Cài đặt PySpark

```
!pip install pyspark

Requirement already satisfied: pyspark in /usr/local/lib/python3.10/dist-packages (3.5.3)
Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.10/dist-packages (from pyspark) (0.10.9.7)

Chuyển đổi file excel sang csv

!pip install gdown pyspark

Requirement already satisfied: gdown in /usr/local/lib/python3.10/dist-packages (5.2.0)
```

```
Requirement already satisfied: gdown in /usr/local/lib/python3.10/dist-packages (5.2.0)
Requirement already satisfied: pyspark in /usr/local/lib/python3.10/dist-packages (3.5.3)
Requirement already satisfied: beautifulsoup4 in /usr/local/lib/python3.10/dist-packages (from gdown) (4.12.3)
Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-packages (from gdown) (3.16.1)
Requirement already satisfied: requests[socks] in /usr/local/lib/python3.10/dist-packages (from gdown) (2.32.3)
Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from gdown) (4.66.6)
Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.10/dist-packages (from pyspark) (0.10.9.7)
Requirement already satisfied: soupsieve>1.2 in /usr/local/lib/python3.10/dist-packages (from beautifulsoup4->gdown) (2.6)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests[socks]->gdown) (3.4.0)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests[socks]->gdown) (2.2.3)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests[socks]->gdown) (2024.8.30)
Requirement already satisfied: PySocks!=1.5.7,>=1.5.6 in /usr/local/lib/python3.10/dist-packages (from requests[socks]->gdown) (1.7.1)
```

Phân tích dữ liêu cơ bản

```
from pyspark.sql import SparkSession
from datetime import datetime
from pyspark.sql.types import StructType, StructField, StringType, IntegerType, DoubleType, TimestampType
spark = SparkSession.builder.appName("Optimized RDD Data Preprocessing").getOrCreate()
# Load the CSV file
df = spark.read.csv('/content/drive/MyDrive/CÜA TÔI/NHOM2_CK/OnlineRetail.csv', header=True, inferSchema=True)
# Convert DataFrame to RDD
rdd = df.rdd
# Filter out rows where InvoiceNo starts with "C"
filtered_rdd = rdd.filter(lambda row: not row.InvoiceNo.startswith("C"))
# Parse the date correctly
def parse_date(row):
    trv:
       date = datetime.strptime(row['InvoiceDate'], '%d-%m-%Y %H:%M')
   except:
        date = None
   return (row['InvoiceNo'], row['StockCode'], row['Description'], row['Quantity'],
            row['InvoiceDate'], row['UnitPrice'], row['CustomerID'], row['Country'], date)
```

```
NHOM2_CK.ipynb - Colab
# Map the RDD to include the parsed date
rdd_with_date = filtered_rdd.map(parse_date)
# Define the schema
schema = StructType([
   StructField("InvoiceNo", StringType(), True),
   StructField("StockCode", StringType(), True),
   StructField("Description", StringType(), True),
   StructField("Quantity", IntegerType(), True),
   StructField("InvoiceDate", StringType(), True),
   StructField("UnitPrice", DoubleType(), True),
   StructField("CustomerID", StringType(), True),
   StructField("Country", StringType(), True),
   StructField("date", TimestampType(), True)
])
# Create DataFrame from the RDD with the correct schema
df_processed = spark.createDataFrame(rdd_with_date, schema)
df_processed.show(5, truncate=False)
    |InvoiceNo|StockCode|Description
                                                         |Quantity|InvoiceDate |UnitPrice|CustomerID|Country
                                                                                                                        date
     536365
              |85123A | WHITE HANGING HEART T-LIGHT HOLDER |6
                                                                    |01-12-2010 08:26|2.55
                                                                                               17850
                                                                                                          |United Kingdom|2010-12-01 08:26:
     536365
              71053
                        WHITE METAL LANTERN
                                                                    |01-12-2010 08:26|3.39
                                                                                               17850
                                                                                                          |United Kingdom|2010-12-01 08:26:
     536365
              84406B
                        CREAM CUPID HEARTS COAT HANGER
                                                                    |01-12-2010 08:26|2.75
                                                                                               17850
                                                                                                          |United Kingdom|2010-12-01 08:26:
                                                           18
     1536365
              184029G
                        |KNITTED UNION FLAG HOT WATER BOTTLE|6
                                                                    |01-12-2010 08:26|3.39
                                                                                               17850
                                                                                                          |United Kingdom|2010-12-01 08:26:
             84029E
                      RED WOOLLY HOTTIE WHITE HEART.
                                                                    |01-12-2010 08:26|3.39
                                                                                               17850
                                                                                                          |United Kingdom|2010-12-01 08:26:
    536365
    only showing top 5 rows
```

```
from pyspark.sql import functions as F
invoice_totals = df_processed.withColumn("TotalCost", F.col("Quantity") * F.col("UnitPrice")) \
                                                                      .groupBy("InvoiceNo") \
                                                                      .agg(F.sum("TotalCost").alias("TotalCost"))
revenue\_per\_customer = df\_processed.withColumn("Revenue", F.col("Quantity") * F.col("UnitPrice")) \setminus (a.c., b.c., b.c.,
                                                                                      .groupBy("CustomerID") \
                                                                                      .agg(F.sum("Revenue").alias("TotalRevenue"))
revenue_per_country = df_processed.withColumn("Revenue", F.col("Quantity") * F.col("UnitPrice")) \
                                                                                   .groupBy("Country") \
                                                                                   .agg(F.sum("Revenue").alias("TotalRevenue"))
orders_per_day = df_processed.filter(F.col("date").isNotNull()) \
                                                                      .withColumn("DayOfWeek", F.date_format(F.col("date"), "EEEE")) \
                                                                       .groupBy("DayOfWeek") \
                                                                      .count() \
                                                                       .orderBy("count", ascending=False)
average_order_value = invoice_totals.agg(F.avg("TotalCost").alias("AverageOrderValue"))
top_selling_products = df_processed.groupBy("StockCode", "Description") \
                                                                                      .agg(F.sum("Quantity").alias("TotalQuantity")) \
                                                                                      .orderBy("TotalQuantity", ascending=False) \
print("Tổng chi phí của từng hóa đơn:")
invoice_totals.show()
print("Doanh thu theo khách hàng:")
revenue_per_customer.show()
print("Doanh thu theo quốc gia:")
revenue_per_country.show()
print("Số lượng đơn hàng theo ngày trong tuần:")
orders_per_day.show()
print("Giá tri trung bình của đơn hàng:")
average_order_value.show()
print("Các sản phẩm bán chạy nhất:")
top_selling_products.show()
 <del>_</del>
```

```
NHOM2_CK.ipynb - Colab
    | AverageOrderValue|
    |482.44019325598475|
    Các sản phẩm bán chạy nhất:
    |StockCode|
                Description|TotalQuantity|
    +----+----
         23843 PAPER CRAFT , LIT...
         23166 MEDIUM CERAMIC TO...
                                         78033
        84077 WORLD WAR 2 GLIDE...
                                         55047
        85099B|JUMBO BAG RED RET...|
                                         48478
       85123A WHITE HANGING HEA...
                                         37603
                   POPCORN HOLDER
         22197
                                         36761
         84879 ASSORTED COLOUR B...
                                         36461
         21212 PACK OF 72 RETROS...
         23084 RABBIT NIGHT LIGHT
                                         30788
        22492 MINI PAINT SET VI...
                                         26633
df_processed.count()
→ 532621
df_processed.select('CustomerID').distinct().count()
→ 4340
from pyspark.sql.functions import countDistinct, desc
df_processed.groupBy('Country').agg(countDistinct('CustomerID').alias('country_count')).orderBy(desc('country_count')).show()
           Country|country_count|
     United Kingdom
                          3921
            Germany|
                              94
             Francel
                              87
                              30 l
              Spain
            Belgium|
                              25
         Switzerland|
                              21
                              19
           Portugal|
              Italy
                              14
            Finland|
                              12
            Austria
                              11
                              10
             Norway
            Denmark
                               9|
     Channel Islands
                               9
                               9
          Australia
         Netherlands
                               91
             Sweden
                               8
                               81
             Cyprus
                               8
              Japan
             Poland
                               6
                               4
             Greecel
    only showing top 20 rows
from pyspark.sql.functions import min, max
date_range = df_processed.select(min("date").alias("EarliestOrderDate"), max("date").alias("LatestOrderDate"))
date_range.show()
    | EarliestOrderDate| LatestOrderDate|
    |2010-12-01 08:26:00|2011-12-09 12:50:00|
    +----
```

Data preprocessing

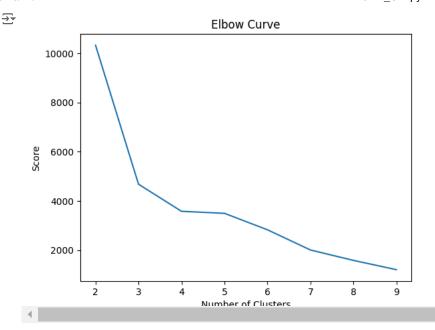
[] L, 12 ô bị ẩn

Chuẩn hóa dữ liệu

Xây dựng mô hình học máy K-means

Tìm số cum K

```
from pyspark.ml.clustering import KMeans
from pyspark.ml.evaluation import ClusteringEvaluator
import numpy as np
from tqdm import tqdm
cost = np.zeros(10)
evaluator = ClusteringEvaluator(predictionCol='prediction', featuresCol='standardized',metricName='silhouette', distanceMeasure='squaredEucl
for i in tqdm(range(2,10)):
    \label{lem:coler} KMeans\_algo=KMeans(featuresCol='standardized', k=i)
    KMeans_fit=KMeans_algo.fit(data_scale_output)
    output=KMeans_fit.transform(data_scale_output)
    cost[i] = KMeans_fit.summary.trainingCost
→ 100%| 8/8 [48:11<00:00, 361.42s/it]
import pandas as pd
import pylab as pl
df_cost = pd.DataFrame(cost[2:])
df_cost.columns = ["cost"]
new_col = range(2,10)
df_cost.insert(0, 'cluster', new_col)
pl.plot(df_cost.cluster, df_cost.cost)
pl.xlabel('Number of Clusters')
pl.ylabel('Score')
pl.title('Elbow Curve')
pl.show()
```



Training K-means với k=4

```
kmeans_algo=KMeans(featuresCol='standardized', k=4)
kmeans_fit=kmeans_algo.fit(data_scale_output)
```

Tiến hành dự đoán cụm

preds=kmeans_fit.transform(data_scale_output)
preds.show(5)

}	++		+	+	·	 	·
	recency	trequency	monetary_value	CustomerID	features	standardized	prediction
	9700860	2	389.440000000000005	16250	[9700860.0,2.0,38	[1.11772206754745	0
	16947300	4	702.25000000000001	15574	[1.69473E7,4.0,70	[1.95264865128936	0
	31218780	16	4805.1699999999994	15555	[3.121878E7,16.0,	[3.59699236231725	3
	31643100	15	2507.0699999999997	15271	[3.16431E7,15.0,2	[3.64588203062519]	3
	4586760	1	153.0	17714	[4586760.0,1.0,15	[0.52848127594295]	0
	++		+			+	·+
	only showi	ng top 5 i	rows				

Trực quan hóa kết quả phân cụm

```
import matplotlib.pyplot as plt
import seaborn as sns

df_viz = preds.select('recency', 'frequency', 'monetary_value', 'prediction')
df_viz = df_viz.toPandas()

avg_df = df_viz.groupby(['prediction'], as_index=False).mean()

list_value = ['recency', 'frequency', 'monetary_value']

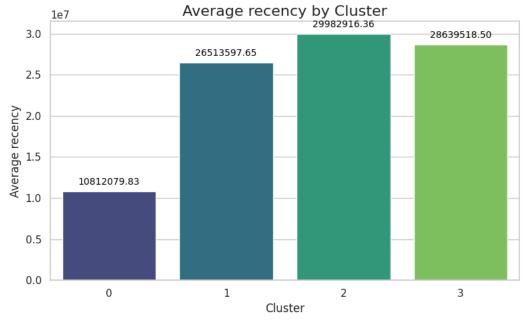
sns.set(style="whitegrid")

for i in list_value:
    plt.figure(figsize=(8, 5))
    bar_plot = sns.barplot(x='prediction', y=str(i), data=avg_df, palette='viridis')

plt.title(f'Average {i} by Cluster', fontsize=16)
    plt.xlabel('Cluster', fontsize=12)
    plt.ylabel(f'Average {i}', fontsize=12)
```

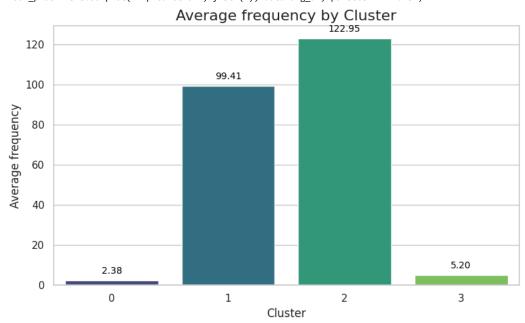
<ipython-input-25-59ccef2e5080>:15: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legor bar_plot = sns.barplot(x='prediction', y=str(i), data=avg_df, palette='viridis')



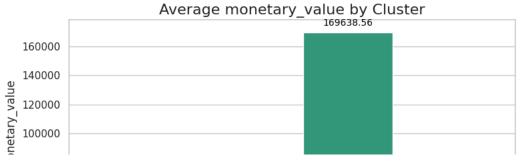
<ipython-input-25-59ccef2e5080>:15: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `lege bar_plot = sns.barplot(x='prediction', y=str(i), data=avg_df, palette='viridis')

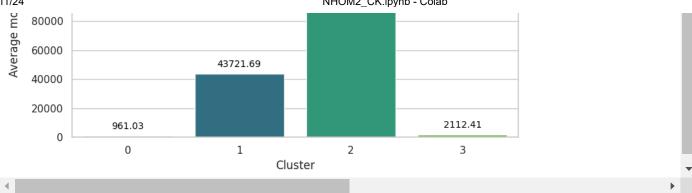


<ipython-input-25-59ccef2e5080>:15: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `lest the control of the co



bar_plot = sns.barplot(x='prediction', y=str(i), data=avg_df, palette='viridis')



RCM

 $\label{eq:df} \textit{df = spark.read.csv('/content/drive/MyDrive/C\dot{U}A T\^OI/NHOM2_CK/OnlineRetail.csv', header=True, inferSchema=True)} \\$

df.show(5)

\rightarrow	+									+
ت	InvoiceNo StockCode			Quantity	Invoi	ceDate	UnitPrice	CustomerID		Country
	536365 536365 536365	85123A 71053	WHITE HANGING HEA WHITE METAL LANTERN CREAM CUPID HEART	6	01-12-2010 01-12-2010 01-12-2010	08:26	3.39	17850	United	Kingdom Kingdom Kingdom
	536365 536365 +	84029G	KNITTED UNION FLA	6	01-12-2010 01-12-2010 01-12-2010	08:26	3.39	17850	United	Kingdom Kingdom

only showing top 5 rows

```
{\tt from\ pyspark.ml.recommendation\ import\ ALS}
best_model = cv_model.bestModel
predictions = best_model.transform(test_data)
rmse = evaluator.evaluate(predictions)
print(f"Best Model's RMSE on test data: {rmse}")
print(f"Best rank: {best_model.rank}")
print(f"Best maxIter: {best_model._java_obj.parent().getMaxIter()}")
print(f"Best regParam: {best_model._java_obj.parent().getRegParam()}")
best_model.userFactors.show()
best_model.itemFactors.show()
best_model.recommendForAllUsers(3).show()
→ Best Model's RMSE on test data: 33.66984784435793
     Best rank: 30
     Best maxIter: 10
     Best regParam: 0.01
     | id|
                   features
     |12350|[0.0, 0.084478065...|
     |12360|[0.19722603, 0.09...|
     |12370|[0.0, 0.0, 0.0914...|
     |12380|[0.0, 0.0, 0.4529...|
     12390|[0.08485217, 0.24...|
     |12410|[0.0, 0.0, 0.0, 0...|
     |12420|[0.0, 0.0, 0.0, 0...|
     |12430|[0.0138090905, 0....|
```