

By . Adam Maulana

Data Analyst

Portfolio.



Hello.





About Me.

Hi! I'm Adam Maulana
an Data enthusiast with
a strong passion for
analysis, digital
content, and
visualization.
Experienced in data
management and
storytelling to support
decisions. Adaptable,
communicative, and
quick to learn new
tools.

Education & Experience





Data Analyst.

Data Cleansing - Data Segmentation -
Retention Cohort Analysis - Time Series
Analysis - Basket Size Analysis

[Click Here To See Full Coding_](#)

Data Cleansing

```
[2]: df_sales = pd.read_csv('Sales Transaction v.4a.csv') # membaca file
```

▼ Data Cleansing

```
[3]: # 1. Ubah kolom Date menjadi tipe datetime
df_sales['Date'] = pd.to_datetime(df_sales['Date'])

# 2. Buang semua transaksi yang memiliki quantity negative atau yang TransactionNo diawali dengan C
df_sales = df_sales[df_sales['TransactionNo'].str[0]!='C']
df_sales
```

This dataset was utilized during my Intermediate Data Analyst certification program at Growia, where I applied various analytical techniques as part of the final project.

	TransactionNo	Date	ProductNo	ProductName	Price	Quantity	CustomerNo	Country
0	581482	2019-12-09	22485	Set Of 2 Wooden Market Crates	21.47	12	17490.0	United Kingdom
1	581475	2019-12-09	22596	Christmas Star Wish List Chalkboard	10.65	36	13069.0	United Kingdom
2	581475	2019-12-09	23235	Storage Tin Vintage Leaf	11.53	12	13069.0	United Kingdom
3	581475	2019-12-09	23272	Tree T-Light Holder Willie Winkie	10.65	12	13069.0	United Kingdom
4	581475	2019-12-09	23239	Set Of 4 Knick Knack Tins Poppies	11.94	6	13069.0	United Kingdom
...
536320	536585	2018-12-01	37449	Ceramic Cake Stand + Hanging Cakes	20.45	2	17460.0	United Kingdom
536321	536590	2018-12-01	22776	Sweetheart 3 Tier Cake Stand	20.45	1	13065.0	United Kingdom
536322	536590	2018-12-01	22622	Box Of Vintage Alphabet Blocks	20.45	2	13065.0	United Kingdom
536323	536591	2018-12-01	37449	Ceramic Cake Stand + Hanging Cakes	20.45	1	14606.0	United Kingdom
536324	536597	2018-12-01	22220	Cake Stand Lovebird 2 Tier White	20.45	1	18011.0	United Kingdom

527765 rows × 8 columns

[Click Here To See Full Coding](#)

Data Segmentation

```
# Menghitung Volume Transaksi
volume_transaksi = df_sales.groupby("ProductNo")["TransactionNo"].nunique().reset_index()
volume_transaksi.columns = ["ProductNo", "Volume"]

# Menghitung Total Revenue (Price * Quantity)
df_sales["Revenue"] = df_sales["Price"] * df_sales["Quantity"]
total_revenue = df_sales.groupby("ProductNo")["Revenue"].sum().reset_index()
total_revenue.columns = ["ProductNo", "Total_Revenue"]

# Menggabungkan kedua metrik
df_segmented = volume_transaksi.merge(total_revenue, on="ProductNo")

# Mengambil nama produk dari df_sales
product_names = df_sales[["ProductNo", "ProductName"]].drop_duplicates()

# Menggabungkan ProductName ke df_segmented
df_segmented = df_segmented.merge(product_names, on="ProductNo", how="left")

# Menentukan batas segmentasi (20% teratas, 20-80%, dan 20% terbawah)
quantile_volume = df_segmented["Volume"].quantile([0.2, 0.8])
quantile_revenue = df_segmented["Total_Revenue"].quantile([0.2, 0.8])
```

```
# Menentukan kategori Volume
def categorize_volume(volume):
    if volume >= quantile_volume[0.8]:
        return "Popular"
    elif volume >= quantile_volume[0.2]:
        return "Normal"
    else:
        return "Low"

df_segmented["Volume_Category"] = df_segmented["Volume"].apply(categorize_volume)

# Menentukan kategori Revenue
def categorize_revenue(revenue):
    if revenue >= quantile_revenue[0.8]:
        return "Popular"
    elif revenue >= quantile_revenue[0.2]:
        return "Normal"
    else:
        return "Low"

df_segmented["Revenue_Category"] = df_segmented["Total_Revenue"].apply(categorize_revenue)

# Menentukan kategori final berdasarkan kombinasi Volume dan Revenue
def categorize_final(row):
    if row["Volume_Category"] == "Popular" and row["Revenue_Category"] == "Popular":
        return "Super Popular"
    elif row["Volume_Category"] == "Popular" or row["Revenue_Category"] == "Popular":
        return "Popular"
    elif row["Volume_Category"] == "Low" and row["Revenue_Category"] == "Low":
        return "Low"
    else:
        return "Normal"

df_segmented["Final_Category"] = df_segmented.apply(categorize_final, axis=1)

# Menampilkan beberapa baris pertama hasil segmentasi
df_segmented
```

In this project, I performed product segmentation based on:

- Transaction Volume: The number of transactions involving a product, regardless of the quantity purchased per transaction.
- Total Revenue: The total dollar revenue generated from the sales of each product.

The segmentation was done using the following criteria:

- Total revenue is calculated as the product of item price and quantity sold.
- Products were segmented by transaction volume as follows:
 - The top 20% highest in transaction volume were labeled "Popular"
 - The middle 60% (20%-80%) were labeled "Normal"
 - The bottom 20% were labeled "Low"
- A similar rule was applied to total revenue:
 - Top 20% in revenue → "Popular"
 - Middle 60% → "Normal"
 - Bottom 20% → "Low"

Then, based on the combination of both segmentations:

- Products that were "Popular" in both transaction volume and total revenue were labeled "Super Popular"
- Products that were "Popular" in one metric and "Normal" in the other were labeled "Popular"
- Products that were "Low" in both metrics were labeled "Low"
- All other combinations were categorized as "Normal"

[Click Here To See Full Coding](#)

Retention Cohort Analysis

Input:

```
# Filter data berdasarkan periode
df_sales_filtered = df_sales[(df_sales["Date"] >= "2019-01-01") & (df_sales["Date"] <= "2019-11-30")].copy()

# Tentukan bulan pertama transaksi (CohortMonth) & bulan transaksi (TransactionMonth)
df_sales_filtered["CohortMonth"] = df_sales_filtered.groupby("CustomerNo")["Date"].transform("min").dt.to_period("M")
df_sales_filtered["TransactionMonth"] = df_sales_filtered["Date"].dt.to_period("M")

# Hitung perbedaan bulan (CohortIndex)
df_sales_filtered["CohortIndex"] = ((df_sales_filtered["TransactionMonth"].dt.year - df_sales_filtered["CohortMonth"].dt.year) * 12 +
                                     (df_sales_filtered["TransactionMonth"].dt.month - df_sales_filtered["CohortMonth"].dt.month))

# Buat matriks kohort retensi
cohort_counts = df_sales_filtered.pivot_table(index="CohortMonth", columns="CohortIndex", values="CustomerNo", aggfunc=pd.Series.nunique)

# Hitung jumlah pelanggan baru (count_new_customer)
cohort_sizes = cohort_counts.iloc[:, 0]
retention_matrix = cohort_counts.divide(cohort_sizes, axis=0)

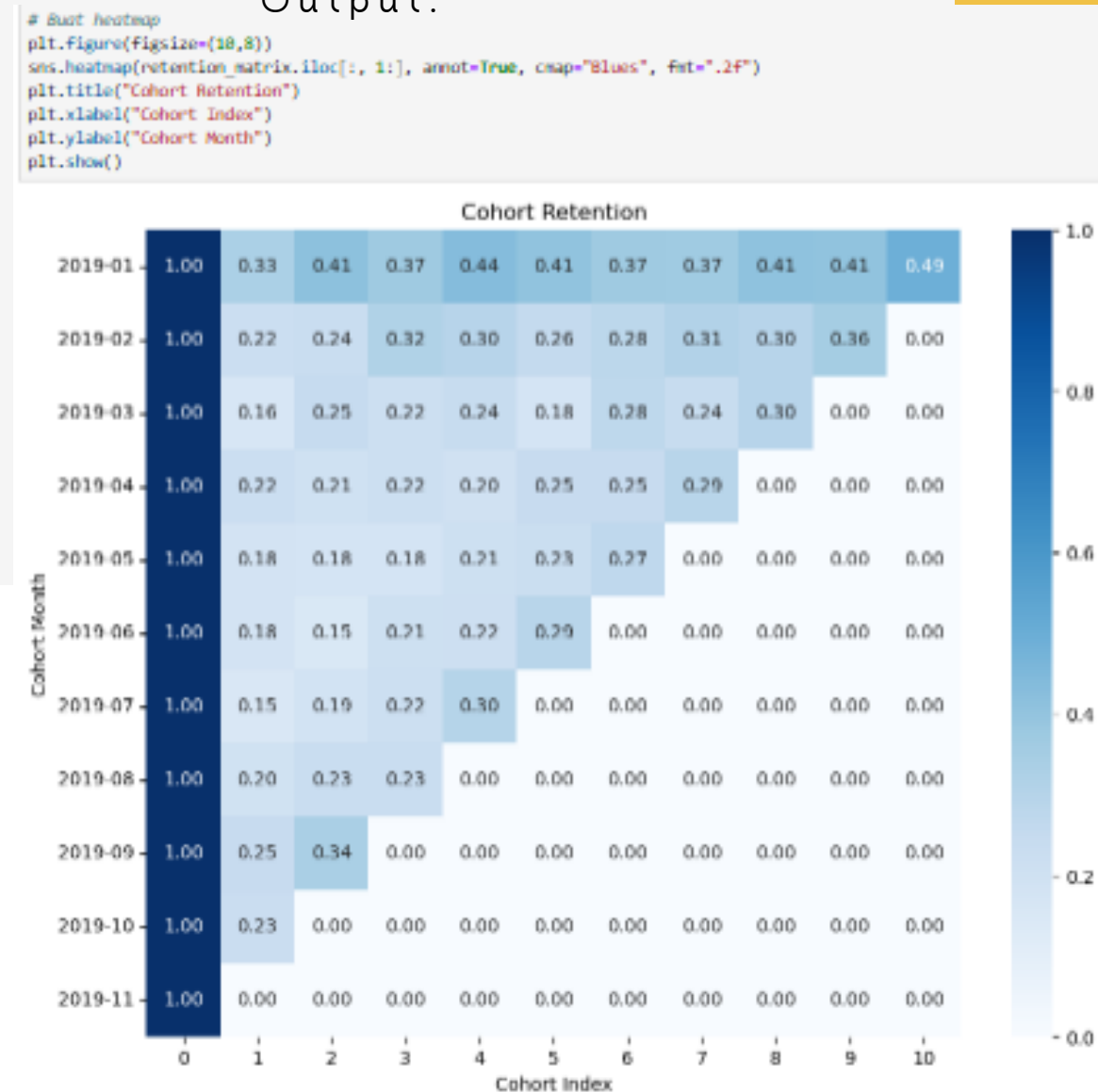
# Tambahkan count_new_customer sebelum normalisasi
retention_matrix.insert(0, "count_new_customer", cohort_sizes)

# Format index agar hanya menampilkan bulan numerik
retention_matrix.index = retention_matrix.index.strftime("%Y-%m")

# Ganti NaN dengan 0 untuk menghindari missing values
retention_matrix = retention_matrix.fillna(0)

# Tampilkan hasil akhir
retention_matrix
```

Output:



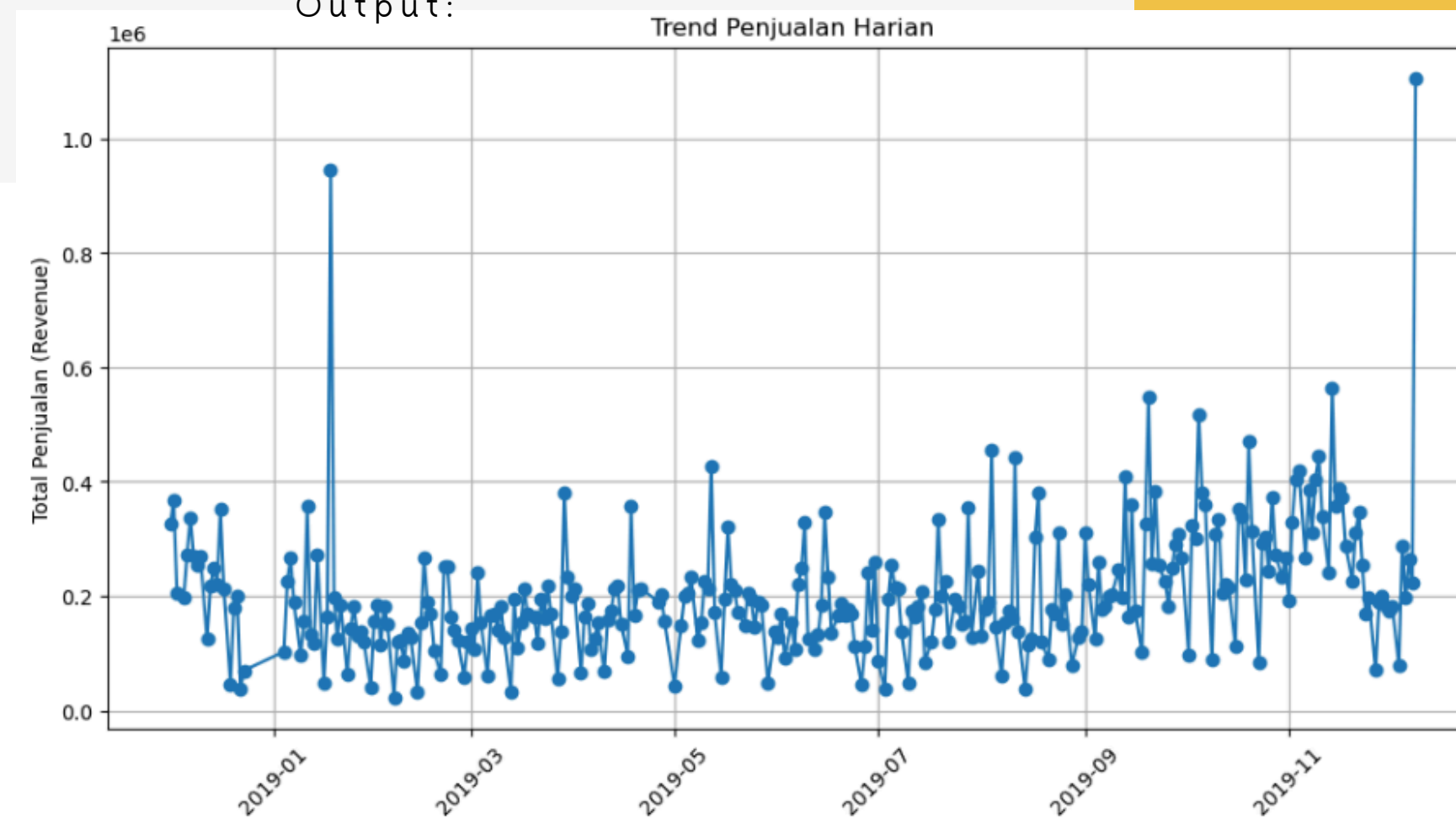
[Click Here To See Full Coding](#)

Time Series Analysis

Input:

```
# Nomor 1
# Agregasi total penjualan per hari
daily_sales = df_sales.groupby('Date')['Revenue'].sum()
# Plot trend penjualan harian
plt.figure(figsize=(12, 6))
plt.plot(daily_sales.index, daily_sales.values, marker='o', linestyle='-')
plt.xlabel("Tanggal")
plt.ylabel("Total Penjualan (Revenue)")
plt.title("Trend Penjualan Harian")
plt.xticks(rotation=45)
plt.grid()
plt.show()
```

Output:



[Click Here To See Full Coding](#)

Basket Size Analysis

Input:

```
# Tambahkan kolom total harga per baris produk
df_sales["TotalPrice"] = df_sales["Price"] * df_sales["Quantity"]

# Hitung total GMV per transaksi
gmv_per_transaction = df_sales.groupby(["DayOfWeek", "TransactionNo"]).agg({
    "TotalPrice": "sum",
    "Country": "first",
    "CustomerNo": "first"
}).reset_index()

# Hitung GMV total dan jumlah transaksi per hari
basket_size_per_day = gmv_per_transaction.groupby("DayOfWeek").agg({
    "TotalPrice": ["sum", "mean", "count"]
}).reset_index()

basket_size_per_day.columns = ["Day", "Total_GMV", "Avg_Basket_Size", "Total_Transactions"]
```

```
ordered_days = ["Monday", "Tuesday", "Wednesday", "Thursday", "Friday", "Saturday", "Sunday"]

basket_size_per_day["Day"] = pd.Categorical(
    basket_size_per_day["Day"],
    categories=ordered_days,
    ordered=True
)

data = basket_size_per_day.set_index("Day").reindex(ordered_days)
y = data["Avg_Basket_Size"].values

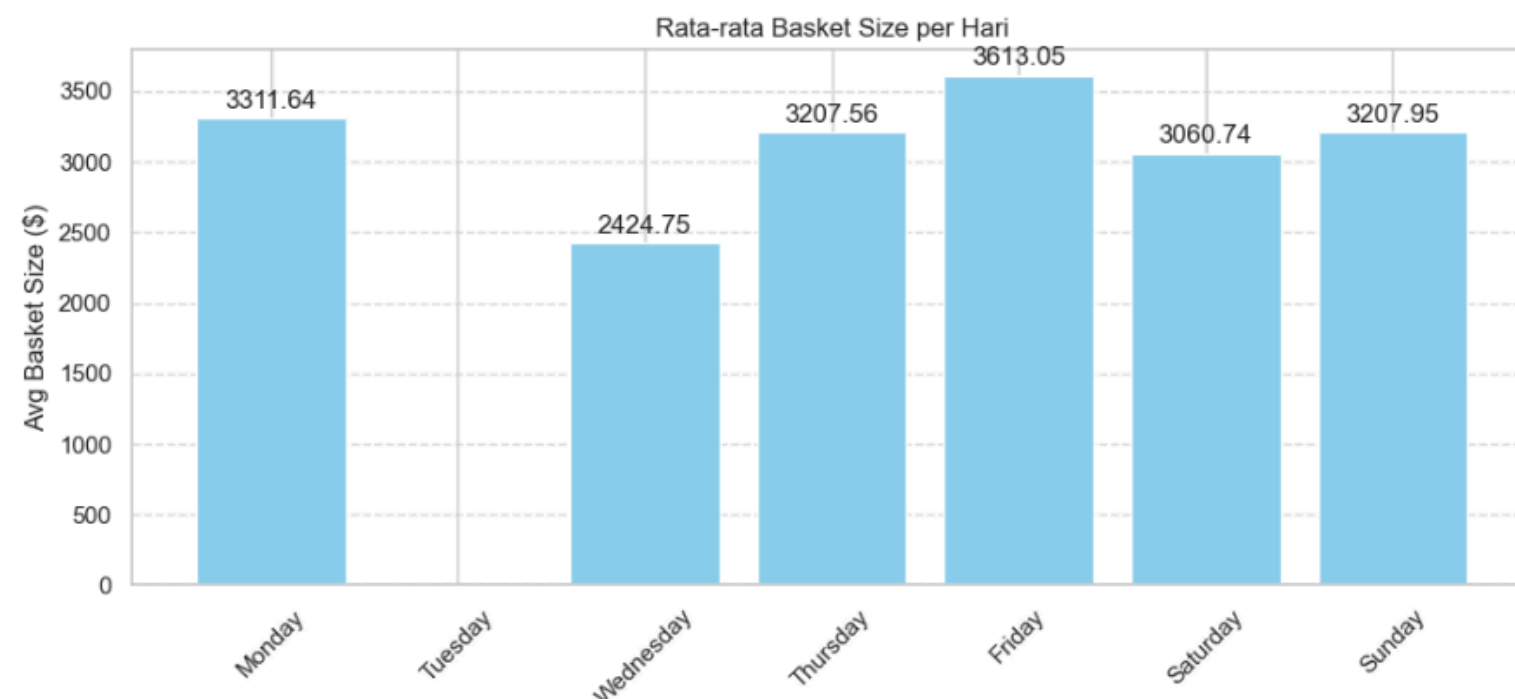
# Plotting seperti sebelumnya
import matplotlib.pyplot as plt
import numpy as np

plt.figure(figsize=(10, 5))
plt.bar(range(len(ordered_days)), y, color='skyblue')
plt.xlabel("Hari dalam Seminggu")
plt.ylabel("Avg Basket Size ($)")
plt.title("Rata-rata Basket Size per Hari")
plt.xticks(range(len(ordered_days)), ordered_days, rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)

for i, value in enumerate(y):
    if not np.isnan(value):
        plt.text(i, value + np.nanmax(y)*0.01, f"{value:.2f}", ha='center', va='bottom')

plt.tight_layout()
plt.show()
```

Output:

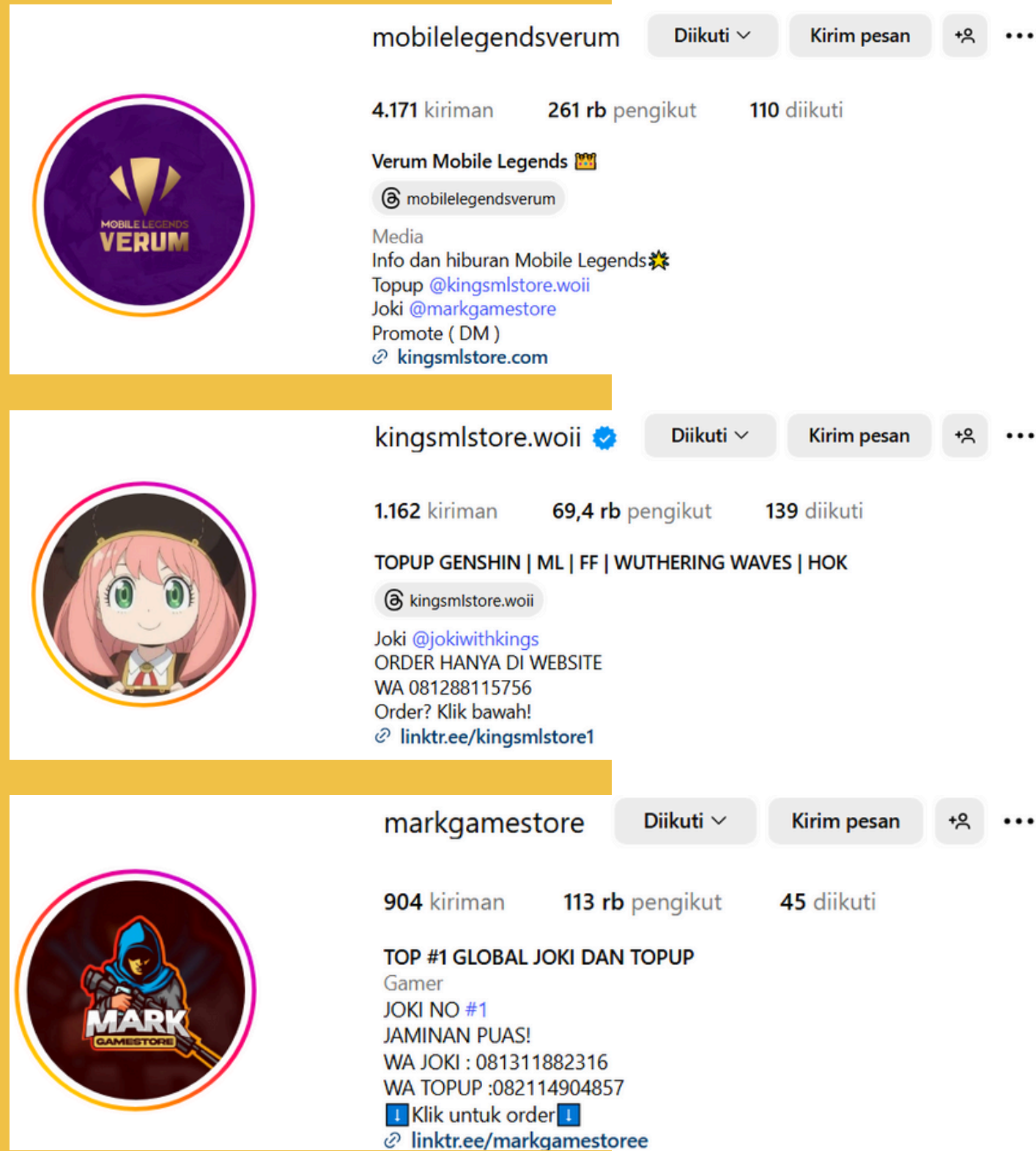




Content Creator.

Account - Content

Account



Responsible for daily content creation across multiple social media accounts, ensuring consistent engagement and brand alignment on each platform.



Content



Published over 250 content pieces across multiple social media platforms.

[check here to see all content](#)

Thank
you.

Contact Me



+62 87-880-925-800



[linkedin.com/in/adammaulana100](https://www.linkedin.com/in/adammaulana100)



adammaulana100@gmail.com