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## Snowflake SnowPro Advanced: Data Scientist Certification Exam Sample

## Questions (Q61-Q66):

### NEW QUESTION # 61

A data scientist is performing exploratory data analysis on a table named 'CUSTOMER TRANSACTIONS'. They need to calculate the standard deviation of transaction amounts (TRANSACTION\_AMOUNT) for different customer segments (CUSTOMER\_SEGMENT). The 'CUSTOMER SEGMENT' column can contain NULL values. Which of the following SQL statements will correctly compute the standard deviation, excluding NULL transaction amounts, and handling NULL customer segments by treating them as a separate segment called 'Unknown'? Consider using Snowflake-specific functions where appropriate.

- `SELECT IFNULL(CUSTOMER_SEGMENT, 'Unknown') AS CUSTOMER_SEGMENT, STDDEV(TRANSACTION_AMOUNT) AS STD_DEV FROM CUSTOMER_TRANSACTIONS WHERE TRANSACTION_AMOUNT IS NOT NULL GROUP BY CUSTOMER_SEGMENT;`
- `SELECT NVL(CUSTOMER_SEGMENT, 'Unknown') AS CUSTOMER_SEGMENT, STDDEV_SAMP(TRANSACTION_AMOUNT) AS STD_DEV FROM CUSTOMER_TRANSACTIONS WHERE TRANSACTION_AMOUNT IS NOT NULL GROUP BY CUSTOMER_SEGMENT;`
- `SELECT COALESCE(CUSTOMER_SEGMENT, 'Unknown') AS CUSTOMER_SEGMENT, STDDEV_POP(TRANSACTION_AMOUNT) AS STD_DEV FROM CUSTOMER_TRANSACTIONS WHERE TRANSACTION_AMOUNT IS NOT NULL GROUP BY CUSTOMER_SEGMENT;`
- `SELECT CASE WHEN CUSTOMER_SEGMENT IS NULL THEN 'Unknown' ELSE CUSTOMER_SEGMENT END AS CUSTOMER_SEGMENT, STDDEV(TRANSACTION_AMOUNT) AS STD_DEV FROM CUSTOMER_TRANSACTIONS WHERE TRANSACTION_AMOUNT IS NOT NULL GROUP BY CUSTOMER_SEGMENT;`
- `SELECT COALESCE(CUSTOMER_SEGMENT, 'Unknown') AS CUSTOMER_SEGMENT, VARIANCE(TRANSACTION_AMOUNT) AS STD_DEV FROM CUSTOMER_TRANSACTIONS WHERE TRANSACTION_AMOUNT IS NOT NULL GROUP BY CUSTOMER_SEGMENT;`

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

**Answer: A,C**

Explanation:

Options B and C correctly calculates the standard deviation. Option B utilizes 'NVL', which is the equivalent of 'COALESCE' or 'IFNULL', to handle NULL Customer Segment values, and 'STDDEV\_SAMP' for sample standard deviation, which is generally the correct function to use when dealing with a sample of the entire population. Option C also uses 'COALESCE' and utilizes the 'STDDEV POP' function, which returns the population standard deviation, assuming the data represents the whole population. Option A uses IFNULL, which works, and STDDEV, which is an alias for either STDDEV\_SAMP or STDDEV\_POP. The exact behavior will depend on session variable setting. Option D also uses 'CASE WHEN' construct which works to identify Unknown segments. STDDEV is again aliased. Option E calculates the variance and not Standard deviation.

### NEW QUESTION # 62

You are tasked with identifying fraudulent transactions from unstructured log data stored in Snowflake. The logs contain various fields, including timestamps, user IDs, and transaction details embedded within free-text descriptions. You plan to use a supervised learning approach, having labeled a subset of transactions as 'fraudulent' or 'not fraudulent.' Which of the following methods best describes the extraction and processing of this data for training a machine learning model within Snowflake?

- A. Extract the entire log description field and train a word embedding model (e.g., Word2Vec) on the entire dataset. Average the word vectors for each transaction's log description to create a document vector. Train a classification model (e.g., Random Forest) on these document vectors within Snowflake.
- B. Use regular expressions within a Snowflake UDF to extract relevant information (e.g., amount, item description) from the log descriptions. Convert extracted data into numerical features using one-hot encoding within the UDF. Then, train a model using the extracted numerical features directly within Snowflake using SQL extensions for machine learning.
- C. Treat the unstructured log description as a categorical feature and directly apply one-hot encoding within Snowflake, then train a classification model. Due to high dimensionality perform PCA for dimensionality reduction before training.
- D. Export the entire log data to an external machine learning platform (e.g., AWS SageMaker) and perform feature extraction, NLP processing, and model training there. Import the trained model back into Snowflake as a UDF for prediction.
- E. Use a combination of regular expressions and natural language processing (NLP) techniques within Snowflake UDFs to extract key features such as transaction amounts, product categories, and sentiment scores from the log descriptions. Then, combine these extracted features with other structured data (e.g., user demographics) and train a classification model using these features. The NLP steps include tokenization, stop word removal, and TF-IDF vectorization.

**Answer: E**

Explanation:

Option C provides the most comprehensive and effective approach. It combines the strengths of both regular expressions (for structured data extraction) and NLP techniques (for understanding the semantic content of the log descriptions). Using Snowflake UDFs keeps the data processing within Snowflake, minimizing data movement. Combining extracted features with other structured data enhances the model's performance.

### NEW QUESTION # 63

You are developing a churn prediction model using Snowpark Python and Scikit-learn. After initial model training, you observe significant overfitting. Which of the following hyperparameter tuning strategies and code snippets, when implemented within a Snowflake Python UDF, would be MOST effective to address overfitting in a Ridge Regression model and how can you implement a reproducible model with minimal code?

- Using 'GridSearchCV' with a wide range of 'alpha' values, without cross-validation, and then selecting the 'alpha' that gives the highest score on the training data. 

```
from sklearn.linear_model import Ridge; from sklearn.model_selection import GridSearchCV; param_grid = {'alpha': [0.001, 0.01, 0.1, 1, 10, 100]}; grid = GridSearchCV(Ridge(), param_grid, cv=None); grid.fit(X_train, y_train); best_alpha = grid.best_params_['alpha']
```
- Using 'RandomizedSearchCV' with a limited number of iterations and a fixed random state, along with cross-validation, and then selecting the 'alpha' that gives the highest average cross-validation score. 

```
from sklearn.linear_model import Ridge; from sklearn.model_selection import RandomizedSearchCV; from scipy.stats import loguniform; param_distributions = {'alpha': loguniform(1e-5, 100)}; rsearch = RandomizedSearchCV(Ridge(), param_distributions, n_iter=10, cv=3, random_state=42); rsearch.fit(X_train, y_train); best_alpha = rsearch.best_params_['alpha']
```
- Manually tuning the 'alpha' parameter by trial and error on the training data, without cross-validation or a structured search. 

```
from sklearn.linear_model import Ridge; alphas = [0.001, 0.01, 0.1, 1, 10, 100]; best_alpha = None; best_score = float('inf'); for alpha in alphas: model = Ridge(alpha=alpha); model.fit(X_train, y_train); score = model.score(X_train, y_train); if score > best_score: best_score = score; best_alpha = alpha
```
- Using 'BayesianSearchCV' with Gaussian Processes and acquisition function optimization and a cross validation with n\_jobs=-1. 

```
from sklearn.linear_model import Ridge; from skopt import BayesSearchCV; from skopt.space import Real; search_space = {'alpha': Real(1e-5, 100, prior='log-uniform')}; opt = BayesSearchCV(Ridge(), search_space, n_trials=5, cv=3, n_jobs=-1, random_state=42); opt.fit(X_train, y_train); best_alpha = opt.best_params_['alpha']
```
- Using 'HalvingGridSearchCV' with successive halving and resource allocation, without random state specification for complete reproducibility. 

```
from sklearn.experimental import enable_halving_search_cv # noqa; from sklearn.model_selection import HalvingGridSearchCV; from sklearn.linear_model import Ridge; param_grid = {'alpha': [0.001, 0.01, 0.1, 1, 10, 100]}; hsearch = HalvingGridSearchCV(Ridge(), param_grid, cv=3, resource='n_samples', max_resources=len(X_train), factor=3).fit(X_train, y_train); best_alpha = hsearch.best_params_['alpha']
```

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

### Answer: A,B

Explanation:

Options B and D are correct because they employ techniques to mitigate overfitting. Option B uses 'RandomizedSearchCV' with cross-validation and a fixed 'random\_state', making the search reproducible and preventing overfitting by evaluating performance on multiple validation sets. Option D leverages 'BayesianSearchCV', which uses a probabilistic model to efficiently explore the hyperparameter space, also with cross-validation and a fixed random state making search reproducible. Both methods aim to find a balance between model complexity and generalization ability. Option A is incorrect because it does not use cross-validation, which is crucial for preventing overfitting. Option C is incorrect because manual tuning without a systematic search and cross-validation is prone to bias and overfitting. Finally, option E is incorrect because while using a modern algorithm, it lacks a random state, making it difficult to reproduce the outcome.

### NEW QUESTION # 64

You have trained a linear regression model in Snowpark ML to predict house prices. After training, you want to assess the overall feature importance using the model's coefficients. Consider the following Snowflake table containing the coefficients:

```
CREATE OR REPLACE TABLE model_coefficients ( feature_name VARCHAR(255), coefficient FLOAT ); INSERT INTO model_coefficients (feature_name, coefficient) VALUES ('sqft', 120.5), ('bedrooms', -30000.0), ('location_score', 80000.0), ('age', -500.0);
```

Which of the following statements are correct interpretations of these coefficients regarding feature impact?

- A. An increase of one square foot (sqft) in house size is associated with an increase of \$120.5 in the predicted house price.
- B. Increasing the number of bedrooms is associated with a decrease in the predicted house price.
- C. The 'location\_score' feature is the most influential predictor in determining house price.
- D. The 'bedrooms' feature has a positive impact on the house price since the coefficient is negative.
- E. The 'age' feature has an insignificant impact because its coefficient is small.

### Answer: A,B,C

### Explanation:

Option A is correct because a positive coefficient for 'sqft' indicates a positive relationship with the target variable (house price). Option C is correct because 'location\_score' has the largest absolute coefficient value. Option E is correct because a negative coefficient for 'bedrooms' indicates an inverse relationship. Option B is incorrect because the negative sign implies a negative impact. Option D is incorrect as the scale of features isn't normalized here, so we cannot conclude about the significance of the 'age' feature based on the magnitude of its coefficient alone without knowing its standard deviation and the standard deviation of other features. We would need to compute the z-score, t-score, or p-value for each coefficient to truly assess significance.

### NEW QUESTION # 65

You are building an automated model retraining pipeline for a sales forecasting model in Snowflake using Snowflake Tasks and Stored Procedures. After retraining, you want to validate the new model against a champion model already deployed. You need to define a validation strategy using the following models: champion model deployed as UDF 'FORECAST UDF', and contender model deployed as UDF 'FORECAST UDF NEW'. Given the following objectives: (1) Minimal impact on production latency, (2) Ability to compare predictions on a large volume of real-time data, (3) A statistically sound comparison metric. Which of the following SQL statements best represents how to efficiently compare the forecasts of the two models on a sample dataset and calculate the Root Mean Squared Error (RMSE) to validate the new model?

- A.

```
CREATE OR REPLACE TEMPORARY TABLE model_comparison AS
SELECT
    sales_data.actual_sales,
    FORECAST_UDF(feature1, feature2) AS champion_forecast,
    FORECAST_UDF_NEW(feature1, feature2) AS challenger_forecast
FROM sales_data
SAMPLE BERNoulli(10);
SELECT
    SQRT(AVG(POWER(actual_sales - challenger_forecast, 2))) AS challenger_rmse,
    SQRT(AVG(POWER(actual_sales - champion_forecast, 2))) AS champion_rmse
FROM model_comparison;
```

```
CREATE OR REPLACE TEMPORARY TABLE model_comparison AS
SELECT
    sales_data.actual_sales,
    FORECAST_UDF(feature1, feature2) AS champion_forecast,
    FORECAST_UDF_NEW(feature1, feature2) AS challenger_forecast
FROM sales_data
LIMIT 1000;

SELECT
    SQRT(AVG(POWER(champion_forecast - challenger_forecast, 2)))
FROM model_comparison;
```

- B.
- C.

```

CREATE OR REPLACE TEMPORARY TABLE model_comparison AS
SELECT
    sales_data.actual_sales,
    FORECAST_UDF(feature1, feature2) AS champion_forecast,
    FORECAST_UDF_NEW(feature1, feature2) AS challenger_forecast
FROM sales_data
SAMPLE BERNoulli(1);

SELECT
    SQRT(AVG(POWER(actual_sales - challenger_forecast, 2))) AS challenger_rmse,
    SQRT(AVG(POWER(actual_sales - champion_forecast, 2))) AS champion_rmse,
    challenger_rmse - champion_rmse AS rmse_diff
FROM model_comparison;

```



```
CREATE OR REPLACE TEMPORARY TABLE model_comparison AS
```

```
ELECT
```

```

    sales_data.actual_sales,
    FORECAST_UDF(feature1, feature2) AS champion_forecast,
    FORECAST_UDF_NEW(feature1, feature2) AS challenger_forecast
FROM sales_data
SAMPLE BERNoulli(10);

```



```
ELECT
```

```

    SQRT(AVG(POWER(actual_sales - challenger_forecast, 2))) AS challenger_rmse
    SQRT(AVG(POWER(actual_sales - champion_forecast, 2))) AS champion_rmse,
    FROM model_comparison

```

- D. HRF challenge rmse < champion rmse;
- E.

```

CREATE OR REPLACE TEMPORARY TABLE model_comparison AS
SELECT
    AVG(POWER(FORECAST_UDF(feature1, feature2) - FORECAST_UDF_NEW(feature1, feature2), 2)) AS mse,
    SQRT(mse) AS rmse
FROM sales_data;

SELECT rmse FROM model_comparison;

```

**Answer: A**

Explanation:

Option E is the best approach. It samples the data using 'SAMPLE BERNoulli(10)' for minimal impact on production. Then, it calculates both the challenger RMSE (new model) and the champion RMSE on this sample data. This provides a direct comparison of the model performance against actual sales and also allows to minimise runtime to compute this metric compared to option C which computes a difference without evaluating if the new model has a better score. Sampling helps with minimal impact while comparison metric in this case needs the actual\_sales column. This provides a statistically relevant comparison within Snowflake, minimizing external processing. Option A does not compare the model to the ground truth (actual sales). Option B only compares the challenger and champion models' predictions against each other on a small, limited dataset (1000 records), which may not be representative. Options C calculates the RMSE difference directly and has a SAMPLE size of 1, which is unlikely to reflect the reality and Option D filters based on RMSE, which makes the approach biased and makes it harder to evaluate if the RMSE is statistically significant.

**NEW QUESTION # 66**

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