sql-session-2

September 23, 2024

[1]:
```
import pandas as pd
import sqlite3 as sql
```

[2]:
```
db = 'factbook.db'
```

[3]:
```
conn = sql.connect(db)
```

[4]:
```
conn
```

[4]:
```
<sqlite3.Connection at 0x18723a054e0>
```

[5]:
```
def read_q(q):
    return pd.read_sql_query(q, conn)
```

[6]:
```
read_q('select * from sqlite_master')
```

[6]:
```
    type             name         tbl_name  rootpage  \
0  table  sqlite_sequence  sqlite_sequence         3
1  table            facts            facts        47
2  table           cities           cities         2

                                                 sql
0             CREATE TABLE sqlite_sequence(name,seq)
1  CREATE TABLE "facts" ("id" INTEGER PRIMARY KEY…
2  CREATE TABLE cities (\n        id integer prim…
```

[7]:
```
read_q('select * from facts')
```

[7]:
```
      id code            name       area  area_land  area_water    population  \
0      1   af     Afghanistan   652230.0   652230.0         0.0  3.256434e+07
1      2   al         Albania    28748.0    27398.0      1350.0  3.029278e+06
2      3   ag         Algeria  2381741.0  2381741.0         0.0  3.954217e+07
3      4   an         Andorra      468.0      468.0         0.0  8.558000e+04
4      5   ao          Angola  1246700.0  1246700.0         0.0  1.962535e+07
..   …  …              …        …          …          …            …
256  257   zh  Atlantic Ocean        NaN        NaN         NaN           NaN
257  258   xo    Indian Ocean        NaN        NaN         NaN           NaN
```

```
258  259   zn   Pacific Ocean        NaN        NaN         NaN           NaN
259  260   oo  Southern Ocean        NaN        NaN         NaN           NaN
260  261   xx           World        NaN        NaN         NaN  7.256490e+09

     population_growth  birth_rate  death_rate  migration_rate
0                 2.32       38.57       13.89            1.51
1                 0.30       12.92        6.58            3.30
2                 1.84       23.67        4.31            0.92
3                 0.12        8.13        6.96            0.00
4                 2.78       38.78       11.49            0.46
..                 …          …          …               …
256                NaN         NaN         NaN             NaN
257                NaN         NaN         NaN             NaN
258                NaN         NaN         NaN             NaN
259                NaN         NaN         NaN             NaN
260               1.08       18.60        7.80             NaN

[261 rows x 11 columns]
```

[9]:
```
read_q('select * from cities')
```

[9]:
```
      id          name  population  capital facts_id
0      1    Oranjestad       37000        1      216
1      2  Saint John'S       27000        1        6
2      3     Abu Dhabi      942000        1      184
3      4         Dubai     1978000        0      184
4      5       Sharjah      983000        0      184
..   …           …          …      …       …
392  393       Mbabane       66000        1      167
393  394         Sanaa     2419000        1      193
394  395          Aden      784000        0      193
395  396        Lusaka     1802000        1      194
396  397        Harare     1542000        1      195

[397 rows x 5 columns]
```

1 Task 1

- write a query that shows all the information about 'Pakistan'

[8]:
```
read_q('select * from facts where name="Pakistan"')
```

[8]:
```
    id code      name    area  area_land  area_water  population  \
0  132   pk  Pakistan  796095     770875       25220   199085847

   population_growth  birth_rate  death_rate  migration_rate
0               1.46       22.58        6.49            1.54
```

- return all cities of country Pakistan

[10]:
```
read_q("""SELECT *
FROM cities
WHERE facts_id=132 """)
```

[10]:

	id	name	population	capital	facts_id
0	273	Karachi	13876000	0	132
1	274	Lahore	7566000	0	132
2	275	Faisalabad	3038000	0	132
3	276	Rawalpindi	2164000	0	132
4	277	Multan	1775000	0	132
5	278	Islamabad	919000	1	132

Primary key is a unique identifier for every record/row. It can not be null and should be unique.

Foreign key is a primary key in its own table. It is being referenced in another table to create a relationship.

ONE - TO - MANY - one Country can have Multiple Cities

[12]:
```
country = read_q('select * from facts')
city = read_q('select * from cities')
```

[13]:
```
country.head()
```

[13]:

	id	code	name	area	area_land	area_water	population	population_growth	birth_rate	death_rate	migration_rate
0	1	af	Afghanistan	652230.0	652230.0	0.0	32564342.0	2.32	38.57	13.89	1.51
1	2	al	Albania	28748.0	27398.0	1350.0	3029278.0	0.30	12.92	6.58	3.30
2	3	ag	Algeria	2381741.0	2381741.0	0.0	39542166.0	1.84	23.67	4.31	0.92
3	4	an	Andorra	468.0	468.0	0.0	85580.0	0.12	8.13	6.96	0.00
4	5	ao	Angola	1246700.0	1246700.0	0.0	19625353.0	2.78	38.78	11.49	0.46

[14]:
```
city.head()
```

[14]:

	id	name	population	capital	facts_id
0	1	Oranjestad	37000	1	216
1	2	Saint John'S	27000	1	6
2	3	Abu Dhabi	942000	1	184

```
3   4         Dubai     1978000         0      184
4   5       Sharjah      983000         0      184
```

[15]:
```
city['facts_id'] = city['facts_id'].astype(int)
```

[16]:
```
df = pd.merge(country, city, left_on='id', right_on='facts_id', how='inner')
```

[17]:
```
df
```

[17]:
```
     id_x code                    name_x       area  area_land  area_water  \
0       1   af               Afghanistan   652230.0   652230.0         0.0
1       2   al                   Albania    28748.0    27398.0      1350.0
2       3   ag                   Algeria  2381741.0  2381741.0         0.0
3       3   ag                   Algeria  2381741.0  2381741.0         0.0
4       4   an                   Andorra      468.0      468.0         0.0
..    …  …                         …         …          …          …
392   243   gq                      Guam      544.0      544.0         0.0
393   245   cq  Northern Mariana Islands      464.0      464.0         0.0
394   246   rq               Puerto Rico    13791.0     8870.0      4921.0
395   247   vq            Virgin Islands     1910.0      346.0      1564.0
396   255   wi            Western Sahara   266000.0   266000.0         0.0

     population_x  population_growth  birth_rate  death_rate  migration_rate  \
0      32564342.0               2.32       38.57       13.89            1.51
1       3029278.0               0.30       12.92        6.58            3.30
2      39542166.0               1.84       23.67        4.31            0.92
3      39542166.0               1.84       23.67        4.31            0.92
4         85580.0               0.12        8.13        6.96            0.00
..            …                  …          …           …               …
392      161785.0               0.54       16.82        5.12            6.34
393       52344.0               2.18       18.32        3.71            7.16
394     3598357.0               0.60       10.86        8.67            8.15
395      103574.0               0.59       10.31        8.54            7.67
396      570866.0               2.82       30.24        8.34             NaN

     id_y            name_y  population_y  capital  facts_id
0       6             Kabul       3097000        1         1
1      10            Tirana        419000        1         2
2       7           Algiers       2916000        1         3
3       8              Oran        783000        0         3
4      12  Andorra La Vella         23000        1         4
..    …               …           …      …        …
392   143           Hagatna        169000        1       243
393    99            Saipan         56000        1       245
394   295          San Juan       2475000        1       246
395   388  Charlotte Amalie         60000        1       247
396   391          Laayoune        237000        0       255
```

[397 rows x 16 columns]

[18]:
```
df[df['id_x'] == 132]
```

[18]:

	id_x	code	name_x	area	area_land	area_water	population_x	\
250	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	
251	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	
252	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	
253	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	
254	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	
255	132	pk	Pakistan	796095.0	770875.0	25220.0	199085847.0	

	population_growth	birth_rate	death_rate	migration_rate	id_y	\
250	1.46	22.58	6.49	1.54	273	
251	1.46	22.58	6.49	1.54	274	
252	1.46	22.58	6.49	1.54	275	
253	1.46	22.58	6.49	1.54	276	
254	1.46	22.58	6.49	1.54	277	
255	1.46	22.58	6.49	1.54	278	

	name_y	population_y	capital	facts_id
250	Karachi	13876000	0	132
251	Lahore	7566000	0	132
252	Faisalabad	3038000	0	132
253	Rawalpindi	2164000	0	132
254	Multan	1775000	0	132
255	Islamabad	919000	1	132

---

2 JOINS

using a common column (which is present in both tables), you can join two or more tables to retrieve information.

Return Pakistan and its cities

[19]:
```
# country name, city name
```

[20]:
```
read_q(""" select f.name, c.name
from facts f inner join
cities c on
f.id = c.facts_id
""")
```

[20]:
```
                    name          name
0                  Aruba    Oranjestad
1    Antigua and Barbuda  Saint John'S
2    United Arab Emirates     Abu Dhabi
3    United Arab Emirates         Dubai
4    United Arab Emirates       Sharjah
..                    …           …
392           Swaziland       Mbabane
393               Yemen         Sanaa
394               Yemen          Aden
395              Zambia        Lusaka
396            Zimbabwe        Harare

[397 rows x 2 columns]
```

[21]:
```
read_q(""" select f.name, c.name
from facts f inner join
cities c on
f.id = c.facts_id
where f.id = 132
""")
```

[21]:

	name	name
0	Pakistan	Karachi
1	Pakistan	Lahore
2	Pakistan	Faisalabad
3	Pakistan	Rawalpindi
4	Pakistan	Multan
5	Pakistan	Islamabad

3 Task 2

- return evry country and its capital city

[36]:
```
read_q(""" select f.name Country, c.name Cities
from facts f inner join
cities c on
f.id = c.facts_id
where c.capital = 1 and f.id = 132
""")
```

[36]:

	Country	Cities
0	Pakistan	Islamabad

4 Task 3

- return the name of the country and its relative city and its population which has a population of more than 10 million.

[43]:
```
read_q(""" select f.name Country, c.name City, c.population City_population
from facts f inner join
cities c on
f.id = c.facts_id
where c.population > 10000000
""")
```

[43]:

	Country	City	City_population
0	Argentina	Buenos Aires	13528000
1	Bangladesh	Dhaka	15391000
2	Brazil	Sao Paulo	19924000
3	China	Shanghai	20208000
4	China	Beijing	15594000
5	China	Guangzhou	10849000
6	Egypt	Cairo	11169000
7	France	Marseille-Aix-en-Provence	14890100
8	India	New Delhi	22654000
9	India	Mumbai	19744000
10	India	Kolkata	14402000
11	Japan	Tokyo	37217000
12	Japan	Osaka-Kobe	11494000
13	Mexico	Mexico City	20446000
14	Nigeria	Lagos	11223000
15	Pakistan	Karachi	13876000
16	Philippines	Manila	11862000
17	Russia	Moscow	11621000
18	Turkey	Istanbul	11253000
19	United States	New York-Newark	20352000
20	United States	Los Angeles-Long Beach-Santa Ana	13395000

5 Task 4

- Country name, Total population, Urban Areas Population

[55]:
```
read_q(""" select f.name Country, f.population Total_Population, sum(c.
 ↪population) Urban_population
from facts f inner join cities c
on f.id = c.facts_id
group by f.name
order by 3 desc
""")
```

[55]:

	Country	Total_Population	Urban_population
0	India	1251695584	82035000
1	China	1367485388	66849000
2	United States	321368864	60116000
3	Japan	126919659	60077000
4	Brazil	204259812	38086000
..	…	…	…
205	Falkland Islands (Islas Malvinas)	3361	2000
206	Anguilla	16418	2000
207	Wallis and Futuna	15613	1000
208	Saint Helena, Ascension, and Tristan da Cunha	7795	1000
209	Palau	21265	1000

[210 rows x 3 columns]

6 Task 5

- find out the percentage of urban to total population

[58]:
```
read_q(""" select f.name Country, f.population Total_Population, sum(c.
 ↪population) Urban_population,
round((sum(c.population)/cast(f.population as float)) * 100, 3) per_Urban
from facts f inner join cities c
on f.id = c.facts_id
group by f.name
order by 3 desc
""")
```

[58]:

	Country	Total_Population	\
0	India	1251695584	
1	China	1367485388	
2	United States	321368864	
3	Japan	126919659	
4	Brazil	204259812	

```
..                                                 …                …
205              Falkland Islands (Islas Malvinas)               3361
206                                       Anguilla              16418
207                                Wallis and Futuna              15613
208  Saint Helena, Ascension, and Tristan da Cunha               7795
209                                         Palau              21265

     Urban_population  per_Urban
0            82035000      6.554
1            66849000      4.888
2            60116000     18.706
3            60077000     47.335
4            38086000     18.646
..                …          …
205              2000     59.506
206              2000     12.182
207              1000      6.405
208              1000     12.829
209              1000      4.703

[210 rows x 4 columns]
```

[ ]: