# **Azure Databricks Assignment**

# **Submitted By-**

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**Date:** 28-11-2024

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

```
✓ 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")

(6) Spark Jobs

* 5. Optionally create a Delta table

df.write.format("delta").mode("overwrite").saveAsTable("people_deltatable")

(6) Spark Jobs
```

#### 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")

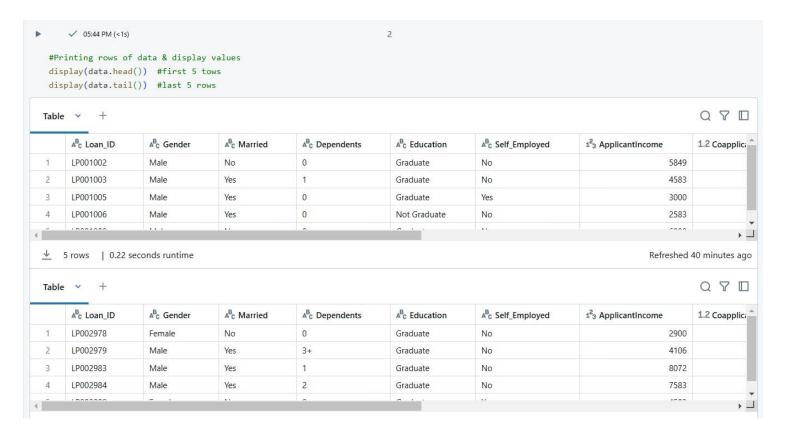
• (6) Spark Jobs

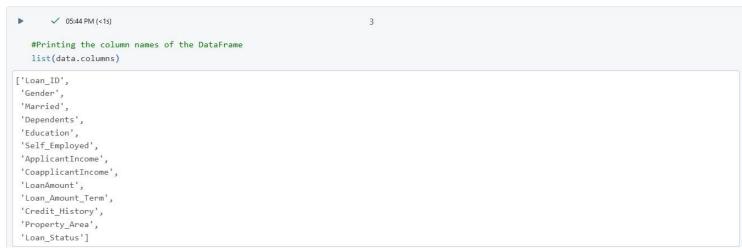
• (6) Spark Jobs

• (7) If the property of the
```

#### 3. Exploratory data analysis (EDA) in Databricks:

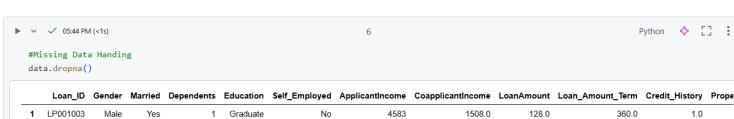
```
▶ ✓ ✓ 05:43 PM (<1s)
                                                                                                                    Python 💠
   import pandas as pd
   spark_df = spark.read.format("delta").load("dbfs:/user/hive/warehouse/loan_data")
   data = spark_df.toPandas()
   print(data)
▶ (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
   LP001002
0
              Male
                      No ...
                                1.0
                                                    Urban
   LP001003
                       Yes ...
                                         1.0
1
              Male
                                                     Rural
                                                                   N
                      Yes ...
2
    LP001005
              Male
                                         1.0
                                                    Urban
3
    LP001006
               Male
                                                    Urban
                       Yes ...
                                         1.0
4
    LP001008
                                         1.0
                                                    Urban
                       No ...
                       ... ...
                       No ...
609 LP002978 Female
                                                    Rural
                                         1.0
                       Yes ...
610 LP002979 Male
                                         1.0
                                                     Rural
611 LP002983 Male
                     Yes ...
                                         1.0
                                                    Urban
                     Yes ...
612 LP002984 Male
                                        1.0
                                                   Urban
613 LP002990 Female
                                        0.0 Semiurban
                      No ...
[614 rows x 13 columns]
```





# #Descriptive Statistical Measures of a DataFrame data.describe()

ApplicantIncome	Coapplicantincome	LoanAmount	Loan_Amount_Term	Credit_History
614.000000	614.000000	592.000000	600.00000	564.000000
5403.459283	1621.245798	146.412162	342.00000	0.842199
6109.041673	2926.248369	85.587325	65.12041	0.364878
150.000000	0.000000	9.000000	12.00000	0.000000
2877.500000	0.000000	100.000000	360.00000	1.000000
3812.500000	1188.500000	128.000000	360.00000	1.000000
5795.000000	2297.250000	168.000000	360.00000	1.000000
81000.000000	41667.000000	700.000000	480.00000	1.000000
	614.000000 5403.459283 6109.041673 150.000000 2877.500000 3812.500000 5795.000000	614.000000 614.000000 5403.459283 1621.245798 6109.041673 2926.248369 150.000000 0.000000 2877.500000 0.000000 3812.500000 1188.500000 5795.000000 2297.250000	614.000000       614.000000       592.000000         5403.459283       1621.245798       146.412162         6109.041673       2926.248369       85.587325         150.000000       0.000000       9.000000         2877.500000       0.000000       100.00000         3812.500000       1188.500000       128.000000         5795.000000       2297.250000       168.000000	614.000000       614.000000       592.000000       600.00000         5403.459283       1621.245798       146.412162       342.00000         6109.041673       2926.248369       85.587325       65.12041         150.000000       0.000000       9.000000       12.00000         2877.500000       0.000000       100.000000       360.00000         3812.500000       1188.500000       128.000000       360.00000         5795.000000       2297.250000       168.000000       360.00000



	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

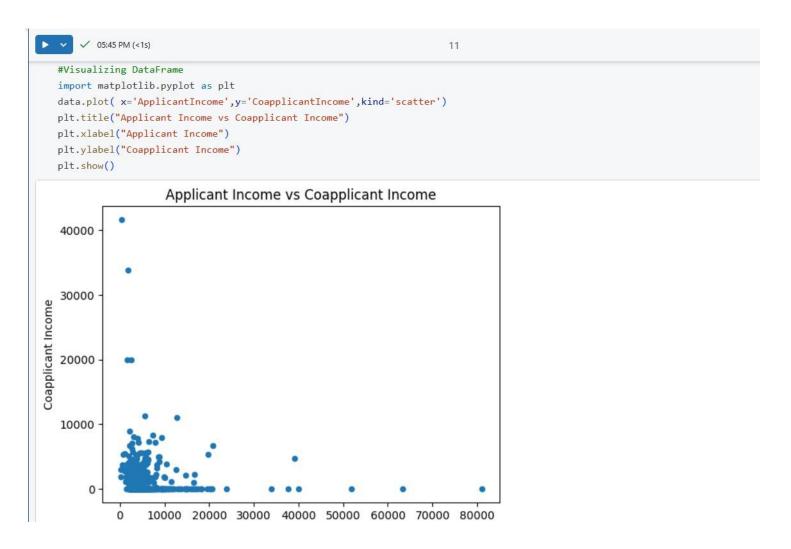
**←** 

```
#Apply Function

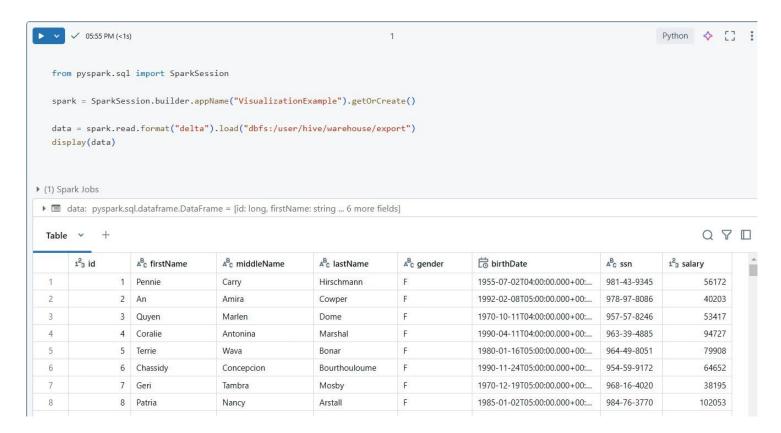
def fun(value):
    if value>3000:
        return 'Yes'
    else:
        return 'No'

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|
+-----+
| 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}")

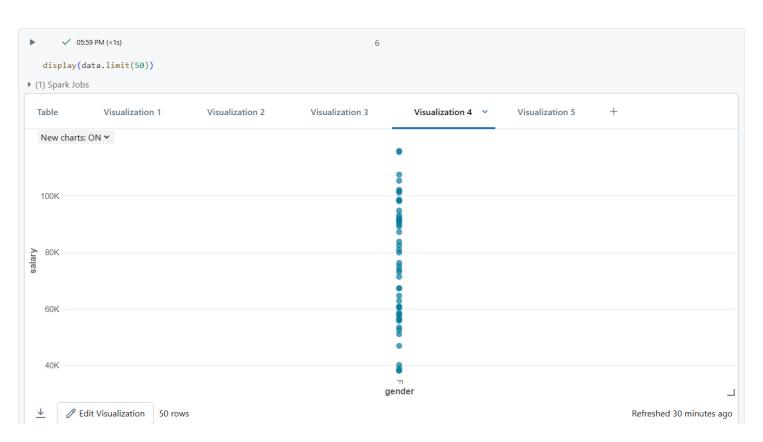
** (2) Spark Jobs
Rows: 1000, Columns: 8
```

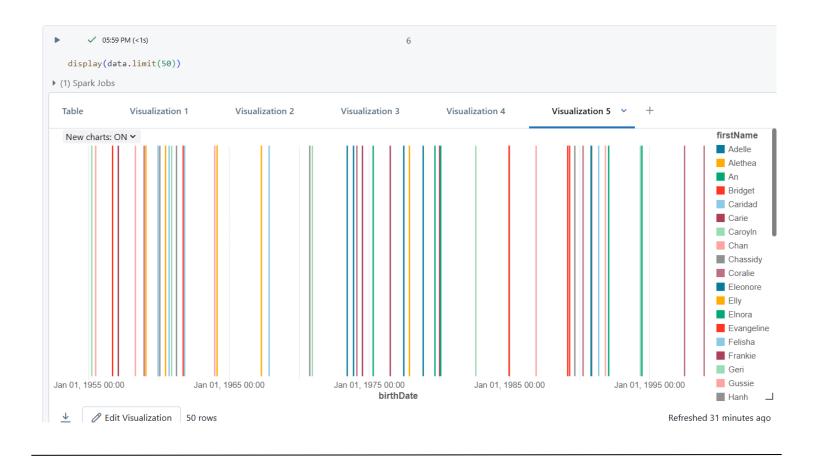
# **Visualization:**











# Thank You!