

SparkRDD 的 Transformations/Actions 操 作实战

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在前面[Spark编程原理及RDD的特性与基本操作](#)介绍了SparkRDD的操作分为两个部分Transformation和Action。这两种操作分为多个算子（即操作函数）。Transformation针对已有的RDD创建一个新的RDD，主要是对数据进行映射，变换，统计，过滤。。。Action主要是对数据进行最后的执行操作，遍历，聚合，保存等操作。下面来看下这些操作的具体实现。

Transformations

Map(func):对RDD中的每个元素通过函数func进行映射。

```
/**  
 * map算子：给集合每个元素*2  
 */  
private static void mapTest() {  
    SparkConf conf = new SparkConf().setAppName("mapTest").setMaster("local");  
    JavaSparkContext jsc = new JavaSparkContext(conf);  
  
    List<Integer> list = Arrays.asList(1, 2, 3, 4, 5, 6, 7, 8);  
    JavaRDD<Integer> numRDD = jsc.parallelize(list);  
    JavaRDD<Integer> num2RDD = numRDD.map(num -> num * 2);  
  
    num2RDD.foreach(num2 -> System.out.printf(num2.toString() + " "));  
    jsc.close();  
}  
/**  
 * map算子：给集合每个元素*2  
 */  
def mapTest(){  
    val conf = new SparkConf().setAppName("mapTest").setMaster("local")  
    val sc  = new SparkContext(conf)  
    val list = Array(1,2,3,4,5,6,7,8)  
    val numRDD = sc.parallelize(list)  
    val num2RDD = numRDD.map(num=>num*2)  
    num2RDD.foreach(num2=>print(num2+" "))  
}  
map结果: 2 4 6 8 10 12 14 16
```

filter:过滤或者选择满足条件的数据

```
/**  
 * filter算子：满足过滤条件的保留  
 */  
private static void filterTest() {  
    SparkConf conf = new SparkConf().setAppName("filterTest").setMaster("local");  
    JavaSparkContext jsc = new JavaSparkContext(conf);  
  
    List<Integer> list = Arrays.asList(1, 2, 3, 4, 5, 6, 7, 8, 9, 10);  
    JavaRDD<Integer> numRDD = jsc.parallelize(list);  
    //过滤偶数集合  
    JavaRDD<Integer> num2RDD = numRDD.filter(num -> num % 2 == 0);  
    num2RDD.foreach(num2 -> System.out.print(num2 + " "));
```

```

        jsc.close();
    }
    /**
     * fileter算子：满足过滤条件的保留
     */
    def filterTest(): Unit ={
        val conf = new SparkConf().setAppName("filterTest").setMaster("local")
        val sc  = new SparkContext(conf)
        val list = Array(1,2,3,4,5,6,7,8,9,10)
        val numRDD = sc.parallelize(list)
        //filter过滤偶数
        val evenNumRDD = numRDD.filter(num=> num % 2 == 0)
        evenNumRDD.foreach(evenNum => print(evenNum + " "))
    }
filter结果: 2 4 6 8 10

```

flatMap:将一个输入映射成0-多个输出

```

    /**
     * flatmap算子：接收RDD中所有元素，并进行运算，然后返回多个元素
     * 这里是对元素分割成单词
     */
    private static void flatMapTest() {
        SparkConf conf = new SparkConf().setAppName("flatMapTest").setMaster("local");
        JavaSparkContext jsc = new JavaSparkContext(conf);

        List<String> list = Arrays.asList("hello world", "hello you", "good morning");
        JavaRDD<String> wordsRDD = jsc.parallelize(list);
        //将每条数据通过空格分隔成多个数据
        JavaRDD<String> wordRDD = wordsRDD.flatMap(
            words -> Arrays.asList(words.split(" ")).listIterator()
        );

        wordRDD.foreach(
            word -> System.out.printf(word+ " ")
        );
        jsc.close();
    }
    /**
     * flatmap算子：接收RDD中所有元素，并进行运算，然后返回多个元素
     * 这里是对元素分割成单词
     */
    def flatMapTest(): Unit ={
        val conf = new SparkConf().setMaster("local").setAppName("flatMapTest")
        val sc = new SparkContext(conf)
        val list = Array("hello world", "hello you", "good morning")
        val wordsRDD = sc.parallelize(list)
        val wordRDD = wordsRDD.flatMap(words=>words.split(" "))
        wordRDD.foreach(word=>println(word).toString)
    }
flatMap结果:
hello

```

world
hello
you
good
morning

groupByKey:按k分组(k,v),(k,v2),(k2,v)(k3,v2) => (k,(v,v)),
sortByKey:按key排序

```
/**  
 * groupByKey算子: 按key分组,  
 * sortByKey:按key排序,无参由小到大, 参数false,由大到小  
 */  
private static void groupByKeyTest() {  
    SparkConf conf = new SparkConf().setAppName("groupByKeyTest").setMaster("local");  
    JavaSparkContext jsc = new JavaSparkContext(conf);  
  
    List<Tuple2<String,Integer>> list = Arrays.asList(  
        new Tuple2<>("A",88),  
        new Tuple2<>("B",78),  
        new Tuple2<>("C",55),  
        new Tuple2<>("A",95),  
        new Tuple2<>("C",34)  
    );  
    JavaPairRDD<String,Integer> scoreRDD = jsc.parallelizePairs(list);  
  
    //不加sortBykey()结果乱序  
    //JavaPairRDD<String,Iterable<Integer>> scoreGroupRDD = scoreRDD.groupByKey();  
    JavaPairRDD<String,Iterable<Integer>> scoreGroupRDD = scoreRDD.groupByKey().sortByKey();  
  
    scoreGroupRDD.foreach(  
        scoreGroup-> {  
            System.out.printf(scoreGroup._1 + ": ");  
            scoreGroup._2.forEach(score-> System.out.printf(score+ " "));  
            System.out.println();  
        }  
    );  
    jsc.close();  
}  
/**  
 * groupByKey算子: 按key分组,  
 * sortByKey:按key排序,无参由小到大, 参数false,由大到小  
 */  
def groupByKeyTest(): Unit ={  
    val conf = new SparkConf().setAppName("groupByKeyTest").setMaster("local")  
    val sc = new SparkContext(conf)  
    val list = Array(  
        Tuple2("A",89),  
        Tuple2("c",59),  
        Tuple2("B",74),  
        Tuple2("c",50),  
        Tuple2("A",98))
```

```

val scoreRDD = sc.parallelize(list)
//不加sortByKey结果乱序
// val groupRDD = scoreRDD.groupByKey()
val groupRDD = scoreRDD.groupByKey().sortByKey()
groupRDD.foreach(
  scoreTP=> {
    print(scoreTP._1 + ": ")
    scoreTP._2.foreach(score=>print(score + " "))
    println()
  })

```

groupByKey结果

A: 88 95

B: 78

C: 55 34

reduceByKey (func) :key相同V用func函数聚合，这

```

/**
 * reduceByKey算子：按key分组并聚合(聚合函数可以自己定义)这里是相加，即key相同则V相加, (k
 ,2) ,(k1,3),(k2,5) => (K1,5),(k2,5)
 */
private static void reduceByKeyTest() {
  SparkConf conf = new SparkConf().setAppName("reduceByKeyTest").setMaster("local");
  JavaSparkContext jsc = new JavaSparkContext(conf);
  List<Tuple2<String,Integer>> list = Arrays.asList(
    new Tuple2<>("A",88),
    new Tuple2<>("B",78),
    new Tuple2<>("C",55),
    new Tuple2<>("A",95),
    new Tuple2<>("C",34)
  );
  JavaPairRDD<String,Integer> scoreRDD = jsc.parallelizePairs(list);
  //定义的聚合函数是：v相加
  JavaPairRDD<String,Integer> scoreGroupRDD = scoreRDD.reduceByKey((v1,v2)->v1+v2);
  scoreGroupRDD.foreach(
    score->System.out.println(score._1+": "+score._2)
  );
  jsc.close();
}
/**
 * reduceByKey算子：按key分组并聚合(聚合函数可以自己定义)这里是相加，即key相同则V相加, (k
 ,2) ,(k1,3),(k2,5) => (K1,5),(k2,5)
 */
def reduceByKeyTest(): Unit ={
  val conf = new SparkConf().setAppName("reduceByKeyTest").setMaster("local")
  val sc = new SparkContext(conf)
  val list = Array(
    Tuple2("A",89),
    Tuple2("c",59),
    Tuple2("B",74),
    Tuple2("c",50),

```

```

Tuple2("A",98))

val scoreRDD = sc.parallelize(list)
val totalRDD = scoreRDD.reduceByKey(_+_).sortByKey()
totalRDD.foreach(
  total=>println(total._1+": "+total._2)
)
}
reduceByKey结果:
(B,78)
(A,183)
(C,89)

```

join和cogroup:对RDD进行连接

```

/*
 * join算子：关联两个RDD，(K,V).join(K,W)=>(K,(V,W)),都存在的K, V保留。其他的丢弃,相当于做
 * 集。
 * cogroup算子:相当于集合+, 做并集。只要k存在就有一条记录。
 */
private static void joinAndCoGroupTest(){
  SparkConf conf = new SparkConf().setMaster("local").setAppName("joinAndCoGroupTest");
  JavaSparkContext jsc = new JavaSparkContext(conf);

  List<Tuple2<Integer,Integer>> scoreList = Arrays.asList(
    new Tuple2<>(1,88),
    new Tuple2<>(2,98),
    new Tuple2<>(3,75),
    new Tuple2<>(4,84),
    new Tuple2<>(6,77)
  );
  List<Tuple2<Integer,String>> stuList = Arrays.asList(
    new Tuple2<>(1,"A"),
    new Tuple2<>(2,"B"),
    new Tuple2<>(3,"C"),
    new Tuple2<>(4,"D"),
    new Tuple2<>(5,"E")
  );
  JavaPairRDD<Integer,Integer> scoreRDD = jsc.parallelizePairs(scoreList);
  JavaPairRDD<Integer,String> stuRDD = jsc.parallelizePairs(stuList);

  //join链接操作(通过Key做链接, 保留都存在的key)
  JavaPairRDD scoreStuRDD= scoreRDD.join(stuRDD).sortByKey();
  scoreStuRDD.foreach(
    score-> System.out.println(score.toString())
  );

  JavaPairRDD stuScoreRDD= stuRDD.join(scoreRDD).sortByKey();
  scoreStuRDD.foreach(
    score-> System.out.println(score.toString())
  );

  //cogroup连接操作:保留所有的key

```

```

JavaPairRDD scoreStuCGRDD= scoreRDD.cogroup(stuRDD).sortByKey();
scoreStuCGRDD.foreach(
    score-> System.out.println(score.toString())
);
JavaPairRDD stuScoreCGRDD= stuRDD.cogroup(scoreRDD).sortByKey();
scoreStuCGRDD.foreach(
    score-> System.out.println(score.toString())
);
}
/** 
 * join算子：关联兩個RDD， (K,V).join(K,W)=>(K,(V,W)),都存在的K， V保留。其他的丢弃,相当于做集。
 * cogroup算子;相当于集合+， 做并集。只要k存在就有一条记录。
*/
def joinAndCoGroupTest(): Unit ={
  val conf = new SparkConf().setAppName("joinAndCoGroupTest").setMaster("local")
  val sc = new SparkContext(conf)

  val scoreList = Array(
    Tuple2(1,89),
    Tuple2(2,59),
    Tuple2(3,74),
    Tuple2(4,50),
    Tuple2(5,98)
  )
  val stuList = Array(
    Tuple2(1,"A"),
    Tuple2(2,"B"),
    Tuple2(3,"C"),
    Tuple2(4,"D"),
    Tuple2(6,"E")
  )
  val scoreRDD = sc.parallelize(scoreList)
  val stuRDD = sc.parallelize(stuList)
  //join联合两个RDD
  val joinRDD = scoreRDD.join(stuRDD).sortByKey()
  joinRDD.foreach(
    join=>println(join.toString())
  )
  //cogroup联合两个RDD
  val cogroupRDD = scoreRDD.cogroup(stuRDD).sortByKey()
  cogroupRDD.foreach(
    co=>println(co.toString())
  )
}
join结果:
(1,(88,A))
(2,(98,B))
(3,(75,C))
(4,(84,D))
cogroup结果:
(1,([88],[A]))
(2,([98],[B]))
(3,([75],[C]))

```

```
(4,([84],[D]))  
(5,([], [E]))  
(6,([77],[]))
```

Actions算子

reduce算子：

```
/**  
 * reduce算子：1, 2聚合结果和3聚合结果和4聚合，递归  
 */  
private static void reduceTest(){  
    SparkConf conf = new SparkConf().setAppName("reduceTest").setMaster("local");  
    JavaSparkContext jsc = new JavaSparkContext(conf);  
  
    List<Integer> numList = Arrays.asList(1,2,3,4,5,6,7,8,9);  
    JavaRDD<Integer> numRDD = jsc.parallelize(numList);  
  
    Integer sumRDD = numRDD.reduce((x1, x2) -> x1 + x2);  
    System.out.printf(sumRDD.toString());  
  
    jsc.close();  
}  
/**  
 * reduce算子：1, 2聚合结果和3聚合结果和4聚合，递归  
 */  
def reduceTest(): Unit = {  
    val conf = new SparkConf().setAppName("reduceTest").setMaster("local")  
    val sc = new SparkContext(conf)  
    val numList = Array(1,2,3,4,5,6,7,8,9)  
    val parRDD = sc.parallelize(numList)  
    val sum = parRDD.reduce(_ + _)  
    println(sum)  
}  
reduce结果： 45
```

count和collect:count统计RDD元素个数， collect将RDD数据拉取到本地

```
/**  
 * count:统计RDD元素数量  
 * collect算子：将RDD数据拉取到本地,大量数据时，性能比较差，高Io，或者oom内存溢出  
 */  
private static void collectTest(){  
    SparkConf conf = new SparkConf().setAppName("collectTest").setMaster("local");  
    JavaSparkContext jsc = new JavaSparkContext(conf);  
  
    List<Integer> numList = Arrays.asList(1,2,3,4,5,6,7,8,9);  
    JavaRDD<Integer> numRDD = jsc.parallelize(numList);
```

```

//count:
long count = numRDD.count();
System.out.printf("元素个数: "+count);

//map:元素翻倍
JavaRDD<Integer> doubleNumRDD = numRDD.map(x->x*2);

//collect:将RDD的数据拉取到本地, 变成了java的List,
List<Integer> listNum= doubleNumRDD.collect();
listNum.forEach(
    num-> System.out.printf(num.toString()+" ")
);

    jsc.close();
}
/***
 * count:统计RDD元素数量
 * collect算子: 将RDD数据拉取到本地,大量数据时, 性能比较差, 高Io, 或者oom内存溢出
 */
def collectTest(): Unit ={
    val conf = new SparkConf().setAppName("collectTest").setMaster("local")
    val sc = new SparkContext(conf)
    val numList = Array(1,2,3,4,5,6,7,8,9)
    val numRDD = sc.parallelize(numList)
    //count
    val count = numRDD.count()
    println("元素个数: "+count);
    val doubleRDD = numRDD.map(x=>x*2)
    //collect
    val listNum = doubleRDD.collect()
    listNum.foreach(
        num=>println(num+" ")
    )
}

```

count结果: 元素个数: 9
 collect结果: 2 4 6 8 10 12 14 16 18

take(N)算子:取RDD的前N条记录

```

/**
 * take(N),类似collect,把RDD拉取到本地, 取前N条数据
 */
private static void takeTest(){
    SparkConf conf = new SparkConf().setAppName("takeTest").setMaster("local");
    JavaSparkContext jsc = new JavaSparkContext(conf);

    List<Integer> numList = Arrays.asList(1,2,3,4,5,6,7,8,9);
    JavaRDD<Integer> numRDD = jsc.parallelize(numList);

    //take(N):
    List<Integer> list = numRDD.take(8);
    list.forEach(

```

```

        num-> System.out.printf(num+ " ")
    );
    jsc.close();
}
/**
 * take(N),类似collect,把RDD拉取到本地, 取前N条数据
 */
def takeTest(): Unit = {
    val conf = new SparkConf().setAppName("takeTest").setMaster("local")
    val sc = new SparkContext(conf)
    val numList = Array(1, 2, 3, 4, 5, 6, 7, 8, 9)
    val numRDD = sc.parallelize(numList)
    val num5 = numRDD.take(5)
    for (n <- num5) {
        println(n)
    }
    // num5.foreach(n => print(n + " "))
}
take(8)结果: 1 2 3 4 5 6 7 8

```

saveAsTextFile算子:保存RDD到本地

```

/**
 * saveAsTextFile:保存到文件
 */
private static void saveAsTextFileTest(){
    SparkConf conf = new SparkConf().setAppName("saveAsTextFile").setMaster("local");
    JavaSparkContext jsc = new JavaSparkContext(conf);

    List<String > list = Arrays.asList("hello world","how are you","nice to");
    JavaRDD<String> strRDD = jsc.parallelize(list);
    //saveAsTextFile
    strRDD.saveAsTextFile("input/str");

    jsc.close();
}
/**
 * saveAsTextFile:保存到文件
 */
def saveAsTextFileTest(): Unit = {
    val conf = new SparkConf().setAppName("takeTest").setMaster("local")
    val sc = new SparkContext(conf)
    val numList = Array("TaskSchedulerImpl","DAGScheduler")
    val numRDD = sc.parallelize(numList)
    numRDD.saveAsTextFile("input/scalasaveAsTextFile")
}

```

countByKey:统计key相同的记录 (相当于对K分组)

```

/**
 * countByKey:统计每个key的数量

```

```

*/
private static void countByKeyTest(){
    SparkConf conf = new SparkConf().setAppName("countByKeyTest").setMaster("local");
    JavaSparkContext jsc = new JavaSparkContext(conf);

    List<Tuple2<Integer,Integer>> numList = Arrays.asList(
        new Tuple2<>(1,100),
        new Tuple2<>(2,50),
        new Tuple2<>(1,100),
        new Tuple2<>(4,80),
        new Tuple2<>(5,50),
        new Tuple2<>(6,80)
    );
    //如果数据是tuple类型的则要用parallelizePairs来并行化，JavaPairRDD来接收类型，而你不是JavaRDD,
    JavaPairRDD<Integer,Integer> numRDD = jsc.parallelizePairs(numList);
    Map<Integer,Long> count = numRDD.countByKey();
    count.forEach((k,v)-> System.out.println(k+"--"+v));
    jsc.close();
}
/***
 * countByKey:统计每个key的数量
 */
def countByKeyTest(): Unit = {
    val conf = new SparkConf().setAppName("takeTest").setMaster("local")
    val sc = new SparkContext(conf)
    val numList = Array("TaskSchedulerImpl","DAGScheduler","TaskSchedulerImpl")
    val numRDD = sc.parallelize(numList)

    numRDD.countByValue().foreach(num=>println(num._1+"--"+num._2));
}
countByKey结果:
5--1
1--2
6--1
2--1
4--1

```