DS 644: Homework 3

Instructions. Answer the following multiple choice questions by selecting the correct choices.

1.	Pri	nciples of (functional) programming
	(a)	Which of the following are programming paradigms? (Select three.) \Box Declarative \Box Functional \Box Hadoop \Box Imperative \Box Scala
	(b)	What three concepts characterize a purely functional programming language? \Box referential transparency \Box input/output \Box no side effects \Box procedural \Box immutability
2.	Big	data properties.
	(a)	In lecture we discussed the meaning of the term "Big Data." We decided that, for simplicity, we will call data "big" when it is □ at least 1Gb □ too big to fit in fast memory (cpu cache + ram) on a single compute node □ too big to fit in all computer memory (whether fast or slow) □ too big to be dealt with by traditional data-processing software
	(b)	"Big Data" concerns which of the following types of data? \Box structured \Box semi-structured \Box unstructured \Box all of these
	(c)	JSON and XML are examples of which type of data? \Box structured \Box unstructured \Box semi-structured \Box none of these
	(d)	Which two of the following statements are true of unstructured data? ☐ It is generally easier to analyze than other types of data. ☐ It is often referred to as "messy" data. ☐ It fits neatly into a schema. ☐ It is the most widespread type of data. ☐ It is usually found in tables.
3.	Lat	ency and fault-tolerance.
	(a)	Latency is degradation in performance due to □ a small number of cores in the central processing unit □ slow data transfer across the network or cluster □ shuffling data between different nodes in a cluster □ failure of one or more nodes in the cluster □ stack overflow caused by recursion
	(b)	Hadoop achieves fault-tolerance by □ using lazy evaluation and garbage collection. □ writing intermediate computations to disk. □ keeping all data immutable and in-memory. □ replaying functional transformations over the original (immutable) dataset.

	(c)	Spark decreases latency while remaining fault-tolerant by
		\square using ideas from imperative programming.
		\square using ideas from functional programming.
		□ discarding data when it's no longer needed.
		\square keeping all data immutable and in-memory.
		$\hfill\Box$ replaying functional transformations over the original (immutable) dataset.
4.	Trai	nsformations and actions.
	(a)	In Spark a transformation on an RDD
		□ is eagerly evaluated.
		□ is lazily evaluated.
		\Box immediately computes and returns a result.
		\Box does not immediately compute a result.
		$\hfill\Box$ usually returns another RDD (once it's evaluated).
	(b)	In Spark an action on an RDD
		\Box is eagerly evaluated.
		\Box is lazily evaluated.
		□ immediately computes and returns a result.
		□ does not immediately compute a result.
		\square always returns another RDD (once it's evaluated).
	,	After performing a series of transformations on an RDD, which of the following methods could you use to make sure those transformations are not repeated (e.g., on each iteration of an algorithm)?
		□ save
		\square persist
		\square memoize
		$\hfill\Box$ There is no such method because of the JVM's garbage collection mechanism.
	(d)	Why does Spark's RDD class not have a foldLeft method?
		□ foldLeft can only be performed on lists of Boolean values.
		☐ foldLeft doesn't work on immutable collections.
		\Box foldLeft is not stack-safe.
		☐ foldLeft is not fault-tolerant.
		□ foldLeft is not parallelizable.
	` '	Why is available in Spark's RDD class that overcomes the limitation of foldLeft mentioned in the previous part of this exercise?
		\square aggregate \square fold \square foldLeft \square join \square leftOuterJoin

5. Read the docs. Navigate to the Spark API documentation at
https://spark.apache.org/docs/3.3.1/api/scala/org/apache/spark/index.html
Enter "RDD" in the search box and select RDD from the results that appear on the left.
(a) Scroll down the resulting RDD API documentation page and find the cache() method. What does it say?
$\hfill\Box$ Persist this RDD with the default storage level (MEMORY_ONLY).
☐ Mark the RDD as non-persistent, and remove all blocks for it from memory and disk.
☐ Set this RDD's storage level to persist its values across operations after the first time it is computed.
$\hfill\Box$ Save this RDD as a Sequence File of serialized objects.
(b) On the RDD API doc page, find the version of persist that takes an argument: def persist(newLevel: StorageLevel). What does it say?
$\hfill\Box$ Persist this RDD with the default storage level (MEMORY_ONLY).
☐ Mark the RDD as non-persistent, and remove all blocks for it from memory and disk.
☐ Set this RDD's storage level to persist its values across operations after the first time it is computed.
$\hfill\Box$ Save this RDD as a Sequence File of serialized objects.
(c) On the RDD API doc page, find the unpersist method. What does it say?
□ Persist this RDD with the default storage level (MEMORY_ONLY).
☐ Mark the RDD as non-persistent, and remove all blocks for it from memory and disk.
☐ Set this RDD's storage level to persist its values across operations after the first time it is computed.
$\hfill\Box$ Save this RDD as a Sequence File of serialized objects.
(d) What's the difference between the sample and takeSample methods of the RDD class?
□ sample always uses a with-replacement sampling method, while takeSample always samples without replacement.
$\hfill\Box$ sample returns an RDD, while takeSample returns an Array.
☐ The second argument specifies the number of samples desired either as a fraction of the size of the RDD (sample) or as an absolute number (takeSample).
$\hfill\Box$ There is no difference; they are just two different names one can use to invoke the same function.