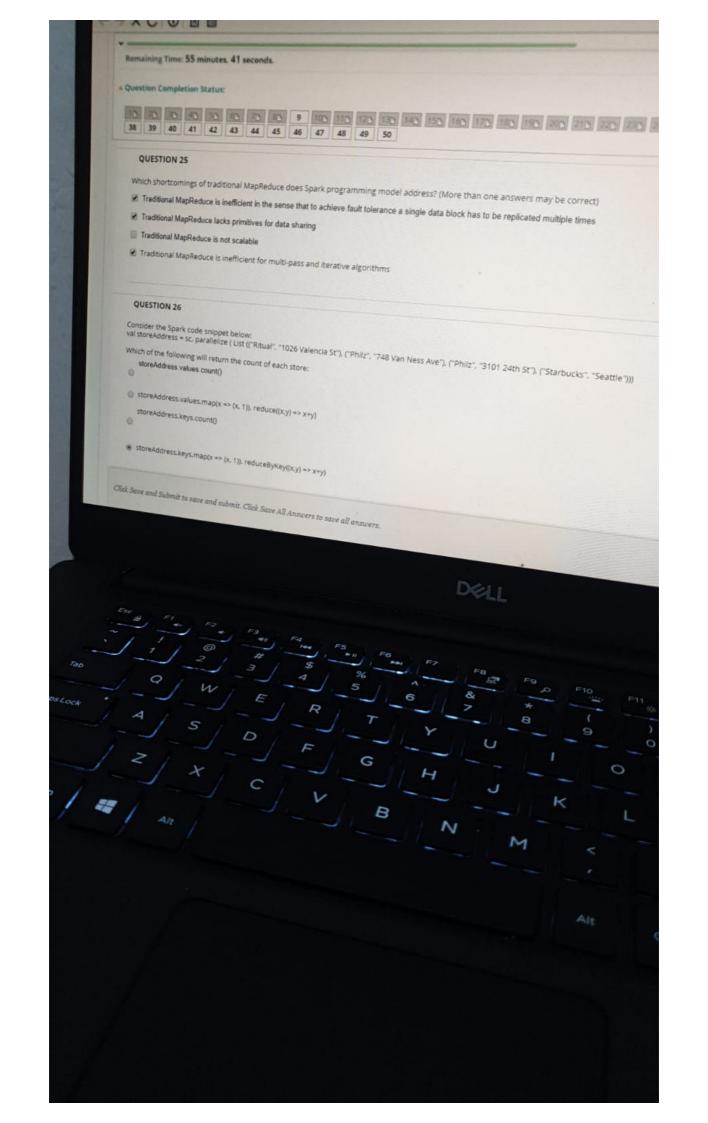
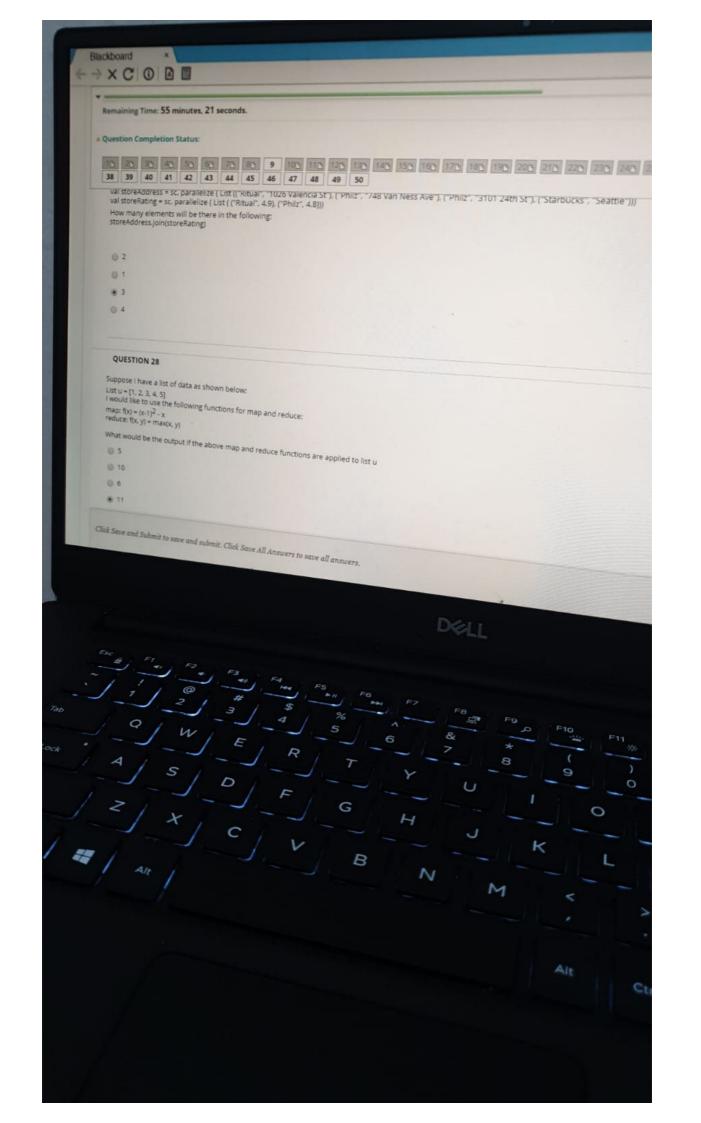
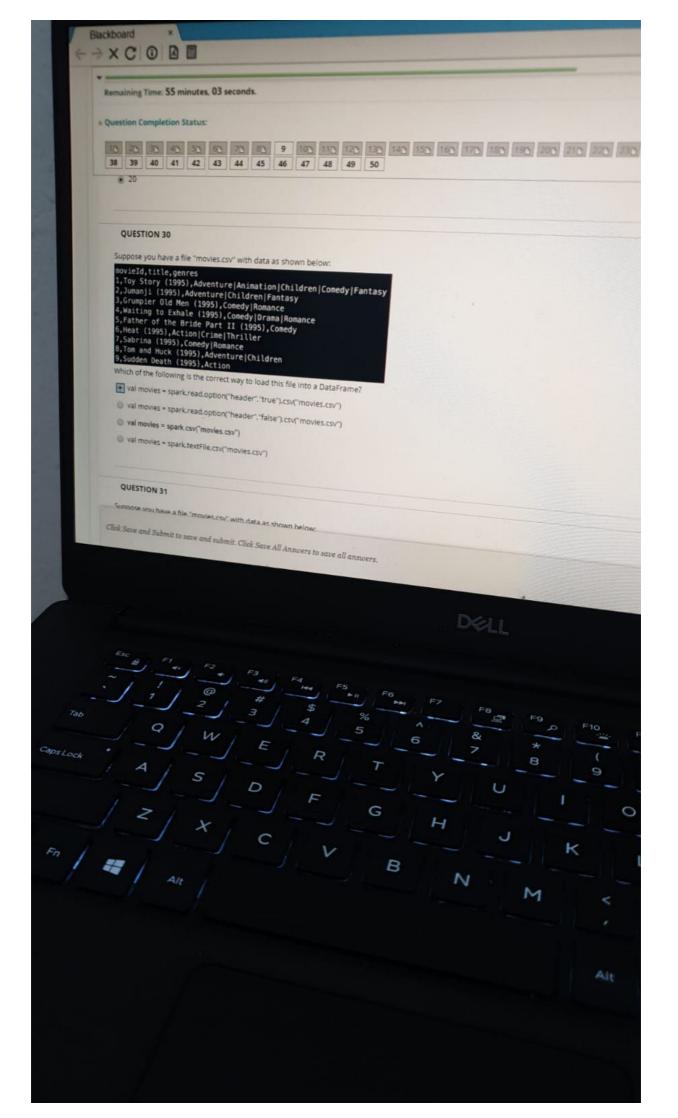
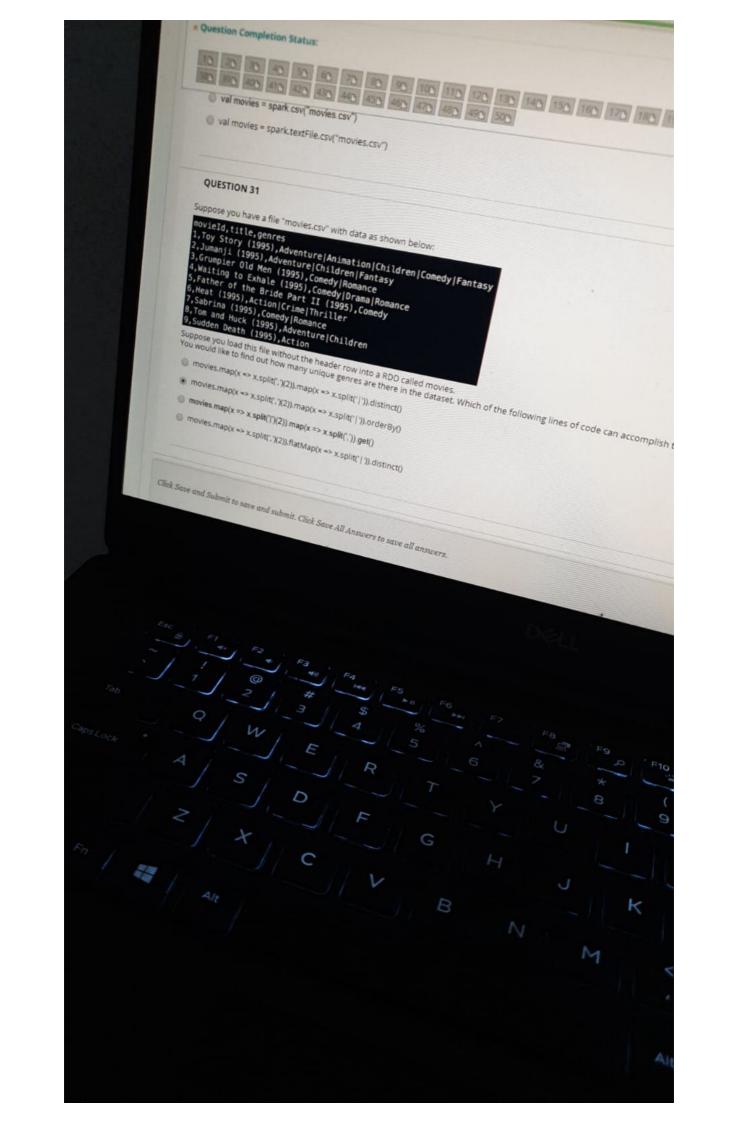


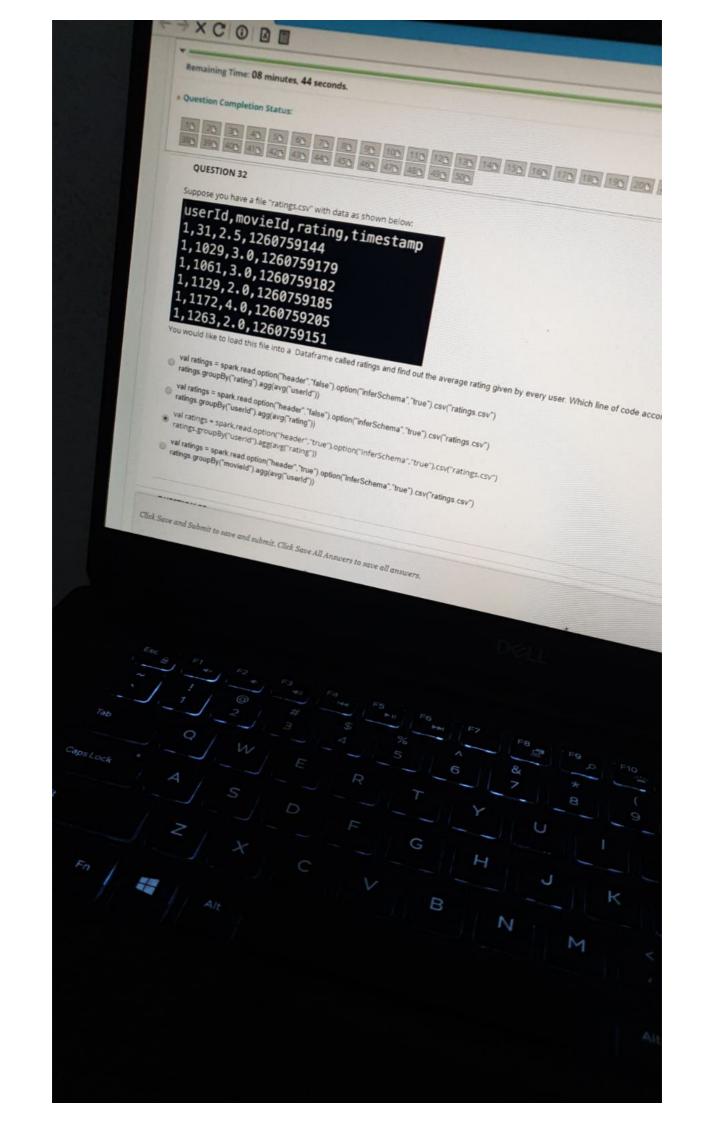
# **QUESTION 23** Which of the following is false? Namenode stores the entire file system metadata in memory. Datanodes send data directly to the clients Datanodes send data to the Namenode, which forwards the data to the clients Clients read data by first asking the Namenode for the location of the blocks. **QUESTION 24** Which of the following is true about the log files of the NameNode • The fsimage log file is used to create a point in time snapshot of the file system, and incremental changes are written to the edits log file. The edits log file is used to create a point in time snapshot of the file system, and incremental changes are written to the fsimage log file. Click Save and Submit to save and submit. Click Save All Answers to save all answers. DELL W E G 0 K B Alt

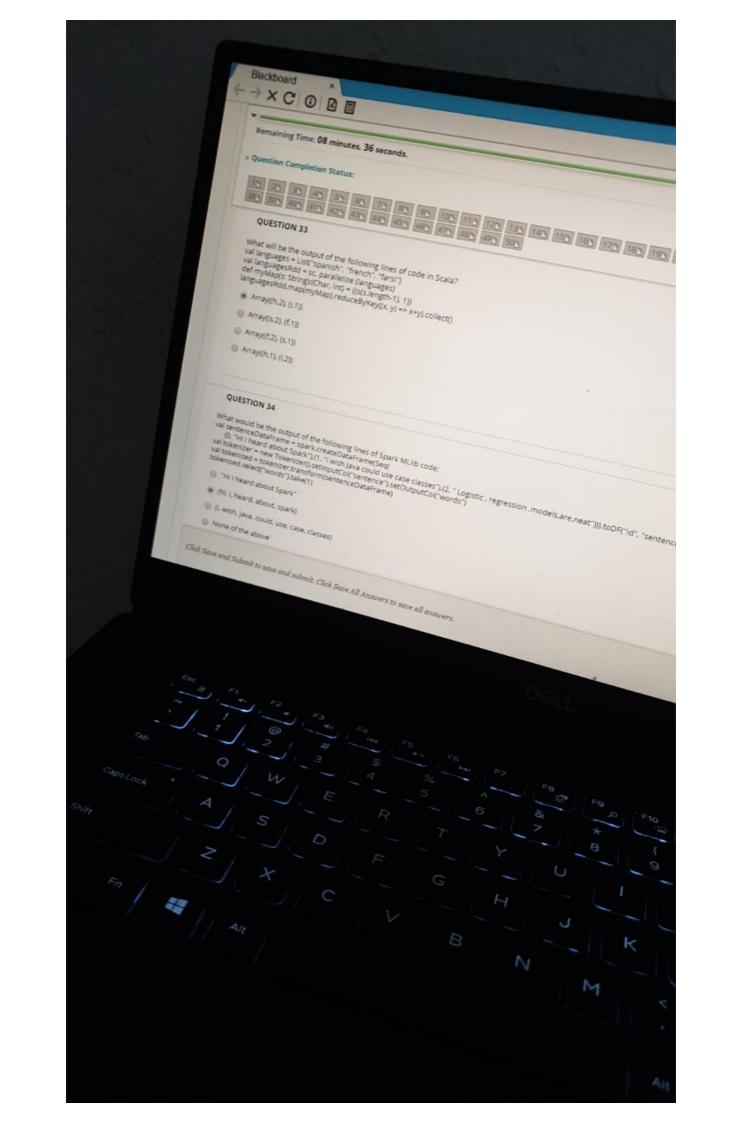


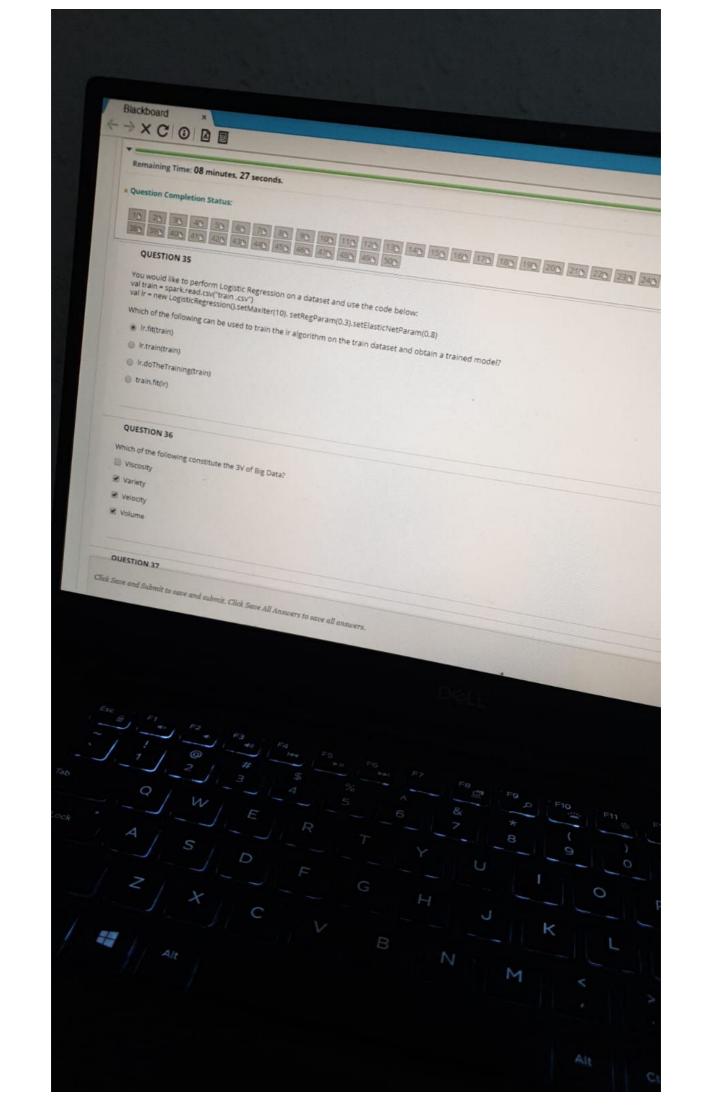


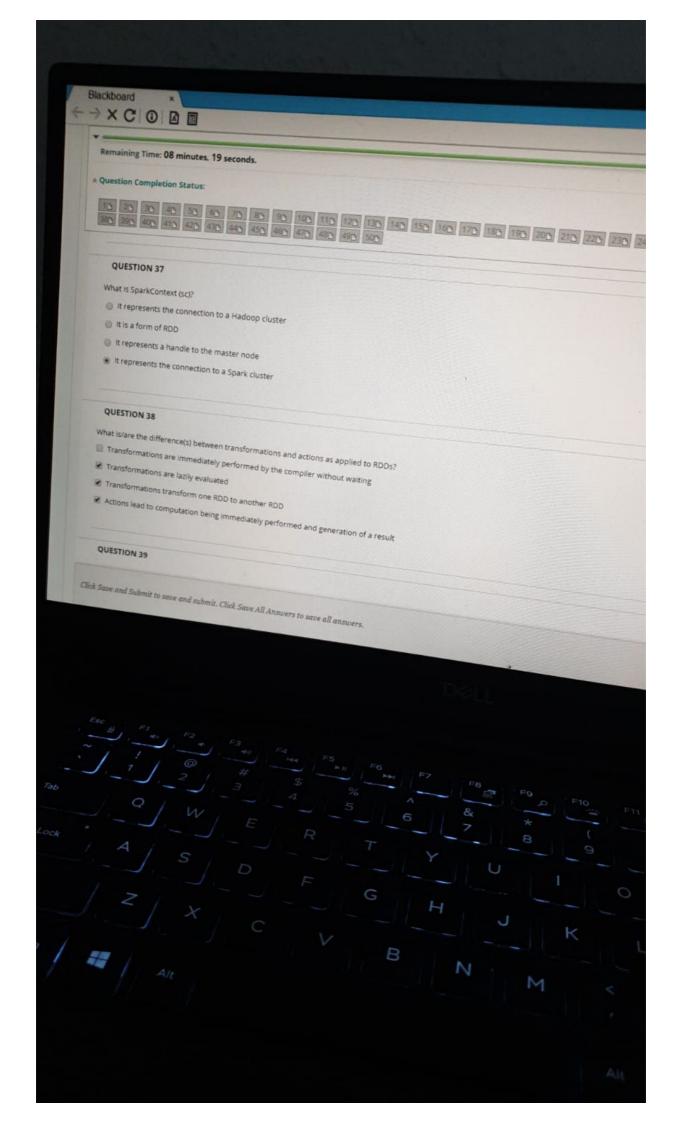


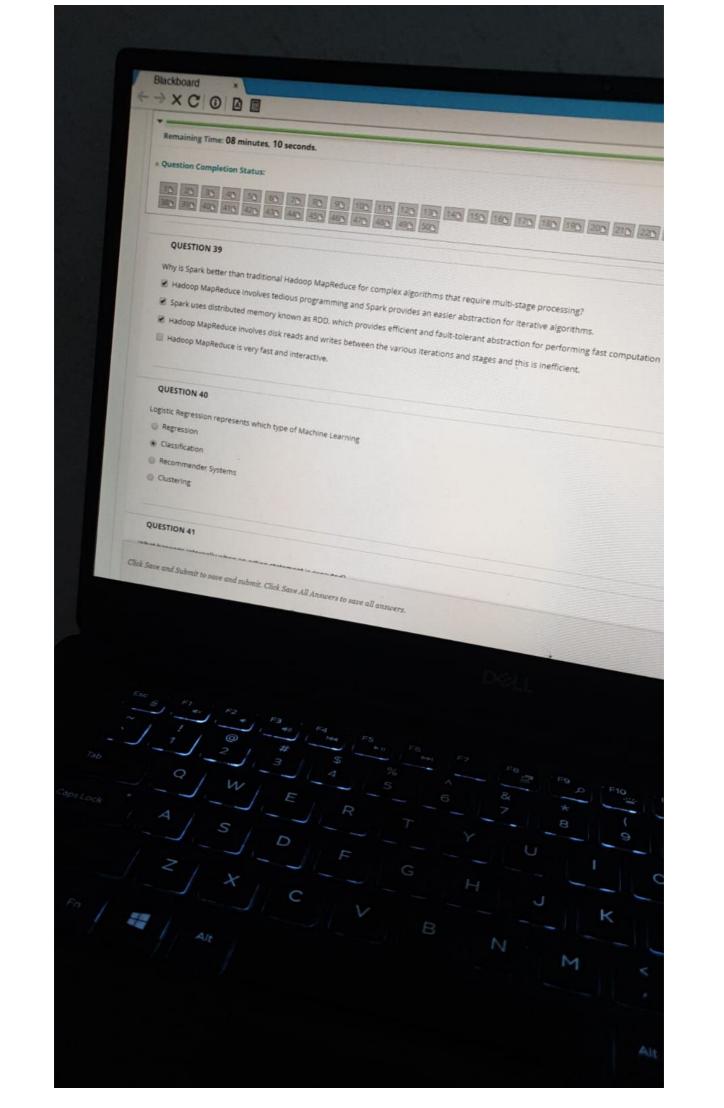


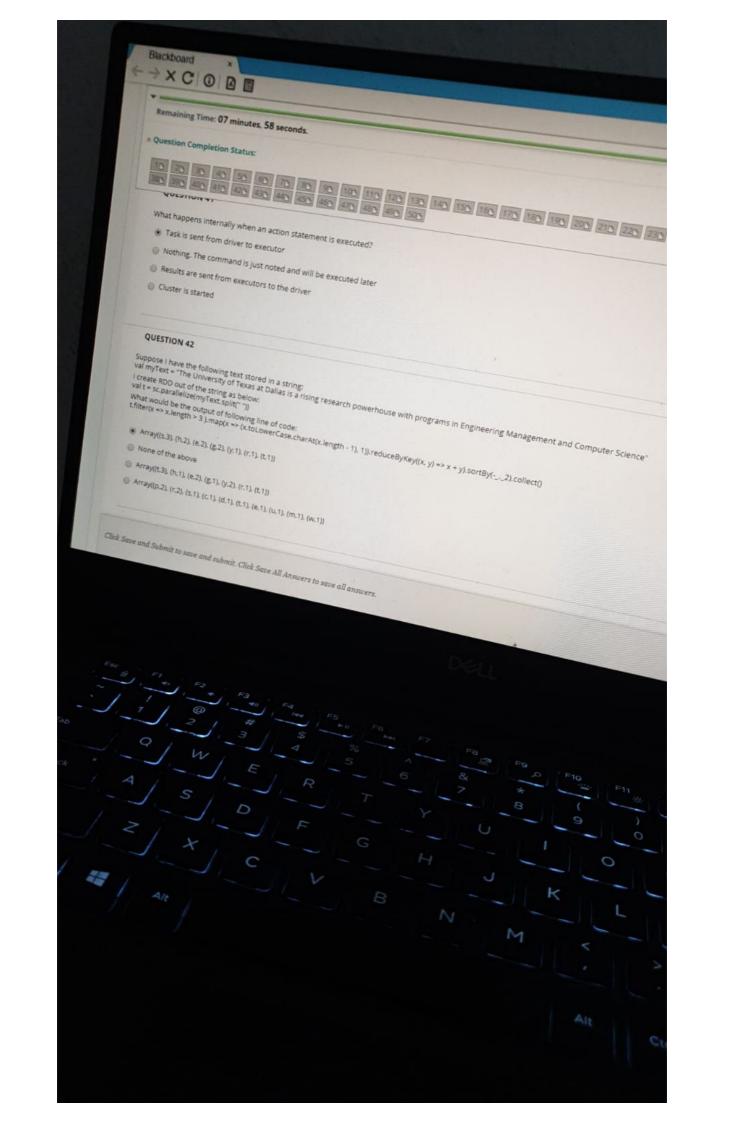












## **QUESTION 9**

Given the following HDFS file called test.txt Hello World the apple is sweet i like sweet apple

You run the following lines on Spark code on it:

val f=sc.textFile("test.txt")

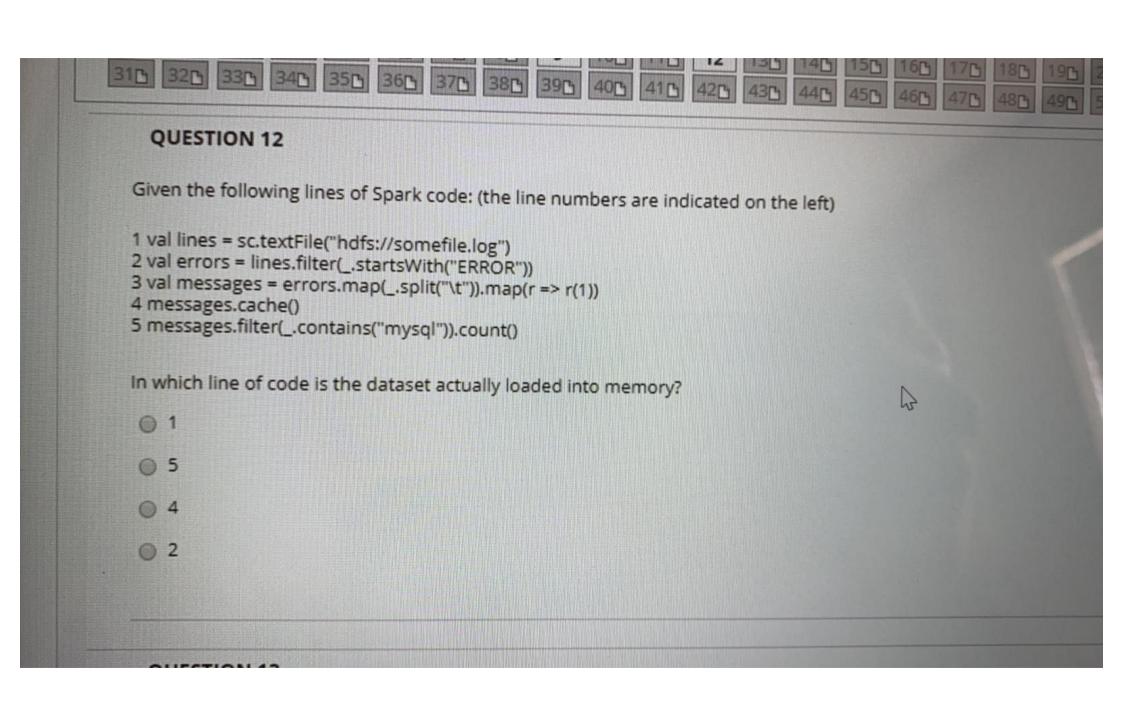
val w = f.flatMap(l => l.split(" ")).map(word => (word, 1)).cache()
w.reduceByKey(\_ + \_ ).saveAsTextFile("out.txt")

What many lines will be there in the output file? Write your answer as a number e.g. 8

N

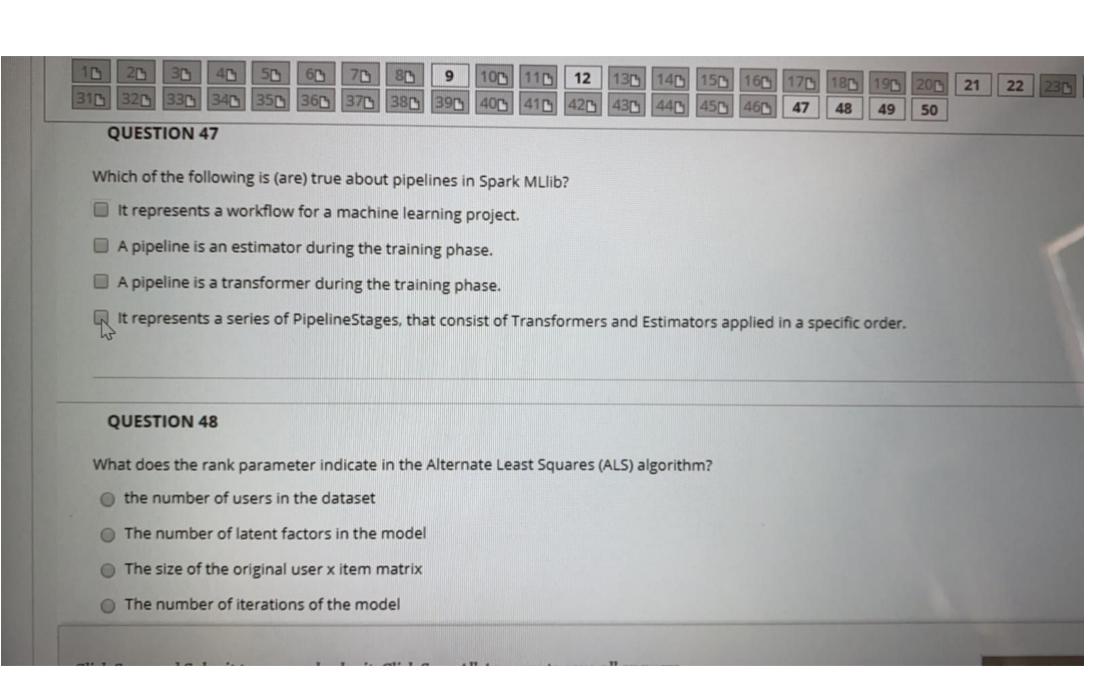
## **QUESTION 10**

Given the following Scala list:



## ♠ Question Completion Status: 4D 5D 6D 7D 8D 9 10D 11D 12 34D 35D 36D 37D 38D 39D 40D 41D 42D **QUESTION 21** Which features of HDFS make it very popular? Ability to handle unstructured data Fault Tolerance Low cost commodity hardware Scalability Faster, interactive environment Faster speed of random access of data

Faster, interactive environment	
Faster speed of random access of data	
QUESTION 22	
Which of the following are differences between traditional distributed shared	memory (DSM) and RDD based arci
RDDs are immutable while DSM can be varied by the applications	
RDD are based on coarse grained transformations, while DSM allow fine gr	ained updates
DSM can recover faster from failure than RDD	
RDDs can be recovered through lineage while DSM requires checkpoints are	nd program rollbacks



#### **QUESTION 49**

Suppose you have a movies dataframe as below:

movield	title	
1	Toy Story	
2	Jumanji	
3	Grumpier Old Men	

and a ratings dataframe as below:

userId	movield	rating
1	1	4.8
1	3	4.0

2

You would like to run the most efficient query to match following criteria:

- 1. the movie should have at least 100 ratings
- 2. we would like to see the title and average rating for those movies
- 3. the data should be sorted by average rating in descending order

Which of the following accomplishes this: (You can assume that the relevant libraries have been imported)

### **QUESTION 50**

Suppose you have a movies dataframe as below:

movield	title	
1	Toy Story	
2	Jumanji	
3	Grumpier Old Men	

and a ratings dataframe as below:

userId	movield	rating
1	1	4.8
1	3	4.0

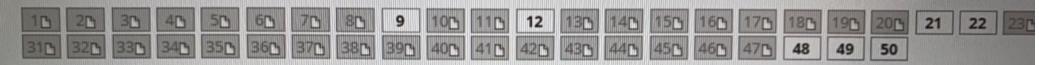


You would like to run the most efficient query to match following criteria:

- 1. the movie should have at least 100 ratings
- 2. average review of the movie should be at least 4.0
- 3. the data should be sorted by count of reviews in descending order

Which of the following accomplishes this: (You can assume that the relevant libraries have been imported)





3. the data should be sorted by count of reviews in descending order

Which of the following accomplishes this: (You can assume that the relevant libraries have been imported)

val counts = ratings.groupBy("movield").count()

- val avgRatings = ratings.groupBy("movield").agg(avg("rating")).toDF("movield","avg")
  val output = counts.join(avgRatings, Seq("movield")).join(movies, Seq("movield")).filter(\$"count" > 100).filter(\$"avg" > 4.0).orderBy(desc("count"))
  - val counts = ratings,groupBy("movield").count().filter(\$"count" > 100)
- val avgRatings = ratings.groupBy("movield").agg(avg("rating")).toDF("movield","avg").filter(\$"avg" > \$4.0) val output = counts.join(avgRatings, Seq("movield")).join(movies, Seq("movield")).orderBy(desc("avg"))
  - val counts = ratings.groupBy("movield").count().filter(\$"count" > 100)
- val avgRatings = ratings.groupBy("movield").agg(avg("(\sting")).toDF("movield","avg").filter(\\$"avg" > \\$4.0)
  val output = counts.join(avgRatings, Seq("movield")).filter(\\$"count" > 100).filter(\\$"avg" > 4.0).join(movies,
  Seq("movield")).orderBy(desc("avg"))
  - val counts = ratings.groupBy("movield").count().filter(\$"count" > 100)
- val avgRatings = ratings.groupBy("movield").agg(avg("rating")).toDF("movield","avg").filter(\$"avg" > 4.0)
   val output = counts.join(avgRatings, Seq("movield")).join(movies, Seq("movield")).orderBy(desc("count"))