


## ✓ Pyspark Practice Question

```
from pyspark.sql import SparkSession
```

```
spark = SparkSession.builder \
    .appName("Pyspark_Practice") \
    .master("local[*]") \
    .getOrCreate()
```

```
spark
```

 **SparkSession - in-memory**  
**SparkContext**

[Spark UI](#)

Version

v3.5.1

Master

local[\*]

AppName


Pyspark\_Practice

## ✓ Basic RDD Operation

```
# initialise spark context
sc = spark.sparkContext
```

```
# create an RDD from a list
rdd = sc.parallelize([1,2,3,4,5])
```


```
# perform a simple action to collect the data
rdd.collect()
```

 [1, 2, 3, 4, 5]


```
# count element in rdd
rdd.count()
```

 5


```
# filter even number in rdd
rdd.filter(lambda x:x%2==0).collect()
```

 [2, 4]

```
# find square of rdd element
rdd.map(lambda x:x**2).collect()
```

 [1, 4, 9, 16, 25]

```
# get distinct element in rdd
rdd.distinct().collect()
```

 [2, 4, 1, 3, 5]

```
# reduce the rdd by summing all element
```

```
rdd.reduce(lambda x,y:x+y)
```

 15

```
# union of two rdd
rdd2 = sc.parallelize([5,6,7,8,9])
rdd.union(rdd2).collect()
```

```
↵ [1, 2, 3, 4, 5, 5, 6, 7, 8, 9]
```

```
# intersection of two rdd
rdd.intersection(rdd2).collect()
```

```
↵ [5]
```

```
# group by key
rdd3 = sc.parallelize([("a",1),("b",2),("a",3),("b",4)])
rdd3.groupByKey().mapValues(list).collect()
```

```
↵ [('b', [2, 4]), ('a', [1, 3])]
```

```
# reduce by key to sum all values
rdd3.reduceByKey(lambda a,b:a+b).collect()
```

```
↵ [('b', 6), ('a', 4)]
```

## ▼ ◆ DataFrame Basics

```
# create dataframe from list of tuples
data = [('Alice',1),('Bob',2),('Charlie',3)]
df = spark.createDataFrame(data,['name','age'])
df.show()
```

```
↵ +-----+----+
|  name|age|
+-----+----+
|  Alice|  1|
|   Bob|  2|
|Charlie|  3|
+-----+----+
```

```
# show schema
df.printSchema()
```

```
↵ root
|-- name: string (nullable = true)
|-- age: long (nullable = true)
```

```
# select a column
df.select('Name').show()
```

```
↵ +-----+
|  Name|
+-----+
|  Alice|
|   Bob|
|Charlie|
+-----+
```

```
#Filter DataFrame rows
df.select('Name').filter(df['age']>2).show()
```

```
↵ +-----+
|  Name|
+-----+
|Charlie|
+-----+
```

```
df.filter(df['age']==3).show()
```

```
↵ +-----+----+
|  name|age|
+-----+----+
|Charlie|  3|
+-----+----+
```

```
#Add new column
df.withColumn('double_age',df['age']*2).show()
```

```
↩️ +-----+-----+
|  name|age|double_age|
+-----+-----+
|  Alice| 1|         2|
|   Bob| 2|         4|
|Charlie| 3|         6|
+-----+-----+
```

```
# show dataframe
df.show()
```

```
↩️ +-----+-----+
|  name|age|
+-----+-----+
|  Alice| 1|
|   Bob| 2|
|Charlie| 3|
+-----+-----+
```

```
# show new column with triple age
df = df.withColumn('triple_age',df['age']*3)
df.show()
```

```
↩️ +-----+-----+
|  name|age|triple_age|
+-----+-----+
|  Alice| 1|         3|
|   Bob| 2|         6|
|Charlie| 3|         9|
+-----+-----+
```

```
# drop a column
df.drop('triple_age').show()
```

```
↩️ +-----+-----+
|  name|age|
+-----+-----+
|  Alice| 1|
|   Bob| 2|
|Charlie| 3|
+-----+-----+
```

```
# rename a column
df.withColumnRenamed('age','Age').show()
```

```
↩️ +-----+-----+
|  name|Age|triple_age|
+-----+-----+
|  Alice| 1|         3|
|   Bob| 2|         6|
|Charlie| 3|         9|
+-----+-----+
```

```
# sort by column
df.sort('Age',ascending = False).show()
```

```
↩️ +-----+-----+
|  name|age|triple_age|
+-----+-----+
|Charlie| 3|         9|
|   Bob| 2|         6|
|  Alice| 1|         3|
+-----+-----+
```

```
# create dataframe from list of tuples
data = [('Alice',1),('Bob',2),('Charlie',3),('Charlie',3),('Bob',4)]
df = spark.createDataFrame(data,['name','age'])
df.show()
```

```

↵
+-----+-----+
|  name|age|
+-----+-----+
|  Alice|  1|
|   Bob|  2|
|Charlie|  3|
|Charlie|  3|
|   Bob|  4|
+-----+-----+

```

# Group by and aggregate

```
df.groupBy('Name').agg({'Age':"sum"}).show()
```

```

↵
+-----+-----+
|  Name|sum(Age)|
+-----+-----+
|   Bob|      6|
|  Alice|      1|
|Charlie|      6|
+-----+-----+

```

# get distinct rows  
df.distinct().show()

```

↵
+-----+-----+
|  name|age|
+-----+-----+
|   Bob|  2|
|  Alice|  1|
|Charlie|  3|
|   Bob|  4|
+-----+-----+

```

## SQL with DataFrames

# Register DataFrame as SQL table  
df.createOrReplaceTempView("people")

# Run a sql query in sparkSQL  
spark.sql("Select \* from people where age > 2").show()

```

↵
+-----+-----+
|  name|age|
+-----+-----+
|Charlie|  3|
|Charlie|  3|
|   Bob|  4|
+-----+-----+

```

# count rows using sparkSQL  
spark.sql("select count(1) as row\_count from people").show()

```

↵
+-----+
|row_count|
+-----+
|      5|
+-----+

```

# case statemetns in sparkSQL  
spark.sql("select name, age, case when age > 2 then 'old' else 'young' end as age\_group from people").show()

```

↵
+-----+-----+-----+
|  name|age|age_group|
+-----+-----+-----+
|  Alice|  1|  young|
|   Bob|  2|  young|
|Charlie|  3|   old|
|Charlie|  3|   old|
|   Bob|  4|   old|
+-----+-----+-----+

```

```
# join two dataframes
df2 = spark.createDataFrame([('Alice',1),('Bob',2)], ['name', 'new_age'])
df2.show()
```

```

+-----+-----+
| name|new_age|
+-----+-----+
| Alice|      1|
|  Bob|      2|
+-----+-----+
```

```
# join dataframes on one column
df = df.join(df2, on='name', how='inner')
df.show()
```

```

+-----+-----+
| name|age|new_age|
+-----+-----+
| Alice|  1|      1|
|  Bob|  2|      2|
|  Bob|  4|      2|
+-----+-----+
```

## Advanced Transformations

```
# Explode array column(transpose)
from pyspark.sql.functions import explode, col
df1 = spark.createDataFrame([[1,2,3],), ["numbers"])
```

```
df1.select(explode(col("numbers"))).show()
```

```

+---+
| col|
+---+
|  1|
|  2|
|  3|
+---+
```

Note : If your df2 DataFrame has a column with arrays (e.g., column "letters"), you should do:

```
data = [("Alice", ["a", "b", "c"]), ("Bob", ["x", "y"])]
df2 = spark.createDataFrame(data, ["name", "letters"])
```

```
# Explode array column
df2.select("name", explode("letters").alias("letter")).show()
```

```

+-----+-----+
| name|letter|
+-----+-----+
| Alice|    a|
| Alice|    b|
| Alice|    c|
|  Bob|    x|
|  Bob|    y|
+-----+-----+
```

```
# pivot dataframe
df.groupBy('name').pivot("age").sum("age").show()
```

```

+-----+-----+
| name|  1|  2|  4|
+-----+-----+
| Alice|  1|NULL|NULL|
|  Bob|NULL|  2|  4|
+-----+-----+
```

```
# pivot with aggregation
df.groupBy('name').pivot('age').sum('new_age').show()
```

```
↵ +-----+-----+-----+
  | name|    1|    2|    4|
  +-----+-----+-----+
  | Alice|    1|NULL|NULL|
  |  Bob|NULL|    2|    2|
  +-----+-----+-----+
```

```
# window function
from pyspark.sql.window import Window
from pyspark.sql.functions import row_number
```

```
# Sample data
```

```
data = [
    ("A", 90),
    ("A", 85),
    ("A", 92),
    ("B", 70),
    ("B", 88),
    ("B", 75)
]
```

```
# Create DataFrame
```

```
df = spark.createDataFrame(data, ["group", "score"])
```

```
# apply window function to assign row numbers within each group based on score
windowSpec = Window.partitionBy("group").orderBy(df['score'].desc())
df.withColumn("row_num", row_number().over(windowSpec)).show()
```

```
↵ +-----+-----+-----+
  |group|score|row_num|
  +-----+-----+-----+
  |  A|   92|      1|
  |  A|   90|      2|
  |  A|   85|      3|
  |  B|   88|      1|
  |  B|   75|      2|
  |  B|   70|      3|
  +-----+-----+-----+
```

```
# Aggregate with multiple functions
```

```
from pyspark.sql.functions import min, max
df.agg(min("score").alias('min_Score'), max("score").alias('max_Score')).show()
```

```
↵ +-----+-----+
  |min_Score|max_Score|
  +-----+-----+
  |       70|       92|
  +-----+-----+
```

```
# Add literal column
```

```
from pyspark.sql.functions import lit
df.withColumn("country", lit("USA")).show()
```

```
↵ +-----+-----+-----+
  |group|score|country|
  +-----+-----+-----+
  |  A|   90|     USA|
  |  A|   85|     USA|
  |  A|   92|     USA|
  |  B|   70|     USA|
  |  B|   88|     USA|
  |  B|   75|     USA|
  +-----+-----+-----+
```

## ▼ ◆ Performance and Optimization

```
# Cache a DataFrame
df.cache()
```

```
↳ DataFrame[group: string, score: bigint]
```

```
# Persist with different storage level
from pyspark import StorageLevel
df.persist(StorageLevel.MEMORY_AND_DISK)
```

```
↳ DataFrame[group: string, score: bigint]
```

```
df2 = df.withColumn('group',lit('A'))
```

```
# Broadcast join
from pyspark.sql.functions import broadcast
df.join(broadcast(df2), "group").show()
```

```
↳
```

group	score	score
A	90	90
A	90	85
A	90	92
A	90	70
A	90	88
A	90	75
A	85	90
A	85	85
A	85	92
A	85	70
A	85	88
A	85	75
A	92	90
A	92	85
A	92	92
A	92	70
A	92	88
A	92	75

```
# Repartition DataFrame
df.repartition(5)
```

```
↳ DataFrame[group: string, score: bigint]
```

```
# Coalesce partitions
df.coalesce(1)
```

```
↳ DataFrame[group: string, score: bigint]
```

## File I/O

```
# read a csv file
uber_df = spark.read.csv("/content/uber_data.csv", header=True, inferSchema=True)
uber_df.show(5)
```

```
↳
```

VendorID	tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	trip_distance	pickup_longitude	pickup_latitude	RatecodeID	stor
1	2016-03-01 00:00:00	2016-03-01 00:07:55	1	2.5	-73.97674560546875	40.765151977539055	1	
1	2016-03-01 00:00:00	2016-03-01 00:11:06	1	2.9	-73.98348236083984	40.767925262451165	1	
2	2016-03-01 00:00:00	2016-03-01 00:31:06	2	19.98	-73.78202056884764	40.64480972290039	1	
2	2016-03-01 00:00:00	2016-03-01 00:00:00	3	10.78	-73.86341857910156	40.769813537597656	1	
2	2016-03-01 00:00:00	2016-03-01 00:00:00	5	30.43	-73.97174072265625	40.79218292236328	3	

only showing top 5 rows

```
# sample json store in file sample_test_json
sample_json = [{"name": "Alice", "age": 30, "city": "New York"},
{"name": "Bob", "age": 25, "city": "San Francisco"}]
```

```
{ "name": "Charlie", "age": 35, "city": "Chicago" }
```

```
# read from a json
sample_df = spark.read.json('/content/sample_test_json.json')
sample_df.show()
```

```
↗ +---+-----+-----+
|age|      city|   name|
+---+-----+-----+
| 30|   New York|  Alice|
| 25|San Francisco|   Bob|
| 35|    Chicago|Charlie|
+---+-----+-----+
```

```
# Write DataFrame to Parquet
uber_df.write.mode('overwrite').parquet("/content/uber_output/")
```

```
# Write DataFrame as CSV
df.write.option("header", True).mode('overwrite').csv("/content/output_csv/")
```

```
# Read Parquet
parquet_df = spark.read.parquet("/content/uber_output/")
parquet_df.show(5)
```

```
↗ +-----+-----+-----+-----+-----+-----+-----+-----+
|VendorID|tpep_pickup_datetime|tpep_dropoff_datetime|passenger_count|trip_distance| pickup_longitude| pickup_latitude|RatecodeID|stor
+-----+-----+-----+-----+-----+-----+-----+-----+
|      1| 2016-03-01 00:00:00| 2016-03-01 00:07:55|           1|         2.5|-73.97674560546875| 40.765151977539055|      1|
|      1| 2016-03-01 00:00:00| 2016-03-01 00:11:06|           1|         2.9|-73.98348236083984| 40.767925262451165|      1|
|      2| 2016-03-01 00:00:00| 2016-03-01 00:31:06|           2|        19.98|-73.78202056884764| 40.64480972290039|      1|
|      2| 2016-03-01 00:00:00| 2016-03-01 00:00:00|           3|         10.78|-73.86341857910156| 40.769813537597656|      1|
|      2| 2016-03-01 00:00:00| 2016-03-01 00:00:00|           5|         30.43|-73.97174072265625| 40.79218292236328|      3|
+-----+-----+-----+-----+-----+-----+-----+-----+
```

only showing top 5 rows

## ✓ ◆ Date and Time Handling

```
from pyspark.sql.functions import current_date
```

```
# current date
df = df.withColumn("today", current_date())
```

```
df.show()
```

```
↗ +-----+-----+-----+
|group|score|  today|
+-----+-----+-----+
|   A|   90|2025-07-29|
|   A|   85|2025-07-29|
|   A|   92|2025-07-29|
|   B|   70|2025-07-29|
|   B|   88|2025-07-29|
|   B|   75|2025-07-29|
+-----+-----+-----+
```

```
# Extract year/month
from pyspark.sql.functions import year, month
```

```
df = df.withColumn("year", year("today"))
df.show()
```

```
↗ +-----+-----+-----+-----+
|group|score|  today|year|
+-----+-----+-----+-----+
|   A|   90|2025-07-29|2025|
|   A|   85|2025-07-29|2025|
|   A|   92|2025-07-29|2025|
|   B|   70|2025-07-29|2025|
|   B|   88|2025-07-29|2025|
+-----+-----+-----+-----+
```



```
|    B|    75|2025-07-29|2025|
+-----+-----+-----+-----+
```

```
# extract month
df = df.withColumn("month", month("today"))
df.show()
```

```
↗
+-----+-----+-----+-----+
|group|score|    today|year|month|
+-----+-----+-----+-----+
|    A|   90|2025-07-29|2025|    7|
|    A|   85|2025-07-29|2025|    7|
|    A|   92|2025-07-29|2025|    7|
|    B|   70|2025-07-29|2025|    7|
|    B|   88|2025-07-29|2025|    7|
|    B|   75|2025-07-29|2025|    7|
+-----+-----+-----+-----+
```

```
# Add a constant column
df = df.withColumn("start", lit('2025-07-20'))
df.show()
```

```
↗
+-----+-----+-----+-----+-----+
|group|score|    today|year|month|    start|
+-----+-----+-----+-----+-----+
|    A|   90|2025-07-29|2025|    7|2025-07-20|
|    A|   85|2025-07-29|2025|    7|2025-07-20|
|    A|   92|2025-07-29|2025|    7|2025-07-20|
|    B|   70|2025-07-29|2025|    7|2025-07-20|
|    B|   88|2025-07-29|2025|    7|2025-07-20|
|    B|   75|2025-07-29|2025|    7|2025-07-20|
+-----+-----+-----+-----+-----+
```

```
# Date difference
from pyspark.sql.functions import datediff
df = df.withColumn("days_diff", datediff("today", "start"))
df.show()
```

```
↗
+-----+-----+-----+-----+-----+-----+
|group|score|    today|year|month|    start|days_diff|
+-----+-----+-----+-----+-----+-----+
|    A|   90|2025-07-29|2025|    7|2025-07-20|        9|
|    A|   85|2025-07-29|2025|    7|2025-07-20|        9|
|    A|   92|2025-07-29|2025|    7|2025-07-20|        9|
|    B|   70|2025-07-29|2025|    7|2025-07-20|        9|
|    B|   88|2025-07-29|2025|    7|2025-07-20|        9|
|    B|   75|2025-07-29|2025|    7|2025-07-20|        9|
+-----+-----+-----+-----+-----+-----+
```

```
# Add days to date
from pyspark.sql.functions import date_add
df = df.withColumn("next_day", date_add("today", 1))
df.show()
```

```
↗
+-----+-----+-----+-----+-----+-----+-----+
|group|score|    today|year|month|    start|days_diff| next_day|
+-----+-----+-----+-----+-----+-----+-----+
|    A|   90|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
|    A|   85|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
|    A|   92|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
|    B|   70|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
|    B|   88|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
|    B|   75|2025-07-29|2025|    7|2025-07-20|        9|2025-07-30|
+-----+-----+-----+-----+-----+-----+-----+
```

```
# print schema
df.printSchema()
```

```
↗ root
 |-- group: string (nullable = true)
 |-- score: long (nullable = true)
 |-- today: date (nullable = false)
 |-- year: integer (nullable = false)
```

```
-- month: integer (nullable = false)
-- start: string (nullable = false)
-- days_diff: integer (nullable = true)
-- next_day: date (nullable = false)
```

```
# Convert string date to date format
from pyspark.sql.functions import to_date
df = df.withColumn('start_date', to_date("start", "yyyy-MM-dd"))
df.show()
```

```
↗ +---+-----+
|group|score|  today|year|month|  start|days_diff|  next_day|start_date|
+---+-----+
|  A|  90|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
|  A|  85|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
|  A|  92|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
|  B|  70|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
|  B|  88|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
|  B|  75|2025-07-29|2025|  7|2025-07-20|  9|2025-07-30|2025-07-20|
+---+-----+
```

```
# print schema to confirm date format changes
df.printSchema()
```

```
↗ root
 |-- group: string (nullable = true)
 |-- score: long (nullable = true)
 |-- today: date (nullable = false)
 |-- year: integer (nullable = false)
 |-- month: integer (nullable = false)
 |-- start: string (nullable = false)
 |-- days_diff: integer (nullable = true)
 |-- next_day: date (nullable = false)
 |-- start_date: date (nullable = true)
```

## ▼ ♦ Miscellaneous

```
# check spark version
spark.version
```

```
↗ '3.5.1'
```

```
# Convert RDD to DataFrame
rdd = sc.parallelize([(1, 'ALICE'), (2, 'BOB'), (3, None)])
df_new = rdd.toDF(["id", "name"])
df_new.show()
```

```
↗ +---+-----+
| id| name|
+---+-----+
|  1|ALICE|
|  2|  BOB|
|  3| NULL|
+---+-----+
```

```
# print first row of DataFrame
df_new.first()
```

```
↗ Row(id=1, name='ALICE')
```

```
# check for null values in a column
from pyspark.sql.functions import col, isnull
df_new.filter(isnull(col("name"))).count()
```

```
↗ 1
```

```
spark.stop()
```