

14 DAYS

AI CHALLENGE

DAY 01

Topic:

Platform Setup & First Steps

Challenge:

1. Create Databricks Community Edition account
2. Navigate Workspace, Compute, Data Explorer
3. Create first notebook
4. Run basic PySpark commands

Day 1 Reflection

1. Experienced collaborative workflows using shared Databricks notebooks
2. Understood how a unified platform connects data, analytics, and AI
3. Built a strong foundation for scalable, team-driven data projects

Challenge Overview



14 Days AI Challenge

An intensive program designed to learn and apply modern AI technologies directly on the Databricks Lakehouse platform.



Daily Format

Structured for consistency:

1. Short, focused learning goal
2. Hands-on notebook execution



Organized By

A collaborative effort by industry leaders:
Indian Data Club • Databricks • Codebasics



Today's Focus (Day 01)

- ✓ Set up Databricks Community Edition
- ✓ Explore Workspace Interface
- ✓ Run First PySpark Commands

What is Databricks?



Unified Data + AI Platform

Built on the Lakehouse architecture, it unifies your data warehousing and data science workloads into a single platform, eliminating data silos.



Core Components

- ⚡ Compute: Apache Spark (distributed processing)
- ☰ Storage: Delta Lake (reliability & performance)
- ⌚ Governance: Unity Catalog



Integrated Workflows

Seamlessly combines data engineering, real-time analytics, and ML/LLM workflows, allowing teams to collaborate on the same data.



The Outcome

Enables faster pipelines, fully governed data, and production-grade AI solutions, all within a single collaborative environment.

Why use Databricks for AI/ML?



Scalability

Seamlessly handle massive datasets with auto-scaling clusters and serverless compute options. Scale from single-node testing to distributed training effortlessly.



Productivity

A unified workspace bringing data, code, and visualizations together. Features collaborative notebooks, integrated git repos, and built-in dashboards.



Performance

Achieve lightning-fast query execution with the Photon engine and Delta Lake storage optimizations, including caching and data skipping.



Governance

Secure your AI assets with **Unity Catalog**. Manage permissions, audit logs, and data lineage across all workspaces from a central point.

How Databricks Works

Lakehouse Platform

INTERFACE

WORKSPACE & APPLICATIONS

- Notebooks
- SQL Editor
- MLflow

COMPUTE

EXECUTION ENGINE

- Spark / Photon

Serverless & Clusters

STORAGE

DATA LAKE STORAGE

- Delta Lake

Parquet Files + Transaction Log

OVERLAY

CATALOG

Unified Storage Layer

Data resides in your cloud account (S3/ADLS/GCS) in open formats. **Delta Lake** adds ACID transactions and versioning on top of Parquet files.

Decoupled Compute

Compute resources (clusters) spin up on demand to process data, then shut down to save costs. **Photon engine** accelerates queries.

Collaborative Workspace

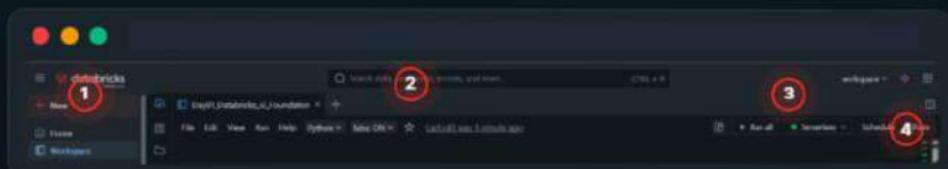
Data Engineers, Scientists, and Analysts work in the same environment using Python, SQL, R, or Scala within interactive notebooks.

Unity Catalog Governance

A single centralized layer to manage permissions, audit logs, and data lineage across all workspaces and clouds.

Explore the Workspace

The Databricks workspace is your unified command center for data engineering, data science, and analytics tasks.



1 Sidebar Navigation

Access core areas: **Workspace** (files), **Catalog** (data), **Workflows**, and **Compute**.

2 Global Search Bar

Quickly find notebooks, tables, dashboards, or documentation across the entire environment.

3 Notebook Controls

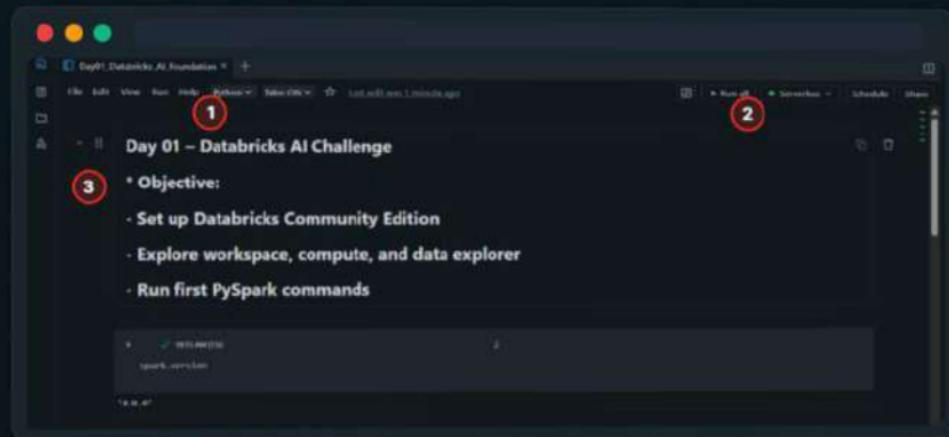
Manage compute attachment, schedule jobs, and share your work with collaborators.

4 Contextual Right Rail

Access comments, version history, variable explorer, and performance metrics.

Tip: Use **Ctrl + P** to open the search palette instantly.

First Steps: Create a Notebook



Before analyzing data, you need to set up your environment. Notebooks are the core workspace for mixing code, visuals, and text.

1 Create a Python Notebook

Navigate to Workspace > Users > Your Folder, right-click and select Create > Notebook. Set the default language to Python.

2 Attach Compute

Use the dropdown in the top-right toolbar to attach your notebook to an active cluster (e.g., Community Edition cluster).

3 Define Objectives

Start with a Markdown cell (`%md`) to document your goals, making your analysis easy to follow.

4 Execute Cells

Run code using `Shift + Enter` or the "Run" button. Keep "Tabs: ON" for easier multitasking.

Tip: Use the "Run All" button to execute the entire notebook from top to bottom when reviewing your work.

PySpark Basics You'll Use Today



SparkSession

The entry point to programming Spark with the Dataset and DataFrame API. In Databricks notebooks, this is automatically available as the variable `spark`.



DataFrame

A distributed collection of data organized into named columns. Think of it like a table in a relational database or a pandas DataFrame, but optimized for distributed processing.



Lazy Evaluation

Spark doesn't compute transformations immediately. Instead, it builds a "logical plan" and waits until an action is called. This optimization is key to performance.



The Common Pattern

- **Create:** Load data or create DataFrame
- **Transform:** `select` , `filter` (Builds Plan)
- ✓ **Action:** `show()` , `count()` (Executes)

Command 1: Check Spark Version

Python 3

Goal

Before running complex transformations, verify that your environment is ready. The `spark` object is automatically available in Databricks notebooks.

```
1 # Check the active Spark version  
spark.version
```

> OUTPUT
'4.0.0'

Day 01 - Databricks AI Challenge

* Objective:

- Set up Databricks Community Edition
- Explore workspace, compute, and data explorer
- Run first PySpark commands

spark.version

Cluster: Active

Note: Seeing a version number confirms that the distributed computing cluster is attached and responsive.

Command 2: Create a DataFrame

Python 3

Goal

Build a small table in memory to understand Spark's structure. We define raw data (rows) and a schema (columns), then convert it into a distributed `DataFrame`.

```
data = [("Ruchi", "Databricks AI Challenge"), ...]
```

OUTPUT PREVIEW

```
+-----+-----+
| Name | Status |
+-----+-----+
| Ruchi | Databricks AI Challe... |
| Day 1 | Completed |
+-----+-----+
```

Note: `df.show()` is an Action that triggers the job execution.

RESULT

```
data = [("Ruchi", "Databricks AI Challenge"), ("Day 1", "Completed")]
columns = ["Name", "Status"]

df = spark.createDataFrame(data, columns)
df.show()
```

The screenshot shows the Databricks interface with a notebook titled "day01_Databricks_AI_Foundation". The code cell contains the creation of a DataFrame from a list of tuples and its display. The resulting DataFrame is shown in the "OUTPUT" section, matching the preview from the Jupyter cell. A status bar at the bottom indicates "Job: Completed (2s)".

Command 3: Inspect Schema

Goal

Spark DataFrames have a defined schema (column names and data types). Using `printSchema()` helps you understand the structure of your data and verify type inference.

```
Cmd 3
1 # View the structure of the DataFrame
df.printSchema()

> OUTPUT
root
|-- Name: string (nullable = true)
|-- Status: string (nullable = true)
```

💡 Insight: Unlike Pandas, Spark is strongly typed. Knowing if a column is nullable is crucial for data quality checks.

The screenshot shows a Databricks notebook interface. On the left, a code cell contains Python code to create a DataFrame and print its schema. On the right, the notebook's sidebar displays the schema information, showing two columns: 'Name' and 'Status'. Below the schema, the 'Inferred' section of the sidebar indicates that the schema was inferred from the data.

```
File Edit View Run Help Python Tab: DB * Last edit was 1 minute ago
Python Run all Serverless Schedules Share
+ Databricks AI Foundation
data = [{"key": "Alice", "Challenge": "Challenge"}, {"key": "Bob", "Challenge": "Completed"}]
df = spark.createDataFrame(data, schema)
df.show()

# Inferred
DF: pyspark.sql.dataframe.DataFrame[Name: string, Status: string]
| Name | Status |
| Alice | Challenge |
| Bob | Completed |
```

Schema: Inferred

Command 4: Select and Show

Python 3

Goal

Demonstrate how to project specific columns using a transformation (`select`) and trigger execution with an action (`show`).

Cmd 4

```
1 # Select "Name" column and show content
df.select("Name").show()
```

OUTPUT

```
+---+
| Name|
+---+
|Ruchi|
|Day 1|
+---+
```

⚡ Insight: Spark lazily builds a plan for `select()` but only executes it when you call `show()`.

The screenshot shows a Databricks workspace interface. At the top right is a Python 3 icon. Below it is a sidebar with a tree view showing a notebook named 'Day01_RuchiKochhar_Foundations' and a 'Run' section. The main area displays a notebook cell with the command: `1 # Select "Name" column and show content
df.select("Name").show()`. Below the cell is the output: `+---+ | Name| +---+ |Ruchi| |Day 1| +---+`. To the right of the notebook is a 'Performance' monitor window. The monitor shows a single query activity: `df.count()`. The table in the monitor includes columns: Statement, Started At, Tasks, Duration, Rows read, Bytes read, and Bytes written. The entry shows: Statement: `df.count()`, Started At: Jan 09, 2020, 10:16 AM, Tasks: 1/1 completed, Duration: 1x 112 ms, Rows read: 0, Bytes read: 0B, Bytes written: 0B. A status message at the bottom right of the monitor says: Execution Time: < 1s.



Command 5: Count Rows

Python 3

Goal

Execute a simple Action to trigger computation. Unlike transformations (which are lazy), `count()` forces Spark to process the data and return a single value to the driver.

Cmd 6

```
1 # Trigger an action to count rows
df.count()
```

↳ OUTPUT

2

⚡ Insight: This is where the actual work happens. The Spark job is launched, tasks are distributed, and the result is sent back.

The screenshot shows the Databricks UI interface. At the top, there's a navigation bar with 'File', 'Edit', 'View', 'Run', 'Help', 'Python...', 'Tab: ON', and a timestamp 'Last 10 days 2 minutes ago'. Below the navigation bar, there's a sidebar with 'Notebook', 'Pandas', and 'Databricks' options. The main area displays a code cell with the command `df.count()`. Below the code cell, there's a table titled 'View of Inquiry History' with columns: Statement, Started At (%), Tasks, Duration, Rows read, Bytes read, and Bytes written. A single row is shown: 'df.count()' with a green status indicator, started at 10:14 AM, duration of 0.4s, and 0 rows and bytes processed. A blue circular icon with a dot is overlaid on the bottom left of the table. A callout box at the bottom right of the table says 'Spark Job: Completed (0.4s)'.

Thank You!!

Presented by: Ruchi Wange