MongoDB

1. Load the file restaurants.json into a MongoDB database. The file contains data on 3,772 restaurants in New York City.

Write the following queries in MongoDB (show the output of your queries):

```
a) Display all the restaurants located in the boroughs Bronx or Brooklyn.
  // 1.a
     {borough: {$in: ['Bronx', 'Brooklyn']}},
     {_id:0, name:1, borough:1}
    b) Find the restaurant id, name, borough and cuisine for those restaurants whose
          name starts with the letters 'Mad'.
                                                                          id": {
"$oid": "6730f2812a1543
           db.res.find(
                                                                         ,
borough": "Manhattan",
cuisine": "French",
name": "Madison Bistro",
restaurant_id": "40657588"
              {name: /^Mad/}, {restaurant_id:1, name :1, borough:1, cuisine:1}
3. Find the restaurants that have received a score between 80 and 90 (inclusive).
                                                          "name": "B.B. Kings"
      db.res.find(
                {$elemMatch:
                   {score: {$gte: 80, $lte:90}}
                                                           "name": "West 79Th Street Boat Basin Cafe"
```

4. Display the restaurant id and name of restaurants which have received a 'C' grade in year 2014.

```
| Comparison of the comparison
```

5. Find the cuisine that has the highest number of restaurants.

6. Find the restaurants that do not prepare an 'American' cuisine and their average grade score is higher than 30. Display the restaurant ids and their average score.

7.(g) For each restaurant display only the grades that were recorded from the year 2014 onwards.

```
2345678991113145167899212222222223333333333344423445
                                                                                    "date": {
    "$date": "2014-11-15T00:00:002"
                                                                                     date": {
    "$date": "2014-05-02T00:00:00Z"
                                                                                 id": {
"$oid": "6730f2812a15430da03bb361"
   db.res.aggregate(
       {sproject: {
                                                                                     date": {
    "$date": "2014-05-29T00:00:00Z"
          name: 1,
              $filter: {
   input: "$grades",
                 as: "grade",
cond: {$gte: ["$$grade.date", ISODate("2014-01-01")]}
                                                                                    "date": {
    "$date": "2014-01-14T00:00:002'
8.(h) Calculate the average score across all the restaurants in the collection.
     //8.(h) Calculate the average score across all the restaurants in the collection.
     db.res.aggregate(
           {$unwind: "$grades"},
           {\sqroup:{_id: null, tot_average_score: {\savg: '\sqrades.score'}}}
     ) ;
  😝 Piayground Result 🛝
     1
     2
                 "_id": null,
     3
      4
                  "tot_average_score": 11.427736743468195
     5
```

2. Create a collection named sales with the following documents:

```
{ "_id": 1, "city": "Berkeley", "state": "CA", "qty": 648 },
{ "_id": 2, "city": "Bend", "state": "OR", "qty": 491 },
{ "_id": 3, "city": "Kensington", "state": "CA", "qty": 233 },
{ "_id": 4, "city": "Eugene", "state": "OR", "qty": 842 },
{ "_id": 5, "city": "Reno", "state": "NV", "qty": 655 },
{ "_id": 6, "city": "Portland", "state": "OR", "qty": 408 },
{ "_id": 7, "city": "Sacramento", "state": "CA", "qty": 574 }
```

a)

```
2
                                                           "total_qty": 1741,
                                                 3
                                                           "state": "OR"
                                                 4
                                                 5
                                                         },
                                                 6
                                                 7
                                                           "total_qty": 1455,
                                                           "state": "CA"
                                                8
                                                 9
                                                         },
                                               10
{$group: {_id: '$state', total_qty: {$sum: '$qty'}}},
{$project:{
                                               11
                                                           "total_qty": 655,
   _id:0,
                                                           "state": "NV"
                                               12
    state:'$_id',
    total_qty:1
                                               13
                                               14
```

b)

c) Print the final sales collection

```
[myDatabase> db.sales.find()
[
    { _id: 2, city: 'Bend', state: 'OR', qty: 541 },
    { _id: 4, city: 'Eugene', state: 'OR', qty: 892 },
    {
        _id: 5,
        city: 'Reno',
        state: 'NV',
        qty: 655,
        salespeople: [ 'David', 'Lisa', 'James' ]
    },
    { _id: 6, city: 'Portland', state: 'OR', qty: 458 }
]
```

3. Load the file books.json into a MongoDB database. Write a Python script that lets the user enter a books category and prints the ISBN and titles of all the books in that category.

```
from pymongo import MongoClient
 2
 3
      client = MongoClient("mongodb://localhost:27017/")
      db = client['myDatabase'] # 데이터베이스 이름
      collection = db['books']
                                      # 컬렉션 이름
      category = input("Enter category: ")
 9
      books = collection.find({"categories": category}, {"isbn": 1, "title": 1, "_id": 0})
10
11
      found = False
12
      for book in books:
13
           print(f"ISBN: {book['isbn']}, title: {book['title']}")
14
           found = True
15
16
      if not found:
17
           print("No books found in this category.")
(base) hunjunsin@hunjunui-MacBookAir mongoDB % /Users/hunjunsin/.pyenv/versions/3.11
  python /Users/hunjunsin/Desktop/box/Introduction_DMB/mongoDB/books.py
 Enter category: Computer Graphics
ISBN: 1884777902, title: 3D User Interfaces with Java 3D
  ISBN: 133034054, title: Graphics File Formats
  ISBN: 1933988398, title: Gnuplot in Action
ISBN: 1884777473, title: The Awesome Power of Direct3D/DirectX ISBN: 138412146, title: Power-3D ISBN: 1930110022, title: Graphics Programming with Perl (base) hunjunsin@hunjunui-MacBookAir mongoDB % □
```

MapReduce

1. Projection $\pi S(R)$: From each tuple of relation R produce only the components for the attributes in S.

```
    1.map(key, value):
    2. projected_value = {attribute in S}
    3. emit(projected_value, None)
    1.reduce(projected_value, _):
    2. emit(projected_value, None)
```

2. Intersection RnS: Return the tuples that are present in both relations R and S. Assume that relations R and S have the same schema (same attributes and same type).

```
1.map(tuple):
2. emit(tuple, "R") // add R tag if tuple comes from R
3. emit(tuple, "S") // add S tag if tuple comes from S

1. reduce(tuple, tags):
2. // tags is the list of tag ['R','S']
3. if "R" in tags and "S" in tags:
4. emit(tuple, None) // tuple 이 R 과 S 모두에 존재하면 반환
```

- 3.Grouping $\gamma A, \theta(B)(R)$. Given a relation R(A, B, C), with one grouping attribute A, one aggregated attribute B, and another attribute C, which is neither grouped or aggregated:
- (a) Partition the tuples of R according to their values in attribute A. (b) For each group, aggregate the values in attribute B and apply function θ on the aggregated value (θ is an aggregation operation such as SUM, COUNT or MAX).

The result of this operation is one tuple for each group. That tuple has a component for the grouping attribute A, with the value common to tuples of that group. It also has a component for each aggregation $\theta(B)$, with the aggregated value for that group.

```
a)
map(tuple):
// tuple: (A, B, C)
key = tuple.A
value = tuple.B
emit(key, value)
b)
reduce(key, value_list):
aggregated_value = sum(value_list)
emit(key, aggregated_value)
```

1. Show the total number of movies in each genre with (a) the DataFrame API, (b) Spark SQL, and (c) RDD operations. Identify the most efficient method.

```
DataFrame API
 from pyspark.sql.functions import explode, col
 genre_count_df = movies_df.select(explode(col("genres")).alias("genre")).groupBy('genre').count()
 genre_count_df.show()
          genre | count |
          Crime| 2678|
        Romance | 3665
      Thriller 2658
     Adventure | 2045
          Drama | 13789
             Wari
                    794
  Documentary 2129
         Family|
                  1311
        Fantasy|
                  1153
        History|
                    999
       Mystery|
                   1259
       Musical
                    487
     Animation|
                    971
          Musicl
                    840
     Film-Noir|
                    105
          Shortl
                    478
         Horrorl
                   1703
        Western|
                    274
     Biography
                   1404
         Comedy |
                  7024
 only showing top 20 rows
```

```
Spark SQL
  # Register DataFrame as a SQL temporary view
  movies_df.createOrReplaceTempView("movies")
 # Execute SQL query
  genre_count_sql = spark.sql("""
      SELECT genre, COUNT(*) as count
     FROM (SELECT EXPLODE(genres) AS genre FROM movies)
     GROUP BY genre
  .....
  genre_count_sql.show()
  24/11/14 13:34:42 WARN Nat
         genre | count |
         Crime| 2678|
       Romance | 3665
      Thriller | 2658 |
     Adventure | 2045
         Drama | 13789
           Warl
                  794
  Documentary 2129
        Family|
                1311
       Fantasy |
                1153
                  999|
       History|
                1259
       Mystery|
       Musical
                  487
     Animation
                  971
         Music
                  840
     Film-Noir|
                  105
         Short
                  478
        Horror 1703
       Westernl
                  2741
     Biography|
                1404
        Comedy | 7024 |
 only showing top 20 rows
```

```
genre_count_rdd = movies_df.select("genres").rdd.\
    filter(lambda row: row["genres"] is not None).\
        flatMap(lambda row: row["genres"]).\
        map(lambda genre: (genre, 1)) .reduceByKey(lambda a, b: a + b)

print(genre_count_rdd.collect())

[("Romance', 3665), ("Horror', 1783), ("Sci-Fi', 1034), ("News', 51), ("History', 999), ("Adventure', 2045), ("Documentary', 2129), ("Crime', 2678), ("Mystery', 1259), ("Film-Noir', 105), ("Talk-Show', 1), ("Comedy', 7024), ("Sport', 390), ("Western', 274), ("Animation', 971), ("Family', 1311), ("Thiller', 2658), ("Drama', 13789), ("Action', 2539), ("Music', 840), ("Musical', 487), ("Short', 478), ("Fantasy', 1153), ("Biography', 1404), ("War', 794)]
```

2. Find the directors who directed the highest number of movies.

3. Determine the genres with the highest average IMDb rating (use the imdb.rating field).

```
from pyspark.sql.functions import explode, col, avg

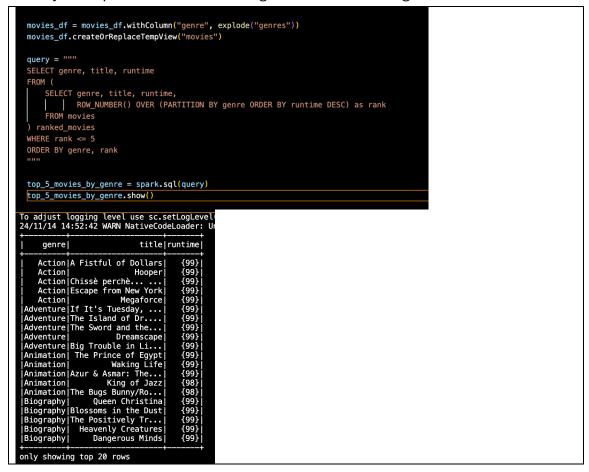
average_rating_df = movies_df \
    .withColumn("genre", explode("genres")) \
    .withColumn("rating", col("imdb.rating.$numberDouble").cast("double")) \
    .groupBy("genre") \
    .agg(avg("rating").alias("avg_rating"))\
    .orderBy('avg_rating', ascending= False)

average_rating_df.show(1)

24/11/14 14:10:24 WARN NativeCodeLoader: Un
    | genre | avg_rating |
    | Film-Noir|7.5041237113402035|
    | Film-Noir|7.5041237113402035|
    | Tow
```

4. Find the month with the most movie releases based on the released date.

5. Identify the top 5 movies with the longest runtime in each genre.



6. Find the top 10 actors who appeared in the most movies. For each actor, list the number of movies and their average IMDb rating.

```
from pyspark.sql.functions import explode, avg, count
movies_df = movies_df.withColumn("rating_float", col("imdb.rating.$numberDouble").cast("double")
movies_df = movies_df.withColumn("actor", explode("cast"))
result_df = movies_df.groupBy("actor")\
     .agg(count("title").alias("movie_count"), avg("rating_float").alias("avg_rating"))\
     .orderBy(col("movie_count").desc())
top_10 = result_df.limit(10)
top_10.show()
                  actor|movie_count|
                                                avg_rating|
                                       6.659375|
6.958490566037736|
     Gèrard Depardieu
       Robert De Niro
Michael Caine
                                   60
                                   53 | 6.604347826086958
50 | 7.1688888888888895
 Marcello Mastroianni
        Max von Sydow
Bruce Willis
                                   49 | 6.940000000000001
49 | 6.4468085106382995
  Morgan Freeman
Samuel L. Jackson
Christopher Plummer
                                       7.172727272727273
6.390476190476191
7.017948717948718
                                   48 j
                                   48
                                   47
          Gene Hackman
                                   46
                                       6.757142857142856
```

7. Plot a bar chart with the number of movies released each year.

