## Spark Lesson 3

6 questions

| 1 point  1. Check all true statements about the Directed Acyclic Graph Scheduler     |   |
|--|---|
|  |   |
|  | A DAG is used to track dependencies of each partition of each RDD   |
|  | If a partition is lost, the DAG is traversed forward to check what other steps are affected                                       |
|  | The DAG is managed by the cluster manager   |
|  | Each transformation is executed as soon as it is called on a RDD  |
| 1 point  |   |
| <ol><li>Why is building a DAG necessary in Spark but not in MapReduce?</li></ol>     |   |
| 0  | In order to make a computation distributed at large scale   |
| 0  | For resiliency: it is necessary to make sure a partition can be recovered in case it is lost.                                     |
| 0  | Because MapReduce always has the same type of workflow, Spark needs to accommodate diverse workflows.                             |
| 1 point 3.   |   |
| What are the differences between an action and a transformation? Mark all that apply |   |
|  | A transformation is lazy, an action instead executes immediately.   |
|  | A transformation is from worker nodes to worker nodes, an action between worker nodes and the Driver (or a data source like HDFS) |
|  | An action always triggers a shuffle.  |
|  | An action always writes the disk.   |

```
1
  point
Generally, which are good stages to mark a RDD for caching in memory?
       The first RDD, just after reading from disk, so we avoid reading from disk again.
       Every 2 or 3 transformations, to keep a recent backup.
       At the start of an iterative algorithm.
       After data cleaning, parsing and validation.
    1
  point
What are good cases for using a broadcast variable? Mark all that apply
       Copy a large configuration dictionary to all worker nodes
       Copy a large lookup table to all worker nodes
       Copy a small/medium sized RDD for a join
       Broadcast a Python module to all worker nodes
    1
  point
We would like to count the number of invalid entries in this example dataset:
       invalid = sc.accumulator(0)
      d = sc.parallelize(["3", "23", "S", "99", "TT"]).foreach(count_invalid)
   2
What would be a good implementation of the count_invalid function?
              def count_invalid(element):
                       int(element)
           3
           4
                   except:
                       invalid = invalid.add(1)
              def count_invalid(element):
           2
```

3

4

int(element)

invalid.add(1)

except:

```
1 def count_invalid(element):
2    try:
3    int(element)
4    except:
5    invalid = invalid + 1
```

```
1 def count_invalid(element):
2    try:
3         int(element)
4    except:
5    invalid.accumulate(1)
```

**/** 

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