

INSIGHT DATA & TOOLS

* This analysis uses the detail_transaksi dataset, and it has the same condition as MySQL analysis report

* This analysis uses Google Colab, Matplotlib-Seaborn Python and Tableau

* The Tableau analysis file was saved in the twb format, in GITHUB and Google drive



```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from matplotlib.ticker import FuncFormatter

df = pd.read_excel('detail_transaksi.xlsx')

filter = df[(df['KOMISI']=='ST') & (df['PAYMENT_STATUS'] == 'PAID')].copy()
```

revenue = filter.groupby('DEPARTMENT')['NTLAT_REVENUE_PRODUCT'].sum().reset_index(

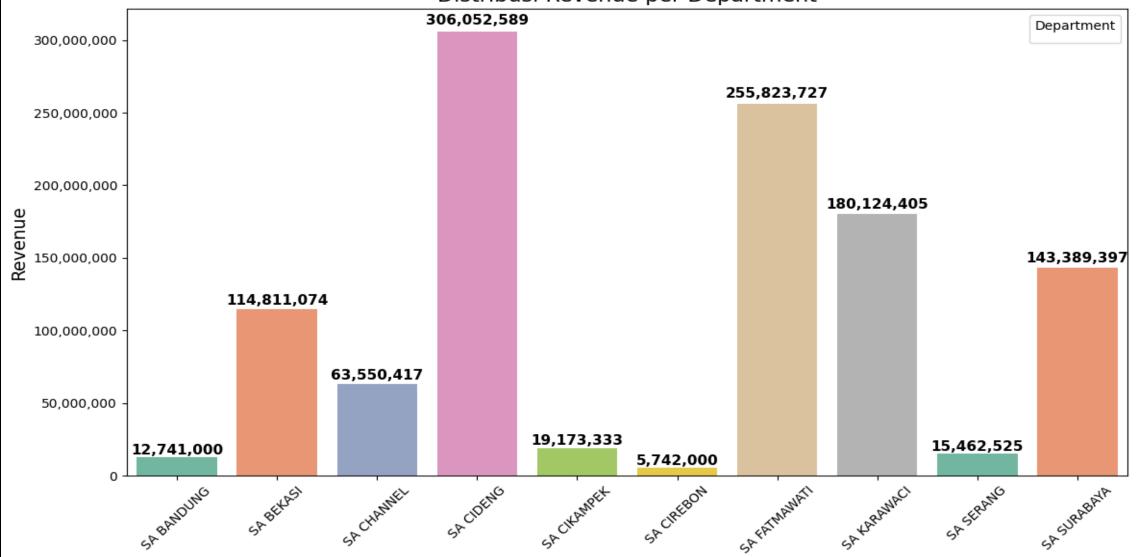


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BAR PLOT

```
#Membuat barplot
plt.figure(figsize=(12, 7))
ax = sns.barplot(data=revenue,
                 x='DEPARTMENT',
                 y = 'NILAI REVENUE PRODUCT',
                 hue='DEPARTMENT',
                 palette='Set2',
                 legend=False)
# fungsi format untuk menampilkan angka dengan koma sebagai pemisah ribuan
def format millions(x, pos):
   # fromat angka dengan koma tanpa desimal
    return f'{int(x):,}'
    # Terapkan format angka pada sumbu Y
ax.yaxis.set major formatter(FuncFormatter(format millions))
    # Manambahkan angka disetiap bar
for p in ax.patches:
        height = p.get height()
        ax.text(p.get_x() + p.get_width() / 2, # horizontal position in the mid bar
                                             # slightly vertical position above the bar
        height + (0.01 * height),
               f'{int(height):,}',
                                             # number text with comma
               ha='center',
                                             # horizontal alignment center
               va='bottom',
                                             # vertical alignment below the text
               fontsize = 11,
               fontweight='bold')
plt.title('Distribusi Revenue per Department',fontsize = 16)
plt.xlabel('Department', fontsize=14)
plt.ylabel('Revenue', fontsize=14)
plt.xticks(rotation=45)
plt.legend(title ='Department')
plt.tight layout()
plt.show()
```



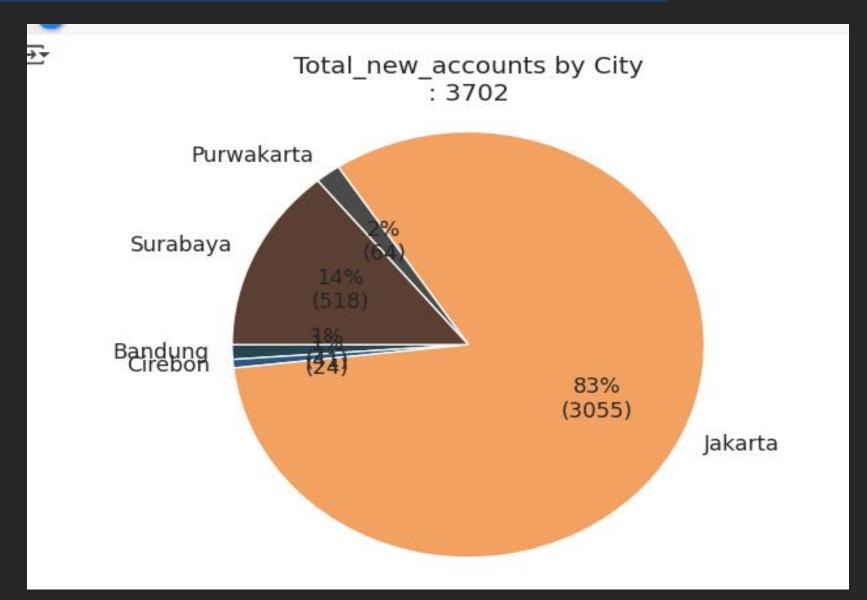


```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
!pip install pandas matplotlib seaborn openpyxl
df = pd.read_excel('detail_transaksi.xlsx')
# Use seaborn style for visual
df = pd.read_excel('detail_transaksi.xlsx')
# filter rows from KOMISI = ST and Payment status = PAID
flt = df[(df['KOMISI']=='ST') & (df['PAYMENT STATUS']=='PAID')]
# assuming each row is one account, so count rows per city
# Group by city and count total accounts ( counting unique CUST ACCT)
group by city = flt.groupby('CITY').size()
# Calculate total accounts
Total new accounts = group by city.sum()
```

PIE CHART

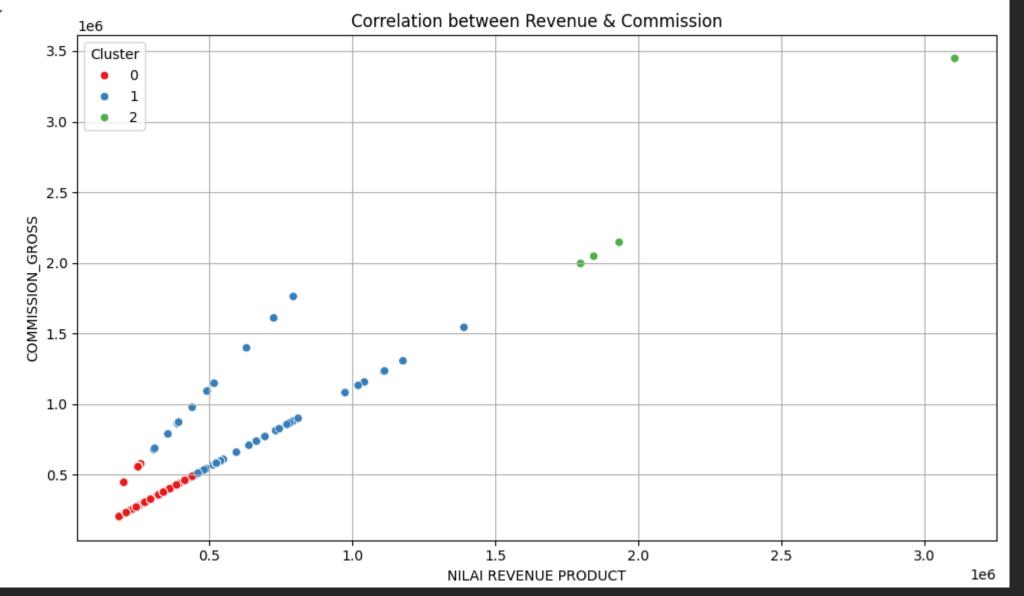
```
# Plot pie chart
plt.figure(figsize=(10, 10))
# plot color pie chart
#cl = sns.color palette(['#264653', '#E63946', '#F4A261', '#2A9D8F'])[:len(group by city)]
cl = sns.color palette(['#264653','#2B5783', '#F4A261', '#4B4B4B', '#5C4033'])[:len(group by city)]
# Define a function to show both percentage and absolute counts in label of pie chart
def autopct format(p):
  absolute = int(round(p * Total new accounts /100.0))
  return f"{p:.0f}%\n({absolute})"
plt.pie(
    group by city,
    labels=group by city.index,
    autopct=autopct format,
    colors=cl,
    startangle=180,
    textprops={'fontsize': 12}
plt.title(f'Total new accounts by City\n: {Total new accounts}', fontsize=14)
plt.axis('equal')
plt.tight layout()
```

```
plt.title(f'Total_new_accounts by City\n: {Total_new_accounts}', fontsize=14)
plt.axis('equal')
plt.tight_layout()
plt.show()
```



```
[2] import pandas as pd
       import numpy as np
       import matplotlib.pyplot as plt
       import seaborn as sns
       from sklearn.cluster import KMeans
  [5] df = pd.read_excel('detail_transaksi.xlsx')
(6] # filter data
       flt = df[(df['KOMISI']=='ST') & (df['PAYMENT_STATUS']=='PAID')]
      # Select relevant columns and drop missng values
       drop_na = flt[['NILAI REVENUE PRODUCT', 'COMMISSION_GROSS']].dropna()
  [8] #Kmeans clustering
       from sklearn.cluster import KMeans
       k means = KMeans(n clusters=3,random state=0)
       cluster = k_means.fit_predict(drop_na)
       # Add cluster label to dataframe
       flt = flt.loc[drop_na.index]
       flt['cluster']=cluster
[12] # Plot with cluster colors
       plt.figure(figsize=(10,6))
       sns.scatterplot(data=flt, x='NILAI REVENUE PRODUCT', y='COMMISSION_GROSS', hue='cluster', palette='Set1')
       plt.title('Correlation between Revenue & Commission')
       plt.xlabel('NILAI REVENUE PRODUCT')
       plt.ylabel('COMMISSION_GROSS')
       plt.legend(title='Cluster')
       plt.grid(True)
       plt.tight layout()
       plt.show()
```

SCATTER PLOT



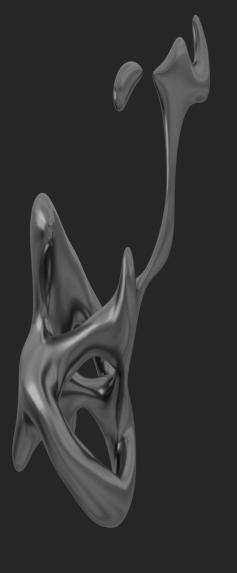
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
# Membaca data
df = pd.read_excel('detail_transaksi.xlsx')
# Filter rows where KOMISI = ST AND PAYMENT_STATUS = PAID
fit = df[(df['KOMISI'] == 'ST') & (df['PAYMENT_STATUS'] == 'PAID')]
# Group by TERM PAYMENT and count unique accounts [CUST_ACCT]
Total_new_accounts = flt.groupby('TERM PAYMENT')['CUST_ACCT'].nunique().reset_index()
# Define the correct order of the months
month_order = ['JAN', 'FEB', 'MAR', 'APR', 'MEI', 'JUNE', 'JULY', 'AUG', 'SEP', 'OKT', 'NOV', 'DEC']
# Ensure 'TERM PAYMENT' is the categorically correct order
Total_new_accounts['TERM PAYMENT'] = pd.Categorical(
  Total_new_accounts['TERM PAYMENT'],
  categories=month order,
  ordered=True
# Sorting by month order
Total_new_accounts = Total_new_accounts.sort_values('TERM PAYMENT')
# Line plotting
plt.figure(figsize=(12, 8))
sns.lineplot(data=Total_new_accounts, x='TERM PAYMENT', y='CUST_ACCT', marker='o', markersize=10)
plt.title('Monthly trend of new accounts')
plt.xlabel('Month')
```

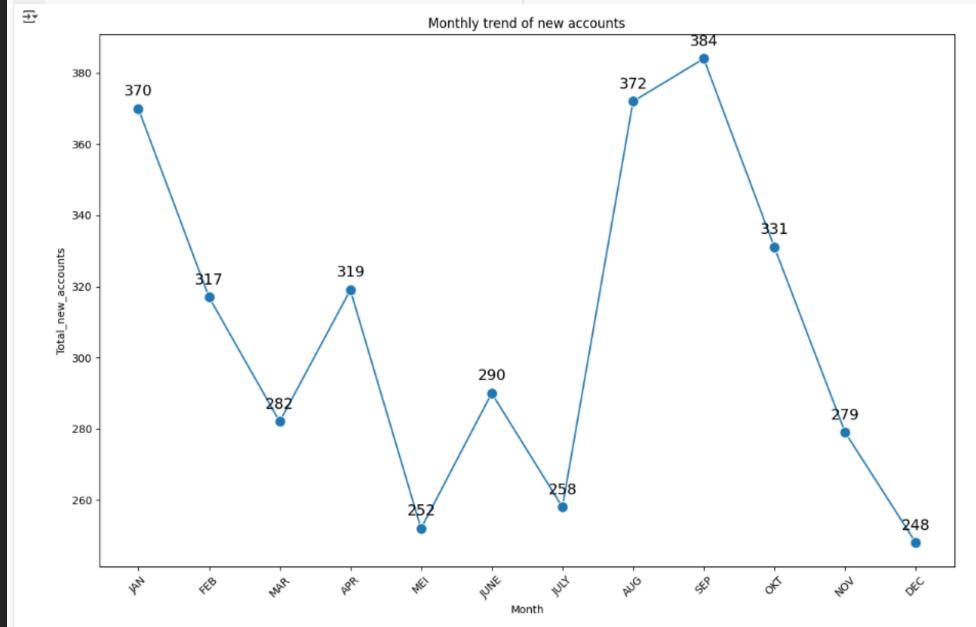
```
plt.ylabel('Total_new_accounts')
plt.grid(True)
plt.xticks(rotation=45)
plt.tight_layout()

# Adding numbered label in each marker
for i, row in Total_new_accounts.iterrows():
    plt.annotate(
        f'{row.CUST_ACCT}',
        (row['TERM PAYMENT'], row.CUST_ACCT),
        textcoords='offset points',
        xytext=(0, 12), # label's position is above the dot
        ha='center',
        fontsize=14
    )

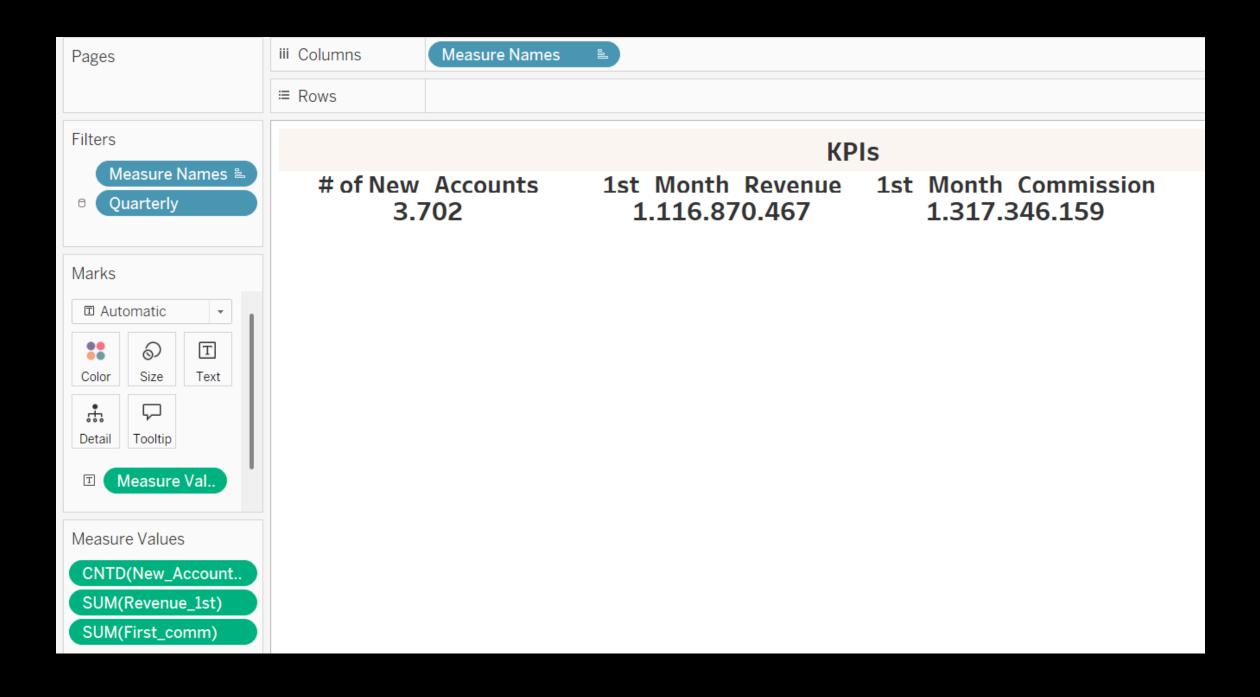
plt.show()
```

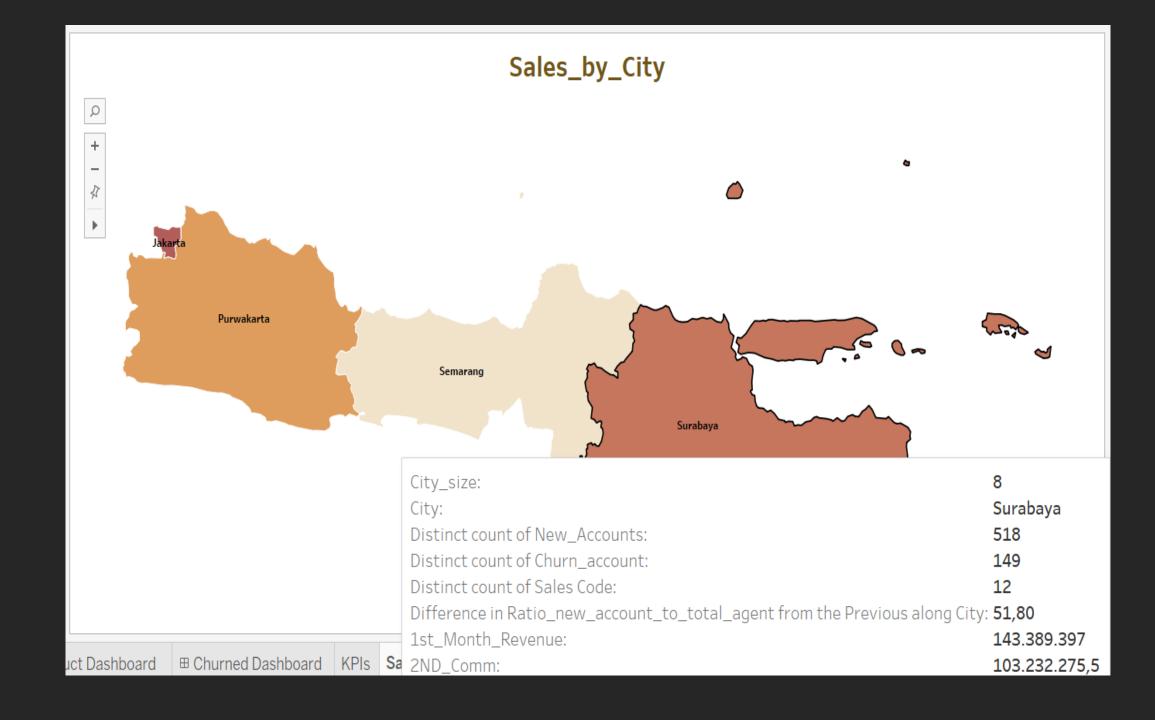
LINE PLOT



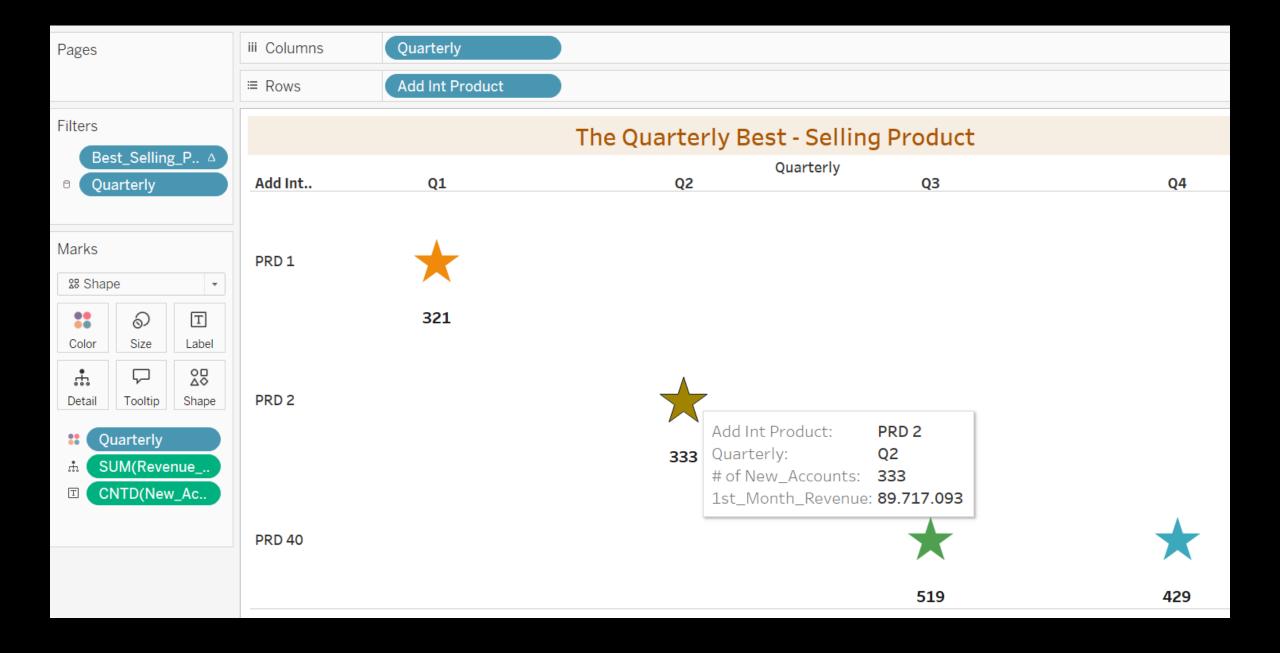


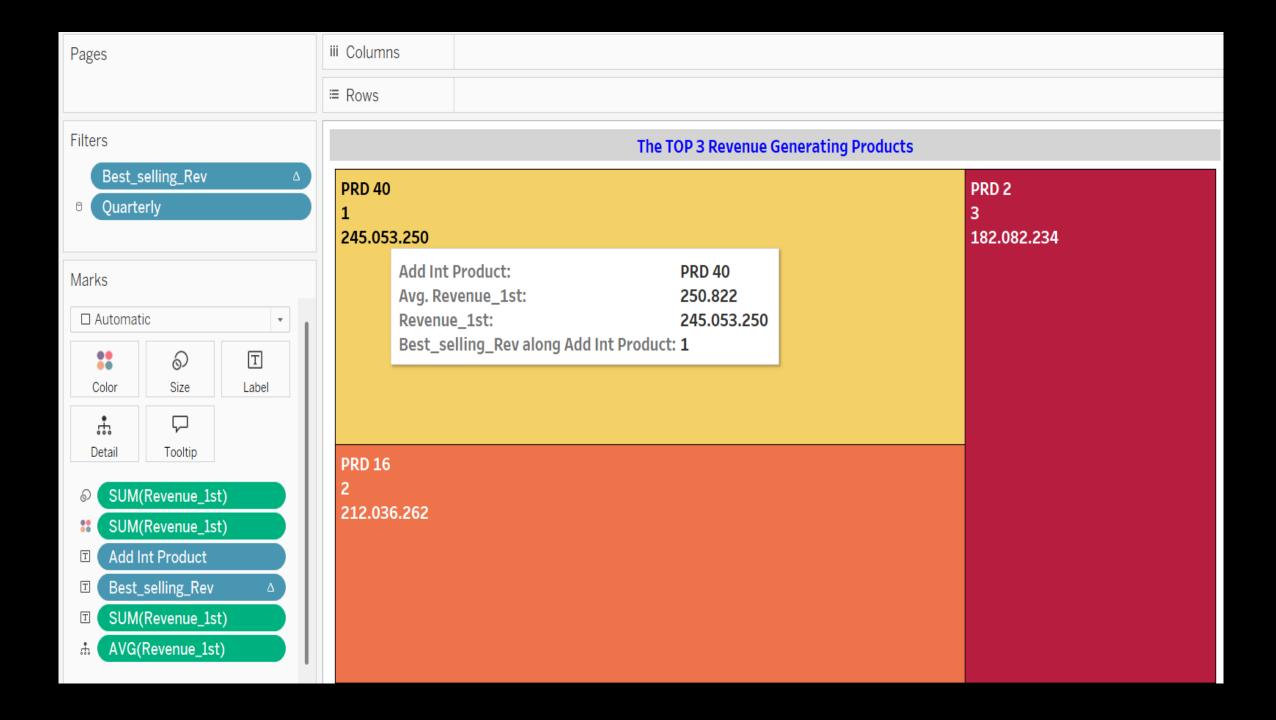


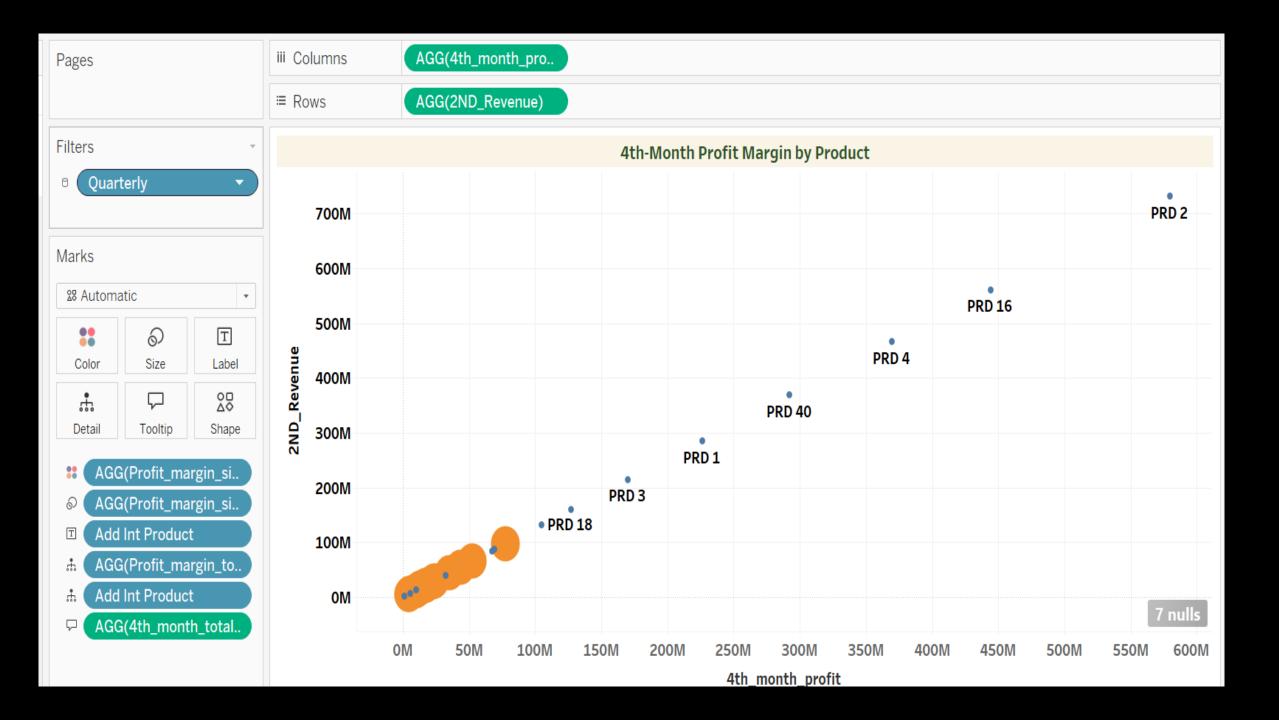


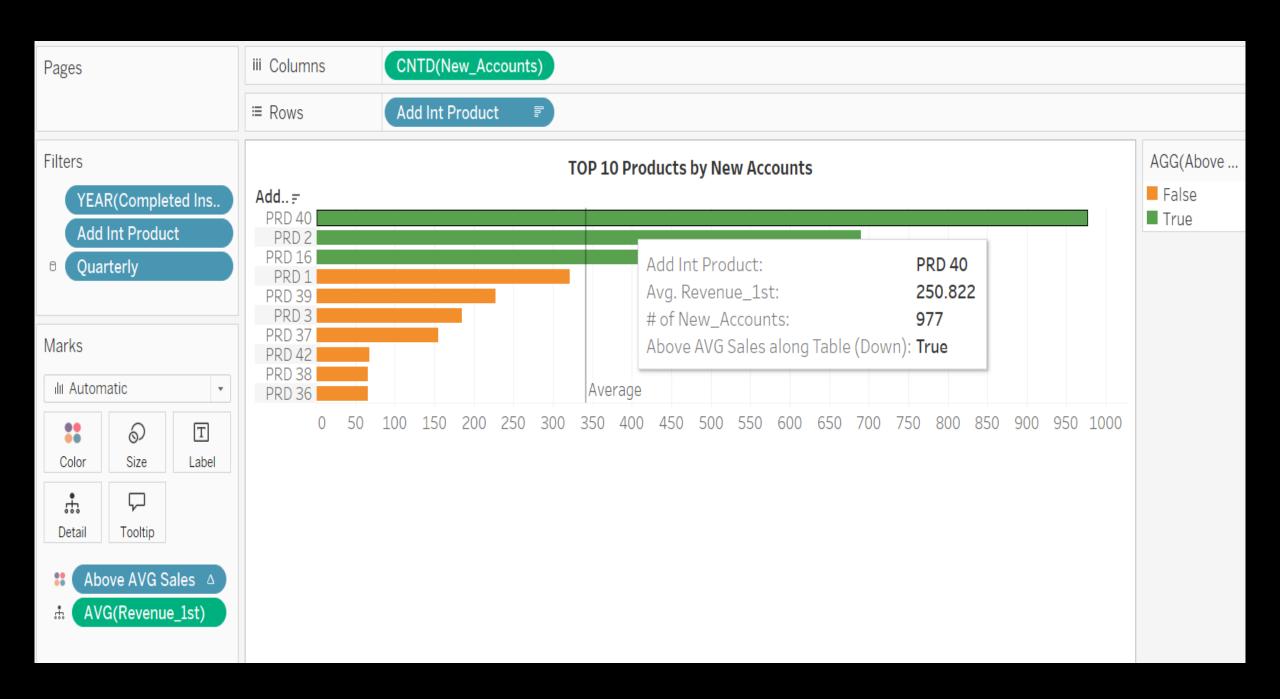


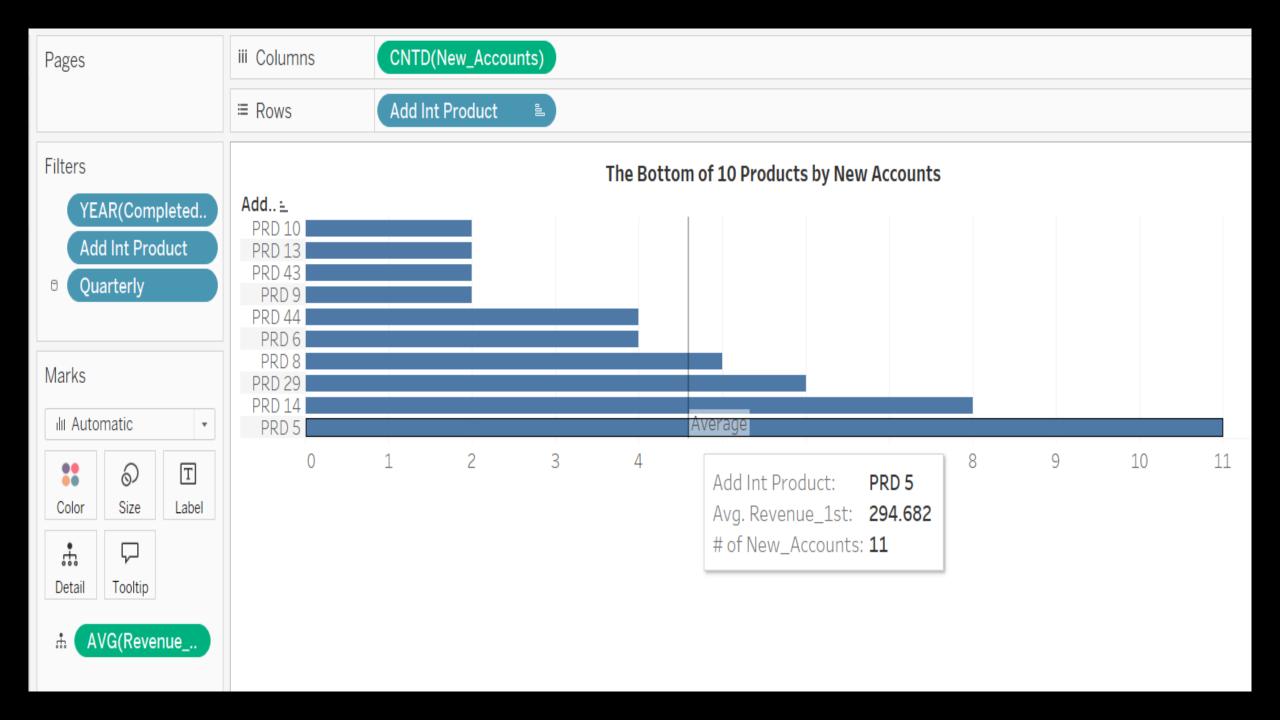


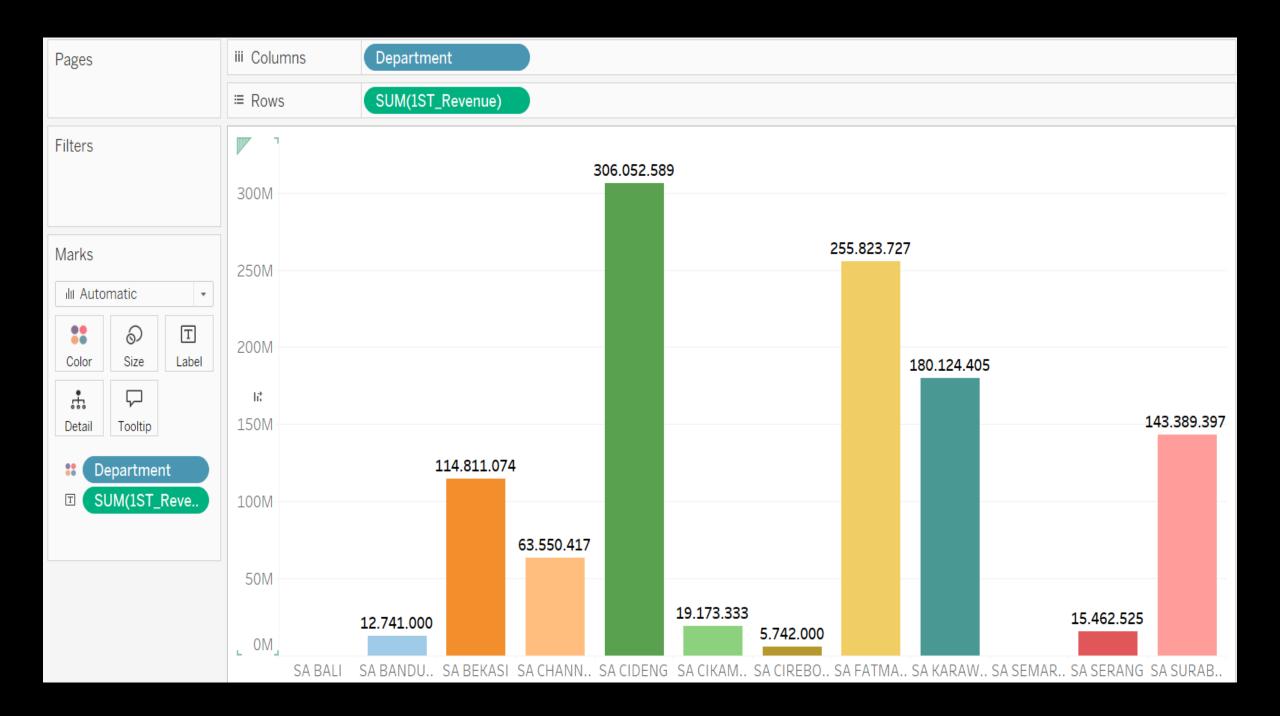


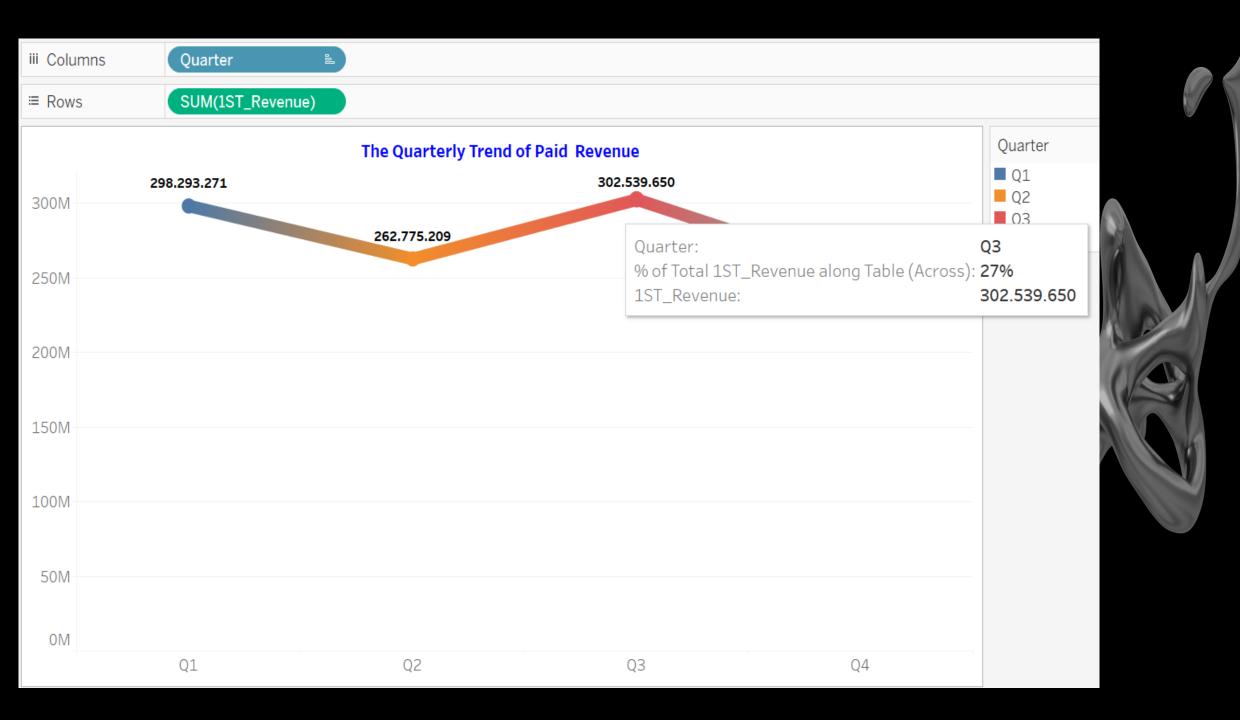


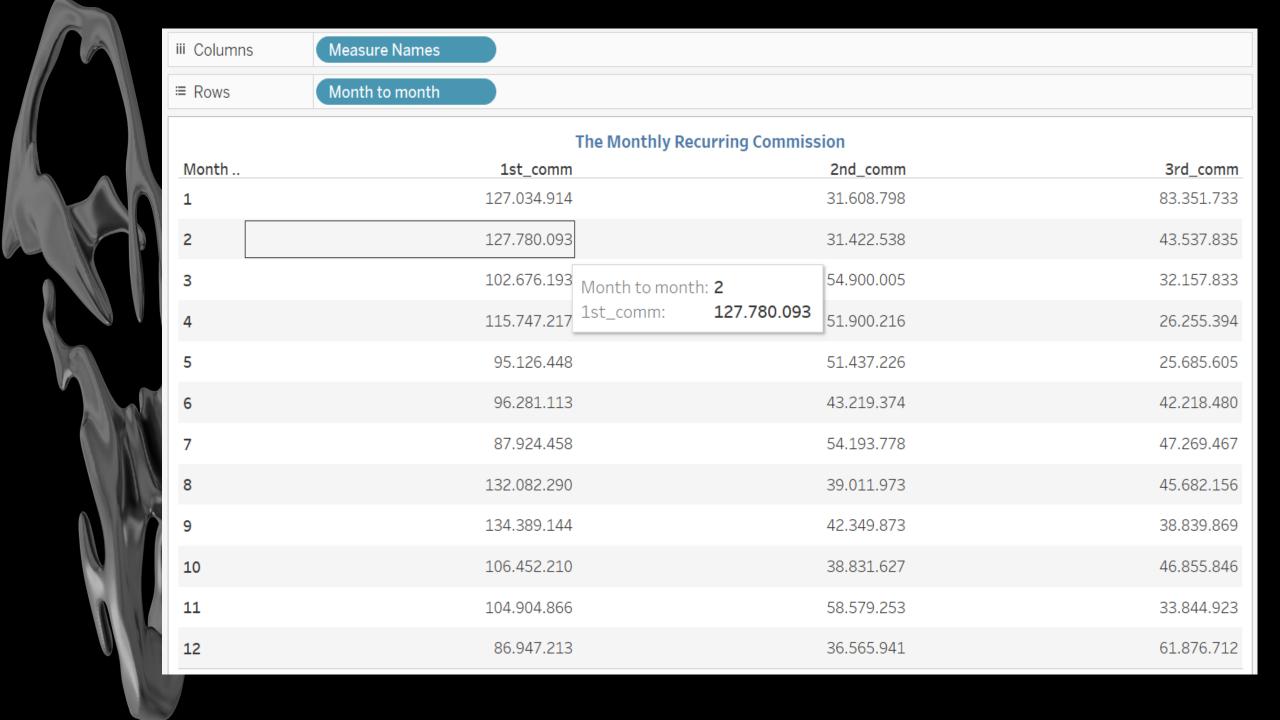


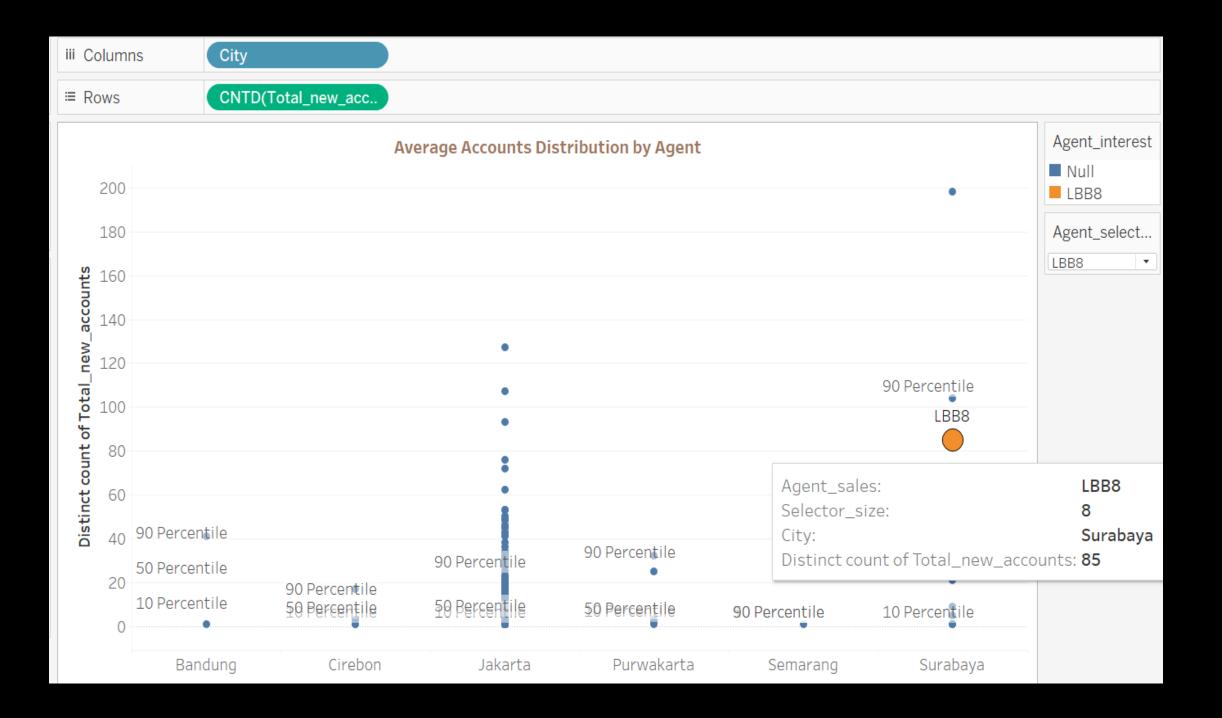


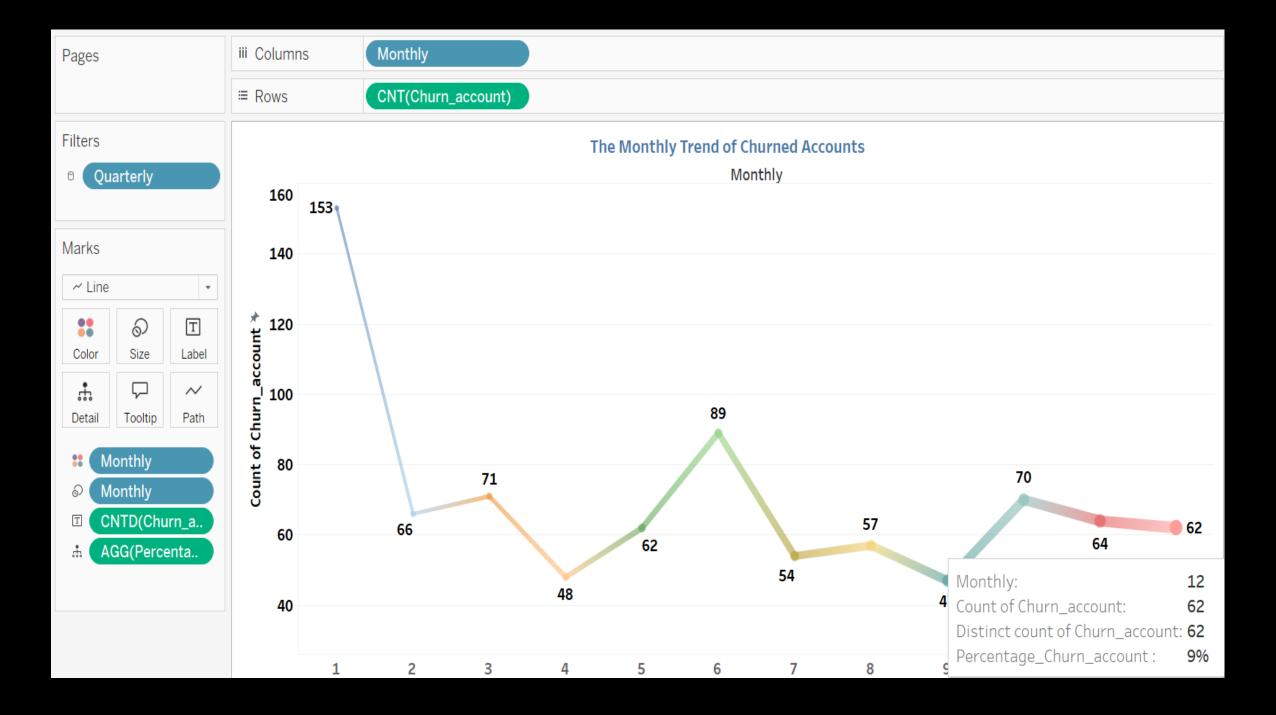


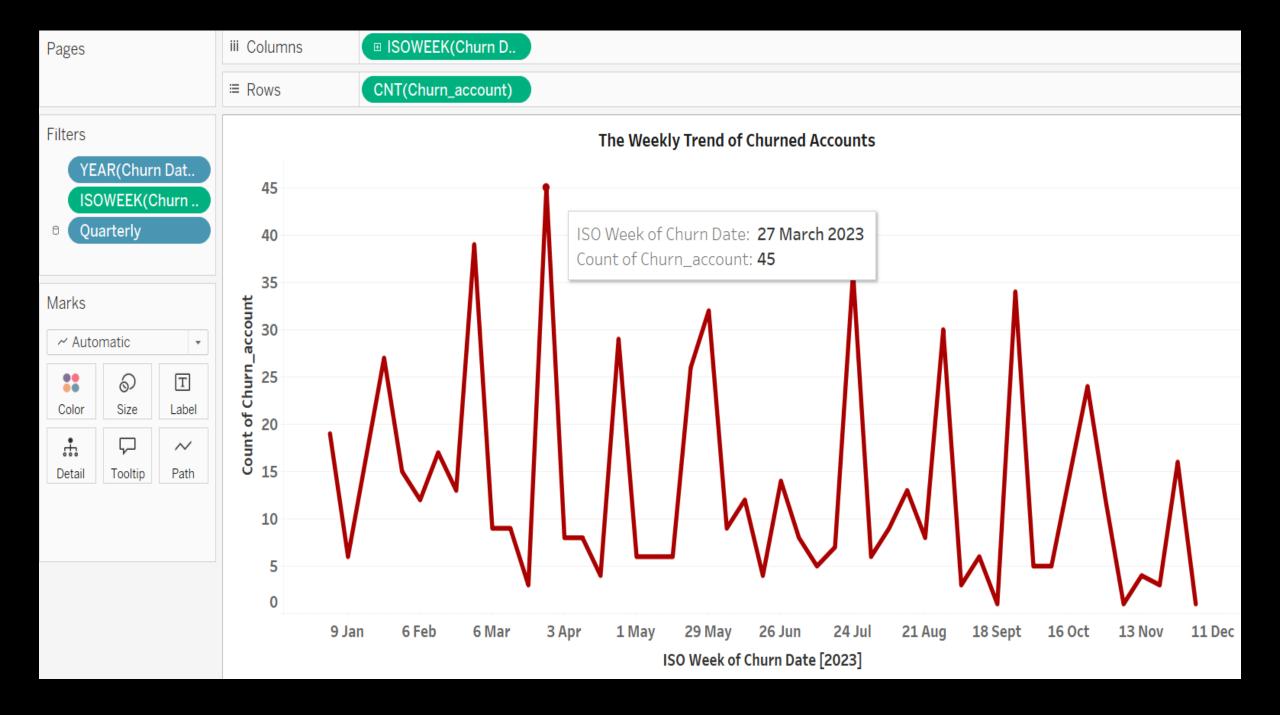






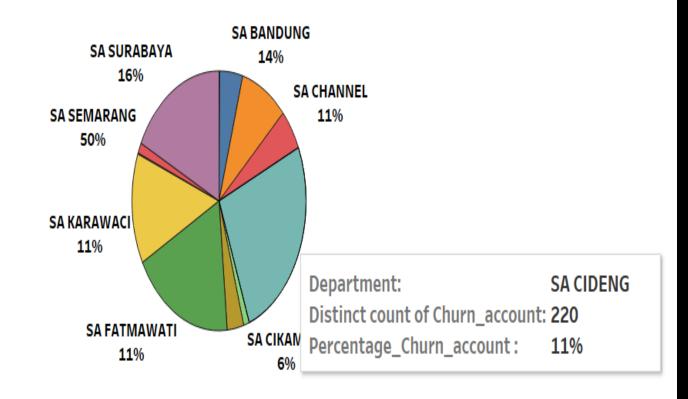






Filters Quarterly Marks O Pie 00 6 T Color Size Label \Box Tooltip Angle Detail Department CNTD(Churn_a.. CNTD(Churn_a.. T Department AGG(Percentag..

The Churned Accounts by Department



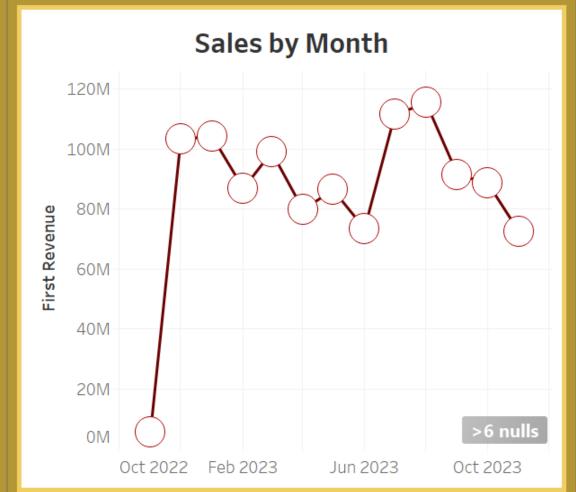
AREA SALES DASHBOARD

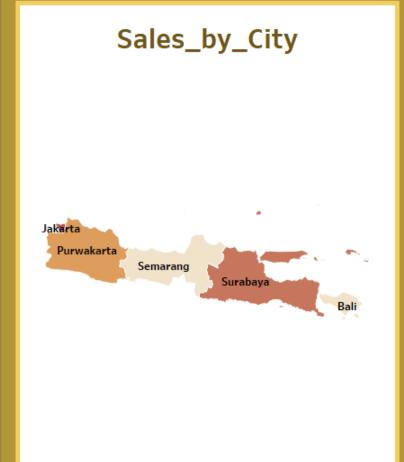
Quarte.. (All)

of New Accounts 3.702

1st Month Revenue 1.116.870.467

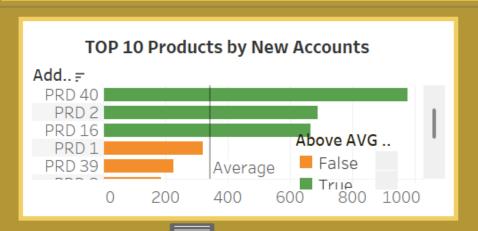
1st Month Commission 1.317.346.159

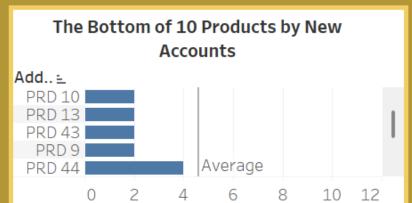




Although Jakarta recorded the highest number of new accounts (3055), with over 300 sales agents, its productivity was not proportional. In contrast, Surabaya achieved 518 new accounts with only around 11-12 agents, resulting in a much higher productivity score of 51. Meanwhile, the highest revenue was recorded in August and the lowest in November, with 72.343.666

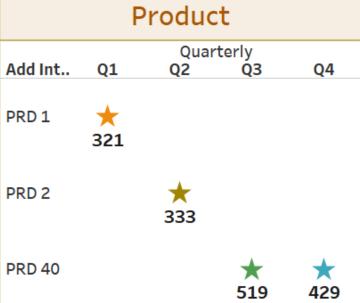
Sales by Product Dashboard

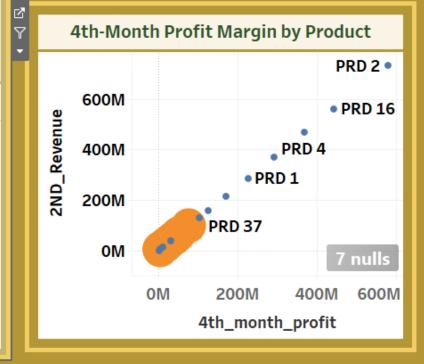




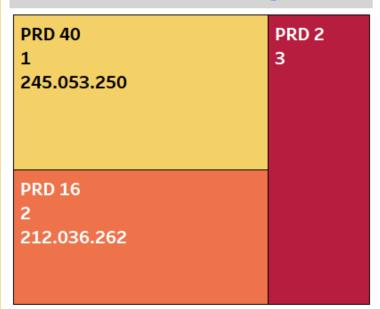
Product 2 dominates nearly all segments, with new accounts and revenue, the best-selling product in Q3, and the highest profit margin. The bottom of some products have an average revenue of more than 250k, it takes further customer segmentation and marketing strategy to delve into the cause of this product to be a favourite

The Quarterly Best - Selling Product

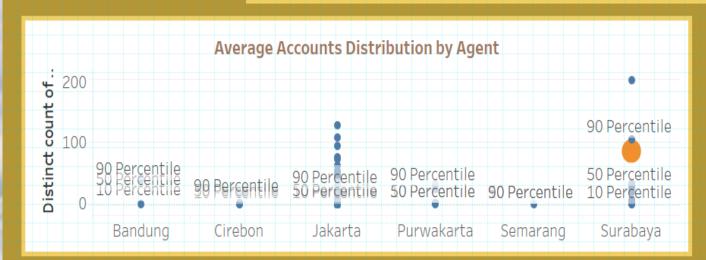




The TOP 3 Revenue Generating Products



Revenue - New Accounts - Commission Sales Dashboard



Agent.. LBB8

Agent_interest

Null I BB8 SA Cideng has the highest revenue.
Furthermore, a few of the sales agents in
Jakarta have an average of more than
percentile- 90 of the account distribution, while
Surabaya is served by a single agent,
nevertheless, the commission amount
consistently decline over the 1st, 4th, and 7th ...

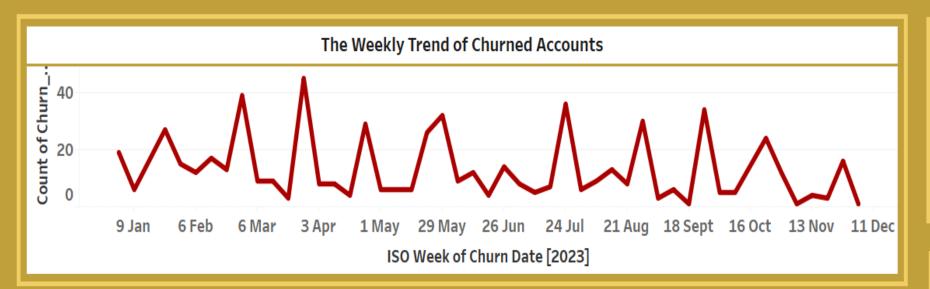


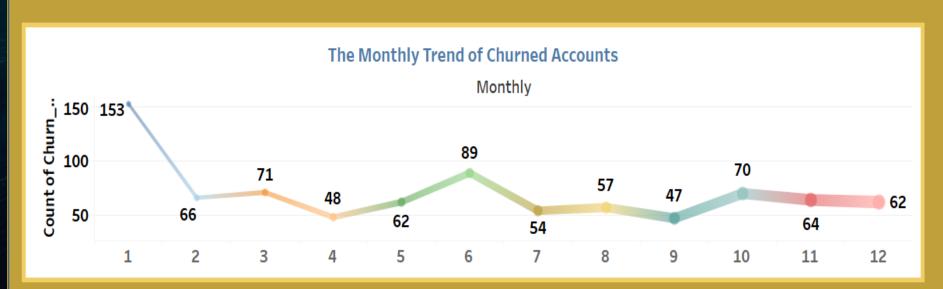


The Monthly Recurring Commission

Month	1st_comm	2nd_comm	3rd_comm
1	127.034.914	31.608.798	83.351.733
2	127.780.093	31.422.538	43.537.835
3	102.676.193	54.900.005	32.157.833
4	115.747.217	51.900.216	26.255.394
5	95.126.448	51.437.226	25.685.605
6	96.281.113	43.219.374	42.218.480
7	87.924.458	54.193.778	47.269.467
8	132.082.290	39.011.973	45.682.156
9	134.389.144	42.349.873	38.839.869
10	106.452.210	38.831.627	46.855.846
11	104.904.866	58.579.253	33.844.923
12	86.947.213	36.565.941	61.876.712

Churn Dashboard





In the opening year, January had 153 churned accounts, following 89 accounts in June. The number of churned accounts rose steadily on a weekly basis during the period from the end of February through March, and SA Cideng had the biggest contribution to churned accounts

