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Snowflake SnowPro Advanced: Data Scientist Certification Exam Sample Questions (Q70-Q75):

NEW QUESTION # 70

You are tasked with developing a Snowpark Python function to identify and remove near-duplicate text entries from a table named 'PRODUCT DESCRIPTIONS'. The table contains a 'PRODUCT ID' (NUMBER) and 'DESCRIPTION' (STRING) column. Near duplicates are defined as descriptions with a Jaccard similarity score greater than 0.9. You need to implement this using Snowpark and UDFs. Which of the following approaches is most efficient, secure, and correct to implement?

- A. Define a Python UDF that calculates the Jaccard similarity. Create a new table, 'PRODUCT DESCRIPTIONS NO DUPES', and insert the distinct descriptions based on the similarity score. Rows in the original table with similar product description must be inserted with lowest product id into new table.
- B. Define a Python UDF to calculate Jaccard similarity. Create a temporary table with a ROW_NUMBER() column partitioned by a hash of the DESCRIPTION column. Calculate the Jaccard similarity between descriptions within each partition. Filter and remove near duplicates based on a tie-breaker (smallest PRODUCT_ID).**
- C. Use the function directly in a SQL query without a UDF. Partition the data by 'PRODUCT_ID' and remove near duplicates where the approximate Jaccard index is above 0.9.
- D. Define a Python UDF that calculates the Jaccard similarity between all pairs of descriptions in the table. Use a cross join to compare all rows, then filter based on the Jaccard similarity threshold. Finally, delete the near-duplicate rows based on a chosen tie-breaker (e.g., smallest PRODUCT_ID).
- E. Define a Python UDF that calculates the Jaccard similarity. Use 'GROUP BY' to group descriptions by the 'PRODUCT ID'. Apply the UDF on this grouped data to remove duplicates with similarity score greater than threshold.

Answer: B

Explanation:

Option D is the most efficient, secure, and correct approach for removing near-duplicate text entries using Snowpark and UDFs. It correctly addresses both the computational complexity and the security implications of the task. - It creates a temporary table because we are doing operations of delete and create a table which is best done via temporary table. - It uses bucketing (hashing descriptions) to reduce the number of comparisons. This significantly improves performance compared to comparing all possible pairs of descriptions which is what options A and B do. - Use ROW_NUMBER() to flag duplicate for deletion with threshold. Option A is not optimal due to the complexity of cross join. Option B is incorrect because there is data and functionality that is lost with the insertion of distinct entries based on score. Also, it would be inefficient as it required re-evaluation of score on insertion. Option C is incorrect because Grouping by Product ID will not allow for similarity calculation across different product IDs. Option E is not applicable because Snowflake does not have a built-in 'APPROX JACCARD INDEX' function to apply directly in a SQL query.

NEW QUESTION # 71

You are using Snowflake Cortex to analyze customer reviews. You have created a vector embedding for each review using a UDF that calls a remote LLM inference endpoint. Now you need to perform a similarity search to identify reviews that are similar to a given query review. Which of the following SQL queries leveraging vector functions in Snowflake is the MOST efficient and appropriate way to achieve this, assuming the 'REVIEW_EMBEDDINGS' table has columns 'review_id' and 'embedding' (a VECTOR column) and 'query_embedding' is a pre-computed vector embedding?

- ☐ SELECT review_id FROM REVIEW_EMBEDDINGS ORDER BY embedding <=> query_embedding LIMIT 10;
- ☐ SELECT review_id FROM REVIEW_EMBEDDINGS WHERE ARRAY_CONTAINS(embedding, query_embedding) LIMIT 10;
- ☐ SELECT review_id FROM REVIEW_EMBEDDINGS QUALIFY ROW_NUMBER() OVER (ORDER BY VECTOR_COSINE_SIMILARITY(embedding, query_embedding) DESC) <= 10;
- ☐ SELECT review_id FROM REVIEW_EMBEDDINGS WHERE VECTOR_L2_DISTANCE(embedding, query_embedding) < 0.5 LIMIT 10;
- ☐ SELECT review_id FROM REVIEW_EMBEDDINGS ORDER BY VECTOR_INNER_PRODUCT(embedding, query_embedding) DESC LIMIT 10;

- A. Option D
- B. Option C
- C. Option A
- D. Option B
- E. Option E**

Answer: E

Explanation:

The most efficient and accurate way to perform a similarity search with vector embeddings is using ordered in descending order

because inner product is the fastest of the vector functions and still gets the vector similarity score. The operator performs an exact match which doesn't consider vector similarity (A). is for array data, not vectors (B). 'QUALIFY' and 'VECTOR COSINE SIMILARITY' works but isn't optimal (C), and L2 distance require some value/threshold to compare. 'ORDER BY ... LIMIT' is efficient with the inner product, it's very fast (E).

NEW QUESTION # 72

You're developing a Python UDTF in Snowflake to perform sentiment analysis on customer reviews. The UDTF uses a pre-trained transformer model from Hugging Face. The code is as follows:

```
import snowflake.snowpark.functions as sf
from snowflake.snowpark.udtf import UDTF
from transformers import pipeline

class SentimentAnalyzer(UDTF):
    def __init__(self):
        self.classifier = pipeline("sentiment-analysis")

    def process(self, text: str):
        result = self.classifier(text)[0]
        yield (result['label'], result['score'])

sentiment_udtf = SentimentAnalyzer(input_types=[sf.StringType()], return_type=sf.StructType([sf.StringType(), sf.FloatType()]))

add_permanent = sf.udtf.register(func=sentiment_udtf,
                                return_type=sf.StructType([sf.StringType(), sf.FloatType()]),
                                input_types=[sf.StringType()],
                                name='SENTIMENT_ANALYZER_UDTF',
                                replace=True,
                                is_permanent=True,
                                stage_location='@my_stage',
                                imports=['/tmp/transformers', '/tmp/torch', '/tmp/tokenizers'])
```



When deploying this UDTF, you encounter a 'ModuleNotFoundError: No module named 'transformers'' error. Considering best practices for managing dependencies in Snowflake UDTFs, what is the most effective way to resolve this issue?

- A. Use the 'snowflake-ml-python' library and its dependency management features to automatically resolve and deploy the 'transformers' dependency.
- B. Install the 'transformers' library directly on the Snowflake compute nodes using Snowpark's 'add_packageS' method at the session level.
- C. Upload all the dependencies of Transformers (manually downloaded libraries) to the internal stage.
- D. Include the 'transformers' library in the same Python file as the UDTF definition. This is acceptable for smaller libraries.
- E. Create a Conda environment containing the 'transformers' library, package it into a zip file, upload it to a Snowflake stage, and specify the stage path in the 'imports' parameter when registering the UDTF.

Answer: E

Explanation:

Option B is the recommended approach for managing dependencies like 'transformers' in Snowflake UDTFs. Creating a Conda environment ensures that all required libraries and their dependencies are packaged together, preventing version conflicts and ensuring reproducibility. Uploading the environment to a stage and specifying it in the 'imports' parameter makes the dependencies available to the UDTF during execution. Option A is incorrect because Snowpark's 'add_packageS' is the ideal way for adding packages. Option C is impractical for large libraries like 'transformers'. Option D, although using snowflake-ml-python is valid, manually creating conda environment will reduce the dependency on other services. Option E is very tedious.

NEW QUESTION # 73

You are tasked with identifying fraudulent transactions in a large financial dataset stored in Snowflake using unsupervised learning. The dataset contains features like transaction amount, merchant ID, location, time, and user ID. You decide to use a combination of clustering and anomaly detection techniques. Which of the following steps and techniques would be MOST effective in achieving this goal while leveraging Snowflake's capabilities and minimizing false positives?

- A. Use a Snowflake Python UDF to perform feature selection, apply a combination of K-means clustering and anomaly detection techniques like Isolation Forest or Local Outlier Factor (LOF), and then score each transaction based on its likelihood of being fraudulent. Tune parameters and use a hold-out validation set to minimize false positives, using a Snowpark DataFrame to retrieve the data.
- B. Perform K-means clustering on the entire dataset using all available features, then flag any transaction that falls outside of any cluster as fraudulent. Ignore any feature selection or engineering to simplify the process.

- C. Apply Principal Component Analysis (PCA) for dimensionality reduction, then use DBSCAN clustering to identify dense regions of normal transactions and flag any transaction that is not within a dense region as potentially fraudulent. After, review the anomalous data points.
- D. Use only the 'transaction amount' feature and perform histogram-based anomaly detection in Snowflake SQL by identifying values outside of the common ranges, disregarding other potentially relevant information.
- E. Implement an Isolation Forest algorithm directly in SQL using complex JOINS and window functions to identify anomalies based on transaction volume and velocity.

Answer: A,C

Explanation:

Option B leverages PCA for dimensionality reduction, improving clustering performance and reducing noise, followed by DBSCAN, which is effective at identifying outliers. Option D provides a comprehensive approach utilizing feature engineering, a combination of clustering and anomaly detection techniques implemented via Python UDF within Snowflake, and proper validation to minimize false positives. These approaches address data preprocessing, algorithm selection, and model evaluation for effective fraud detection.

Option A lacks feature selection/engineering and may lead to poor clustering. Option C is inefficient and impractical. Option E is too simplistic and ignores crucial information.

NEW QUESTION # 74

You are building a model to predict loan defaults using a dataset stored in Snowflake. After training your model and calculating residuals, you create a scatter plot of the residuals against the predicted values. The plot shows a cone-shaped pattern, with residuals spreading out more as the predicted values increase. Which of the following SQL queries, run within a Snowpark Python session, could be used to address the underlying issue indicated by this residual pattern, assuming the predicted values are stored in a column named and the residuals in a column named 'loan_default_residuar' in a Snowflake table named 'loan_predictions'?

- A.

```
SELECT predicted_loan_default_prob, loan_default_residual FROM loan_predictions QUALIFY ROW_NUMBER() OVER (ORDER BY loan_default_residual DESC)
<= 0.95 COUNT( ) OVER();
```

- B.

```
SELECT predicted_loan_default_prob, LN(loan_default_residual) FROM loan_predictions;
```

- C.

```
SELECT predicted_loan_default_prob, loan_default_residual FROM loan_predictions WHERE ABS(loan_default_residual) < 3 * STDDEV(loan_default_residual);
```

- D.

```
SELECT predicted_loan_default_prob, loan_default_residual FROM loan_predictions;
```

- E.

```
SELECT predicted_loan_default_prob, loan_default_residual FROM loan_predictions;
```

Answer: E

Explanation:

A cone-shaped pattern in the residuals plot (heteroscedasticity) indicates that the variance of the errors is not constant. Applying a transformation like Box-Cox to the target variable before retraining the model (Option D) is the most appropriate way to address this. Option A attempts to filter outliers based on the residuals, but does not address the heteroscedasticity itself and requires statistical functions unavailable within standard SQL. Option B attempts to take the natural log of the residuals, which is nonsensical as residuals can be negative. Option C attempts to filter based on the rank of residuals, which is similarly unhelpful, does not fix the problem, and uses inappropriate outlier removal with SQL QUALIFY clause. Option E scaling the features might sometimes improve model performance, but it does not directly address heteroscedasticity.

NEW QUESTION # 75

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