

### Final Project

Kalbe Nutritionals Data Scientist Project Based Internship

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### **Case Study Description**

In this project, as a Data Scientist at Kalbe Nutritionals, I am responsible for addressing two key objectives from the inventory team and marketing team:

- 1. Daily Sales Prediction
  - Assist the inventory team in forecasting daily sales quantity for all Kalbe products.
  - Goal: Ensure optimal stock availability to meet daily demand efficiently.
- 2. Customer Segmentation
  - Group customers based on various criteria for the marketing team.
  - Goal: Enhance promotional effectiveness through personalized marketing strategies.

This project leverages Python, Tableau, MySQL for data analysis, predictive modeling, and visualization



Exploratory Data Analysis (EDA) with MySQL



#### Query 1: Average Age of Customers based on Marital Status

```
SELECT * FROM `case study - customer`;

SELECT `Marital Status`, AVG(Age) AS "Average Age of Customers"
FROM `case study - customer`
GROUP BY `Marital Status`;
```

	Marital Status	Average Age of Customers	
Þ	31.3333		
	Married	43.0382	
	Single	29.3846	

#### Query 2: Average Age of Customers based on Gender

```
SELECT * FROM `case study - customer`;

SELECT `Gender`, AVG(Age) AS 'Average Age of Customers'
FROM `case study - customer`
GROUP BY `Gender`;
```

	Gender	Average Age of Customers
١	0	40.3264
	1	39.1415



#### Query 3: The store with the highest total quantity sold

```
SELECT s.StoreName, SUM(t.Qty) AS TotalQuantity
FROM `case study - transaction` t
JOIN `case study - store` s ON t.StoreID = s.StoreID
GROUP BY s.StoreName
ORDER BY TotalQuantity DESC
LIMIT 1;
```

```
StoreName TotalQuantity
Lingga 2777
```

#### Query 4: The best-selling product based on the highest total amount

```
SELECT p. Product Name, SUM(t.TotalAmount) AS TotalSales
FROM 'case study - transaction' t

JOIN 'case study - product' p ON t.ProductID = p.ProductID
GROUP BY p. Product Name'

ORDER BY TotalSales DESC

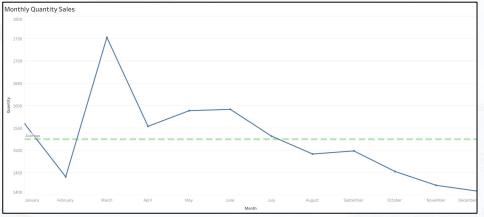
LIMIT 1;
```

	Product Name	TotalSales	
١	Cheese Stick	27615000	



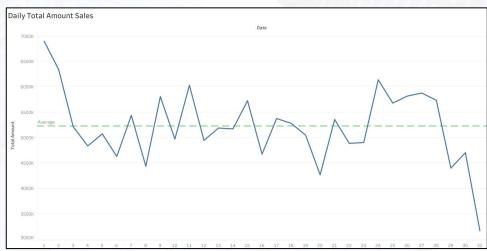
Data Visualization and Dashboard creation using Tableau

### 1. Monthly Quantity Sales



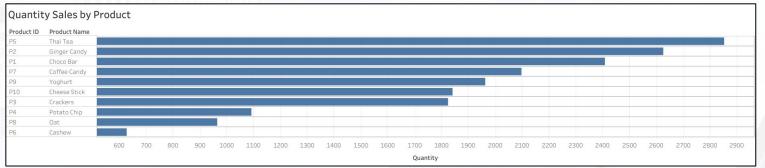


### 2. Daily Total Amount Sales

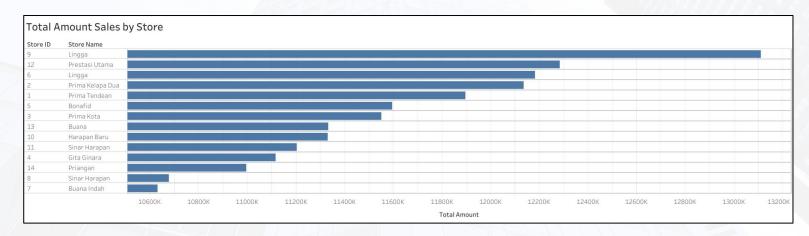


#### 3. Total Sales by Product



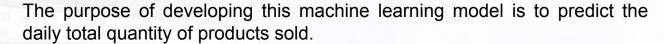


#### 4. Total Amount Sales by Store





Machine Learning Regression (Time Series) Using ARIMA





```
V Data Cleansing

[7] df_customer['Income'] = df_customer['Income'].replace('[,]','.',regex=True).astype(float)
    df_customer.head()

[8] df_store['Latitude'] = df_store['Latitude'].replace('[,]','.',regex=True).astype(float)
    df_store['Longitude'] = df_store['Longitude'].replace('[,]','.',regex=True).astype(float)
    df_store.head()

[9] df_transaction['Date'] = pd.to_datetime(df_transaction['Date'], format='%d/%m/%Y')
    df_transaction.head()
```

Data cleansing first, then adjust data types accordingly.

```
Data Merge

[10] df_merge = pd.merge(df_transaction, df_customer, on='CustomerID')
    df merge = pd.merge(df merge, df store, on='StoreID')
```

df merge.head()

df merge = pd.merge(df merge, df product, on='ProductID')

Data merging to combine all datasets



#### Data Frame Regression

- Create a new dataset for regression by grouping data by date and aggregating the quantity using sum.
- The final dataset will have approximately 365 rows.

#### Prediction ARIMA

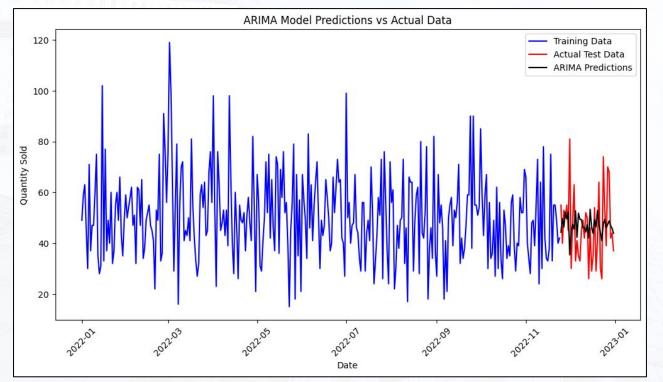
```
[115] # Generate forecast
    y_pred = ARIMA_model.get_forecast(len(df_test))

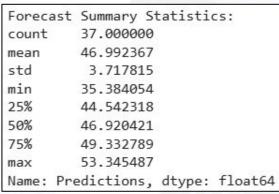
# Convert predictions to a DataFrame
    y_pred_df = y_pred.conf_int()
    y_pred_df['Predictions'] = ARIMA_model.predict(start=y_pred_df.index[0], end=y_pred_df.index[-1])
    y_pred_df.index = df_test.index
    y_pred_out = y_pred_df['Predictions']
```

The data is split into training and testing sets based on time, then the ARIMA model is trained using the training data to capture historical patterns, and finally, it is used to predict values on the testing data.

Mean Absolute Error (MAE): 11.30 Root Mean Squared Error (RMSE): 14.30 (11.30067669959999, 14.2991667067729) The ARIMA model's performance is evaluated using MAE (11.30) and RMSE (14.30), indicating its accuracy in predicting the test data.







The ARIMA model predicts that for the next month (January 1-31, 2023), the inventory team should prepare approximately 47 stocks/day.



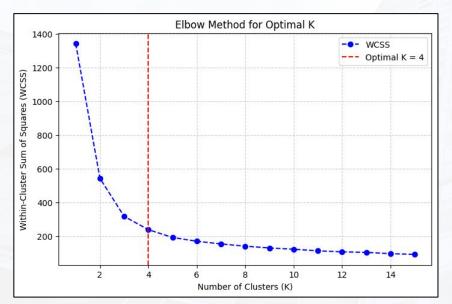
Machine Learning Clustering Using K-Means

The purpose of building this machine learning model is to create clusters of similar customers.

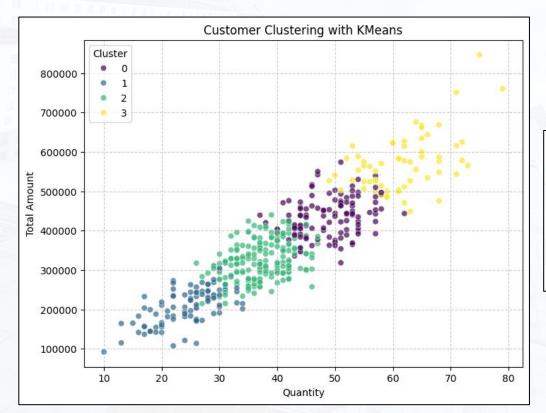


Create a new dataset for clustering by grouping by CustomerID, then aggregate the following:

- TransactionID → Count
- Qty  $\rightarrow$  Sum
- TotalAmount → Sum



The Elbow Method is used to determine the optimal number of clusters (K), and based on the graph, the best value for K is identified as 4.





	CustomerID	TransactionID	Qty	TotalAmount	
Cluster					
0	114	13.254386	49.078947	436203.508772	
1	93	7.021505	24.505376	208283.870968	
2	180	10.427778	37.350000	325663.333333	
3	60	16.316667	61.650000	572100.000000	



#### **Customer Clustering Insights**

- Cluster 3 High Value Customers
  - Highest transactions (16.3 times) & largest spending (Rp572,100)
  - Potentially loyal customers or wholesale buyers
- Cluster 0 Frequent & High Spenders
  - Large customer base (114 people), high transactions (13.25 times)
  - Frequently make large purchases (Rp436,203)
- Cluster 2 Moderate Customers
  - Largest customer group (180), moderate transactions & spending
  - Potential target for loyalty programs
- Cluster 1 Low Spenders
- Lowest transactions & spending (Rp208,283)
- Possibly new or less active customers

#### **Recommended Actions**

- ✓ Cluster 3 → Offer exclusive deals to retain high-value customers.
- ✓ Clusters 0 & 2 → Provide discounts or membership programs to boost loyalty.
- ✓ Cluster 1 
  → Launch special promotions or retargeting campaigns to increase engagement.

### **Thank You**





