Azure Databricks Assignment

Submitted By: Aniroop Gupta DE Batch 1

1. Write Data to CSV, JSON, Parquet, delta:

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✓ ✓ 05:31 PM (<1s)
                                                                                                                                     Python 💠 []
   # Sample DataFrame
   data = [("Alice", 34), ("Bob", 45), ("Cathy", 25)]
   columns = ["Name", "Age"]
   df = spark.createDataFrame(data, columns)
 ▶ ■ df: pyspark.sql.dataframe.DataFrame = [Name: string, Age: long]
       ✓ 05:32 PM (1s)
   # 1. Write DataFrame to CSV
   df.write.option("header", "true").csv("/dbfs/mnt/path/to/outputs/csv")
▶ (1) Spark Jobs
       ✓ 05:32 PM (1s)
   # 2. Write DataFrame to JSON
   df.write.json("/dbfs/mnt/path/to/outputs/json")
▶ (1) Spark Jobs
       ✓ 05:32 PM (1s)
   # 3. Write DataFrame to Parquet
   df.write.parquet("/dbfs/mnt/path/to/outputs/parquet")
▶ (1) Spark Jobs
```

```
# 4. Write DataFrame to Delta format
    df.write.format("delta").save("/dbfs/mnt/path/to/outputs/delta")

**Notice**
**Notice*
```

2. Writing dataframe to Delta Table:

```
# Sample DataFrame
data = [("Alice", 34), ("Bob", 45), ("Cathy", 25)]
columns = ["Name", "Age"]
df = spark.createDataFrame(data, columns)

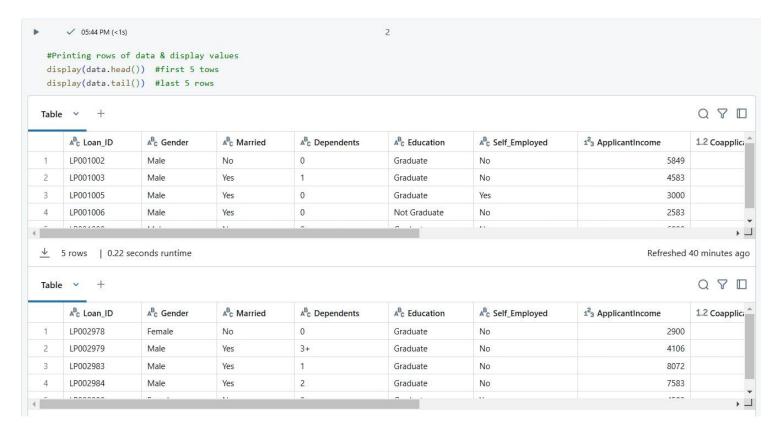
# Write DataFrame to Delta table
df.write.format("delta").mode("overwrite").save("/dbfs/mnt/path/to/delta_table")

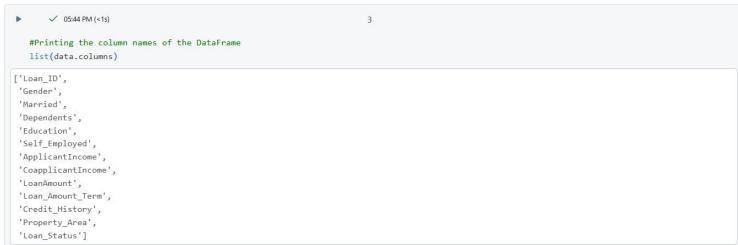
> (6) Spark Jobs

| df: pyspark.sql.dataframe.DataFrame = [Name: string, Age: long]
```

3. Exploratory data analysis (EDA) in Databricks:

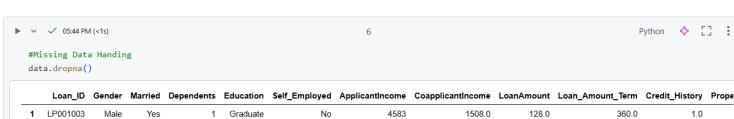
```
► ✓ ✓ 05:43 PM (<1s)
                                                                                                                  Python 💠
   import pandas as pd
   spark_df = spark.read.format("delta").load("dbfs:/user/hive/warehouse/loan_data")
   data = spark_df.toPandas()
   print(data)
▶ (1) Spark Jobs
 ▶ ■ spark_df: pyspark.sql.dataframe.DataFrame = [Loan_ID: string, Gender: string ... 11 more fields]
     Loan ID Gender Married ... Credit History Property Area Loan Status
0
    LP001002
                      No ...
1
    LP001003
              Male
                       Yes ...
             Male
                                        1.0
                     Yes ...
2
   LP001005
                                                   Urban
   LP001006 Male
                     Yes ...
                                       1.0
3
                                                  Urban
4
   LP001008 Male
                      No ...
                                        1.0
                                                   Urban
609 LP002978 Female
                                                   Rural
                      No ...
                                        1.0
610 LP002979 Male
                                        1.0
611 LP002983 Male Yes ...
                                        1.0
                                                    Urban
612 LP002984 Male
                     Yes ...
                                        1.0
                                                   Urban
613 LP002990 Female
                                        0.0 Semiurban
                     No ...
[614 rows x 13 columns]
```





√ 05:44 PM (<1s) 5 #Descriptive Statistical Measures of a DataFrame data.describe() </p>

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.00000	564.000000
mean	5403.459283	1621.245798	146.412162	342.00000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.00000	0.000000
25%	2877.500000	0.000000	100.000000	360.00000	1.000000
50%	3812.500000	1188.500000	128.000000	360.00000	1.000000
75%	5795.000000	2297.250000	168.000000	360.00000	1.000000
max	81000.000000	41667.000000	700.000000	480.00000	1.000000



	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	Applicantincome	Coapplicantincome	LoanAmount	Loan_Amount_Term	Credit_History	Prope
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	
5	LP001011	Male	Yes	2	Graduate	Yes	5417	4196.0	267.0	360.0	1.0	
609	LP002978	Female	No	0	Graduate	No	2900	0.0	71.0	360.0	1.0	
610	LP002979	Male	Yes	3+	Graduate	No	4106	0.0	40.0	180.0	1.0	
611	LP002983	Male	Yes	1	Graduate	No	8072	240.0	253.0	360.0	1.0	
612	LP002984	Male	Yes	2	Graduate	No	7583	0.0	187.0	360.0	1.0	
613	LP002990	Female	No	0	Graduate	Yes	4583	0.0	133.0	360.0	0.0	S
400 -	40											

480 rows × 13 columns

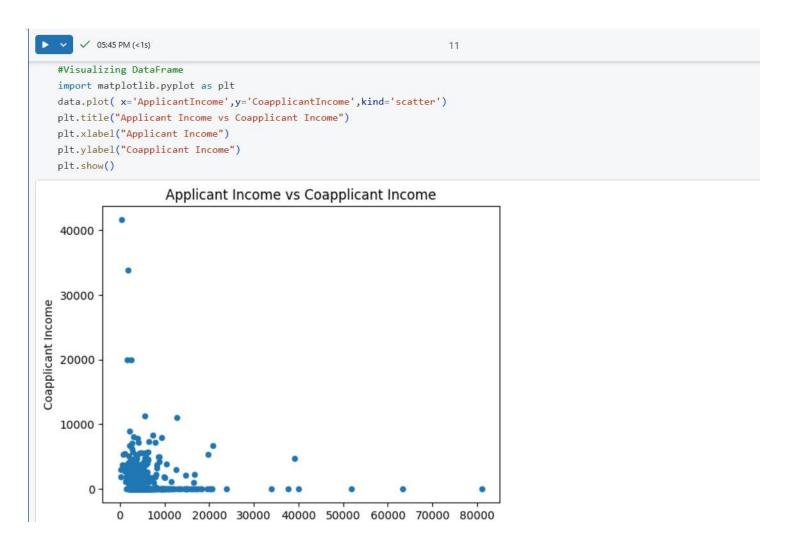
←

```
#Apply Function

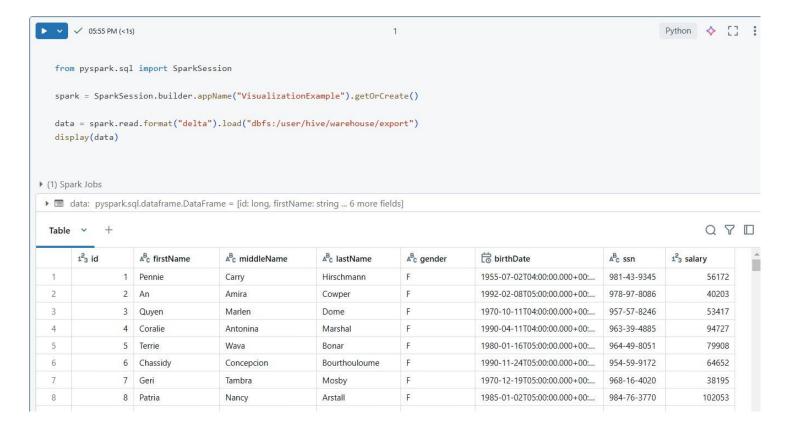
def fun(value):
    if value>3000:
        return 'Yes'
    else:
        return 'No'

data['newColumn'] = data['ApplicantIncome'].apply(fun)
data.head()
```

Ochlaci	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Propert
Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	
Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	
	Male Male Male	Male Yes Male Yes Male Yes	Male Yes 1 Male Yes 0 Male Yes 0	MaleYes1GraduateMaleYes0GraduateMaleYes0Not Graduate	MaleYes1GraduateNoMaleYes0GraduateYesMaleYes0Not GraduateNo	MaleYes1GraduateNo4583MaleYes0GraduateYes3000MaleYes0Not GraduateNo2583	Male Yes 1 Graduate No 4583 1508.0 Male Yes 0 Graduate Yes 3000 0.0 Male Yes 0 Not Graduate No 2583 2358.0	Male Yes 1 Graduate No 4583 1508.0 128.0 Male Yes 0 Graduate Yes 3000 0.0 66.0 Male Yes 0 Not Graduate No 2583 2358.0 120.0	Male Yes 1 Graduate No 4583 1508.0 128.0 360.0 Male Yes 0 Graduate Yes 3000 0.0 66.0 360.0 Male Yes 0 Not Graduate No 2583 2358.0 120.0 360.0	Male Yes 1 Graduate No 4583 1508.0 128.0 360.0 1.0 Male Yes 0 Graduate Yes 3000 0.0 66.0 360.0 1.0 Male Yes 0 Not Graduate No 2583 2358.0 120.0 360.0 1.0



4. Data Exploration and Visualization in Databricks:



```
✓ 05:52 PM (<1s)
  # Print the schema to understand data types and column structure
  data.printSchema()
root
|-- id: long (nullable = true)
|-- firstName: string (nullable = true)
|-- middleName: string (nullable = true)
|-- lastName: string (nullable = true)
|-- gender: string (nullable = true)
|-- birthDate: timestamp (nullable = true)
|-- ssn: string (nullable = true)
|-- salary: long (nullable = true)
      ✓ 05:52 PM (1s)
  # Show summary statistics for numeric columns
  data.describe().show()
▶ (2) Spark Jobs
+-----+
             id|firstName|middleName|lastName|gender| ssn|
                                                                       salary
+-----+
| count| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| | stddev|288.8194360957494| NULL| NULL| NULL| NULL| NULL| NULL| NULL|20670.644326853664|
  min| 1| Abbie| Adella| Abrahim| F|666-15-8671| -6523|
                1000 | Zita | Zulma | Zannutti | F | 999-93-2044 |
   max
```

```
# Count the number of rows and columns
num_rows = data.count()
num_cols = len(data.columns)
print(f"Rows: {num_rows}, Columns: {num_cols}")

** (2) Spark Jobs
Rows: 1000, Columns: 8
```

Visualization:

