```
In [160... import pandas as pd import matplotlib.pyplot as plt import numpy as np import seaborn as sns
```

1. Merge taxi_owners with taxi_veh on the column vid, and save the result to taxi_own_veh. Set the left and right table suffixes for overlapping columns of the merge to _own and _veh, respectively. Select the fuel_type column from taxi_own_veh and print the value_counts() to find the most popular fuel_types used. Answer: Hybrid

```
In [161... taxi_owners=pd.read_pickle("taxi_owners.p")
         taxi_veh=pd.read_pickle("taxi_vehicles.p")
In [162... print(taxi owners.value counts(),"\n\n")
         print(taxi_veh.value_counts())
         rid
                vid
                      owner
                                                      address
                                                                              zip
                                                                              60618
         Т1
                      YC1 LLC
                                                      3351 W. ADDISON ST.
                1
         T5162
                5162 VERONIQUE ARIELLE CAB CORP.
                                                      2617 S. WABASH AVE.
                                                                              60616
         1
         T5140
                5140 IRA C BERGER
                                                      6311 N. LAWNDALE AVE.
                                                                              60659
         1
         T5145 5145 HHH & I COMPANY
                                                      3351 W. ADDISON ST.
                                                                              60618
         1
         T5146 5146 AMR CAB CORP.
                                                      3351 W. ADDISON AVE.
                                                                              60618
         T3015
                3015
                      TANAKA TRANS CORP.
                                                      4626 W. CORNELIA AVE.
                                                                              60641
         T3016
                3016 DRAKE GLOBAL ENTERPRISES INC.
                                                      2945 W. PETERSON AVE.
                                                                              60659
         1
                3018 DERLY'S CAB CO., INC.
                                                      4626 W. CORNELIA AVE.
         T3018
                                                                              60641
         1
         T3019
                3019
                      HERETIC INC
                                                      3351 W. ADDISON ST.
                                                                              60618
         1
                997
                      PLAYING POLO IN CHICAGO INC
         T997
                                                      2617 S. WABASH AVE.
                                                                              60616
         Length: 3519, dtype: int64
         vid
               make
                       model
                                year
                                      fuel type
                                                 owner
         1
               TOYOTA
                       CAMRY
                                2014
                                     HYBRID
                                                 YC1 LLC
                                                                                   1
                                                 VERONIQUE ARIELLE CAB CORP.
         5162
               FORD
                       ESCAPE
                                2012
                                     HYBRID
                                                                                   1
         5140
               FORD
                       FUSION
                                2012
                                      HYBRID
                                                 IRA C BERGER
                                                                                   1
         5145
                                2012
               TOYOTA
                       CAMRY
                                     HYBRID
                                                 HHH & I COMPANY
                                                                                   1
         5146
               TOYOTA PRIUS
                                2013
                                      HYBRID
                                                 AMR CAB CORP.
                                                                                   1
         3015
               TOYOTA CAMRY
                                2014
                                                 TANAKA TRANS CORP.
                                                                                   1
                                      HYBRID
         3016
               TOYOTA SIENNA 2019
                                      GASOLINE
                                                 DRAKE GLOBAL ENTERPRISES INC.
                                                                                   1
         3018
               TOYOTA
                       CAMRY
                                2016
                                      HYBRID
                                                 DERLY'S CAB CO., INC.
                                                                                   1
         3019
               TOYOTA CAMRY
                                2010
                                     HYBRID
                                                                                   1
                                                 HERETIC INC
         997
               FORD
                       ESCAPE 2012
                                      HYBRID
                                                 PLAYING POLO IN CHICAGO INC
                                                                                   1
         Length: 3519, dtype: int64
```

```
In [163...
         taxi own veh=taxi veh.merge(taxi owners,on="vid")
         print(taxi_own_veh.head())
         print("\n\n======\n\n")
         print(taxi_own_veh.columns) #note that there is two duplicate coulmn --own
             vid
                          model year fuel_type
                                                            owner_x
                                                                       rid
            2767
                 TOYOTA
                          CAMRY
                                 2013
                                         HYBRID
                                                     SEYED M. BADRI
                                                                    T2767
         1
            1411
                 TOYOTA
                           RAV4 2017
                                        HYBRID
                                                        DESZY CORP.
                                                                    T1411
            6500 NISSAN SENTRA 2019 GASOLINE
         2
                                                     AGAPH CAB CORP T6500
         3 2746 TOYOTA
                          CAMRY 2013 HYBRID MIDWEST CAB CO, INC T2746
            5922 TOYOTA
                          CAMRY 2013
                                        HYBRID
                                                     SUMETTI CAB CO T5922
                                             address
                       owner y
                                                        zip
         0
                 SEYED M. BADRI
                                  510 W. BELMONT AVE.
                                                      60657
         1
                                  3351 W. ADDISON ST.
                   DESZY CORP.
                                                      60618
                AGAPH CAB CORP 2945 W. PETERSON AVE.
                                                      60659
         3 MIDWEST CAB CO, INC 4020 W. GLENLAKE AVE.
                                                      60646
                SUMETTI CAB CO
                                    6550 N. CLARK ST.
                                                      60626
         Index(['vid', 'make', 'model', 'year', 'fuel_type', 'owner_x', 'rid',
                owner_y', 'address', 'zip'],
               dtype='object')
In [164...
        taxi own veh=taxi veh.merge(taxi owners,on="vid",suffixes=(" own"," veh"))
         print(taxi_own_veh.head())
             vid
                   make
                          model year fuel type
                                                          owner own
                                                                       rid \
            2767 TOYOTA
                                                     SEYED M. BADRI T2767
                          CAMRY 2013
                                        HYBRID
                                                        DESZY CORP.
           1411 TOYOTA
         1
                          RAV4 2017
                                        HYBRID
                                                                     Т1411
                                                     AGAPH CAB CORP T6500
           6500 NISSAN SENTRA 2019 GASOLINE
         3
            2746
                 TOYOTA
                          CAMRY
                                 2013
                                        HYBRID MIDWEST CAB CO, INC
                                                                    T2746
            5922 TOYOTA
                                                     SUMETTI CAB CO T5922
                          CAMRY 2013
                                        HYBRID
                     owner_veh
                                             address
                                                        zip
         0
                 SEYED M. BADRI
                                  510 W. BELMONT AVE.
                                                      60657
         1
                   DESZY CORP.
                                  3351 W. ADDISON ST.
                                                      60618
         2
                AGAPH CAB CORP 2945 W. PETERSON AVE.
                                                      60659
         3
           MIDWEST CAB CO, INC 4020 W. GLENLAKE AVE.
                                                      60646
                SUMETTI CAB CO
                                    6550 N. CLARK ST.
                                                      60626
In [165...
         print(taxi_own_veh['fuel_type'].value_counts())
         HYBRID
                                  2792
         GASOLINE
                                   611
         FLEX FUEL
                                    89
         COMPRESSED NATURAL GAS
         Name: fuel_type, dtype: int64
```

1. Wards_Offices_Altered.csv is available which contains only 46 wards. Load this .csv to wards_altered and merge it with census on the column ward. Save the result to wards_census_altered and observe the number of rows.

(46, 4)
(50, 6)

```
In [167... wards_cenus=wards.merge(cenus,on="ward")
         print(wards cenus.head())
         print("======Total number of rows and columns in wards cenus =", wards
            ward
                             alderman
                                                              address_x zip_x pop_200
         0
                  Proco "Joe" Moreno
                                             2058 NORTH WESTERN AVENUE
         0
                                                                         60647
                                                                                   5295
         1
         1
                        Brian Hopkins
                                            1400 NORTH ASHLAND AVENUE
                                                                         60622
                2
                                                                                   5436
         1
                                               5046 SOUTH STATE STREET
         2
                3
                           Pat Dowell
                                                                         60609
                                                                                   4038
         5
         3
                     William D. Burns 435 EAST 35TH STREET, 1ST FLOOR
                                                                         60616
                                                                                   5195
         3
                  Leslie A. Hairston
                                                                         60649
         4
                                                 2325 EAST 71ST STREET
                                                                                   5530
         2
            pop_2010 change
                                                                        zip_y
                                                             address_y
         0
                56149
                                          2765 WEST SAINT MARY STREET
                                                                        60647
                          6%
         1
                55805
                          3%
                                             WM WASTE MANAGEMENT 1500
                                                                        60622
         2
                53039
                         31%
                                                  17 EAST 38TH STREET
                                                                        60653
         3
                         5%
                             31ST ST HARBOR BUILDING LAKEFRONT TRAIL
                54589
                         -7%
                             JACKSON PARK LAGOON SOUTH CORNELL DRIVE
                                                                        60637
                51455
         =====Total number of rows and columns in wards cenus = (46, 9) =====
In [168...
         wards cenus altered=cenus.merge(wards,on="ward")
         print(wards cenus altered.head())
         print("======Total number of rows and columns in wards cenus =",wards
            ward pop_2000 pop_2010 change
                                                                             address x
         \
         0
               1
                      52951
                                56149
                                          6%
                                                          2765 WEST SAINT MARY STREET
         1
                2
                      54361
                                55805
                                          3%
                                                              WM WASTE MANAGEMENT 1500
         2
                3
                      40385
                                53039
                                         31%
                                                                   17 EAST 38TH STREET
         3
                4
                      51953
                                54589
                                          5%
                                              31ST ST HARBOR BUILDING LAKEFRONT TRAIL
         4
                5
                      55302
                                51455
                                         -7%
                                              JACKSON PARK LAGOON SOUTH CORNELL DRIVE
                              alderman
            zip_x
                                                               address_y
                                                                          zip_y
                   Proco "Joe" Moreno
         0
            60647
                                              2058 NORTH WESTERN AVENUE
                                                                          60647
         1
            60622
                         Brian Hopkins
                                             1400 NORTH ASHLAND AVENUE
                                                                          60622
         2
            60653
                            Pat Dowell
                                                5046 SOUTH STATE STREET
                                                                          60609
         3
            60653
                     William D. Burns 435 EAST 35TH STREET, 1ST FLOOR
                                                                          60616
                   Leslie A. Hairston
                                                  2325 EAST 71ST STREET
                                                                          60649
         =====Total number of rows and columns in wards cenus = (46, 9) =====
           -----
         lincenses=pd.read_csv("Business_Licenses.csv")
In [169...
         print( lincenses.head())
         print(lincenses.shape)
```

```
account ward aid
                                       business
                                                               address
0
   307071
             3
                 743
                           REGGIE'S BAR & GRILL
                                                      2105 S STATE ST
1
       10
             10 829
                                     HONEYBEERS 13200 S HOUSTON AVE
2
    10002
             14
                 775
                                                    5089 S ARCHER AVE
                                     CELINA DELI
3
    10005
             12 NaN
                      KRAFT FOODS NORTH AMERICA
                                                       2005 W 43RD ST
4
    10044
             44
                 638
                      NEYBOUR'S TAVERN & GRILLE 3651 N SOUTHPORT AVE
       zip
0
  60616.0
1
  60633.0
2
  60632.0
  60609.0
4
  60613.0
(10000, 6)
```

In [170... ward_lincenses=wards.merge(lincenses,on="ward",suffixes=("_ward","_lic")) print(ward_lincenses)

	ward		a	lderman			addre	ss_ward	zip_ward	accoun
t \ 0	1	Proco	"Joe"	Moreno	2058	NORTI	H WESTERN	AVENUE	60647	1202
4 1 6	1	Proco	"Joe"	Moreno	2058	NORTI	H WESTERN	AVENUE	60647	1444
2	1	Proco	"Joe"	Moreno	2058	NORTI	H WESTERN	AVENUE	60647	1462
3 7	1	Proco	"Joe"	Moreno	2058	NORTI	H WESTERN	AVENUE	60647	1498
4 2	1	Proco	"Joe"	Moreno	2058	NORTH	H WESTERN	AVENUE	60647	1564
• • •	• • •			• • •				• • •	•••	
9285 7	46	Jam	es Ca	ppleman	4544	NORTH	BROADWAY	AVENUE	60640	6478
9286 4	46	Jam	es Ca	ppleman	4544	NORTH	BROADWAY	AVENUE	60640	676
9287 7	46	Jam	es Ca	ppleman	4544	NORTH	BROADWAY	AVENUE	60640	678
9288 9	46	Jam	es Ca	ppleman	4544	NORTH	BROADWAY	AVENUE	60640	832
9289 2	46	Jam	es Ca	ppleman	4544	NORTH	BROADWAY	AVENUE	60640	861
	aid				k	ousines	ss		address_lic	: \
0	NaN						CS		ASHLAND AVE	
1 2	743 775								ERN AVE 157 IFORNIA AVE	
3	NaN				ROWN'S				AGO AVE 187	
4	814					Kitche			DIVISION ST	
							•		• • •	•
9285		CUPID'S		SURES OF					HALSTED ST	
9286 9287	775 775		THO.		Bar o		AL 850 I		F PARK RD 1 W BUENA AVE	
9288	774				MARKE				ROVE AVE 1	
9289	775		A	-1 DONUT	& SNA	CK SHO)P	3938 N	SHERIDAN RI)
	zip_l	ic								
0	60622	.0								
1	60622									
2	60647									
3 4	60622 60622									
• • •		••								
9285	60657	.0								
9286	60613									
9287 9288	60613 60613									
9289	60613									
[9290	[9290 rows x 9 columns]									

1. Complete the following tasks. • Starting with the licenses table on the left, merge it to the biz_owners table on the column account, and save the results to a variable named licenses_owners. • Group licenses_owners by title and count the number of accounts for each title using .agg({'account':'count'}). Save the result as counted_df • Sort counted_df by the number of accounts in descending order, and save this as a variable named sorted_df. • Use the .head() method to print the first few rows of the sorted_df.

```
In [171...
          biz owners=pd.read pickle("business owners.p")
          lincenses=pd.read_pickle("licenses.p")
          print(biz owners)
          print(lincenses)
                 account first_name
                                     last_name
                                                           title
          0
                      10
                              PEARL
                                        SHERMAN
                                                       PRESIDENT
          1
                      10
                              PEARL
                                        SHERMAN
                                                       SECRETARY
          2
                   10002
                             WALTER
                                         MROZEK
                                                         PARTNER
          3
                   10002
                             CELINA
                                         BYRDAK
                                                         PARTNER
          4
                   10005
                              IRENE ROSENFELD
                                                       PRESIDENT
          21347
                    9513
                             THOMAS
                                         MAHLUM
                                                       SECRETARY
          21348
                    9563
                          GUADALUPE
                                          PEREZ
                                                       PRESIDENT
                    9638
          21349
                             STEVEN
                                         JENSEN VICE PRESIDENT
          21350
                    9922
                             MANUEL
                                        SALAZAR
                                                       SECRETARY
          21351
                    9933
                            MICHAEL
                                           KNAB
                                                       PRESIDENT
          [21352 rows x 4 columns]
               account ward
                              aid
                                                      business
                                                                                      addres
          s
          0
                 307071
                           3
                              743
                                         REGGIE'S BAR & GRILL
                                                                             2105 S STATE S
          т
                                                                         13200 S HOUSTON AV
          1
                     10
                          10
                              829
                                                    HONEYBEERS
          Е
          2
                 10002
                          14
                              775
                                                   CELINA DELI
                                                                           5089 S ARCHER AV
          Е
          3
                 10005
                                    KRAFT FOODS NORTH AMERICA
                                                                              2005 W 43RD S
                          12
                              NaN
          Т
                                    NEYBOUR'S TAVERN & GRILLE
                  10044
                                                                       3651 N SOUTHPORT AV
          4
                              638
                          44
          Е
                               . . .
          . . .
          9995
                   8634
                                                    J M V CORP
                                                                              2717 E 95TH S
                          10
                              NaN
          Т
          9996
                  86350
                          42
                              708
                                          AMERICASH LOANS LLC
                                                                          103 N WELLS ST 1S
          Т
          9997
                  86352
                          23
                              775
                                                   MC DONALD'S
                                                                       6720 W ARCHER AVE
          1
          9998
                  86355
                          43
                              638
                                         VOSGES HAUT-CHOCOLAT
                                                                       951 W ARMITAGE AVE
          1
          9999
                                      ALMUFLIHI FOOD & LIQUOR 5400-5404 S HALSTED ST 1S
                  86373
                          20
                              775
          Т
                   zip
          0
                 60616
          1
                 60633
          2
                 60632
          3
                 60609
          4
                 60613
                   . . .
          9995
                60617
          9996
                60606
          9997
                60638
          9998
                60614
          9999
                60609
          [10000 rows x 6 columns]
In [172...
          lincenses_onwers=lincenses.merge(biz_owners,on="account")
          counted_df=lincenses_onwers.groupby("title").agg({"account":"count"})
In [173...
          counted df
```

Out [173]: account

title	
ASST. SECRETARY	111
BENEFICIARY	4
CEO	110
DIRECTOR	146
EXECUTIVE DIRECTOR	10
GENERAL PARTNER	21
INDIVIDUAL	268
LIMITED PARTNER	26
MANAGER	134
MANAGING MEMBER	878
MEMBER	884
NOT APPLICABLE	11
OTHER	1200
PARTNER	451
PRESIDENT	6259
PRINCIPAL OFFICER	63
SECRETARY	5205
SHAREHOLDER	590
SOLE PROPRIETOR	1658
SPOUSE	34
TREASURER	447
TRUSTEE	6
VICE PRESIDENT	970

```
In [174... sorted_df=counted_df.sort_values("account",ascending=False)
    print(sorted_df.head())
```

```
account
title
PRESIDENT 6259
SECRETARY 5205
SOLE PROPRIETOR 1658
OTHER 1200
VICE PRESIDENT 970
```

1. Complete the following tasks. • Merge the ridership and cal tables together, starting with the ridership table on the left and save the result to the variable ridership_cal. • Extend the previous merge to three tables by also merging the stations table. • Create a variable called filter_criteria to select the appropriate rows from the merged table so that you can sum the rides column.

```
In [175...
          cal=pd.read pickle("cta calendar.p")
          ridership=pd.read_pickle("cta_ridership.p")
          stations=pd.read pickle("stations.p")
In [176... print(cal.head())
          print(ridership.head())
          print(stations.head())
             year month
                          day
                                      day_type
          0
             2019
                       1
                            1
                               Sunday/Holiday
          1
            2019
                       1
                            2
                                      Weekday
          2
            2019
                       1
                            3
                                      Weekday
          3 2019
                       1
                            4
                                      Weekday
            2019
                       1
                            5
                                      Saturday
            station_id year month day rides
          0
                 40010 2019
                                             576
                               1
                                       1
          1
                 40010 2019
                                  1
                                        2
                                            1457
          2
                 40010 2019
                                        3
                                            1543
                                  1
          3
                 40010 2019
                                  1
                                        4
                                            1621
                 40010 2019
                                  1
                                        5
                                             719
            station_id
                              station_name
                                                            location
          0
                 40010 Austin-Forest Park (41.870851, -87.776812)
          1
                 40020
                               Harlem-Lake (41.886848, -87.803176)
          2
                 40030
                              Pulaski-Lake (41.885412, -87.725404)
                                              (41.878723, -87.63374)
          3
                 40040
                              Quincy/Wells
                 40050
                                      Davis
                                              (42.04771, -87.683543)
In [177...
         #ridership cal=ridership.merge(cal)
          #merging the ridersip ,cal and stations tables
          ridership_cal_stations=ridership.merge(cal,on=["year","month","day"]).merge(
          print(ridership_cal_stations)
               station_id year month day
                                              rides
                                                           day_type
                                                                            station_name
          \
          0
                    40010
                           2019
                                      1
                                           1
                                                576
                                                     Sunday/Holiday
                                                                      Austin-Forest Park
                    40010
                           2019
                                      1
                                           2
          1
                                               1457
                                                            Weekday
                                                                     Austin-Forest Park
          2
                    40010
                           2019
                                      1
                                           3
                                               1543
                                                            Weekday
                                                                     Austin-Forest Park
          3
                    40010
                           2019
                                      1
                                           4
                                               1621
                                                            Weekday Austin-Forest Park
                           2019
                                           5
                                                           Saturday Austin-Forest Park
          4
                    40010
                                      1
                                                719
                      . . .
                            . . .
                                    . . .
                                         . . .
                                                . . .
                                                                 . . .
          3280
                    41660
                           2019
                                    12
                                          27
                                              13898
                                                            Weekday
                                                                              Lake/State
          3281
                    41660
                           2019
                                    12
                                          28
                                               9485
                                                           Saturday
                                                                              Lake/State
          3282
                    41660 2019
                                     12
                                          29
                                              7581
                                                     Sunday/Holiday
                                                                              Lake/State
          3283
                    41660 2019
                                     12
                                          30 15332
                                                            Weekday
                                                                              Lake/State
          3284
                                     12
                    41660 2019
                                          31 13430
                                                            Weekday
                                                                              Lake/State
                               location
          0
                (41.870851, -87.776812)
          1
                (41.870851, -87.776812)
          2
                (41.870851, -87.776812)
                (41.870851, -87.776812)
          3
          4
                (41.870851, -87.776812)
          . . .
          3280 (41.884809, -87.627813)
          3281
                (41.884809, -87.627813)
          3282
                (41.884809, -87.627813)
         3283
                (41.884809, -87.627813)
                (41.884809, -87.627813)
          3284
          [3285 rows x 8 columns]
In [178...
         ridership cal stations=ridership.merge(cal,on=["year","month","day"]).merge(
          print(ridership cal stations)
```

140005

```
station id
                           year month
                                         day
                                              rides
                                                            day_type
                                                                             station name
          \
          0
                    40010
                           2019
                                      1
                                           1
                                                576
                                                      Sunday/Holiday
                                                                      Austin-Forest Park
          1
                    40010
                           2019
                                      1
                                           2
                                               1457
                                                             Weekday Austin-Forest Park
          2
                    40010
                           2019
                                      1
                                           3
                                               1543
                                                             Weekday Austin-Forest Park
          3
                    40010
                           2019
                                      1
                                           4
                                               1621
                                                             Weekday Austin-Forest Park
          4
                    40010
                           2019
                                      1
                                           5
                                                719
                                                            Saturday
                                                                      Austin-Forest Park
                      . . .
                            . . .
                                         . . .
          3280
                    41660
                           2019
                                     12
                                          27
                                              13898
                                                             Weekday
                                                                               Lake/State
          3281
                    41660
                           2019
                                     12
                                          28
                                               9485
                                                            Saturday
                                                                               Lake/State
          3282
                    41660
                           2019
                                     12
                                          29
                                               7581 Sunday/Holiday
                                                                               Lake/State
          3283
                    41660
                           2019
                                     12
                                          30 15332
                                                             Weekday
                                                                               Lake/State
          3284
                    41660
                           2019
                                     12
                                          31 13430
                                                             Weekday
                                                                               Lake/State
                                location
          0
                (41.870851, -87.776812)
          1
                (41.870851, -87.776812)
          2
                (41.870851, -87.776812)
          3
                (41.870851, -87.776812)
                (41.870851, -87.776812)
          4
          . . .
                (41.884809, -87.627813)
          3280
          3281
                (41.884809, -87.627813)
                (41.884809, -87.627813)
          3282
          3283
                (41.884809, -87.627813)
                (41.884809, -87.627813)
          3284
          [3285 rows x 8 columns]
In [179...
         filter criteria=((ridership cal stations["month"]==7)
                             &(ridership_cal_stations["day_type"] == "Weekday")
                             &(ridership_cal_stations["station_name"]=="Wilson"))
          ############caution#####################case / name senstitice
          print(ridership_cal_stations.loc[filter_criteria, "rides"].sum())
```

1. Complete the following tasks. • Starting with the licenses table, merge to it the zip_demo table on the zip column. Then merge the resulting table to the wards table on the ward column. Save result of the three merged tables to a variable named licenses_zip_ward. • Group the results of the three merged tables by the column alderman and find the median income (agg({'income':'median'}).

```
In [180... lincenses=pd.read_pickle("licenses.p")
    wards=pd.read_pickle("ward.p")
    zip_demo=pd.read_pickle("zip_demo.p")

In [181... print(lincenses.head())
    print(wards.head())
    print(zip_demo.head())
```

```
zip
  account ward
                 aid
                                         business
                                                                  address
                 743
                            REGGIE'S BAR & GRILL
0
   307071
             3
                                                         2105 S STATE ST
                                                                           60616
1
       10
             10
                 829
                                      HONEYBEERS
                                                     13200 S HOUSTON AVE
                                                                           60633
2
    10002
            14
                 775
                                     CELINA DELI
                                                       5089 S ARCHER AVE
                                                                           60632
3
    10005
                 NaN
                                                          2005 W 43RD ST
                                                                           60609
            12
                      KRAFT FOODS NORTH AMERICA
4
    10044
             44
                 638
                      NEYBOUR'S TAVERN & GRILLE
                                                   3651 N SOUTHPORT AVE
                                                                           60613
                   alderman
                                                        address
  ward
                                                                    zip
        Proco "Joe" Moreno
0
                                    2058 NORTH WESTERN AVENUE
                                                                  60647
     1
1
     2
             Brian Hopkins
                                   1400 NORTH ASHLAND AVENUE
                                                                  60622
2
     3
                 Pat Dowell
                                      5046 SOUTH STATE STREET
                                                                  60609
3
     4
          William D. Burns
                              435 EAST 35TH STREET, 1ST FLOOR
                                                                  60616
4
        Leslie A. Hairston
                                         2325 EAST 71ST STREET
                                                                  60649
          income
     zip
0
   60630
           70122
1
   60640
           50488
2
   60622
           87143
3
   60614
          100116
   60608
           41226
```

In [182... lincenses_zip_wards=lincenses.merge(zip_demo,on="zip").merge(wards,on="ward"
print(lincenses_zip_wards)

```
account ward
                    aid
                                            business
                                                                      address_x
\
0
      307071
                 3
                    743
                               REGGIE'S BAR & GRILL
                                                               2105 S STATE ST
1
       11280
                    763
                                           PRIME WAY
                                                           2251 S STATE ST 1ST
                 3
2
       15015
                    NaN
                              SOUTHVIEW MANOR, INC.
                                                           3311 S MICHIGAN AVE
                 3
3
       19168
                 3
                    666
                                            BP AMOCO
                                                      3101 S MICHIGAN AVE 1ST
      205980
                    763
                                     FISH & CHICKEN
                                                                 8 E CERMAK RD
4
                 3
                              J & J
9989
      278535
               18
                               LOOP DEVELOPMENT PTR
                                                                3135 W 71ST ST
                    NaN
9990
       35801
                18
                    NaN
                           MOTHER'S TOUCH DAY CARE
                                                            2501 W 71ST ST 1ST
9991
                    894
      395464
                18
                              Brother's Barber Shop
                                                              2445 W 71ST ST 1
9992
       42012
                18
                    NaN
                                       WINSTON LOTT
                                                            2957 W 71ST ST 1ST
9993
       85634
                18
                    NaN
                         KIDS "R" US OUTLET CENTER
                                                          7455 S CICERO AVE
      zip x
             income
                                alderman
                                                          address y
                                                                      zip y
0
      60616
               46340
                              Pat Dowell
                                           5046 SOUTH STATE STREET
                                                                      60609
1
                                           5046 SOUTH STATE STREET
      60616
               46340
                              Pat Dowell
                                                                      60609
2
      60616
               46340
                              Pat Dowell
                                           5046 SOUTH STATE STREET
                                                                      60609
3
                                           5046 SOUTH STATE STREET
      60616
               46340
                              Pat Dowell
                                                                      60609
      60616
               46340
                                           5046 SOUTH STATE STREET
4
                              Pat Dowell
                                                                      60609
. . .
        . . .
                 . . .
                                     . . .
                                                                 . . .
                                                                        . . .
9989
      60629
               41856
                     Derrick G. Curtis
                                          8359 SOUTH PULASKI ROAD
                                                                      60652
9990
      60629
               41856
                      Derrick G. Curtis
                                           8359 SOUTH PULASKI ROAD
                                                                      60652
                      Derrick G. Curtis
9991
      60629
               41856
                                           8359 SOUTH PULASKI ROAD
                                                                      60652
9992
               41856
      60629
                      Derrick G. Curtis
                                          8359 SOUTH PULASKI ROAD
                                                                      60652
9993
      60629
               41856
                     Derrick G. Curtis
                                          8359 SOUTH PULASKI ROAD
                                                                      60652
```

[9994 rows x 10 columns]

```
In [183... lincenses zip wards["alderman"]
```

```
Pat Dowell
Out[183]:
          1
                          Pat Dowell
          2
                          Pat Dowell
          3
                          Pat Dowell
                          Pat Dowell
                         . . .
          9989
                   Derrick G. Curtis
          9990
                  Derrick G. Curtis
          9991
                   Derrick G. Curtis
          9992
                   Derrick G. Curtis
          9993
                   Derrick G. Curtis
          Name: alderman, Length: 9994, dtype: object
```

1. Complete the following tasks. • What column is likely the best column to merge the two tables on? • Merge the movies table, as the left table, with the financials table using a left join, and save the result to movies_financials. • Count the number of rows in movies_financials with a null value in the budget column

```
In [184... movies=pd.read_pickle("movies.p")
         finicials=pd.read_pickle("financials.p")
         print(movies.head())
         print(finicials.head())
               id
                                 title popularity release date
         0
              257
                                                   2005-09-23
                          Oliver Twist 20.415572
         1
           14290 Better Luck Tomorrow
                                         3.877036
                                                     2002-01-12
         2
           38365
                             Grown Ups 38.864027
                                                     2010-06-24
         3
             9672
                                         3.680896 2006-11-16
                              Infamous
         4 12819
                       Alpha and Omega 12.300789 2010-09-17
               id
                      budget
                                   revenue
         0
             19995 237000000 2.787965e+09
               285 300000000 9.610000e+08
         1
         2 206647 245000000 8.806746e+08
         3
             49026 250000000 1.084939e+09
             49529 260000000 2.841391e+08
In [185...
         movies finicial=movies.merge(finicials,on="id",how="left")
         #merging movies and finicials with a left join
         print(movies_finicial)
```

```
id
                             title popularity release date
                                                                    budget
0
        257
                                                   2005-09-23
                                      20.415572
                                                               50000000.0
                      Oliver Twist
1
      14290 Better Luck Tomorrow
                                       3.877036
                                                   2002-01-12
2
      38365
                         Grown Ups
                                      38.864027
                                                   2010-06-24
                                                                8000000.0
3
                                                   2006-11-16
       9672
                          Infamous
                                       3.680896
                                                                1300000.0
4
      12819
                   Alpha and Omega
                                      12.300789
                                                   2010-09-17
                                                                20000000.0
. . .
        . . .
                                             . . .
                                                                 300000.0
       3089
                                       5.344815
                                                   1948-08-26
4798
                         Red River
4799
      11934
              The Hudsucker Proxy
                                      14.188982
                                                   1994-03-11
                                                                       NaN
4800
      13807
                                       8.486390
                                                   2006-09-06
                                                                       NaN
                            Exiled
4801
      73873
                      Albert Nobbs
                                       7.802245
                                                   2011-12-21
                                                                 8000000.0
4802
      11622
              Blast from the Past
                                       8.737058
                                                   1999-02-12
                                                                35000000.0
          revenue
0
       42093706.0
1
              NaN
2
      271430189.0
3
        1151330.0
4
       39300000.0
4798
        9012000.0
4799
              NaN
4800
              NaN
4801
        5634828.0
4802
       40263020.0
[4803 rows x 6 columns]
number_of_missing_vals=movies_finicial["budget"].isna().sum()
print(number of missing vals)
```

In [186...

1574

Right Join

```
In [187...
         movie_to_genres = pd.read_pickle("movie_to_genres.p")
          movie_to_genres.to_csv("tdmb_to_genres.csv")
          #merge from csv to p
In [188...
          movie_to_genres = pd.read_csv("tdmb_movie_to genres.csv")
          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
          11096
                   205321
                           TV Movie
          11282
                   231617
                           TV Movie
In [189...
         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
0
                                                                 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
                                                  2011-11-12
   78814
               We Have Your Husband
                                       0.102003
                                                                 78814
 153397
                           Restless
                                       0.812776
                                                  2012-12-07
                                                                153397
      genre
0
  TV Movie
1
  TV Movie
2 TV Movie
  TV Movie
  TV Movie
```

Outer Join

```
In [190... movie_to_genres.to_csv("tdmb_to_genres.csv")
    m = movie_to_genres['genre'] == 'Family'
    family = movie_to_genres[m].head(3)
    family
Out[190]: movie_id genre
```

```
        out [190]:
        movie_id
        genre

        5
        12
        Family

        33
        35
        Family

        111
        105
        Family
```

```
In [191... m = movie_to_genres['genre'] == 'Comedy'
comedy = movie_to_genres[m].head(3)
comedy
```

```
        Out [191]:
        movie_id
        genre

        1
        5
        Comedy

        7
        13
        Comedy

        35
        35
        Comedy
```

```
In [192... family_comedy = family.merge(comedy, on='movie_id', how='outer', suffixes=('
    print(family_comedy)
```

```
movie_id genre_fam genre_com
0
         12
                Family
                              NaN
                Family
1
         35
                           Comedy
2
        105
                Family
                              NaN
3
          5
                   NaN
                           Comedy
         13
                   NaN
                           Comedy
```

8

```
In [193... movies=pd.read_pickle("movies.p")
    movie_to_genres = pd.read_csv("tdmb_movie_to_genres.csv")

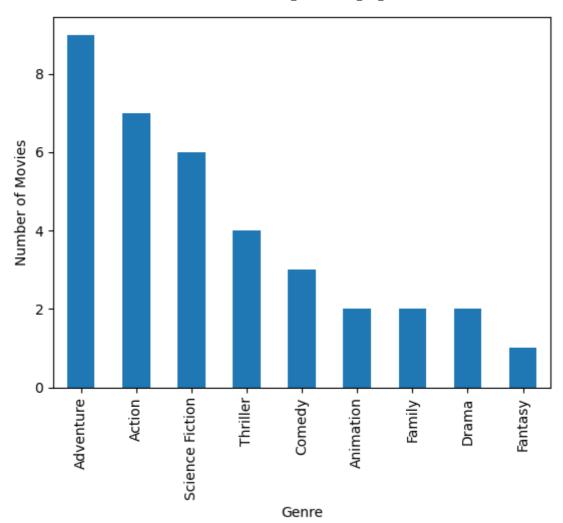
m = movie_to_genres['genre'] == 'Science Fiction'
    scifi_movies = movie_to_genres[m]
```

Out[194]:	id		title popularity		release_date	movie_id	genre_act	genre_sci
	0	18841	The Lost Skeleton of Cadavra	1.680525	2001-09-12	18841	NaN	Science Fiction
	1	26672	The Thief and the Cobbler	2.439184	1993-09-23	26672	NaN	Science Fiction
	2	15301	Twilight Zone: The Movie	12.902975	1983-06-24	15301	NaN	Science Fiction
	3	8452	The 6th Day	18.447479	2000-11-17	8452	NaN	Science Fiction
	4	1649	Bill & Ted's Bogus Journey	11.349664	1991-07-19	1649	NaN	Science Fiction
	•••		•••	•••	•••	•••	•••	
	253	245703	Midnight Special	32.717853	2016-02-18	245703	NaN	Science Fiction
	254	3509	A Scanner Darkly	26.093043	2006-05-25	3509	NaN	Science Fiction
	255	42188	Never Let Me Go	30.983397	2010-09-15	42188	NaN	Science Fiction
	256	18045	The Dark Hours	1.428483	2005-03-11	18045	NaN	Science Fiction
	257	11058	Godsend	12.102350	2004-04-30	11058	NaN	Science Fiction

258 rows × 7 columns

Exercise 9

```
In [200... pop_movies = pd.read_csv("pop_movies.csv")
    movie_to_genres = pd.read_csv("tdmb_movie_to_genres.csv")
    genres_movies = movie_to_genres.merge(pop_movies, left_on='movie_id',right_c
    genre_counts = genres_movies['genre'].value_counts()
    genres_movies.groupby('genre').agg({'id':'count'})
    genre_counts.plot(kind='bar')
    plt.xlabel('Genre')
    plt.ylabel('Number of Movies')
    plt.show()
```



In [103... sequel=pd.read_pickle("sequels.p")
 print(sequel.head())

id title sequel 0 19995 Avatar <NA> 1 862 Toy Story 863 2 863 Toy Story 2 10193 3 597 Titanic <NA> 24428 The Avengers <NA>

Out[109]:	idorg		titleorg	sequelorg	idseq	titleseq	sequelseq
	0	862	Toy Story	863	863	Toy Story 2	10193
	1	863	Toy Story 2	10193	10193	Toy Story 3	<na></na>
	2	675	Harry Potter and the Order of the Phoenix	767	767	Harry Potter and the Half-Blood Prince	<na></na>
	3	121	The Lord of the Rings: The Two Towers	122	122	The Lord of the Rings: The Return of the King	<na></na>
	4	120	The Lord of the Rings: The Fellowship of the Ring	121	121	The Lord of the Rings: The Two Towers	122

```
In [111... print(original_sequels[,['title_org','title_seq']].head())
```

```
File "/var/folders/4j/bnvctt7152z6l5l6szd4m7wh0000gn/T/ipykernel_30085/200
7738660.py", line 1
    print(original_sequels[,['title_org','title_seq']].head())

SyntaxError: invalid syntax

In [129... original_sequels = sequel.merge(sequel, left_on="sequel", right_on='id',how=original_sequels
```

out[129]:		id_org	title_org	sequel_org	id_seq	title_seq	sequel_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>
	4	24428	The Avengers	<na></na>	<na></na>	NaN	<na></na>
	•••				•••		
	4798	133931	Zambezia	<na></na>	<na></na>	NaN	<na></na>
	4799	309503	Zipper	<na></na>	<na></na>	NaN	<na></na>
	4800	34592	ZMD: Zombies of Mass Destruction	<na></na>	<na></na>	NaN	<na></na>
	4801	206213	Zombie Hunter	<na></na>	<na></na>	NaN	<na></na>

Zulu

<NA> <NA>

NaN

<NA>

4803 rows × 6 columns

4802 185567

Exercise 10

Out[130]:		id	department_dir	job_dir	name_dir	department_crew	job_crew	name_crew
	156	19995	Directing	Director	James Cameron	Editing	Editor	Stephen E. Rivkin
	157	19995	Directing	Director	James Cameron	Sound	Sound Designer	Christopher Boyes
	158	19995	Directing	Director	James Cameron	Production	Casting	Mali Finn
	160	19995	Directing	Director	James Cameron	Writing	Writer	James Cameron
	161	19995	Directing	Director	James Cameron	Art	Set Designer	Richard F. Mays

Concatenate DataFrame Together Vertically.

EXERCISE 11

```
In [131... import pandas as pd

tracks_master = pd.read_csv('tracks_master.csv')
    tracks_ride = pd.read_csv('tracks_ride.csv')
    tracks_st = pd.read_csv('tracks_st.csv')

concatenated_with_sort = pd.concat([tracks_master, tracks_ride, tracks_st],
    concatenated_reset_index = pd.concat([tracks_master, tracks_ride, tracks_st])

concatenated_common_columns = pd.concat([tracks_master, tracks_ride, tracks_st])

print("Concatenated with Sorting:")
    print("Concatenated_with_sort.head())

print("\nConcatenated_reset_index.head())

print("\nConcatenated_reset_index.head())

print("\nConcatenated_common_columns.head())
```

```
Concatenated with Sorting:
                  composer gid mtid
  aid
                                                            tid u_price
                                                      name
  152 J.Hetfield/L.Ulrich
                                                    Battery 1853
                                                                     0.99
1
  152
                 K.Hammett
                              3
                                    1
                                          Master Of Puppets 1854
                                                                     0.99
2 152 J.Hetfield/L.Ulrich
                             3
                                    1
                                          Disposable Heroes 1857
                                                                     0.99
0 154
                       NaN
                              3
                                    1 Fight Fire With Fire 1874
                                                                     0.99
1 154
                       NaN
                              3
                                    1
                                         Ride The Lightning 1875
                                                                     0.99
Concatenated with Reset Index:
                        name
                             aid mtid gid
                                                        composer u_price
  1853
0
                     Battery
                              152
                                     1
                                           3
                                              J.Hetfield/L.Ulrich
                                                                     0.99
1
  1854
           Master Of Puppets
                              152
                                      1
                                           3
                                                       K.Hammett
                                                                     0.99
2 1857
           Disposable Heroes
                             152
                                      1
                                           3
                                             J.Hetfield/L.Ulrich
                                                                     0.99
3 1874 Fight Fire With Fire 154
                                           3
                                                                     0.99
                                      1
                                                             NaN
4 1875
          Ride The Lightning
                                           3
                                                             NaN
                                                                     0.99
Concatenated with Common Columns:
   tid
                        name aid mtid gid
                                             u price
0
  1853
                     Battery 152
                                           3
                                                 0.99
                                      1
1 1854
           Master Of Puppets 152
                                           3
                                                 0.99
                                      1
2 1857
           Disposable Heroes 152
                                           3
                                                 0.99
0 1874 Fight Fire With Fire 154
                                           3
                                                 0.99
                                      1
1 1875
          Ride The Lightning 154
                                           3
                                      1
                                                 0.99
```

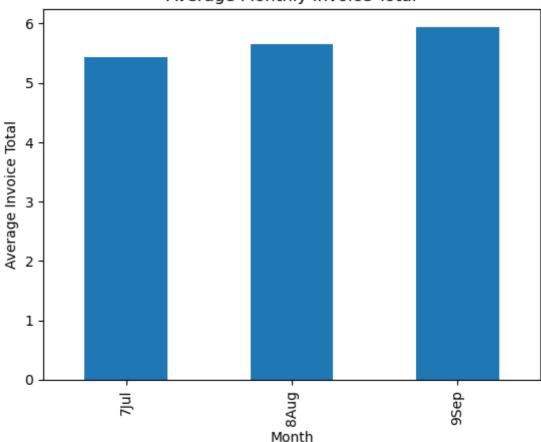
EXERCISE 12

```
inv_jul = pd.read_csv('inv_jul.csv')
inv_aug = pd.read_csv('inv_aug.csv')
inv_sep = pd.read_csv('inv_sep.csv')

avg_inv_by_month = pd.concat([inv_jul, inv_aug, inv_sep], keys=['7Jul', '8Au

average_inv_by_month = avg_inv_by_month.groupby(level=0)['total'].agg('mean'
average_inv_by_month.plot(kind='bar')
plt.title('Average Monthly Invoice Total')
plt.xlabel('Month')
plt.ylabel('Average Invoice Total')
plt.show()
```

Average Monthly Invoice Total



EXERCISE 13

```
In [158...
         gdp = pd.read_csv('GDP.csv')
          sp500 = pd.read csv('S&P500.csv')
          gdp_sp500 = pd.merge_ordered(gdp, sp500, left_on='year', right_on='date', ho
          gdp_sp500[gdp_sp500['year'] == 2018]
          print(gdp_sp500[gdp_sp500['year'] == 2018])
             Unnamed: 0 country code year
                                                           date returns
                                 USA
                                      2018 2.050000e+13
                                                            NaN
In [159...
         gdp_sp500 = pd.merge_ordered(gdp, sp500, left_on='year', right_on='date', ho
          gdp_returns = gdp_sp500[['gdp', 'returns']]
          correlation_matrix = gdp_returns.corr()
          correlation_matrix
Out[159]:
                      gdp
                            returns
             gdp 1.000000 0.220321
          returns 0.220321 1.000000
```

EXERCISE 14

```
In [157... unemployment = pd.read_csv('unemployment.csv')
    inflation = pd.read_csv('inflation.csv')
```

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

plt.figure(figsize=(8, 6))
plt.scatter(inflation_unemploy['unemployment_rate'], inflation_unemploy['cpi
plt.title('Phillips Curve')
plt.xlabel('Unemployment Rate')
plt.ylabel('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
   1/1/2016
             237.833
                      CUSR0000SA0
                                    SEASONALLY ADJUSTED INDEX
3
   1/1/2017
             243.780
                      CUSR0000SA0
                                    SEASONALLY ADJUSTED INDEX
   1/1/2018 248.884
4
                      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
9
                 4.0
```

Phillips Curve

