

CODING CHALLENGE 4 (PySpark)

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```
✓ 0s [17] import pyspark
      from pyspark import SparkContext
      from pyspark.sql import SparkSession

      sc = SparkContext.getOrCreate()
      spark = SparkSession.builder.appName("Actions and Transformations in Pyspark").getOrCreate()

      sample_data = sc.parallelize([1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20])
```

Actions:

1. Collect

```
✓ 0s [18] # Collect Action
      print(sample_data.collect())

⇒ [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20]
```

2. First

```
✓ 1s [19] # First Action
      print(sample_data.first())

⇒ 1
```

3. Count

```
✓ 0s [20] # Count Action
      print("Count of values in the given dataframe:",sample_data.count())

⇒ Count of values in the given dataframe: 20
```

4. Take

```
✓ 0s [22] # Take Action
      print(sample_data.take(10))

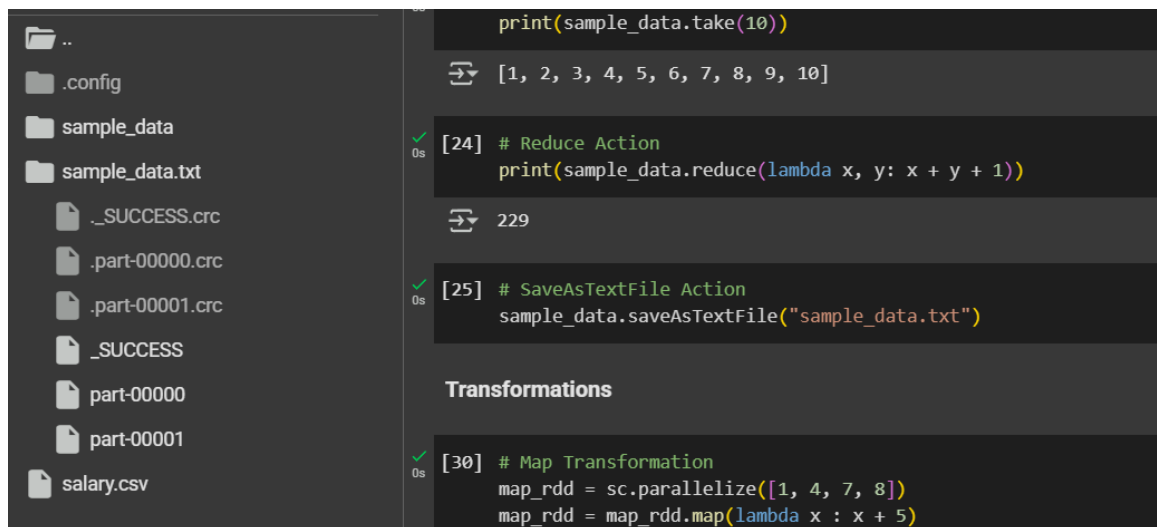
⇒ [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
```

5. Reduce

```
✓ 0s [24] # Reduce Action
      print(sample_data.reduce(lambda x, y: x + y + 1))

⇒ 229
```

6. Save As Text File

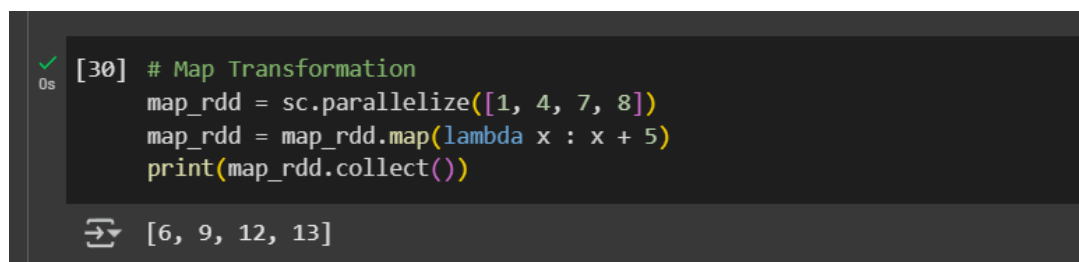


The screenshot shows a file explorer on the left with a directory structure including `..`, `.config`, `sample_data`, `sample_data.txt`, `._SUCCESS.crc`, `.part-00000.crc`, `.part-00001.crc`, `._SUCCESS`, `part-00000`, `part-00001`, and `salary.csv`. The code editor on the right displays the following Spark code:

```
print(sample_data.take(10))  
[1, 2, 3, 4, 5, 6, 7, 8, 9, 10]  
[24] # Reduce Action  
print(sample_data.reduce(lambda x, y: x + y + 1))  
229  
[25] # SaveAsTextFile Action  
sample_data.saveAsTextFile("sample_data.txt")  
  
Transformations  
[30] # Map Transformation  
map_rdd = sc.parallelize([1, 4, 7, 8])  
map_rdd = map_rdd.map(lambda x : x + 5)
```

Transformations


1. Map



The screenshot shows a code editor with the following Spark code:

```
[30] # Map Transformation  
map_rdd = sc.parallelize([1, 4, 7, 8])  
map_rdd = map_rdd.map(lambda x : x + 5)  
print(map_rdd.collect())  
[6, 9, 12, 13]
```

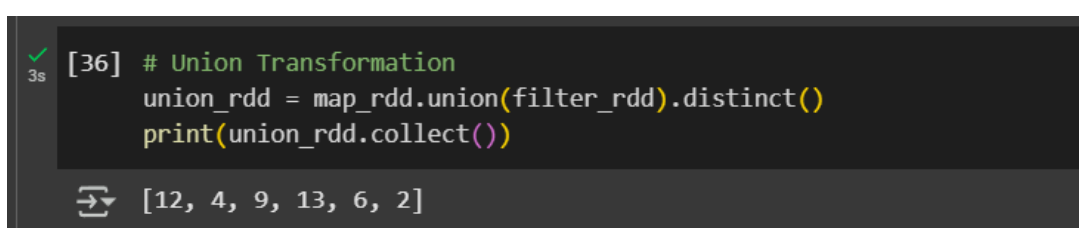
2. Filter



The screenshot shows a code editor with the following Spark code:

```
[31] # Filter Transformation  
filter_rdd = sc.parallelize([1,2,3,4,5,6])  
filter_rdd = filter_rdd.filter(lambda x: x% 2 == 0)  
print(filter_rdd.collect())  
[2, 4, 6]
```

3. Union



The screenshot shows a code editor with the following Spark code:

```
[36] # Union Transformation  
union_rdd = map_rdd.union(filter_rdd).distinct()  
print(union_rdd.collect())  
[12, 4, 9, 13, 6, 2]
```

4. FlatMap

```
[34] # Flat Map Transformation
flatmap_rdd = sc.parallelize(["Hi! I am Swathi", "I am currently undergoing Data Engineering training"])
flatmap_rdd = flatmap_rdd.flatMap(lambda x: x.split(" "))
print(flatmap_rdd.collect())
```

['Hi!', 'I', 'am', 'Swathi', 'I', 'am', 'currently', 'undergoing', 'Data', 'Engineering', 'training']

Joins

Joins

```
[73] # Joining of 2 dataframes
emp = [(1,"Smith",-1,"2018","10","M",3000),
       (2, "Rose",1, "2010", "20","M", 4000),
       (3,"Williams",1,"2010","10","M",1000),
       (4, "Jones",2, "2005","10","F",2000),
       (5,"Brown",2,"2010","40","", -1),
       (6, "Brown", 2, "2010","50","", -1)]
empColumns = ["emp_id","name","superior_emp_id","year_joined", "emp_dept_id","gender","salary"]

dept = [("Finance",10),("Marketing",20),("Sales",30),("IT",40)]
deptColumns = ["dept_name","dept_id"]

empDF = spark.createDataFrame(data = emp, schema = empColumns)
deptDF = spark.createDataFrame(data = dept, schema = deptColumns)
```

1. Left Join

```
[53] # Joins - Left
empDF.join(deptDF, empDF.emp_dept_id == deptDF.dept_id, "left").show()
```

emp_id	name	superior_emp_id	year_joined	emp_dept_id	gender	salary	dept_name	dept_id
1	Smith	-1	2018	10	M	3000	Finance	10
3	Williams	1	2010	10	M	1000	Finance	10
2	Rose	1	2010	20	M	4000	Marketing	20
6	Brown	2	2010	50		-1	NULL	NULL
4	Jones	2	2005	10	F	2000	Finance	10
5	Brown	2	2010	40		-1	IT	40

2. Right Join

```
[55] # Joins - Right
empDF.join(deptDF, empDF.emp_dept_id == deptDF.dept_id, "right").show()
```

emp_id	name	superior_emp_id	year_joined	emp_dept_id	gender	salary	dept_name	dept_id
4	Jones	2	2005	10	F	2000	Finance	10
3	Williams	1	2010	10	M	1000	Finance	10
1	Smith	-1	2018	10	M	3000	Finance	10
2	Rose	1	2010	20	M	4000	Marketing	20
NULL	NULL	NULL	NULL	NULL	NULL	NULL	Sales	30
5	Brown	2	2010	40		-1	IT	40

3. Full Outer Join

```
[56] # Joins - Full Outer
empDF.join(deptDF, empDF.emp_dept_id == deptDF.dept_id, "fullouter").show()
```

emp_id	name	superior_emp_id	year_joined	emp_dept_id	gender	salary	dept_name	dept_id
1	Smith	-1	2018	10	M	3000	Finance	10
3	Williams	1	2010	10	M	1000	Finance	10
4	Jones	2	2005	10	F	2000	Finance	10
2	Rose	1	2010	20	M	4000	Marketing	20
NULL	NULL	NULL	NULL	NULL	NULL	NULL	Sales	30
5	Brown	2	2010	40		-1	IT	40
6	Brown	2	2010	50		-1	NULL	NULL

4. Inner Join

```
[57] # Joins - Inner
empDF.join(deptDF, empDF.emp_dept_id == deptDF.dept_id, "inner").show()
```

emp_id	name	superior_emp_id	year_joined	emp_dept_id	gender	salary	dept_name	dept_id
1	Smith	-1	2018	10	M	3000	Finance	10
3	Williams	1	2010	10	M	1000	Finance	10
4	Jones	2	2005	10	F	2000	Finance	10
2	Rose	1	2010	20	M	4000	Marketing	20
5	Brown	2	2010	40		-1	IT	40

Aggregations

1. Count

```
Aggregations

[48] # Simple Aggregations - Count
from pyspark.sql.functions import count
empDF.select(count("emp_id").alias("Count of Records")).show()
```

Count of Records
6

2. Sum

```
✓ [49] # Simple Aggregations - Sum
0s from pyspark.sql.functions import sum
empDF.select(sum("Salary").alias("Total Salary Distributed")).show()
```

Total Salary Distributed
9998

3. Average

```
✓ [50] # Simple Aggregations - Avg
0s from pyspark.sql.functions import avg
empDF.select(avg("Salary").alias("Average Salary")).show()
```

Average Salary
1666.3333333333333

4. Maximum

```
✓ [52] # Simple Aggregations - Max
0s from pyspark.sql.functions import max
empDF.select(max("Salary").alias("Maximum Salary")).show()
```

Maximum Salary
4000

5. Minimum

```
✓ [51] # Simple Aggregations - Min
0s from pyspark.sql.functions import min
empDF.select(min("Salary").alias("Minimum Salary")).show()
```

Minimum Salary
-1

Group By Functions

```
Group By Functions

[65] emp_dept_DF = empDF.join(deptDF, empDF.emp_dept_id == deptDF.dept_id, "fullouter")
emp_dept_DF.show()
```

emp_id	name	superior_emp_id	year_joined	emp_dept_id	gender	salary	dept_name	dept_id
1	Smith	-1	2018	10	M	3000	Finance	10
3	Williams	1	2010	10	M	1000	Finance	10
4	Jones	2	2005	10	F	2000	Finance	10
2	Rose	1	2010	20	M	4000	Marketing	20
NULL	NULL	NULL	NULL	NULL	NULL	NULL	Sales	30
5	Brown	2	2010	40		-1	IT	40
6	Brown	2	2010	50		-1	NULL	NULL

1. Sum

```
[69] # GroupBy Functions - Sum
from pyspark.sql.functions import col
emp_dept_DF.filter(col("dept_name").isNotNull() & col("salary").isNotNull()).groupBy("dept_name").agg(sum("salary").alias("SUM(Salary)")).show()
```

dept_name	SUM(Salary)
Finance	6000
Marketing	4000
IT	-1

2. Average

```
[72] # GroupBy Functions - Avg
from pyspark.sql.functions import col
emp_dept_DF.filter(col("dept_name").isNotNull() & col("salary").isNotNull()).groupBy("dept_name").agg(avg("salary").alias("AVG(Salary)")).show()
```

dept_name	AVG(Salary)
Finance	2000.0
Marketing	4000.0
IT	-1.0

3. Maximum

```
[70] # GroupBy Functions - Max
from pyspark.sql.functions import col
emp_dept_DF.filter(col("dept_name").isNotNull() & col("salary").isNotNull()).groupBy("dept_name").agg(max("salary").alias("MAX(Salary)")).show()
```

dept_name	MAX(Salary)
Finance	3000
Marketing	4000
IT	-1

4. Minimum

```
[71] # GroupBy Functions - Min
from pyspark.sql.functions import col
emp_dept_DF.filter(col("dept_name").isNotNull() & col("salary").isNotNull()).groupBy("dept_name").agg(min("salary").alias("MIN(Salary)")).show()
```

dept_name	MIN(Salary)
Finance	1000
Marketing	4000
IT	-1

Window Functions

Window Functions

```
[97] data = [("James", "Potter", "Finance", 90000),
            ("Harry", "Thomas", "Finance", 70000),
            ("Henry", "Finch", "Finance", 70000),
            ("Lily", "Evans", "Management", 80000),
            ("Ambika", "Prathap", "Education", 65000),
            ("Preethi", "Sivam", "Education", 50000),
            ("Swathi", "Baskaran", "Technology", 65000),
            ("Thangam", "Meghanathan", "Management", 109000),
            ("Vani", "Megham", "Technology", 80000),
            ("Ravi", "Chandran", "Technology", 110000)]

columns = ["FirstName", "LastName", "Department", "Salary"]
dataDF = spark.createDataFrame(data = data, schema = columns)

[102] from pyspark.sql.window import Window
      from pyspark.sql.functions import row_number, rank, dense_rank, percent_rank, desc

      windowSpec = Window.partitionBy("Department").orderBy(("Salary"), ("Department"))
```

1. Row Number

```
[103] # WindowFunctions - RowNumber
      dataDF.withColumn("Row_Number", row_number().over(windowSpec)).show()
```

FirstName	LastName	Department	Salary	Row_Number
Preethi	Sivam	Education	50000	1
Ambika	Prathap	Education	65000	2
Harry	Thomas	Finance	70000	1
Henry	Finch	Finance	70000	2
James	Potter	Finance	90000	3
Lily	Evans	Management	80000	1
Thangam	Meghanathan	Management	109000	2
Swathi	Baskaran	Technology	65000	1
Vani	Megham	Technology	80000	2
Ravi	Chandran	Technology	110000	3

2. Rank

```
0s # Window Function - Rank
dataDF.withColumn("Rank", rank().over(windowSpec)).show()
```

FirstName	LastName	Department	Salary	Rank
Preethi	Sivam	Education	50000	1
Ambika	Prathap	Education	65000	2
Harry	Thomas	Finance	70000	1
Henry	Finch	Finance	70000	1
James	Potter	Finance	90000	3
Lily	Evans	Management	80000	1
Thangam	Meghanathan	Management	109000	2
Swathi	Baskaran	Technology	65000	1
Vani	Megham	Technology	80000	2
Ravi	Chandran	Technology	110000	3

3. Dense Rank

```
0s [105] # Window Function - Dense Rank
dataDF.withColumn("Dense Rank", dense_rank().over(windowSpec)).show()
```

FirstName	LastName	Department	Salary	Dense Rank
Preethi	Sivam	Education	50000	1
Ambika	Prathap	Education	65000	2
Harry	Thomas	Finance	70000	1
Henry	Finch	Finance	70000	1
James	Potter	Finance	90000	2
Lily	Evans	Management	80000	1
Thangam	Meghanathan	Management	109000	2
Swathi	Baskaran	Technology	65000	1
Vani	Megham	Technology	80000	2
Ravi	Chandran	Technology	110000	3

4. Percent Rank

```
✓ [106] # Window Function - Percent Rank  
0s dataDF.withColumn("Percent Rank", percent_rank().over(windowSpec)).show()
```

```
⇒ +-----+-----+-----+-----+-----+  
|FirstName|  LastName|Department|Salary|Percent Rank|  
+-----+-----+-----+-----+-----+  
|   Preethi|    Sivam| Education| 50000|         0.0|  
|   Ambika|   Prathap| Education| 65000|         1.0|  
|    Harry|   Thomas|  Finance| 70000|         0.0|  
|    Henry|    Finch|  Finance| 70000|         0.0|  
|    James|   Potter|  Finance| 90000|         1.0|  
|    Lily|    Evans|Management| 80000|         0.0|  
| Thangam|Meghanathan|Management|109000|         1.0|  
|   Swathi|   Baskaran|Technology| 65000|         0.0|  
|    Vani|    Megham|Technology| 80000|         0.5|  
|    Ravi|   Chandran|Technology|110000|         1.0|  
+-----+-----+-----+-----+-----+
```