Technical Test Junior Data Analyst





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Certified Python Associate | Statistics and Data Science Graduate | Data Enthusiast Task 1

Predictive Feature Analysis For Missing Data

Description:

Identify patterns in missing values across the dataset by writing an SQL query. Specifically:

- Find columns with more than 20% missing values.
- Calculate and return the following metrics for each such column: Table name, Column name, Total rows in the table, Number of missing values, Percentage of missing values
- Order the results by the percentage of missing values in descending order.

Github SQL

Predictive Feature Analysis For Missing Data

```
/* TASK 1 : Predictive Feature Analysis for Missing Data */
— Membuat tabel customer
 CREATE TABLE public.customer (
     id INT PRIMARY KEY,
     name VARCHAR(255),
     email VARCHAR(255),
     phone VARCHAR(255),
     address VARCHAR(255)
➡ INSERT INTO public.customer (id, name, email, phone, address) VALUES
 (1, 'Rizaldy Uto', 'uto@example.com', '1234567890', NULL),
 (2, 'Caemila', NULL, NULL, NULL),
 (3, NULL, 'caem@example.com', '9876543210', 'Elm Street'),
 (4, NULL, NULL, NULL, NULL),
 (5, 'Johan Chris', 'jo@example.com', '555555555', 'Pine Street');
 CREATE TABLE public.orders (
     id INT PRIMARY KEY,
     customer id INT.
     order_date DATE,
     delivery_date DATE,
     tracking_number VARCHAR(128),
     FOREIGN KEY (customer id) REFERENCES customer(id)
■ INSERT INTO public orders (id, customer_id, order_date, delivery_date, tracking_number) VALUES
 (1, 1, '2025-01-01', '2025-01-03', '123-ABC'),
 (2, 1, '2025-01-02', NULL, '456-DEF'),
 (3, 2, NULL, NULL, NULL),
 (4, 3, '2025-01-03', '2025-01-04', NULL),
  (5, 4, NULL, NULL, NULL);
```

```
SELECT 'customer' AS table_name,
       'name' AS column_name,
COUNT(*) AS total_rows,
COUNT(*) FILTER (WHERE name IS NULL) AS missing_value
   FROM public customer
   UNION ALL
       COUNT(*),
COUNT(*) FILTER (WHERE address IS NULL)
   UNION ALL
       COUNT(*) FILTER (WHERE phone IS NULL)
   UNION ALL
        'order_date',
       COUNT(*) FILTER (WHERE order date IS NULL)
   UNION ALL
        'delivery_date',
       COUNT(*) FILTER (WHERE delivery_date IS NULL)
   FROM public orders UNION ALL
        'tracking_number',
        COUNT(*) FILTER (WHERE tracking_number IS NULL)
   ROUND((missing_value::DECIMAL / total_rows) * 100, 2) || '%' AS missing_percentage
ORDER BY missing_percentage DESC;
```

Task 2

Scenario:

You are working as a data analyst for a company that tracks dealer performance and product sales. The management has requested a comprehensive report to gain insights into dealer performance, product profitability, and forecast accuracy. This report must include detailed segmentation and ranking based on various metrics.

Github SQL

```
a. Indentify the top-performing and least-performing dealers in terms of total sales. Include dealer ID, dealer name, total sales, and their rank. with cite as c. select dealer_id, dealer_id, dealer_name as total_sales from public.cealers dealer_id) term to the control of the control of
```

Output:

123 dealer_id 🔻	A-Z dealer_name 🔻	123 total_sales 🔻	123 dealer_rank 🔻
25	Boyd, Rivas and Fisher	4,522	1
77	Sims-Butler	899	100
ī			

1. Identifikasi Dealer Terbaik dan Terburuk Berdasarkan Total Penjualan

Dealer terbaik dan terburuk dalam hal total penjualan telah diidentifikasi dengan menggunakan total penjualan yang dihitung dari data penjualan yang ada.

Dealer_id = 25 menunjukkan performa penjualan terbaik dengan total penjualan mencapai 4.522, sementara **Dealer_id = 77** berada di peringkat terbawah dengan hanya 899.

Output:

123 dealer_id	•	A-z dealer_name •	123 total_sales 🔻	123 decile
2	25	Boyd, Rivas and Fisher	4,522	
5	51	Baker-Davis	4,477	1
2	28	Robles Inc	4,351	1
8	34	Smith, Martin and Vargas	4,158	1
8	81	Walker Inc	3,998	1
9	91	Mueller Inc	3,980	1
5	57	Johnson, Chan and Yang	3,904	1
	a	Miles I td	3 883	1

2. Pengelompokan Dealer ke dalam Decile Berdasarkan Total Penjualan

Dealer dikelompokkan dalam 10 kelompok (deciles) berdasarkan total penjualan mereka, yang memberikan gambaran distribusi performa penjualan

Dealer dengan penjualan tertinggi masuk ke decile 1, sementara dealer dengan penjualan terendah masuk ke decile 10.

Output:

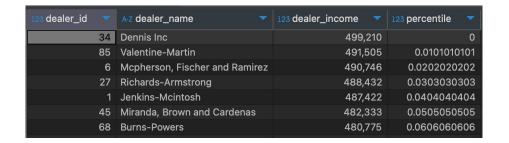
123 product_id	*	A-Z product_name		123 total_sales 🔻	123 quartiles	*
	40	e-enable world-class partnerships		7,010		
	37	enable B2C bandwidth		6,939		
Ī	19	e-enable value-added experiences		6,915		
i	4	aggregate seamless synergies		6,797		
!	50	whiteboard one-to-one supply-chains		6,618		
	21	integrate killer users		6,572		
:	26	integrate extensible markets		6,417		1

3. Segmentasi Produk ke dalam Kuartil Berdasarkan Total Penjualan

Produk dikelompokkan dalam 4 kuartil berdasarkan total penjualan. Produk dengan penjualan tertinggi berada di kuartil pertama, dan produk dengan penjualan terendah berada di kuartil keempat.

```
0— 4. Rank dealers based on their income percentile and list the percentile rank for each dealer. SELECT dealer_id, dealer_name, dealer_name, dealer_income , percent_rank() OVER (ORDER BY dealer_income DESC) AS percentile FROM public.dealers
```

Output:



4. Peringkat Dealer Berdasarkan Persentil Penghasilan

Dealer peringkat berdasarkan penghasilan mereka menggunakan persentil. Dealer dengan penghasilan terendah berada di persentil 100.

```
0—— 5. Categorize dealers into 5 age groups based on their age and display the group for each dealer. select dealer_id, dealer_name, dealer_name, dealer_age , ntile(5) over (order by dealer_age DESC) as age_group from public.dealers d
```

5. Kategorisasi Dealer ke dalam 5 Kelompok Berdasarkan Usia

Dealer dikelompokkan dalam 5 kelompok usia untuk memberikan gambaran mengenai usia mereka.

Output:



```
0—— 5. Categorize dealers into 5 age groups based on their age and display the group for each dealer. select dealer_id, dealer_name, dealer_name, dealer_age , ntile(5) over (order by dealer_age DESC) as age_group from public.dealers d
```

5. Kategorisasi Dealer ke dalam 5 Kelompok Berdasarkan Usia

Dealer dikelompokkan dalam 5 kelompok usia untuk memberikan gambaran mengenai usia mereka.

Output:



```
--- 6. Rank products by sales performance within each dealer and identify the top-performing product for each dealer.

with cte as (
select

dealer_id,

dealer_name,

product_id,

product_name,

SUM(sale_amount) as total_sales

from public.dealers d

left join public.products p using (dealer_id)

left join public.products p using (product_id)

group by dealer_id, dealer_name, product_id , product_name
),

cte_2 as (
select

*,

rank() over (partition by dealer_name order by total_sales DESC) as rank_dealer_product

from cte
}

select *

from cte_2
where rank_dealer_product = 1
```

Output:

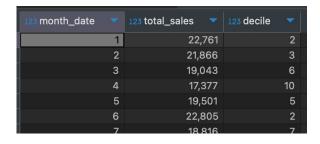
123 dealer_id 🔻	AZ dealer_name V	123 product_id 🔻	AZ product_name	123 total_sales 🔻	123 rank_dealer_product	•
64	Allen Group		disintermediate end-to-end networks	463		
82	Armstrong-Harris		target 24/365 solutions	638		
51	Baker-Davis		orchestrate distributed platforms	852		
65	Baker-Jones		integrate extensible markets	494		
3	Bishop and Sons		cultivate real-time e-tailers	481		
25	Boyd, Rivas and Fisher	43	enable next-generation solutions	814		1

6. Peringkat Produk Berdasarkan Performa Penjualan di Setiap Dealer

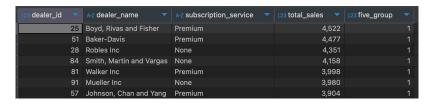
Setiap dealer memiliki produk yang berbeda, dan performa penjualan produk diurutkan untuk mengetahui produk mana yang terlaris dalam masing-masing dealer.

7. Segmentasi Penjualan Bulanan ke dalam Decile Penjualan bulanan dealer dikelompokkan ke dalam decile (10 kelompok)

Output:



Output:



8. Analisis Jenis Langganan dan Performa Penjualan Dealer

Dealer dikelompokkan berdasarkan jenis langganan mereka dan performa penjualan, untuk menentukan korelasi antara keduanya.

Output:

123 🔻	A-Z dealer_n; ▼	123 actu 🔻	123 forecasted_sales 🔻	123 forecast_error	123 forecast_accuracy ▼	123 accuracy_group	•
28	Robles Inc	4,351	4,372	0.48	99.52		
64	Allen Group	1,175	1,166	0.77	99.23		
	Deleon-Rogers	3,328	3,384	1.68	98.32		
	Harris-Graham	2,719	2,659	2.21	97.79		
56	Sanders-Lewis	1,808	1,766	2.32	97.68		
95	Ellis-Rogers	2,261	2,323	2.74	97.26		
85	Valentine-Martir	2,371	2,304	2.83	97.17		1

9. Evaluasi Akurasi Perkiraan Penjualan untuk Setiap Dealer

Akurasi perkiraan penjualan dihitung dengan menghitung selisih antara perkiraan dan penjualan aktual.

```
-- 10. Rank products into deciles based on their profit margin.

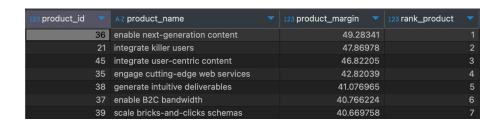
select

*,

rank() over (order by product_margin DESC) as rank_product
from public.products p
```

10. Peringkat Produk Berdasarkan Margin Laba Produk diperingkat berdasarkan margin laba mereka untuk mengidentifikasi produk yang paling menguntungkan.

Output:



Task 3

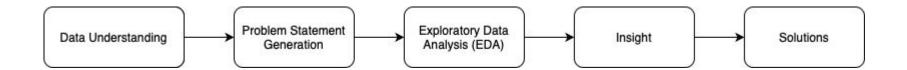
■ Telco Company Customer Behavior Analysis

Problem Statement:

A telecommunication company wants to analyze its customer's behavior. Given the following dataset, build a comprehensive analysis that provide meaningful insights to the telecommunication company.

Tableau Dashboard

■ Step

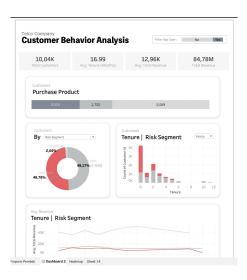


Problem Statement Generation

Problem Statement

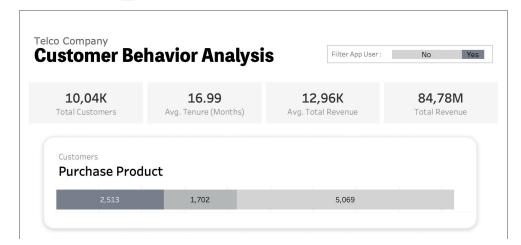
How can the company leverage insights about customer segmentation, product purchasing behavior, and revenue trends to improve customer retention, maximize revenue, and tailor their offerings effectively?

EDA With Tableu Dashboard



"I used Tableau for EDA, focusing on KPI indicators, pie charts, bar charts, line charts, and detailed tables to analyze customer behavior, segment distributions, and trends effectively."

Tableau Dashboard

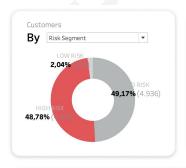


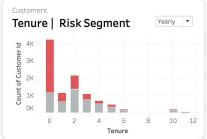
KPI Metrics:

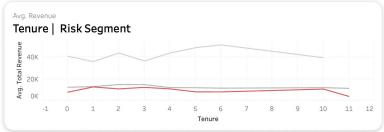
- Total Customers: 10K
- Average Tenure: **16.99 months**
- Average Total Revenue: **12.96K**
- Total Revenue: **84.78M**

Product Analysis:

• **Product C** is the most purchased, followed by Product A and Product B.

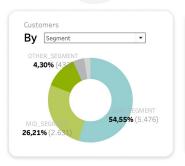


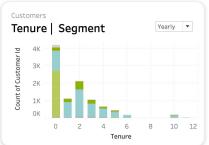


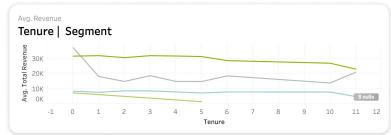


Customer by Risk Segment:

- Most customers are in the MID RISK category, followed by HIGH RISK and LOW RISK.
- LOW RISK customers generate the highest average revenue by tenure, followed by MID RISK and HIGH RISK.



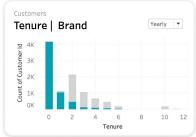


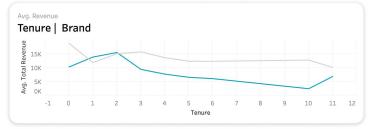


Customer by Segment:

- Most customers belong to the LOW_SEGMENT, followed by MID SEGMENT, HIGH SEGMENT, and OTHER SEGMENT.
- Customers in the MID SEGMENT have a tenure of less than one year.
- Average revenue is highest in the HIGH SEGMENT, followed by OTHER SEGMENT, LOW SEGMENT, and MID SEGMENT.



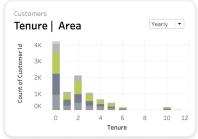


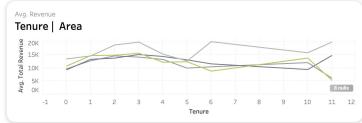


Customer by Brand:

- Most customers use Brand B, but Brand A customers show higher loyalty with longer tenure.
- Brand A also generates higher average revenue by tenure compared to Brand B.

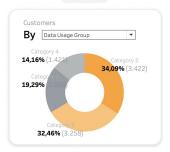


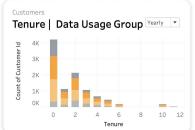


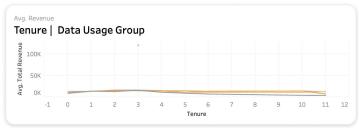


Customer by Area:

- **Area 2** has the highest total number of customers.
- Area 1 generates the highest average revenue.







Customer by Data Usage Group:

 Category 2 has the most customers among the data usage groups.

Solutions

Solutions for the Company

1. Product Strategy:

- Promote Product C aggressively and bundle it with Product A or B to boost sales.
- Investigate and revamp marketing for **Product B** to enhance its appeal.

2. Risk Management:

- Retain LOW RISK customers with loyalty programs.
- Engage MID RISK and HIGH RISK customers with targeted retention campaigns.

3. Segment Strategy:

- Extend MID SEGMENT customer tenure with onboarding campaigns.
- Sustain HIGH SEGMENT revenue by offering exclusive benefits.

4. Brand Strategy:

- Position Brand A as premium to capitalize on loyalty.
- Improve **Brand B** offerings to increase retention and revenue.

5. Regional Focus:

 Focus on growth in Area 2 and high-value campaigns in Area 1.

6. **Data Usage Optimization**:

 Tailor data plans for Category 2 users and encourage higher usage in other categories. ■ Terima kasih