In [102	# Import packages import pandas as PD import numpy as NP import matplotlib.pyplot as PLT import seaborn as SNS # Input data files are available in the "/input/" directory import time, warnings import datetime as DT #Visualizations import matplotlib pyplot as PLT
In [103 Out[103]:	<pre>import matplotlib.pyplot as PLT from pandas.plotting import scatter_matrix %matplotlib inline import seaborn as SNS warnings.filterwarnings("ignore") #Load the dataset retail_DF = PD.read_csv('/kaggle/input/customer-segmentation/customer segmintation.csv',encoding="ISO-8859-1",dtype={'CustomerID': str,'InvoiceID': str}) retail_DF.head() InvoiceNo StockCode</pre>
In [105	<pre>retail_DF.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 541909 entries, 0 to 541908 Data columns (total 8 columns): # Column Non-Null Count Dtype</class></pre>
	<pre># Is USA on file? retail_USA = retail_DF[retail_DF['Country']=='USA'] #check the shape retail_USA.shape</pre>
	So I'll choose Germany to do the RFM analysis Prepare the Data تانا البيانات # Remove canceled orders
Out[108]:	# interior land lan
Out[109]: In [110	<pre>#restrict the data to one full year because it's better to use a metric per Months or Years in RFM retail_G = retail_G[retail_G['InvoiceDate']>= "2010-1-12"]</pre>
Out[110]: In [111…	<pre>#check the shape retail_G.shape (4757, 8) print("Summary") print("") #exploring the unique values of each attribute print("Number of transactions: ", retail_G['InvoiceNo'].nunique()) print("Number of products bought: ",retail_G['StockCode'].nunique())</pre>
	<pre>print("Number of customers:", retail_G['CustomerID'].nunique()) print("Percentage of customers NA: ", round(retail_G['CustomerID'].isnull().sum() * 100 / len(retail_DF),2),"%") Summary</pre>
In [112 Out[112]:	#المجودة عندنا المجودة عندنا المجودة عندنا المجودة عندنا retail_G['InvoiceDate'].max()
	# اغر تارین موجود عو 2011-8-9 اغر تارین موجود عو 2011-8-9 امستخدمه کمرجع (استخدمه کمرجع الله علائل کذا حنستخدمه کمرجع (استخدمه کمرجع الله الله الله الله الله الله الله الل
In [115 Out[115]:	# Double check the data # "Double check the data # "Double check the data # "Double check the data retail_G.head() InvoiceNo StockCode Description Quantity InvoiceDate UnitPrice CustomerID Country date
In [116 Out[116]:	105888 545295 22354 RETROSPOT PADDED SEAT CUSHION 20 3/1/2011 12:09 3.75 12709 Germany 2011-03-01 105889 545295 22333 RETROSPOT PARTY BAG + STICKER SET 30 3/1/2011 12:09 1.65 12709 Germany 2011-03-01 # Group by customers and check last date of purshace # العملاء المعالاء العملاء العم
In [118	<pre>3 12472 2011-07-24 4 12473 2011-08-17 # Calculate Recency recency_DF['Recency'] = recency_DF['LastPurshaceDate'].apply(lambda x: (now - x).days)</pre> recency_DF.head()
Out[118]:	CustomerID LastPurshaceDate Recency 0 12426 2011-05-29 72 1 12468 2011-06-05 65 2 12471 2011-09-22 -44 3 12472 2011-07-24 16 4 12473 2011-08-17 -8
In [119	How many invoices are registered by the same customer # Drop duplicates retail_G_copy = retail_G retail_G_copy.drop_duplicates(subset=['InvoiceNo', 'CustomerID'], keep="first", inplace=True) #calculate frequency of purchases frequency_DF = retail_G_copy.groupby(by=['CustomerID'], as_index=False)['InvoiceNo'].count() frequency_DF.columns = ['CustomerID', 'Frequency']
Out[119]:	EustomerID Frequency 0 12426 1 1 12468 1 2 12471 14 3 12472 3 4 12473 2
In [120	<pre>Monetary How much money did the customer spent over time? # Create column total cost retail_G['TotalCost'] = retail_G['Quantity'] * retail_G['UnitPrice'] monetary_DF = retail_G.groupby(by='CustomerID', as_index=False).agg({'TotalCost': 'sum'})</pre>
Out[120]:	0 12426 17.70 1 12468 13.20 2 12471 856.68 3 12472 45.30
In [121	Treate RFM Table # Merge recency dataframe with frequency dataframe # frequency dataframe و recency dataframe و المسوى دمن بين temp_DF = recency_DF.merge(frequency_DF, on='CustomerID') temp_DF.head()
Out[121]:	<pre># Merge with monetary dataframe to get a table with the 3 columns rfm_DF = temp_DF.merge(monetary_DF, on='CustomerID') # Use CustomerID as index rfm_DF.set_index('CustomerID', inplace=True) # Check the head rfm_DF.head() LastPurshaceDate Recency Frequency Monetary CustomerID</pre>
	12426 2011-05-29 72 1 17.70 12468 2011-06-05 65 1 13.20 12471 2011-09-22 -44 14 856.68 12472 2011-07-24 16 3 45.30 12473 2011-08-17 -8 2 28.20 RFM Table Correctness verification
In [122 Out[122]: In [123 Out[123]:	207305 554985 20665 RED RETROSPOT PURSE 6 5/29/2011 12:26 2.95 12426 Germany 2011-05-29 17.7 (now - DT.date(2011,5,29)).days == 72
	Customer segments with RFM Model قر ائح من العملاء مع RFM We assign a score from 1 to 4 to Recency, Frequency and Monetary. Four is the best/highest value, and one is the lowest/worst value عنقسر العملاء حسب درجا تهر من 1 الى 4.4 تعتبر افضل و اعلى درجه, 1 تعتبر ادنى و اسوء درجه
In [124 Out[124]:	quantiles
In [125 Out[125]:	<pre>quantiles.to_dict() {'Recency': {0.25: -41.0, 0.5: -16.0, 0.75: 29.25}, 'Frequency': {0.25: 1.0, 0.5: 2.0, 0.75: 4.0}, 'Monetary': {0.25: 25.53, 0.5: 53.74, 0.75: 140.31}} - Creation of RFM Segments Create two segmentation classes since, high recency is bad, while high frequency and monetary value is good</pre>
In [126	<pre># Arguments (x = value, p = recency, monetary_value, frequency, d = quartiles dict) def RScore(x,p,d): if x <= d[p][0.25]: return 4 elif x <= d[p][0.50]: return 3 elif x <= d[p][0.75]: return 2 else: return 1 # Arguments (x = value, p = recency, monetary_value, frequency, k = quartiles dict) def FMScore(x,p,d): if x <= d[p][0.25]: return 1 elif x <= d[p][0.50]: return 2</pre>
	<pre>elif x <= d[p][0.75]: return 3 else: return 4 #create rfm segmentation table rfm_segmentation = rfm_DF rfm_segmentation['R_Quartile'] = rfm_segmentation['Recency'].apply(RScore, args=('Recency', quantiles,)) rfm_segmentation['F_Quartile'] = rfm_segmentation['Frequency'].apply(FMScore, args=('Frequency', quantiles,)) rfm_segmentation['M_Quartile'] = rfm_segmentation['Monetary'].apply(FMScore, args=('Monetary', quantiles,))</pre>
Out[126]:	LastPurshaceDate Recency Frequency Monetary R_Quartile F_Quartile M_Quartile
In [127 Out[127]:	rfm_segmentation['RFMScore'] = rfm_segmentation.R_Quartile.map(str) \
In [128 Out[128]:	CustomerID 12471 2011-09-22 -44 14 856.68 4 4 4 444 12619 2011-09-23 -45 7 821.77 4 4 4 444 12621 2011-09-19 -41 13 467.90 4 4 4 444 12709 2011-09-29 -51 9 297.30 4 4 444
In [129 Out[129]:	12720 2011-09-20 -42 14 264.78 4 4 4 444 12647 2011-09-22 -44 7 252.53 4 4 4 444 # Total RFM score rfm_segmentation['Total Score'] = rfm_segmentation['R_Quartile'] + rfm_segmentation['F_Quartile'] + \ rfm_segmentation.head() LastPurshaceDate Recency Frequency Monetary R_Quartile F_Quartile M_Quartile RFMScore Total Score
Out[129]:	LastPurshaceDate Recency Frequency Monetary R_Quartile R_Quartile RFMScore Total Score CustomerID 12426 2011-05-29 72 1 17.70 1 1 111 3 12468 2011-06-05 65 1 13.20 1 1 111 3 12471 2011-09-22 -44 14 856.68 4 4 4 444 12 12472 2011-07-24 16 3 45.30 2 3 2 232 7 12473 2011-08-17 -8 2 28.20 2 2 2 22 6 rfm_segmentation.groupby('Total Score')['Monetary'].mean().plot(kind='bar', colormap='Blues_r')
In [130 Out[130]:	<aver: score!="" vlahel-!tetal=""></aver:>
In [131 Out[131]:	
In [132 Out[132]:	rfm_segmentation.groupby('Total Score')['Recency'].mean().plot(kind='bar', colormap='Blues_r') <axes: xlabel="Total Score"></axes:>
	100 - 80 - 60 - 40 - 20 - 40 - 20 - 40 - 50 - 70 Total Score
In [133 Out[133]:	CustomerID 12426 2011-05-29 72 1 17.70 1 1 1 111 3 111 12468 2011-06-05 65 1 13.20 1 1 1 111 3 111
In [134	12471 2011-09-22 -44 14 856.68 4 4 4 4 444 12 444 12472 2011-07-24 16 3 45.30 2 3 2 232 7 232 12473 2011-08-17 -8 2 28.20 2 2 2 222 6 222 seg_map = { r'[1-2][1-2]': 'hibernating', r'[1-2][3-4]': 'at_Risk', r'[1-2]5': 'cant_loose', r'3[1-2]': 'about_to_sleep', r'31': 'need_attention',
	<pre>r'33': 'need_attention', r'[3-4][4-5]': 'loyal_customers', r'41': 'promising', r'51': 'new_customers', r'[4-5][2-3]': 'potential_loyalists', r'5[4-5]': 'champions' } rfm_DF['Segment'] = rfm_DF['Segment'].replace(seg_map, regex=True) rfm_DF.reset_index(inplace=True) rfm_DF.head()</pre>
Out[134]:	0 12426 2011-05-29 72 1 17.70 1 1 1 111 3 hibernating1 1 12468 2011-06-05 65 1 13.20 1 1 111 3 hibernating1 2 12471 2011-09-22 -44 14 856.68 4 4 4 444 12 loyal_customers4 3 12472 2011-07-24 16 3 45.30 2 3 2 232 7 at_Risk2 4 12473 2011-08-17 -8 2 28.20 2 2 2 222 6 hibernating2
In [135	- How many customers do we have in each segment? Description: Descri
	Best Customers: 6 Loyal Customers: 15 Big Spenders: 18 Almost Lost: 0 Lost Customers: 0 Lost Cheap Customers: 7