Resumen

0.1 Initial project description

We are going to create a database which use is going to be directly related to aliments. The goal of our database is to provide a determined patient with a diet following its main characteristics (age, height, weight) Therefore, we will need a group of nutritionists. They are the ones who will be in charge of the development of the diets. We will storage data for aliments mainly, and we will use this data in order to create new diets for our patients in the nutritional center.

We will use a relational database, a type of database that stores and provides access to data points that are related to one another.

In order to generate data for our database, we will use csv files for some of our tables. For the rest of the tables, we will try to generate data automatically with the tools we will see in class (Python)

0.2 RDBMS Deployment

In the schema of our database, we will have the following tables: Food, Patient, Nutritionist, Diet and List Of Food. One Diet will consist of different kinds of Food, a Diet will be created by a Nutritionist (although a Nutrionist can create more than one) and a Nutritionist will take care of one or more Patients. The table List Of Food is created as a consequence of the relation between Diet and Food (a relation M:N creates a new table including the primary keys of both tables included in the relation).

Firstly, we have introduced the table Food, which has the food ID as the primary key. The rest of columns of the table are related with information such as the energy in kcal and KJ that provides the food, the quantity of Protein and Fiber in food in every 100 grams... Then the table to register the information about the Patient has been created and fulfilled with relevant information about it in order

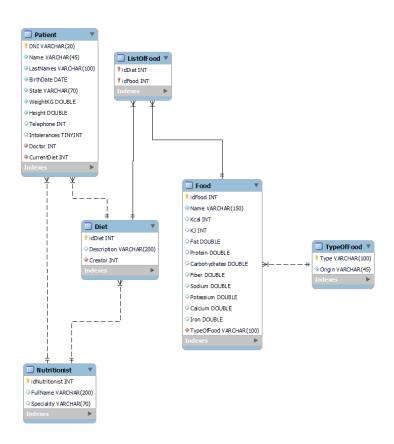


Figure 1: EER model

to organize the new diet. It also stores the current diet that the person is following and the intolerances, information that can be quite useful in order to design a new diet.

The table of the Nutritionist is created also with an ID as the primary key. It also contains information about the name of the nutritionist and the specialty that he or she has (obesity, for example)

For the diets we have another table. As an identifier, we have and ID for the diet (primary key) Also, a description and a Creator are columns for this table. The Description will contain the different steps to follow in order to do the diet. The data which has been introduced in the table Food has been obtained by a csv file in https://www.fao.org/infoods/infoods/tablas-y-bases-de-datos/es/

0.3 Query Design

· Food with more than 20 grams of protein (Fig. 2):

· All the names of patients treated by the doctor Pablo Moreno Garcia-Espina (Fig. 3):

· Description of the diets created by the doctor Susana Rocio Fernandez Giac-comassi (Fig. 4)

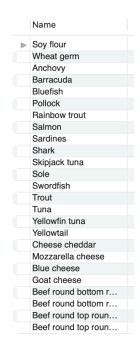


Figure 2: First query output

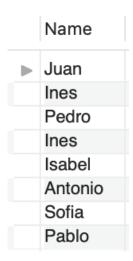


Figure 3: Second query output

	Description		
	Dieta sin lactosa para veganos		
	Dieta rica en fibra para mejorar la digestion		
	Dieta sin sal para mejorar la circulacion		
	Dieta para mejorar la salud intestinal		
	Dieta para reducir el estres		
	Dieta para mejorar la salud ocular		
	Dieta para mejorar la circulacion sanguinea		
	Dieta para mejorar la salud del higado		
	Dieta para mejorar la salud de las articulaciones		

Figure 4: Third query output

· Obtain the description and the name of their creator of all diets with more than 10 foods (Fig. 5):
than 10 100ds (11g. 3).
\cdot Get the names and protein content of foods that have more protein than the
average protein of all foods (Fig. 6):

	Diets with more than ten foods	Creator
	Dieta rica en proteinas para deportistas	Pablo Moreno Garcia-Espina
	Dieta para reducir el colesterol	Pablo Moreno Garcia-Espina
	Dieta para aumentar la masa muscular	Pablo Moreno Garcia-Espina
	Dieta para reducir la inflamacion	Pablo Moreno Garcia-Espina
	Dieta para mejorar la concentracion	Pablo Moreno Garcia-Espina
	Dieta para aliviar la artritis	Pablo Moreno Garcia-Espina
	Dieta para mejorar la digestion	Pablo Moreno Garcia-Espina
	Dieta para reducir la hinchazon abdominal	Pablo Moreno Garcia-Espina
	Dieta para reducir la retencion de liquidos	Pablo Moreno Garcia-Espina
	Dieta sin lactosa para veganos	Susana Rocio Fernandez Giaccomass
	Dieta rica en fibra para mejorar la digestion	Susana Rocio Fernandez Giaccomass
	Dieta sin sal para mejorar la circulacion	Susana Rocio Fernandez Giaccomass
	Dieta para mejorar la salud intestinal	Susana Rocio Fernandez Giaccomass
	Dieta para reducir el estres	Susana Rocio Fernandez Giaccomass
	Dieta para mejorar la salud ocular	Susana Rocio Fernandez Giaccomass
	Dieta para mejorar la circulacion sanguinea	Susana Rocio Fernandez Giaccomass
	Dieta para mejorar la salud del higado	Susana Rocio Fernandez Giaccomass
	Dieta para mejorar la salud de las articul	Susana Rocio Fernandez Giaccomass
	Dieta baja en grasas	Ismael Navas Delgado
	Dieta equilibrada para mantener el peso	Ismael Navas Delgado
	Dieta sin cafeina para mejorar el sueno	Ismael Navas Delgado

Figure 5: Fourth query ouput

	Name	Protein
	Bagel	11
	Buckwheat	13.3
	Einkorn wheat	16.5
	Kamut	14.3
	Millet	11
	Quinoa	14.1
	Soy flour	39.6
	Soybeans cooked	16.6
	Spelt raw	14.6
	Wheat bran	15.6
	Wheat germ	27.2
	Wheat whole grain	12.6
	Whole-grain bread	10.4
	Anchovy	20.3
	Barracuda	25.4
	Bluefish	20.4
	Catfish	18

Figure 6: Fifth query output

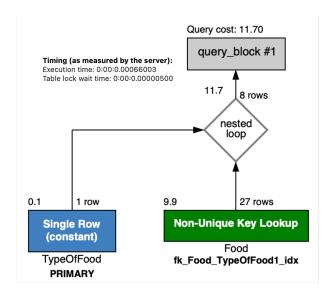


Figure 7: First query without index

0.4 RDBMS Optimization

Firstly, we have changed the first query in order to make a deeper search. We can also search in the database according to the type of food, as we made a join between the table of Food and the table of Type of Food.

In order to make the search easier when it comes to diets related with sporty people, we have decided to design two indexes: one with Name and Kcal and another with Protein and Name. This index becomes useful with our first query, as we are looking firstly at the food with protein's value of more than 20 grams.

For the previous index, when we try to execute the first query apparently the index is not used. That could be because the improvement when the index is used is not considerably high when you compare it with the performance without the index.

This picture (Fig. 7) shows the performance of the first query without the index.

This picture (Fig. 8) shows the performance of the first query with the index.

Furthermore, as we will frequently look for the different diets that one nutritionist has created, we will create an index for the column "Full Name" in the table Nutritionist. We already have an index for the id of a nutritionist, but we

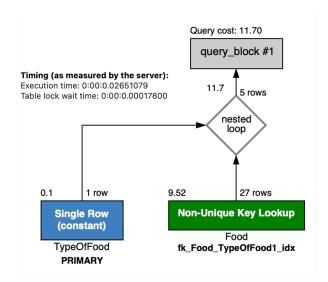


Figure 8: First query with index

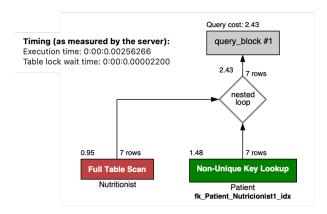


Figure 9: Second query without index

will typically get a diet of a nutritionist by his concrete name. With this index we will get an advantage in the second and third query, as they use the full name of the nutritionist in order to make the query.

This picture (Fig. 9) shows the time which lasts the second query without the index.

This picture (Fig. 10) shows the time which lasts the second query with the index.

This picture (Fig. 11) shows the performance of the first query without the index.

This picture (Fig. 12) shows the performance of the third query with the index.

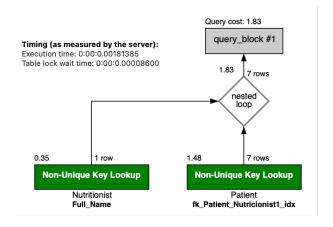


Figure 10: Second query with index

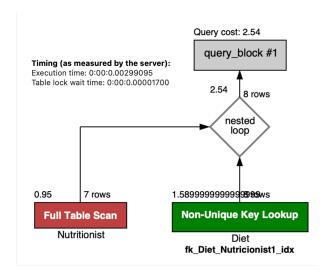


Figure 11: Third query without index

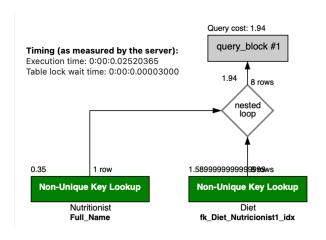


Figure 12: Third query with index

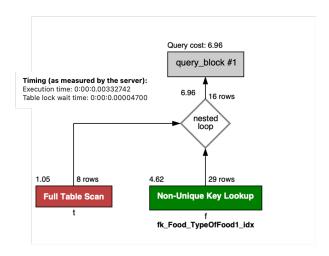


Figure 13: Extra query with index

We see how the time elapsed in the second time is lower than in the first time. The existence of the index makes the performance of the query higher.

We will also do queries which involve the name of the different patients we have in our database. Therefore, it would be suitable too to make an index with the name of the patients. Every time that we do a query in which the name of the patient is involved, we will get an advantage in terms of time, as we can see before.

We think that nutritionists may want to filter the aliments by the nutritional information and origin, so the query would be (Fig.9):

In this case, we have created an index for the table Food including the columns Protein and Iron, but it did not used it, neither the protein index created before.

0.5 Python

In order to develop the Python application to test the queries from an user interface, we have created a new project in PyCharm named NutritionalDB.

Specifically in the class NutritionalDB.py we have created three definitions (Fig. 9): one to initiate the connection to the database in SQL, one close the connection and another one to print the result of the queries we will implement next.

```
import pymysql
import pymysql.cursors
from pymysql import DatabaseError
class NutritionalQuerys(object):
    def __init__(self, host, user, password, database):
        self.connection = pymysql.connect(host=host,
                                            user=user,
                                            password=password,
                                            db=database,
                                            charset = 'utf8mb4',
                                            cursorclass =
   pymysql.cursors.DictCursor)
        print("Successfully Connected!")
    def close(self):
        self.connection.close()
        print("Connection Closed!")
    def print_query(self, query):
        cursor = self.exec_query(query)
        print()
        for row in cursor:
            print(row)
```

However, if we analyze the code for the "print query" definition we see that another definition called "exec query" is called. This new definition is used in order to send the query to the cursor created with the connection (Fig. 10)

```
def exec_query(self, query):
try:
```

After these first definitions, we have the definitions for each of the queries. When we made the queries in SQL we asked for concrete values of the columns (for example 20 grams of protein), however here we have parameterized the functions in order to make the query according to the user's goal.

When the user in Python calls the definition in order to ask for a food whose protein content is greater than X (Fig. 11), he has to specify this X value in the call to the method.

```
def gtProtein(self, protein):
    sql = 'SELECT NutritionalDB.Food.Name FROM Food
WHERE Food.Protein > ' + str(protein) + '; '
    print()
    print('Food with more than ' + str(protein) + '
grams of protein:')
    self.print_query(sql)
```

As we can see (Fig. 12), all of the definitions we have implemented in Python have the same structure. Firstly we create a string called sql in which we write the proper query. We have to insert here the different parameters we have in the header of the definition. Secondly, we print an empty line in order to make a separation in the terminal between queries. Then we print a text in order to make the user understand about the meaning of the query, and finally we execute the method "print query" in order to show in the terminal the results of the query.

We have created a method that receives a doctor as input and returns the patients treated by the input doctor.

```
def treatedBy(self, doctor):
    sql = "SELECT NutritionalDB.Patient.Name FROM
NutritionalDB.Patient JOIN Nutritionist " \
        "ON Patient.Doctor = Nutritionist.
idNutritionist " \
        "WHERE Nutritionist.FullName = '" + doctor + "
';"
    print()
    print('Patients treated by ' + str(doctor) + ':')
    self.print_query(sql)
```

The following definition receives an integer value and returns the diets with more than that number of food.

```
def dietsWithMoreFood(self, quantity):
     sql = "SELECT NutritionalDB.Diet.Description as " \
           "'Diets with more than " + str(quantity) + "
foods', "\
             "NutritionalDB. Nutritionist. FullName as '
Creator' FROM " \
             "NutritionalDB. Diet JOIN NutritionalDB.
Nutritionist ON " \
             "Diet. Creator = Nutritionist.idNutritionist
" \
             "WHERE NutritionalDB.Diet.idDiet IN (SELECT
" \
             "NutritionalDB.Diet.idDiet FROM
NutritionalDB.ListOfFood " \
             "GROUP BY idDiet HAVING COUNT(*) > " \
              "" + str (quantity) + ");"
     print()
     print('Diets with more than ' + str(quantity) + '
foods: ')
```

```
self.print_query(sql)
```

We can also get the food with more than the average of an input column. And the last definition receives two columns and two values, so we can filter by columns and grams.

```
def moreThanAvg(self, column):
     sql = "SELECT NutritionalDB. Food. Name, NutritionalDB
.Food." + column + \
           " FROM NutritionalDB.Food" \
           " WHERE NutritionalDB.Food." + column + " > (
SELECT AVG(NutritionalDB.Food." \
           + column + ") FROM NutritionalDB.Food);"
     print()
     print('Food with more ' + str(column) + ' than the
average: ')
     self.print query(sql)
 def filterByInfoAndOrigin(self, info1, info2, value1,
value2, origin):
     sql = "SELECT NutritionalDB.Food.Name FROM
NutritionalDB.Food" \
           " JOIN NutritionalDB. TypeOfFood ON Food.
TypeOfFood = TypeOfFood.Type" \
           " WHERE TypeOfFood.Origin = '" + origin + \
           "' AND (Food." + info1 + " > " + str(value1) +
 " OR Food." + info2 + \
           " > " + str (value2) + ");"
     print()
     print('Food with more than ' + str(value1) + ' grams
 of ' + str(info1) + ' or more than '
           + str(value2) + 'grams of ' + str(info2) + '
from ' + str(origin) + ' origin : ')
```

```
Food with more than 20 grams of protein:

{'Name': 'Pollock'}
{'Name': 'Sole'}
{'Name': 'Incut'}
{'Name': 'Sluefish'}
{'Name': 'Bluefish'}
{'Name': 'Salmon'}
{'Name': 'Salmon'}
{'Name': 'Shoulder noast'}
{'Name': 'Shoulder noast'}
{'Name': 'Shoulder noast'}
{'Name': 'Shoulder noast'}
{'Name': 'Beef round bottom round roast separable lean and fat trimmed to 0" fat all grades raw'}
{'Name': 'Shoulder noast'}
{'Name': 'Shorese'}
{'Name': 'Beef round bottom round roast separable lean and fat trimmed to 0" fat all grades raw'}
{'Name': 'Boek ground bottom round roast separable lean and fat trimmed to 0" fat all grades raw'}
{'Name': 'Boek ground 96% lean 4% fat raw'}
{'Name': 'Yellowtail'}
{'Name': 'Wild boar roast'}
{'Name': 'Wild boar roast'}
{'Name': 'Wild boar roast'}
{'Name': 'Beef shoulder pot roast or steak boneless separable lean only trimmed to 0" fat all grades raw'}
{'Name': 'Sardines'}
{'Name': 'Yeal chop'}
{'Name': 'Yeal chop'}
{'Name': 'Yeal chop'}
{'Name': 'Yeal chop'}
{'Name': 'Yeal shank'}
{'Name': 'Yeal shank'}
{'Name': 'Sarracuda'}
{'Name': 'Strip steak'}
{'Name': 'Ti-pone steak'}
{'Name': 'Ti-pone steak'}
{'Name': 'Yenison steak'}
```

Figure 14: gtProtein() output

```
self.print_query(sql)
```

Lastly, we have created a main class to test these definitions:

```
if __name__ == '__main__':
    password = 'Your_Password'
    ndb = NutritionalQuerys('localhost', 'root', password, '
    NutritionalDB')
    ndb.gtProtein(20)
    ndb.treatedBy('Pablo Moreno Garcia-Espina')
    ndb.dietsWithMoreFood(10)
    ndb.moreThanAvg('Protein')
    ndb.filterByInfoAndOrigin('Protein', 'Iron', 15, 5, '
    animal')
```

A preview of the outputs can be seen below in figures 15-18:

```
Patients treated by Pablo Moreno Garcia-Espina:

{'Name': 'Juan'}

{'Name': 'Ines'}

{'Name': 'Pedro'}

{'Name': 'Isabel'}

{'Name': 'Antonio'}

{'Name': 'Sofia'}

{'Name': 'Pablo'}
```

Figure 15: treatedBy() output

```
Clasts with more than 10 foods:

(Dists with more than 10 foods: 'Oieta vegetariana equilibrada', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods: 'Oieta detox para eliminar toxinas', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods: 'Oieta para aumentar not para edelgazar nagloa', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods: 'Oieta para aumentar la energia', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods: 'Oieta para aumentar la energia', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods: 'Oieta para aumentar la energia', 'Creator': 'Antonio Rodriguez Horcas')

(Dists with more than 10 foods': 'Oieta para reducir la hipertension', 'Creator': 'Antonio Rodriguez Horcas')

('Oiets with more than 10 foods': 'Oieta para nejorar la salud oses', 'Creator': 'Antonio Rodriguez Horcas')

('Oiets with more than 10 foods': 'Oieta para nejorar la salud oses', 'Creator': 'Antonio Rodriguez Horcas')

('Oiets with more than 10 foods': 'Oieta para nejorar la salud oses', 'Creator': 'Antonio Rodriguez Horcas')

('Oiets with more than 10 foods': 'Oieta para nejorar la salud oses', 'Creator': 'Cecilia Gonzalez Perez')

('Oiets with more than 10 foods': 'Oieta para nejorar la salud selective.' 'Creator': 'Cecilia Gonzalez Perez')

('Oiets with more than 10 foods': 'Oieta para fortalezer et sistema immunologio*, 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 foods': 'Oieta para mejorar la salud del cerebor, 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 foods': 'Oieta para mejorar la salud media', 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 foods': 'Oieta para mejorar la salud becal', 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 foods': 'Oieta para mejorar la salud becal', 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 foods': 'Oieta para mejorar la salud becal', 'Creator': 'Cecilia Gonzalez Perez')

('Dists with more than 10 fo
```

Figure 16: dietsWithMoreFood() output

```
{'Name': 'Whole-grain bread', 'Protein': 10.4}
{'Name': 'Reese's Peanut Butter Cups', 'Protein': 10.5}
{'Name': 'Reese's Peanut Butter Cups', 'Protein': 10.5}
{'Name': 'Reese's Peanut Butter Cups', 'Protein': 10.5}
{'Name': 'Rageqt', 'Protein': 11.0}
{'Name': 'Millet', 'Protein': 11.0}
{'Name': 'Ricotta cheese', 'Protein': 11.3}
{'Name': 'Soybeans', 'Protein': 12.0}
{'Name': 'Subcan', 'Protein': 13.3}
{'Name': 'Buckwheat', 'Protein': 13.3}
{'Name': 'Feta cheese', 'Protein': 14.2}
{'Name': 'Feta cheese', 'Protein': 14.2}
{'Name': 'Feta cheese', 'Protein': 14.3}
{'Name': 'Spult raw', 'Protein': 14.6}
{'Name': 'Sayuid', 'Protein': 15.6}
{'Name': 'Sussage', 'Protein': 15.6}
{'Name': 'Wheat bran', 'Protein': 15.6}
{'Name': 'Wheat bran', 'Protein': 16.5}
{'Name': 'Soybeans cooked', 'Protein': 16.6}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lea
```

Figure 17: moreThanAvg() output

0.6 NoSQL Database Design

In this section we are going to create an alternative to our SQL database. This is going to be done in MongoDB, using JSON documents to create the collections. The structure that our NoSQL database is going to have is simple, having one collection for each table on the relational SQL database.

To export our tables in the JSON format, we used a powerful tool of MySQL Workbench. We selected each table and chose the option "export wizard table". Then, we selected the csv format, and changed it for JSON file when saving it into the repository folder.

In order to create the different collections in MongoDB, we opened a new connection in MongoDB Compass and started to create new collections in a folder. We created the collections with the same name as the tables in SQL and import the data from the JSON files we have just generated.

Before we add the data to the collections we have to do a change in the JSON files. As MongoDB always renames the primary key with id, in order to relate

```
{'Name': 'Whole-grain bread', 'Protein': 10.4}
{'Name': "Reese's Peanut Butter Cups", 'Protein': 10.5}
{'Name': "Bagel', 'Protein': 11.0}
{'Name': 'Millet', 'Protein': 11.0}
{'Name': 'Millet', 'Protein': 11.0}
{'Name': 'Ricotta cheese', 'Protein': 11.3}
{'Name': 'Soybeans', 'Protein': 12.0}
{'Name': 'Wheat whole grain', 'Protein': 12.6}
{'Name': 'Buckwheat', 'Protein': 13.3}
{'Name': 'Quinoa', 'Protein': 14.1}
{'Name': 'Guinoa', 'Protein': 14.3}
{'Name': 'Spelt raw', 'Protein': 14.6}
{'Name': 'Squid', 'Protein': 14.6}
{'Name': 'Squid', 'Protein': 15.6}
{'Name': 'Sausage', 'Protein': 16.1}
{'Name': 'Wild boar sausage', 'Protein': 16.5}
{'Name': 'Beef ground 85% lean meat 15% fat cooked pan-broiled', 'Protein': 17.34}
{'Name': 'Beef ground 85% lean meat 15% fat raw', 'Protein': 17.34}
{'Name': 'Carp', 'Protein': 17.5}
{'Name': 'Carp', 'Protein': 17.5}
{'Name': 'Carp', 'Protein': 18.0}
{'Name': 'Shad', 'Protein': 18.0}
{'Name': 'Shad', 'Protein': 18.0}
{'Name': 'Halibut', 'Protein': 18.0}
{'Name': 'Halibut', 'Protein': 18.2}
{'Name': 'Halibut', 'Protein': 18.2}
{'Name': 'Ele', 'Protein': 18.5}
{'Name': 'Lilefish', 'Protein': 18.6}
```

Figure 18: filterByInfoAndOrigin() output

```
_id: 1
Description: "Dieta baja en grasas"
Creator: 3

_id: 2
Description: "Dieta sin lactosa para veganos"
Creator: 2

_id: 3
Description: "Dieta rica en proteinas para deportistas"
Creator: 1

_id: 4
Description: "Dieta vegetariana equilibrada"
Creator: 4
```

Figure 19: Diet Collection in MongoDB

the primary key in SQL with the primary key in MongoDB we have renamed the primary key in the JSON files as id.

These is the appearance of the different collections in MongoDB: (Fig. 19-24)

```
_id: 1
Name: "Asparagus"
Kcal: 20
KJ: 84
Fat: 0.2
Protein: 2.2
Carbohydrates: 3.9
Fiber: 2.1
Sodium: 2
Potassium: 202
Calcium: 24
Iron: 1.1
TypeOfFood: "vegetables"
_id: 2
Name: "Bell Pepper"
Kcal: 20
KJ: 84
Fat: 0.2
Protein: 0.9
Carbohydrates: 4.6
Fiber: 1.7
```

Figure 20: Food Collection in MongoDB

Sodium: 3

Potassium: 211

```
_id: ObjectId('6438338635aa65b68feebd89')
idDiet: 26
idFood: 1

_id: ObjectId('6438338635aa65b68feebd8a')
idDiet: 46
idFood: 1

_id: ObjectId('6438338635aa65b68feebd8b')
idDiet: 30
idFood: 4

_id: ObjectId('6438338635aa65b68feebd8c')
idDiet: 1
idFood: 5
```

Figure 21: ListOfFood Collection in MongoDB

```
_id: 1
FulName: "Pablo Moreno Garcia-Espina"
Speciality: "Deporte"

_id: 2
FulName: "Susana Rocio Fernandez Giaccomassi"
Speciality: "Descenso"

_id: 3
FulName: "Ismael Navas Delgado"
Speciality: "Deporte"

_id: 4
FulName: "Antonio Rodriguez Horcas"
Speciality: "Ascenso"
```

Figure 22: Nutritionist Collection in MongoDB

```
_id: "02358124Z"
Name: "Carmen"
LastNames: "Ruiz Soto"
BirthDate: "1952-08-08"
State: "descenso"
WeightKG: 89.71
Height: 1.86
Telephone: 608765432
Intolerances: 0
Doctor: 3
CurrentDiet: 44

_id: "05622134W"
Name: "Pilar"
LastNames: "Ortega Jimenez"
BirthDate: "1959-12-21"
```

Figure 23: Patient Collection in MongoDB

State: "ascenso"
WeightKG: 161.1
Height: 1.91

Intolerances: 0

CurrentDiet: 27

Doctor: 7

Telephone: 650123456

```
_id: "cereals and bread"
Origin: "processed"

_id: "chocolates and sweets"
Origin: "processed"

_id: "dairy"
Origin: "animal"

_id: "fish"
Origin: "animal"

_id: "fruits"
Origin: "vegetal"

_id: "legumes"
Origin: "vegetal"
```

Figure 24: TypeOfFood Collection in MongoDB