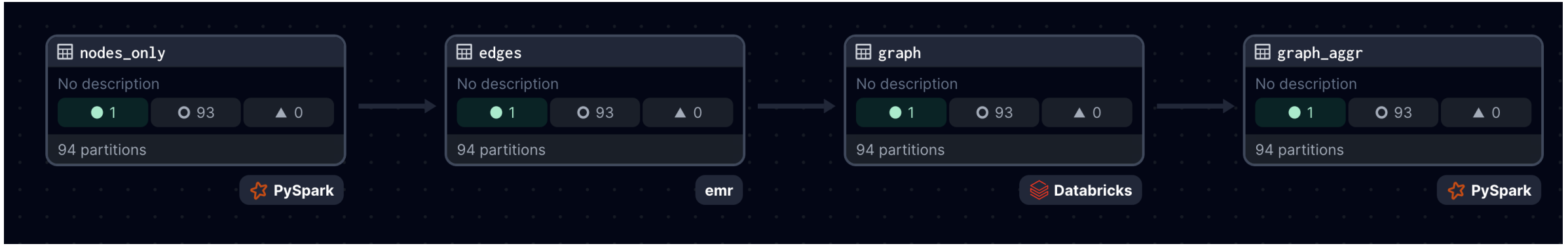
```python
def main():
    orders_df = pd.DataFrame({"order_id": [1, 2
    total_orders = len(orders_df)
    context = PipesContext.get()
    print(context.get_extra("foo"))
    context.log.info("Here from remote")
    context.report_asset_materialization(
        metadata={"num_orders": len(orders_df)}
    )

if __name__ == "__main__":
    with open_dagster_pipes():
        main()
```

Internal asset shim orchestrating the execution of external script

```python
@asset
def subprocess_asset(
    context: AssetExecutionContext, pipes_subprocess_client: PipesSubprocessClien
) -> MaterializeResult:
    cmd = [shutil.which("python"), file_relative_path(__file__, "external_code.py
    return pipes_subprocess_client.run(
        command=cmd,
        context=context,
        extras={"foo": "bar"},
        env={
            "MY_ENV_VAR_IN_SUBPROCESS": "my_value",
        },
    ).get_materialize_result()
```

Results


nodes_only
No description
1
93
0
94 partitions
PySpark
edges
No description
1
93
0
94 partitions
emr
graph
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Databricks
graph_aggr
No description
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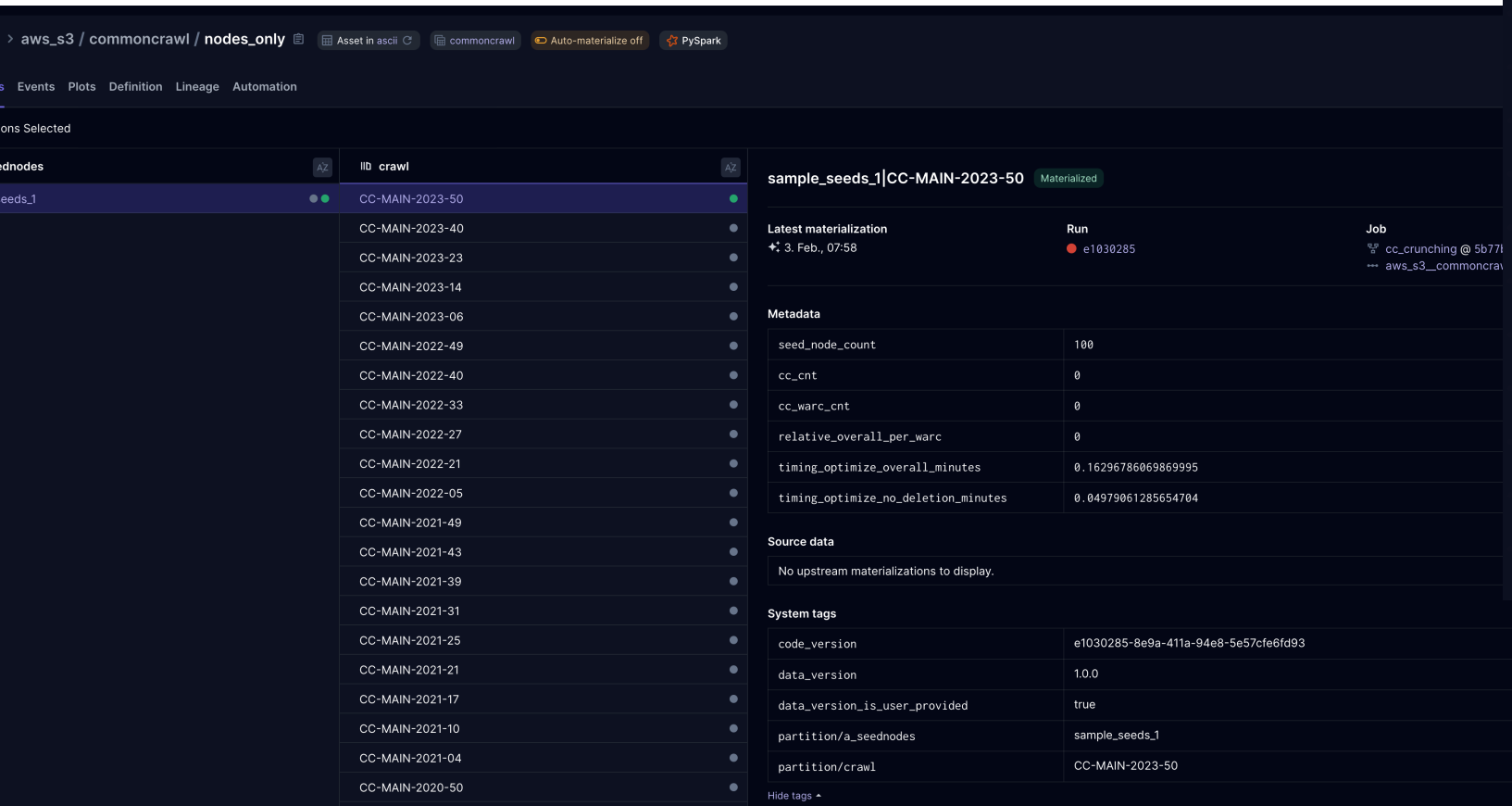
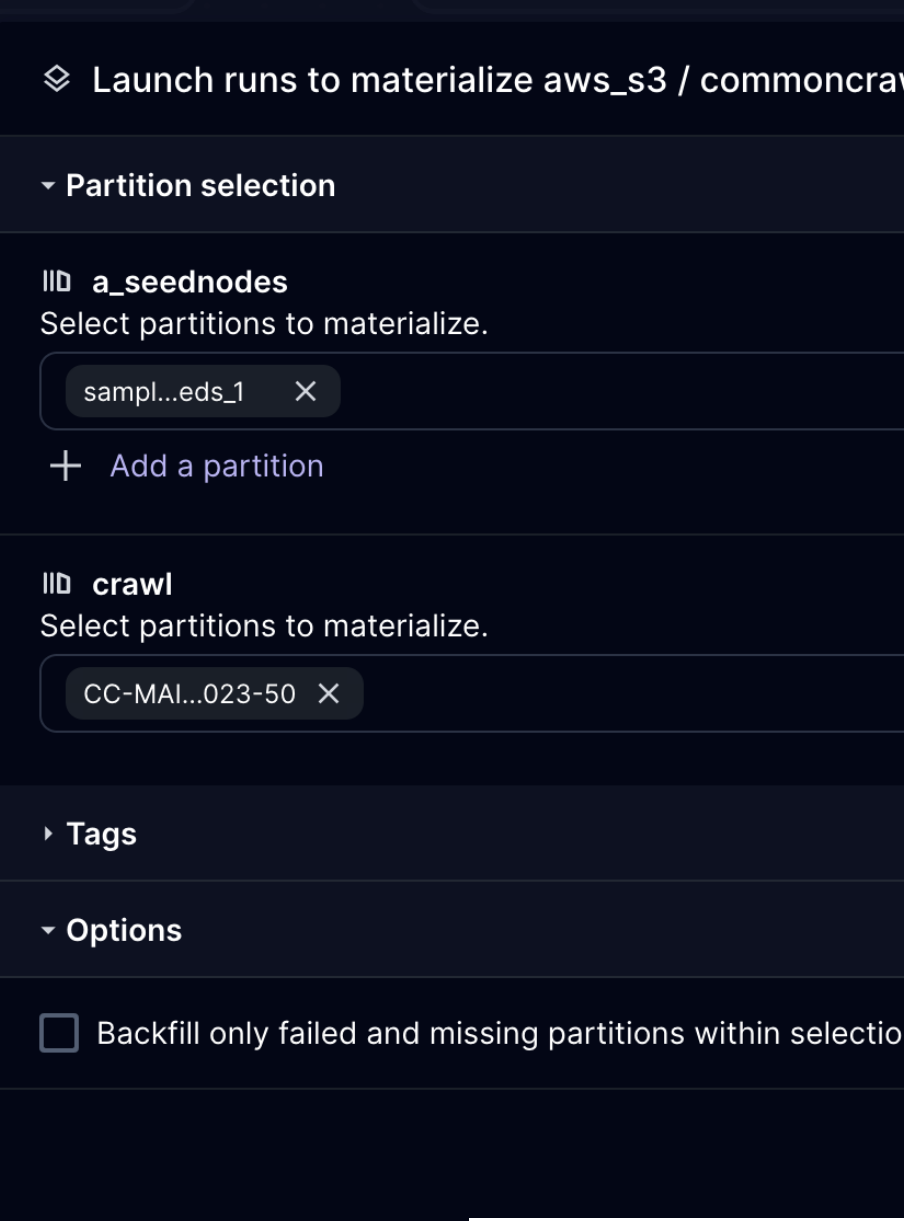
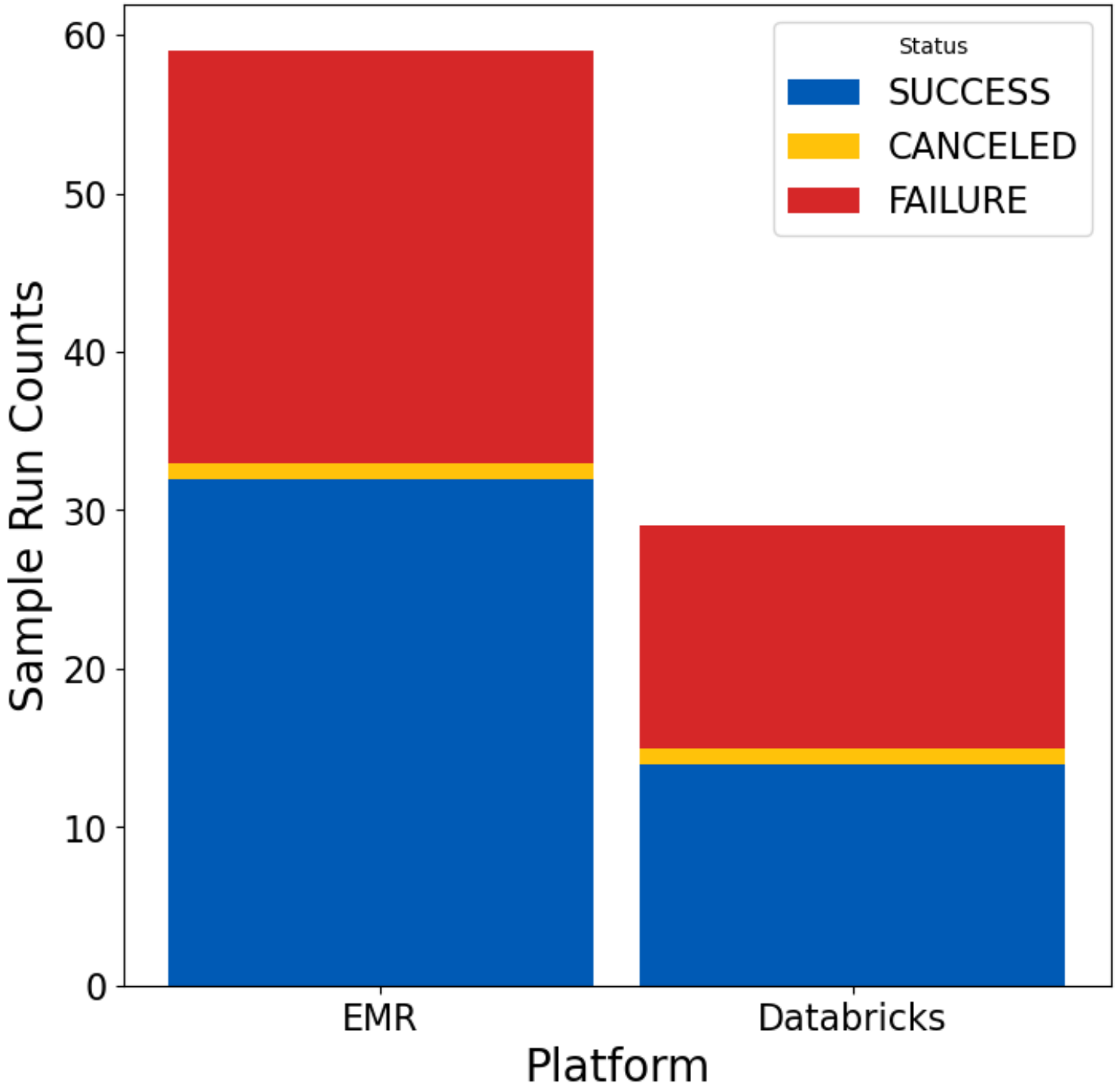
PySpark

Partitioned UI

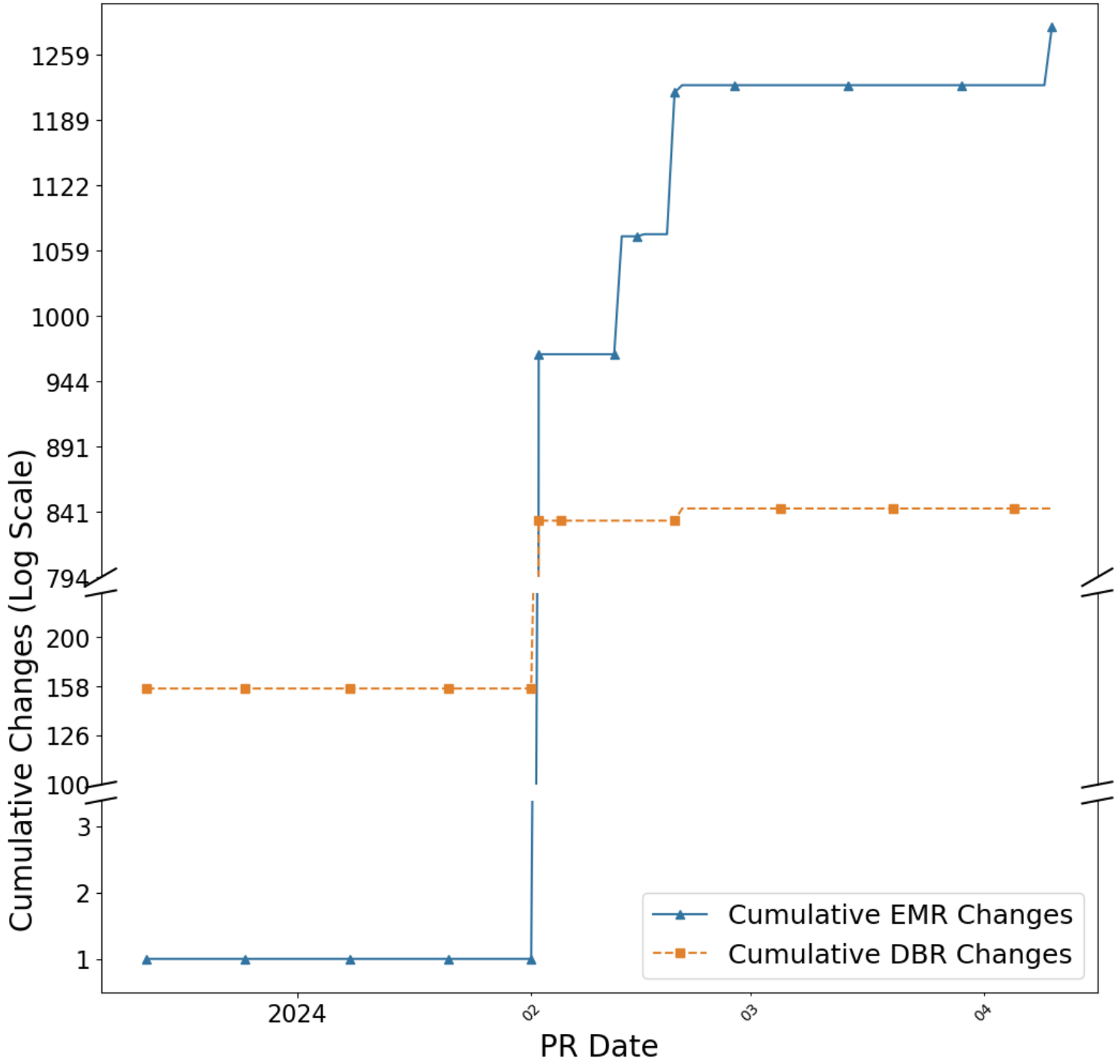




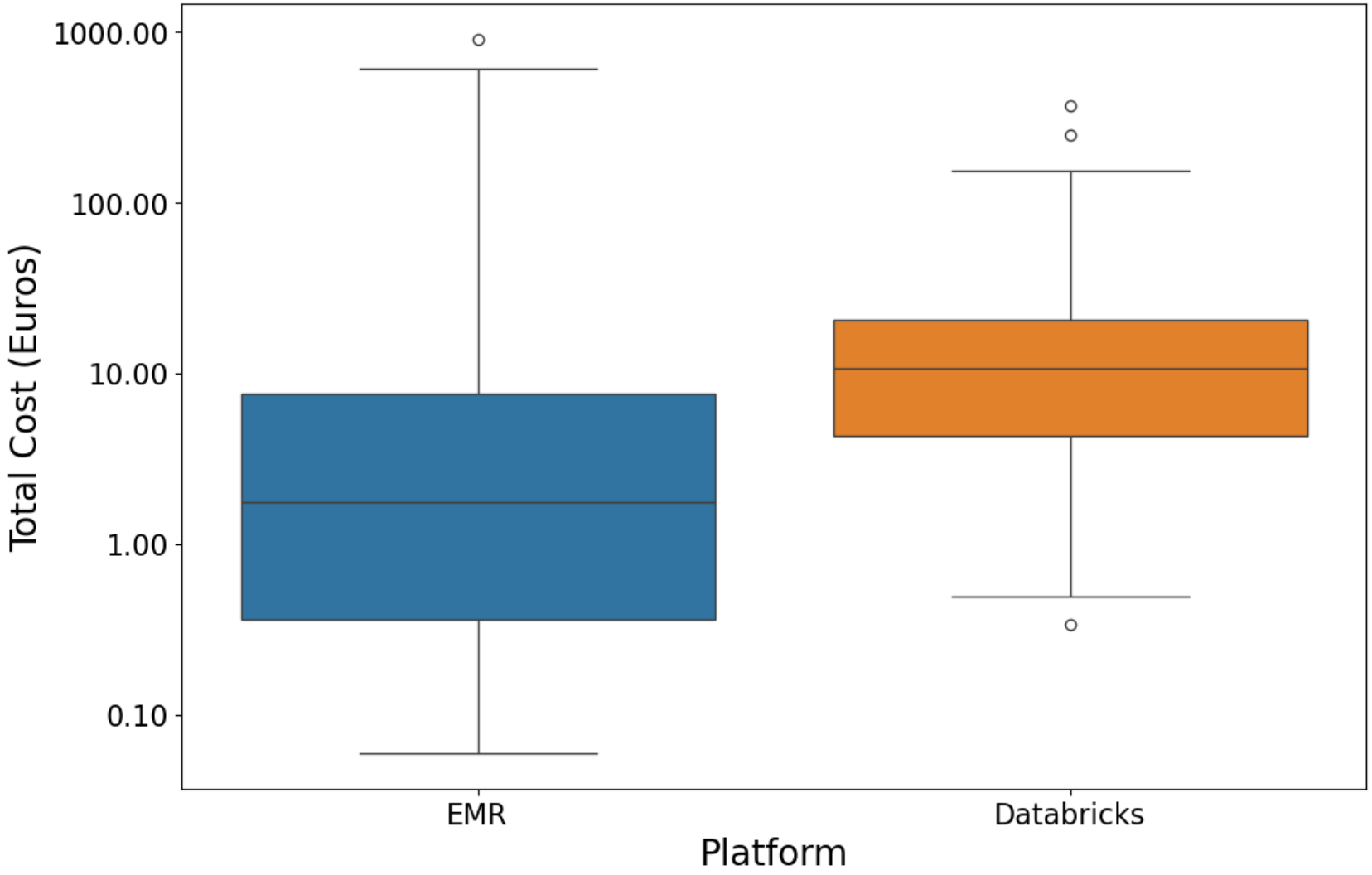
Implementation time of DBR is lower



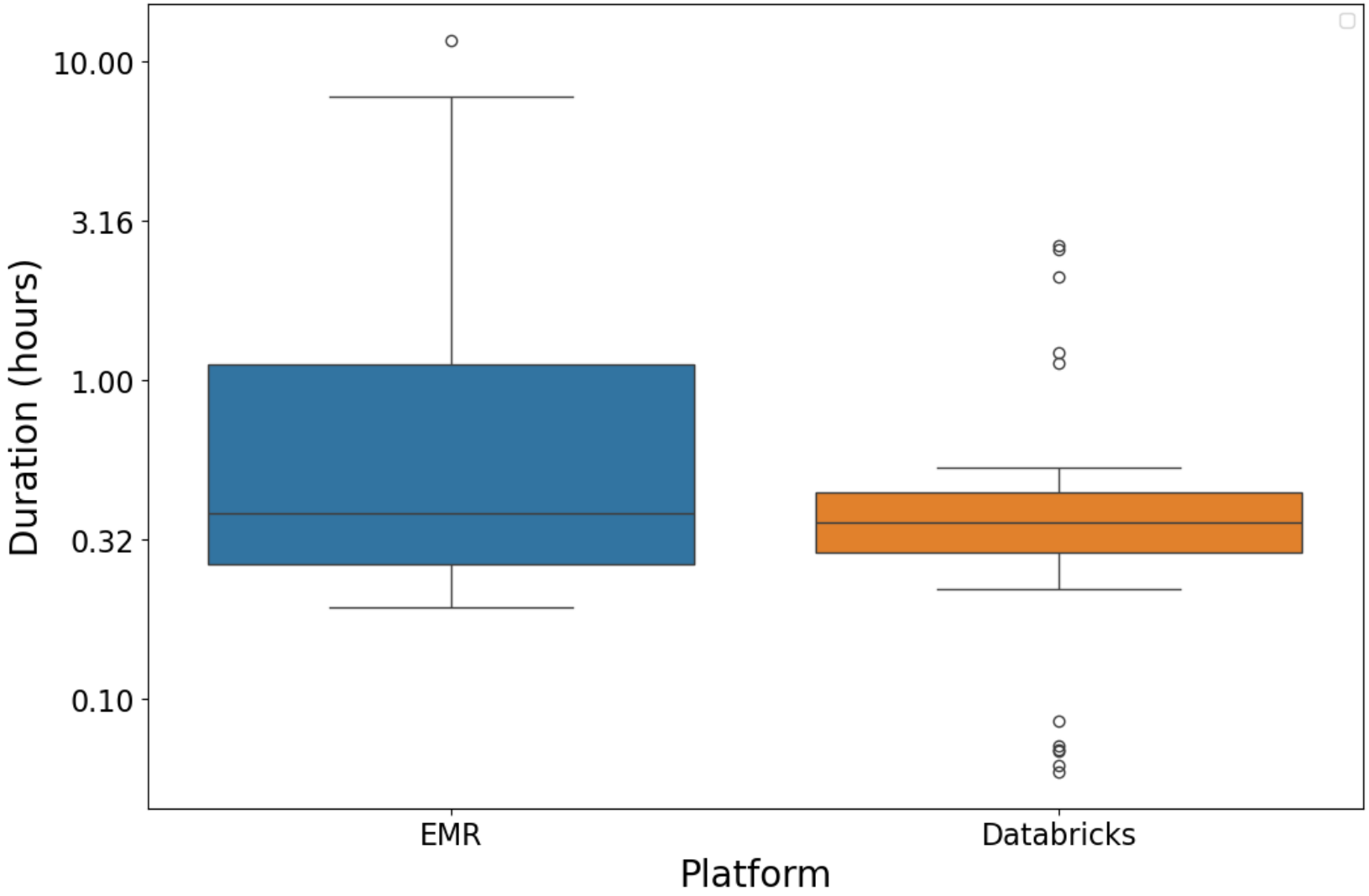
Implementation complexity of DBR is lower more & more frequent commits for EMR integration



Median cost of DBR is higher than EMR



Variability of execution time of DBR is lower



Implementation learnings

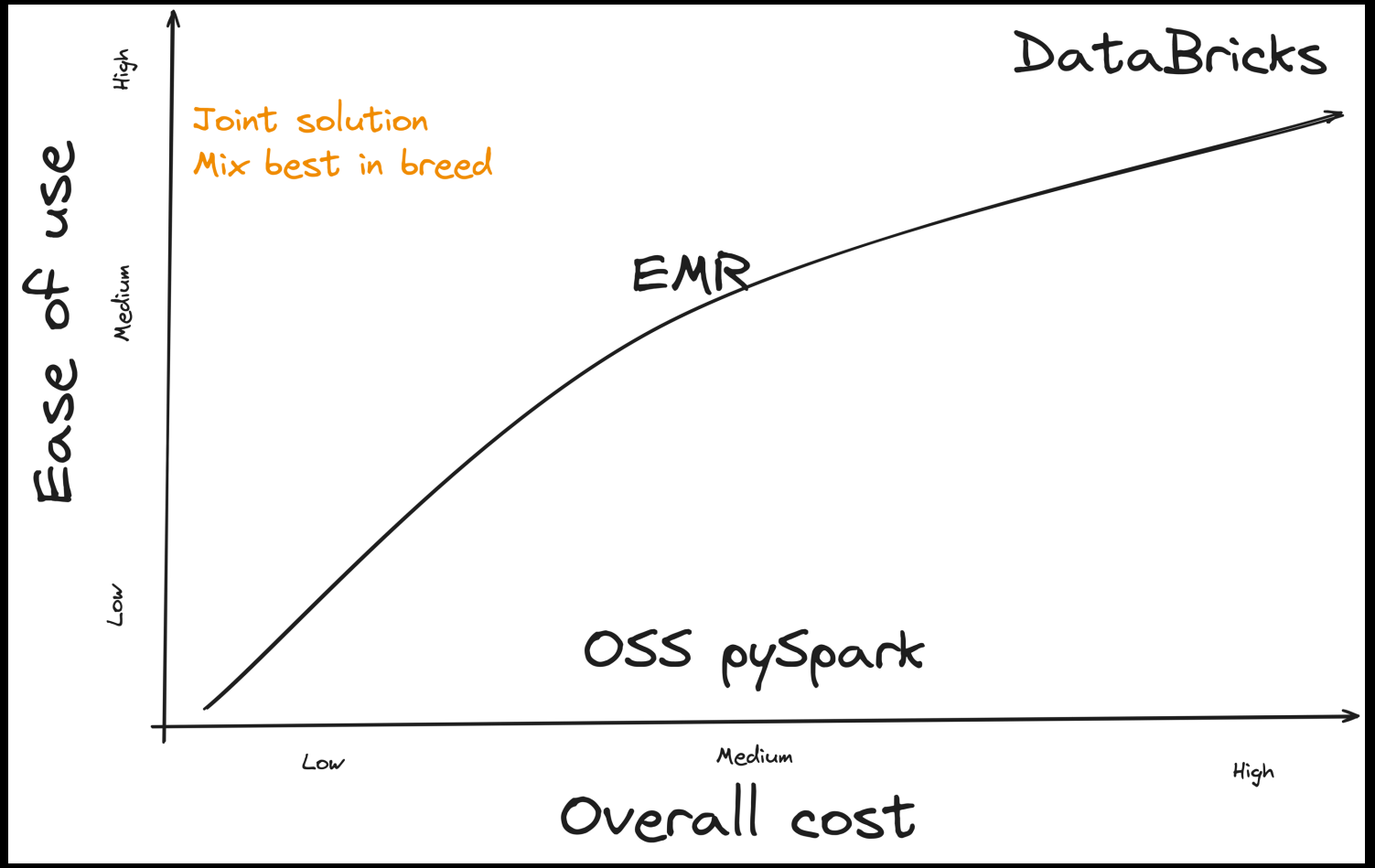
- Complexity of AWS EMR: Many low level details about AWS, spot instances, networking required (master on spot instance => 💥💥)
- Abstracting the PaaS requires deep understanding of their APIs

Tips

- maximizeResourceAllocation
- LZO
- Delta zorder on partition
- spark.databricks.delta.vacuum.parallelDelete.enabled=true

Summary

- Money saved – 43%
- Bring back software engineering best practices for data
- Flexibility
  - Data PaaS as a commodity
  - Take back control
  - Best in breed



COST EFFICIENCY FOR SPARK

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