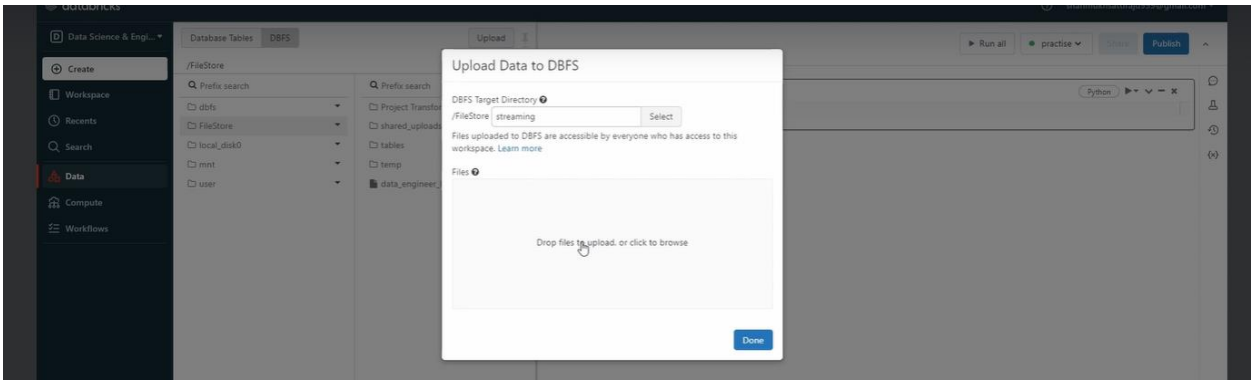
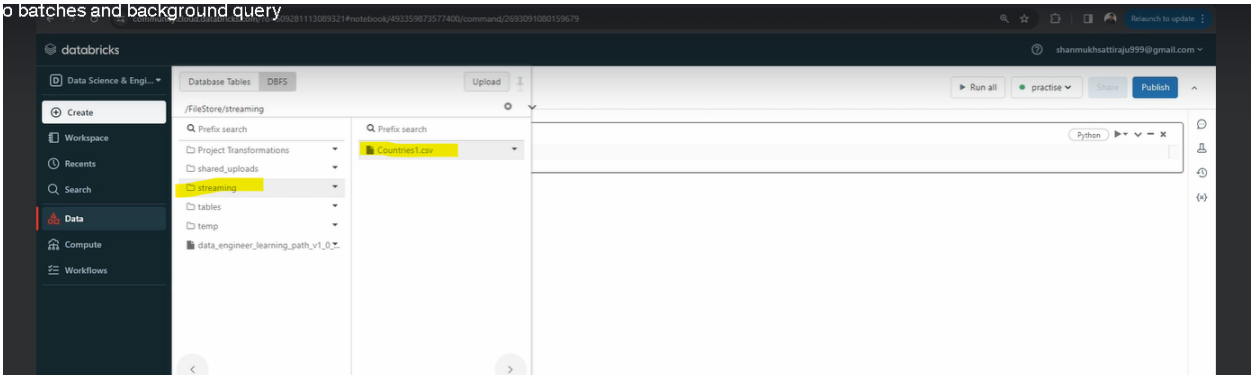
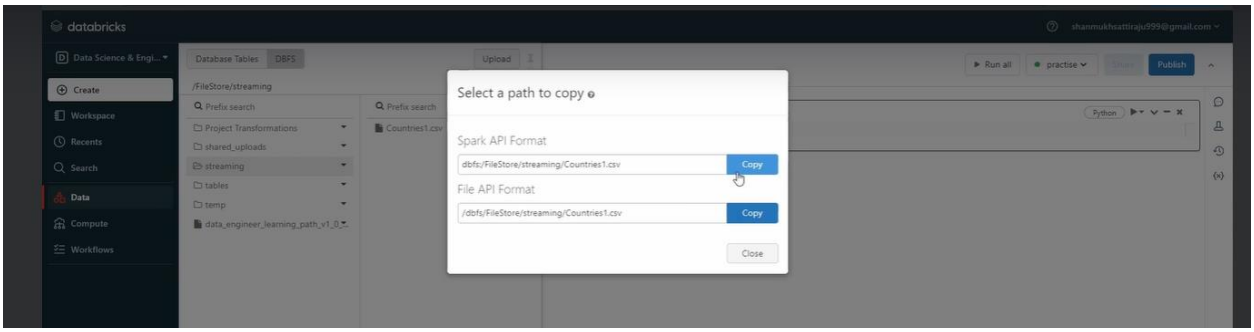
Streaming_practice

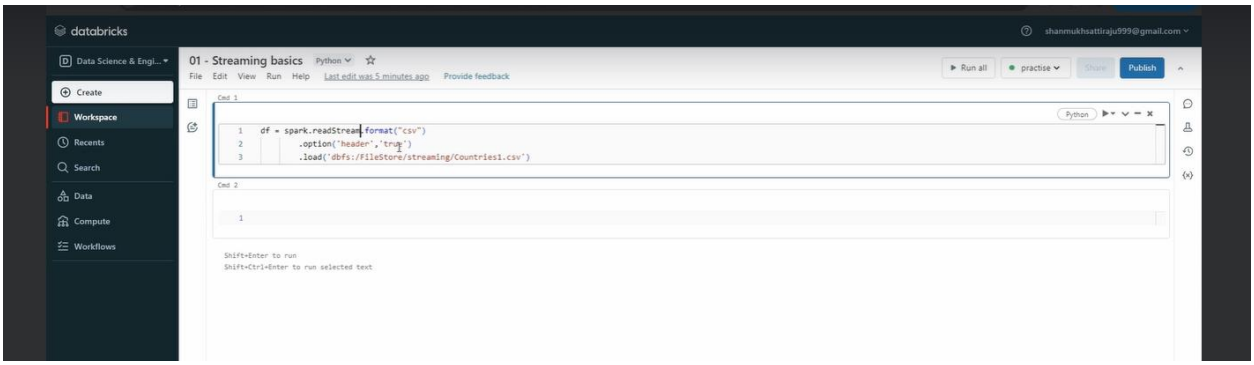


Upload a file to dbfs in the workspace creating a folder



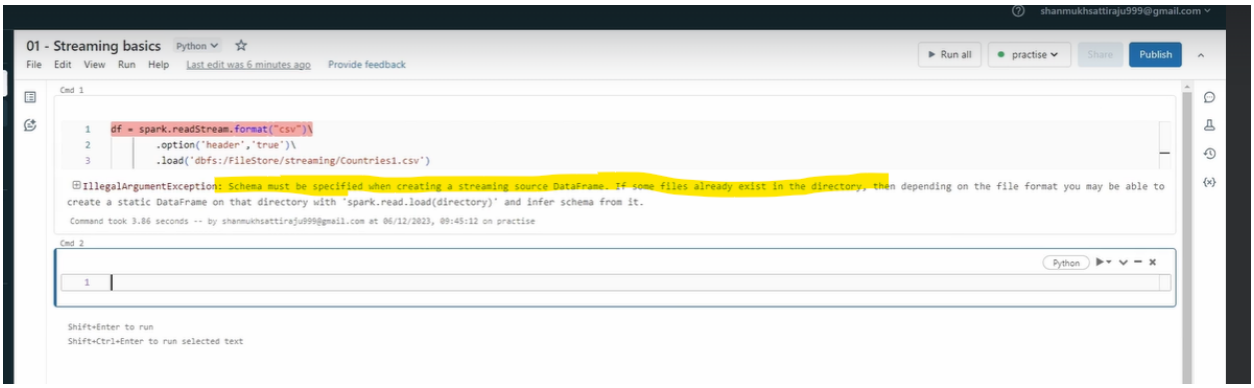


Reading the data



Reading a streaming dataset

If we try to read the dataframe directly without specifying or inferring the schema, we will get the below error.




```
from pyspark.sql.types import StructType, StructField, StringType , IntegerType, FloatType

schema = StructType([
                     StructField('Country',StringType()),
                     StructField('Citizens',IntegerType())
])
```

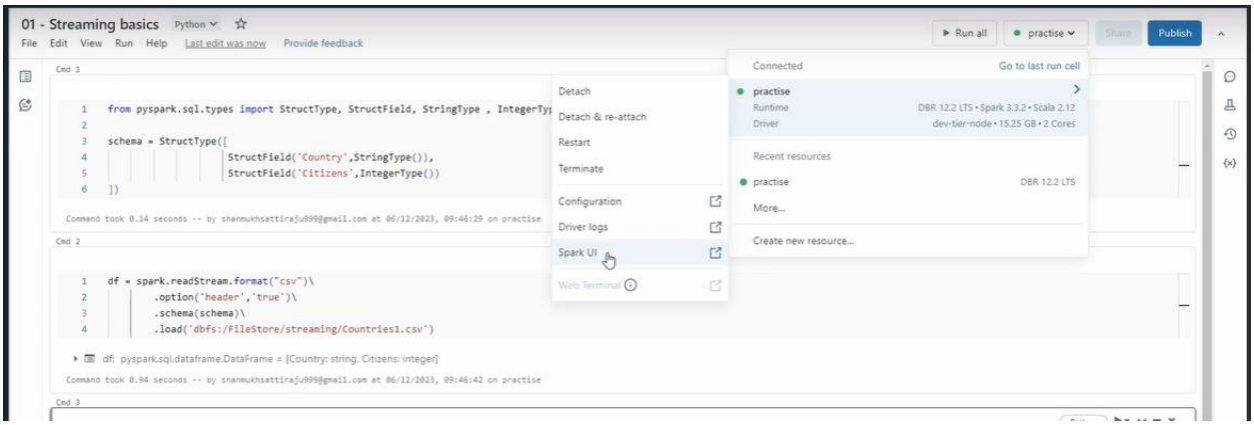
Define the schema and re-run the data frame

```
df = spark.readStream.format("csv")\
    .option('header','true')\
    .schema(schema)\
    .load('dbfs:/FileStore/streaming/Countries1.csv')
```

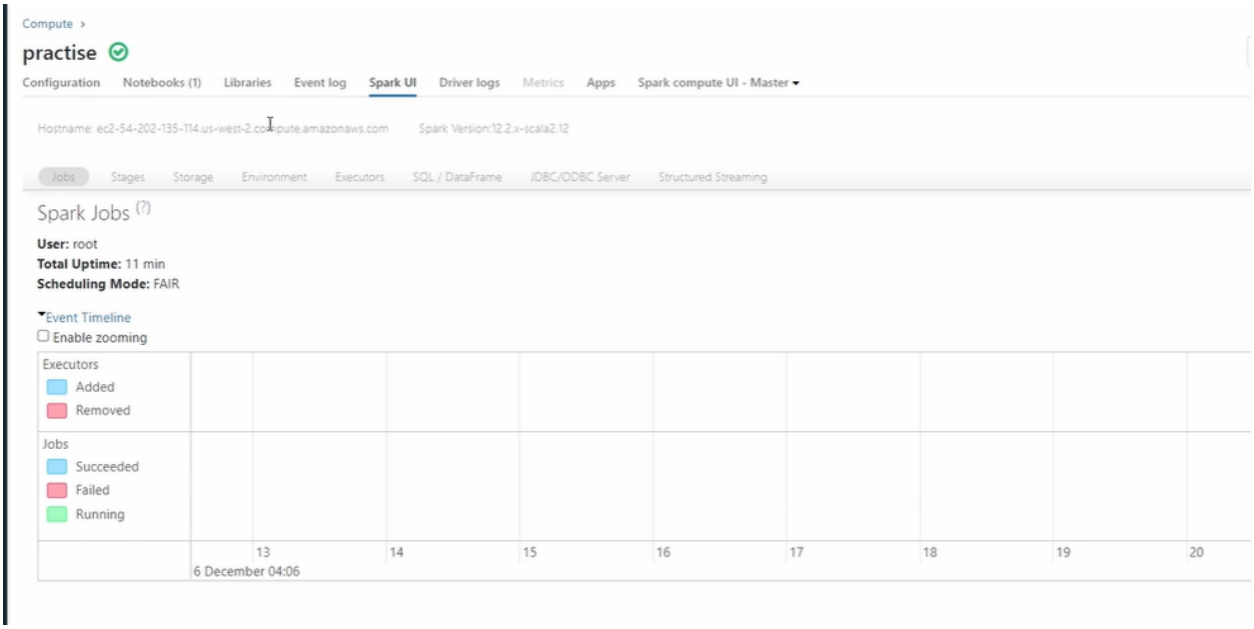
df: pyspark.sql.dataframe.DataFrame = [Country: string, Citizens: integer]

Command took 0.94 seconds -- by shanmukhsattiraju999@gmail.com at 06/12/2023, 09:46:42 on practise

Rerunning the data frame, specifying the schema will read the stream data

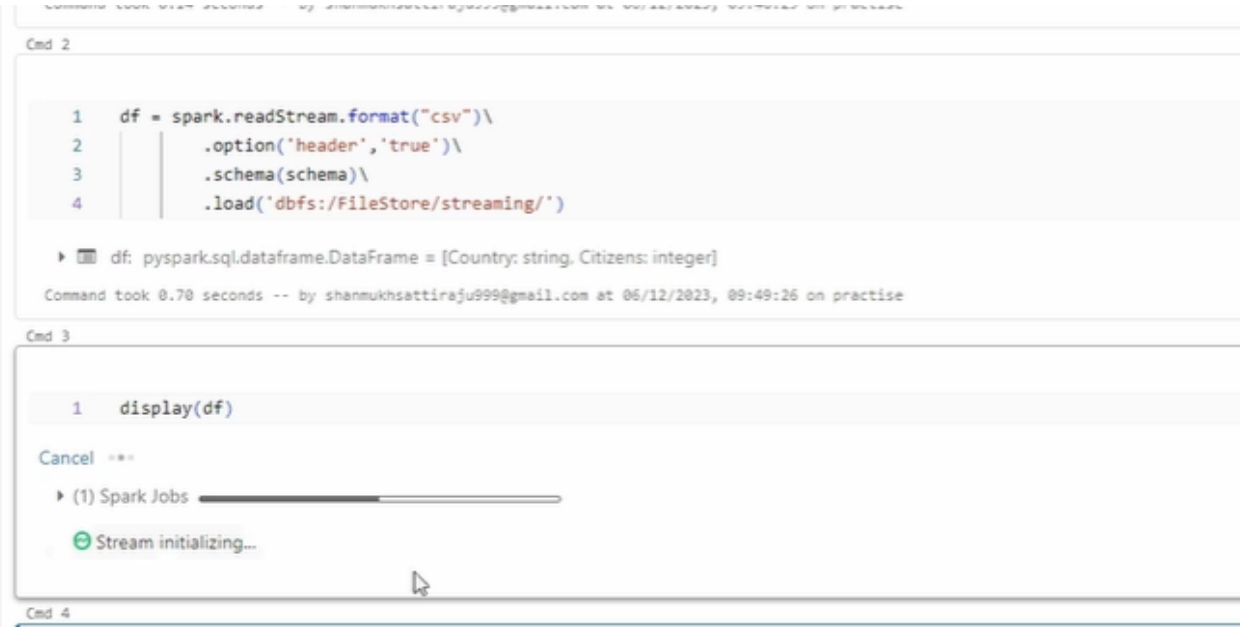


Jobs initiated can be read through the spark UI interface

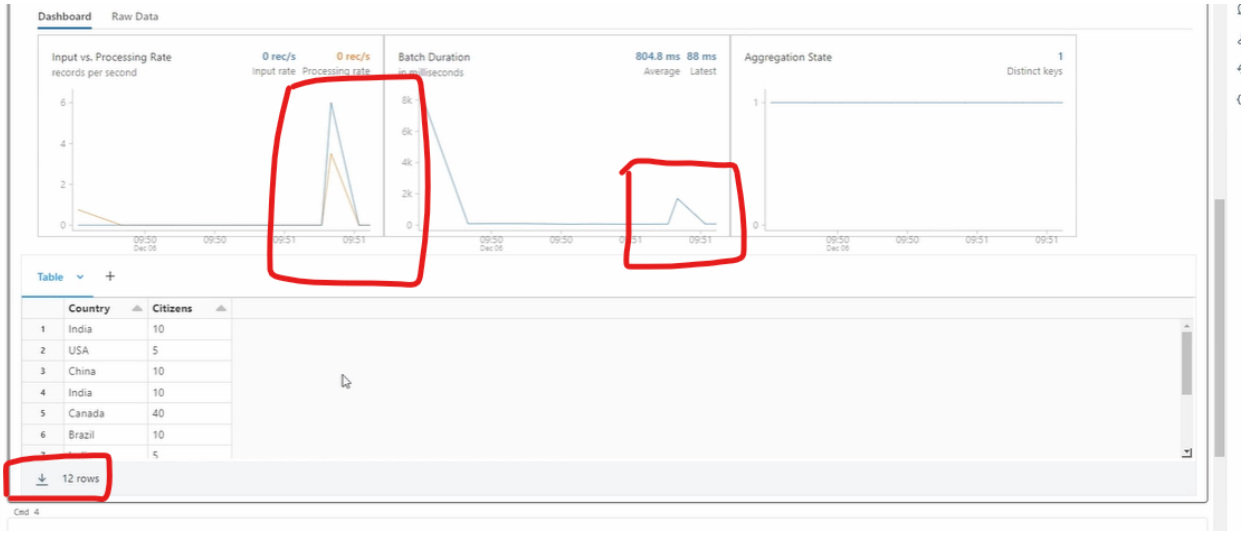


Initially no jobs were initiated, jobs will be initiated only when trying to fetch some data.

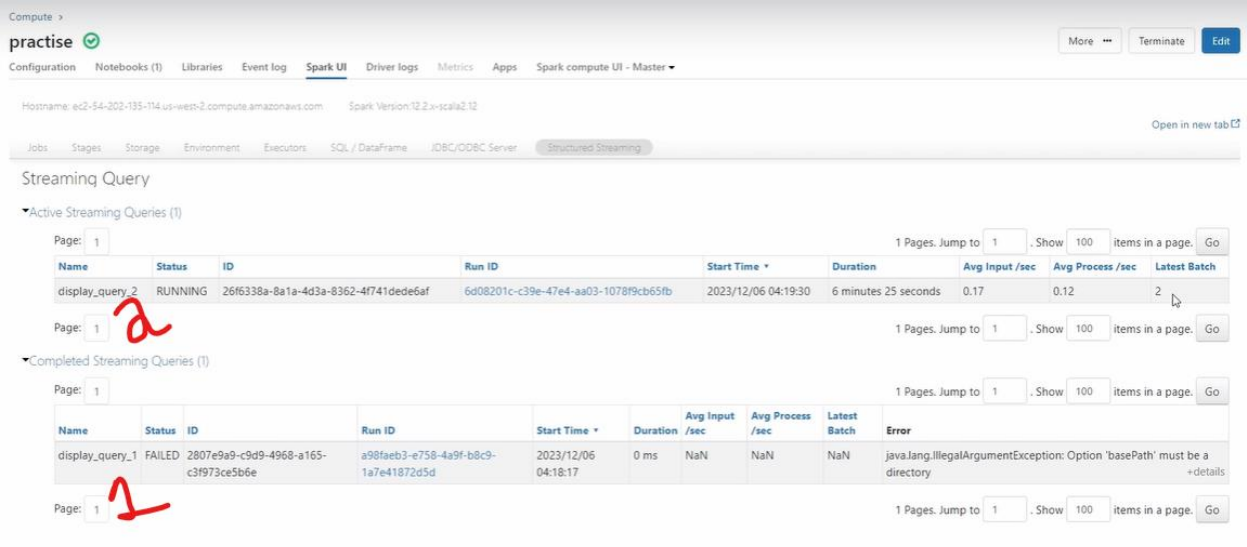
Streaming data can be read through display commad, however basepath must be a directory.



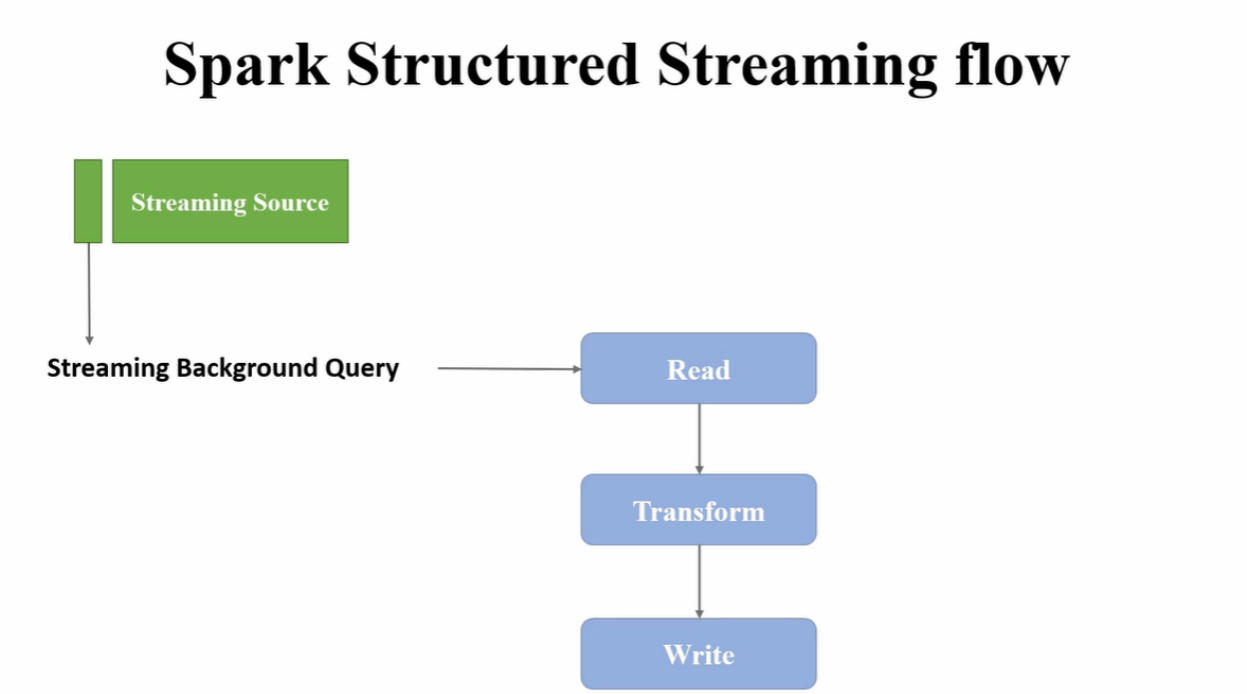
To fix it change the load path to directly to directory level.



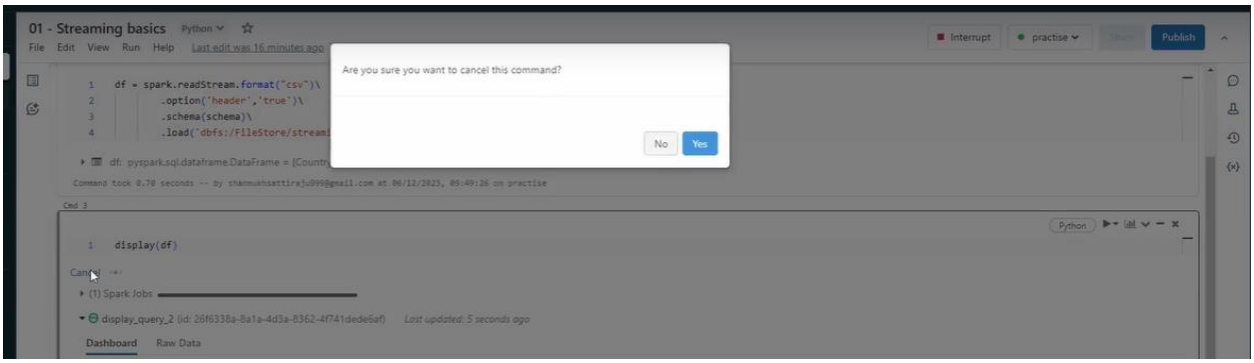
To observe the change in the streaming input data, add a new file to the directory. With in few seconds will be noticed in the input and batch streaming graph, without re-executing the display command. Record count is changed to 12 from 6.



Each file addition is read in micro batches.



The background streaming query acts as a file watcher, whenever there is a new file added to the directory, a new job will be invoked, the file will be read, transformed and written. The query will be running continuously, until it is stopped explicitly.



Streaming queries can be stopped explicitly by cancelling the query.

Supported sources and sinks for the Spark Streaming.

StreamWriter

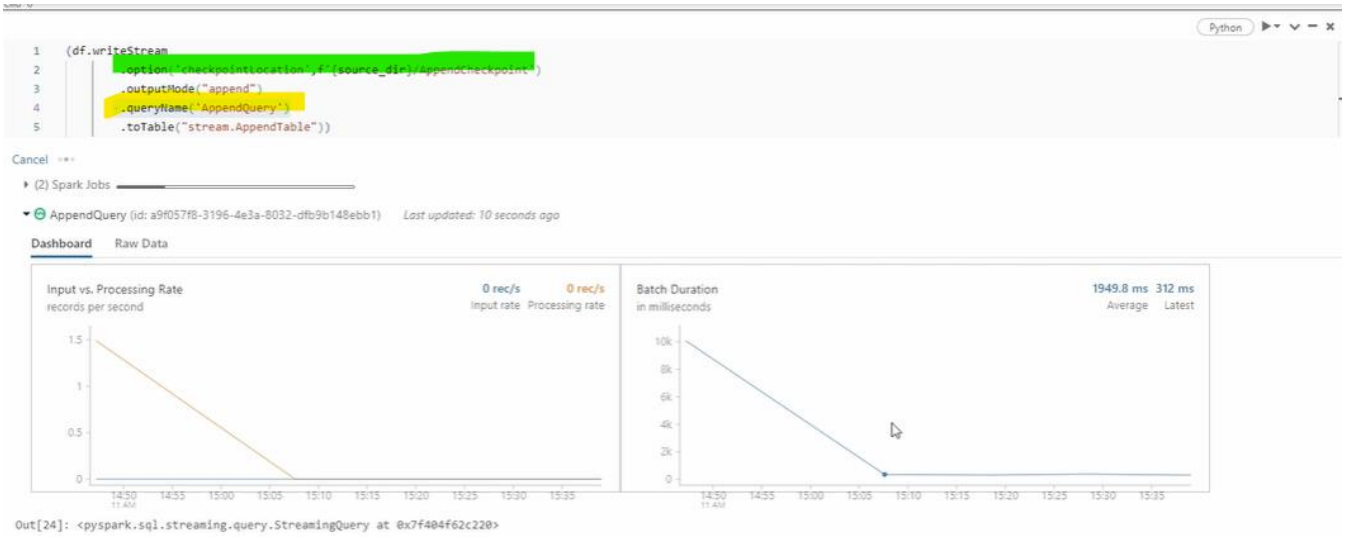
```
<StreamingDataframe>.writeStream
                    .option('checkpointLocation',<Location>)
                    .outputMode('append')
                    .toTable('<TableName>')
```

Checkpoint

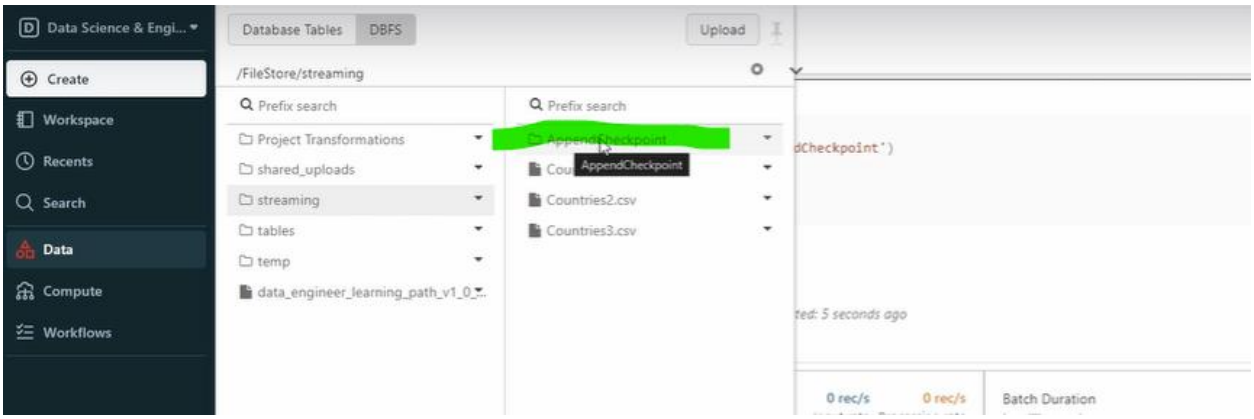
To develop fault tolerant and resilient Spark APPS

It maintains intermediate state on fault-tolerant compatible files like DBFS, ADLS and S3 storage systems to recover from the failures.

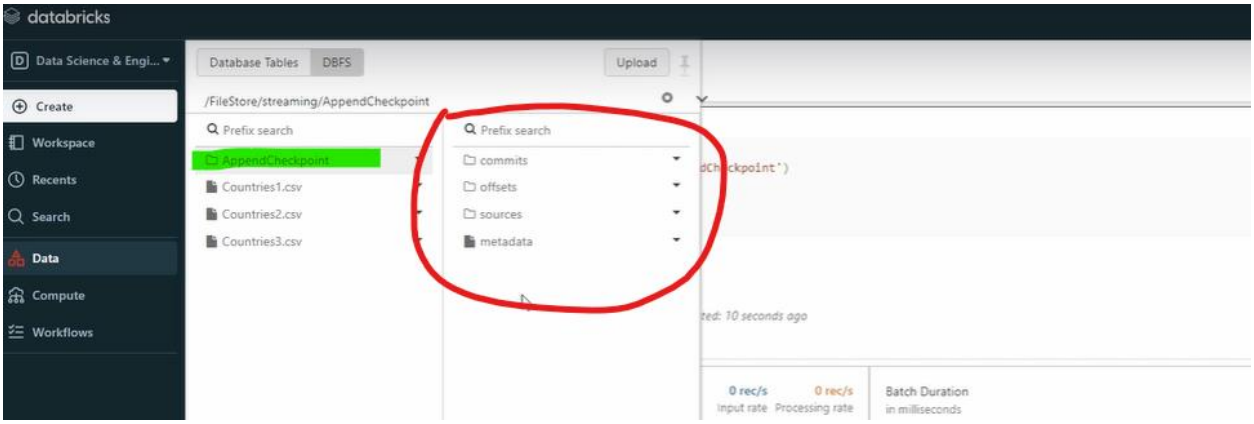
It is unique to each stream.



Streaming data frame, data is read, checkpoint location is specified and query name as well specified.

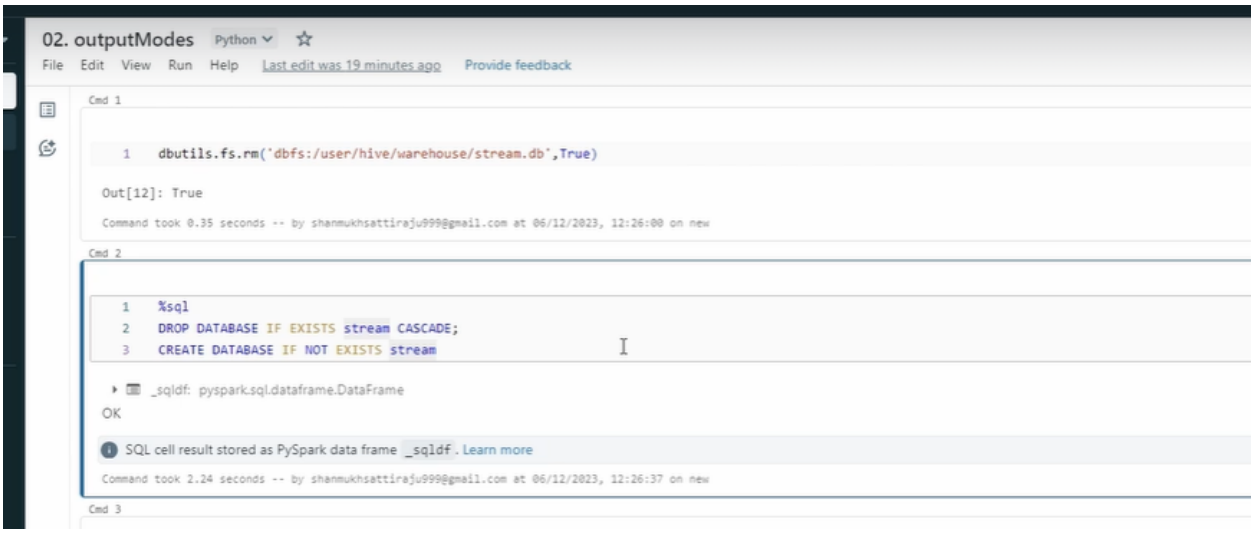


Now after executing the streaming query, a check point location folder is added in the location specified in the above query.

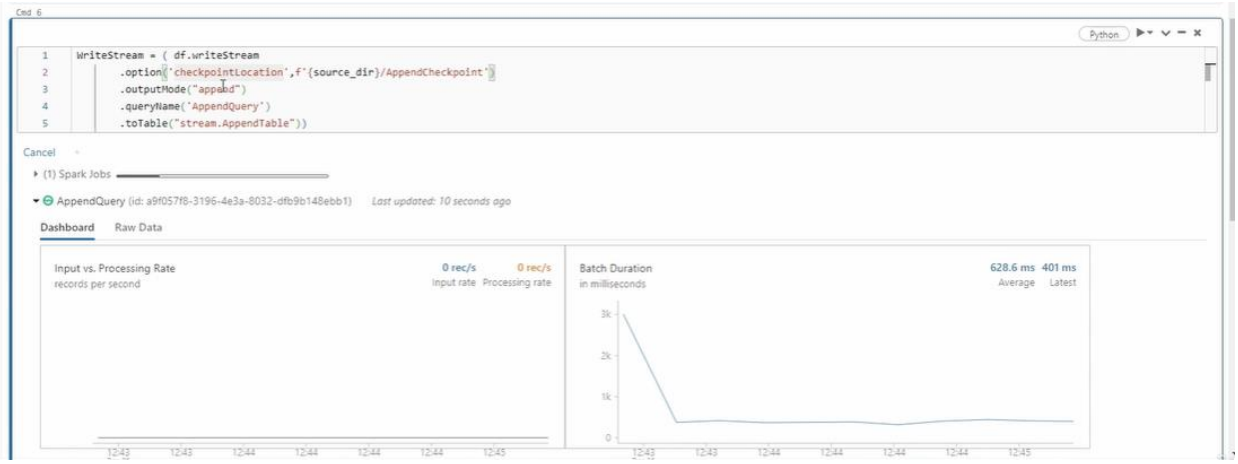


Contains the metadata of the streaming

Assign the streaming query to a variable and to stop the streaming use the command **<var>.stop()**



In the community edition when the cluster is terminated all the schema/database created in the working session will be terminated. Therefore, while using the community edition we should remove the files in the dbfs path, drop the database if exists and re-create it.



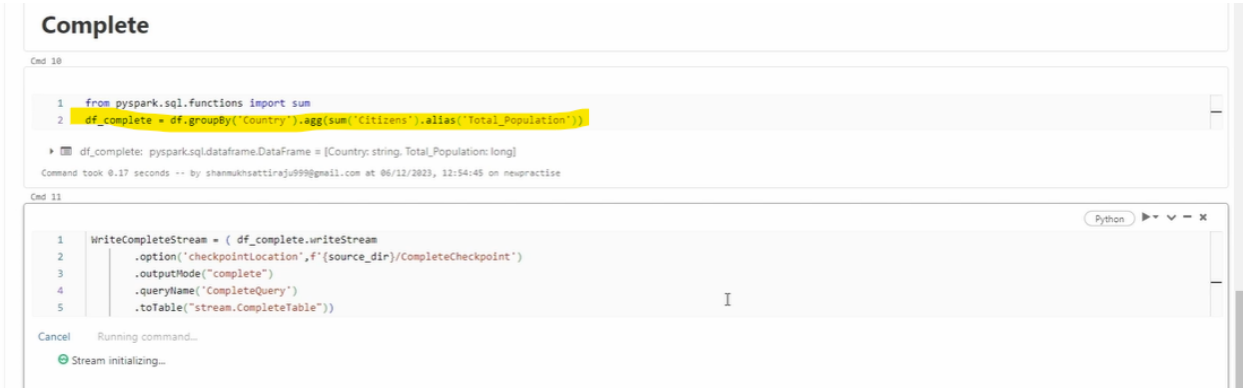
When we are trying to run the stream query initially it will verify the check point location, if there are any unprocessed files, only then it will show the updates in the graph

Output Modes

Deleting the checkpoint location will help to process the older data, without removing the data, just by removing the checkpoint, previous data will be processed.

Complete Mode

To use the complete output mode aggregation should be used on the data frame.



Applied aggregation to execute the ***complete output mode***

Use case of complete mode

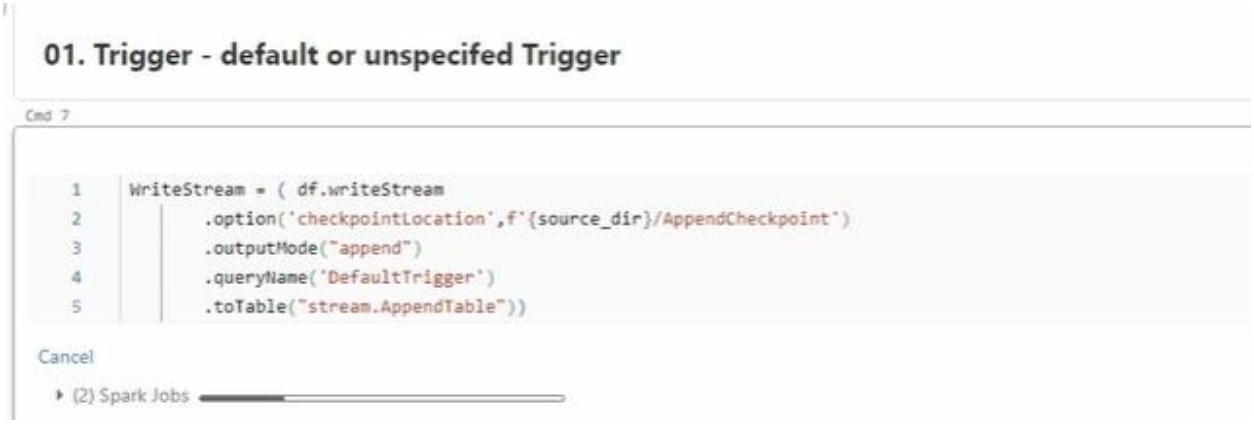
When a new file or dataset is added, and we need to apply the aggregation on top it, to fetch the results of the query entire files should be read.

Countries1	
Country	Citizens
India	10
USA	5
China	10
India	10
Canada	40
Brazil	10

Countries2	
Country	Citizens
India	5
USA	10
China	5
India	5
Canada	10
Brazil	50

Result of Countries 1 and 2	
Country	Total Pop
India	30
USA	15
China	15
Canada	50
Brazil	60

Trigger types



If no trigger time is specified, by default for every 5sec, streaming query will look for new data. If the new file is added every 5 to 10 min, default trigger is not a good option.

Cmd 9

02. Trigger - processingTime

Cmd 10

```
WriteStream = ( df.writeStream
      .option('checkpointLocation',f'{source_dir}/AppendCheckpoint')
      .outputMode("append")
      .trigger(processingTime='2 minutes')
      .queryName('ProcessingTime')
      .toTable("stream.AppendTable"))
```

Every 2 min, a micro batch will be processed.

At 13:08 initial processing occurred, even though data is available only after 2 min next micro batch will be processed.

03. Trigger - availablenow

Cmd 12

```
WriteStream = ( df.writeStream
      .option('checkpointLocation',f'{source_dir}/AppendCheckpoint')
      .outputMode("append")
      .trigger(availableNow=True)
      .queryName('AvailableNow')
      .toTable("stream.AppendTable"))
```

▸ (1) Spark Jobs

▸ AvailableNow (id: 9150eaac-ace9-4fd7-a9dc-07c5dc687ac4) *Last updated: 45 seconds ago*

Is equivalent to incremental batch processing, whenever there is a new data available query will process and stop the streaming

Triggers

Triggers	Usage	Description
Unspecified (default)		will trigger the microbatch for every 500 ms or half a second
processingTime (Fixed Interval)	. trigger(processingTime='2 minutes')	You can set processing time or time interval for each execution .
availableNow (OneTime)	.trigger(availableNow = True)	consumes all available records from previous execution as an incremental batch
Continuous (experimental)	.trigger(continuous = '1 second')	For ~1ms latency

Types of streaming triggers

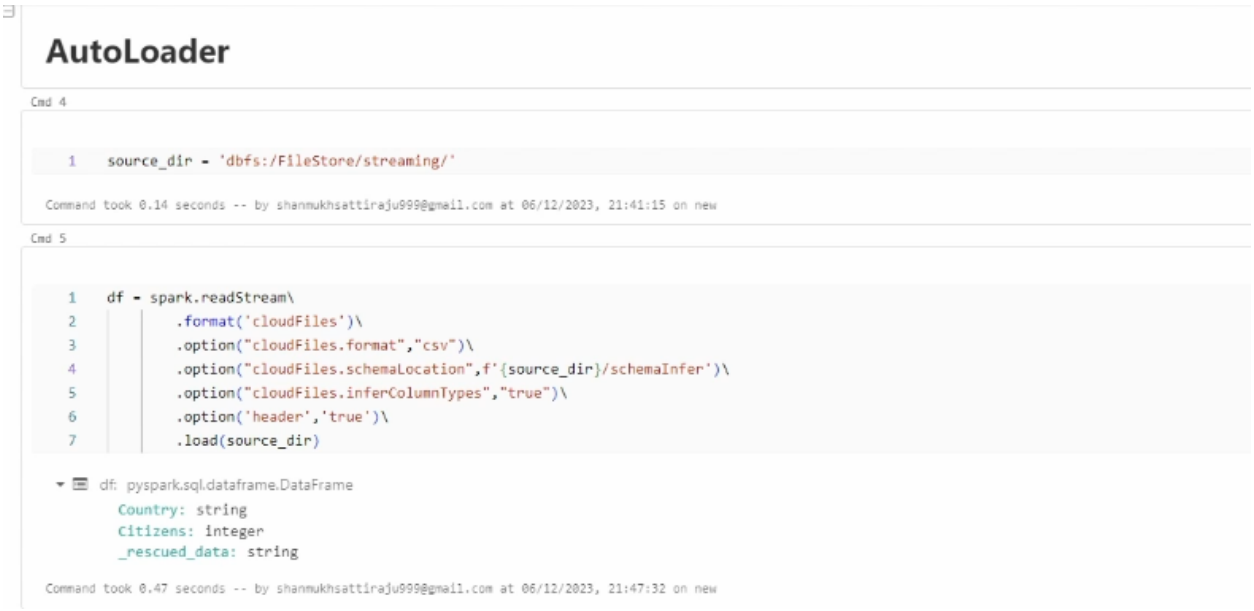
Autoloader

Autoloader

- Autoloader is an **optimized data ingestion tool** that incrementally and efficiently processes new data files as they arrive in the cloud storage built into the Databricks Lakehouse.
- Auto Loader incrementally and efficiently processes new data files as they arrive in cloud storage without any additional setup.
- Auto Loader can load data files from Cloud Storages without being vendor specific (AWS S3 , Azure ADLS , Google Cloud Storage, DBFS).
- Auto Loader can ingest JSON, CSV, PARQUET, AVRO, ORC, TEXT, and BINARYFILE file formats
- This Auto loader is beneficial when you are ingesting data into your lakehouse particularly into bronze layer as a streaming query.

```
df_str = (spark.readStream
          .format("cloudFiles")     ## This will tell the spark to use AutoLoader.
          .option("cloudFiles.format","csv")  ## Tells Autoloader to expect csv files
          .option('header','true')
          .schema(schema)
          .load(f'{source_dir}')
 )
```

To use the autoloader format(”cloudFiles”) needs to be specified



To get the correct data types, schema must be inferred.

Schema can be referred through the path, rescued data is the additional field created.

Schema Hints

For explicitly defining the data type of a column

For explicitly defining the data type of a column

Schema Evolution

- Schema evolution is the process of managing changes in data schema as it evolves over time, often due to updates in software or changing business requirements, which can cause schema drift

- Ways to handle schema changes
  - Fail the stream
  - Manually change the existing schema
  - Evolve automatically with change in schema

There are 4 schema evolution types

1. AddNewColunms

   Stream will fail, if new column is added and existing columns do not evolve.

2. FailOnNewColumns

   Stream fails, stream does not restart until schema is updated or offending data is removed.

3. Rescue
   Schema never evolved; stream doesn't change due to schema changes. All new columns are recorded in the rescued data.
4. None

Ignore new columns, (doesn't evolve the schema, new columns are ignored, data is not rescued unless the rescuedDataColumn option is set. Stream doesn't fail due to schema changes)

Using the **rescue type schema evolution**

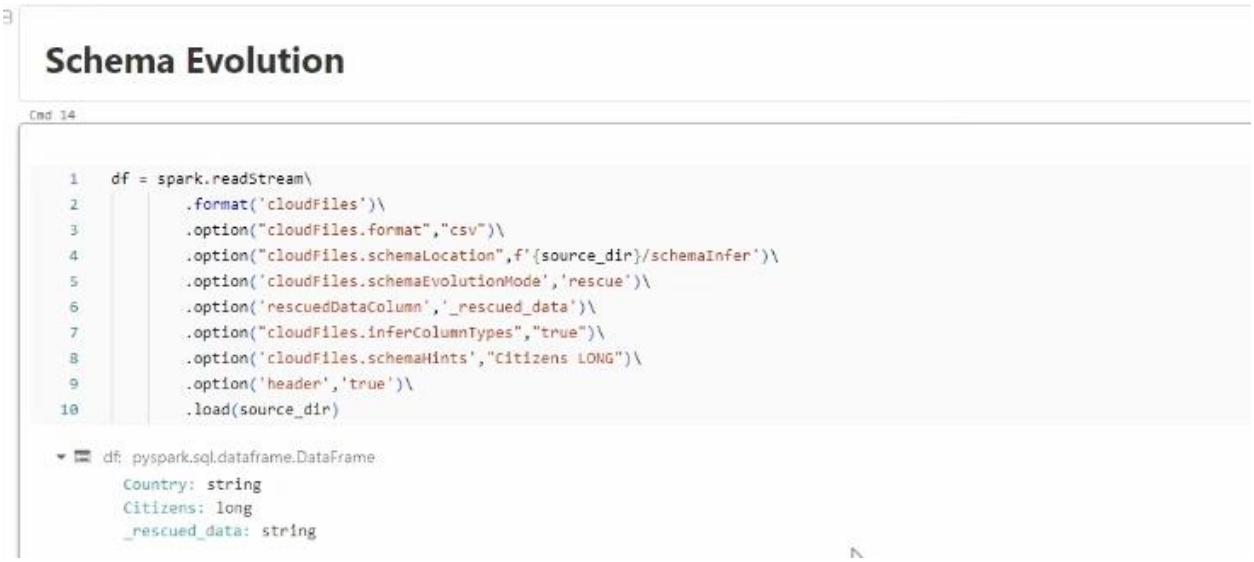
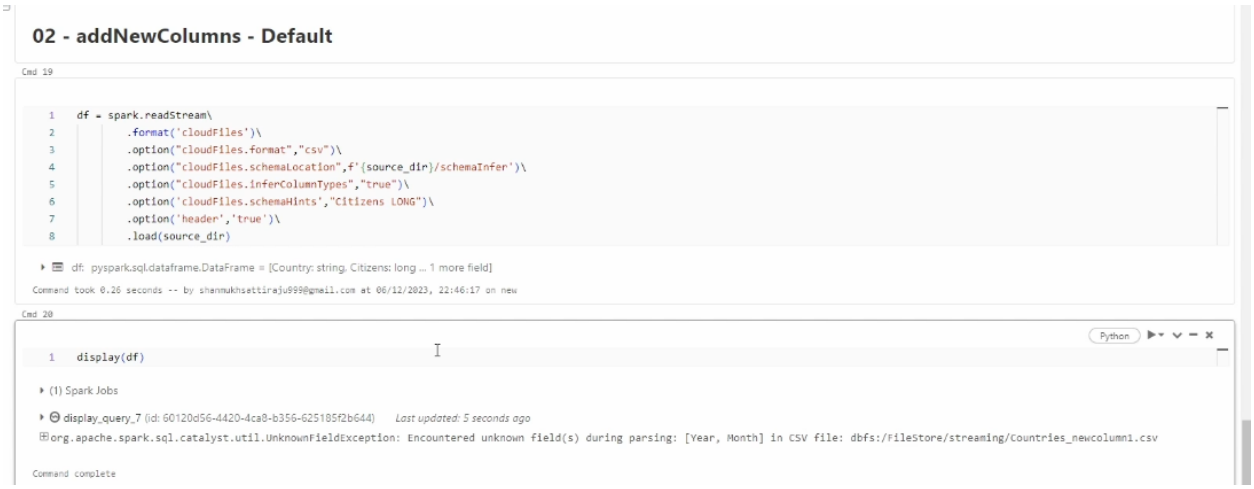


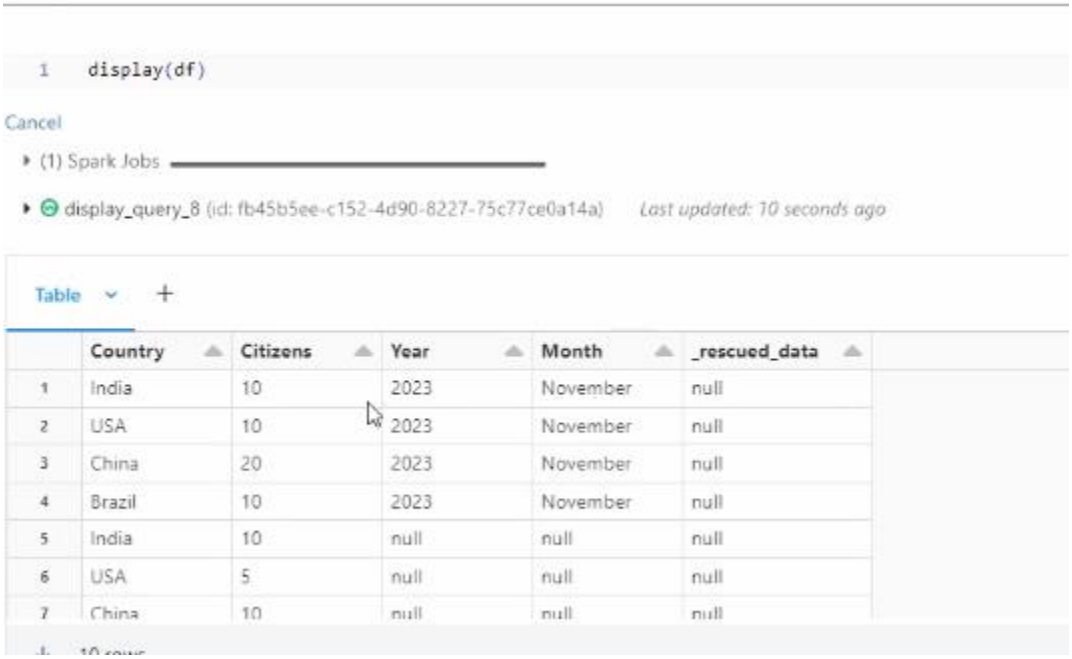
Table +

	Country	Citizens	_rescued_data
1	India	10	{"Month":"November","Year":"2023","_file_path":"dbfs:/FileStore/streaming/Countries_newcolumn1.csv"}
2	USA	10	{"Month":"November","Year":"2023","_file_path":"dbfs:/FileStore/streaming/Countries_newcolumn1.csv"}
3	China	20	{"Month":"November","Year":"2023","_file_path":"dbfs:/FileStore/streaming/Countries_newcolumn1.csv"}
4	Brazil	10	{"Month":"November","Year":"2023","_file_path":"dbfs:/FileStore/streaming/Countries_newcolumn1.csv"}
5	India	10	null
6	USA	5	null
7	China	10	null

Extra columns that are added handled by the rescued_data column

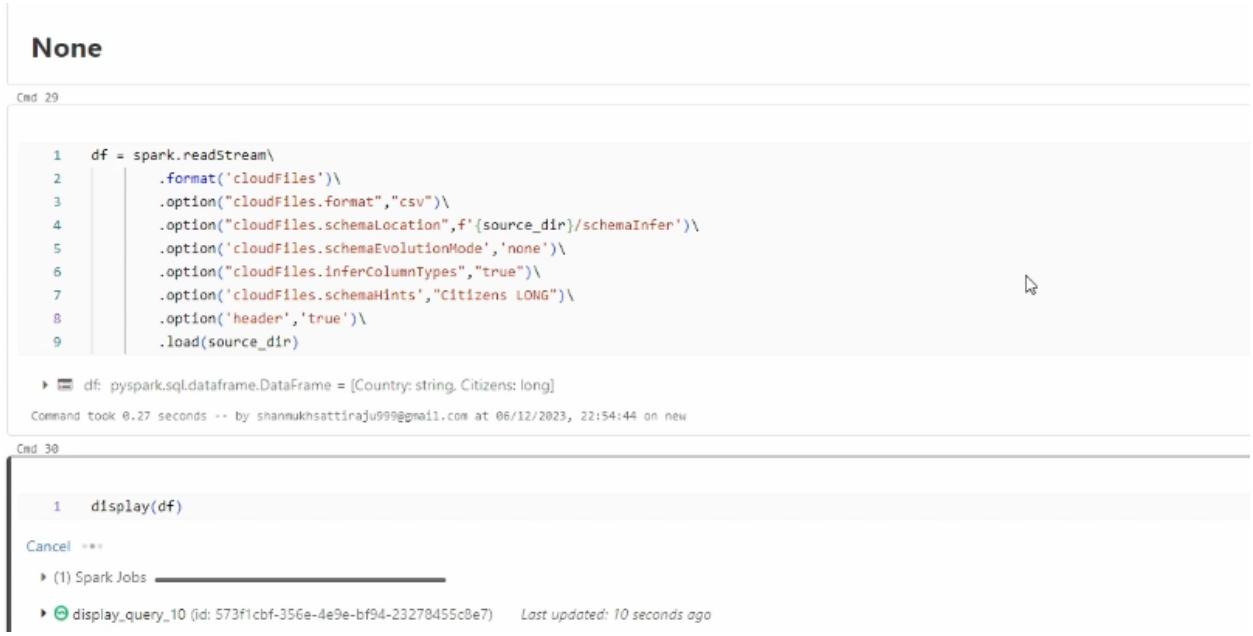


In the addNewColumns(default) type, after adding a file with extra columns, dataframe will be executed, however when we do a display it will fail.



To fix this need to rerun the data frame and display to see the data with evolved schema.

FailOnNewcolumns schema evolution mode, when a new file is added, display will fail, to fix either schema must be redefined or data should be changed.



None type of schema evolution ignores any changes in the schema, add the data to the existing schema.