

# 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.



## Organized By

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



## Daily Format

Structured for consistency:

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



## 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



## 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.

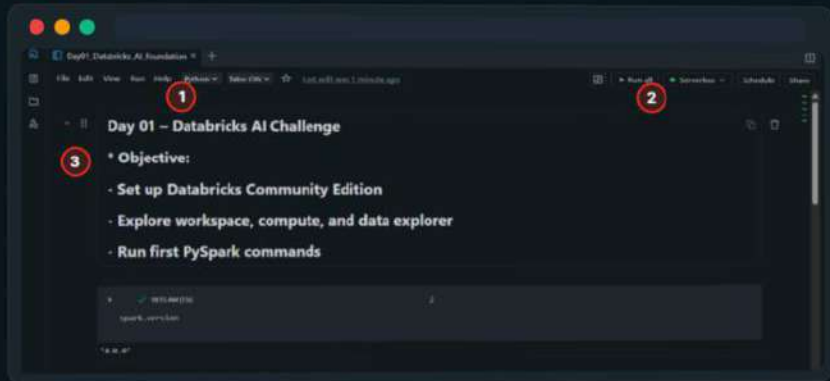
## 4 Contextual Right Rail

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 (Xmd) 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



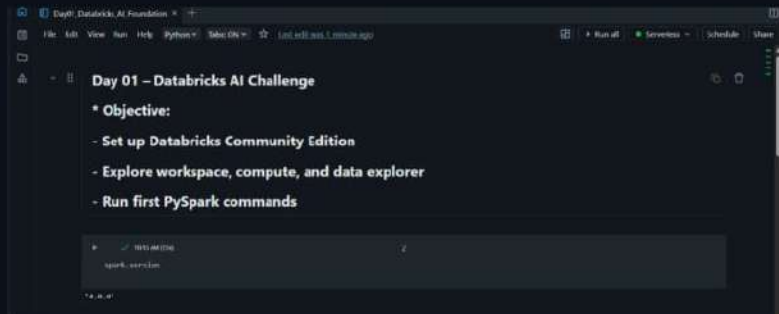
## Goal

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'
```

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



Cluster: Active



# 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**.

```
1 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

The screenshot shows a Databricks notebook with a Python script and its execution results. The script defines a list of data, creates a schema, and creates a DataFrame. It then uses `df.show()` to display the data. The results section shows the performance of the job and the output of the `df.show()` command.

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

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

Job performance (1)

```
dt: jupyterLab.commands.databricks.databricks: (Name: string, Status: string)
```

Name	Status
Ruchi Databricks AI Cha...	Completed

```
df.write.csv('/tmp/df.write.csv')
```

```
df.write.csv('/tmp/df.write.csv')
```

rust

```
1: Name: string (nullable = true)
1: Status: string (nullable = true)
```

Job: Completed (2s)



# Command 3: Inspect Schema

Python 3



## 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.

```
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. The top bar includes a file explorer, a menu (File, Edit, View, Run, Help), and tabs for Python, Scala, and SQL. The main editor area contains the following code:

```
# Create a DataFrame
data = [{"Name": "Databricks AI Challenge", "Day 1": "Completed"}]
columns = ["Name", "Status"]

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

Below the code, there is a section titled "See performance (1)" which shows the execution plan for the DataFrame. The plan indicates that the DataFrame has a schema of (Name: string, Status: string) and contains one row: [Databricks AI Cha... | Day 1 | Completed].

At the bottom, the output of the `df.printSchema()` command is displayed:

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

A blue button labeled "Schema: Inferred" is located at the bottom of the notebook interface.



# 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 notebook interface. The top toolbar includes buttons for 'Run all', 'Serverless', 'Schedule', and 'Share'. The notebook cell contains the following code:

```
df.select("Name").show()
```

Below the code, the 'View performance (1)' tab is active, displaying a table with execution details:

Statement	Started At	Tasks	Duration	Rows read	Bytes read	Bytes written
1: df.select()	Jan 05, 2025, 10:16 AM	1/1 completed	1 x 112 ms	0	0 B	0 B

At the bottom right, a status bar indicates: **Execution Time: < 1s**



# Command 5: Count Rows



## 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.

```
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 IDE interface. At the top, there's a toolbar with 'Run all', 'Serverless', 'Schedule', and 'Share' buttons. Below that, a Python notebook cell is shown with the code `df.count()`. The cell is labeled '2 minutes ago' and '5' lines. Below the code, there's a 'View performance' link. The execution results are displayed below the code, showing a table with columns: Statement, Started At, Tasks, Duration, Rows read, Bytes read, and Bytes written. The table shows one statement, 'df.count()', which started at 'Jan 28, 2026, 10:11 AM', has '1/1 completed' tasks, a duration of '1 x 112 ms', and 0 rows read, 0 bytes read, and 1 B bytes written. Below the table, there's a status bar that says 'Spark Job: Completed (0.4s)'.

Statement	Started At	Tasks	Duration	Rows read	Bytes read	Bytes written
df.count()	Jan 28, 2026, 10:11 AM	1/1 completed	1 x 112 ms	0	0 B	1 B



# Thank You!!

Presented by: Ruchi Wange