

RFM Analysis Using Python

```
In [1]: import pandas as pd
import plotly.express as px
import plotly.io as pio
import plotly.graph_objects as go
%matplotlib inline
import warnings
warnings.simplefilter(action='ignore', category=FutureWarning)

pio.templates.default = "plotly_white"

data = pd.read_csv(r"C:\Users\maddh\Downloads\rfm_data.csv")
print(data.head())
```

	CustomerID	PurchaseDate	TransactionAmount	ProductInformation	OrderID \
0	8814	2023-04-11	943.31	Product C	890075
1	2188	2023-04-11	463.70	Product A	176819
2	4608	2023-04-11	80.28	Product A	340062
3	2559	2023-04-11	221.29	Product A	239145
4	9482	2023-04-11	739.56	Product A	194545

	Location
0	Tokyo
1	London
2	New York
3	London
4	Paris

```
In [2]: from datetime import datetime

# Ensure 'PurchaseDate' is in datetime format
data['PurchaseDate'] = pd.to_datetime(data['PurchaseDate'])

# Calculate Recency (days since last purchase)
today_date = datetime.now()
data['Recency'] = (today_date - data['PurchaseDate']).dt.days # Apply .dt.days cor

# Calculate Frequency (Number of orders per Customer)
frequency_data = data.groupby('CustomerID')['OrderID'].count().reset_index()
frequency_data.rename(columns={'OrderID': 'Frequency'}, inplace=True)

# Merge Frequency Data
data = data.merge(frequency_data, on='CustomerID', how='left')

# Calculate Monetary Value (Total spending per Customer)
monetary_data = data.groupby('CustomerID')['TransactionAmount'].sum().reset_index()
monetary_data.rename(columns={'TransactionAmount': 'MonetaryValue'}, inplace=True)

# Merge Monetary Value Data
data = data.merge(monetary_data, on='CustomerID', how='left')
```

In [3]: `print(data.head())`

	CustomerID	PurchaseDate	TransactionAmount	ProductInformation	OrderID \
0	8814	2023-04-11	943.31	Product C	890075
1	2188	2023-04-11	463.70	Product A	176819
2	4608	2023-04-11	80.28	Product A	340062
3	2559	2023-04-11	221.29	Product A	239145
4	9482	2023-04-11	739.56	Product A	194545

	Location	Recency	Frequency	MonetaryValue
0	Tokyo	666	1	943.31
1	London	666	1	463.70
2	New York	666	1	80.28
3	London	666	1	221.29
4	Paris	666	1	739.56

In [4]: *# Define scoring criteria for each RFM value*
`recency_scores = [5, 4, 3, 2, 1] # Higher score for lower recency (more recent)`
`frequency_scores = [1, 2, 3, 4, 5] # Higher score for higher frequency`
`monetary_scores = [1, 2, 3, 4, 5] # Higher score for higher monetary value`

Calculate RFM scores
`data['RecencyScore'] = pd.cut(data['Recency'], bins=5, labels=recency_scores)`
`data['FrequencyScore'] = pd.cut(data['Frequency'], bins=5, labels=frequency_scores)`
`data['MonetaryScore'] = pd.cut(data['MonetaryValue'], bins=5, labels=monetary_score)`

In [5]: *# Convert RFM scores to numeric type*
`data['RecencyScore'] = data['RecencyScore'].astype(int)`
`data['FrequencyScore'] = data['FrequencyScore'].astype(int)`
`data['MonetaryScore'] = data['MonetaryScore'].astype(int)`

In [6]: *# Calculate RFM score by combining the individual scores*
`data['RFM_Score'] = data['RecencyScore'] + data['FrequencyScore'] + data['MonetaryS`

Create RFM segments based on the RFM score
`segment_labels = ['Low-Value', 'Mid-Value', 'High-Value']`
`data['Value Segment'] = pd.qcut(data['RFM_Score'], q=3, labels=segment_labels)`

In [7]: `print(data.head())`

	CustomerID	PurchaseDate	TransactionAmount	ProductInformation	OrderID	\
0	8814	2023-04-11	943.31	Product C	890075	
1	2188	2023-04-11	463.70	Product A	176819	
2	4608	2023-04-11	80.28	Product A	340062	
3	2559	2023-04-11	221.29	Product A	239145	
4	9482	2023-04-11	739.56	Product A	194545	

	Location	Recency	Frequency	MonetaryValue	RecencyScore	FrequencyScore	\
0	Tokyo	666	1	943.31	1	1	
1	London	666	1	463.70	1	1	
2	New York	666	1	80.28	1	1	
3	London	666	1	221.29	1	1	
4	Paris	666	1	739.56	1	1	

	MonetaryScore	RFM_Score	Value	Segment
0	2	4	Low-Value	
1	1	3	Low-Value	
2	1	3	Low-Value	
3	1	3	Low-Value	
4	2	4	Low-Value	

```
In [26]: # RFM Segment Distribution
segment_counts = data['Value Segment'].value_counts().reset_index()
segment_counts.columns = ['Value Segment', 'Count']

pastel_colors = px.colors.qualitative.Pastel

# Create the bar chart
fig_segment_dist = px.bar(segment_counts, x='Value Segment', y='Count',
                           color='Value Segment', color_discrete_sequence=pastel_col,
                           title='RFM Value Segment Distribution')

# Update the Layout
fig_segment_dist.update_layout(xaxis_title='RFM Value Segment',
                                yaxis_title='Count',
                                showlegend=False,
                                width=800, # Adjust the width of the figure
                                height=400)

# Show the figure
fig_segment_dist.show()
fig_segment_dist.write_html("segment_distribution.html")
```

In [17]: *# Create a new column for RFM Customer Segments*

```
data['RFM Customer Segments'] = ''

# Assign RFM segments based on the RFM score
data.loc[data['RFM_Score'] >= 9, 'RFM Customer Segments'] = 'Champions'
data.loc[(data['RFM_Score'] >= 6) & (data['RFM_Score'] < 9), 'RFM Customer Segments'] = 'Potential Champions'
data.loc[(data['RFM_Score'] >= 5) & (data['RFM_Score'] < 6), 'RFM Customer Segments'] = 'Potential Loyalists'
data.loc[(data['RFM_Score'] >= 4) & (data['RFM_Score'] < 5), 'RFM Customer Segments'] = 'Loyalists'
data.loc[(data['RFM_Score'] >= 3) & (data['RFM_Score'] < 4), 'RFM Customer Segments'] = 'Lost'

# Print the updated data with RFM segments
print(data[['CustomerID', 'RFM Customer Segments']].head(10))
```

	CustomerID	RFM Customer Segments
0	8814	Can't Lose
1	2188	Lost
2	4608	Lost
3	2559	Lost
4	9482	Can't Lose
5	8483	Lost
6	8317	Potential Loyalists
7	6911	Lost
8	8993	Lost
9	3519	Lost

In [27]: `segment_product_counts = data.groupby(['Value Segment', 'RFM Customer Segments']).s`

```
segment_product_counts = segment_product_counts.sort_values('Count', ascending=False)
fig_treemap_segment_product = px.treemap(segment_product_counts,
```

```

path=['Value Segment', 'RFM Customer Segme
values='Count',
color='Value Segment', color_discrete_sequ
title='RFM Customer Segments by Value')

# Update layout to set figure size
fig_treemap_segment_product.update_layout(
    width=800, # Adjust the width of the figure
    height=500 # Adjust the height of the figure
)

fig_treemap_segment_product.show()
fig_treemap_segment_product.write_html("treemap_segment_product.html")

```

In [29]:

```

# Filter the data to include only the customers in the Champions segment
champions_segment = data[data['RFM Customer Segments'] == 'Champions']

fig = go.Figure()
fig.add_trace(go.Box(y=champions_segment['RecencyScore'], name='Recency'))
fig.add_trace(go.Box(y=champions_segment['FrequencyScore'], name='Frequency'))
fig.add_trace(go.Box(y=champions_segment['MonetaryScore'], name='Monetary'))

fig.update_layout(title='Distribution of RFM Values within Champions Segment',
                  yaxis_title='RFM Value',

```

```
        showlegend=True,  
        width=800, # Adjust the width of the figure  
        height=400 )  
  
fig.show()  
fig.write_html("champions_segment_rfm_distribution.html")
```

In [30]:

```
correlation_matrix = champions_segment[['RecencyScore', 'FrequencyScore', 'Monetary  
  
# Visualize the correlation matrix using a heatmap  
fig_heatmap = go.Figure(data=go.Heatmap(  
    z=correlation_matrix.values,  
    x=correlation_matrix.columns,  
    y=correlation_matrix.columns,  
    colorscale='RdBu',  
    colorbar=dict(title='Correlation'))))  
  
# Update layout to set figure size  
fig_heatmap.update_layout(  
    title='Correlation Matrix of RFM Values within Champions Segment',  
    width=800, # Adjust the width of the figure  
    height=500 ) # Adjust the height of the figure  
  
fig_heatmap.show()  
fig_heatmap.write_html("champions_segment_rfm_heatmap.html")
```