



Databricks

Exam Questions Databricks-Certified-Professional-Data-Engineer

Databricks Certified Data Engineer Professional Exam

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NEW QUESTION 1

A junior data engineer seeks to leverage Delta Lake's Change Data Feed functionality to create a Type 1 table representing all of the values that have ever been valid for all rows in a bronze table created with the property `delta.enableChangeDataFeed = true`. They plan to execute the following code as a daily job: Which statement describes the execution and results of running the above query multiple times?

- A. Each time the job is executed, newly updated records will be merged into the target table, overwriting previous values with the same primary keys.
- B. Each time the job is executed, the entire available history of inserted or updated records will be appended to the target table, resulting in many duplicate entries.
- C. Each time the job is executed, the target table will be overwritten using the entire history of inserted or updated records, giving the desired result.
- D. Each time the job is executed, the differences between the original and current versions are calculated; this may result in duplicate entries for some records.
- E. Each time the job is executed, only those records that have been inserted or updated since the last execution will be appended to the target table giving the desired result.

Answer: B

Explanation:

Reading table's changes, captured by CDF, using `spark.read` means that you are reading them as a static source. So, each time you run the query, all table's changes (starting from the specified `startingVersion`) will be read.

NEW QUESTION 2

Which of the following technologies can be used to identify key areas of text when parsing Spark Driver log4j output?

- A. Regex
- B. Julia
- C. `pyspark.ml.feature`
- D. Scala Datasets
- E. C++

Answer: A

Explanation:

Regex, or regular expressions, are a powerful way of matching patterns in text. They can be used to identify key areas of text when parsing Spark Driver log4j output, such as the log level, the timestamp, the thread name, the class name, the method name, and the message. Regex can be applied in various languages and frameworks, such as Scala, Python, Java, Spark SQL, and Databricks notebooks. References:

- ? <https://docs.databricks.com/notebooks/notebooks-use.html#use-regular-expressions>
- ? <https://docs.databricks.com/spark/latest/spark-sql/udf-scala.html#using-regular-expressions-in-udfs>
- ? https://docs.databricks.com/spark/latest/sparkr/functions/regexp_extract.html
- ? https://docs.databricks.com/spark/latest/sparkr/functions/regexp_replace.html

NEW QUESTION 3

A data engineer is configuring a pipeline that will potentially see late-arriving, duplicate records.

In addition to de-duplicating records within the batch, which of the following approaches allows the data engineer to deduplicate data against previously processed records as it is inserted into a Delta table?

- A. Set the configuration `delta.deduplicate = true`.
- B. VACUUM the Delta table after each batch completes.
- C. Perform an insert-only merge with a matching condition on a unique key.
- D. Perform a full outer join on a unique key and overwrite existing data.
- E. Rely on Delta Lake schema enforcement to prevent duplicate records.

Answer: C

Explanation:

To deduplicate data against previously processed records as it is inserted into a Delta table, you can use the merge operation with an insert-only clause. This allows you to insert new records that do not match any existing records based on a unique key, while ignoring duplicate records that match existing records. For example, you can use the following syntax:

```
MERGE INTO target_table USING source_table ON target_table.unique_key = source_table.unique_key WHEN NOT MATCHED THEN INSERT *
```

This will insert only the records from the source table that have a unique key that is not present in the target table, and skip the records that have a matching key.

This way, you can avoid inserting duplicate records into the Delta table.

References:

- ? <https://docs.databricks.com/delta/delta-update.html#upsert-into-a-table-using-merge>
- ? <https://docs.databricks.com/delta/delta-update.html#insert-only-merge>

NEW QUESTION 4

Which statement describes the default execution mode for Databricks Auto Loader?

- A. New files are identified by listing the input directory; new files are incrementally and idempotently loaded into the target Delta Lake table.
- B. Cloud vendor-specific queue storage and notification services are configured to track newly arriving files; new files are incrementally and idempotently into the target Delta Lake table.
- C. Webhook trigger Databricks job to run anytime new data arrives in a source directory; new data automatically merged into target tables using rules inferred from the data.
- D. New files are identified by listing the input directory; the target table is materialized by directory querying all valid files in the source directory.

Answer: A

Explanation:

Databricks Auto Loader simplifies and automates the process of loading data into Delta Lake. The default execution mode of the Auto Loader identifies new files by listing the input directory. It incrementally and idempotently loads these new files into the target Delta Lake table. This approach ensures that files are not missed and are processed exactly once, avoiding data duplication. The other options describe different mechanisms or integrations that are not part of the default

behavior of the Auto Loader.

References:

? Databricks Auto Loader Documentation: Auto Loader Guide

? Delta Lake and Auto Loader: Delta Lake Integration

NEW QUESTION 5

Incorporating unit tests into a PySpark application requires upfront attention to the design of your jobs, or a potentially significant refactoring of existing code. Which statement describes a main benefit that offset this additional effort?

- A. Improves the quality of your data
- B. Validates a complete use case of your application
- C. Troubleshooting is easier since all steps are isolated and tested individually
- D. Yields faster deployment and execution times
- E. Ensures that all steps interact correctly to achieve the desired end result

Answer: A

NEW QUESTION 6

A Delta Lake table was created with the below query:

Consider the following query:

```
DROP TABLE prod.sales_by_store -
```

If this statement is executed by a workspace admin, which result will occur?

- A. Nothing will occur until a COMMIT command is executed.
- B. The table will be removed from the catalog but the data will remain in storage.
- C. The table will be removed from the catalog and the data will be deleted.
- D. An error will occur because Delta Lake prevents the deletion of production data.
- E. Data will be marked as deleted but still recoverable with Time Travel.

Answer: C

Explanation:

When a table is dropped in Delta Lake, the table is removed from the catalog and the data is deleted. This is because Delta Lake is a transactional storage layer that provides ACID guarantees. When a table is dropped, the transaction log is updated to reflect the deletion of the table and the data is deleted from the underlying storage. References:

? <https://docs.databricks.com/delta/quick-start.html#drop-a-table>

? <https://docs.databricks.com/delta/delta-batch.html#drop-table>

NEW QUESTION 7

A Delta Lake table representing metadata about content posts from users has the following schema:

```
user_id LONG, post_text STRING, post_id STRING, longitude FLOAT, latitude FLOAT, post_time TIMESTAMP, date DATE
```

This table is partitioned by the date column. A query is run with the following filter: `longitude < 20 & longitude > -20`

Which statement describes how data will be filtered?

- A. Statistics in the Delta Log will be used to identify partitions that might include files in the filtered range.
- B. No file skipping will occur because the optimizer does not know the relationship between the partition column and the longitude.
- C. The Delta Engine will use row-level statistics in the transaction log to identify the files that meet the filter criteria.
- D. Statistics in the Delta Log will be used to identify data files that might include records in the filtered range.
- E. The Delta Engine will scan the parquet file footers to identify each row that meets the filter criteria.

Answer: D

Explanation:

This is the correct answer because it describes how data will be filtered when a query is run with the following filter: `longitude < 20 & longitude > -20`. The query is run on a Delta Lake table that has the following schema: `user_id LONG, post_text STRING, post_id STRING, longitude FLOAT, latitude FLOAT, post_time TIMESTAMP, date DATE`. This table is partitioned by the date column. When a query is run on a partitioned Delta Lake table, Delta Lake uses statistics in the Delta Log to identify data files that might include records in the filtered range. The statistics include information such as min and max values for each column in each data file. By using these statistics, Delta Lake can skip reading data files that do not match the filter condition, which can improve query performance and reduce I/O costs. Verified References: [Databricks Certified Data Engineer Professional], under "Delta Lake" section; Databricks Documentation, under "Data skipping" section.

NEW QUESTION 8

A developer has successfully configured credential for Databricks Repos and cloned a remote Git repository. They do not have privileges to make changes to the main branch, which is the only branch currently visible in their workspace.

Use Response to pull changes from the remote Git repository commit and push changes to a branch that appeared as a changes were pulled.

- A. Use Repos to merge all differences and make a pull request back to the remote repository.
- B. Use repos to merge all difference and make a pull request back to the remote repository.
- C. Use Repos to create a new branch commit all changes and push changes to the remote Git repository.
- D. Use repos to create a fork of the remote repository commit all changes and make a pull request on the source repository

Answer: C

Explanation:

In Databricks Repos, when a user does not have privileges to make changes directly to the main branch of a cloned remote Git repository, the recommended approach is to create a new branch within the Databricks workspace. The developer can then make changes in this new branch, commit those changes, and push the new branch to the remote Git repository. This workflow allows for isolated development without affecting the main branch, enabling the developer to propose changes via a pull request from the new branch to the main branch in the remote repository. This method adheres to common Git collaboration workflows, fostering code review and collaboration while ensuring the integrity of the main branch.

References:

? Databricks documentation on using Repos with Git: <https://docs.databricks.com/repos.html>

NEW QUESTION 9

A table named user_ltv is being used to create a view that will be used by data analysis on various teams. Users in the workspace are configured into groups, which are used for setting up data access using ACLs.

The user_ltv table has the following schema:

```
email STRING, age INT, ltv INT
```

The following view definition is executed:

```
CREATE VIEW user_ltv_no_minors AS
SELECT email, age, ltv
FROM user_ltv
WHERE
CASE
WHEN is_member('auditing') THEN TRUE
ELSE age >= 18
END
```

An analyze who is not a member of the auditing group executing the following query:

```
SELECT * FROM user_ltv_no_minors
```

Which result will be returned by this query?

- A. All columns will be displayed normally for those records that have an age greater than 18; records not meeting this condition will be omitted.
- B. All columns will be displayed normally for those records that have an age greater than 17; records not meeting this condition will be omitted.
- C. All age values less than 18 will be returned as null values all other columns will be returned with the values in user_ltv.
- D. All records from all columns will be displayed with the values in user_ltv.

Answer: A

Explanation:

Given the CASE statement in the view definition, the result set for a user not in the auditing group would be constrained by the ELSE condition, which filters out records based on age. Therefore, the view will return all columns normally for records with an age greater than 18, as users who are not in the auditing group will not satisfy the is_member('auditing') condition. Records not meeting the age > 18 condition will not be displayed.

NEW QUESTION 10

A data team's Structured Streaming job is configured to calculate running aggregates for item sales to update a downstream marketing dashboard. The marketing team has introduced a new field to track the number of times this promotion code is used for each item. A junior data engineer suggests updating the existing query as follows: Note that proposed changes are in bold.

Original query:

```
df.groupBy("item")
  .agg(count("item").alias("total_count"),
       mean("sale_price").alias("avg_price"))
  .writeStream
  .outputMode("complete")
  .option("checkpointLocation", "/item_agg/__checkpoint")
  .start("/item_agg")
```

Proposed query:

```
df.groupBy("item")
  .agg(count("item").alias("total_count"),
       mean("sale_price").alias("avg_price"),
       count("promo_code = 'NEW_MEMBER'").alias("new_member_promo"))
  .writeStream
  .outputMode("complete")
  .option('mergeSchema', 'true')
  .option("checkpointLocation", "/item_agg/__checkpoint")
  .start("/item_agg")
```

Which step must also be completed to put the proposed query into production?

- A. Increase the shuffle partitions to account for additional aggregates
- B. Specify a new checkpointLocation
- C. Run REFRESH TABLE delta, /item_agg'
- D. Remove .option (mergeSchema', true') from the streaming write

Answer: B

Explanation:

When introducing a new aggregation or a change in the logic of a Structured Streaming query, it is generally necessary to specify a new checkpoint location. This is because the checkpoint directory contains metadata about the offsets and the state of the aggregations of a streaming query. If the logic of the query changes, such as including a new aggregation field, the state information saved in the current checkpoint would not be compatible with the new logic, potentially leading to incorrect results or failures. Therefore, to accommodate the new field and ensure the streaming job has the correct starting point and state information for aggregations, a new checkpoint location should be specified. References:

? Databricks documentation on Structured Streaming:

<https://docs.databricks.com/spark/latest/structured-streaming/index.html>

? Databricks documentation on streaming checkpoints: <https://docs.databricks.com/spark/latest/structured-streaming/production.html#checkpointing>

NEW QUESTION 10

A Data engineer wants to run unit's tests using common Python testing frameworks on python functions defined across several Databricks notebooks currently used in production.

How can the data engineer run unit tests against function that work with data in production?

- A. Run unit tests against non-production data that closely mirrors production
- B. Define and unit test functions using Files in Repos
- C. Define units test and functions within the same notebook
- D. Define and import unit test functions from a separate Databricks notebook

Answer: A

Explanation:

The best practice for running unit tests on functions that interact with data is to use a dataset that closely mirrors the production data. This approach allows data engineers to validate the logic of their functions without the risk of affecting the actual production data. It's important to have a representative sample of production data to catch edge cases and ensure the functions will work correctly when used in a production environment.

References:

? Databricks Documentation on Testing: Testing and Validation of Data and Notebooks

NEW QUESTION 13

A Spark job is taking longer than expected. Using the Spark UI, a data engineer notes that the Min, Median, and Max Durations for tasks in a particular stage show the minimum and median time to complete a task as roughly the same, but the max duration for a task to be roughly 100 times as long as the minimum.

Which situation is causing increased duration of the overall job?

- A. Task queueing resulting from improper thread pool assignment.
- B. Spill resulting from attached volume storage being too small.
- C. Network latency due to some cluster nodes being in different regions from the source data
- D. Skew caused by more data being assigned to a subset of spark-partitions.
- E. Credential validation errors while pulling data from an external system.

Answer: D

Explanation:

This is the correct answer because skew is a common situation that causes increased duration of the overall job. Skew occurs when some partitions have more data than others, resulting in uneven distribution of work among tasks and executors. Skew can be caused by various factors, such as skewed data distribution, improper partitioning strategy, or join operations with skewed keys. Skew can lead to performance issues such as long-running tasks, wasted resources, or even task failures due to memory or disk spills. Verified References: [Databricks Certified Data Engineer Professional], under "Performance Tuning" section; Databricks Documentation, under "Skew" section.

NEW QUESTION 15

The data engineer is using Spark's MEMORY_ONLY storage level.

Which indicators should the data engineer look for in the spark UI's Storage tab to signal that a cached table is not performing optimally?

- A. Size on Disk is > 0
- B. The number of Cached Partitions > the number of Spark Partitions
- C. The RDD Block Name included the " annotation signaling failure to cache
- D. On Heap Memory Usage is within 75% of off Heap Memory usage

Answer: C

Explanation:

In the Spark UI's Storage tab, an indicator that a cached table is not performing optimally would be the presence of the `_disk` annotation in the RDD Block Name. This annotation indicates that some partitions of the cached data have been spilled to disk because there wasn't enough memory to hold them. This is suboptimal because accessing data from disk is much slower than from memory. The goal of caching is to keep data in memory for fast access, and a spill to disk means that this goal is not fully achieved.

NEW QUESTION 17

Assuming that the Databricks CLI has been installed and configured correctly, which Databricks CLI command can be used to upload a custom Python Wheel to object storage mounted with the DBFS for use with a production job?

- A. `configure`
- B. `fs`
- C. `jobs`
- D. `libraries`
- E. `workspace`

Answer: B

Explanation:

The `libraries` command group allows you to install, uninstall, and list libraries on Databricks clusters. You can use the `libraries install` command to install a custom Python Wheel on a cluster by specifying the `--whl` option and the path to the wheel file. For example, you can use the following command to install a custom Python Wheel named `mylib-0.1-py3-none-any.whl` on a cluster with the id `1234-567890-abcde123`:

```
databricks libraries install --cluster-id 1234-567890-abcde123 --whl dbfs:/mnt/mylib/mylib-0.1-py3-none-any.whl
```

This will upload the custom Python Wheel to the cluster and make it available for use with a production job. You can also use the `libraries uninstall` command to uninstall a library from a cluster, and the `libraries list` command to list the libraries installed on a cluster. References:

? Libraries CLI (legacy): <https://docs.databricks.com/en/archive/dev-tools/cli/libraries-cli.html>

? Library operations: <https://docs.databricks.com/en/dev-tools/cli/commands.html#library-operations>

? Install or update the Databricks CLI: <https://docs.databricks.com/en/dev-tools/cli/install.html>

NEW QUESTION 22

The DevOps team has configured a production workload as a collection of notebooks scheduled to run daily using the Jobs UI. A new data engineering hire is onboarding to the team and has requested access to one of these notebooks to review the production logic.

What are the maximum notebook permissions that can be granted to the user without allowing accidental changes to production code or data?

- A. Can manage
- B. Can edit
- C. Can run
- D. Can Read

Answer: D

Explanation:

Granting a user 'Can Read' permissions on a notebook within Databricks allows them to view the notebook's content without the ability to execute or edit it. This level of permission ensures that the new team member can review the production logic for learning or auditing purposes without the risk of altering the notebook's code or affecting production data and workflows. This approach aligns with best practices for maintaining security and integrity in production environments, where strict access controls are essential to prevent unintended modifications. References: Databricks documentation on access control and permissions for notebooks within the workspace (<https://docs.databricks.com/security/access-control/workspace-acl.html>).

NEW QUESTION 27

The data architect has mandated that all tables in the Lakehouse should be configured as external (also known as "unmanaged") Delta Lake tables.

Which approach will ensure that this requirement is met?

- A. When a database is being created, make sure that the LOCATION keyword is used.
- B. When configuring an external data warehouse for all table storage, leverage Databricks for all ELT.
- C. When data is saved to a table, make sure that a full file path is specified alongside the Delta format.
- D. When tables are created, make sure that the EXTERNAL keyword is used in the CREATE TABLE statement.
- E. When the workspace is being configured, make sure that external cloud object storage has been mounted.

Answer: D

Explanation:

To create an external or unmanaged Delta Lake table, you need to use the EXTERNAL keyword in the CREATE TABLE statement. This indicates that the table is not managed by the catalog and the data files are not deleted when the table is dropped. You also need to provide a LOCATION clause to specify the path where the data files are stored. For example:

```
CREATE EXTERNAL TABLE events ( date DATE, eventId STRING, eventType STRING, data STRING) USING DELTA LOCATION '/mnt/delta/events';
```

This creates an external Delta Lake table named events that references the data files in the '/mnt/delta/events' path. If you drop this table, the data files will remain intact and you can recreate the table with the same statement.

References:

? <https://docs.databricks.com/delta/delta-batch.html#create-a-table>

? <https://docs.databricks.com/delta/delta-batch.html#drop-a-table>

NEW QUESTION 30

Which statement regarding spark configuration on the Databricks platform is true?

- A. Spark configuration properties set for an interactive cluster with the Clusters UI will impact all notebooks attached to that cluster.
- B. When the same spark configuration property is set for an interactive to the same interactive cluster.
- C. Spark configuration set within a notebook will affect all SparkSession attached to the same interactive cluster
- D. The Databricks REST API can be used to modify the Spark configuration properties for an interactive cluster without interrupting jobs.

Answer: A

Explanation:

When Spark configuration properties are set for an interactive cluster using the Clusters UI in Databricks, those configurations are applied at the cluster level. This means that all notebooks attached to that cluster will inherit and be affected by these configurations. This approach ensures consistency across all executions within that cluster, as the Spark configuration properties dictate aspects such as memory allocation, number of executors, and other vital execution parameters. This centralized configuration management helps maintain standardized execution environments across different notebooks, aiding in debugging and performance optimization.

References:

? Databricks documentation on configuring clusters: <https://docs.databricks.com/clusters/configure.html>

NEW QUESTION 31

A member of the data engineering team has submitted a short notebook that they wish to schedule as part of a larger data pipeline. Assume that the commands provided below produce the logically correct results when run as presented.

```

Cmd 1
rawDF = spark.table("raw_data")

Cmd 2
rawDF.printSchema()

Cmd 3
flattenedDF = rawDF.select("?", "values.*")

Cmd 4
finalDF = flattenedDF.drop("values")

Cmd 5
display(finalDF)

Cmd 6

finalDF.write.mode("append").saveAsTable("flat_data")

```

Which command should be removed from the notebook before scheduling it as a job?

- A. Cmd 2
- B. Cmd 3
- C. Cmd 4
- D. Cmd 5
- E. Cmd 6

Answer: E

Explanation:

Cmd 6 is the command that should be removed from the notebook before scheduling it as a job. This command is selecting all the columns from the finalDF dataframe and displaying them in the notebook. This is not necessary for the job, as the finalDF dataframe is already written to a table in Cmd 7. Displaying the dataframe in the notebook will only consume resources and time, and it will not affect the output of the job. Therefore, Cmd 6 is redundant and should be removed.

The other commands are essential for the job, as they perform the following tasks:

? Cmd 1: Reads the raw_data table into a Spark dataframe called rawDF.

? Cmd 2: Prints the schema of the rawDF dataframe, which is useful for debugging and understanding the data structure.

? Cmd 3: Selects all the columns from the rawDF dataframe, as well as the nested columns from the values struct column, and creates a new dataframe called flattenedDF.

? Cmd 4: Drops the values column from the flattenedDF dataframe, as it is no longer needed after flattening, and creates a new dataframe called finalDF.

? Cmd 5: Explains the physical plan of the finalDF dataframe, which is useful for optimizing and tuning the performance of the job.

? Cmd 7: Writes the finalDF dataframe to a table called flat_data, using the append mode to add new data to the existing table.

NEW QUESTION 32

The data architect has decided that once data has been ingested from external sources into the Databricks Lakehouse, table access controls will be leveraged to manage permissions for all production tables and views. The following logic was executed to grant privileges for interactive queries on a production database to the core engineering group:
 GRANT USAGE ON DATABASE prod TO eng; GRANT SELECT ON DATABASE prod TO eng;

Assuming these are the only privileges that have been granted to the eng group and that these users are not workspace administrators, which statement describes their privileges?

- A. Group members have full permissions on the prod database and can also assign permissions to other users or groups.
- B. Group members are able to list all tables in the prod database but are not able to see the results of any queries on those tables.
- C. Group members are able to query and modify all tables and views in the prod database, but cannot create new tables or views.
- D. Group members are able to query all tables and views in the prod database, but cannot create or edit anything in the database.
- E. Group members are able to create, query, and modify all tables and views in the prod database, but cannot define custom functions.

Answer: D

Explanation:

The GRANT USAGE ON DATABASE prod TO eng command grants the eng group the permission to use the prod database, which means they can list and access the tables and views in the database. The GRANT SELECT ON DATABASE prod TO eng command grants the eng group the permission to select data from the tables and views in the prod database, which means they can query the data using SQL or DataFrame API. However, these commands do not grant the eng group any other permissions, such as creating, modifying, or deleting tables and views, or defining custom functions. Therefore, the eng group members are able to query all tables and views in the prod database, but cannot create or edit anything in the database. References:

? Grant privileges on a database: <https://docs.databricks.com/en/security/auth-authorization/table-acls/grant-privileges-database.html>

? Privileges you can grant on Hive metastore objects: <https://docs.databricks.com/en/security/auth-authorization/table-acls/privileges.html>

NEW QUESTION 37

A data pipeline uses Structured Streaming to ingest data from kafka to Delta Lake. Data is being stored in a bronze table, and includes the Kafka_generated timesamp, key, and value. Three months after the pipeline is deployed the data engineering team has noticed some latency issued during certain times of the day. A senior data engineer updates the Delta Table's schema and ingestion logic to include the current timestamp (as recoded by Apache Spark) as well the Kafka topic and partition. The team plans to use the additional metadata fields to diagnose the transient processing delays:

Which limitation will the team face while diagnosing this problem?

- A. New fields not be computed for historic records.
- B. Updating the table schema will invalidate the Delta transaction log metadata.

- C. Updating the table schema requires a default value provided for each file added.
- D. Spark cannot capture the topic partition fields from the kafka source.

Answer: A

Explanation:

When adding new fields to a Delta table's schema, these fields will not be retrospectively applied to historical records that were ingested before the schema change. Consequently, while the team can use the new metadata fields to investigate transient processing delays moving forward, they will be unable to apply this diagnostic approach to past data that lacks these fields.

References:

? Databricks documentation on Delta Lake schema management: <https://docs.databricks.com/delta/delta-batch.html#schema-management>

NEW QUESTION 39

A production cluster has 3 executor nodes and uses the same virtual machine type for the driver and executor.

When evaluating the Ganglia Metrics for this cluster, which indicator would signal a bottleneck caused by code executing on the driver?

- A. The five Minute Load Average remains consistent/flat
- B. Bytes Received never exceeds 80 million bytes per second
- C. Total Disk Space remains constant
- D. Network I/O never spikes
- E. Overall cluster CPU utilization is around 25%

Answer: E

Explanation:

This is the correct answer because it indicates a bottleneck caused by code executing on the driver. A bottleneck is a situation where the performance or capacity of a system is limited by a single component or resource. A bottleneck can cause slow execution, high latency, or low throughput. A production cluster has 3 executor nodes and uses the same virtual machine type for the driver and executor. When evaluating the Ganglia Metrics for this cluster, one can look for indicators that show how the cluster resources are being utilized, such as CPU, memory, disk, or network. If the overall cluster CPU utilization is around 25%, it means that only one out of the four nodes (driver + 3 executors) is using its full CPU capacity, while the other three nodes are idle or underutilized. This suggests that the code executing on the driver is taking too long or consuming too much CPU resources, preventing the executors from receiving tasks or data to process. This can happen when the code has driver-side operations that are not parallelized or distributed, such as collecting large amounts of data to the driver, performing complex calculations on the driver, or using non-Spark libraries on the driver. Verified References: [Databricks Certified Data Engineer Professional], under "Spark Core" section; Databricks Documentation, under "View cluster status and event logs - Ganglia metrics" section; Databricks Documentation, under "Avoid collecting large RDDs" section.

In a Spark cluster, the driver node is responsible for managing the execution of the Spark application, including scheduling tasks, managing the execution plan, and interacting with the cluster manager. If the overall cluster CPU utilization is low (e.g., around 25%), it may indicate that the driver node is not utilizing the available resources effectively and might be a bottleneck.

NEW QUESTION 42

The data engineering team maintains the following code:

```
accountDF = spark.table("accounts")
orderDF = spark.table("orders")
itemDF = spark.table("items")

orderWithItemDF = (orderDF.join(
    itemDF,
    orderDF.itemID == itemDF.itemID)
    .select(
        orderDF.accountID,
        orderDF.itemID,
        itemDF.itemName))

finalDF = (accountDF.join(
    orderWithItemDF,
    accountDF.accountID == orderWithItemDF.accountID)
    .select(
        orderWithItemDF["*"],
        accountDF.city))

(finalDF.write
    .mode("overwrite")
    .table("enriched_itemized_orders_by_account"))
```

Assuming that this code produces logically correct results and the data in the source tables has been de-duplicated and validated, which statement describes what will occur when this code is executed?

- A. A batch job will update the enriched_itemized_orders_by_account table, replacing only those rows that have different values than the current version of the table, using accountID as the primary key.
- B. The enriched_itemized_orders_by_account table will be overwritten using the current valid version of data in each of the three tables referenced in the join logic.
- C. An incremental job will leverage information in the state store to identify unjoined rows in the source tables and write these rows to the enriched_itemized_orders_by_account table.
- D. An incremental job will detect if new rows have been written to any of the source tables; if new rows are detected, all results will be recalculated and used to

overwrite the enriched_itemized_orders_by_account table.

E. No computation will occur until enriched_itemized_orders_by_account is queried; upon query materialization, results will be calculated using the current valid version of data in each of the three tables referenced in the join logic.

Answer: B

Explanation:

This is the correct answer because it describes what will occur when this code is executed. The code uses three Delta Lake tables as input sources: accounts, orders, and order_items. These tables are joined together using SQL queries to create a view called new_enriched_itemized_orders_by_account, which contains information about each order item and its associated account details. Then, the code uses write.format("delta").mode("overwrite") to overwrite a target table called enriched_itemized_orders_by_account using the data from the view. This means that every time this code is executed, it will replace all existing data in the target table with new data based on the current valid version of data in each of the three input tables. Verified References: [Databricks Certified Data Engineer Professional], under "Delta Lake" section; Databricks Documentation, under "Write to Delta tables" section.

NEW QUESTION 47

The data governance team is reviewing user for deleting records for compliance with GDPR. The following logic has been implemented to propagate deleted requests from the user_lookup table to the user_aggregate table.

```
(spark.read
  .format("delta")
  .option("readChangeData", True)
  .option("startingTimestamp", '2021-08-22 00:00:00')
  .option("endingTimestamp", '2021-08-29 00:00:00')
  .table("user_lookup")
  .createOrReplaceTempView("changes"))
```

```
spark.sql("""
DELETE FROM user_aggregates
WHERE user_id IN (
  SELECT user_id
  FROM changes
  WHERE _change_type='delete'
)
""")
```

Assuming that user_id is a unique identifying key and that all users have requested deletion have been removed from the user_lookup table, which statement describes whether successfully executing the above logic guarantees that the records to be deleted from the user_aggregates table are no longer accessible and why?

- A. No: files containing deleted records may still be accessible with time travel until a BACUM command is used to remove invalidated data files.
- B. Yes: Delta Lake ACID guarantees provide assurance that the DELETE command succeeded fully and permanently purged these records.
- C. No: the change data feed only tracks inserts and updates not deleted records.
- D. No: the Delta Lake DELETE command only provides ACID guarantees when combined with the MERGE INTO command

Answer: A

Explanation:

The DELETE operation in Delta Lake is ACID compliant, which means that once the operation is successful, the records are logically removed from the table. However, the underlying files that contained these records may still exist and be accessible via time travel to older versions of the table. To ensure that these records are physically removed and compliance with GDPR is maintained, a VACUUM command should be used to clean up these data files after a certain retention period. The VACUUM command will remove the files from the storage layer, and after this, the records will no longer be accessible.

NEW QUESTION 50

A small company based in the United States has recently contracted a consulting firm in India to implement several new data engineering pipelines to power artificial intelligence applications. All the company's data is stored in regional cloud storage in the United States.

The workspace administrator at the company is uncertain about where the Databricks workspace used by the contractors should be deployed.

Assuming that all data governance considerations are accounted for, which statement accurately informs this decision?

- A. Databricks runs HDFS on cloud volume storage; as such, cloud virtual machines must be deployed in the region where the data is stored.
- B. Databricks workspaces do not rely on any regional infrastructure; as such, the decision should be made based upon what is most convenient for the workspace administrator.
- C. Cross-region reads and writes can incur significant costs and latency; whenever possible, compute should be deployed in the same region the data is stored.
- D. Databricks leverages user workstations as the driver during interactive development; as such, users should always use a workspace deployed in a region they are physically near.
- E. Databricks notebooks send all executable code from the user's browser to virtual machines over the open internet; whenever possible, choosing a workspace region near the end users is the most secure.

Answer: C

Explanation:

This is the correct answer because it accurately informs this decision. The decision is about where the Databricks workspace used by the contractors should be deployed. The contractors are based in India, while all the company's data is stored in regional cloud storage in the United States. When choosing a region for deploying a Databricks workspace, one of the important factors to consider is the proximity to the data sources and sinks. Cross-region reads and writes can incur significant costs and latency due to network bandwidth and data transfer fees. Therefore, whenever possible, compute should be deployed in the same region the data is stored to optimize performance and reduce costs. Verified References: [Databricks Certified Data Engineer Professional], under "Databricks Workspace"

section; Databricks Documentation, under "Choose a region" section.

NEW QUESTION 52

The view updates represents an incremental batch of all newly ingested data to be inserted or updated in the customers table. The following logic is used to process these records.

```
MERGE INTO customers USING (  
SELECT updates.customer_id as merge_key, updates .* FROM updates  
UNION ALL  
SELECT NULL as merge_key, updates .* FROM updates JOIN customers  
ON updates.customer_id = customers.customer_id  
WHERE customers.current = true AND updates.address <> customers.address  
) staged_updates  
ON customers.customer_id = mergekey  
WHEN MATCHED AND customers.current = true AND customers.address <> staged_updates.address THEN  
UPDATE SET current = false, end_date = staged_updates.effective_date  
WHEN NOT MATCHED THEN  
INSERT (customer_id, address, current, effective_date, end_date)  
VALUES (staged_updates.customer_id, staged_updates.address, true, staged_updates.effective_date, null)
```

Which statement describes this implementation?

- A. The customers table is implemented as a Type 2 table; old values are overwritten and new customers are appended.
- B. The customers table is implemented as a Type 1 table; old values are overwritten by new values and no history is maintained.
- C. The customers table is implemented as a Type 2 table; old values are maintained but marked as no longer current and new values are inserted.
- D. The customers table is implemented as a Type 0 table; all writes are append only with no changes to existing values.

Answer: C

Explanation:

The provided MERGE statement is a classic implementation of a Type 2 SCD in a data warehousing context. In this approach, historical data is preserved by keeping old records (marking them as not current) and adding new records for changes. Specifically, when a match is found and there's a change in the address, the existing record in the customers table is updated to mark it as no longer current (current = false), and an end date is assigned (end_date = staged_updates.effective_date). A new record for the customer is then inserted with the updated information, marked as current. This method ensures that the full history of changes to customer information is maintained in the table, allowing for time-based analysis of customer data. References: Databricks documentation on implementing SCDs using Delta Lake and the MERGE statement (<https://docs.databricks.com/delta/delta-update.html#upsert-into-a-table-using-merge>).

NEW QUESTION 56

All records from an Apache Kafka producer are being ingested into a single Delta Lake table with the following schema:
key BINARY, value BINARY, topic STRING, partition LONG, offset LONG, timestamp LONG

There are 5 unique topics being ingested. Only the "registration" topic contains Personal Identifiable Information (PII). The company wishes to restrict access to PII. The company also wishes to only retain records containing PII in this table for 14 days after initial ingestion. However, for non-PII information, it would like to retain these records indefinitely.

Which of the following solutions meets the requirements?

- A. All data should be deleted biweekly; Delta Lake's time travel functionality should be leveraged to maintain a history of non-PII information.
- B. Data should be partitioned by the registration field, allowing ACLs and delete statements to be set for the PII directory.
- C. Because the value field is stored as binary data, this information is not considered PII and no special precautions should be taken.
- D. Separate object storage containers should be specified based on the partition field, allowing isolation at the storage level.
- E. Data should be partitioned by the topic field, allowing ACLs and delete statements to leverage partition boundaries.

Answer: B

Explanation:

Partitioning the data by the topic field allows the company to apply different access control policies and retention policies for different topics. For example, the company can use the Table Access Control feature to grant or revoke permissions to the registration topic based on user roles or groups. The company can also use the DELETE command to remove records from the registration topic that are older than 14 days, while keeping the records from other topics indefinitely.

Partitioning by the topic field also improves the performance of queries that filter by the topic field, as they can skip reading irrelevant partitions. References:

? Table Access Control: <https://docs.databricks.com/security/access-control/table-acls/index.html>

? DELETE: <https://docs.databricks.com/delta/delta-update.html#delete-from-a-table>

NEW QUESTION 59

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