



# PySpark Scenario-Based Interview Questions

DAY 5 — Handling NULLs,  
Missing Data & Data Quality  
Checks



**Karthik Kondpak**  
**9989454737**

# PySpark Scenario-Based Interview

## Questions

### DAY 5 — Handling NULLs, Missing Data & Data Quality Checks

#### Concepts Covered Today

- Identifying NULL values
- Handling NULLs using fillna and dropna
- Conditional NULL handling
- Data quality validation checks
- Real-time pipeline scenarios

#### Sample Data: employee\_df

emp_id	emp_name	department	salary	joining_date
1	Rahul	IT	60000	2022-01-10
2	Priya	NULL	55000	2022-02-15
3	Amit	HR	NULL	2022-03-01
4	Neha	IT	65000	NULL



## Question 1: Identify Records Containing NULL Values

### ◆ Scenario

Before analytics reporting, identify records that contain **any NULL values**.



### PySpark Solution

```
null_records_df = employee_df.filter(
    employee_df.emp_name.isNull() |
    employee_df.department.isNull() |
    employee_df.salary.isNull() |
    employee_df.joining_date.isNull()
)

null_records_df.show()
```



### Explanation

- isNull() checks for missing values
- Common first step in data quality validation

## Question 2: Replace NULL Values with Default Values

### Scenario

Replace missing values based on business rules:

- Department → "Unknown"
- Salary → 0

### PySpark Solution

```
filled_df = employee_df.fillna({
    "department": "Unknown",
    "salary": 0
})

filled_df.show()
```

### Explanation

- `fillna()` is efficient for bulk NULL replacement

- Often used before aggregations

### **Question 3: Drop Records with Critical NULL Values**

#### **Scenario**

If emp\_name or joining\_date is missing, the record should be **rejected**.

#### **PySpark Solution**

```
clean_df = employee_df.dropna(subset=["emp_name",  
"joining_date"])  
  
clean_df.show()
```

#### **Explanation**

- subset ensures only critical columns are validated
- Helps maintain data accuracy



## Question 4: Conditional NULL Handling Using when()

### ◆ Scenario

If salary is NULL, assign default based on department:

- IT → 50000
- HR → 40000



### PySpark Solution

```
from pyspark.sql.functions import when, col

conditional_df = employee_df.withColumn(
    "salary",
    when(col("salary").isNull() & (col("department")
== "IT"), 50000)
    .when(col("salary").isNull() & (col("department")
== "HR"), 40000)
    .otherwise(col("salary"))
)

conditional_df.show()
```



## Explanation

- Business-driven NULL handling
- Frequently discussed in real projects



## Question 5: Validate Data Quality Rules



### Scenario

Apply data quality checks before loading into the warehouse:

- Salary must be  $> 0$
- Joining date must not be NULL



## PySpark Solution

```
invalid_records_df = employee_df.filter(  
    (col("salary") <= 0) |  
    col("joining_date").isNull()  
)  
  
valid_records_df =
```

```
employee_df.subtract(invalid_records_df)

invalid_records_df.show()
valid_records_df.show()
```



### **Explanation**

- Invalid records can be redirected to error tables
- Demonstrates pipeline robustness





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