Data Processing in Databricks

Leveraging Pandas, PySpark, and SQL

Marcelino Mayorga Quesada



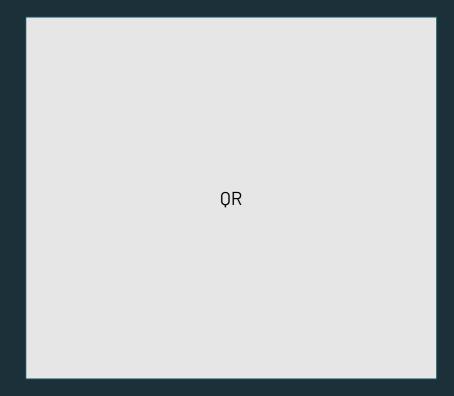


Marcelino Mayorga Quesada

- 14 Years of Experience in Software in finance, marketing and video games sectors, and over the last 3 years in AI.
- Experienced in technical and delivery management roles.
- Clients: Blackstone, Cambridge Associates, EA
 Sports, Citibank and BAC Credomatic.
- Technical Instructor on GCP and .NET courses.
- Passionate about Artificial Intelligence in music
 and video games in.

Agenda

- 1. Introductions
 - Instructor, Topic, Audience
- 2. Data Processing
 - Concept, Operations
- 3. Databricks
 - Solution, Pyspark, Pandas, SQL
- 4. Demo
 - Community Version
 - Use case
- **5.** Key Takeaways
- 6. Q&A

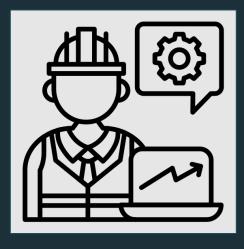




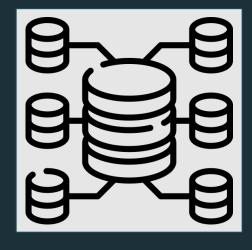
Data Processing

Concept

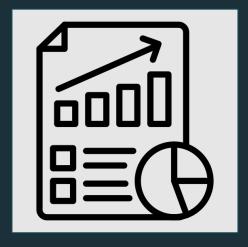
A series of operations to convert <u>raw</u> data into <u>meaningful information</u>.



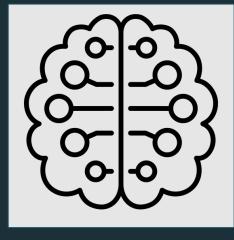
Data Engineers



Storage



Analytics



DL & ML Models



Data Processing

Operations

1. Cleaning

- Removing duplicates
- Impute or delete missing values
- Correct errors and inconsistencies

2. Integration

- ETL(Extract Transform Load)
- Merge and Join data warehousing
- Augmentation

3. Transformation

- Normalization and Standardization
- Aggregation (Summing, Averaging)
- Pivoting tables
- Encoding categorical values

4. Reduction

- Dimensionality Reduction: PCA, t-SNE
- Feature Selection & Extraction
- Sampling
- Compression

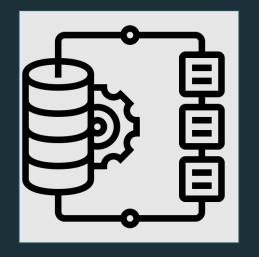


Data Processing

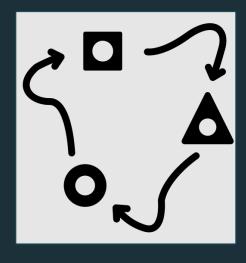
Operations



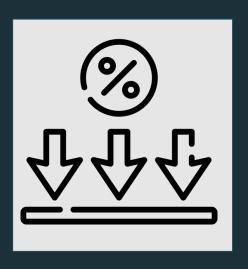
Cleaning



Integration



Transformation



Reduction



Databricks

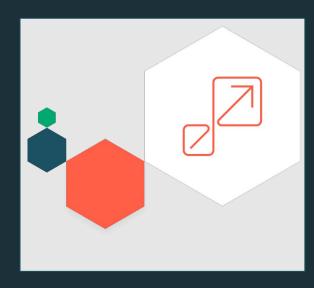
Data Lakehouse Architecture



Unified



Open



Scalable



Pandas

Data Analysis and Manipulation Library

- Data cleaning, transformation and visualization
- Low volume data for single computing
- Prototyping
- Available:
 - Stand alone library
 - Pandas API in Spark
 - Pandas Dataframes





PySpark

Big Data Processing Framework

- Python API for Apache Spark
- High volume data (TB, PB)
- Distributed computing (Clusters)
 - Parallel processing
 - Lazy Evaluation
 - Fault Tolerance
- Complex data transformations
- Used via Spark Session & Context and Spark Dataframes





SQL

Structure Query Language

- Managing and manipulating relational databases
- Queries and Transactions over multiple tables.
- Available:
 - Stand alone DBMS (mysql)
 - SQL in Spark
 - SQLContext





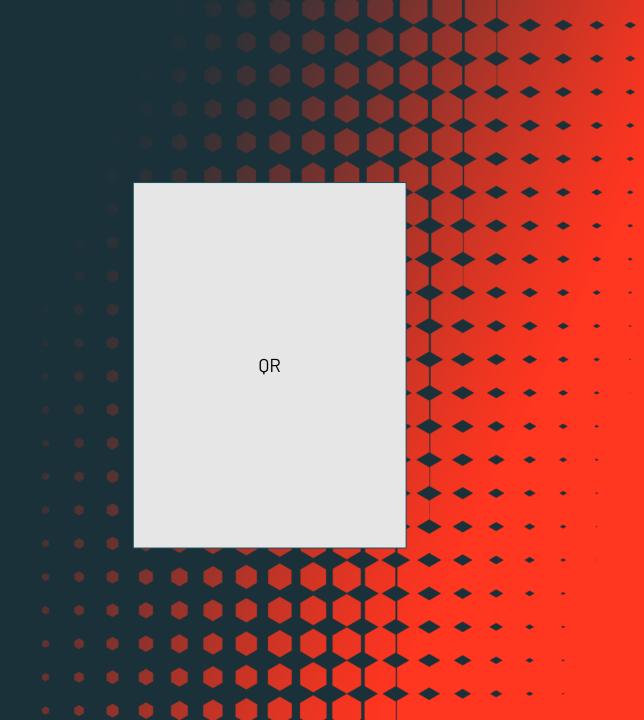
Databricks

Summary

Name	Type	Purpose	Usage via	Ideal
Pandas	Data Analysis and Manipulation Library	Data HandlingTransformationEager Execution	 Standalone Library Pandas API on Spark Dataframes 	Low Volume DataPrototyping
PySpark	Python Interface for Big Data Processing Framework	 Distributed Computing Parallel Processing Lazy Evaluation Fault Tolerance 	PysparkContext & SessionDataframes	 Big Data Scalability Performance Integration with multiple sources
SQL	Structured Query Language	QueryingTransformationsTransactionsStorage	SQLContextDatasets	QueryingManaging relational databases

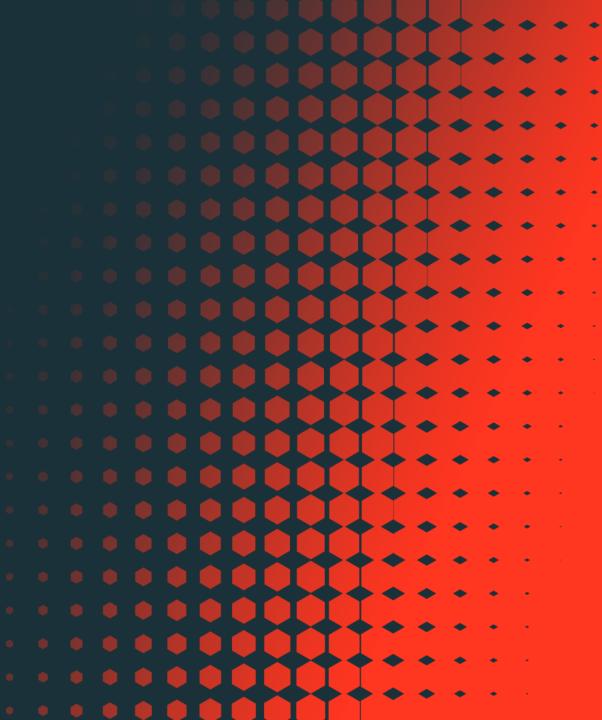


POP QUIZ!





Demo



Databricks

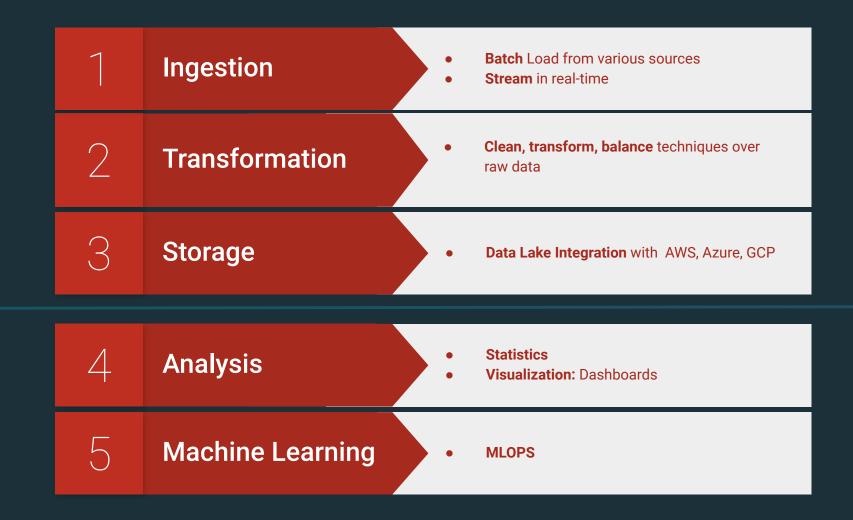
Community Edition

- Free access: https://community.cloud.databricks.com/
- Collaborative Notebook Environment
- Cluster Management
 - 1 Driver
 - 15.3 GB Memory, 2 Cores, 1 DBU
- Experiment Tracking



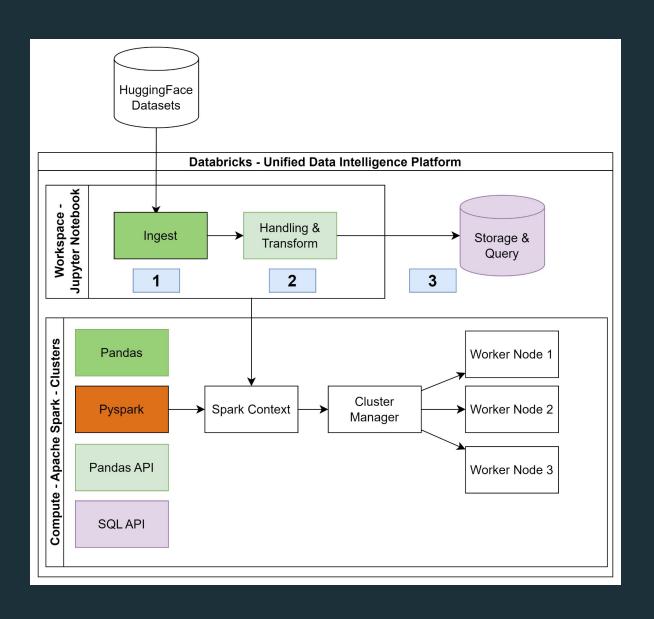
Data Workflow

Stages





Use Case





Key Takeaways

- Data Processing crucial step in Data Engineering
- Databricks Data Intelligence Platform
- Pandas is a powerful for data handling library aimed for single computing.
- Pyspark supports and enhances Pandas and SQL



A&Q



