

#1. Import Libraries

```
from pyspark.sql.functions import col
from pyspark.sql.types import IntegerType, DoubleType, BooleanType,
DateType
```

#2. Need to create connections [Databricks <--> Azure Data Lake]

1. Create configuration format
2. Create mount

#1. create configuration format

```
configs = {"fs.azure.account.auth.type": "OAuth",
"fs.azure.account.oauth.provider.type":
"org.apache.hadoop.fs.azurebfs.oauth2.ClientCredsTokenProvider",
"fs.azure.account.oauth2.client.id": "4e5fde0e-f726-4e0d-9e79-
4685413e2078", #clientid
"fs.azure.account.oauth2.client.secret":
'hWm8Q~Kbovgxczxcg2yx.wZl6n_dv4Up.-w8YubU.', #secretid
"fs.azure.account.oauth2.client.endpoint":
"https://login.microsoftonline.com/9075e7fb-176e-4971-ae19-
2443cd22c3f2/oauth2/token"} #tenantid
```

#2. Mount

```
dbutils.fs.mount(
source = "abfss://ade-data@adeprojectdata.dfs.core.windows.net/", #
container@storageacc
mount_point = "/mnt/tokyoolympic",
extra_configs = configs)
```

#3. To Check if the mount is successful or not

```
%fs
ls "/mnt/tokyoolympic"
```

#4. Read Each File's Data & Schema

1. Athletes
2. Coaches
3. Entries Gender
4. Medals
5. Teams

```
#filename =
spark.read.format("csv:fileformat").options("header","true").option("inferSchema","true").load("@(mountpoint:/mnt/tokyoolympic)/@datalakename/athletes.csv")
```

```

athletes =
spark.read.format("csv").option("header","true").option("inferSchema",
"true").load("/mnt/tokyoolympic/Raw Data Lake/Athletes.csv")
coaches =
spark.read.format("csv").option("header","true").option("inferSchema",
"true").load("/mnt/tokyoolympic/Raw Data Lake/Coaches.csv")
entriesgender =
spark.read.format("csv").option("header","true").option("inferSchema",
"true").load("/mnt/tokyoolympic/Raw Data Lake/EntriesGender.csv")
medals =
spark.read.format("csv").option("header","true").option("inferSchema",
"true").load("/mnt/tokyoolympic/Raw Data Lake/Medals.csv")
teams =
spark.read.format("csv").option("header","true").option("inferSchema",
"true").load("/mnt/tokyoolympic/Raw Data Lake/Teams.csv")

```

#1. Athletes Data

```
athletes.show()
```

PersonName	Country	Discipline
AALERUD Katrine	Norway	Cycling Road
ABAD Nestor	Spain	Artistic Gymnastics
ABAGNALE Giovanni	Italy	Rowing
ABALDE Alberto	Spain	Basketball
ABALDE Tamara	Spain	Basketball
ABALO Luc	France	Handball
ABAROA Cesar	Chile	Rowing
ABASS Abobakr	Sudan	Swimming
ABBASALI Hamideh	Islamic Republic of Iran	Karate
ABBASOV Islam	Azerbaijan	Wrestling
ABBINGH Lois	Netherlands	Handball
ABBOT Emily	Australia	Rhythmic Gymnastics
ABBOTT Monica	United States of America	Baseball/Softball
ABDALLA Abubaker ...	Qatar	Athletics
ABDALLA Maryam	Egypt	Artistic Swimming
ABDALLAH Shahd	Egypt	Artistic Swimming
ABDALRASOOL Mohamed	Sudan	Judo
ABDEL LATIF Radwa	Egypt	Shooting
ABDEL RAZEK Samy	Egypt	Shooting
ABDELAZIZ Abdalla	Egypt	Karate

only showing top 20 rows

#Schema

```
athletes.printSchema()
```

```
root
|-- PersonName: string (nullable = true)
|-- Country: string (nullable = true)
|-- Discipline: string (nullable = true)
```

#2. Coaches Data

```
coaches.show()
```

Name	Country	Discipline	Event
ABDELMAGID Wael	Egypt	Football	null
ABE Junya	Japan	Volleyball	null
ABE Katsuhiko	Japan	Basketball	null
ADAMA Cherif	Côte d'Ivoire	Football	null
AGEBA Yuya	Japan	Volleyball	null
AIKMAN Siegfried ...	Japan	Hockey	Men
AL SAADI Kais	Germany	Hockey	Men
ALAMEDA Lonni	Canada	Baseball/Softball	Softball
ALEKNO Vladimir	Islamic Republic ...	Volleyball	Men
ALEKSEEV Alexey	ROC	Handball	Women
ALLER CARBALLO Ma...	Spain	Basketball	null
ALSHEHRI Saad	Saudi Arabia	Football	Men
ALY Kamal	Egypt	Football	null
AMAYA GAITAN Fabian	Puerto Rico	Basketball	null
AMO AGUADO Pablo	Spain	Football	null
ANDONOVSKI Vlatko	United States of ...	Football	Women
ANNAN Alyson	Netherlands	Hockey	Women
ARNAU CREUS Xavier	Japan	Hockey	Women
ARNOLD Graham	Australia	Football	Men
AXNER Tomas	Sweden	Handball	Women

only showing top 20 rows

#Schema

```
coaches.printSchema()
```

```
root
|-- Name: string (nullable = true)
|-- Country: string (nullable = true)
|-- Discipline: string (nullable = true)
|-- Event: string (nullable = true)
```

#3. EntriesGender Data

entriesgender

Discipline	Female	Male	Total
3x3 Basketball	32	32	64
Archery	64	64	128
Artistic Gymnastics	98	98	196
Artistic Swimming	105	0	105
Athletics	969	1072	2041
Badminton	86	87	173
Baseball/Softball	90	144	234
Basketball	144	144	288
Beach Volleyball	48	48	96
Boxing	102	187	289
Canoe Slalom	41	41	82
Canoe Sprint	123	126	249
Cycling BMX Frees...	10	9	19
Cycling BMX Racing	24	24	48
Cycling Mountain ...	38	38	76
Cycling Road	70	131	201
Cycling Track	90	99	189
Diving	72	71	143
Equestrian	73	125	198
Fencing	107	108	215

only showing top 20 rows

#Schema

```
entriesgender.printSchema()
```

```
root
|-- Discipline: string (nullable = true)
|-- Female: integer (nullable = true)
|-- Male: integer (nullable = true)
|-- Total: integer (nullable = true)
```

Observation: As we can see in the output that Female & Male should have Integer datatype, but it is string. Therefore we need to change the datatype from string to integer.

#To fix the datatype from string to integer. [run 1 for col]

```
entriesgender =
entriesgender.withColumn("Female",col("Female").cast(IntegerType()))\
.withColumn("Male",col("Male").cast(IntegerType()))\
```

```
.withColumn("Total",col("Total").cast(IntegerType()))
```

#After fixing the data type

```
entriesgender.printSchema()
```

```
root
|-- Discipline: string (nullable = true)
|-- Female: integer (nullable = true)
|-- Male: integer (nullable = true)
|-- Total: integer (nullable = true)
```

#4. Medals Data

```
medals.show()
```

Rank	TeamCountry	Gold	Silver	Bronze	Total	Rank by Total
1	United States of ...	39	41	33	113	1
2	People's Republic...	38	32	18	88	2
3	Japan	27	14	17	58	5
4	Great Britain	22	21	22	65	4
5	ROC	20	28	23	71	3
6	Australia	17	7	22	46	6
7	Netherlands	10	12	14	36	9
8	France	10	12	11	33	10
9	Germany	10	11	16	37	8
10	Italy	10	10	20	40	7
11	Canada	7	6	11	24	11
12	Brazil	7	6	8	21	12
13	New Zealand	7	6	7	20	13
14	Cuba	7	3	5	15	18
15	Hungary	6	7	7	20	13
16	Republic of Korea	6	4	10	20	13
17	Poland	4	5	5	14	19
18	Czech Republic	4	4	3	11	23
19	Kenya	4	4	2	10	25
20	Norway	4	2	2	8	29

only showing top 20 rows

#Schema

```
medals.printSchema()
```

```
root
|-- Rank: integer (nullable = true)
```

```

|-- TeamCountry: string (nullable = true)
|-- Gold: integer (nullable = true)
|-- Silver: integer (nullable = true)
|-- Bronze: integer (nullable = true)
|-- Total: integer (nullable = true)
|-- Rank by Total: integer (nullable = true)

```

#5. Teams Data

```
teams.show()
```

TeamName	Discipline	Country	Event
Belgium	3x3 Basketball	Belgium	Men
China	3x3 Basketball	People's Republic...	Men
China	3x3 Basketball	People's Republic...	Women
France	3x3 Basketball	France	Women
Italy	3x3 Basketball	Italy	Women
Japan	3x3 Basketball	Japan	Men
Japan	3x3 Basketball	Japan	Women
Latvia	3x3 Basketball	Latvia	Men
Mongolia	3x3 Basketball	Mongolia	Women
Netherlands	3x3 Basketball	Netherlands	Men
Poland	3x3 Basketball	Poland	Men
ROC	3x3 Basketball	ROC	Men
ROC	3x3 Basketball	ROC	Women
Romania	3x3 Basketball	Romania	Women
Serbia	3x3 Basketball	Serbia	Men
United States	3x3 Basketball	United States of ...	Women
Australia	Archery	Australia	Men's Team
Australia	Archery	Australia	Mixed Team
Bangladesh	Archery	Bangladesh	Mixed Team
Belarus	Archery	Belarus	Women's Team

only showing top 20 rows

#Schema

```
teams.printSchema()
```

```

root
|-- TeamName: string (nullable = true)
|-- Discipline: string (nullable = true)
|-- Country: string (nullable = true)
|-- Event: string (nullable = true)

```

All done for the files and schema

#5. Few Insights Extraction

#Find countries with highest number of gold medals.

```
top_gold_medal_by_country =  
medals.select("TeamCountry", "Gold").orderBy("Gold",  
ascending=False).show()
```

```
+-----+-----+  
| TeamCountry | Gold |  
+-----+-----+  
| United States of ... | 39 |  
| People's Republic... | 38 |  
| Japan | 27 |  
| Great Britain | 22 |  
| ROC | 20 |  
| Australia | 17 |  
| Netherlands | 10 |  
| France | 10 |  
| Germany | 10 |  
| Italy | 10 |  
| Canada | 7 |  
| Brazil | 7 |  
| New Zealand | 7 |  
| Cuba | 7 |  
| Hungary | 6 |  
| Republic of Korea | 6 |  
| Poland | 4 |  
| Czech Republic | 4 |  
| Kenya | 4 |  
| Norway | 4 |  
+-----+-----+  
only showing top 20 rows
```

Calculate the average number of entries by gender for each discipline

```
average_entries_by_gender = entriesgender.withColumn(  
    'Avg_Female', entriesgender['Female'] / entriesgender['Total']  
)  
)  
average_entries_by_gender.withColumn(  
    'Avg_Male', entriesgender['Male'] / entriesgender['Total']  
)  
average_entries_by_gender.show()
```

```
+-----+-----+-----+-----+  
+-----+  
| Discipline | Female | Male | Total | Avg_Female |  
Avg_Male |  
+-----+-----+-----+-----+
```

```

+-----+
|      3x3 Basketball|      32|      32|      64|      0.5|
0.5|
|      Archery|      64|      64|     128|      0.5|
0.5|
| Artistic Gymnastics|      98|      98|     196|      0.5|
0.5|
|  Artistic Swimming|     105|       0|     105|      1.0|
0.0|
|      Athletics|     969|    1072|    2041| 0.4747672709456149|
0.5252327290543851|
|      Badminton|      86|      87|     173|0.49710982658959535|
0.5028901734104047|
|  Baseball/Softball|      90|     144|     234|0.38461538461538464|
0.6153846153846154|
|      Basketball|     144|     144|     288|      0.5|
0.5|
|  Beach Volleyball|      48|      48|      96|      0.5|
0.5|
|      Boxing|     102|     187|     289|0.35294117647058826|
0.6470588235294118|
|      Canoe Slalom|      41|      41|      82|      0.5|
0.5|
|      Canoe Sprint|     123|     126|     249| 0.4939759036144578|
0.5060240963855421|
|Cycling BMX Frees...|      10|       9|      19| 0.5263157894736842|
0.47368421052631576|
|  Cycling BMX Racing|      24|      24|      48|      0.5|
0.5|
|Cycling Mountain ...|      38|      38|      76|      0.5|
0.5|
|      Cycling Road|      70|     131|     201| 0.3482587064676617|
0.6517412935323383|
|      Cycling Track|      90|      99|     189|0.47619047619047616|
0.5238095238095238|
|      Diving|      72|      71|     143| 0.5034965034965035|
0.4965034965034965|
|      Equestrian|      73|     125|     198| 0.3686868686868687|
0.6313131313131313|
|      Fencing|     107|     108|     215|0.49767441860465117|
0.5023255813953489|
+-----+-----+-----+-----+-----+

```

only showing top 20 rows

All the connections, reading, transformation and analysis done. NEXT --> dump the transformed data into Transformed Data Lake

#Move the Transformed Cleaned Data to Transformed Data Lake.


```
#format
#filename.repartition(1).write.mode("overwrite").option("header",'true')
.csv("/mnt/tokyoolympic/@transformed Data Lake/filename")

athletes.repartition(1).write.mode("overwrite").option("header",'true')
.csv("/mnt/tokyoolympic/Transformed Data Lake/Athletes")
coaches.repartition(1).write.mode("overwrite").option("header","true")
.csv("/mnt/tokyoolympic/Transformed Data Lake/Coaches")
entriesgender.repartition(1).write.mode("overwrite").option("header","true")
.csv("/mnt/tokyoolympic/Transformed Data Lake/EntriesGender")
medals.repartition(1).write.mode("overwrite").option("header","true").
.csv("/mnt/tokyoolympic/Transformed Data Lake/Medals")
teams.repartition(1).write.mode("overwrite").option("header","true").c
sv("/mnt/tokyoolympic/Transformed Data Lake/Teams")
```

Pulled the Data from 'Raw Data Lake' -> Transformed in Databricks using Apache Spark ->
Dumped the transformed data in "Transformed Data Lake"

Next: Transformed ADLS --> Azure Synapse