PANDAS PART - 1 BY VIRAT TIWARI

October 10, 2023

1 PANDAS - IT ALLOWS US TO ANALYZE THE DATA AND MANIPULATION OF THE DATA . IT ALSO HELP IN CLEANING THE MESSY DATA OR PRESENT THEM IN A STRUCTED WAY - VIRAT TIWARI

PANDAS DATA READING COMPLETE CONCEPT AND DATAFRAMES (df) OPERATIONS

```
[4]: import pandas as pd

[5]: pip install pandas

Requirement already satisfied: pandas in /opt/conda/lib/python3.10/site-packages (1.5.2)

Requirement already satisfied: python-dateutil>=2.8.1 in /opt/conda/lib/python3.10/site-packages (from pandas) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in /opt/conda/lib/python3.10/site-packages (from pandas) (2022.6)

Requirement already satisfied: numpy>=1.21.0 in /opt/conda/lib/python3.10/site-packages (from pandas) (1.23.5)

Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.1->pandas) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

IMPORTANT - This is how we read the CSV file by using the "pd.read_csv() "function

[44]: pd.read_csv("services.csv")
```

```
[44]:
              location_id program_id
                                                     accepted_payments
      0
            1
                           1
                                       NaN
                                                                     NaN
      1
            2
                           2
                                       NaN
                                                                     NaN
      2
            3
                           3
                                       NaN
                                                                     NaN
            4
                           4
      3
                                       NaN
                                                                     NaN
      4
            5
                           5
                                       NaN
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      5
            6
                           6
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            7
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                           7
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      7
            8
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            9
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                                       NaN
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           10
                          10
                                       NaN
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```

10	11 11	NaN	NaN	
11	12 12	NaN	NaN	
12	13 13	NaN	NaN	
13	14 14	NaN	NaN	
14	15 15	NaN	NaN	
15	16 16	NaN	NaN	
16	17 17	NaN	NaN	
17	18 18	${\tt NaN}$	NaN	
18	19 19	NaN	NaN	
19	20 20	${\tt NaN}$	NaN	
20	21 21	NaN	NaN	
21	22 22	NaN	Cash, Check, Credit Card	
22	23 22	NaN	NaN	
	alternate_name		application_process	\
0	NaN		Walk in or apply by phone.	
1	NaN		Apply by phone for an appointment.	
2	NaN		Phone for information (403-4300 Ext. 4322).	
3	NaN		Apply by phone.	
4	NaN		Phone for information.	
5	NaN		in or apply by phone for membership appli	
6	NaN	Apply	by phone or be referred by a doctor, soc	
7	NaN		Apply by phone.	
8	NaN		in. Proof of residency in California requ	
9	NaN		in. Proof of California residency to rece	
10	NaN N- N		in. Proof of California residency require	
11 12	NaN NaN		in or apply by phone, email or webpage re	
13	NaN		in. Proof of California residency require	
14	NaN	Call	for appointment. Referral from human serv Walk in or through other agency referral.	
15	NaN	Walk	in. Written application, identification r	
16	NaN	Waik	Call for information.	
17	NaN	Call	for screening appointment. Medical visits	
18	NaN		all for screening appointment (650-347-3648).	
19	NaN		Walk in.	
20	NaN		By phone during business hours.	
21	Fotos para pasaportes		Walk in or apply by phone or mail	
22	NaN		Walk in or apply by phone or mail	
0	01.4 414 55		audience \	
0	Older adults age 55 or			
1	Residents of San Mateo County age 55 or over			
2 3	Older adults age 55 or over who can benefit fr Parents, children, families with problems of c			
			-	
4 5	Low-income working fam	TTTES		
5 6	Older adults who have	momove	Any age	
U	order addres with Have	шешот й	or pensory ross,	

7	Senior citizens age 60 or over, disabled indiv	
8	Ethnic minorities, especially Spanish speaking	
9	NaN	
10	NaN	
11	Adults, parents, children in 1st-12th grades i	
12	NaN	
13	Individuals or families with low or no income	
14	Adult alcoholic/drug addictive men and women w	
15	NaN	
16	NaN	
17	NaN	
18	NaN	
19	NaN	
20	NaN	
21	Profit and nonprofit businesses, the public, m	
22	Second service and nonprofit businesses, the p	
	, ,	
	description	\
0	A walk-in center for older adults that provide	
1	Provides training and job placement to eligibl	
2	Offers supportive counseling services to San M	
3	Provides supervised visitation services and a	
4	Provides fixed 8% short term loans to eligible	
5	A multipurpose center offering a wide variety	
6	Rosener House is a day center for older adults	
7	Delivers a hot meal to the home of persons age	
8	Provides general reading material, including b	
9	Provides general reading and media materials,	
10	Provides general reading materials, including	
11	Offers an intergenerational literacy program f	
12	Provides general reading materials, including	
13	Provides food, clothing, bus tokens and shelte	
14	Provides a long-term (6-12 month) residential	
15	Provides emergency assistance including food a	
16	Provides emergency food, clothing and furnitur	
17	By appointment only, Project Smile provides a	
18	Provides free medical and dental care to those	
19	no unrequired fields for this service	
20	just a test service	
21	[NOTE THIS IS NOT A REAL SERVICETHIS IS FOR	
22	[NOTE THIS IS NOT A REAL ORGANIZATIONTHIS IS	
_	_	
	eligibility	email \
0	Age 55 or over for most programs, age 60 or ov	NaN
1	Age 55 or over, county resident and willing an	NaN
2	Resident of San Mateo County age 55 or over	NaN
3	None	NaN

```
4
    Eligibility: Low-income family with legal cust...
                                                                            NaN
5
                                                                              NaN
                                                    None
6
                                         Age 18 or over
                                                                              NaN
7
               Homebound person unable to cook or shop
                                                                              NaN
8
      Resident of California to obtain a library card
                                                                              NaN
              Resident of California to obtain a card
9
                                                                              NaN
10
   Resident of California to obtain a library car...
                                                                           NaN
11
    English-speaking adult reading at or below 7th...
                                                                            NaN
12
      Resident of California to obtain a library card
                                                                              NaN
13
    None for most services. For emergency assistan...
                                                                            NaN
    Age 21-60, detoxed, physically able and willin...
                                                                            NaN
15
                         None for emergency assistance
                                                                              NaN
16
                                    Low-income families
                                                                              NaN
      Low-income person without access to health care
17
                                                                              NaN
18
      Low-income person without access to health care
                                                                              NaN
19
                                                     NaN
                                                                              NaN
20
                                                     NaN
                                                                              NaN
21
                                                    None
                                                          passports@example.org
22
                                                    None
                                                                              NaN
                                   interpretation_services
                                                        NaN
0
1
                                                        NaN
2
                                                        NaN
3
                                                        NaN
4
                                                        NaN
5
                                                        NaN
6
                                                        NaN
7
                                                        NaN
                                                        NaN
8
9
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10
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11
12
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14
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15
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16
                                                        NaN
17
                                                        NaN
                                                        NaN
18
19
                                                        NaN
20
                                                        NaN
21
       We offer 3-way interpretation services over th...
22
                                                        NaN
                                                keywords languages \
```

 ${\tt NaN}$

ADULT PROTECTION AND CARE SERVICES, Meal Sites...

0

```
1
    EMPLOYMENT/TRAINING SERVICES, Job Development, ...
                                                             NaN
2
    Geriatric Counseling, Older Adults, Gay, Lesbi...
                                                             NaN
3
    INDIVIDUAL AND FAMILY DEVELOPMENT SERVICES, Gr ...
                                                             NaN
       COMMUNITY SERVICES, Speakers, Automobile Loans
4
                                                               NaN
5
    ADULT PROTECTION AND CARE SERVICES, In-Home Su...
                                                             NaN
    ADULT PROTECTION AND CARE SERVICES, Adult Day ...
6
                                                             NaN
7
    ADULT PROTECTION AND CARE SERVICES, Meal Sites...
                                                             NaN
    EDUCATION SERVICES, Library, Libraries, Public...
8
                                                             NaN
    EDUCATION SERVICES, Library, Libraries, Public...
9
                                                             NaN
10
   EDUCATION SERVICES, Library, Libraries, Public...
                                                             NaN
    EDUCATION SERVICES, Adult, Alternative, Litera...
11
                                                             NaN
   EDUCATION SERVICES, Library, Libraries, Public...
                                                             NaN
13
    COMMUNITY SERVICES, Interpretation/Translation...
                                                             NaN
14
   ALCOHOLISM SERVICES, Residential Care, DRUG AB...
                                                             NaN
    COMMODITY SERVICES, Clothing/Personal Items, C...
15
                                                             NaN
    COMMODITY SERVICES, Clothing/Personal Items, C...
16
                                                             NaN
17
    HEALTH SERVICES, Outpatient Care, Community Cl...
                                                             NaN
    HEALTH SERVICES, Outpatient Care, Community Cl...
18
                                                             NaN
19
                                                               NaN
                                                    NaN
20
                                                    NaN
                                                               NaN
21
                                       Salud, Medicina
                                                           Spanish
22
                                                               NaN
           Ruby on Rails/Postgres/Redis, testing, wic
                                     name \
0
         Fair Oaks Adult Activity Center
1
        Second Career Employment Program
2
                  Senior Peer Counseling
3
                Family Visitation Center
4
       Economic Self-Sufficiency Program
5
    Little House Recreational Activities
6
        Rosener House Adult Day Services
7
          Meals on Wheels - South County
8
                         Fair Oaks Branch
9
                             Main Library
10
                          Schaberg Branch
11
                             Project Read
12
                    Redwood Shores Branch
13
                       Redwood City Corps
14
             Adult Rehabilitation Center
15
                          Sunnyvale Corps
16
       South San Francisco Citadel Corps
17
                            Project Smile
18
           San Mateo Free Medical Clinic
19
               Service with blank fields
20
         Service for Admin Test Location
21
                          Passport Photos
22
                     Example Service Name
```

```
required_documents
0
                                            NaN
1
                                            NaN
2
                                            NaN
3
                                            NaN
4
                                            NaN
5
                                            NaN
6
                                            NaN
7
                                            NaN
8
                                            NaN
9
                                            NaN
10
                                            NaN
11
                                            NaN
12
                                            NaN
13
                                            NaN
14
                                            NaN
15
                                            NaN
16
                                            NaN
17
                                            NaN
18
                                            NaN
19
                                            {\tt NaN}
20
                                            NaN
21
    Government-issued picture identification
22
                                            NaN
                                                 status
                              service_areas
0
                                       Colma
                                                 active
1
                           San Mateo County
                                                 active
2
                           San Mateo County
                                                 active
3
                           San Mateo County
                                                 active
4
                           San Mateo County
                                                 active
5
                           San Mateo County
                                                 active
6
      Belmont, Burlingame, East Palo Alto
                                                 active
7
                   Belmont, East Palo Alto
                                                 active
8
                           San Mateo County
                                                 active
9
                           San Mateo County
                                                 active
10
                           San Mateo County
                                                 active
11
                                   Daly City
                                                 active
12
                           San Mateo County
                                                 active
13
      Belmont, Burlingame, East Palo Alto
                                                 active
14
          Alameda County, San Mateo County
                                                 active
                                                 active
15
16
    Colma, Daly City, South San Francisco
                                                 active
17
                             East Palo Alto
                                                 active
18
                        Belmont, Burlingame
                                                 active
19
                                         NaN
                                                defunct
```

20	San Mateo County inactive	
21	Alameda County, San Mateo County active	
22	San Mateo County, Alameda County active	
	wait_time	website \
0	No wait.	NaN
1	Varies.	NaN
2	Varies.	NaN
3	No wait.	NaN
4	NaN	NaN
5	No wait.	NaN
6	No wait.	NaN
7	No wait.	NaN
8	No wait.	NaN
9	No wait.	NaN
10	No wait.	NaN
11	Depends on availability of tutors for small gr	NaN
12	No wait.	NaN
13	Up to 20 minutes.	NaN
14	Varies according to available beds for men and	NaN
15	No wait.	NaN
16	NaN	NaN
17	Varies.	NaN
18	Varies.	NaN
19	NaN	NaN
	NaN NaN	NaN NaN
19 20 21	NaN	NaN
20	NaN No wait to 2 weeks.	NaN http://www.example.com
20 21	NaN	NaN
20 21	NaN No wait to 2 weeks.	NaN http://www.example.com
20 21	NaN No wait to 2 weeks. No wait to 2 weeks	NaN http://www.example.com
20 21 22	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids	NaN http://www.example.com
20 21 22 0	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN	NaN http://www.example.com
20 21 22 0 1	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN	NaN http://www.example.com
20 21 22 0 1 2	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN	NaN http://www.example.com
20 21 22 0 1 2 3	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN	NaN http://www.example.com
20 21 22 0 1 2 3 4	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN	NaN http://www.example.com
20 21 22 0 1 2 3 4 5	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN NaN NaN	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10	NaN No wait to 2 weeks. No wait to 2 weeks taxonomy_ids NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10 11	Nan No wait to 2 weeks. No wait to 2 weeks taxonomy_ids Nan	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10 11 12	Nan No wait to 2 weeks No wait to 2 weeks taxonomy_ids Nan	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10 11 12 13	Nan No wait to 2 weeks. No wait to 2 weeks taxonomy_ids Nan	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14	Nan No wait to 2 weeks No wait to 2 weeks taxonomy_ids Nan	NaN http://www.example.com
20 21 22 0 1 2 3 4 5 6 7 8 9 10 11 12 13	Nan No wait to 2 weeks. No wait to 2 weeks taxonomy_ids Nan	NaN http://www.example.com

```
18
                                                NaN
     19
                                                NaN
     20
                                                NaN
         105, 108, 108-05, 108-05-01, 111, 111-05
     21
     22
                                                NaN
     [23 rows x 22 columns]
[]: | # By giving the " df " as a variable we store the file in df
     df=pd.read_csv("services.csv")
    Head ( ) - When we want to read first five or first four record of data than we use ".head ( ) "
    function and by default it gives first five record of data
[8]: df.head()
            location_id program_id accepted_payments alternate_name
[8]:
         1
                                 NaN
                                                     NaN
                                                                    NaN
     0
                       1
     1
         2
                       2
                                 NaN
                                                                    NaN
                                                     NaN
     2
         3
                       3
                                                                    NaN
                                 NaN
                                                     NaN
                       4
     3
         4
                                 NaN
                                                     NaN
                                                                    NaN
     4
         5
                       5
                                 NaN
                                                     NaN
                                                                    NaN
                                 application_process \
     0
                          Walk in or apply by phone.
     1
                 Apply by phone for an appointment.
       Phone for information (403-4300 Ext. 4322).
     3
                                      Apply by phone.
                              Phone for information.
     4
                                                   audience \
        Older adults age 55 or over, ethnic minorities...
     0
             Residents of San Mateo County age 55 or over
     1
     2 Older adults age 55 or over who can benefit fr...
     3 Parents, children, families with problems of c...
     4 Low-income working families with children tran...
                                                description \
     O A walk-in center for older adults that provide...
     1 Provides training and job placement to eligibl...
     2 Offers supportive counseling services to San M...
     3 Provides supervised visitation services and a ...
     4 Provides fixed 8% short term loans to eligible...
                                                eligibility email ... \
```

NaN

17

```
Age 55 or over, county resident and willing an...
                                                             NaN
              Resident of San Mateo County age 55 or over
                                                               NaN
                                                               {\tt NaN}
     3
       Eligibility: Low-income family with legal cust...
                                                             NaN
       interpretation_services
                                                                             keywords \
     0
                            NaN
                                 ADULT PROTECTION AND CARE SERVICES, Meal Sites...
                                 EMPLOYMENT/TRAINING SERVICES, Job Development,...
     1
                            NaN
     2
                                 Geriatric Counseling, Older Adults, Gay, Lesbi ...
                            NaN
                                 INDIVIDUAL AND FAMILY DEVELOPMENT SERVICES, Gr ...
     3
                            NaN
     4
                            NaN
                                     COMMUNITY SERVICES, Speakers, Automobile Loans
                                                 name required_documents
       languages
     0
             NaN
                     Fair Oaks Adult Activity Center
                                                                      NaN
     1
             NaN
                    Second Career Employment Program
                                                                      NaN
     2
             NaN
                              Senior Peer Counseling
                                                                      NaN
     3
             NaN
                            Family Visitation Center
                                                                      NaN
             {\tt NaN}
                  Economic Self-Sufficiency Program
                                                                      NaN
           service_areas status wait_time website taxonomy_ids
     0
                    Colma
                           active No wait.
                                                 NaN
                                                               NaN
     1 San Mateo County
                                                               NaN
                           active
                                    Varies.
                                                 NaN
     2 San Mateo County
                           active
                                    Varies.
                                                 NaN
                                                               NaN
     3 San Mateo County
                           active
                                   No wait.
                                                 {\tt NaN}
                                                               NaN
     4 San Mateo County
                           active
                                                 NaN
                                                               NaN
     [5 rows x 22 columns]
[9]: df.head(2)
            location_id program_id accepted_payments alternate_name
[9]:
     0
         1
                       1
                                 NaN
                                                    NaN
                                                                    NaN
         2
                       2
     1
                                 NaN
                                                                    NaN
                                                    NaN
                        application_process
     0
                Walk in or apply by phone.
        Apply by phone for an appointment.
        Older adults age 55 or over, ethnic minorities...
     1
             Residents of San Mateo County age 55 or over
                                                description \
       A walk-in center for older adults that provide...
     1 Provides training and job placement to eligibl...
```

NaN

Age 55 or over for most programs, age 60 or ov...

```
O Age 55 or over for most programs, age 60 or ov...
      1 Age 55 or over, county resident and willing an...
                                                               NaN
                                                                                keywords \
        interpretation_services
      0
                                   ADULT PROTECTION AND CARE SERVICES, Meal Sites...
                              NaN
                                   EMPLOYMENT/TRAINING SERVICES, Job Development,...
      1
                              NaN
                                                  name required documents
        languages
      0
                     Fair Oaks Adult Activity Center
               NaN
                    Second Career Employment Program
      1
               NaN
                                                                        NaN
             service_areas status wait_time website taxonomy_ids
      0
                     Colma active No wait.
                                                   NaN
                                                                 NaN
         San Mateo County active
                                      Varies.
                                                   NaN
                                                                 NaN
      [2 rows x 22 columns]
     Tail () - When we want to see the last record of data than we have to used ".tail ()" function
     it gives the last five record of data by default and we also manualy find the last record by giving a
     number inside the tail function
[10]: df.tail()
[10]:
                                                 accepted_payments
          id
               location_id program_id
      18
          19
                        19
                                    NaN
                                                                NaN
      19
          20
                        20
                                    NaN
                                                                NaN
      20
          21
                        21
                                    NaN
                                                                NaN
      21
          22
                        22
                                    NaN
                                          Cash, Check, Credit Card
                        22
      22
          23
                                    NaN
                                                                NaN
                                                                application_process
                  alternate_name
      18
                                   Call for screening appointment (650-347-3648).
                              {\tt NaN}
      19
                              NaN
      20
                                                   By phone during business hours.
                              NaN
                                                 Walk in or apply by phone or mail
      21
          Fotos para pasaportes
      22
                              NaN
                                                 Walk in or apply by phone or mail
                                                       audience
      18
                                                            NaN
      19
                                                            NaN
      20
          Profit and nonprofit businesses, the public, m...
      21
          Second service and nonprofit businesses, the p...
                                                   description \
         Provides free medical and dental care to those...
```

eligibility email ... \

```
19
                no unrequired fields for this service
20
                                    just a test service
21
    [NOTE THIS IS NOT A REAL SERVICE--THIS IS FOR ...
    [NOTE THIS IS NOT A REAL ORGANIZATION--THIS IS...
                                                                        email
                                          eligibility
    Low-income person without access to health care
                                                                           NaN
19
                                                                           NaN
20
                                                  NaN
                                                                           NaN
21
                                                 None
                                                        passports@example.org
22
                                                 None
                                                                           NaN
                                  interpretation_services
18
                                                        NaN
19
                                                        NaN
20
                                                        NaN
21
       We offer 3-way interpretation services over th...
22
                                                        NaN
                                               keywords languages \
    HEALTH SERVICES, Outpatient Care, Community Cl...
                                                             NaN
19
                                                               NaN
                                                    NaN
20
                                                    NaN
                                                               NaN
21
                                        Salud, Medicina
                                                           Spanish
22
           Ruby on Rails/Postgres/Redis, testing, wic
                                                               NaN
                                                              required_documents
18
      San Mateo Free Medical Clinic
                                                                              NaN
19
          Service with blank fields
                                                                              NaN
    Service for Admin Test Location
20
                                                                              NaN
                     Passport Photos
21
                                      Government-issued picture identification
22
               Example Service Name
                                                                              NaN
                        service_areas
                                          status
                                                             wait_time
18
                 Belmont, Burlingame
                                                               Varies.
                                          active
19
                                  NaN
                                         defunct
                                                                   NaN
20
                     San Mateo County
                                       inactive
                                                                   NaN
    Alameda County, San Mateo County
21
                                          active
                                                  No wait to 2 weeks.
    San Mateo County, Alameda County
                                                   No wait to 2 weeks
                                          active
                    website
                                                           taxonomy_ids
18
                        NaN
                                                                    NaN
19
                        NaN
                                                                    NaN
20
                        NaN
                                                                    NaN
                             105, 108, 108-05, 108-05-01, 111, 111-05
21
    http://www.example.com
    http://www.example.com
                                                                    NaN
```

[5 rows x 22 columns]

```
[11]: df.tail(3)
[11]:
              location_id program_id
                                               accepted_payments \
      20
                       21
          21
                                   NaN
      21
          22
                       22
                                   NaN
                                        Cash, Check, Credit Card
      22
          23
                       22
                                   NaN
                                                              NaN
                 alternate_name
                                                 application_process \
      20
                                    By phone during business hours.
                             NaN
      21
          Fotos para pasaportes
                                  Walk in or apply by phone or mail
                             NaN
                                  Walk in or apply by phone or mail
                                                     audience \
      20
                                                          NaN
         Profit and nonprofit businesses, the public, m...
      21
          Second service and nonprofit businesses, the p...
                                                  description eligibility \
      20
                                         just a test service
                                                                      NaN
      21
          [NOTE THIS IS NOT A REAL SERVICE--THIS IS FOR ...
                                                                   None
      22
          [NOTE THIS IS NOT A REAL ORGANIZATION--THIS IS...
                                                                   None
                           email
      20
                             NaN
      21
          passports@example.org
      22
                             NaN
                                     interpretation_services \
      20
                                                          NaN
      21
          We offer 3-way interpretation services over th...
      22
                                             keywords languages
      20
                                                   NaN
                                                             NaN
      21
                                      Salud, Medicina
                                                         Spanish
      22
          Ruby on Rails/Postgres/Redis, testing, wic
                                                             NaN
                                                                   required_documents
                                      name
      20
          Service for Admin Test Location
                           Passport Photos
                                            Government-issued picture identification
      21
                     Example Service Name
                                                                                   NaN
                              service_areas
                                               status
                                                                  wait_time \
      20
                           San Mateo County
                                            inactive
         Alameda County, San Mateo County
                                               active No wait to 2 weeks.
```

```
22 San Mateo County, Alameda County
                                            active No wait to 2 weeks
                         website
                                                               taxonomy_ids
      20
                             NaN
      21 http://www.example.com 105, 108, 108-05, 108-05-01, 111, 111-05
      22 http://www.example.com
                                                                        NaN
      [3 rows x 22 columns]
[12]: | # Df - It stands for the Data Frame that present in a tabular form
      type(df)
[12]: pandas.core.frame.DataFrame
[13]: # This is how we get all columns of data
      df.columns
[13]: Index(['id', 'location_id', 'program_id', 'accepted_payments',
             'alternate_name', 'application_process', 'audience', 'description',
             'eligibility', 'email', 'fees', 'funding_sources',
             'interpretation_services', 'keywords', 'languages', 'name',
             'required_documents', 'service_areas', 'status', 'wait_time', 'website',
             'taxonomy_ids'],
            dtype='object')
[14]: # This is how we get all the columns of data in the list by using simply list (
       →) function
      list(df.columns)
[14]: ['id',
       'location_id',
       'program_id',
       'accepted_payments',
       'alternate_name',
       'application_process',
       'audience',
       'description',
       'eligibility',
       'email',
       'fees',
       'funding_sources',
       'interpretation_services',
       'keywords',
       'languages',
```

```
'required_documents',
       'service_areas',
       'status',
       'wait_time',
       'website',
       'taxonomy_ids']
[15]: # This is how we exctrat the data of particular column only single colimn by
       →giving their name inside the df [" "] function
      df['service_areas']
[15]: 0
                                             Colma
      1
                                  San Mateo County
      2
                                  San Mateo County
      3
                                  San Mateo County
      4
                                  San Mateo County
      5
                                  San Mateo County
              Belmont, Burlingame, East Palo Alto
      6
      7
                           Belmont, East Palo Alto
      8
                                  San Mateo County
      9
                                  San Mateo County
      10
                                  San Mateo County
                                         Daly City
      11
      12
                                  San Mateo County
              Belmont, Burlingame, East Palo Alto
      13
      14
                 Alameda County, San Mateo County
      15
            Colma, Daly City, South San Francisco
      16
      17
                                    East Palo Alto
      18
                               Belmont, Burlingame
      19
                                                NaN
      20
                                  San Mateo County
      21
                 Alameda County, San Mateo County
      22
                 San Mateo County, Alameda County
      Name: service_areas, dtype: object
[18]: df['website']
[18]: 0
                                NaN
      1
                                NaN
      2
                                NaN
      3
                                NaN
      4
                                NaN
      5
                                NaN
      6
                                NaN
```

'name',

```
7
                                    {\tt NaN}
      8
                                    {\tt NaN}
       9
                                    NaN
       10
                                    {\tt NaN}
       11
                                    NaN
       12
                                    {\tt NaN}
       13
                                    NaN
       14
                                    {\tt NaN}
       15
                                    NaN
       16
                                    NaN
       17
                                    NaN
       18
                                    NaN
       19
                                    NaN
       20
                                    {\tt NaN}
       21
             http://www.example.com
             http://www.example.com
       22
      Name: website, dtype: object
[19]: # NOTE - In pandas series is equivalent to list
      type(df["website"])
[19]: pandas.core.series.Series
[20]: list(df["website"])
[20]: [nan,
       nan,
       nan,
```

```
'http://www.example.com']
[21]: # PANDAS NOTE - IN LIST IT DOES NOT SHOW THE INDEXING BUT IN SERIES IT SHOWS
       → THE INDEXING
[22]: # This is how we pass the column name iside a list
      df[["website"]]
[22]:
                          website
      0
                              NaN
                              NaN
      1
      2
                              NaN
      3
                              NaN
      4
                              NaN
      5
                              NaN
      6
                              NaN
      7
                              NaN
      8
                              NaN
      9
                              NaN
      10
                              NaN
      11
                              NaN
      12
                              NaN
      13
                              NaN
      14
                              NaN
                              NaN
      15
      16
                              NaN
      17
                              NaN
      18
                              NaN
      19
                              NaN
      20
                              NaN
          http://www.example.com
      21
         http://www.example.com
[26]: type(df[["website"]])
[26]: pandas.core.frame.DataFrame
[27]: df.columns
[27]: Index(['id', 'location_id', 'program_id', 'accepted_payments',
             'alternate_name', 'application_process', 'audience', 'description',
             'eligibility', 'email', 'fees', 'funding_sources',
             'interpretation_services', 'keywords', 'languages', 'name',
             'required_documents', 'service_areas', 'status', 'wait_time', 'website',
             'taxonomy_ids'],
```

'http://www.example.com',

dtype='object')

```
1
                         NaN
                                           2
2
                         NaN
                                           3
3
                                           4
                         NaN
4
                         NaN
                                           5
5
                                           6
                         NaN
6
                         NaN
                                           7
7
                         NaN
                                           8
8
                         NaN
                                           9
9
                         NaN
                                          10
10
                         NaN
                                          11
                                          12
11
                         NaN
12
                         NaN
                                          13
13
                         NaN
                                          14
14
                         NaN
                                          15
15
                         NaN
                                          16
16
                         NaN
                                          17
17
                         NaN
                                          18
18
                                          19
                         NaN
19
                         NaN
                                          20
20
                                          21
                         NaN
    passports@example.org
                                          22
21
22
                         NaN
                                          22
```

```
[29]: # Here we pass three columns

df[["email","keywords","location_id"]]
```

```
[29]:
                                                                                   keywords \
                             email
                                    ADULT PROTECTION AND CARE SERVICES, Meal Sites...
      0
                               {\tt NaN}
      1
                               {\tt NaN}
                                    EMPLOYMENT/TRAINING SERVICES, Job Development, ...
      2
                                    Geriatric Counseling, Older Adults, Gay, Lesbi ...
                               {\tt NaN}
      3
                                    INDIVIDUAL AND FAMILY DEVELOPMENT SERVICES, Gr ...
                               {\tt NaN}
      4
                                        COMMUNITY SERVICES, Speakers, Automobile Loans
                               NaN
      5
                                    ADULT PROTECTION AND CARE SERVICES, In-Home Su...
                               NaN
                                    ADULT PROTECTION AND CARE SERVICES, Adult Day ...
                               NaN
```

```
ADULT PROTECTION AND CARE SERVICES, Meal Sites...
8
                             EDUCATION SERVICES, Library, Libraries, Public...
                        NaN
                             EDUCATION SERVICES, Library, Libraries, Public...
9
                        NaN
10
                             EDUCATION SERVICES, Library, Libraries, Public...
                        {\tt NaN}
11
                             EDUCATION SERVICES, Adult, Alternative, Litera...
                        NaN
12
                             EDUCATION SERVICES, Library, Libraries, Public...
                        {\tt NaN}
                             COMMUNITY SERVICES, Interpretation/Translation...
13
                        NaN
14
                             ALCOHOLISM SERVICES, Residential Care, DRUG AB...
                        {\tt NaN}
                             COMMODITY SERVICES, Clothing/Personal Items, C...
15
                        NaN
16
                        NaN
                             COMMODITY SERVICES, Clothing/Personal Items, C...
17
                             HEALTH SERVICES, Outpatient Care, Community Cl...
                        {\tt NaN}
18
                        {\tt NaN}
                             HEALTH SERVICES, Outpatient Care, Community Cl...
19
                        NaN
20
                        NaN
                                                                                NaN
21
    passports@example.org
                                                                   Salud, Medicina
22
                        {\tt NaN}
                                     Ruby on Rails/Postgres/Redis, testing, wic
    location_id
0
               2
1
2
               3
3
               4
4
               5
5
               6
6
               7
7
               8
8
               9
9
              10
10
              11
              12
11
12
              13
13
              14
14
              15
15
              16
16
              17
17
              18
18
              19
19
              20
20
              21
21
              22
22
              22
```

 ${\tt NaN}$

```
[30]: id
                                      int64
      location_id
                                      int64
      program_id
                                    float64
      accepted_payments
                                     object
      alternate_name
                                     object
      application_process
                                     object
      audience
                                     object
      description
                                     object
      eligibility
                                     object
      email
                                     object
      fees
                                     object
                                     object
      funding_sources
      interpretation_services
                                     object
      keywords
                                     object
      languages
                                     object
      name
                                     object
      required_documents
                                     object
      service_areas
                                     object
      status
                                     object
      wait time
                                     object
      website
                                     object
      taxonomy_ids
                                     object
      dtype: object
[31]: # Note - Object data type is nothing but a string
     IMPORTANT _ This is how we read the EXCEL file by using the "pd.read_excel ()" function
[43]: pd.read_excel("LUSID Excel - Setting up your market data.xlsx")
[43]:
          Unnamed: 0
                       Unnamed: 1
                                    Unnamed: 2
      0
                  NaN
                               NaN
                                            NaN
      1
                  NaN
                               NaN
                                            NaN
      2
                  NaN
                               NaN
                                            NaN
      3
                  NaN
                               NaN
                                            NaN
      4
                  NaN
                               NaN
                                            NaN
      5
                  NaN
                               NaN
                                            NaN
      6
                  NaN
                               NaN
                                            NaN
      7
                  NaN
                               NaN
                                            {\tt NaN}
      8
                  NaN
                               NaN
                                            {\tt NaN}
      9
                  NaN
                               NaN
                                            NaN
                               NaN
      10
                  NaN
                                            NaN
      11
                  NaN
                               NaN
                                            NaN
      12
                  NaN
                               NaN
                                            NaN
      13
                  NaN
                               NaN
                                            NaN
      14
                  NaN
                               NaN
                                            NaN
      15
                                            NaN
                  NaN
                               NaN
```

16 17 18 19 20 21 22 23 24 25 26 27	NaN	NaN	NaN	
0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27		-	Datetimes in LUS Normat datetimes for tions please visit out Normat datetimes for Normat datetimes in LUS Normat datetimes for Norm	aN SID SaN
0 1 2			N	4 \ TaN TaN

3 4 5 6 7 8 9	NaN NaN Getting Started tutorials Knowledge Base articles or Contact us NaN NaN
11	LUSID accepts datetimes in any recognised Exce
12 13 14	NaN Excel formats to show datetime
14 15	NaN Enter the date to convert
16	NaN
17 18	This date has no time component NaN
19	Specify time component
20 21	NaN Add to your date and display the complete date
22 23	NaN
23 24	Although the date can still appears without time NaN
25 26	Add an hour to your datetime NaN
27	Subtract a minute from your datetime
	Unnamed: 5 Unnamed: 6 \
0	NaN NaN
1	NaN NaN
2	
3	NaN NaN
	NaN NaN
4	NaN NaN NaN NaN
4 5	NaN NaN NaN NaN NaN NaN
4 5 6	NaN NaN NaN NaN NaN NaN NaN NaN
4 5 6 7	NaN NaN NaN NaN NaN NaN NaN NaN NaN NaN
4 5 6 7 8	NaN
4 5 6 7 8 9	NaN
4 5 6 7 8 9 10	NaN
4 5 6 7 8 9 10 11	NaN
4 5 6 7 8 9 10	NaN
4 5 6 7 8 9 10 11 12	NaN NaN
4 5 6 7 8 9 10 11 12 13	NaN NaN
4 5 6 7 8 9 10 11 12 13 14	NaN NaN
4 5 6 7 8 9 10 11 12 13 14 15 16	NaN NaN 2019-04-10 00:00:00 NaN NaN 2019-04-10 00:00:00 NaN NaN
4 5 6 7 8 9 10 11 12 13 14 15	NaN NaN 2019-04-10 00:00:00 NaN NaN NaN NaN

```
20
                              NaN
                                            NaN
21
    2019-04-10 13:30:45.550000
                                            NaN
22
                                            NaN
23
    2019-04-10 13:30:45.550000
                                            NaN
24
                                            NaN
    2019-04-10 14:30:45.550000
25
                                            NaN
26
                                            NaN
27
   2019-04-10 14:29:45.550000
                                            NaN
                                                 Unnamed: 7 \
0
                                                         NaN
                                                         NaN
1
                                                         NaN
2
3
                                                         {\tt NaN}
4
                                                         NaN
5
                                                         {\tt NaN}
6
                                                         NaN
7
                                                         {\tt NaN}
8
                                                         NaN
9
                                                         {\tt NaN}
10
                                                         {\tt NaN}
    LUSID also accepts UTC, UTS offsets and cutlabels
12
                                                         NaN
13
                                        Additional formats
14
                                                         NaN
15
                                  Standard UTC timestamp:
16
17
                              Convert date into a string
18
19
          UTC offset (not recognised as an excel date)
20
                                                         {\tt NaN}
21
                                                   Cutlabel
22
                                                         NaN
23
                                                         NaN
24
                                                         NaN
25
                                                         NaN
26
                                                         NaN
27
                                                         {\tt NaN}
                     Unnamed: 8 \
0
                             NaN
                             NaN
1
2
                             NaN
3
                             NaN
4
                             NaN
5
                             NaN
                             NaN
```

```
7
                               NaN
8
                               NaN
9
                               NaN
10
                               NaN
11
                               NaN
12
                               NaN
13
                               NaN
14
                               NaN
15
     "YYYY-MM-DDTHH:MM:SS.OOZ"
16
17
      2019-04-10T 13:30:45.55Z
18
19
    2019-04-10T13:30:45+04:00
20
21
    2019-04-10NSingaporeClose
22
                               NaN
23
                               NaN
24
                               NaN
25
                               NaN
26
                               NaN
27
                               {\tt NaN}
                                                   Unnamed: 9
0
                                                           NaN
1
                                                           NaN
2
                                                           NaN
3
                                                           NaN
4
                                                           NaN
5
                                                           {\tt NaN}
6
                                                           NaN
7
                                                           NaN
8
                                                           NaN
9
                                                           NaN
10
                                                           NaN
11
                                                           NaN
12
                                                           {\tt NaN}
13
                                                           NaN
14
                                                           {\tt NaN}
15
    Where Z indicates Zulu time (UTC+0), and is eq...
16
                                                           NaN
17
                                                           NaN
18
                                                           NaN
19
                                                           NaN
20
                                                           {\tt NaN}
21
                                                           NaN
22
                                                           NaN
23
                                                           NaN
```

```
24
                                                              {\tt NaN}
      25
                                                              {\tt NaN}
      26
                                                              NaN
      27
                                                              NaN
 []: df1=pd.read_excel("LUSID Excel - Setting up your market data.xlsx")
[34]: type(df1)
[34]: pandas.core.frame.DataFrame
[35]: df1.dtypes
[35]: Unnamed: 0
                      float64
      Unnamed: 1
                      float64
      Unnamed: 2
                      float64
      Unnamed: 3
                       object
      Unnamed: 4
                       object
      Unnamed: 5
                       object
      Unnamed: 6
                      float64
      Unnamed: 7
                       object
      Unnamed: 8
                       object
      Unnamed: 9
                       object
      dtype: object
[36]: df1.columns
[36]: Index(['Unnamed: 0', 'Unnamed: 1', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4',
              'Unnamed: 5', 'Unnamed: 6', 'Unnamed: 7', 'Unnamed: 8', 'Unnamed: 9'],
             dtype='object')
[39]: df1["Unnamed: 1"]
[39]: 0
            NaN
      1
            NaN
      2
            NaN
      3
            NaN
      4
            NaN
      5
            {\tt NaN}
      6
            {\tt NaN}
      7
            {\tt NaN}
      8
            {\tt NaN}
      9
            {\tt NaN}
      10
            {\tt NaN}
      11
            NaN
      12
            NaN
      13
            NaN
```

```
14
       NaN
15
       NaN
16
       {\tt NaN}
17
       {\tt NaN}
18
       {\tt NaN}
19
       NaN
20
       {\tt NaN}
21
       {\tt NaN}
22
       NaN
23
       {\tt NaN}
       NaN
24
25
       {\tt NaN}
26
       {\tt NaN}
27
       {\tt NaN}
Name: Unnamed: 1, dtype: float64
```

[42]: df1[["Unnamed: 1","Unnamed: 3"]]

[42]:	Unnamed: 1	Unnamed: 3
0	NaN	NaN
1	NaN	Datetimes in LUSID
2	NaN	NaN
3	NaN	This sheet allows you to format datetimes for
4	NaN	If you have any questions please visit our:
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	NaN	NaN
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	NaN	NaN
26	NaN	NaN
27	NaN	NaN

```
⇔titanic.csv")
[45]:
            PassengerId
                          Survived
                                   Pclass
      0
                       1
                                 0
                                          3
      1
                       2
                                 1
                                          1
      2
                       3
                                 1
                                          3
      3
                       4
                                 1
                                          1
      4
                      5
                                 0
                                          3
      . .
                                          2
                                 0
      886
                    887
      887
                    888
                                 1
                                          1
                    889
      888
                                 0
                                          3
      889
                    890
                                 1
                                          1
      890
                    891
                                          3
                                                             Name
                                                                      Sex
                                                                                 SibSp \
                                                                             Age
      0
                                        Braund, Mr. Owen Harris
                                                                            22.0
                                                                     male
                                                                                       1
      1
            Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
                                                                                     1
      2
                                         Heikkinen, Miss. Laina
                                                                   female
                                                                            26.0
                                                                                       0
      3
                 Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                                   female
                                                                            35.0
                                                                                       1
      4
                                       Allen, Mr. William Henry
                                                                     male
                                                                            35.0
                                          Montvila, Rev. Juozas
      886
                                                                     male
                                                                            27.0
                                                                                       0
      887
                                  Graham, Miss. Margaret Edith
                                                                            19.0
                                                                                       0
                                                                   female
      888
                     Johnston, Miss. Catherine Helen "Carrie"
                                                                   female
                                                                             NaN
                                                                                       1
      889
                                          Behr, Mr. Karl Howell
                                                                            26.0
                                                                                       0
                                                                     male
      890
                                            Dooley, Mr. Patrick
                                                                     male
                                                                            32.0
                                                                                       0
           Parch
                              Ticket
                                          Fare Cabin Embarked
      0
                           A/5 21171
                                        7.2500
                                                  NaN
                0
                                                              S
      1
                0
                            PC 17599
                                       71.2833
                                                  C85
                                                              C
      2
                   STON/02. 3101282
                                        7.9250
                                                  NaN
                                                              S
      3
                0
                                       53.1000
                                                C123
                                                              S
                              113803
                                                              S
      4
                0
                              373450
                                        8.0500
                                                  NaN
      . .
                                           •••
                                                  •••
                               •••
      886
                0
                              211536
                                       13.0000
                                                  NaN
                                                              S
```

[45]: pd.read_csv("https://raw.githubusercontent.com/datasciencedojo/datasets/master/

[891 rows x 12 columns]

0

0

0

887

888

889

890

[48]: df3=pd.read_csv("https://raw.githubusercontent.com/datasciencedojo/datasets/

--master/titanic.csv")

30.0000

23.4500

30.0000

7.7500

112053

111369

370376

W./C. 6607

B42

NaN

NaN

C148

S

S

С

Q

```
[50]: df3
[50]:
           PassengerId Survived Pclass
                       1
                                  0
                                          3
      0
      1
                       2
                                  1
                                          1
      2
                       3
                                  1
                                          3
      3
                       4
                                  1
                                          1
      4
                       5
                                  0
                                          3
                                          2
      886
                    887
                                  0
      887
                    888
                                           1
                                  1
                                          3
      888
                    889
                                  0
      889
                    890
                                  1
                                          1
                                  0
      890
                    891
                                          3
                                                             Name
                                                                       Sex
                                                                              Age SibSp \
      0
                                        Braund, Mr. Owen Harris
                                                                      male
                                                                             22.0
                                                                                        1
      1
            Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
                                                                                     1
      2
                                         Heikkinen, Miss. Laina
                                                                    female
                                                                             26.0
                                                                                        0
      3
                 Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                                    female
                                                                             35.0
                                                                                        1
                                       Allen, Mr. William Henry
      4
                                                                      male
                                                                             35.0
                                                                                        0
      . .
                                                                     •••
                                                                               •••
      886
                                          Montvila, Rev. Juozas
                                                                      male
                                                                             27.0
                                                                                        0
      887
                                   Graham, Miss. Margaret Edith
                                                                            19.0
                                                                    female
                                                                                        0
      888
                      Johnston, Miss. Catherine Helen "Carrie"
                                                                    female
                                                                              NaN
                                                                                        1
      889
                                          Behr, Mr. Karl Howell
                                                                      male
                                                                            26.0
                                                                                        0
      890
                                             Dooley, Mr. Patrick
                                                                      male
                                                                            32.0
                                                                                        0
                                          Fare Cabin Embarked
           Parch
                              Ticket
      0
                0
                           A/5 21171
                                        7.2500
                                                  NaN
      1
                0
                            PC 17599
                                       71.2833
                                                  C85
                                                              C
      2
                0
                   STON/02. 3101282
                                        7.9250
                                                  NaN
                                                              S
      3
                0
                              113803
                                       53.1000
                                                 C123
                                                              S
      4
                0
                              373450
                                        8.0500
                                                              S
                                                  NaN
      . .
                                                   •••
      886
                0
                              211536
                                       13.0000
                                                  NaN
                                                              S
                                                  B42
                                                              S
      887
                0
                              112053
                                       30.0000
      888
                2
                          W./C. 6607
                                       23.4500
                                                  NaN
                                                              S
      889
                0
                                                 C148
                                                              С
                              111369
                                       30.0000
                                                              Q
      890
                              370376
                                        7.7500
                                                  NaN
      [891 rows x 12 columns]
[51]: df3.head()
```

[51]:

0

PassengerId Survived Pclass \

0

3

1

```
1
                   2
                                      1
                              1
      2
                   3
                                      3
                              1
      3
                   4
                              1
                                      1
      4
                   5
                              0
                                      3
                                                        Name
                                                                  Sex
                                                                        Age SibSp \
      0
                                    Braund, Mr. Owen Harris
                                                                 male
                                                                       22.0
                                                                                  1
      1
         Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
                                                                                1
      2
                                     Heikkinen, Miss. Laina
                                                                                  0
                                                               female
                                                                       26.0
      3
              Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                                       35.0
                                                               female
                                                                                  1
                                    Allen, Mr. William Henry
      4
                                                                 male
                                                                       35.0
                                                                                  0
         Parch
                           Ticket
                                      Fare Cabin Embarked
      0
             0
                        A/5 21171
                                    7.2500
                                              NaN
                                                         S
                         PC 17599
                                   71.2833
                                                         С
      1
             0
                                              C85
      2
                                    7.9250
                                                         S
             0
                STON/02. 3101282
                                              NaN
      3
                                                         S
                                   53.1000
                                             C123
             0
                           113803
      4
             0
                           373450
                                    8.0500
                                                         S
                                              NaN
[52]: df3.columns
[52]: Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',
             'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
            dtype='object')
[53]: type(df3)
[53]: pandas.core.frame.DataFrame
[54]:
      df3["Name"]
[54]: 0
                                         Braund, Mr. Owen Harris
      1
             Cumings, Mrs. John Bradley (Florence Briggs Th ...
      2
                                         Heikkinen, Miss. Laina
      3
                  Futrelle, Mrs. Jacques Heath (Lily May Peel)
      4
                                        Allen, Mr. William Henry
      886
                                           Montvila, Rev. Juozas
      887
                                   Graham, Miss. Margaret Edith
      888
                       Johnston, Miss. Catherine Helen "Carrie"
                                           Behr, Mr. Karl Howell
      889
      890
                                             Dooley, Mr. Patrick
      Name: Name, Length: 891, dtype: object
[56]: df3[["Age","Ticket"]]
```

```
[56]:
            Age
                             Ticket
            22.0
                          A/5 21171
      0
      1
            38.0
                           PC 17599
      2
           26.0
                  STON/02. 3101282
      3
            35.0
                             113803
      4
            35.0
                             373450
      . .
            •••
                              •••
      886
           27.0
                             211536
           19.0
      887
                             112053
      888
            NaN
                         W./C. 6607
      889
           26.0
                             111369
      890
           32.0
                             370376
      [891 rows x 2 columns]
[59]:
     df3.tail()
            PassengerId
                          Survived
[59]:
                                   Pclass
                                                                                     Name
                                                                  Montvila, Rev. Juozas
      886
                    887
                                 0
                                          2
      887
                    888
                                 1
                                                           Graham, Miss. Margaret Edith
                                          1
                                             Johnston, Miss. Catherine Helen "Carrie"
      888
                    889
                                 0
                                          3
      889
                    890
                                 1
                                          1
                                                                  Behr, Mr. Karl Howell
                                          3
      890
                    891
                                 0
                                                                    Dooley, Mr. Patrick
               Sex
                     Age
                           SibSp
                                  Parch
                                              Ticket
                                                        Fare Cabin Embarked
                    27.0
      886
              male
                               0
                                       0
                                               211536
                                                       13.00
                                                                NaN
                                                                            S
      887
           female
                    19.0
                               0
                                       0
                                                       30.00
                                                                B42
                                                                            S
                                               112053
                                       2
                                          W./C. 6607
                                                                            S
      888
           female
                     NaN
                               1
                                                       23.45
                                                                NaN
      889
                    26.0
                                                                            С
              male
                               0
                                       0
                                               111369
                                                       30.00
                                                               C148
      890
              male
                    32.0
                               0
                                       0
                                               370376
                                                        7.75
                                                                NaN
                                                                            Q
[62]: df3.tail(3)
[62]:
            PassengerId
                          Survived
                                    Pclass
                                                                                     Name
                                             Johnston, Miss. Catherine Helen "Carrie"
      888
                    889
                                 0
                                          3
      889
                    890
                                          1
                                                                  Behr, Mr. Karl Howell
                                 1
      890
                    891
                                 0
                                          3
                                                                    Dooley, Mr. Patrick
               Sex
                           SibSp
                                  Parch
                                                        Fare Cabin Embarked
                     Age
                                               Ticket
      888
           female
                     NaN
                               1
                                       2
                                          W./C. 6607
                                                       23.45
                                                                NaN
      889
              male
                    26.0
                               0
                                       0
                                               111369
                                                       30.00
                                                               C148
                                                                            С
      890
                    32.0
              male
                               0
                                       0
                                               370376
                                                        7.75
                                                                NaN
                                                                            Q
[64]:
      pip install lxml
```

Collecting lxml

Downloading lxml-4.9.3-cp310-cp310-manylinux_2_28_x86_64.whl (7.9 MB)

7.9/7.9 MB

56.6 MB/s eta 0:00:0000:0100:01
Installing collected packages: lxml

```
⇔NBA_2015_totals.html")
 [6]: pip install lxml
      Requirement already satisfied: lxml in /opt/conda/lib/python3.10/site-packages
      Note: you may need to restart the kernel to use updated packages.
 [7]: type(url_data)
 [7]: list
     len(url data)
 [8]:
 [8]: 1
      df3=url_data[0]
[10]:
      df3
[10]:
                                                                                        ORB
             Rk
                           Player Pos Age
                                              Tm
                                                    G
                                                       GS
                                                              MP
                                                                   FG
                                                                        FGA
                                                                                  FT%
                                                                                              \
      0
              1
                      Quincy Acy
                                   PF
                                        24
                                             NYK
                                                  68
                                                       22
                                                            1287
                                                                   152
                                                                        331
                                                                                  .784
                                                                                         79
      1
              2
                    Jordan Adams
                                    SG
                                        20
                                             MEM
                                                  30
                                                        0
                                                             248
                                                                    35
                                                                         86
                                                                                 .609
                                                                                          9
      2
                    Steven Adams
              3
                                     C
                                        21
                                             OKC
                                                  70
                                                       67
                                                            1771
                                                                   217
                                                                        399
                                                                                  .502
                                                                                        199
      3
                     Jeff Adrien PF
                                                             215
              4
                                        28
                                             MIN
                                                   17
                                                        0
                                                                    19
                                                                         44
                                                                                  .579
                                                                                         23
      4
                   Arron Afflalo
                                    SG
                                        29
                                             TOT
                                                  78
                                                       72
                                                            2502
                                                                   375
                                                                                  .843
                                                                        884
                                                                                         27
                                    . .
      670
            490
                  Thaddeus Young
                                   PF
                                        26
                                             TOT
                                                  76
                                                       68
                                                            2434
                                                                  451
                                                                        968
                                                                                 .655
                                                                                        127
      671
            490
                  Thaddeus Young
                                    PF
                                        26
                                             MIN
                                                  48
                                                       48
                                                            1605
                                                                   289
                                                                        641
                                                                                  .682
                                                                                         75
      672
            490
                  Thaddeus Young
                                        26
                                             BRK
                                                       20
                                                             829
                                                                   162
                                                                        327
                                                                                 .606
                                                                                         52
                                    PF
                                                  28
      673
                     Cody Zeller
            491
                                     C
                                        22
                                             CHO
                                                  62
                                                       45
                                                            1487
                                                                   172
                                                                        373
                                                                                 .774
                                                                                         97
      674
            492
                    Tyler Zeller
                                     С
                                        25
                                             BOS
                                                  82
                                                       59
                                                                        619
                                                                                  .823
                                                            1731
                                                                   340
                                                                                        146
            DRB
                             STL BLK
                                       TOV
                                              PF
                                                    PTS
                  TRB
                       AST
      0
            222
                  301
                        68
                              27
                                   22
                                        60
                                             147
                                                    398
      1
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                        16
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                                    7
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                                              24
                                                     94
      2
            324
                  523
                              38
                                   86
                                        99
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                                                    537
                        66
      3
             54
                   77
                        15
                               4
                                    9
                                         9
                                              30
                                                     60
      4
                                    7
            220
                  247
                       129
                              41
                                       116
                                             167
                                                   1035
      670
            284
                  411
                       173
                             124
                                   25
                                       117
                                             171
                                                   1071
      671
            170
                  245
                       135
                              86
                                   17
                                        75
                                             115
                                                    685
      672
                  166
                                        42
                                              56
                                                    386
            114
                        38
                              38
                                    8
                                                    472
      673
            265
                  362
                       100
                              34
                                   49
                                        62
                                             156
      674
            319
                  465
                                   52
                                        76
                                             205
                                                    833
                       113
                              18
```

url_data=pd.read_html("https://www.basketball-reference.com/leagues/

```
[11]: df3.head()
[11]:
        Rk
                     Player Pos Age
                                        Tm
                                              G
                                                 GS
                                                        MP
                                                              FG
                                                                  FGA
                                                                            FT%
                                                                                  ORB
                                                                                        DRB
                                                                                             \
          1
                Quincy Acy
                              PF
                                   24
                                       NYK
                                             68
                                                 22
                                                      1287
                                                             152
                                                                  331
                                                                            .784
                                                                                   79
                                                                                        222
      0
      1
          2
              Jordan Adams
                                                                           .609
                              SG
                                  20
                                       MEM
                                             30
                                                  0
                                                       248
                                                              35
                                                                   86
                                                                                    9
                                                                                         19
      2
          3
              Steven Adams
                               С
                                  21
                                       OKC
                                             70
                                                 67
                                                      1771
                                                             217
                                                                  399
                                                                            .502
                                                                                        324
                                                                                  199
          4
                              PF
      3
               Jeff Adrien
                                  28
                                       MIN
                                             17
                                                  0
                                                       215
                                                              19
                                                                   44
                                                                            .579
                                                                                   23
                                                                                         54
          5
             Arron Afflalo
                              SG
                                  29
                                       TOT
                                             78
                                                 72
                                                      2502
                                                             375
                                                                  884
                                                                            .843
                                                                                   27
                                                                                        220
          TRB
               AST STL BLK
                              TOV
                                     PF
                                           PTS
      0
          301
                68
                     27
                          22
                               60
                                    147
                                           398
           28
                     16
                          7
                                     24
      1
                16
                               14
                                           94
      2
          523
                66
                     38
                         86
                               99
                                    222
                                           537
      3
           77
                15
                      4
                           9
                                9
                                     30
                                           60
                          7
          247
               129
                     41
                                    167
                                         1035
                              116
      [5 rows x 30 columns]
[12]: df3.tail()
[12]:
             Rk
                          Player Pos Age
                                              Tm
                                                   G
                                                       GS
                                                             MP
                                                                   FG
                                                                        FGA
                                                                                  FT%
                                                                                        ORB
      670
            490
                 Thaddeus Young PF
                                             TOT
                                                  76
                                                       68
                                                           2434
                                                                  451
                                                                        968
                                                                                 .655
                                                                                        127
                                        26
            490
                  Thaddeus Young
      671
                                   PF
                                        26
                                             MIN
                                                  48
                                                       48
                                                           1605
                                                                  289
                                                                        641
                                                                                 .682
                                                                                         75
      672
            490
                  Thaddeus Young
                                   PF
                                        26
                                                  28
                                                       20
                                                             829
                                                                  162
                                                                        327
                                                                                 .606
                                                                                         52
                                             BRK
      673
                     Cody Zeller
                                                                                 .774
            491
                                        22
                                             CHO
                                                  62
                                                       45
                                                                  172
                                                                        373
                                                                                         97
                                     С
                                                            1487
      674
            492
                    Tyler Zeller
                                     С
                                        25
                                             BOS
                                                  82
                                                       59
                                                            1731
                                                                  340
                                                                        619
                                                                                 .823
                                                                                        146
            DRB
                  TRB
                       AST
                             STL BLK
                                       TOV
                                              PF
                                                   PTS
      670
            284
                  411
                             124
                                  25
                                                  1071
                       173
                                       117
                                             171
      671
            170
                  245
                       135
                                  17
                                        75
                                             115
                                                   685
                              86
      672
            114
                  166
                        38
                              38
                                   8
                                        42
                                              56
                                                   386
      673
            265
                                                   472
                  362
                       100
                              34
                                  49
                                        62
                                             156
      674
            319
                  465
                       113
                              18
                                  52
                                        76
                                             205
                                                   833
      [5 rows x 30 columns]
[13]: df3.columns
[13]: Index(['Rk', 'Player', 'Pos', 'Age', 'Tm', 'G', 'GS', 'MP', 'FG', 'FGA', 'FG%',
              '3P', '3PA', '3P%', '2P', '2PA', '2P%', 'eFG%', 'FT', 'FTA', 'FT%',
               'ORB', 'DRB', 'TRB', 'AST', 'STL', 'BLK', 'TOV', 'PF', 'PTS'],
             dtype='object')
[14]: df3.dtypes
```

```
object
[14]: Rk
                 object
      Player
      Pos
                 object
      Age
                 object
                 object
      Tm
      G
                 object
      GS
                 object
      MΡ
                 object
      FG
                 object
      FGA
                 object
      FG%
                 object
      3P
                 object
      3PA
                 object
      3P%
                 object
      2P
                 object
      2PA
                 object
      2P%
                 object
      eFG%
                 object
      FT
                 object
      FTA
                 object
      FT%
                 object
      ORB
                 object
      DRB
                 object
      TRB
                 object
      AST
                 object
      STL
                 object
      BLK
                 object
      TOV
                 object
      PF
                 object
      PTS
                 object
      dtype: object
```

MOST IMPORTANT - We use ".to_csv() or .to_excel etc " for function for saving the data in file or any format like if we want to save the players data than we have to execute df3.to_csv("players_data.csv") than it will save that in csv in another file and we also download that file in our local system as well

```
[15]: df3.to_csv("players_data.csv")
[16]: df3.to_csv("players_data.csv",index=False)
```

THANK YOU SO MUCH!!

YOURS VIRAT TIWARI :)