## Script 2

In script2, data is read from two files student\_data.csv and major.csv. These tables have a common attribute, i.e. major\_id. Student\_data table has 25,000 rows and major table has 10 rows. The scripts performs a inner join, full outer join, a left outer join on the two tables on the key 'major id'.

In the above screen shot, the physical plan of full outer join where Sort Merge method is used is shown.

```
Endoministance CAMAINDOWN System 22 considers - spack-shell

Scala's spark.time(df3.join(df2,Seq("school_id"),"left").join(df1,Seq("major_id"),"left"))

Time taken: 20 ms

ress5: org.apacine.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]

scala's join_innee.explain()

--- Physical Plan

---- Physical Pl
```

The screen shot above shows the query plan for inner join where Broadcast join is used.

```
Administrator C.\Windows\System32\cmd.exe-spak-shell

calls join left.explain()
-- Physical Plan --
"3) Project [major_id#121, school_id#128, stud_id#111, first_name#112, last_name#113, birthday#114, gender#115, age#116, stud_email#1
17, stud_phone#118, stud_address#119, school_name#48, school_email#41, school_phone#42, street#43, city#44, state#45, zip_code#46, maj
on_name#11, course_duna*id=#12], [major_id#121], [major_id#12], [last_name#112, last_name#13], birthday#114, gender#115, age#116, stud_email#17, stud
phone#118, stud_address#119, major_id#121, school_name#48, school_email#41, school_phone#42, street#43, city#44, state#45, zip_code#4

i -- *(3) Project [school_id#128] major_id#121, school_id#39], LeftOuter, BuildRigh

: -- *(3) FloeScan_csv [stud_id#111,first_name#112, last_name#113, birthday#114, gender#115, age#116, stud_email#17, stud_phone#118,

: -- *(3) FloeScan_csv [stud_id#111,first_name#112,last_name#113,birthday#114.gender#115,age#116,stud_email#17,stud_phone#118,

stud_address#119,school_id#128, major_id#121 Batched: false, Format: CSV, Location: inMemory#ileindex[file:/Users/Administrator/Deskto
pystudent_data.csv), PartitionFilters: [], PushedFilters: [], Readschems: struct(stud_distring,first_name:string,last_name:string,

: +- *(3) Project [school_id#39, school_name#40, school_email#41, school_phone#42, street#43, city#44, state#45, zip_code#46

: +- *(1) Filter isnotnull(school_id#39)

: +- *(1) Filter isnotnull(school_id#39)

: +- *(2) Project [school_id#39, school_name#40, school_email#41, school_phone#42, street#43, city#44, state#45, zip_code#46

| +- *(2) Project [school_id#39], Readschems: struct(school_distring,school_mame*2, street#43, city#44, state#45, zip_code#46

| -- *(3) Project [major_id#16], major_name#11, course_duration#12]

+- *(2) Project [major_id#10], major_name#11, course_duration#12]

+- *(2) Project [major_id#10], major_name#11, course_duration#12]

+- *(2) Filter isnotnull(major_id#10), major_name#11, course_duration#12]

+- *(2) Filter isnotnull(major_id#10)
```

The screenshot above shows the plan for left outer join where Broadcast join is used.

The screen shot above explains the plan for right outer join with Sort Merge Join.

## **Performance Tuning:**

```
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Administrator: C:\Windows\System32\cmd.exe - spark-shell
scala> val join_inner=df3.join(df2,Seq("school_id"),"inner").join(df1,Seq("major_id"),"inner")
join_inner: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
 cala> val join_fullouter=df3.join(df2,Seq("school_id"),"fullouter").join(df1,Seq("major_id"),"fullouter")
oin_fullouter: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> val join_right=df3.join(df2,Seq("school_id"),"right").join(df1.Seq("major_id"),"right")
(console>:29: error: value Seq is not a member of org.apache.spark.sql.DataFrame
val join_right=df3.join(df2,Seq("school_id"),"right").join(df1.Seq("major_id"),"right")
scala> val join_right=df3.join(df2,Seq("school_id"),"right").join(df1,Seq("major_id"),"right")
join_right: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> val join_left=df3.join(df2,Seq("school_id"),"left").join(df1,Seq("major_id"),"left")
join_left: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"inner").join(df1,Seq("major_id"),"inner"))
res22: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
        spark.time(df3.join(df2,Seq("school_id"),"fullouter").join(df1,Seq("major_id"),"fullouter"))
res23: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"right").join(df1,Seq("major_id"),"right"))
res24: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"left").join(df1,Seq("major_id"),"left"))
 es25: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> join_inner.explain()
== Physical Plan ==
*(3) Project [major_id#121, school_id#120, stud_id#111, first_name#112, last_name#113, birthday#114, gender#115, age#116, stud_email#1
17, stud_phone#118, stud_address#119, school_name#40, school_email#41, school_phone#42, street#43, city#44, state#45, zip_code#46, maj
Select Administrator: C:\Windows\System32\cmd.exe - spark-shell
     . 49 elided
 cala> spark.conf.get("spark.sql.autoBroadcastJoinThreshold").toInt
es33: Int = 100000
scala> spark.conf.set("spark.sql.autoBroadcastJoinThreshold",-1)
scala> val join_inner=df3.join(df2,Seq("school_id"),"inner").join(df1,Seq("major_id"),"inner")
join_inner: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> val join_fullouter=df3.join(df2,Seq("school_id"),"fullouter").join(df1,Seq("major_id"),"fullouter")
join_fullouter: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> val join_right=df3.join(df2,Seq("school_id"),"right").join(df1,Seq("major_id"),"right")
join_right: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> val join_left=df3.join(df2,Seq("school_id"),"left").join(df1,Seq("major_id"),"left")
join_left: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"inner").join(df1,Seq("major_id"),"inner"))
es35: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"fullouter").join(df1,Seq("major_id"),"fullouter"))
 es36: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"right").join(df1,Seq("major_id"),"right"))
Time taken: 16 ms
 es37: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> spark.time(df3.join(df2,Seq("school_id"),"left").join(df1,Seq("major_id"),"left"))
res38: org.apache.spark.sql.DataFrame = [major_id: string, school_id: string ... 18 more fields]
scala> join_inner.explain()
== Physical Plan ==
 (9) Project [major_id#121, school_id#120, stud_id#111, first_name#112, last_name#115 Generate Test CSV or JSON Data - Mozilla Firefox 16, stud_email#1
```

Sr. No	Join Type	Broad Cast Join	Sort Merge Join
1	Inner	23 ms	26 ms
2	Full outer	23ms	17ms
3	Right Outer	17ms	16ms
4	Left Outer	20ms	11ms

Here we can say that the output time of the join varies for different joins . Incase the table size for the joins are almost equal , a sort merge join is preferred over Broad Cast join. If there is a comparable difference between the sizes of the table then Broad Cast join is preferred .