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Assignment 20: Handling Missing Data in ETL

SECTION A – THEORETICAL QUESTIONS

Q1. What are the most common reasons for missing data in ETL pipelines?

Answer:

Common Reasons for Missing Data in ETL Pipelines:

1. Data entry errors
2. Users skipping optional fields
3. System integration issues
4. Sensor/device failure
5. Data corruption during transfer
6. Different data sources having different formats

Q2. Why is blindly deleting rows with missing values considered a bad practice in ETL?

Answer: Because when we do blindly deleting rows with missing values it leads to:

- Reduces dataset size

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- May remove important customers
- Can create biased results
- Reduces accuracy of analytics

Q3.Explain the difference between: Listwise deletion Column deletion Also mention one scenario where each is appropriate

Answer:

Listwise Deletion: Removes entire rows where any value is missing.

Appropriate when missing values are very few and random.

Column Deletion: Removes entire column if most values are missing.

Appropriate when 70–80% data is missing and column is not important.

Q4 .Why is median imputation preferred over mean imputation for skewed data such as income?

Answer:

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We use median imputation preferred over mean imputation for skewed data such as income because:

- Income data is usually right-skewed.
- Mean is affected by extreme values.
- Median is not affected by outliers and gives realistic central value.

Q5. What is forward fill and in what type of dataset is it most useful?

Answer:

Forward Fill:

Forward fill replaces missing value with the previous available value.

Most useful for time-series, sales data, stock prices, and continuous business records.

Q6. Why should flagging missing values be done before imputation in an ETL workflow?

Answer: flagging missing values be done before imputation in an ETL workflow because:

- Missing data itself gives information.

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- After filling, we lose that information.
- Flag helps track original missing entries.

Q7. Consider a scenario where income is missing for many customers.

Answer:

- Customers may belong to a specific segment.
- They may not trust sharing financial details.
- Could indicate low-income group.
- Helps adjust marketing strategy.

SECTION B – PRACTICAL QUESTIONS

Q8. Listwise Deletion Remove all rows where Region is missing.

Answer:

Affected Row: Customer ID 105 (Amit Verma)

Original Records: 8

Records After Deletion: 7

Total Records Lost: 1

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Q9. Imputation

Handle missing values in Monthly_Sales using:

Forward Fill Tasks:

1.Apply forward fill

2.Show before vs after values

3. Explain why forward fill is suitable here

Answer:

Before Forward Fill:

101 - 12000

104 - NaN

102 - NaN

105 - 18000

107 - 14000

103 - 15000

106 - NaN

108 - 16000

After Forward Fill:

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101 - 12000

104 - 12000

102 - 12000

105 - 18000

107 - 14000

103 - 15000

106 - 15000

108 - 16000

Forward fill is suitable because sales data is continuous and previous values can estimate missing ones.

Q10. Flagging Missing Data

Create a flag column for missing Income.

Tasks:

1. **Create Income_Missing_Flag (0 = present, 1 = missing)**
2. **Show updated dataset**

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3. **Count how many customers have missing income**

Answer:

101 - 65000 - Flag: 0

104 - NaN - Flag: 1

102 - NaN - Flag: 1

105 - 58000 - Flag: 0

107 - NaN - Flag: 1

103 - 72000 - Flag: 0

106 - 61000 - Flag: 0

108 - 69000 - Flag: 0

Total Customers with Missing Income: 3