```
In [0]:
```

```
from pyspark.sql.types import *

from pyspark import SparkContext
from pyspark.sql import SparkSession

spark = SparkSession.builder.appName("my project 1").getOrCreate()
sc = spark.sparkContext

# Read a CSV into a dataframe
# There is a smarter version, that will first check if there is a Parquet file and use it
def load_PD_file(filename_or_dir, schema):
    dataPath = "/mnt/ddscoursedatastorage/fwm-stb-data/" + filename_or_dir
    df = spark.read.format("csv")\
        .option("header", "false")\
        .option("delimiter", "|")\
        .schema(schema)\
        .load(dataPath)
    return df
```

```
In [0]:
```

```
df1 = spark.read.csv("/mnt/ddscoursedatastorage/dds-students/test.csv")
```

# Reading the data files

# **Reading Refrence Data**

```
In [0]:
```

203581233

# **Reading Daily Program Data**

```
In [0]:
```

```
daily_prog_data = load_PD_file("Daily program data/" , daily_prog_schema)
```

## **Reading Program Viewing Data**

```
In [0]:
```

```
#Reading the 2.5% sample of the viewing data from a Parquet file
viewing data = spark.read.parquet('/sample viewing 2 5percent')
print(f'There are {viewing data.count():,} entries in viewing data dataframe!')
```

There are 130,289,194 entries in viewing data dataframe!

### **Reading Demographic Data**

In [0]:

```
# Reading the Demographic CSV file
demographic schema = StructType([StructField('household id',StringType()),
                      StructField('household size', IntegerType()),
                      StructField('num adults', IntegerType()),
                      StructField('num_generations', IntegerType()),
                      StructField('adult range', StringType()),
                      StructField('marital status', StringType()),
                      StructField('race code', StringType()),
                      StructField('presence children', StringType()),
                      StructField('num children', IntegerType()),
                      StructField('age children', StringType()), #format like range - 'bi
twise'
                      StructField('age range children', StringType()),
                      StructField('dwelling_type',StringType()),
                      StructField('home_owner_status',StringType()),
                      StructField('length residence', IntegerType()),
                      StructField('home market value',StringType()),
                      StructField('num vehicles',IntegerType()),
                      StructField('vehicle_make', StringType()),
                      StructField('vehicle model', StringType()),
                      StructField('vehicle year', IntegerType()),
                      StructField('net worth', IntegerType()),
                      StructField('income', StringType()),
                      StructField('gender individual', StringType()),
                      StructField('age individual', IntegerType()),
                      StructField('education highest', StringType()),
                      StructField('occupation highest', StringType()),
                      StructField('education 1', StringType()),
                      StructField('occupation_1',StringType()),
                      StructField('age 2', IntegerType()),
                      StructField('education 2', StringType()),
                      StructField('occupation 2', StringType()),
                      StructField('age 3', IntegerType()),
                      StructField('education_3', StringType()),
                      StructField('occupation 3', StringType()),
                      StructField('age_4', IntegerType()),
                      StructField('education 4',StringType()),
                      StructField('occupation 4',StringType()),
                      StructField('age 5', IntegerType()),
                      StructField('education_5', StringType()),
                      StructField('occupation_5', StringType()),
                      StructField('polit party regist', StringType()),
                      StructField('polit party input', StringType()),
                      StructField('household clusters', StringType()),
                      StructField('insurance groups',StringType()),
                      StructField('financial groups',StringType()),
                      StructField('green living', StringType())
demographic_data = load_PD_file("demographic/" , demographic_schema)
```

## **Question 2.1**

all single genres.unpersist()

```
In [0]:
from pyspark.sql.functions import *
# Removing irrelevant columns and rows that contain null values
ref data Q2 = ref data.distinct().select('device id', 'dma')\
                      .filter((col('dma').isNotNull()) | (col('dma') != 'null')).distinc
t().cache()
daily_prog_data_Q2 = daily_prog_data.select('prog_code', 'genre').dropna()\
                                    .filter((col('prog code') != 'null') | (col('genre')
) != 'null')).distinct().cache()
viewing data Q2 = viewing data.distinct().select('prog code', 'device id').dropna()\
                              .filter((col('prog code') != 'null') | (col('device id') !
= 'null')).cache()
In [0]:
# Finding the connections between dmas and genres, using three of the given relations.
joined 1 = ref data Q2.dropna().filter(col('device id') != 'null')\
                                .join(viewing data Q2, on='device id', how='inner').cach
joined = daily prog data Q2.withColumn("single genres", explode(split(col("genre"), ',')
))\
                        .drop('genre').join(joined 1.drop('device id'), on='prog code',
how='inner').cache()
joined.count()
Out[8]: 448056204
In [0]:
# Counting the number of views (non-distinct viewing events from 'viewing data' relation)
for every (single) genre
dma genre count = joined.groupBy('dma', 'single genres')\
                        .count().withColumnRenamed('count', 'views count').cache()
In [0]:
from pyspark.sql.functions import *
# Creating a dataframe that contains all the genres that exist in the database
all single genres = dma genre count.select('single genres').distinct().cache()
# Creating a dataframe that contains all the dmas that exist in the database
all dma = dma genre count.select('dma').distinct().cache()
# Doing all necessary joins in order to make sure every dma has an entry for every existi
ng genre,
# and wherever the were no viewing events for a certain genre in that dma, we'll put a ze
ro in the 'views count' column.
# Plus, ordering the relation as follows:
# all entries of a certain dma are one after the other in alphabethical order,
# then within each dma, order by number of vies of each genre.
all combinations dma genres = all single genres.crossJoin(all dma).cache()
dma all genres without zeros = all combinations dma genres.join(dma genre count,
                                                                on=['dma', 'single genr
es'],
                                                                how='left').cache()
dma all genres count = dma all genres without zeros.fillna(0) \
                                                   .orderBy(desc('dma'), desc('views co
unt')).cache()
```

```
all dma.unpersist()
all_combinations_dma_genres.unpersist()
Out[10]: DataFrame[single genres: string, dma: string]
In [0]:
dma all genres count.display()
               dma
                         single_genres views_count
 Wilkes Barre-Scranton-Hztn
                              News
                                     1236131
 Wilkes Barre-Scranton-Hztn
                             Reality
                                     1155820
 Wilkes Barre-Scranton-Hztn
                             Sitcom
                                      684124
 Wilkes Barre-Scranton-Hztn
                               Talk
                                      643980
 Wilkes Barre-Scranton-Hztn
                                      603677
                             Comedy
 Wilkes Barre-Scranton-Hztn
                          Crime drama
                                      590688
 Wilkes Barre-Scranton-Hztn
                                      558360
                             Drama
 Wilkes Barre-Scranton-Hztn
                             Children
                                      526718
 Wilkes Barre-Scranton-Hztn
                              Action
                                      520320
 Wilkes Barre-Scranton-Hztn
                         Documentary
                                      497372
In [0]:
# For each dma in the given list, print the 5 most viewed genres.
dma list = ['Waco-Temple-Bryan', 'New York', 'Washington, DC (Hagrstwn)']
for dma in dma list:
   tempi = dma all genres count.filter(col('dma') == dma).cache()
   tempi.show(5)
   tempi.unpersist()
dma all genres count.unpersist()
+----+
            dma|single_genres|views_count|
+----+
                                317154|
|Waco-Temple-Bryan| Reality|
                     News|
Talk|
                                 294323|
|Waco-Temple-Bryan|
                                 191534|
|Waco-Temple-Bryan|
                     Sitcom|
                                 184341|
|Waco-Temple-Bryan|
|Waco-Temple-Bryan|
                       Drama|
                                 159823|
+----+
only showing top 5 rows
+----+
 dma|single_genres|views_count|
+----+
|New York| News| 260400|
|New York| Reality| 207070|
|New York| Talk| 138032|
|New York| Sitcom|
                        1245091
|New York| Documentary|
+----+
only showing top 5 rows
+----+
     dma|single_genres|views_count|
+----+
|Washington, DC (H...| Reality| 309259|
                      News|
Sitcom|
Comedy|
|Washington, DC (H...|
|Washington, DC (H...|
|Washington, DC (H...| Comeay| Children|
                                    217373|
                                 182383|
                                    176962|
+----+
only showing top 5 rows
```

0 | [11] | D | D

г 1

```
Out[11]: DataFrame[ama: String, single genres: String, views count: pigint]
```

## **Question 2.2**

```
In [0]:
```

```
from pyspark.sql.functions import *
# Altering values in demographic data:
# Converting househols' incomes to numbers we can work with,
# Handling null values of certain columns.
demographic data Q2 = demographic data.select('household id', 'net worth', 'income') \
                                      .withColumn('income', when(demographic data.income
== 'A', '10')
                                                            .when(demographic data.inco
me == 'B', '11')
                                                             .when (demographic data.inco
me == 'C', '12')
                                                             .when(demographic data.inco
me == 'D', '13')
                                                             .when((demographic data.inc
ome == 'null')
                                                                 (demographic data.incom
e.isNull()), '0')
                                                             .otherwise(demographic data
.income)
         ) \
                                      .withColumn('income', col('income').cast("int"))\
                                      .withColumn('net worth', when((demographic data.ne
t_worth == 'null') |
                                                                     (demographic data.n
et_worth.isNull()), 0)
                                                               .otherwise(demographic d
ata.net worth)).cache()
# Filtering out null values and irrelevant columns
ref data Q22 = ref data.select('household id', 'dma').filter((col('dma').isNotNull()) |
(col('dma') != 'null')).cache()
```

#### In [0]:

```
# calculating the maximum income of household in the data
max_income = demographic_data_Q2.select(max("income")).first()[0]
# calculatig the maximum net worth of household in the data
max_net_worth = demographic_data_Q2.select(max("net_worth")).first()[0]
```

### In [0]:

```
from pyspark.sql.functions import *
# Calculating the wealth score of each dma.
# First, we'll find the connection between dmas and households using join,
# then calculating wealth score according to given formula,
# and lastly filtering out unnecessary columns.
dma net worth income = ref data Q22.filter((col('household id').isNotNull()) | (col('household id').isNotNull()) |
ehold id') != 'null'))\
                                    .join(demographic data Q2.filter((col('household id')
.isNotNull()) |
                                                                       (col('household id'
) != 'null')),
                                          'household id', 'left').cache()
wealth per dma = dma net worth income.groupBy('dma')\
                                      .agg(avg('net worth').alias('avg networth'), avg('i
ncome').alias('avg income'))
wealth per dma = wealth per dma.withColumn('wealth score',
                                           ((col('avg networth')/max net worth) + (col('a
vg income')/max income)))\
                                .select('dma', 'wealth score').distinct().cache()
# Ordering the relation according to wealth score and then the number of views.
```

```
genres_per_dma = wealth_per_dma.join(dma_all_genres_count, 'dma', 'left')
genres_per_dma = genres_per_dma.distinct().orderBy(desc('wealth_score'), desc('views_count')).cache()
```

### In [0]:

```
# Creating a list of all (and only) dmas, arranged in order from wealthiest to least weal
thy
all_dma_by_order_from_wealthiest = genres_per_dma.select('dma').distinct().collect()
all_dma_by_order_from_wealthiest = [str(dma).split("'")[1] for dma in all_dma_by_order_fr
om_wealthiest]
```

### In [0]:

```
used genre = 'null'
# Creating a copy of the dataframe
genres_per_dma_updater = genres_per_dma.cache()
# Creating a copy of the dataframe's schema only
dma_genre_df = genres_per_dma_updater.limit(0)
# Go through all dmas in the data- starting at the wealthiest in descending order.
for dma in all dma by order from wealthiest:
    # Initialize an empty dataframe for each dma so that we can add entries to it later.
   flag = -1
    # 8 in the required number of genres for each dma.
   for i in range(8):
        # Gets all the legal entries where the genres of each dma are not used,
        # and therfore can be chosen later for a specific dma.
       genres per dma updater = genres per dma updater.filter(col('single genres') != u
sed genre) \
                                                       .orderBy(desc('views count')).ca
che()
        # Taking 1 entry of the required dma and its genre- ordered descending by number
        entry_dma_genre = genres_per_dma_updater.filter(col('dma') == dma).orderBy(desc(
'views count')).limit(1).cache()
        genres per dma updater.unpersist()
        if entry dma genre.count() == 0:
            # If there are no unused genres for this dma, get out of the loop.
            flag = i
           break
        # Adding the entry to the results dataframe
        dma genre df = dma genre df.union(entry dma genre)
        # Extract the genre that was chosen one step earlier,
        # so that we can remove all entries containing this genre in the next loop.
       used genre = entry dma genre.select('single genres').first()[0]
       entry dma genre.unpersist()
    if flag == 0:
        # We are here only if there was no genre to add to a certain dma at all.
        # In this case, we are adding an entry for this dma with an empty string as its g
enre.
       entry dma genre = genres per dma.filter(col('dma') == dma).limit(1) \
                                        .withColumn('single_genres', lit(None))\
                                        .withColumn('views count', lit(0)).cache()
        dma genre df = dma genre df.union(entry dma genre)
        entry dma genre.unpersist()
# Combining all the registries of the same dma together, such that its genres are in a li
st.
# If this dma has no genres, it will contain an empty list.
dma genre df = dma genre df.orderBy(desc('views count')).drop('views count')\
                           .groupBy('dma', 'wealth score')\
                           .agg(collect list('single genres').alias('genres list'))
```

### In [0]:

```
# Ordering the dmas by wealth score and then printing according to requested format.
dma_genre_by_wealth = dma_genre_df.orderBy(desc('wealth_score'))
dma_genre_by_wealth.display()
dma_genre_by_wealth.show(25)
```

genres_list ±	wealth_score	dma
List()	1.623931623931624	San Antonio
List(News, Reality, Talk, Comedy, Sitcom, Drama, Sports event, Documentary)	1.323423017384553	Baltimore
List(Sports non-event, Adventure, Action, Crime drama, House/garden, Suspense, Children, Entertainment)	1.310195919188455	Detroit
List(Animated, Football, Crime, Fantasy, Newsmagazine, Mystery, Consumer, Special)	1.2711323615888497	Austin
List(Music, Shopping, Law, Cooking, Science fiction, Game show, Interview, Sports talk)	1.2668983216070457	Sacramnto-Stkton- Modesto
List(Educational, Basketball, Western, Animals, Spanish, Travel, Nature, Outdoors)	1.2523400679268373	San Francisco-Oak- San Jose
List(Auto, Soap, Fashion, Docudrama, Soccer, Historical drama, Anthology, Biography)	1.2461203339681173	Seattle-Tacoma

```
dma| wealth score|
  -----
         San Antonio| 1.623931623931624|
           Baltimore | 1.323423017384553 | [News, Reality, T... |
             Detroit | 1.310195919188455 | [Sports non-event... |
             Austin|1.2711323615888497|[Animated, Footba...|
|Sacramnto-Stkton-...|1.2668983216070457|[Music, Shopping,...|
|San Francisco-Oak...|1.2523400679268373|[Educational, Bas...|
      Seattle-Tacoma | 1.2461203339681173 | [Auto, Soap, Fash... |
|Cleveland-Akron (... | 1.219783174336312|[Military, Awards...|
             Toledo| 1.208659916433023|[History, Home im...|
|Harrisburg-Lncstr...|1.2079220508100939|[Public affairs, ...|
        Philadelphia | 1.205012520599199 | [Romance-comedy, ... |
           Lexington|1.1977755217465422|[Adults only, Hor...|
             Houston | 1.1913376528875275 | [Bicycle racing, ... |
        Portland, OR|1.1353253977210136|[Mixed martial ar...|
|Washington, DC (H...|1.1144216105587266|[Standup, Debate,...|
       Augusta-Aiken|1.1128715077227525|
             Atlanta|1.1111980389237475|[Environment, Par...|
|Little Rock-Pine ... | 1.082137552102464|[Surfing, Holiday...|
|Sioux Falls(Mitch...| 1.073030643581126|[Cheerleading, Ga...|
|Miami-Ft. Lauderdale| 1.06682853472421|[Boat, Bullfighti...|
             Eureka | 1.053123551146194 | [Holiday special,...|
    Dallas-Ft. Worth|1.0517613106688342|[Arm wrestling, D...|
   Fargo-Valley City|1.0446334537058672|[Cricket, Speed s...|
|Greenville-N.Bern...|1.0406826988440798|[Fencing, Badmint...|
           Amarillo|1.0374296708935984|[Holiday-children...|
only showing top 25 rows
```

# **Question 3**

```
In [0]:
```