

Manipulating data using pandas

Introduction to Pandas

In this section of the course we will learn how to use pandas for data analysis. Pandas is a fast, powerful, flexible and easy to use open source data analysis and manipulation tool, built on top of the Python programming language. You can think of pandas as an extremely powerful version of Excel, with a lot more features. In this section of the course, you should go through the notebooks in this order:

- Introduction to Pandas
- Series
- DataFrames
- Missing Data
- GroupBy
- Merging,Joining,and Concatenating
- Operations
- Data Input and Output

First you must have pandas library which can be installed using this command

In [91]:

```
!pip install pandas
```

```
Requirement already satisfied: pandas in c:\anaconda3\lib\site-packages (0.24.2)
Requirement already satisfied: pytz>=2011k in c:\anaconda3\lib\site-packages (from pandas) (2019.1)
Requirement already satisfied: python-dateutil>=2.5.0 in c:\anaconda3\lib\site-packages (from pandas) (2.8.0)
Requirement already satisfied: numpy>=1.12.0 in c:\anaconda3\lib\site-packages (from pandas) (1.16.4)
Requirement already satisfied: six>=1.5 in c:\anaconda3\lib\site-packages (from python-dateutil>=2.5.0->pandas) (1.12.0)
Could not fetch URL https://pypi.org/simple/pip/: (https://pypi.org/simple/pip/) There was a problem confirming the ssl certificate: HTTPSConnectionPool(host='pypi.org', port=443): Max retries exceeded with url: /simple/pip/ (Caused by SSLError(SSLCertVerificationError(1, '[SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: unable to get local issuer certificate (_ssl.c:1056)')))) - skipping
```

Import Pandas library using this command

In [92]:

```
import pandas as pd
```

Series

The first main data type we will learn about for pandas is the Series data type.

A Series is very similar to a NumPy array (it is built on top of the NumPy array object).

What differentiates the NumPy array from a Series?

- 1) is that a Series can have axis labels, meaning it can be indexed by a label, instead of just a number location.
- 2) It also doesn't need to hold numeric data, it can hold any arbitrary Python Object.

Let's explore this concept through some examples:

In [93]:

```
import numpy as np
import pandas as pd
```

You can convert a list, numpy array, or dictionary to a Series:

In [94]:

```
labels = ['a', 'b', 'c']
my_list = [10, 20, 30]
arr = np.array([10, 20, 30])
d = {'a': 10, 'b': 20, 'c': 30}
```

Using Lists

In [95]:

```
pd.Series(data=my_list)
```

Out[95]:

```
0    10
1    20
2    30
dtype: int64
```

In [96]:

```
pd.Series(data=my_list, index=labels)
```

Out[96]:

```
a    10
b    20
c    30
dtype: int64
```

NumPy Arrays

In [97]:

```
pd.Series(arr)
```

Out[97]:

```
0    10
1    20
2    30
dtype: int32
```

In [98]:

```
pd.Series(arr,labels)
```

Out[98]:

```
a    10
b    20
c    30
dtype: int32
```

Dictionary

In [99]:

```
pd.Series(d)
```

Out[99]:

```
a    10
b    20
c    30
dtype: int64
```

Using an Index

The key to using a Series is understanding its index. Pandas makes use of these index names or numbers by allowing for fast look ups of information

In [100]:

```
ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan'])
```

In [101]:

```
ser2 = pd.Series([6,7,8,9],index = ['USA', 'Germany','Italy', 'Japan'])
```

In [102]:

```
ser1
```

Out[102]:

```
USA      1
Germany  2
USSR     3
Japan    4
dtype: int64
```

In [103]:

```
ser2
```

Out[103]:

```
USA      6
Germany  7
Italy    8
Japan    9
dtype: int64
```

In [104]:

```
ser1[0]
```

Out[104]:

```
1
```

In [105]:

```
ser2[3]
```

Out[105]:

```
9
```

Operations are then also done based off of index:

In [106]:

```
ser1 + ser2
```

Out[106]:

```
Germany    9.0
Italy      NaN
Japan     13.0
USA        7.0
USSR       NaN
dtype: float64
```

DataFrames

One basic structure that you get with pandas is a data frame. A data frame is a two dimensional grid, rather similar to a relational database table except in memory.

DataFrames are the workhorse of pandas and are directly inspired by the R programming language.

We can think of a DataFrame as a bunch of Series objects put together to share the same index

In [107]:

```
from numpy.random import randn
np.random.seed(101)
```

In [108]:

```
df = pd.DataFrame(randn(5,4),index='A B C D E'.split(),columns='W X Y Z'.split())
```

In [109]:

```
df
```

Out[109]:

| | W | X | Y | Z |
|---|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 |
| E | 0.190794 | 1.978757 | 2.605967 | 0.683509 |

Selection and Indexing

How to grab data from a DataFrame

In [110]:

```
df['W']
```

Out[110]:

```
A    2.706850
B    0.651118
C   -2.018168
D    0.188695
E    0.190794
Name: W, dtype: float64
```

In [111]:

```
# Pass a list of column names
df[['W', 'Z']]
```

Out[111]:

| | W | Z |
|---|-----------|-----------|
| A | 2.706850 | 0.503826 |
| B | 0.651118 | 0.605965 |
| C | -2.018168 | -0.589001 |
| D | 0.188695 | 0.955057 |
| E | 0.190794 | 0.683509 |

DataFrame Columns are just Series

In [112]:

```
type(df['W'])
```

Out[112]:

pandas.core.series.Series

Creating a new column:

In [113]:

```
df['new'] = df['W'] + df['Y']
```

In [114]:

```
df
```

Out[114]:

| | W | X | Y | Z | new |
|---|-----------|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 | 3.614819 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 | -0.196959 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 | -1.489355 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 | -0.744542 |
| E | 0.190794 | 1.978757 | 2.605967 | 0.683509 | 2.796762 |

Removing Columns

In [115]:

```
df.drop('new',axis=1)
```

axis = 1 is referring to column

Out[115]:

| | W | X | Y | Z |
|---|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 |
| E | 0.190794 | 1.978757 | 2.605967 | 0.683509 |

but the 'new' column not permanently deleted from memory

In [116]:

```
df
```

Out[116]:

| | W | X | Y | Z | new |
|---|-----------|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 | 3.614819 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 | -0.196959 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 | -1.489355 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 | -0.744542 |
| E | 0.190794 | 1.978757 | 2.605967 | 0.683509 | 2.796762 |

Not inplace unless specified!

In [117]:

```
df.drop('new',axis=1,inplace=True)
```

In [118]:

```
df
```

Out[118]:

| | W | X | Y | Z |
|---|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 |
| E | 0.190794 | 1.978757 | 2.605967 | 0.683509 |

Removing Rows

In [119]:

```
df.drop('E',axis=0)
```

Out[119]:

| | W | X | Y | Z |
|---|-----------|-----------|-----------|-----------|
| A | 2.706850 | 0.628133 | 0.907969 | 0.503826 |
| B | 0.651118 | -0.319318 | -0.848077 | 0.605965 |
| C | -2.018168 | 0.740122 | 0.528813 | -0.589001 |
| D | 0.188695 | -0.758872 | -0.933237 | 0.955057 |

Selecting Rows

In [120]:

```
df.loc['A']
```

Out[120]:

```
W    2.706850
X    0.628133
Y    0.907969
Z    0.503826
Name: A, dtype: float64
```

Or you can select based off of position instead of label

In [121]:

```
df.iloc[2]
```

Out[121]:

```
W    -2.018168
X     0.740122
Y     0.528813
Z    -0.589001
Name: C, dtype: float64
```

**Selecting subset of rows and columns **

In [122]:

```
df.loc['B', 'Y']
```

Out[122]:

```
-0.8480769834036315
```

In [123]:

```
df.loc[['A', 'B'], ['W', 'Y']]
```

Out[123]:

| | W | Y |
|---|----------|-----------|
| A | 2.706850 | 0.907969 |
| B | 0.651118 | -0.848077 |

Example 1

In [124]:

```
##{key : list[:value[]]}

data = {'state': ['Jakarta', 'Jakarta', 'Jakarta', 'Selangor', 'Selangor', 'Kelantan', 'Kelantan'],
        'year': [2000, 2001, 2002, 2001, 2002, 2001, 2002],
        'pop': [1.5, 1.7, 3.6, 2.4, 2.9, 1.2, 3.2]}
frame = pd.DataFrame(data)
frame
```

Out[124]:

| | state | year | pop |
|---|----------|------|-----|
| 0 | Jakarta | 2000 | 1.5 |
| 1 | Jakarta | 2001 | 1.7 |
| 2 | Jakarta | 2002 | 3.6 |
| 3 | Selangor | 2001 | 2.4 |
| 4 | Selangor | 2002 | 2.9 |
| 5 | Kelantan | 2001 | 1.2 |
| 6 | Kelantan | 2002 | 3.2 |

In [125]:

```
frame = pd.DataFrame(data, columns=['year', 'state', 'pop'])
frame
```

Out[125]:

| | year | state | pop |
|---|------|----------|-----|
| 0 | 2000 | Jakarta | 1.5 |
| 1 | 2001 | Jakarta | 1.7 |
| 2 | 2002 | Jakarta | 3.6 |
| 3 | 2001 | Selangor | 2.4 |
| 4 | 2002 | Selangor | 2.9 |
| 5 | 2001 | Kelantan | 1.2 |
| 6 | 2002 | Kelantan | 3.2 |

Adding new column

In [126]:

```
frame2 = pd.DataFrame(data, columns=['year', 'state', 'pop', 'debt'],
                      index = ['one', 'two', 'three', 'four', 'five', 'six', 'seven'])
frame2
```

Out[126]:

| | year | state | pop | debt |
|-------|------|----------|-----|------|
| one | 2000 | Jakarta | 1.5 | NaN |
| two | 2001 | Jakarta | 1.7 | NaN |
| three | 2002 | Jakarta | 3.6 | NaN |
| four | 2001 | Selangor | 2.4 | NaN |
| five | 2002 | Selangor | 2.9 | NaN |
| six | 2001 | Kelantan | 1.2 | NaN |
| seven | 2002 | Kelantan | 3.2 | NaN |

Selecting column

In [127]:

```
frame2.columns
```

Out[127]:

```
Index(['year', 'state', 'pop', 'debt'], dtype='object')
```

In [128]:

```
frame2['state']
```

Out[128]:

```
one      Jakarta
two      Jakarta
three    Jakarta
four     Selangor
five     Selangor
six      Kelantan
seven    Kelantan
Name: state, dtype: object
```

In [129]:

```
frame2.year
```

Out[129]:

```
one      2000
two      2001
three    2002
four     2001
five     2002
six      2001
seven    2002
Name: year, dtype: int64
```

In [130]:

```
frame2.loc['two']
```

Out[130]:

```
year      2001
state     Jakarta
pop        1.7
debt      NaN
Name: two, dtype: object
```

loc is location will list all elements under loc [two]. loc will call base on assignee name

In [131]:

```
frame2.loc['two', 'state']
```

Out[131]:

```
'Jakarta'
```

In [132]:

```
## iloc base on index

frame2.iloc[1,1]
```

Out[132]:

```
'Jakarta'
```

In [133]:

```
frame2.loc['two':, : 'state']
```

Out[133]:

| | year | state |
|--------------|------|----------|
| two | 2001 | Jakarta |
| three | 2002 | Jakarta |
| four | 2001 | Selangor |
| five | 2002 | Selangor |
| six | 2001 | Kelantan |
| seven | 2002 | Kelantan |

In [134]:

```
frame2
```

Out[134]:

| | year | state | pop | debt |
|--------------|------|----------|-----|------|
| one | 2000 | Jakarta | 1.5 | NaN |
| two | 2001 | Jakarta | 1.7 | NaN |
| three | 2002 | Jakarta | 3.6 | NaN |
| four | 2001 | Selangor | 2.4 | NaN |
| five | 2002 | Selangor | 2.9 | NaN |
| six | 2001 | Kelantan | 1.2 | NaN |
| seven | 2002 | Kelantan | 3.2 | NaN |

The debt value is NaN. We can assign value for 'debt'

In [135]:

```
frame2['debt'] = 16.5  
frame2
```

Out[135]:

| | year | state | pop | debt |
|--------------|------|----------|-----|------|
| one | 2000 | Jakarta | 1.5 | 16.5 |
| two | 2001 | Jakarta | 1.7 | 16.5 |
| three | 2002 | Jakarta | 3.6 | 16.5 |
| four | 2001 | Selangor | 2.4 | 16.5 |
| five | 2002 | Selangor | 2.9 | 16.5 |
| six | 2001 | Kelantan | 1.2 | 16.5 |
| seven | 2002 | Kelantan | 3.2 | 16.5 |

In [136]:

```
frame2['debt'] = np.arange(7.)  
frame2
```

Out[136]:

| | year | state | pop | debt |
|-------|------|----------|-----|------|
| one | 2000 | Jakarta | 1.5 | 0.0 |
| two | 2001 | Jakarta | 1.7 | 1.0 |
| three | 2002 | Jakarta | 3.6 | 2.0 |
| four | 2001 | Selangor | 2.4 | 3.0 |
| five | 2002 | Selangor | 2.9 | 4.0 |
| six | 2001 | Kelantan | 1.2 | 5.0 |
| seven | 2002 | Kelantan | 3.2 | 6.0 |

In [137]:

```
frame2['debt'] = [10,20,15,13,11,67,87]  
frame2
```

Out[137]:

| | year | state | pop | debt |
|-------|------|----------|-----|------|
| one | 2000 | Jakarta | 1.5 | 10 |
| two | 2001 | Jakarta | 1.7 | 20 |
| three | 2002 | Jakarta | 3.6 | 15 |
| four | 2001 | Selangor | 2.4 | 13 |
| five | 2002 | Selangor | 2.9 | 11 |
| six | 2001 | Kelantan | 1.2 | 67 |
| seven | 2002 | Kelantan | 3.2 | 87 |

In [138]:

```
'Jakarta' in frame2.columns
```

Out[138]:

False

Use Case Exercise

In [139]:

```
import numpy as np  
import pandas as pd
```

DATA 1

In [140]:

```
df = pd.read_csv('data2.csv')
```

Checking Top 10 and bottom 10 data

In [141]:

```
df.head()
```

Out[141]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|-----------|------------|---------|----------|------------|------------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 |

In [142]:

```
df.tail()
```

Out[142]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|----|------------|------------|---------|---------|------------|------------|
| 15 | Argentina | 44.94 | 2780.40 | 637.49 | S.America | 1816-07-09 |
| 16 | Algeria | 43.38 | 2381.74 | 167.56 | Africa | 5/7/1962 |
| 17 | Canada | 37.59 | 9984.67 | 1647.12 | N.America | 1867-07-01 |
| 18 | Australia | 25.47 | 7692.02 | 1408.68 | Oceania | NaN |
| 19 | Kazakhstan | 18.53 | 2724.90 | 159.41 | Asia | 16/12/1991 |

In [143]:

```
df.dtypes
```

Out[143]:

```
COUNTRY      object
POPULATION   float64
AREA          float64
GDP           float64
CONTINENTS    object
IND_DAY       object
dtype: object
```

In [144]:

```
df.shape
```

Out[144]:

```
(20, 6)
```

In [145]:

```
pd.isnull(df)
```

Out[145]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|---------|------------|-------|-------|------------|---------|
| 0 | False | False | False | False | False | True |
| 1 | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False |
| 3 | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False |
| 5 | False | False | False | False | False | False |
| 6 | False | False | False | False | False | False |
| 7 | False | False | False | False | False | False |
| 8 | False | False | False | False | True | False |
| 9 | False | False | False | False | False | False |

In [146]:

```
newdf1=df[['COUNTRY', 'POPULATION', 'CONTINENTS', 'IND_DAY']]  
newdf1
```

Out[146]:

| | COUNTRY | POPULATION | CONTINENTS | IND_DAY |
|---|------------|------------|------------|------------|
| 0 | China | 1398.72 | Asia | NaN |
| 1 | India | 1351.16 | Asia | 15/8/1947 |
| 2 | US | 329.74 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | Asia | 17/8/1945 |
| 4 | Brazil | 210.32 | S.America | 1822-09-07 |
| 5 | Pakistan | 205.71 | Asia | 14/8/1947 |
| 6 | Nigeria | 200.96 | Africa | 1/10/1960 |
| 7 | Bangladesh | 167.09 | Asia | 26/3/1971 |
| 8 | Russia | 146.79 | NaN | 12/6/1992 |
| 9 | Mexico | 126.58 | N.America | 1810-09-16 |

In [147]:

```
newdf1['CONTINENTS'].fillna('transcontinental')
```

Out[147]:

```
0          Asia
1          Asia
2      N.America
3          Asia
4      S.America
5          Asia
6          Africa
7          Asia
8  transcontinental
9      N.America
10         Asia
11         Europe
12         Europe
13         Europe
14         Europe
15      S.America
16         Africa
17      N.America
```

In [148]:

```
newdf1
#no changes in Rusia
```

Out[148]:

| | COUNTRY | POPULATION | CONTINENTS | IND_DAY |
|---|------------|------------|------------|------------|
| 0 | China | 1398.72 | Asia | NaN |
| 1 | India | 1351.16 | Asia | 15/8/1947 |
| 2 | US | 329.74 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | Asia | 17/8/1945 |
| 4 | Brazil | 210.32 | S.America | 1822-09-07 |
| 5 | Pakistan | 205.71 | Asia | 14/8/1947 |
| 6 | Nigeria | 200.96 | Africa | 1/10/1960 |
| 7 | Bangladesh | 167.09 | Asia | 26/3/1971 |
| 8 | Russia | 146.79 | NaN | 12/6/1992 |
| 9 | Mexico | 126.58 | N.America | 1810-09-16 |

In [149]:

```
newdf1['CONTINENTS'].fillna('Transcontinental', inplace=True)
newdf1
```

C:\Anaconda3\lib\site-packages\pandas\core\generic.py:6130: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy>)

```
self._update_inplace(new_data)
```

Out[149]:

| | COUNTRY | POPULATION | CONTINENTS | IND_DAY |
|---|-----------|------------|------------|------------|
| 0 | China | 1398.72 | Asia | NaN |
| 1 | India | 1351.16 | Asia | 15/8/1947 |
| 2 | US | 329.74 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | Asia | 17/8/1945 |
| 4 | ... | ... | ... | ... |

In [150]:

```
newdf1.dtypes
```

Out[150]:

```
COUNTRY      object
POPULATION    float64
CONTINENTS    object
IND_DAY       object
dtype: object
```

How to change date format

In [151]:

```
newdf1['IND_DAY']=pd.to_datetime(newdf1['IND_DAY'])
newdf1
```

C:\Anaconda3\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy>)

"""Entry point for launching an IPython kernel.

Out[151]:

| | COUNTRY | POPULATION | CONTINENTS | IND_DAY |
|---|-----------|------------|------------|------------|
| 0 | China | 1398.72 | Asia | NaT |
| 1 | India | 1351.16 | Asia | 1947-08-15 |
| 2 | US | 329.74 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | Asia | 1945-08-17 |

In [152]:

```
newdf1.dtypes
```

Out[152]:

```
COUNTRY          object
POPULATION       float64
CONTINENTS        object
IND_DAY          datetime64[ns]
dtype: object
```

How to fill in missing date

In [153]:

```
newdf1['IND_DAY'].fillna(pd.Timestamp("20210423"))
```

Out[153]:

```
0    2021-04-23
1    1947-08-15
2    1776-07-04
3    1945-08-17
4    1822-09-07
5    1947-08-14
6    1960-01-10
7    1971-03-26
8    1992-12-06
9    1810-09-16
10   2021-04-23
11   2021-04-23
12   1789-07-14
13   2021-04-23
14   2021-04-23
15   1816-07-09
16   1962-05-07
17   1867-07-01
```

In [154]:

```
newdf1['IND_DAY'].astype(str).replace({'NaT': "No date"})
```

Out[154]:

```
0      No date
1    1947-08-15
2    1776-07-04
3    1945-08-17
4    1822-09-07
5    1947-08-14
6    1960-01-10
7    1971-03-26
8    1992-12-06
9    1810-09-16
10     No date
11     No date
12   1789-07-14
13     No date
14     No date
15   1816-07-09
16   1962-05-07
17   1867-07-01
```

In [155]:

```
newdf1['IND_DAY'].fillna(value = 'No date')
```

Out[155]:

```
0          No date
1    1947-08-15 00:00:00
2    1776-07-04 00:00:00
3    1945-08-17 00:00:00
4    1822-09-07 00:00:00
5    1947-08-14 00:00:00
6    1960-01-10 00:00:00
7    1971-03-26 00:00:00
8    1992-12-06 00:00:00
9    1810-09-16 00:00:00
10         No date
11         No date
12    1789-07-14 00:00:00
13         No date
14         No date
15    1816-07-09 00:00:00
16    1962-05-07 00:00:00
17    1867-07-01 00:00:00
```

In [156]:

```
newdf1['IND_DAY'].fillna(value = 'No date', inplace = True)
newdf1
```

Out[156]:

| | COUNTRY | POPULATION | CONTINENTS | IND_DAY |
|---|------------|------------|------------------|---------------------|
| 0 | China | 1398.72 | Asia | No date |
| 1 | India | 1351.16 | Asia | 1947-08-15 00:00:00 |
| 2 | US | 329.74 | N.America | 1776-07-04 00:00:00 |
| 3 | Indonesia | 268.07 | Asia | 1945-08-17 00:00:00 |
| 4 | Brazil | 210.32 | S.America | 1822-09-07 00:00:00 |
| 5 | Pakistan | 205.71 | Asia | 1947-08-14 00:00:00 |
| 6 | Nigeria | 200.96 | Africa | 1960-01-10 00:00:00 |
| 7 | Bangladesh | 167.09 | Asia | 1971-03-26 00:00:00 |
| 8 | Russia | 146.79 | Transcontinental | 1992-12-06 00:00:00 |
| 9 | Mexico | 126.58 | N.America | 1810-09-16 00:00:00 |

References:

- [1] <https://stackoverflow.com/questions/42818262/pandas-dataframe-replace-nat-with-none>
(<https://stackoverflow.com/questions/42818262/pandas-dataframe-replace-nat-with-none>)
- [2] https://pandas.pydata.org/pandas-docs/stable/user_guide/timedeltas.html
(https://pandas.pydata.org/pandas-docs/stable/user_guide/timedeltas.html)
- [3] <https://stackoverflow.com/questions/32327314/how-to-rearrange-a-date-in-python>
(<https://stackoverflow.com/questions/32327314/how-to-rearrange-a-date-in-python>)

Add new data set call 'data3'

In [157]:

```
df1 = pd.read_csv('data3.csv')
df1
```

UnicodeDecodeError

Traceback (most recent call last)

pandas/_libs/parsers.pyx in pandas._libs.parsers.TextReader._convert_tokens()

pandas/_libs/parsers.pyx in pandas._libs.parsers.TextReader._convert_with_dtype()

pandas/_libs/parsers.pyx in pandas._libs.parsers.TextReader._string_convert()

pandas/_libs/parsers.pyx in pandas._libs.parsers._string_box_utf8()

UnicodeDecodeError: 'utf-8' codec can't decode byte 0xf4 in position 1: in valid continuation byte

During handling of the above exception, another exception occurred:

In [158]:

```
df1 = pd.read_csv('data3.csv',encoding='latin-1')
df1
```

Out[158]:

| | country | cases | deaths | region |
|---|----------------|------------|---------|---------------|
| 0 | United States | 32,669,121 | 584,226 | North America |
| 1 | India | 16,257,309 | 186,928 | Asia |
| 2 | Brazil | 14,172,139 | 383,757 | South America |
| 3 | France | 5,408,606 | 102,164 | Europe |
| 4 | Russia | 4,736,121 | 107,103 | Europe |
| 5 | Turkey | 4,501,382 | 37,329 | Asia |
| 6 | United Kingdom | 4,398,431 | 127,345 | Europe |
| 7 | Italy | 3,920,945 | 118,357 | Europe |
| 8 | Spain | 3,456,886 | 77,496 | Europe |
| 9 | Germany | 3,238,054 | 81,693 | Europe |

In [159]:

```
df1.dtypes
```

Out[159]:

```
country    object
cases      object
deaths     object
region     object
dtype: object
```

In [160]:

```
df1.shape
```

Out[160]:

```
(220, 4)
```

Change Data Type

change cases and death to float

In [161]:

```
df1['deaths'] = df1['deaths'].str.replace(',', '')
df1['deaths'] = df1.deaths.astype(float)
df1['cases'] = df1['cases'].str.replace(',', '')
df1['cases'] = df1.cases.astype(float)
```

In [162]:

```
df1.dtypes
```

Out[162]:

```
country    object
cases      float64
deaths     float64
region     object
dtype: object
```

In [163]:

```
df.head(2)
```

Out[163]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|---------|------------|---------|----------|------------|-----------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 |

In [164]:

```
df1.head(2)
```

Out[164]:

| | country | cases | deaths | region |
|---|---------------|------------|----------|---------------|
| 0 | United States | 32669121.0 | 584226.0 | North America |
| 1 | India | 16257309.0 | 186928.0 | Asia |

Setting header to data case

Different way writing the header. Let's change it.

In [165]:

```
# .capitalize to change first letter as capital letter.  
df1.columns=df1.columns.str.capitalize()  
df1.head(2)
```

Out[165]:

| | Country | Cases | Deaths | Region |
|---|---------------|------------|----------|---------------|
| 0 | United States | 32669121.0 | 584226.0 | North America |
| 1 | India | 16257309.0 | 186928.0 | Asia |

In [166]:

```
df1.columns=df1.columns.str.upper()  
  
#.upper() to change header to uppercase  
  
df1.head(2)
```

Out[166]:

| | COUNTRY | CASES | DEATHS | REGION |
|---|---------------|------------|----------|---------------|
| 0 | United States | 32669121.0 | 584226.0 | North America |
| 1 | India | 16257309.0 | 186928.0 | Asia |

In [167]:

```
df1.shape
```

Out[167]:

(220, 4)

In [168]:

```
df.shape
```

Out[168]:

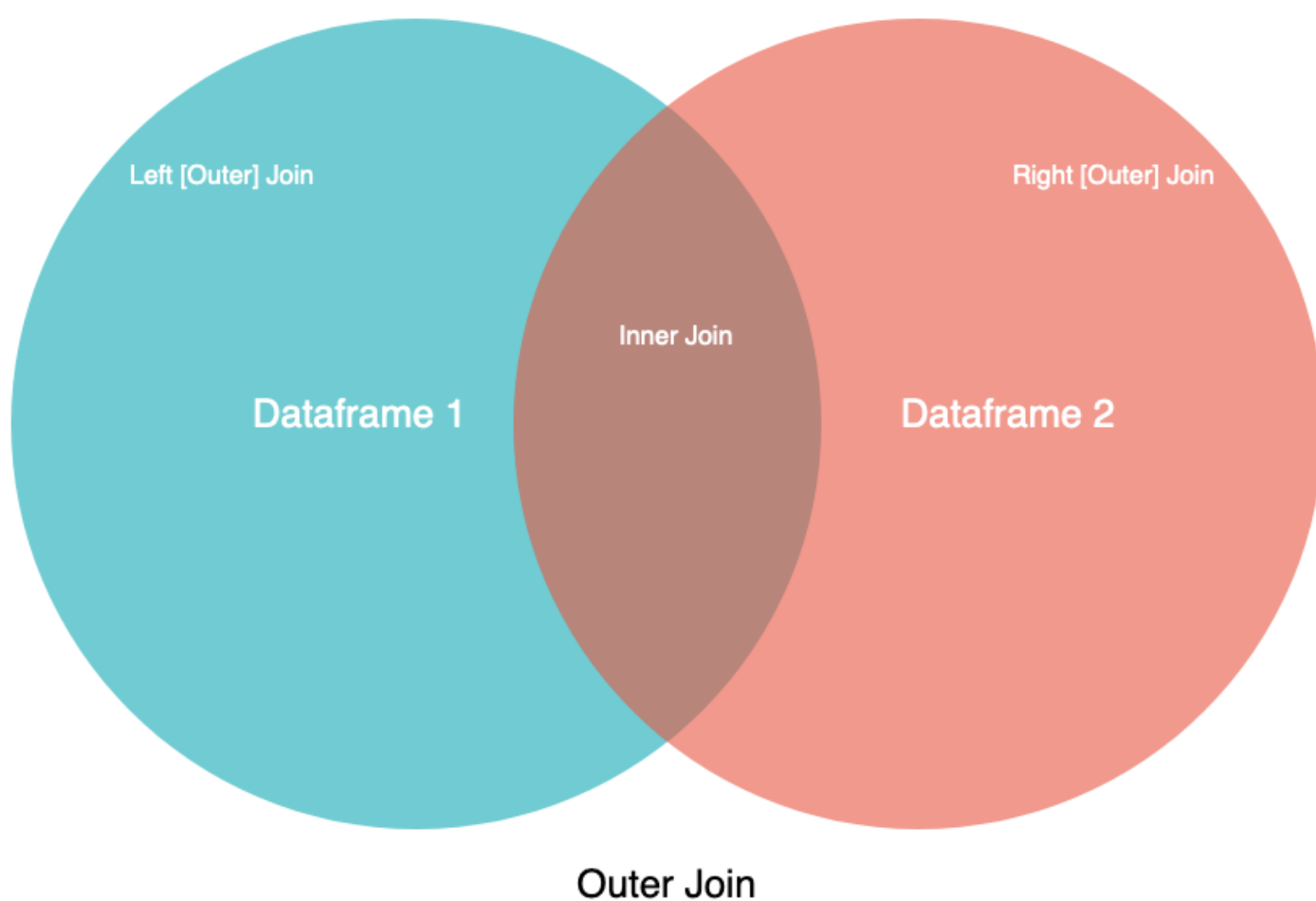
(20, 6)

Merge

Pandas provides a single function, `merge()`, as the entry point for all standard database join operations between `DataFrame` or named `Series` objects.

MERGE combining data on common columns or indices.

You can achieve both many-to-one and many-to-many joins with `merge()`



When gluing together multiple `DataFrames`, you have a choice of how to handle the other axes (other than the one being concatenated). This can be done in the following two ways

Take the union of them all, `join='outer'`. This is the default option as it results in zero information loss.

Take the intersection, `join='inner'`.

In [169]:

```
merge_df=pd.merge(df, df1)
merge_df
```

Out[169]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | R |
|---|------------|------------|----------|----------|------------|------------|------------|----------|---------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 3 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 | South A |
| 4 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 5 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 6 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 7 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 | 4736121.0 | 107103.0 | |
| 8 | Mexico | 126.58 | 1964.38 | 1158.23 | N.America | 1810-09-12 | 2319519.0 | 214095.0 | North A |

By default, how = inner, which will merge only match data.

In [170]:

```
merge_df.shape
```

Out[170]:

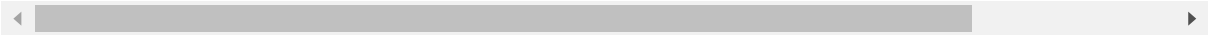
(18, 9)

In [171]:

```
merge_df.CONTINENTS=merge_df.CONTINENTS.replace(['N.America','S.America'],['North America'],
merge_df
```

Out[171]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEAT |
|----|------------|------------|----------|----------|---------------|------------|------------|-------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 463 |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 18692 |
| 2 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 4417 |
| 3 | Brazil | 210.32 | 8515.77 | 2055.51 | South America | 1822-09-07 | 14172139.0 | 38375 |
| 4 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 1684 |
| 5 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 206 |
| 6 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 1078 |
| 7 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 | 4736121.0 | 10710 |
| 8 | Mexico | 126.58 | 1964.38 | 1158.23 | North America | 1810-09-16 | 2319519.0 | 21409 |
| 9 | Japan | 126.22 | 377.97 | 4872.42 | Asia | NaN | 547137.0 | 977 |
| 10 | Germany | 83.02 | 357.11 | 3693.20 | Europe | NaN | 3238054.0 | 8169 |
| 11 | France | 67.02 | 640.68 | 2582.49 | Europe | 1789-07-14 | 5408606.0 | 10216 |
| 12 | Italy | 60.36 | 301.34 | 1943.84 | Europe | NaN | 3920945.0 | 11835 |
| 13 | Argentina | 44.94 | 2780.40 | 637.49 | South America | 1816-07-09 | 2796768.0 | 6062 |
| 14 | Algeria | 43.38 | 2381.74 | 167.56 | Africa | 5/7/1962 | 120363.0 | 318 |
| 15 | Canada | 37.59 | 9984.67 | 1647.12 | North America | 1867-07-01 | 1155834.0 | 2382 |
| 16 | Australia | 25.47 | 7692.02 | 1408.68 | Oceania | NaN | 29626.0 | 91 |
| 17 | Kazakhstan | 18.53 | 2724.90 | 159.41 | Asia | 16/12/1991 | 300733.0 | 351 |



In [172]:

```
test1merge_df=pd.merge(df, df1, how='inner')
test1merge_df
```

Out[172]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | R |
|---|------------|------------|----------|----------|------------|------------|------------|----------|---------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 3 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 | South A |
| 4 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 5 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 6 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 7 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 | 4736121.0 | 107103.0 | |
| 8 | Mexico | 126.58 | 1964.38 | 1158.23 | N.America | 1810-09-22 | 2319519.0 | 214095.0 | North A |

In [173]:

```
test1merge_df.shape
```

Out[173]:

(18, 9)

In [174]:

```
merge_df=pd.merge(df, df1, how='outer')
merge_df
# will merge all data
```

Out[174]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | |
|---|------------|------------|----------|----------|------------|------------|------------|----------|-------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 | NaN | NaN | |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 | South |
| 5 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 6 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 7 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 8 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 | 4736121.0 | 107103.0 | |

In [175]:

```
merge_df.shape
```

Out[175]:

(222, 9)

In [176]:

```
# Let's try how='left' or 'right'

test1=pd.merge(df, df1, how='left')
test1
```

Out[176]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | R |
|---|------------|------------|---------|----------|------------|------------|------------|----------|---------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 | NaN | NaN | |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 | South A |
| 5 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 6 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 7 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 8 | ... | ... | ... | ... | ... | ... | ... | ... | ... |

how='left', will merge base on left file, in this example is df

In [177]:

```
test2=pd.merge(df, df1, how='right')
test2
```

Out[177]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | |
|---|------------|------------|----------|----------|------------|------------|------------|----------|-------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 3 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 | South |
| 4 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 5 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 6 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 7 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 | 4736121.0 | 107103.0 | |
| 8 | Mexico | 126.58 | 1964.38 | 1158.23 | N.America | 1810-09-22 | 2319519.0 | 214095.0 | North |

how='right', will merge base on right file, in this example is df1

Concatenating

With concatenation, your datasets are just stitched together along an axis — either the row axis or column axis.

In [178]:

```
concat_df=pd.concat([df, df1], axis=1)
concat_df
```

Out[178]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | COUNTRY | CASES | DEATHS |
|---|-----------|------------|---------|----------|------------|------------|----------------|------------|----------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | United States | 32669121.0 | 584220.0 |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | India | 16257309.0 | 186928.0 |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 | Brazil | 14172139.0 | 383757.0 |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | France | 5408606.0 | 102160.0 |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | Russia | 4736121.0 | 107103.0 |
| 5 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | Turkey | 4501382.0 | 37320.0 |
| 6 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | United Kingdom | 4398431.0 | 127340.0 |

Let's try append the data

Let's call add new dataset call data4

In [179]:

```
df2=pd.read_csv('data4.csv')
df2
```

Out[179]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|---------|------------|--------|--------|------------|------------|
| 0 | Egypt | 93 | 640.68 | 375.77 | Asia | 1867-07-01 |
| 1 | Germany | 81 | 242.50 | 245.63 | Europe | 1789-07-14 |
| 2 | Iran | 80 | 301.34 | 143.00 | Europe | NaN |
| 3 | Turkey | 79 | NaN | 250.00 | NaN | NaN |

In [180]:

```
test3 = df.append(df2, ignore_index=True, sort=False)
test3
```

Out[180]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|------------|------------|----------|----------|------------|------------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 |
| 5 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 |
| 6 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 |
| 7 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 |
| 8 | Russia | 146.79 | 17098.25 | 1530.75 | NaN | 12/6/1992 |
| 9 | Mexico | 126.58 | 1964.38 | 1158.23 | N.America | 1810-09-16 |

data from df 1 and data 4 are combine at row level

LET'S MOVE TO GROUPBY

In [181]:

```
df.head(2)
```

Out[181]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|---------|------------|---------|----------|------------|-----------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 |

In [182]:

```
df1.head(2)
```

Out[182]:

| | COUNTRY | CASES | DEATHS | REGION |
|---|---------------|------------|----------|---------------|
| 0 | United States | 32669121.0 | 584226.0 | North America |
| 1 | India | 16257309.0 | 186928.0 | Asia |

In [183]:

```
df2.head()
```

Out[183]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY |
|---|---------|------------|--------|--------|------------|------------|
| 0 | Egypt | 93 | 640.68 | 375.77 | Asia | 1867-07-01 |
| 1 | Germany | 81 | 242.50 | 245.63 | Europe | 1789-07-14 |
| 2 | Iran | 80 | 301.34 | 143.00 | Europe | NaN |
| 3 | Turkey | 79 | NaN | 250.00 | NaN | NaN |

In [184]:

```
merge_df.head()
```

Out[184]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS |
|---|-----------|------------|---------|----------|------------|------------|------------|----------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 | NaN | NaN |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 | 14172139.0 | 383757.0 |

In [185]:

```
merge_df.shape
```

Out[185]:

(222, 9)

In [186]:

```
merge_df.dtypes
```

Out[186]:

```
COUNTRY      object
POPULATION   float64
AREA         float64
GDP          float64
CONTINENTS   object
IND_DAY      object
CASES        float64
DEATHS       float64
REGION       object
dtype: object
```

CONTINENT and REGION actually referring to the same thing. Let's try to fix it

Checking the elements

In [187]:

```
reg=merge_df.groupby('REGION').sum()
reg
```

Out[187]:

| | POPULATION | AREA | GDP | CASES | DEATHS |
|-------------------|------------|----------|----------|------------|----------|
| REGION | | | | | |
| Africa | 244.34 | 3305.51 | 543.33 | 4513248.0 | 119538.0 |
| Asia | 3535.50 | 18927.50 | 21405.59 | 35616438.0 | 482532.0 |
| Australia/Oceania | 25.47 | 7692.02 | 1408.68 | 61971.0 | 1184.0 |
| Europe | 357.19 | 18397.38 | 9750.28 | 43483441.0 | 989869.0 |
| North America | 164.17 | 11949.05 | 2805.35 | 37760260.0 | 852502.0 |
| South America | 255.26 | 11296.17 | 2693.00 | 23897427.0 | 639607.0 |

In [188]:

```
con=merge_df.groupby('CONTINENTS').sum()  
con
```

Out[188]:

| | POPULATION | AREA | GDP | CASES | DEATHS |
|------------|------------|----------|----------|------------|----------|
| CONTINENTS | | | | | |
| Africa | 244.34 | 3305.51 | 543.33 | 284951.0 | 5242.0 |
| Asia | 3535.50 | 18927.50 | 21405.59 | 20342739.0 | 276648.0 |
| Europe | 276.84 | 1541.63 | 10850.76 | 12567605.0 | 302214.0 |
| N.America | 493.91 | 21782.57 | 22290.74 | 3475353.0 | 237917.0 |
| Oceania | 25.47 | 7692.02 | 1408.68 | 29626.0 | 910.0 |
| S.America | 255.26 | 11296.17 | 2693.00 | 16968907.0 | 444377.0 |

In [189]:

```
merge_df.CONTINENTS=merge_df.CONTINENTS.replace(['N.America','S.America'], ['North America',  
merge_df
```

Out[189]:

| | COUNTRY | POPULATION | AREA | GDP | CONTINENTS | IND_DAY | CASES | DEATHS | |
|---|------------|------------|---------|----------|---------------|------------|------------|----------|-------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 | |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 | |
| 2 | US | 329.74 | 9833.52 | 19485.39 | North America | 1776-07-04 | NaN | NaN | |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 | |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | South America | 1822-09-07 | 14172139.0 | 383757.0 | South |
| 5 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 | |
| 6 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 | |
| 7 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 | |
| 8 | Denmark | 148.78 | 4300.05 | 1500.75 | NaN | 10/6/1900 | 1700101.0 | 107100.0 | |

In [190]:

```
reg=merge_df.groupby('REGION').sum()  
reg
```

Out[190]:

| | POPULATION | AREA | GDP | CASES | DEATHS |
|-------------------|------------|----------|----------|------------|----------|
| REGION | | | | | |
| Africa | 244.34 | 3305.51 | 543.33 | 4513248.0 | 119538.0 |
| Asia | 3535.50 | 18927.50 | 21405.59 | 35616438.0 | 482532.0 |
| Australia/Oceania | 25.47 | 7692.02 | 1408.68 | 61971.0 | 1184.0 |
| Europe | 357.19 | 18397.38 | 9750.28 | 43483441.0 | 989869.0 |
| North America | 164.17 | 11949.05 | 2805.35 | 37760260.0 | 852502.0 |
| South America | 255.26 | 11296.17 | 2693.00 | 23897427.0 | 639607.0 |

In [191]:

```
con=merge_df.groupby('CONTINENTS').sum()  
con
```

Out[191]:

| | POPULATION | AREA | GDP | CASES | DEATHS |
|---------------|------------|----------|----------|------------|----------|
| CONTINENTS | | | | | |
| Africa | 244.34 | 3305.51 | 543.33 | 284951.0 | 5242.0 |
| Asia | 3535.50 | 18927.50 | 21405.59 | 20342739.0 | 276648.0 |
| Europe | 276.84 | 1541.63 | 10850.76 | 12567605.0 | 302214.0 |
| North America | 493.91 | 21782.57 | 22290.74 | 3475353.0 | 237917.0 |
| Oceania | 25.47 | 7692.02 | 1408.68 | 29626.0 | 910.0 |
| South America | 255.26 | 11296.17 | 2693.00 | 16968907.0 | 444377.0 |

In [192]:

```
df = df.rename(columns={'CONTINENTS': 'REGION'})  
df.head()
```

Out[192]:

| | COUNTRY | POPULATION | AREA | GDP | REGION | IND_DAY |
|---|-----------|------------|---------|----------|-----------|------------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 |
| 2 | US | 329.74 | 9833.52 | 19485.39 | N.America | 1776-07-04 |
| 3 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 |
| 4 | Brazil | 210.32 | 8515.77 | 2055.51 | S.America | 1822-09-07 |

In [193]:

```
df1.head()
```

Out[193]:

| | COUNTRY | CASES | DEATHS | REGION |
|---|---------------|------------|----------|---------------|
| 0 | United States | 32669121.0 | 584226.0 | North America |
| 1 | India | 16257309.0 | 186928.0 | Asia |
| 2 | Brazil | 14172139.0 | 383757.0 | South America |
| 3 | France | 5408606.0 | 102164.0 | Europe |
| 4 | Russia | 4736121.0 | 107103.0 | Europe |

In [194]:

```
df101=pd.merge(df, df1)
df101
```

Out[194]:

| | COUNTRY | POPULATION | AREA | GDP | REGION | IND_DAY | CASES | DEATHS |
|----|------------|------------|---------|----------|--------|------------|------------|----------|
| 0 | China | 1398.72 | 9596.96 | 12234.78 | Asia | NaN | 90566.0 | 4636.0 |
| 1 | India | 1351.16 | 3287.26 | 2575.67 | Asia | 15/8/1947 | 16257309.0 | 186928.0 |
| 2 | Indonesia | 268.07 | 1910.93 | 1015.54 | Asia | 17/8/1945 | 1626812.0 | 44172.0 |
| 3 | Pakistan | 205.71 | 881.91 | 302.14 | Asia | 14/8/1947 | 784108.0 | 16842.0 |
| 4 | Nigeria | 200.96 | 923.77 | 375.77 | Africa | 1/10/1960 | 164588.0 | 2061.0 |
| 5 | Bangladesh | 167.09 | 147.57 | 245.63 | Asia | 26/3/1971 | 736074.0 | 10781.0 |
| 6 | Japan | 126.22 | 377.97 | 4872.42 | Asia | NaN | 547137.0 | 9777.0 |
| 7 | Germany | 83.02 | 357.11 | 3693.20 | Europe | NaN | 3238054.0 | 81693.0 |
| 8 | France | 67.02 | 640.68 | 2582.49 | Europe | 1789-07-14 | 5408606.0 | 102164.0 |
| 9 | Italy | 60.36 | 301.34 | 1943.84 | Europe | NaN | 3920945.0 | 118357.0 |
| 10 | Algeria | 43.38 | 2381.74 | 167.56 | Africa | 5/7/1962 | 120363.0 | 3181.0 |
| 11 | Kazakhstan | 18.53 | 2724.90 | 159.41 | Asia | 16/12/1991 | 300733.0 | 3512.0 |

In [195]:

```
reg=df101.groupby('REGION').sum()  
reg
```

Out[195]:

| | POPULATION | AREA | GDP | CASES | DEATHS |
|--------|------------|----------|----------|------------|----------|
| REGION | | | | | |
| Africa | 244.34 | 3305.51 | 543.33 | 284951.0 | 5242.0 |
| Asia | 3535.50 | 18927.50 | 21405.59 | 20342739.0 | 276648.0 |
| Europe | 210.40 | 1299.13 | 8219.53 | 12567605.0 | 302214.0 |

In []:

In []: