```
ward
                  alderman
                                                     address
                                                               zip
     1 Proco "Joe" Moreno
0
                                  2058 NORTH WESTERN AVENUE
                                                             60647
1
     2
              Brian Hopkins
                                 1400 NORTH ASHLAND AVENUE
                                                             60622
2
     3
                Pat Dowell
                                     5046 SOUTH STATE STREET
                                                             60609
          William D. Burns 435 EAST 35TH STREET, 1ST FLOOR 60616
3
     4
     5 Leslie A. Hairston
                                      2325 EAST 71ST STREET
                                                             60649
(50, 4)
```

	ward	pop_2000	pop_2010	change	addre
SS	\				
0	1	52951	56149	6%	2765 WEST SAINT MARY STRE
ET					
1	2	54361	55805	3%	WM WASTE MANAGEMENT 15
90					
2	3	40385	53039	31%	17 EAST 38TH STRE
ET					
3	4	51953	54589	5%	31ST ST HARBOR BUILDING LAKEFRONT TRA
ΙL					
4	5	55302	51455	-7%	JACKSON PARK LAGOON SOUTH CORNELL DRI
VE					

```
In [5]:
         wards_census = wards.merge(census, on="ward")
            print(wards_census.head(4))
               ward
                               alderman
                                                                address_x
                                                                           zip_x pop_
            2000
                  1
                    Proco "Joe" Moreno
                                                2058 NORTH WESTERN AVENUE
                                                                           60647
                                                                                      5
            2951
                  2
                                               1400 NORTH ASHLAND AVENUE
                                                                           60622
                                                                                      5
            1
                          Brian Hopkins
            4361
            2
                             Pat Dowell
                                                  5046 SOUTH STATE STREET
                                                                           60609
                                                                                      4
            0385
                  4
                                                                           60616
                                                                                      5
            3
                       William D. Burns 435 EAST 35TH STREET, 1ST FLOOR
            1953
                                                               address_y
               pop_2010 change
                                                                          zip_y
            0
                  56149
                            6%
                                             2765 WEST SAINT MARY STREET
                                                                          60647
            1
                            3%
                  55805
                                                WM WASTE MANAGEMENT 1500 60622
            2
                           31%
                                                     17 EAST 38TH STREET
                  53039
                                                                          60653
            3
                  54589
                            5% 31ST ST HARBOR BUILDING LAKEFRONT TRAIL
                                                                          60653
         ▶ print(wards_census.shape)
In [6]:
            (50, 9)
In [7]:
         #Ex 1
            taxi owners = pd.read pickle("taxi owners.p")
            taxi vehicles = pd.read pickle("taxi vehicles.p")
            taxi_own_veh = taxi_owners.merge(taxi_vehicles, on = "vid")
            print(taxi own veh["fuel type"].value counts())
            print()
            print("The highest is : ", taxi_own_veh["fuel_type"].max())
                                       2792
            HYBRID
            GASOLINE
                                        611
            FLEX FUEL
                                         89
            COMPRESSED NATURAL GAS
                                         27
            Name: fuel_type, dtype: int64
            The highest is: HYBRID
```

wa		alderman	address_x	Z
ip_x 0 0647		Proco "Joe" Moreno	2058 NORTH WESTERN AVENUE	6
1 0622	2	Brian Hopkins	1400 NORTH ASHLAND AVENUE	6
2 0609	3	Pat Dowell	5046 SOUTH STATE STREET	6
3 0616	4	William D. Burns	435 EAST 35TH STREET, 1ST FLOOR	6
	5	Leslie A. Hairston	2325 EAST 71ST STREET	6
5 0619	6	Roderick T. Sawyer	8001 S. MARTIN LUTHER KING DRIVE	6
6 0617	7	Gregory I. Mitchell	2249 EAST 95TH STREET	6
	8	Michelle A. Harris	8539 SOUTH COTTAGE GROVE AVENUE	6
	9	Anthony A. Beale	34 EAST 112TH PLACE	6
	10	Susan Sadlowski Garza	10500 SOUTH EWING AVENUE	6
	11	Patrick Daley Thompson	3659 SOUTH HALSTED STREET	6
	12	George Cardenas	3476 SOUTH ARCHER AVENUE	6
	13	Marty Quinn	6500 SOUTH PULASKI ROAD	6
	14	Edward M. Burke	2650 WEST 51ST STREET	6
	15	Raymond A. Lopez	1650 WEST 63RD STREET	6
	16	Toni L. Foulkes	3045 WEST 63RD STREET	6
	17	David H. Moore	7313 SOUTH ASHLAND AVENUE	6
	18	Derrick G. Curtis	8359 SOUTH PULASKI ROAD	6
	19	Matthew J. O'Shea	10400 SOUTH WESTERN AVENUE	6
	20	Willie B. Cochran	6357 SOUTH COTTAGE GROVE AVENUE	6
	21	Howard B. Brookins, Jr.	9011 SOUTH ASHLAND AVENUE, UNIT B	6
	22	Ricardo Munoz	2500 SOUTH ST. LOUIS AVENUE	6
	23	Michael R. Zalewski	6247 SOUTH ARCHER AVENUE	6

23 0624	24	Michael Scott, Jr.	1158 SOUTH KEELER AVENUE	6
24 0608	25	Daniel "Danny" Solis	1800 SOUTH BLUE ISLAND AVENUE	6
25 0622	26	Roberto Maldonado	2511 WEST DIVISION STREET	6
26 0612	27	Walter Burnett, Jr.	4 NORTH WESTERN AVENUE	6
27 0612	28	Jason C. Ervin	2602 WEST 16TH STREET	6
28 0639	29	Chris Taliaferro	6272 WEST NORTH AVENUE	6
29 0641	30	Ariel E. Reyboras	3559 NORTH MILWAUKEE AVENUE	6
30 0639	31	Milagros "Milly" Santiago	2521 NORTH PULASKI ROAD	6
31 0614	32	Scott Waguespack	2657 NORTH CLYBOURN AVENUE	6
32 0618	33	Deborah Mell	3001 WEST IRVING PARK ROAD	6
33 0628	34	Carrie M. Austin	507 WEST 111TH STREET	6
34 0647	35	Carlos Ramirez-Rosa	2710 NORTH SAWYER AVENUE	6
35 0607	36	Gilbert Villegas	6934 WEST DIVERSEY	6
36 0651	37	Emma M. Mitts	4924 WEST CHICAGO AVENUE	6
37	38	Nicholas Sposato	3821 NORTH HARLEM AVENUE	6
0634 38 0630	39	Margaret Laurino	4404 WEST LAWRENCE AVENUE	6
39 0659	40	Patrick J. O'Connor	5850 NORTH LINCOLN AVENUE	6
40 0631	41	Anthony V. Napolitano	7442 NORTH HARLEM AVENUE	6
41 0654	42	Brendan Reilly	325 WEST HURON STREET, SUITE 510	6
42 0614	43	Michelle Smith	2523 NORTH HALSTED STREET	6
43 0657	44	Tom Tunney	3223 NORTH SHEFFIELD AVENUE	6
44	45	John S. Arena	4754 NORTH MILWAUKEE AVENUE	6
0630 45 0640	46	James Cappleman	4544 NORTH BROADWAY AVENUE	6
0 1 2 3 4 5 6 7	52 54 40 51 55 54	000 pop_2010 change \ 951 56149 6% 361 55805 3% 385 53039 31% 953 54589 5% 302 51455 -7% 989 52341 -5% 593 51581 -6% 039 51687 -4%		

8

52008

51519

-1%

52008	51519	-1%		
56613	51535	-9%		
64228	51497	-20%		
68922	52235	-24%		
64382	53722	-17%		
80143	54031	-33%		
57204	55184			
63695	55598	-13%		
49922	51599	3%		
57588	55281	-4%		
63376	54766	-14%		
56120	51538	-8%		
66011	56001	-15%		
64291	55882	-13%		
58652	55319	-6%		
56127	55991	0%		
68102	55870	-18%		
	56170	- 3%		
58758	56058	-5%		
58758 60653	56058 55967	-5% -8%		
58758	56058	-5% -8%		
58758 60653	56058 55967	-5% -8%	address v	zip v
58758 60653	56058 55967	-5% -8% -5%	address_y 5 WEST SAINT MARY STREET	
58758 60653	56058 55967	-5% -8% -5%	WEST SAINT MARY STREET	60647
58758 60653	56058 55967	-5% -8% -5%	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500	60647 60622
58758 60653	56058 55967 53784	-5% -8% -5% 2765	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET	60647 60622 60653
58758 60653	56058 55967 53784 31ST ST HA	-5% -8% -5% 2765 V ARBOR E	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL	60647 60622 60653 60653
58758 60653	56058 55967 53784 31ST ST HA	-5% -8% -5% 2765 V ARBOR E	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE	60647 60622 60653 60653 60637
58758 60653	56058 55967 53784 31ST ST HA	-5% -8% -5% 2765 W ARBOR E	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET	60647 60622 60653 60653 60637 60636
58758 60653	56058 55967 53784 31ST ST HA	-5% -8% -5% 2765 V ARBOR E ARK LAG	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE	60647 60622 60653 60653 60637 60636 60617
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG	WEST SAINT MARY STREET WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET	60647 60622 60653 60653 60637 60636 60617 60649
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG	WEST SAINT MARY STREET WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE	60647 60622 60653 60653 60637 60636 60617 60649 60628
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 WARBOR E ARK LAG 85 134 9-11059	WEST SAINT MARY STREET WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE F 943-947 WEST 14TH PLACE	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394 60607
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG 85 134 9-11059	WEST SAINT MARY STREET WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE F 943-947 WEST 14TH PLACE 46 STEVENSON EXPRESSWAY	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394 60607 60632
58758 60653	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG 85 134 9-11059	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE 10534 SOUTH AVENUE 46 STEVENSON EXPRESSWAY MMP SOUTH LARAMIE AVENUE	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394 60607 60632 60638
58758 60653 56587	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG 85 134 9-11059 CP	WEST SAINT MARY STREET WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE 10534 SOUTH AVENUE 46 STEVENSON EXPRESSWAY AMP SOUTH LARAMIE AVENUE 4540 WEST 51ST STREET	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394 60607 60632 60638 60632
58758 60653 56587	56058 55967 53784 31ST ST HA JACKSON PA	-5% -8% -5% 2765 V ARBOR E ARK LAG 85 134 9-11059 CP	WEST SAINT MARY STREET WM WASTE MANAGEMENT 1500 17 EAST 38TH STREET BUILDING LAKEFRONT TRAIL GOON SOUTH CORNELL DRIVE 150 WEST 74TH STREET 649 SOUTH OGLESBY AVENUE 16-1352 EAST 75TH STREET 9 SOUTH WENTWORTH AVENUE 10534 SOUTH AVENUE 10534 SOUTH AVENUE 46 STEVENSON EXPRESSWAY MMP SOUTH LARAMIE AVENUE	60647 60622 60653 60653 60637 60636 60617 60649 60628 46394 60607 60632 60638
	56613 64228 68922 64382 80143 56057 50205 49264 55043 54546 51854 51751 59734 63691 50879 55954 56841 61287 49423 61949 72698 65045 57204 63695 49922 57588 63376 56120 66011 64291 58652 56127 68102	56613 51535 64228 51497 68922 52235 64382 53722 80143 54031 56057 51501 50205 51954 49264 51846 55043 52992 54546 51525 51854 52372 51751 51632 59734 53515 63691 53728 50879 54909 55954 54539 56841 53516 61287 52939 49423 55199 61949 55267 72698 55560 65045 53724 57204 55184 63695 55598 49922 51599 57588 55281 63376 54766 56120 51538 66011 56001 64291 55882 58652 55319 56127 55991 68102 55870	56613 51535 -9% 64228 51497 -20% 68922 52235 -24% 64382 53722 -17% 80143 54031 -33% 56057 51501 -8% 50205 51954 3% 49264 51846 5% 55043 52992 -4% 54546 51525 -6% 51854 52372 1% 51751 51632 0% 59734 53515 -10% 63691 53728 -16% 50879 54909 8% 55954 54539 -3% 56841 53516 -6% 61287 52939 -14% 49423 55199 12% 61949 55267 -11% 72698 55560 -24% 65045 53724 -17% 57204 55184 -4% 63695 55598 -13% 49922 51599 3% 57588	56613 51535 -9% 64228 51497 -20% 68922 52235 -24% 64382 53722 -17% 80143 54031 -33% 56057 51501 -8% 50205 51954 3% 49264 51846 5% 55043 52992 -4% 54546 51525 -6% 51854 52372 1% 51751 51632 0% 59734 53515 -10% 63691 53728 -16% 50879 54909 8% 55954 54539 -3% 56841 53516 -6% 61287 52939 -14% 49423 55199 12% 61949 55267 -11% 72698 55560 -24% 63045 53724 -17% 57204 55184 -4% 63695 55598 -13% 49922 51599 3% 57588

```
16
                         7216 SOUTH WINCHESTER AVENUE
                                                        60636
17
                            3286 WEST COLUMBUS AVENUE
                                                        60652
18
                          9999 SOUTH FRANCISCO AVENUE
                                                        60805
19
                       DAN RYAN EXPRESSWAY PARK MANOR
                                                       60621
20
                       8852-8854 SOUTH EMERALD AVENUE
                                                        60620
21
                                4233 WEST 36TH STREET
                                                        60632
22
    CHICAGO MIDWAY INTERNATIONAL AIRPORT WEST 62ND...
                                                        60629
23
                         1635 SOUTH CHRISTIANA AVENUE
                                                        60623
24
                        1632-1746 SOUTH MILLER STREET
                                                        60608
25
               LITTLE CUBS FIELD COMFORT STATION 1400
                                                        60622
26
                        2151-2153 WEST CHICAGO AVENUE
                                                        60651
                          RML SPECIALTY HOSPITAL 3435
27
                                                        60624
28
                          1241 NORTH RIDGELAND AVENUE
                                                        60302
29
                            5118 WEST FLETCHER STREET
                                                        60641
30
                            2854 NORTH KEATING AVENUE
                                                        60641
31
                          2901 NORTH WASHTENAW AVENUE
                                                        60618
32
                      4041-4043 NORTH RICHMOND STREET
                                                        60625
                      11544-11546 SOUTH PEORIA STREET
33
                                                        60827
34
                             3634 WEST BELMONT AVENUE
                                                        60618
35
                         2918 NORTH RUTHERFORD AVENUE
                                                        60634
36
                           4738-4748 WEST RICE STREET
                                                        60651
37
                      7307-7331 WEST IRVING PARK ROAD
                                                        60706
38
                    QUEEN OF ALL SAINTS BASILICA 6280
                                                        60646
39
                           5536 NORTH ARTESIAN AVENUE
                                                        60645
40
                            1652 SOUTH CLIFTON AVENUE
                                                        60068
41
                            410-420 WEST GRAND AVENUE
                                                        60654
42
                                 LINCOLN PARK ZOO 2001
                                                        60614
43
                           507-513 WEST ALDINE AVENUE
                                                        60657
44
         CONGREGATIONAL CHURCH OF JEFFERSON PARK 5320
                                                        60630
45
                   UPTOWN BROADWAY BUILDING 4743-4763
                                                        60640
```

In [9]: | licenses = pd.read_csv('Business_Licenses.csv')
 print(licenses.head())
 print(licenses.shape)

```
account ward aid
                                                               address
                                        business
0
    307071
              3
                743
                           REGGIE'S BAR & GRILL
                                                       2105 S STATE ST
1
       10
             10 829
                                     HONEYBEERS
                                                   13200 S HOUSTON AVE
2
             14
                 775
                                                     5089 S ARCHER AVE
    10002
                                     CELINA DELI
3
    10005
             12
                 NaN
                      KRAFT FOODS NORTH AMERICA
                                                        2005 W 43RD ST
    10044
              44 638 NEYBOUR'S TAVERN & GRILLE 3651 N SOUTHPORT AVE
```

```
zip
0 60616.0
1 60633.0
2 60632.0
3 60609.0
4 60613.0
(10000, 6)
```

```
In [10]:
          ward_licenses = wards.merge(licenses, on = "ward", suffixes = ("_ward",
             print(ward_licenses.head())
                ward
                                alderman
                                                       address_ward
                                                                     zip_ward
             0
                   1 Proco "Joe" Moreno 2058 NORTH WESTERN AVENUE
                                                                                 12024
                                                                        60647
                   1 Proco "Joe" Moreno 2058 NORTH WESTERN AVENUE
             1
                                                                        60647
                                                                                 14446
             2
                   1 Proco "Joe" Moreno 2058 NORTH WESTERN AVENUE
                                                                                 14624
                                                                        60647
                   1 Proco "Joe" Moreno 2058 NORTH WESTERN AVENUE
                                                                        60647
                                                                                 14987
                   1 Proco "Joe" Moreno 2058 NORTH WESTERN AVENUE
                                                                        60647
                                                                                 15642
                aid
                                                       address_lic zip_lic
                                 business
             0 NaN
                      DIGILOG ELECTRONICS
                                                1038 N ASHLAND AVE
                                                                    60622.0
               743
                         EMPTY BOTTLE INC
                                            1035 N WESTERN AVE 1ST
                                                                    60622.0
                775 LITTLE MEL'S HOT DOG
                                           2205 N CALIFORNIA AVE
                                                                    60647.0
             3
                       MR. BROWN'S LOUNGE
                                            2301 W CHICAGO AVE 1ST
                NaN
                                                                    60622.0
                814
                             Beat Kitchen 2000-2100 W DIVISION ST 60622.0
In [11]:

    print(wards.shape)

             print(ward_licenses.shape)
             (50, 4)
             (10000, 9)
 In [ ]:
          H
```

```
In [12]:
         #Ex 3
           licenses = pd.read_pickle("licenses.p")
           biz_owners = pd.read_pickle("business_owners.p")
           licenses_owners = licenses.merge(biz_owners, on = "account")
           print(licenses_owners.head())
           print("----")
           counted_df = licenses_owners.groupby("title").agg({"account" : "count"})
           print(counted_df.head())
           print("----")
           sorted_df = counted_df.sort_values(["account"], ascending = False)
           print(sorted_df.head())
             account ward aid
                                        business
                                                           address
                                                                     zip \
           0 307071 3 743 REGGIE'S BAR & GRILL
                                                    2105 S STATE ST 60616
                 10 10 829
                                     HONEYBEERS 13200 S HOUSTON AVE 60633
           1
           2
                 10 10 829
                                     HONEYBEERS 13200 S HOUSTON AVE 60633
                                   CELINA DELI 5089 S ARCHER AVE 60632
           3 10002 14 775
                                    CELINA DELI 5089 S ARCHER AVE 60632
              10002 14 775
             first_name last_name title
    ROBERT GLICK MEMBER
           1
                 PEARL SHERMAN PRESIDENT
           2
                 PEARL SHERMAN SECRETARY
           3
                WALTER MROZEK PARTNER
                CELINA
                         BYRDAK
                                  PARTNER
                             account
           title
           ASST. SECRETARY
                                111
           BENEFICIARY
                                  4
           CEO
                                110
           DIRECTOR
                                146
           EXECUTIVE DIRECTOR
                                 10
                          account
           title
           PRESIDENT
                             6259
                             5205
           SECRETARY
           SOLE PROPRIETOR
                             1658
           OTHER
                             1200
           VICE PRESIDENT
                            970
```

```
grants = pd.read_csv("Smalllajdf;ajdsf; . csv ")
In [13]:
             print(grants.head())
             FileNotFoundError
                                                        Traceback (most recent call las
             Input In [13], in <cell line: 1>()
             ----> 1 grants = pd.read_csv("Smalllajdf;ajdsf; . csv ")
                   2 print(grants.head())
             File ~\anaconda3\lib\site-packages\pandas\util\_decorators.py:311, in dep
             recate_nonkeyword_arguments.<locals>.decorate.<locals>.wrapper(*args, **k
             wargs)
                 305 if len(args) > num_allow_args:
                         warnings.warn(
                 307
                             msg.format(arguments=arguments),
                 308
                             FutureWarning,
                             stacklevel=stacklevel,
                 309
                 310
             --> 311 return func(*args, **kwargs)
             File ~\anaconda3\lib\site-packages\pandas\io\parsers\readers.py:680, in r
             ead_csv(filepath_or_buffer, sep, delimiter, header, names, index_col, use
             cols, squeeze, prefix, mangle_dupe_cols, dtype, engine, converters, true_
             values, false_values, skipinitialspace, skiprows, skipfooter, nrows, na_v
             alues, keep_default_na, na_filter, verbose, skip_blank_lines, parse_date
             s, infer_datetime_format, keep_date_col, date_parser, dayfirst, cache_dat
             es, iterator, chunksize, compression, thousands, decimal, lineterminator,
             quotechar, quoting, doublequote, escapechar, comment, encoding, encoding_
             errors, dialect, error_bad_lines, warn_bad_lines, on_bad_lines, delim_whi
             tespace, low_memory, memory_map, float_precision, storage_options)
                 665 kwds_defaults = _refine_defaults_read(
                 666
                         dialect,
                         delimiter,
                 667
                (\ldots)
                         defaults={"delimiter": ","},
                 676
                 677 )
                 678 kwds.update(kwds_defaults)
             --> 680 return _read(filepath_or_buffer, kwds)
             File ~\anaconda3\lib\site-packages\pandas\io\parsers\readers.py:575, in _
             read(filepath_or_buffer, kwds)
                 572 _validate_names(kwds.get("names", None))
                 574 # Create the parser.
             --> 575 parser = TextFileReader(filepath_or_buffer, **kwds)
                 577 if chunksize or iterator:
                 578
                         return parser
             File ~\anaconda3\lib\site-packages\pandas\io\parsers\readers.py:933, in T
             extFileReader.__init__(self, f, engine, **kwds)
                 930
                         self.options["has_index_names"] = kwds["has_index_names"]
                 932 self.handles: IOHandles | None = None
             --> 933 self._engine = self._make_engine(f, self.engine)
             File ~\anaconda3\lib\site-packages\pandas\io\parsers\readers.py:1217, in
```

```
TextFileReader._make_engine(self, f, engine)
          mode = "rb"
   1214 # error: No overload variant of "get handle" matches argument typ
   1215 # "Union[str, PathLike[str], ReadCsvBuffer[bytes], ReadCsvBuffer
[str]]"
   1216 # , "str", "bool", "Any", "Any", "Any", "Any", "Any"
-> 1217 self.handles = get_handle( # type: ignore[call-overload]
   1218
            f,
   1219
            mode,
   1220
            encoding=self.options.get("encoding", None),
            compression=self.options.get("compression", None),
   1221
   1222
            memory_map=self.options.get("memory_map", False),
   1223
            is_text=is_text,
   1224
            errors=self.options.get("encoding errors", "strict"),
            storage_options=self.options.get("storage_options", None),
   1225
   1226 )
   1227 assert self.handles is not None
   1228 f = self.handles.handle
File ~\anaconda3\lib\site-packages\pandas\io\common.py:789, in get_handle
(path_or_buf, mode, encoding, compression, memory_map, is_text, errors, s
torage_options)
    784 elif isinstance(handle, str):
            # Check whether the filename is to be opened in binary mode.
    786
            # Binary mode does not support 'encoding' and 'newline'.
            if ioargs.encoding and "b" not in ioargs.mode:
    787
                # Encoding
    788
--> 789
                handle = open(
    790
                    handle,
    791
                    ioargs.mode,
    792
                    encoding=ioargs.encoding,
    793
                    errors=errors,
    794
                    newline="",
    795
                )
            else:
    796
    797
                # Binary mode
    798
                handle = open(handle, ioargs.mode)
FileNotFoundError: [Errno 2] No such file or directory: 'Smalllajdf;ajds
f; . csv '
```

```
    grants_license = grants.merge(licenses, on = 'zip')

In [14]:
             print(grants_licenses.loc[grants_licenses["business"]== "REGGIE'S BAR & GR"
                                        ["grant", "company", "account", "ward", "busines:
             NameError
                                                        Traceback (most recent call las
             t)
             Input In [14], in <cell line: 1>()
             ----> 1 grants_license = grants.merge(licenses, on = 'zip')
                   2 print(grants_licenses.loc[grants_licenses["business"]== "REGGIE'S")
             BAR & GRILL",
                                                ["grant", "company", "account", "ward",
                   3
             "business"]])
             NameError: name 'grants' is not defined

    | grants.merge(licenses, on = ["address", "zip"])
In [15]:
             NameError
                                                        Traceback (most recent call las
             t)
             Input In [15], in <cell line: 1>()
             ----> 1 grants.merge(licenses, on = ["address", "zip"])
             NameError: name 'grants' is not defined
          | grants_license_ward = grants.merge(license, on = ["address", "zip"]) \
In [16]:
             .merge(wards, on = "ward", suffixes = ("bus", "_wards"))
               Input In [16]
                 grants_license_ward = grants.merge(license, on = ["address", "zip"])
             SyntaxError: unexpected character after line continuation character
```

```
In [17]:

    grant_license_ward. groupby('ward').agg('sum').plot(kind = 'bar', y = 'grant')

             plt.show()
             NameError
                                                        Traceback (most recent call las
             t)
             Input In [17], in <cell line: 1>()
             ----> 1 grant_license_ward. groupby('ward').agg('sum').plot(kind = 'bar',
             y = 'grant')
                   2 plt.show()
             NameError: name 'grant_license_ward' is not defined
 In [ ]:
          #Ex 4
In [18]:
             cal = pd.read_pickle("cta_calendar.p")
             ridership = pd.read_pickle("cta_ridership.p")
             stations = pd.read_pickle("stations.p")
             ridership_cal_station = ridership.merge(cal, on = [ "year", "month", "day"
             fil = ((ridership_cal_station [ "month" ] == 7)
                   & (ridership_cal_station [ 'day_type'] == 'Weekday')
                   & (ridership_cal_station [ "station_name"] == "Wilson"))
             print(ridership_cal_station.loc[fil, 'rides'].sum())
```

140005

```
In [19]:
           #Ex 5
              licenses = pd.read_pickle("licenses.p")
              wards = pd.read_pickle("ward.p")
              zip_demo = pd.read_pickle("zip_demo.p")
              licenses_zip_ward = licenses.merge(zip_demo, on = "zip").merge(wards, on =
              income = licenses_zip_ward.groupby("alderman").agg(({"income": "median"}))
    Out[19]:
                                      income
                            alderman
                        Ameya Pawar
                                      66246.0
                     Anthony A. Beale
                                      38206.0
                 Anthony V. Napolitano
                                     82226.0
                     Ariel E. Reyboras
                                      41307.0
                       Brendan Reilly
                                     110215.0
                        Brian Hopkins
                                     87143.0
                  Carlos Ramirez-Rosa
                                      66246.0
                      Carrie M. Austin
                                     38206.0
                       Chris Taliaferro
                                     55566.0
                  Daniel "Danny" Solis
                                     41226.0
In [20]:
           movies = pd.read_csv("tmdb_movies.csv")
              print(movies.head())
              print(movies.shape)
                    id
                                         title popularity release_date
              0
                   257
                                 Oliver Twist
                                                 20.415572
                                                                23/9/2005
              1
                 14290 Better Luck Tomorrow
                                                  3.877036
                                                                12/1/2002
                 38365
              2
                                    Grown Ups
                                                 38.864027
                                                                24/6/2010
                  9672
              3
                                      Infamous
                                                  3.680896
                                                              16/11/2006
              4
                12819
                              Alpha and Omega
                                                 12.300789
                                                                17/9/2010
              (4803, 4)

★ taglines = pd.read_csv('tmdb_movies.csv')

In [21]:
              print(taglines.head())
              print(taglines.shape)
                    id
                                         title
                                                popularity release_date
              0
                   257
                                 Oliver Twist
                                                  20.415572
                                                                23/9/2005
              1
                14290
                       Better Luck Tomorrow
                                                  3.877036
                                                                12/1/2002
              2
                                    Grown Ups
                                                                24/6/2010
                 38365
                                                 38.864027
              3
                  9672
                                      Infamous
                                                              16/11/2006
                                                  3.680896
              4
                 12819
                              Alpha and Omega
                                                 12.300789
                                                               17/9/2010
              (4803, 4)
```

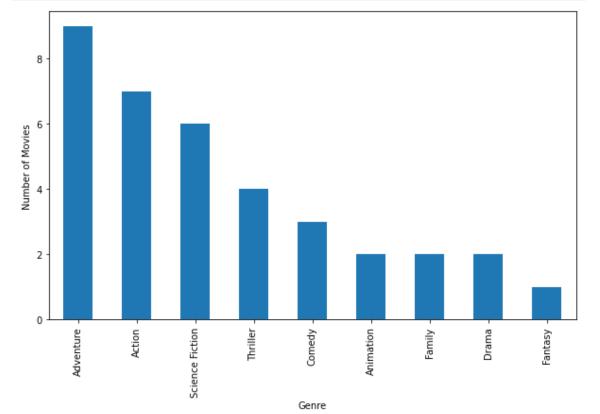
```
movies_taglines = movies.merge(taglines, on = "id", how = 'left')
In [22]:
              print(movies_taglines.head())
                    id
                                               popularity_x release_date_x
                                      title_x
              0
                   257
                                 Oliver Twist
                                                   20.415572
                                                                   23/9/2005
              1
                 14290 Better Luck Tomorrow
                                                    3.877036
                                                                   12/1/2002
              2
                38365
                                    Grown Ups
                                                   38.864027
                                                                   24/6/2010
              3
                  9672
                                     Infamous
                                                    3.680896
                                                                  16/11/2006
              4
                 12819
                             Alpha and Omega
                                                   12.300789
                                                                   17/9/2010
                               title_y
                                        popularity_y release_date_y
              0
                         Oliver Twist
                                           20.415572
                                                            23/9/2005
              1
                 Better Luck Tomorrow
                                             3.877036
                                                            12/1/2002
              2
                             Grown Ups
                                            38.864027
                                                            24/6/2010
              3
                              Infamous
                                             3.680896
                                                          16/11/2006
              4
                      Alpha and Omega
                                           12.300789
                                                            17/9/2010
In [23]:
           ▶ print(movies_taglines.shape)
              (4803, 7)
In [24]:
              #Ex 6
              movies = pd.read_pickle("movies.p")
              financials = pd.read_pickle("financials.p")
              movies_financials = movies.merge(financials, on = "id", how = "left")
              movies financials["budget"].isnull().sum()
    Out[24]: 1574
In [25]:
           #Ex 7
              toy_story = pd.read_csv("toy_story.csv")
              taglines = pd.read_pickle("taglines.p")
              toy_story_tag = toy_story.merge(taglines, on = "id", how="left")
              #toy_story_tag = toy_story.merge(taglines, on = "id")
              toy_story_tag
    Out[25]:
                    id
                             title
                                 popularity release_date
                                                                  tagline
              0 10193 Toy Story 3
                                    59.995
                                              16/6/2010 No toy gets left behind.
              1
                   863
                       Toy Story 2
                                    73.575
                                             30/10/1999
                                                          The toys are back!
              2
                                             30/10/1995
                   862
                         Toy Story
                                    73.640
                                                                    NaN
```

```
movie_to_genres = pd.read_csv('tdmb_movie_to_genres.csv')
In [26]:
             tv_genre = movie_to_genres[movie_to_genres['genre'] == 'TV Movie']
             print(tv_genre)
                    movie_id
                                 genre
             4998
                       10947 TV Movie
             5994
                       13187 TV Movie
             7443
                       22488 TV Movie
             10061
                       78814 TV Movie
             10790
                      153397 TV Movie
             10835
                      158150 TV Movie
                      205321 TV Movie
             11096
             11282
                      231617 TV Movie
In [27]:
          tv_movies = movies.merge(tv_genre, how = 'right', left_on = 'id', right_on
             print(tv_movies.head())
                    id
                                            title popularity release_date
                                                                             movie_id
             \
                 10947
                              High School Musical
                                                    16.536374
                                                                 2006-01-20
                                                                                10947
             1
                 13187 A Charlie Brown Christmas
                                                     8.701183
                                                                 1965-12-09
                                                                                13187
             2
                 22488
                               Love's Abiding Joy
                                                     1.128559
                                                                 2006-10-06
                                                                                22488
             3
                             We Have Your Husband
                 78814
                                                     0.102003
                                                                 2011-11-12
                                                                                78814
                153397
                                         Restless
                                                     0.812776
                                                                 2012-12-07
                                                                               153397
                   genre
             0 TV Movie
             1
               TV Movie
             2 TV Movie
             3 TV Movie
             4 TV Movie
In [28]:
          | m = movie_to_genres['genre'] == "Family"
             family = movie_to_genres[m].head(3)
             family
             m = movie_to_genres['genre'] == "Comedy"
             comedy = movie_to_genres[m].head(3)
             comedy
   Out[28]:
                 movie_id
                           genre
                       5 Comedy
               1
               7
                      13
                         Comedy
              35
                      35 Comedy
```

```
| family_comedy = family.merge(comedy, on = "movie_id", how = "outer", suffi
In [29]:
             print(family_comedy)
                movie_id genre_fam genre_com
             0
                             Family
                      12
             1
                      35
                             Family
                                       Comedy
             2
                     105
                             Family
                                          NaN
             3
                       5
                                NaN
                                       Comedy
             4
                      13
                                NaN
                                       Comedy
In [30]:
          #Ex 8
             import pandas as pd
             # Load data
             movies = pd.read_pickle('movies.p')
             # Subset scifi_movies and action_movies
             scifi_movies = movie_to_genres[movie_to_genres['genre'] == 'Science Fiction')
             action_movies = movie_to_genres[movie_to_genres['genre'] == 'Action']
             # Merge action_movies and scifi_movies with a right join and add suffixes
             action_scifi = pd.merge(action_movies, scifi_movies, on='movie_id', how='r
             # Subset rows where genre_act column is null (only science fiction)
             scifi_only = action_scifi[action_scifi['genre_act'].isnull()]
             # Merge movies and scifi_only with an inner join
             result = pd.merge(movies, scifi_only, left_on='id', right_on='movie_id', he
             # Display the result
             print(result[['id', 'title']])
                      id
                                                  title
             0
                   18841
                          The Lost Skeleton of Cadavra
             1
                   26672
                              The Thief and the Cobbler
             2
                   15301
                               Twilight Zone: The Movie
             3
                    8452
                                            The 6th Day
                            Bill & Ted's Bogus Journey
             4
                    1649
                     . . .
                  245703
                                       Midnight Special
             253
             254
                                       A Scanner Darkly
                    3509
             255
                                        Never Let Me Go
                   42188
             256
                   18045
                                         The Dark Hours
             257
                                                Godsend
                   11058
```

[258 rows x 2 columns]

```
In [31]:
          #Ex 9
             import pandas as pd
             import matplotlib.pyplot as plt
             # Load data from CSV files
             pop_movies = pd.read_csv('pop_movies.csv')
             movie_to_genres = pd.read_csv('tdmb_movie_to_genres.csv')
             # Merge using right join
             genres_movies = pd.merge(movie_to_genres, pop_movies, left_on='movie_id',
             # Group by genre and count the number of movies
             genre_counts = genres_movies['genre'].value_counts().head(10)
             # Plot the bar chart
             plt.figure(figsize=(10, 6))
             genre_counts.plot(kind='bar')
             plt.xlabel('Genre')
             plt.ylabel('Number of Movies')
             plt.show()
```



	id_org	title_org					sequel_	org	
\									
0	862	Toy Story 86							
1	863				Toy Stor	y 2	10	193	
2	675	Harry	Potter and	the Orde	r of the Phoe	nix		767	
3	121	Т	he Lord of t	he Rings	: The Two Tow	ers		122	
4	120	The Lord of t	he Rings: Th	e Fellow	ship of the R	ing		121	
	id_seq				title_seq	sequ	el_seq		
0	863				Toy Story 2		10193		
1	10193	Toy Story 3 <na></na>							
2	767	Harry Potter and the Half-Blood Prince <na></na>							
3	122	The Lord of the Rings: The Return of the King <na></na>							
4	121	The Lord of the Rings: The Two Towers 122							
	id_org	title_org	sequel_org	id_seq	title_seq	sequ	el_seq		
0	19995	Avatar	<na></na>	<na></na>	NaN		<na></na>		
1	862	Toy Story	863	863	Toy Story 2		10193		
2	863	Toy Story 2	10193	10193	Toy Story 3		<na></na>		
3	597	Titanic	<na></na>	<na></na>	NaN		<na></na>		

<NA>

<NA>

NaN

<NA>

24428 The Avengers

```
In [33]:
          #Ex 10
             import pandas as pd
             # Load the crews table
             crews = pd.read_pickle('crews.p')
             crews_self_merged = pd.merge(crews, crews, on='id', how='inner', suffixes=
             boolean_filter = ((crews_self_merged['job_dir'] == 'Director') &
                               (crews_self_merged['job_crew'] != 'Director'))
             direct_crews = crews_self_merged[boolean_filter]
             print(direct_crews.head())
                     id department_dir    job_dir
                                                      name_dir department_crew \
             156 19995
                             Directing Director James Cameron
                                                                       Editing
             157 19995
                            Directing Director James Cameron
                                                                         Sound
             158 19995
                            Directing Director James Cameron
                                                                    Production
                             Directing Director James Cameron
             160 19995
                                                                       Writing
             161 19995
                            Directing Director James Cameron
                                                                           Art
                        job_crew
                                          name_crew
             156
                          Editor Stephen E. Rivkin
             157 Sound Designer Christopher Boyes
             158
                         Casting
                                         Mali Finn
             160
                          Writer
                                      James Cameron
             161
                                   Richard F. Mays
                    Set Designer
In [52]:
          import pandas as pd
             pd.concat([inv_jan, inv_feb, inv_mar], ignor_index =True, keys=['jan', 'fel
             NameError
                                                      Traceback (most recent call las
             t)
             Input In [52], in <cell line: 2>()
                   1 import pandas as pd
             ----> 2 pd.concat([inv_jan, inv_feb, inv_mar], ignor_index =True, keys=['
             jan', 'feb', 'mar'])
             NameError: name 'inv_jan' is not defined
```

```
#Ex11
In [54]:
              tracks_master = pd.read_csv("tracks_master.csv")
              tracks_ride = pd.read_csv("tracks_ride.csv")
              tracks_st = pd.read_csv("tracks_st.csv")
              print(tracks_master)
              print()
              print()
              print(tracks_ride)
              print()
              print(tracks_st)
              pd.concat([tracks_master, tracks_ride, tracks_st], sort = True, keys = ['n
                   tid
                                                   mtid
                                                          gid
                                       name
                                              aid
                                                                            composer
                                                                                       u_price
                                    Battery
              0
                  1853
                                              152
                                                       1
                                                            3
                                                                J.Hetfield/L.Ulrich
                                                                                           0.99
                  1854
                        Master Of Puppets
                                                            3
                                                                                           0.99
              1
                                              152
                                                       1
                                                                           K.Hammett
                        Disposable Heroes
              2
                  1857
                                              152
                                                       1
                                                            3
                                                                J.Hetfield/L.Ulrich
                                                                                           0.99
                   tid
                                              name
                                                    aid
                                                          mtid
                                                                 gid
                                                                      u_price
              0
                 1874
                            Fight Fire With Fire
                                                    154
                                                              1
                                                                   3
                                                                          0.99
              1
                  1875
                              Ride The Lightning
                                                    154
                                                              1
                                                                   3
                                                                          0.99
              2
                  1876
                        For Whom The Bell Tolls
                                                    154
                                                              1
                                                                   3
                                                                          0.99
              3
                                    Fade To Black
                                                    154
                                                              1
                                                                   3
                                                                          0.99
                  1877
              4
                  1878
                               Trapped Under Ice
                                                    154
                                                              1
                                                                   3
                                                                          0.99
                   tid
                                           name
                                                 aid
                                                      mtid
                                                              gid
                                                                   u_price
              0
                  1882
                                       Frantic
                                                 155
                                                                3
                                                          1
                                                                      0.99
              1
                 1883
                                                                3
                                                                      0.99
                                     St. Anger
                                                 155
                                                          1
                 1884
              2
                        Some Kind Of Monster
                                                 155
                                                          1
                                                                3
                                                                      0.99
                                                                3
                                                                      0.99
              3
                 1885
                                 Dirty Window
                                                 155
                                                          1
              4
                  1886
                                Invisible Kid
                                                 155
                                                          1
                                                                3
                                                                      0.99
    Out[54]:
                         aid
                                  composer gid mtid
                                                                     name
                                                                             tid u_price
                     0 152
                             J.Hetfield/L.Ulrich
                                              3
                                                    1
                                                                           1853
                                                                                    0.99
               name
                                                                    Battery
                        152
                                  K.Hammett
                                              3
                                                    1
                                                           Master Of Puppets
                                                                           1854
                                                                                    0.99
                      2 152 J.Hetfield/L.Ulrich
                                              3
                                                    1
                                                           Disposable Heroes
                                                                           1857
                                                                                    0.99
                 aid 0 154
                                       NaN
                                              3
                                                    1
                                                          Fight Fire With Fire
                                                                           1874
                                                                                    0.99
                        154
                                       NaN
                                              3
                                                    1
                                                           Ride The Lightning
                                                                           1875
                                                                                    0.99
                      1
                                                      For Whom The Bell Tolls
                      2 154
                                       NaN
                                              3
                                                                          1876
                                                                                    0.99
                       154
                                       NaN
                                              3
                                                    1
                                                              Fade To Black 1877
                                                                                    0.99
```

3

3

1

1

Trapped Under Ice

1878

Frantic 1882

0.99

0.99

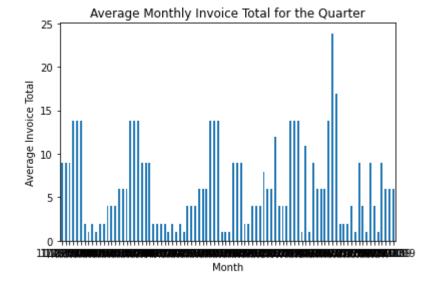
NaN

NaN

154

mtid 0 155

	aid	composer	gid	mtid	name	tid	u_price
1	155	NaN	3	1	St. Anger	1883	0.99
2	155	NaN	3	1	Some Kind Of Monster	1884	0.99
_	4	A. A.	^	4	D: ()A!: 1	1005	2 22



```
In [56]:
          pd.merge_ordered(appl, mcd, on = 'date', suffixes = ('_appl', '_mcd'))
             pd.merge_ordered(appl, mcd, on = 'date', suffixes = ('_appl', '_mcd'), fill
             NameError
                                                      Traceback (most recent call las
             t)
             Input In [56], in <cell line: 2>()
                   1 import pandas as pd
             ----> 2 pd.merge_ordered(appl, mcd, on = 'date', suffixes = ('_appl', '_m
             cd'))
                   3 pd.merge_ordered(appl, mcd, on = 'date', suffixes = ('_appl', '_m
             cd'), fill_method = 'ffill')
             NameError: name 'appl' is not defined
In [69]:
          #Ex 13
             import pandas as pd
             sp500 = pd.read_csv("S&P500.csv")
             gdp = pd.read_csv("GDP.csv")
             gdp_sp500 = pd.merge_ordered(gdp, sp500, left_on='year', right_on='date', |
             print(gdp_sp500[gdp_sp500['year'] == 2018])
             gdp_sp500 = pd.merge_ordered(gdp, sp500, left_on='year', right_on='date', |
             gdp_returns = gdp_sp500[['gdp', 'returns']]
             correlation_matrix = gdp_returns.corr()
             print(correlation_matrix)
                Unnamed: 0 country code year
                                                            date
                                                                  returns
                                                       gdp
             9
                                   USA 2018 2.050000e+13
                                                             NaN
                                                                      NaN
                           gdp
                                returns
                      1.000000 0.220321
             gdp
             returns 0.220321 1.000000
```

9

```
import pandas as pd
import matplotlib.pyplot as plt

unemployment = pd.read_csv("unemployment.csv")
inflation = pd.read_csv("inflation.csv")

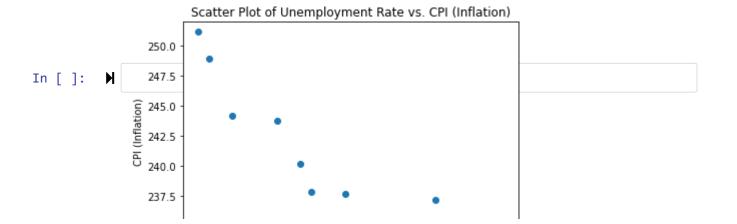
inflation_unemploy = pd.merge_ordered(inflation, unemployment, on='date'
print(inflation_unemploy)

plt.scatter(inflation_unemploy['unemployment_rate'], inflation_unemploy['cplt.xlabel('Unemployment Rate')
plt.ylabel('CPI (Inflation)')
plt.title('Scatter Plot of Unemployment Rate vs. CPI (Inflation)')
plt.show()
```

```
date
                cpi
                        seriesid
                                                 data_type \
0 1/1/2014 235.288 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
1 1/1/2015 234.718 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
2 1/1/2016 237.833 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
3 1/1/2017 243.780 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
4 1/1/2018 248.884 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
5 1/6/2014 237.231 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
6 1/6/2015 237.684 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
7 1/6/2016 240.167 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
8 1/6/2017 244.182 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
  1/6/2018 251.134 CUSR0000SA0 SEASONALLY ADJUSTED INDEX
   unemployment_rate
0
                6.7
1
                5.6
2
                5.0
3
                4.7
4
                4.1
5
                6.1
6
                5.3
7
                4.9
8
                4.3
```

23 of 24 8/25/2023, 3:14 PM

4.0



24 of 24