Data Science amb Python

Sprint 18

S18 T01: NoSQL database task

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Description

We are starting to get acquainted with NoSQL databases !!! Let's start with a few basic exercises

Nivel 1

Exercise 1

Create a NoSQL database using MongoDB. Add some sample data that allows you to check that you are able to process the information in a basic way.

Exercise 2

Connect the NoSQL database to Python using for example pymongo.

```
! pip install pymongo
In [1]:
        Requirement already satisfied: pymongo in /Applications/anaconda3/lib/pytho
        n3.8/site-packages (3.11.4)
         # import libraies
In [7]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         import pymongo
        # Creating a Database
In [8]:
         myclient = pymongo.MongoClient("mongodb://localhost:27017/")
         mydb = myclient["mydatabase"]
         mydb
```

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Out[8]: Database(MongoClient(host=['localhost:27017'], document_class=dict, tz_awar

e=False, connect=True), 'mydatabase')

```
#Check if Database Exists
 In [9]:
           print(myclient.list database names())
          ['admin', 'config', 'local']
 In [ ]:
          df teste = pd.read csv('test scores.csv')
In [41]:
           df_teste.values.tolist()
Out[41]: [['ANKYI',
            'Urban',
            'Non-public',
            '60L',
            'Standard',
            20.0,
            '2FHT3',
            'Female',
            'Does not qualify',
            62.0,
            72.0],
           ['ANKYI',
            'Urban',
            'Non-public',
            '60L',
            'Standard',
            20.0,
            '3JIVH',
            'Female',
            'Does not qualify',
            66.0,
            79.0],
           ['ANKYI',
'Urban',
            'Non-public',
            '60L',
            'Standard',
            20.0,
            '3XOWE',
            'Male',
            'Does not qualify',
            64.0,
            76.0],
           ['ANKYI',
            'Urban',
            'Non-public',
            '60L',
            'Standard',
            20.0,
            '55600',
            'Female',
            'Does not qualify',
            61.0,
            77.0],
           ['ANKYI',
            'Urban',
            'Non-public',
```

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'60L',

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'Standard',
 20.0,
 '74LOE',
 'Male',
 'Does not qualify',
 64.0,
 76.0],
['ANKYI',
 'Urban',
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 '60L',
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 20.0,
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 'Female',
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74.0],
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 'Non-public',
 '60L',
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 20.0,
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 'Male',
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['ANKYI',
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 20.0,
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 'Female',
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 63.0,
72.0],
['ANKYI',
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 'Non-public',
 '60L',
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 20.0,
 'CS5QP',
 'Male',
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77.0],
['ANKYI',
 'Urban',
 'Non-public',
 '60L',
 'Standard',
 20.0,
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 'Female',
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 61.0,
```

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```
72.0],
['ANKYI',
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 'Male',
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73.01,
['ANKYI',
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 'Non-public',
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74.0],
['ANKYI',
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 '60L',
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 'Male',
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78.0],
['ANKYI',
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 '60L',
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 20.0,
 'JQM2W',
 'Female',
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71.0],
['ANKYI',
 'Urban',
 'Non-public',
 '60L',
 'Standard',
 20.0,
 'MEUC4',
 'Female',
 'Does not qualify',
 64.0,
 77.0],
['ANKYI',
'Urban',
 'Non-public',
 '60L',
 'Standard',
```

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```
20.0,
 'R4U8H',
 'Male',
 'Does not qualify',
 64.0,
73.0],
['ANKYI',
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 '60L',
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70.0],
['ANKYI',
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 'Male',
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 73.0],
['ANKYI',
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 '60L',
 'Standard',
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 'Female',
 'Does not qualify',
 63.0,
71.0],
['ANKYI',
 'Urban',
 'Non-public',
 '60L',
 'Standard',
 20.0,
 'ZBQ4T',
 'Male',
 'Does not qualify',
 64.0,
73.0],
['ANKYI',
'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '0CRO6',
 'Male',
 'Does not qualify',
 60.0,
 68.0],
```

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```
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '1QMDI',
 'Male',
 'Does not qualify',
 60.0,
 70.01,
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '3CFUK',
 'Male',
 'Does not qualify',
 57.0,
 66.01,
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '44700',
 'Male',
 'Does not qualify',
 57.0,
70.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '4IDEM',
 'Female',
 'Qualifies for reduced/free lunch',
 56.0,
 65.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 '5UQNP',
 'Male',
 'Does not qualify',
 58.0,
 67.0],
['ANKYI',
'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
```

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```
'AV95M',
 'Female',
 'Does not qualify',
60.0,
70.01,
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
21.0,
 'BX6I6',
 'Female',
 'Does not qualify',
60.0,
67.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
21.0,
 'EYFXR',
 'Male',
 'Does not qualify',
54.0,
63.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
21.0,
 'FFC9M',
 'Male',
 'Does not qualify',
60.0,
75.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
21.0,
 'GGT4A',
 'Male',
 'Does not qualify',
58.0,
66.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
21.0,
 'HRZOJ',
 'Female',
 'Does not qualify',
66.0,
71.0],
['ANKYI',
```

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```
'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'IFC62',
 'Male',
 'Does not qualify',
 60.0,
 76.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'LLQI3',
 'Female',
 'Does not qualify',
 59.0,
 69.0],
['ANKYI',
'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'OCJE3',
 'Female',
 'Does not qualify',
 57.0,
69.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'OGW1F',
 'Female',
 'Does not qualify',
 60.0,
 68.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'Q5QRY',
 'Female',
 'Does not qualify',
 61.0,
 71.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'TJRIV',
```

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```
'Female',
 'Does not qualify',
 61.0,
 66.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'U24U5',
 'Female',
 'Does not qualify',
 59.0,
 69.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'VKX2N',
 'Female',
 'Does not qualify',
 62.0,
 72.0],
['ANKYI',
 'Urban',
 'Non-public',
 'ZNS',
 'Standard',
 21.0,
 'ZTROF',
 'Male',
 'Does not qualify',
 60.0,
 66.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 '1IALS',
 'Female',
 'Does not qualify',
 61.0,
 75.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 '5NDXD',
 'Male',
 'Qualifies for reduced/free lunch',
 58.0,
 78.0],
['CCAAW',
 'Suburban',
```

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```
'Non-public',
 '2B1',
 'Experimental',
 18.0,
 '6DCTV',
 'Female',
 'Qualifies for reduced/free lunch',
 64.0,
82.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 '60B0S',
 'Male',
 'Qualifies for reduced/free lunch',
58.0,
77.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'AITPY',
 'Male',
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65.0,
87.0],
['CCAAW',
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 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'B8B6G',
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65.0,
80.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
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 'Female',
 'Qualifies for reduced/free lunch',
62.0,
75.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'CD1MB',
 'Male',
```

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```
'Qualifies for reduced/free lunch',
58.0,
73.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'F4ZHY',
 'Male',
 'Qualifies for reduced/free lunch',
59.0,
74.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'GT9F2',
 'Male',
 'Qualifies for reduced/free lunch',
 59.0,
74.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'L5T1W',
 'Male',
 'Qualifies for reduced/free lunch',
63.0,
76.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
18.0,
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 'Female',
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63.0,
78.0],
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 'Non-public',
 '2B1',
 'Experimental',
18.0,
 'N7XOT',
 'Male',
 'Does not qualify',
59.0,
77.0],
['CCAAW',
 'Suburban',
 'Non-public',
```

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```
'2B1',
 'Experimental',
 18.0,
 'NKP72',
 'Male',
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 61.0,
77.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'QJVAM',
 'Male',
 'Qualifies for reduced/free lunch',
55.0,
72.0],
['CCAAW',
 'Suburban',
 'Non-public',
 '2B1',
 'Experimental',
18.0,
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 'Male',
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66.0,
83.0],
['CCAAW',
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 '2B1',
 'Experimental',
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 'Female',
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81.0],
['CCAAW',
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 'Non-public',
 '2B1',
 'Experimental',
 18.0,
 'YEJFP',
 'Female',
 'Does not qualify',
68.0,
79.0],
['CCAAW',
 'Suburban',
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 'EPS',
 'Experimental',
 20.0,
 '2RA1H',
 'Female',
 'Does not qualify',
```

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```
63.0,
80.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
 20.0,
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63.0,
83.01,
['CCAAW',
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 'Non-public',
 'EPS',
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66.0,
82.0],
['CCAAW',
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 'EPS',
 'Experimental',
 20.0,
 '3RJO7',
 'Female',
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67.0,
84.0],
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 'Non-public',
 'EPS',
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20.0,
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67.0,
83.0],
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 'Non-public',
 'EPS',
 'Experimental',
20.0,
 '4B72M',
 'Male',
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73.0,
85.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
```

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```
'Experimental',
 20.0,
 '7D9J0',
 'Male',
 'Does not qualify',
68.0,
87.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
 20.0,
 '7QNXG',
 'Female',
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70.0,
91.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
20.0,
 '8HMCH',
 'Female',
 'Qualifies for reduced/free lunch',
74.0,
85.0],
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 'Non-public',
 'EPS',
 'Experimental',
20.0,
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83.0],
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 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
20.0,
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 'Male',
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 62.0,
78.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
 20.0,
 'CFICH',
 'Female',
 'Does not qualify',
 76.0,
```

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84.01,
['CCAAW',
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 'Non-public',
 'EPS',
 'Experimental',
 20.0,
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73.0,
86.01,
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 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
20.0,
 'EEJ6S',
 'Male',
 'Does not qualify',
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87.0],
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 'Non-public',
 'EPS',
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79.0],
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 'EPS',
 'Experimental',
20.0,
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 'Female',
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66.0,
84.0],
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 'Non-public',
 'EPS',
 'Experimental',
 20.0,
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 'Female',
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82.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
```

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```
20.0,
 'SAKZI',
 'Female',
 'Qualifies for reduced/free lunch',
 65.0,
83.0],
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 'Suburban',
 'Non-public',
 'EPS',
 'Experimental',
20.0,
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 'Female',
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63.0,
78.0],
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 'Non-public',
 'EPS',
 'Experimental',
20.0,
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 'Female',
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68.0,
84.0],
['CCAAW',
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 'Non-public',
 'IQN',
 'Experimental',
15.0,
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 'Female',
 'Does not qualify',
65.0,
79.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
15.0,
 '4CE6S',
 'Female',
 'Does not qualify',
70.0,
83.0],
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 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'DB15U',
 'Female',
 'Qualifies for reduced/free lunch',
65.0,
81.0],
```

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```
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'GFIP5',
 'Female',
 'Qualifies for reduced/free lunch',
61.0,
74.0],
['CCAAW',
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 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'GJ8SE',
 'Male',
 'Qualifies for reduced/free lunch',
63.0,
77.0],
['CCAAW',
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 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'GYLNC',
 'Female',
 'Does not qualify',
62.0,
82.0],
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 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
15.0,
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 'Male',
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69.0,
81.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'JPSF7',
 'Male',
 'Qualifies for reduced/free lunch',
63.0,
75.01,
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
```

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```
'KW69V',
 'Male',
 'Qualifies for reduced/free lunch',
 62.0,
81.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'MY8XT',
 'Female',
 'Qualifies for reduced/free lunch',
63.0,
77.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'NKEK5',
 'Male',
 'Qualifies for reduced/free lunch',
59.0,
74.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'PCELB',
 'Male',
 'Does not qualify',
63.0,
80.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'VLKJU',
 'Male',
 'Does not qualify',
66.0,
76.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'WI916',
 'Female',
 'Does not qualify',
59.0,
77.0],
['CCAAW',
```

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```
'Suburban',
 'Non-public',
 'IQN',
 'Experimental',
 15.0,
 'WO5E2',
 'Male',
 'Qualifies for reduced/free lunch',
74.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'PGK',
 'Standard',
 21.0,
 '0U8DI',
 'Female',
 'Qualifies for reduced/free lunch',
 68.0,
76.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'PGK',
 'Standard',
21.0,
 '18YOG',
 'Male',
 'Qualifies for reduced/free lunch',
74.0,
82.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'PGK',
 'Standard',
21.0,
 '4YN68',
 'Female',
 'Qualifies for reduced/free lunch',
63.0,
73.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'PGK',
 'Standard',
21.0,
 '905ZT',
 'Female',
 'Qualifies for reduced/free lunch',
73.0,
78.0],
['CCAAW',
 'Suburban',
 'Non-public',
 'PGK',
 'Standard',
 21.0,
 '9WYVP',
```

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```
'Does not qualify',
            75.0,
            83.01,
           ['CCAAW',
            'Suburban',
            'Non-public',
            'PGK',
            'Standard',
            21.0,
            'A1554',
            'Female',
            'Does not qualify',
            78.0,
            84.0]]
           # Add some sample data
In [14]:
           mycol = mydb["customers"]
          mylist = [
               {'Country':'United States', 'First Name': 'Marshall', 'Last Name': 'Berno
               {'Country':'Ghana', 'First Name': 'Celinda', 'Last Name': 'Malkin', 'Te
               {'Country':'Ukraine', 'First Name' :'Guillermo', 'Last Name': 'Furze',''
               {'Country':'Greece','First Name':'Aharon','Last Name': 'Tunnow', 'Test
               {'Country':'Russia', 'First Name': 'Bail', 'Last Name': 'Goodwin', 'Tes
               {'Country':'Poland', 'First Name' :'Cole', 'Last Name': 'Winteringham' {'Country':'Sweden', 'First Name' :'Emlyn', 'Last Name': 'Erricker', 'Te
               {'Country':'Russia','First Name': 'Cathee', 'Last Name': 'Sivewright'
               {'Country':'China', 'First Name' :'Barny', 'Last Name': 'Ingerson', 'Tes
               {'Country':'Uganda', 'First Name': 'Sharla', 'Last Name': 'Papaccio',
           1
In [15]:
          x = mycol.insert_many(mylist)
          #print list of the id values of the inserted documents:
In [16]:
           print(x.inserted ids)
          [ObjectId('60ddbb84dd473efd9b0c27b3'), ObjectId('60ddbb84dd473efd9b0c27b4')
          , ObjectId('60ddbb84dd473efd9b0c27b5'), ObjectId('60ddbb84dd473efd9b0c27b6'
          ), ObjectId('60ddbb84dd473efd9b0c27b7'), ObjectId('60ddbb84dd473efd9b0c27b8
```

Nivel 2

bc')]

Exercises 1

'Male',

Load some simple queries to a Pandas Dataframe.

```
In [17]: # select data from a collection
    x1 = mycol.find_one()
    print(x1)

{'_id': ObjectId('60ddbb84dd473efd9b0c27b3'), 'Country': 'United States', '
    First Name': 'Marshall', 'Last Name': 'Bernadot', 'Test Score': 54}
```

'), ObjectId('60ddbb84dd473efd9b0c27b9'), ObjectId('60ddbb84dd473efd9b0c27ba'), ObjectId('60ddbb84dd473efd9b0c27bb'), ObjectId('60ddbb84dd473efd9b0c27bb'), ObjectId('60ddbb84dd473efd9b0c27bb')

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```
In [19]:
          x2 = mycol.find()
          entries = list(x2)
          entries[:5]
Out[19]: [{'_id': ObjectId('60ddbb84dd473efd9b0c27b3'),
            'Country': 'United States',
            'First Name': 'Marshall',
            'Last Name': 'Bernadot',
            'Test Score': 54},
          {' id': ObjectId('60ddbb84dd473efd9b0c27b4'),
            'Country': 'Ghana',
            'First Name': 'Celinda',
           'Last Name': 'Malkin',
            'Test Score': 51},
            '_id': ObjectId('60ddbb84dd473efd9b0c27b5'),
            'Country': 'Ukraine',
            'First Name': 'Guillermo',
           'Last Name': 'Furze',
           'Test Score': 53},
          {' id': ObjectId('60ddbb84dd473efd9b0c27b6'),
            'Country': 'Greece',
            'First Name': 'Aharon',
            'Last Name': 'Tunnow',
            'Test Score': 48},
          {' id': ObjectId('60ddbb84dd473efd9b0c27b7'),
            'Country': 'Russia',
            'First Name': 'Bail',
            'Last Name': 'Goodwin',
            'Test Score': 46}]
In [43]:
          df scores = pd.DataFrame(entries)
          df_scores
```

Out[43]:		_id	Country	First Name	Last Name	Test Score
	0	60ddbb84dd473efd9b0c27b3	United States	Marshall	Bernadot	54
	1	60ddbb84dd473efd9b0c27b4	Ghana	Celinda	Malkin	51
	2	60ddbb84dd473efd9b0c27b5	Ukraine	Guillermo	Furze	53
	3	60ddbb84dd473efd9b0c27b6	Greece	Aharon	Tunnow	48
	4	60ddbb84dd473efd9b0c27b7	Russia	Bail	Goodwin	46
	5	60ddbb84dd473efd9b0c27b8	Poland	Cole	Winteringham	49
	6	60ddbb84dd473efd9b0c27b9	Sweden	Emlyn	Erricker	55
	7	60ddbb84dd473efd9b0c27ba	Russia	Cathee	Sivewright	49
	8	60ddbb84dd473efd9b0c27bb	China	Barny	Ingerson	57
	9	60ddbb84dd473efd9b0c27bc	Uganda	Sharla	Papaccio	55

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```
In [23]: #Find document(s) with the country "Greece"

myquery_1 = { "Country": "Greece" }

mydoc = mycol.find(myquery_1)

df_list1 = []
    for x in mydoc:
        df_list1.append(x)

df_scores1 = pd.DataFrame(df_list1)
    df_scores1
```

```
Out[23]: __id Country First Name Last Name Test Score

O 60ddbb84dd473efd9b0c27b6 Greece Aharon Tunnow 48
```

```
In [24]: #Find documents where the First Name starts with the letter "G" or higher

myquery_2 = { "First Name": { "$gt": "G" } }

mydoc = mycol.find(myquery_2)

df_list2 = []
for x in mydoc:
    df_list2.append(x)

df_scores2 = pd.DataFrame(df_list2)
df_scores2
```

Out[24]:		_id	Country	First Name	Last Name	Test Score
	0	60ddbb84dd473efd9b0c27b3	United States	Marshall	Bernadot	54
	1	60ddbb84dd473efd9b0c27b5	Ukraine	Guillermo	Furze	53
	2	60ddbb84dd473efd9b0c27bc	Uganda	Sharla	Papaccio	55

Nivel 3

Exercise 1

Generates a statistical summary of the information contained in the database In this task, I will use other dataset because there are sufficient entries on it for the exercise.

```
In [25]: df_scores.info()
```

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```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 10 entries, 0 to 9
         Data columns (total 5 columns):
               Column
                           Non-Null Count Dtype
          0
               id
                           10 non-null
                                            object
              Country
                           10 non-null
          1
                                            object
           2
              First Name 10 non-null
                                            object
           3
              Last Name
                           10 non-null
                                            object
              Test Score 10 non-null
                                            int64
         dtypes: int64(1), object(4)
         memory usage: 528.0+ bytes
In [46]: # How many unique countries there are in the dataframe
          df_scores['Country'].nunique()
Out[46]: 9
In [47]:
         # How many entries per country
          df scores['Country'].value counts()
Out[47]: Russia
                           2
         Poland
         Sweden
                           1
         United States
         Uganda
                           1
         Greece
         Ukraine
                           1
                           1
         Ghana
         China
                           1
         Name: Country, dtype: int64
In [48]: | score_mean = pd.DataFrame(df_scores.groupby('Country')['Test Score'].mean(
          score_mean
                Country Test Score
Out[48]:
          0
                  China
                              57.0
          1
                  Ghana
                              51.0
          2
                 Greece
                             48.0
          3
                 Poland
                             49.0
          4
                  Russia
                             47.5
          5
                 Sweden
                             55.0
                 Uganda
          6
                             55.0
          7
                 Ukraine
                             53.0
         8 United States
                             54.0
          print('The maximum average score is', score_mean['Test Score'].max())
In [72]:
          print('The country with the highest average score is',
                 score_mean.at[score_mean['Test Score'].idxmax(),'Country'])
```

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The maximum average score is 57.0 The country with the highest average score is China

We can see that China has the highest average test score of this dataframe.

```
score_mean.at[score_mean['Test Score'].idxmax(),'Country']
In [70]:
          'China'
Out[70]:
In [73]:
           # Average test score of all students
           df scores['Test Score'].mean()
Out[73]: 51.7
           plt.figure(figsize = (10,8))
In [66]:
           sns.set theme(style="whitegrid")
           sns.barplot(x='Last Name', y ='Test Score', hue="Country", data = df_scores
           plt.legend(bbox_to_anchor=(1.05, 1.0),loc='upper left')
           plt.title('Test')
           plt.axhline(df scores['Test Score'].mean(), color='blue', linewidth=2);
                                                                                   United States
                                                                                   Ghana
                                                                                   Ukraine
                                                                                   Greece
           50
                                                                                   Russia
                                                                                   Poland
                                                                                   Sweden
                                                                                   Uganda
           40
           20
```

Individually, the largest note belongs to Ingerson of China, while the smallest belongs to Goodwin of Russia.

GoodwinWinteringham Erricker Sivewright Ingerson Papaccio

We can also see that 50% of students are above the average of 51.7.

10

0

Bernadot

Malkin

```
In [ ]:
```

Last Name

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