



# PySpark Scenario-Based Interview Questions (Complete Notes Series)

DAY 11 – Data Skew, Salting & Advanced Join Optimization



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# PySpark Scenario-Based Interview Questions (Complete Notes Series)

## DAY 11 — Data Skew, Salting & Advanced Join Optimization

### Concepts Covered Today

- What is data skew and why it happens
- How to identify skew
- Join skew vs aggregation skew
- Salting technique (interview favourite)
- Broadcast vs AQE vs skew hints

### Scenario

You work for an **Indian UPI / fintech platform**.

Table: transactions

- 2 billion rows

- 40% of transactions belong to **customer\_id = 'PAYTM\_MERCHANT'**

Job: Calculate **daily total amount per merchant**.



## **Problem: What is Data Skew?**

### **◆ Interview Definition**

Data skew happens when **few keys contain a very large portion of data**, causing:

- Some partitions to be extremely large
- Few tasks running very long
- Other executors staying idle



## **Question 1: How to Identify Data Skew?**

### **◆ Scenario**

Spark job is slow and stuck at 90%.

### PySpark Solution

```
transactions_df.groupBy("merchant_id")  
    .count()  
    .orderBy("count", ascending=False)  
    .show()
```

### Explanation

- One key dominating output → skew confirmed
- Interviewers love this simple detection logic

## Question 2: Why Skewed Joins Are Dangerous?

### ◆ Explanation

- Skewed key sent to **single reducer**
- Causes OOM or long-running tasks
- Cluster resources wasted



## Question 3: Salting Technique (MOST ASKED)

### ◆ Scenario

Join transactions with merchants table.



### Step 1: Add Salt Column to Large Table

```
from pyspark.sql.functions import rand, floor

salted_txn_df = transactions_df.withColumn(
    "salt",
    floor(rand() * 10)
)
```



### Step 2: Expand Small Table

```
from pyspark.sql.functions import explode, array

salted_merchants_df = merchants_df.withColumn(
    "salt",
    explode(array([0,1,2,3,4,5,6,7,8,9]))
)
```

)

### **Step 3: Perform Salted Join**

```
final_df = salted_txn_df.join(  
    salted_merchants_df,  
    ["merchant_id", "salt"],  
    "inner"  
)
```

### **Why This Works**

- Skewed key distributed across multiple partitions
- Load balanced across executors

## Question 4: Broadcast vs Salting — When to Use What?

### Broadcast

- One table is small (< few 100 MB)
- Avoids shuffle completely

### Salting

- Both tables are large
- Skewed keys exist



## Question 5: AQE (Adaptive Query Execution)

### Scenario

Spark 3.x environment



### Enable AQE

```
spark.conf.set("spark.sql.adaptive.enabled", "true")
```



## What AQE Does

- Automatically handles skew joins
- Changes join strategy at runtime
- Splits skewed partitions





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