

LTIMindtree Data Engineer Interview Guide – Experienced 3+

Round 1 - Technical

1. Project Explanation and Technologies Used

- Describe the projects emphasizing Spark, Hadoop, or Azure for large-scale data processing.
- Discuss tools for ETL, orchestration (Airflow, ADF), and real-time pipelines (Kafka, Spark Streaming).

2. Performance Tuning Techniques

- Optimize shuffles: Use appropriate partitioning.
- Use `persist()/cache()` for reused datasets.
- Leverage broadcast joins for smaller datasets.
- Optimize query plans using `explain()`.

3. Accumulator and Broadcast Variables

- Accumulator: Write-only, for aggregating data across tasks (e.g., counters).
- Broadcast Variables: Read-only, for sharing smaller datasets among executors.

4. Difference Between `SparkSession` and `SparkContext`

- `SparkSession`: Unified entry point for `DataFrame`, SQL, and streaming (introduced in Spark 2.0).
- `SparkContext`: Core entry point for RDD-based operations.

5. Difference Between `Dataset` and `DataFrame`

- `Dataset`: Provides compile-time type safety (Scala/Java).
- `DataFrame`: Schema-based API; simpler for SQL-like operations.

6. Spark Session Command

```
from pyspark.sql import SparkSession
spark = SparkSession.builder.appName("App").getOrCreate()
```

7. Command to Read JSON Data and Options

```
df = spark.read.option("multiline", "true").json("path_to_json")
```

8. CSV Without Column Names/Schema

```
df = spark.read.option("header", "false").csv("path_to_csv")  
df = df.toDF("col1", "col2", "col3") # Assign custom column names
```

9. Find 3rd Highest Salary

```
from pyspark.sql.functions import col, dense_rank  
from pyspark.sql.window import Window  
  
window_spec = Window.orderBy(col("salary").desc())  
df = df.withColumn("rank", dense_rank().over(window_spec))  
third_highest = df.filter(col("rank") == 3)
```

10. Filter Rows Where Employee Salary > Manager Salary

```
emp_df = df.alias("emp")  
mgr_df = df.alias("mgr")  
  
result = emp_df.join(mgr_df, emp_df["mngName"] == mgr_df["Empname"]) \  
    .filter(emp_df["salary"] > mgr_df["salary"]) \  
    .
```

11. Palindrome

```
def is_palindrome(s):  
    return s == s[::-1]  
print(is_palindrome("madam")) # Example
```

12. Spark Submit

```
spark-submit --class <MainClass> --master <ClusterURL> <AppJar>
```

13. Memory Tuning

- Adjust **executor memory** (--executor-memory).
- Use **storage levels** (MEMORY_AND_DISK).
- Tune **GC settings** for large heaps.

14. Created JARs

Describe building custom JARs for Spark jobs using Maven/SBT.

15. Worked with UDFs

Share examples of custom UDFs using Python or Scala.

16. Dynamic Resource Allocation

Automatically adjusts resources based on workload. Enable with:

```
--conf spark.dynamicAllocation.enabled=true
```

17. Daily Data Volume

Quantify daily data (e.g., 1TB/day) and its source (e.g., logs, transactions).

18. Production Experience

Discuss deploying Spark jobs and monitoring them using tools like Airflow or YARN.

Round 2 – Technical

1. Difference Between DataFrame and Dataset

DataFrame: Untyped, optimized for SQL operations.

Dataset: Typed, provides compile-time safety.

2. Load CSV from HDFS

```
df = spark.read.option("header", "true").csv("hdfs://path_to_csv")
```

3. Syntax for CSV Loading

```
df = spark.read.option("header", "true").csv("hdfs://path_to_csv")
```

4. What is Multiline?

Multiline option handles JSON files with nested records across multiple lines.

5. No Column Names in CSV

```
df = spark.read.option("header", "false").csv("path_to_csv")
df = df.toDF("col1", "col2", "col3") # Assign column names
```

6. Case Class and StructType Syntax

```
# Scala Case Class
case class Employee(name: String, salary: Double, dept: String)

# Python StructType
from pyspark.sql.types import StructType, StructField, StringType, DoubleType
schema = StructType([
    StructField("name", StringType(), True),
    StructField("salary", DoubleType(), True),
    StructField("dept", StringType(), True)
])
```

7. Partitioning vs. Bucketing

Partitioning: Divides data into directories based on keys.

Bucketing: Splits data into fixed-size buckets based on hash functions.

8. Closure Function

A function that references variables from its enclosing scope.

9. Count of Alphabets in String

```
from collections import Counter
s = "I am a data engineer"
counts = Counter(c for c in s.lower() if c.isalpha())
print(counts)
```

10. Difference Between List and Tuple

List: Mutable, slower.

Tuple: Immutable, faster.

11. List Comprehension

```
squares = [x**2 for x in range(10)]
```

Glassdoor LTI Mindtree Review –

https://www.glassdoor.co.in/Reviews/LTIMindtree-work-life-balance-Reviews-EI_IE8441464.0,11_KH12,29.htm

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