NoSQL_MongoDB_with_Python

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1 Course Section : AIT614-005

1.1 Lab 2 : NoSQL MongoDB with Python

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[1]:
```
!pip install pymongo
```

```
Requirement already satisfied: pymongo in c:\users\rashmika\anaconda3\lib\site-
packages (4.3.3)
Requirement already satisfied: dnspython<3.0.0,>=1.16.0 in
c:\users\rashmika\anaconda3\lib\site-packages (from pymongo) (2.3.0)
```

Importing the required libraries

[2]:
```
import pymongo
import pandas as pd
import json
```

Connect to MongoDB

[3]:
```
client = pymongo.MongoClient("mongodb://localhost:27017/")
```

Load the csv file

[4]:
```
df = pd.read_csv("EmployeeAttrition.csv")
df.head(10)
```

[4]:

	Age	Attrition	BusinessTravel	DailyRate	Department	\
0	41	Yes	Travel_Rarely	1102	Sales	
1	49	No	Travel_Frequently	279	Research & Development	
2	37	Yes	Travel_Rarely	1373	Research & Development	
3	33	No	Travel_Frequently	1392	Research & Development	
4	27	No	Travel_Rarely	591	Research & Development	
5	32	No	Travel_Frequently	1005	Research & Development	
6	59	No	Travel_Rarely	1324	Research & Development	
7	30	No	Travel_Rarely	1358	Research & Development	
8	38	No	Travel_Frequently	216	Research & Development	
9	36	No	Travel_Rarely	1299	Research & Development	

	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	\
0	1	2	Life Sciences	1	1	
1	8	1	Life Sciences	1	2	
2	2	2	Other	1	4	
3	3	4	Life Sciences	1	5	
4	2	1	Medical	1	7	
5	2	2	Life Sciences	1	8	
6	3	3	Medical	1	10	
7	24	1	Life Sciences	1	11	
8	23	3	Life Sciences	1	12	
9	27	3	Medical	1	13	

	…	RelationshipSatisfaction	StandardHours	StockOptionLevel	\
0	…	1	80	0	
1	…	4	80	1	
2	…	2	80	0	
3	…	3	80	0	
4	…	4	80	1	
5	…	3	80	0	
6	…	1	80	3	
7	…	2	80	1	
8	…	2	80	0	
9	…	2	80	2	

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance	YearsAtCompany	\
0	8	0	1	6	
1	10	3	3	10	
2	7	3	3	0	
3	8	3	3	8	
4	6	3	3	2	
5	8	2	2	7	
6	12	3	2	1	
7	1	2	3	1	
8	10	2	3	9	
9	17	3	2	7	

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7
2	0	0	0
3	7	3	0
4	2	2	2
5	7	3	6
6	0	0	0
7	0	0	0
8	7	1	8
9	7	7	7

```
[10 rows x 35 columns]
```

[5]:
```
df.shape
```

[5]:
```
(1470, 35)
```

Converting the dataframe to JSON format and Loading to MongoDB

[6]:
```
# Converting the df to json
emp_data = json.loads(df.to_json(orient='records'))
```

Creating a Database

[7]:
```
mongo_db = client["myDB"]
mongo_db
```

[7]:
```
Database(MongoClient(host=['localhost:27017'], document_class=dict,
tz_aware=False, connect=True), 'myDB')
```

Creating a collection

[8]:
```
collection_nm = "Empl_Attrition"
```

[9]:
```
collection_nm = mongo_db[collection_nm]
```

Insert the data into MongoDB Collection

[10]:
```
collection_nm.insert_many(emp_data)
```

[10]:
```
<pymongo.results.InsertManyResult at 0x2285f173100>
```

Query MongoDB

Count the total no of documents in the collection

[11]:
```
collection_nm.count_documents({})
```

[11]:
```
1470
```

1. Count the no of employees whose TotalWorkingYears are greater than 20.

[12]:
```
collection_nm.count_documents({
    "TotalWorkingYears" : {"$gt" : 20}  })
```

[12]:
```
207
```

2. Find EmployeeNumber, EducationField, JobRole for all the employees whose Age is between 25 and 30 and Education is 5. Display only EmployeeNumber, EducationField, and JobRobe in the output.

[13]:
```
res = collection_nm.find({'Age' : {'$gte' : 25, '$lte' : 30},
                          "Education" : 5},
                          {'EmployeeNumber', 'EducationField','JobRole'}
                         )
print('EmployeeNumber', '\t', 'EducationField', '\t\t', 'JobRole')
print('-------------------------------------------------------')
for r in res:
    print(r['EmployeeNumber'], '\t\t', r['JobRole'], '\t\t',␣
 ↪r['EducationField'])
```

```
EmployeeNumber   EducationField                  JobRole
-------------------------------------------------------
455              Laboratory Technician          Other
565              Research Scientist             Technical Degree
747              Sales Executive                Marketing
1094             Laboratory Technician          Life Sciences
```

3. For all the women employees having Age between 35 and 40 and TotalWorkingYears < 5, sort EmployeeNumber in an ascending order. Print only Department and EmployeeNumber in the output.

[14]:
```
# Adding conditions to the find function
emp_res= collection_nm.find(
    {"$and": [
        {"Gender" : 'Female'},
        {'Age' : {'$gte' : 35}},
        {'Age' : {'$lte' : 40}},
        {'TotalWorkingYears' : {'$lt':5}}
    ]},
    {'EmployeeNumber','Department'}
).sort('EmployeeNumber',1)
```

[15]:
```
# Converting the cursor to a list
emp_res_list = list(emp_res)
```

[16]:
```
# Converting the list to a dataframe
emp_df_3 = pd.DataFrame(emp_res_list)
emp_df_3.shape
```

[16]:
```
(9, 3)
```

[17]:
```
#Displaying the results
emp_df_3[['EmployeeNumber','Department']]
```

[17]:

	EmployeeNumber	Department
0	49	Sales
1	75	Research & Development
2	245	Research & Development
3	805	Sales
4	1569	Research & Development
5	1662	Research & Development
6	1675	Research & Development
7	1886	Research & Development
8	2052	Research & Development

4. Find employees whose HourlyRate is greater than or equal to 100 or DailyRate is greater than 1490. Display Age, HourlyRate, DailyRate, and Department only and sort DailyRate in an ascending order.

[18]:
```
# Adding conditions to the find function
emp_res4 = collection_nm.find(
    {'$or': [
        {'HourlyRate' : {'$gte':100}},
        {'DailyRate': {'$gt':1490}}
    ]},
    {'Age','HourlyRate','DailyRate','Department'}
).sort('DailyRate',1) # 1 means ascending order
```

[19]:
```
# Converting the cursor to a list
emp_res_list4 = list(emp_res4)
emp_res_list4
```

[19]:
```
[{'_id': ObjectId('63f5348794ccd61cbfa4e495'),
  'Age': 31,
  'DailyRate': 218,
  'Department': 'Sales',
  'HourlyRate': 100},
 {'_id': ObjectId('63f5348794ccd61cbfa4e79b'),
  'Age': 29,
  'DailyRate': 224,
  'Department': 'Research & Development',
  'HourlyRate': 100},
 {'_id': ObjectId('63f5348794ccd61cbfa4e55e'),
  'Age': 45,
  'DailyRate': 306,
  'Department': 'Sales',
  'HourlyRate': 100},
 {'_id': ObjectId('63f5348794ccd61cbfa4e911'),
  'Age': 38,
  'DailyRate': 345,
  'Department': 'Sales',
```

```
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e6a9'),
 'Age': 35,
 'DailyRate': 528,
 'Department': 'Human Resources',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e3e6'),
 'Age': 22,
 'DailyRate': 594,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e411'),
 'Age': 19,
 'DailyRate': 602,
 'Department': 'Sales',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e752'),
 'Age': 26,
 'DailyRate': 652,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e5c4'),
 'Age': 34,
 'DailyRate': 702,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e457'),
 'Age': 32,
 'DailyRate': 976,
 'Department': 'Sales',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e478'),
 'Age': 21,
 'DailyRate': 996,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e3ce'),
 'Age': 37,
 'DailyRate': 1040,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e4d5'),
 'Age': 50,
 'DailyRate': 1046,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e913'),
```

```
 'Age': 36,
 'DailyRate': 1120,
 'Department': 'Sales',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e607'),
 'Age': 33,
 'DailyRate': 1198,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e46a'),
 'Age': 32,
 'DailyRate': 1311,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e86f'),
 'Age': 38,
 'DailyRate': 1336,
 'Department': 'Human Resources',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e76f'),
 'Age': 31,
 'DailyRate': 1445,
 'Department': 'Research & Development',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e628'),
 'Age': 40,
 'DailyRate': 1479,
 'Department': 'Sales',
 'HourlyRate': 100},
{'_id': ObjectId('63f5348794ccd61cbfa4e677'),
 'Age': 40,
 'DailyRate': 1492,
 'Department': 'Research & Development',
 'HourlyRate': 61},
{'_id': ObjectId('63f5348794ccd61cbfa4e53d'),
 'Age': 38,
 'DailyRate': 1495,
 'Department': 'Research & Development',
 'HourlyRate': 76},
{'_id': ObjectId('63f5348794ccd61cbfa4e77a'),
 'Age': 49,
 'DailyRate': 1495,
 'Department': 'Research & Development',
 'HourlyRate': 96},
{'_id': ObjectId('63f5348794ccd61cbfa4e818'),
 'Age': 38,
 'DailyRate': 1495,
```

```
 'Department': 'Research & Development',
 'HourlyRate': 87},
{'_id': ObjectId('63f5348794ccd61cbfa4e401'),
 'Age': 29,
 'DailyRate': 1496,
 'Department': 'Research & Development',
 'HourlyRate': 41},
{'_id': ObjectId('63f5348794ccd61cbfa4e786'),
 'Age': 28,
 'DailyRate': 1496,
 'Department': 'Sales',
 'HourlyRate': 92},
{'_id': ObjectId('63f5348794ccd61cbfa4e740'),
 'Age': 39,
 'DailyRate': 1498,
 'Department': 'Sales',
 'HourlyRate': 44},
{'_id': ObjectId('63f5348794ccd61cbfa4e511'),
 'Age': 60,
 'DailyRate': 1499,
 'Department': 'Sales',
 'HourlyRate': 80}]
```

[20]:
```
# Converting the list to a dataframe
emp_df_4 = pd.DataFrame(emp_res_list4)
emp_df_4.shape
```

[20]:
```
(27, 5)
```

[21]:
```
#Displaying the results
emp_df_4.loc[ : , emp_df_4.columns != '_id']
```

[21]:

	Age	DailyRate	Department	HourlyRate
0	31	218	Sales	100
1	29	224	Research & Development	100
2	45	306	Sales	100
3	38	345	Sales	100
4	35	528	Human Resources	100
5	22	594	Research & Development	100
6	19	602	Sales	100
7	26	652	Research & Development	100
8	34	702	Research & Development	100
9	32	976	Sales	100
10	21	996	Research & Development	100
11	37	1040	Research & Development	100
12	50	1046	Research & Development	100
13	36	1120	Sales	100

```
14   33       1198  Research & Development         100
15   32       1311  Research & Development         100
16   38       1336         Human Resources         100
17   31       1445  Research & Development         100
18   40       1479                   Sales         100
19   40       1492  Research & Development          61
20   38       1495  Research & Development          76
21   49       1495  Research & Development          96
22   38       1495  Research & Development          87
23   29       1496  Research & Development          41
24   28       1496                   Sales          92
25   39       1498                   Sales          44
26   60       1499                   Sales          80
```

5. For each JobRole, find the average MonthlyIncome. Print out the formatted monthly incomes in hundredth and arrange them in descending order.

[22]:
```
emp_res5 = collection_nm.aggregate([
    {"$group": {
        "_id" : "$JobRole",
        "avg_monthly_income" : {"$avg" : '$MonthlyIncome'}
    }},
    {"$sort" : {
        "avg_monthly_income" : -1 }
    }
])


#Printing the results
print('Job Role', '\t\t\t\t', 'Average Monthly Income')
print('-----------------------------------------------------------------')
for r in emp_res5:
    print(f"{r['_id'] : <25}{'{:.2f}' : >30}".format(r['avg_monthly_income']))
```

```
Job Role                                  Average Monthly Income
-----------------------------------------------------------------
Manager                                                17181.68
Research Director                                      16033.55
Healthcare Representative                               7528.76
Manufacturing Director                                  7295.14
Sales Executive                                         6924.28
Human Resources                                         4235.75
Research Scientist                                      3239.97
Laboratory Technician                                   3237.17
Sales Representative                                    2626.00
```

6. Count the different MaritalStatus when Attrition is YES and AGE is greater than 35 in the dataset. Arrange the count in descending order.

[23]:
```
emp_res6 = collection_nm.aggregate([
    {
        '$match' : {
            '$and': [
                {"Attrition":'Yes'},
                {'Age': {'$gt' : 35}}
            ]
        }
    },
    {
        "$group" : {
            "_id" : "$MaritalStatus",
            "count_emp" : {"$sum" : 1}
        }
    },
    {"$sort" : {
        "count_emp" : -1 }
    }
])
```

[24]:
```
list(emp_res6)
```

[24]:
```
[{'_id': 'Married', 'count_emp': 33},
 {'_id': 'Single', 'count_emp': 30},
 {'_id': 'Divorced', 'count_emp': 14}]
```

Delete All Documents in a Collection

[25]:
```
collection_del = collection_nm.delete_many({})
```

Delete the Collection

[26]:
```
collection_nm.drop()
```

References [1] Dr. Liao's lab tutorials and code examples: Blackboard/Liao_PyMongo.html

[2] Python MongoDB - https://www.w3schools.com/python/python_mongodb_getstarted.asp

[3] PyMongoDB Documentation - https://pymongo.readthedocs.io/en/stable/