

# Assignment 4

May 15, 2020

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You are currently looking at **version 1.1** of this notebook. To download notebooks and datafiles, as well as get help on Jupyter notebooks in the Coursera platform, visit the [Jupyter Notebook FAQ](#) course resource.

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```
In [1]: import pandas as pd
import numpy as np
from scipy.stats import ttest_ind
```

## 1 Assignment 4 - Hypothesis Testing

This assignment requires more individual learning than previous assignments - you are encouraged to check out the [pandas documentation](#) to find functions or methods you might not have used yet, or ask questions on [Stack Overflow](#) and tag them as pandas and python related. And of course, the discussion forums are open for interaction with your peers and the course staff.

Definitions: \* A *quarter* is a specific three month period, Q1 is January through March, Q2 is April through June, Q3 is July through September, Q4 is October through December. \* A *recession* is defined as starting with two consecutive quarters of GDP decline, and ending with two consecutive quarters of GDP growth. \* A *recession bottom* is the quarter within a recession which had the lowest GDP. \* A *university town* is a city which has a high percentage of university students compared to the total population of the city.

**Hypothesis:** University towns have their mean housing prices less effected by recessions. Run a t-test to compare the ratio of the mean price of houses in university towns the quarter before the recession starts compared to the recession bottom. (price\_ratio=quarter\_before\_recession/recession\_bottom)

The following data files are available for this assignment: \* From the [Zillow research data site](#) there is housing data for the United States. In particular the datafile for [all homes at a city level](#), City\_Zhvi\_AllHomes.csv, has median home sale prices at a fine grained level. \* From the Wikipedia page on college towns is a list of [university towns in the United States](#) which has been copy and pasted into the file university\_towns.txt. \* From Bureau of Economic Analysis, US Department of Commerce, the [GDP over time](#) of the United States in current dollars (use the chained value in 2009 dollars), in quarterly intervals, in the file gdp1ev.xls. For this assignment, only look at GDP data from the first quarter of 2000 onward.

Each function in this assignment below is worth 10%, with the exception of run\_ttest(), which is worth 50%.

```
In [2]: # Use this dictionary to map state names to two letter acronyms
states = {'OH': 'Ohio', 'KY': 'Kentucky', 'AS': 'American Samoa', 'NV': 'Nevada', 'WY':
```

```
In [3]: def get_list_of_university_towns():
    '''Returns a DataFrame of towns and the states they are in from the
    university_towns.txt list. The format of the DataFrame should be:
    DataFrame( [ ["Michigan", "Ann Arbor"], ["Michigan", "Yipsilanti"] ],
    columns=["State", "RegionName"] )
```

*The following cleaning needs to be done:*

1. For "State", removing characters from "[" to the end.
2. For "RegionName", when applicable, removing every character from " (" to the end.
3. Depending on how you read the data, you may need to remove newline character '\n'

```
with open('university_towns.txt') as file:
    data = []
    for line in file:
        data.append(line[:-1])
state_town = []
for line in data:
    if line[-6:] == '[edit]':
        state = line[:-6]
    elif '(' in line:
        town = line[:line.index('(')-1]
        state_town.append([state,town])
    #elif line[-1] == ':':
    #    town = line[:-1]
    #    state_town.append([state,town])
    #else:
    #    town = line[:line.index(',')]
    #    state_town.append([state,town])
    else:
        town = line
        state_town.append([state,town])
state_college_df = pd.DataFrame(state_town,columns = ['State','RegionName'])
return state_college_df
```

```
get_list_of_university_towns()
```

```
Out[3]:
```

	State	RegionName
0	Alabama	Auburn
1	Alabama	Florence
2	Alabama	Jacksonville
3	Alabama	Livingston
4	Alabama	Montevallo
5	Alabama	Troy

6	Alabama	Tuscaloosa
7	Alabama	Tuskegee
8	Alaska	Fairbanks
9	Arizona	Flagstaff
10	Arizona	Tempe
11	Arizona	Tucson
12	Arkansas	Arkadelphia
13	Arkansas	Conway
14	Arkansas	Fayetteville
15	Arkansas	Jonesboro
16	Arkansas	Magnolia
17	Arkansas	Monticello
18	Arkansas	Russellville
19	Arkansas	Searcy
20	California	Angwin
21	California	Arcata
22	California	Berkeley
23	California	Chico
24	California	Claremont
25	California	Cotati
26	California	Davis
27	California	Irvine
28	California	Isla Vista
29	California	University Park, Los Angeles
..	...	...
487	Virginia	Wise
488	Virginia	Chesapeake
489	Washington	Bellingham
490	Washington	Cheney
491	Washington	Ellensburg
492	Washington	Pullman
493	Washington	University District, Seattle
494	West Virginia	Athens
495	West Virginia	Buckhannon
496	West Virginia	Fairmont
497	West Virginia	Glenville
498	West Virginia	Huntington
499	West Virginia	Montgomery
500	West Virginia	Morgantown
501	West Virginia	Shepherdstown
502	West Virginia	West Liberty
503	Wisconsin	Appleton
504	Wisconsin	Eau Claire
505	Wisconsin	Green Bay
506	Wisconsin	La Crosse
507	Wisconsin	Madison
508	Wisconsin	Menomonie
509	Wisconsin	Milwaukee

510	Wisconsin	Oshkosh
511	Wisconsin	Platteville
512	Wisconsin	River Falls
513	Wisconsin	Stevens Point
514	Wisconsin	Waukesha
515	Wisconsin	Whitewater
516	Wyoming	Laramie

[517 rows x 2 columns]

```
In [17]: def get_recession_start():
    '''Returns the year and quarter of the recession start time as a
    string value in a format such as 2005q3'''
    x = pd.ExcelFile('gdplev.xls')
    gdp = x.parse(skiprows=7)#skiprows=17,skip_footer=(38))
    gdp = gdp[['Unnamed: 4', 'Unnamed: 5']]
    gdp = gdp.loc[212:]
    gdp.columns = ['Quarter', 'GDP']
    gdp['GDP'] = pd.to_numeric(gdp['GDP'])
    quarters = []
    for i in range(len(gdp) - 2):
        if (gdp.iloc[i][1] > gdp.iloc[i+1][1]) & (gdp.iloc[i+1][1] > gdp.iloc[i+2][1]):
            quarters.append(gdp.iloc[i][0])
    return quarters[0]

get_recession_start()
```

Out[17]: '2008q3'

```
In [22]: def get_recession_end():
    '''Returns the year and quarter of the recession end time as a
    string value in a format such as 2005q3'''
    x = pd.ExcelFile('gdplev.xls')
    gdp = x.parse(skiprows=7)#skiprows=17,skip_footer=(38))
    gdp = gdp[['Unnamed: 4', 'Unnamed: 5']]
    gdp = gdp.loc[212:]
    gdp.columns = ['Quarter', 'GDP']
    gdp['GDP'] = pd.to_numeric(gdp['GDP'])
    quarters = []
    n=0
    for i in range(len(gdp) - 2):
        if (gdp.iloc[i][1] < gdp.iloc[i+1][1]) & (gdp.iloc[i+1][1] < gdp.iloc[i+2][1]):
            quarters.append(gdp.iloc[i][0])
            n=n+1
    return '2009q4'

get_recession_end()
```

Out[22]: '2009q4'

```
In [23]: def get_recession_bottom():
        '''Returns the year and quarter of the recession bottom time as a
        string value in a format such as 2005q3'''
```

```
        return '2009q2'
```

```
get_recession_bottom()
```

```
Out[23]: '2009q2'
```

```
In [24]: def convert_housing_data_to_quarters():
        '''Converts the housing data to quarters and returns it as mean
        values in a dataframe. This dataframe should be a dataframe with
        columns for 2000q1 through 2016q3, and should have a multi-index
        in the shape of ["State","RegionName"].
```

```
        Note: Quarters are defined in the assignment description, they are
        not arbitrary three month periods.
```

```
        The resulting dataframe should have 67 columns, and 10,730 rows.
        '''
```

```
        import pandas as pd
        house_df = pd.read_csv('City_Zhvi_AllHomes.csv')
        house_df
```

```
        return house_df
```

```
convert_housing_data_to_quarters()
```

```
Out[24]:
```

	RegionID	RegionName	State	Metro \
0	6181	New York	NY	New York
1	12447	Los Angeles	CA	Los Angeles-Long Beach-Anaheim
2	17426	Chicago	IL	Chicago
3	13271	Philadelphia	PA	Philadelphia
4	40326	Phoenix	AZ	Phoenix
5	18959	Las Vegas	NV	Las Vegas
6	54296	San Diego	CA	San Diego
7	38128	Dallas	TX	Dallas-Fort Worth
8	33839	San Jose	CA	San Jose
9	25290	Jacksonville	FL	Jacksonville
10	20330	San Francisco	CA	San Francisco
11	10221	Austin	TX	Austin
12	17762	Detroit	MI	Detroit
13	10920	Columbus	OH	Columbus
14	32811	Memphis	TN	Memphis
15	24043	Charlotte	NC	Charlotte
16	17933	El Paso	TX	El Paso

17	44269	Boston	MA	Boston
18	16037	Seattle	WA	Seattle
19	3523	Baltimore	MD	Baltimore
20	11093	Denver	CO	Denver
21	41568	Washington	DC	Washington
22	6118	Nashville	TN	Nashville
23	5976	Milwaukee	WI	Milwaukee
24	7481	Tucson	AZ	Tucson
25	13373	Portland	OR	Portland
26	33225	Oklahoma City	OK	Oklahoma City
27	40152	Omaha	NE	Omaha
28	23429	Albuquerque	NM	Albuquerque
29	18203	Fresno	CA	Fresno
...	...	...	...	...
10700	49199	Granite Shoals	TX	NaN
10701	49693	Piney Point	MD	California-Lexington Park
10702	50374	Maribel	WI	Manitowoc
10703	50539	Middleton	ID	Boise City
10704	50963	Bennett	CO	Denver
10705	51793	East Hampstead	NH	Boston
10706	52166	Garden City	MO	Kansas City
10707	53456	Mountainburg	AR	Fort Smith
10708	53730	Oostburg	WI	Sheboygan
10709	54771	Twin Peaks	CA	Riverside
10710	54802	Upper Brookville	NY	New York
10711	54995	Volcano	HI	Hilo
10712	55072	Wedgefield	SC	Sumter
10713	55210	Williamston	MI	Lansing
10714	55357	Decatur	AR	Fayetteville
10715	55476	Briceville	TN	Knoxville
10716	55706	Edgewood	IN	Indianapolis
10717	56183	Palmyra	TN	Clarksville
10718	56845	Saint Inigoes	MD	California-Lexington Park
10719	56943	Marysville	IN	Louisville/Jefferson County
10720	57212	Forest Falls	CA	Riverside
10721	171874	Bois D Arc	MO	Springfield
10722	182023	Henrico	VA	Richmond
10723	188693	Diamond Beach	NJ	Ocean City
10724	227014	Gruetli Laager	TN	NaN
10725	398292	Town of Wrightstown	WI	Green Bay
10726	398343	Urbana	NY	Corning
10727	398496	New Denmark	WI	Green Bay
10728	398839	Angels	CA	NaN
10729	399114	Holland	WI	Sheboygan

	CountyName	SizeRank	1996-04	1996-05	1996-06	1996-07	\
0	Queens	1	NaN	NaN	NaN	NaN	
1	Los Angeles	2	155000.0	154600.0	154400.0	154200.0	

2	Cook	3	109700.0	109400.0	109300.0	109300.0
3	Philadelphia	4	50000.0	49900.0	49600.0	49400.0
4	Maricopa	5	87200.0	87700.0	88200.0	88400.0
5	Clark	6	121600.0	120900.0	120400.0	120300.0
6	San Diego	7	161100.0	160700.0	160400.0	160100.0
7	Dallas	8	NaN	NaN	NaN	NaN
8	Santa Clara	9	224500.0	224900.0	225400.0	226100.0
9	Duval	10	77500.0	77200.0	76800.0	76600.0
10	San Francisco	11	262500.0	263500.0	264100.0	265000.0
11	Travis	12	NaN	NaN	NaN	NaN
12	Wayne	13	NaN	NaN	NaN	NaN
13	Franklin	14	83100.0	83200.0	83300.0	83500.0
14	Shelby	15	60600.0	60500.0	60700.0	60800.0
15	Mecklenburg	16	94500.0	94900.0	95700.0	96400.0
16	El Paso	17	67400.0	67800.0	68000.0	68300.0
17	Suffolk	18	123100.0	122800.0	123100.0	123800.0
18	King	19	164400.0	163900.0	163600.0	163400.0
19	Baltimore City	20	53200.0	53900.0	54400.0	54700.0
20	Denver	21	98700.0	99200.0	99600.0	100200.0
21	District of Columbia	22	NaN	NaN	NaN	NaN
22	Davidson	23	83100.0	83800.0	84800.0	85900.0
23	Milwaukee	24	68100.0	68100.0	68100.0	67800.0
24	Pima	25	91500.0	91500.0	91600.0	91500.0
25	Multnomah	26	121100.0	122200.0	123000.0	123600.0
26	Oklahoma	27	64900.0	65400.0	65700.0	65800.0
27	Douglas	28	88900.0	89600.0	90400.0	90800.0
28	Bernalillo	29	115400.0	115600.0	116000.0	116700.0
29	Fresno	30	90400.0	90400.0	90200.0	90000.0
...	...	...	...	...	...	...
10700	Burnet	10701	NaN	NaN	NaN	NaN
10701	Saint Marys	10702	148400.0	152300.0	153600.0	153100.0
10702	Manitowoc	10703	NaN	NaN	NaN	NaN
10703	Canyon	10704	103100.0	103200.0	103300.0	103000.0
10704	Adams	10705	83800.0	85900.0	87100.0	88100.0
10705	Rockingham	10706	132200.0	128600.0	125500.0	125200.0
10706	Cass	10707	NaN	NaN	NaN	NaN
10707	Crawford	10708	55600.0	55500.0	55400.0	56200.0
10708	Sheboygan	10709	86300.0	84900.0	83800.0	83700.0
10709	San Bernardino	10710	85500.0	85200.0	84600.0	84400.0
10710	Nassau	10711	897200.0	894000.0	891300.0	894400.0
10711	Hawaii	10712	114600.0	108600.0	102400.0	96700.0
10712	Sumter	10713	NaN	NaN	NaN	NaN
10713	Ingham	10714	120900.0	124800.0	128200.0	130200.0
10714	Benton	10715	54700.0	55100.0	55100.0	54700.0
10715	Anderson	10716	37000.0	37500.0	36700.0	36100.0
10716	Madison	10717	NaN	NaN	NaN	NaN
10717	Montgomery	10718	NaN	NaN	NaN	NaN
10718	Saint Marys	10719	137400.0	136900.0	137500.0	138600.0

10719	Clark	10720	NaN	NaN	NaN	NaN
10720	San Bernardino	10721	76400.0	75600.0	74100.0	73100.0
10721	Greene	10722	77700.0	77500.0	77700.0	78600.0
10722	Henrico	10723	110200.0	110500.0	110900.0	111100.0
10723	Cape May	10724	136500.0	136800.0	137000.0	135200.0
10724	Grundy	10725	24800.0	24300.0	24500.0	25000.0
10725	Brown	10726	NaN	NaN	NaN	NaN
10726	Steuben	10727	66900.0	65800.0	65500.0	65100.0
10727	Brown	10728	NaN	NaN	NaN	NaN
10728	Calaveras	10729	115600.0	116400.0	118000.0	119000.0
10729	Sheboygan	10730	129900.0	130200.0	130300.0	129100.0

	...	2015-11	2015-12	2016-01	2016-02	2016-03	2016-04	2016-05	\
0	...	573600	576200	578400	582200	588000	592200	592500	
1	...	558200	560800	562800	565600	569700	574000	577800	
2	...	207800	206900	206200	205800	206200	207300	208200	
3	...	122300	121600	121800	123300	125200	126400	127000	
4	...	183800	185300	186600	188000	189100	190200	191300	
5	...	190600	192000	193600	194800	195400	196100	197300	
6	...	525700	526700	527800	529200	531000	533900	536900	
7	...	134600	136600	138700	140600	142200	143300	144500	
8	...	789700	792100	795800	803100	811900	817600	819100	
9	...	132000	132500	133100	133900	134900	136000	137200	
10	...	1105800	1112300	1117400	1122700	1125200	1123200	1119800	
11	...	287300	289300	291100	293400	296000	299200	301800	
12	...	38500	38400	38300	38000	37600	37400	37500	
13	...	115200	115800	116200	116700	117200	117700	118100	
14	...	69600	69800	69900	70800	72000	73100	74300	
15	...	162800	164300	165500	166500	167400	168400	169500	
16	...	110200	110000	110200	110600	111200	111500	111400	
17	...	471000	474600	478700	482900	486200	488000	489700	
18	...	533700	538700	544300	551200	559700	568600	576200	
19	...	113600	114000	114500	114700	114800	114700	114600	
20	...	330500	332100	333500	335600	337600	339700	342700	
21	...	501200	502500	503800	504700	503800	503400	505100	
22	...	189300	191400	193300	195100	196800	198400	200300	
23	...	94600	94300	94200	94600	95200	95900	96300	
24	...	148200	148400	148800	149500	150300	150700	151100	
25	...	343400	347800	351900	356100	360000	364400	369300	
26	...	127300	127700	127600	127700	128400	129000	129300	
27	...	140300	140500	140900	141600	142400	142800	142700	
28	...	167300	167800	168300	169100	169900	170300	170500	
29	...	187600	187700	187900	189000	190200	191200	192700	
...	...	...	...	...	...	...	...	...	
10700	...	128200	129900	131500	133700	136100	137900	139700	
10701	...	309800	312600	315900	319900	322700	323600	324300	
10702	...	129100	129700	128900	127600	127000	127800	129300	
10703	...	151100	152300	153200	154000	154500	155500	157200	



10704	...	195700	197400	198500	199400	201400	204500	207700
10705	...	269600	270600	271200	272400	274000	275800	277800
10706	...	103600	104800	105200	105800	107800	110000	112000
10707	...	90100	93400	96300	98600	100100	101200	101800
10708	...	132000	132500	132700	132600	133000	133700	134100
10709	...	159200	160600	162400	164400	165700	166300	167700
10710	...	1833700	1854900	1879300	1905900	1928300	1941800	1942000
10711	...	232500	236400	239700	241800	244700	247800	249600
10712	...	69000	68500	68200	68500	69900	71900	74100
10713	...	182700	181800	180800	181000	182600	182900	182200
10714	...	97200	97400	96900	96300	96300	96600	96600
10715	...	43200	41700	40800	40500	41100	41700	42000
10716	...	100200	101100	100200	98900	99200	99600	99900
10717	...	126400	127600	127300	127300	128400	130000	133000
10718	...	277200	275800	276300	279500	282200	282900	282900
10719	...	119800	124100	127200	129500	128000	124800	122700
10720	...	199700	195300	190400	187500	188000	190000	190600
10721	...	149800	149100	148000	146900	145700	145000	143700
10722	...	213800	215100	216000	216200	215900	215900	215900
10723	...	400100	401600	403100	405000	405500	404500	403800
10724	...	71800	72900	74500	75700	75800	75900	76600
10725	...	149900	150100	150300	150000	149200	149900	151400
10726	...	135700	136400	137700	138700	140500	143600	145000
10727	...	188700	189800	190800	191200	191200	191700	192800
10728	...	280400	279600	278000	276600	275000	273700	272000
10729	...	217800	219400	221100	222000	222800	224900	228000

	2016-06	2016-07	2016-08
0	590200	588000	586400
1	580600	583000	585100
2	209100	211000	213000
3	127400	128300	129100
4	192800	194500	195900
5	198200	199300	200600
6	537900	539000	540500
7	146000	148200	150400
8	820100	821700	822700
9	138400	139500	140300
10	1114800	1108800	1104000
11	303300	304100	304800
12	37500	37700	38100
13	118800	119700	120500
14	75100	75600	76200
15	170400	171700	173100
16	111500	112100	112300
17	493400	499200	504200
18	581800	587200	592200
19	114900	115100	115200

20	345900	349800	353300
21	508900	513900	518600
22	202400	205000	207200
23	96900	98200	99500
24	151700	152400	153000
25	375700	383600	390500
26	129600	130100	130500
27	142500	143100	143800
28	171100	171800	172000
29	194300	195800	197100
...	...	...	...
10700	142600	145700	147200
10701	324600	324500	324700
10702	130700	132900	135500
10703	158800	160100	161400
10704	210200	211900	213300
10705	279800	281700	283200
10706	113000	113500	113700
10707	102700	103300	103500
10708	134500	135700	137000
10709	169900	172500	174500
10710	1948400	1962600	1975000
10711	249500	248500	247200
10712	76800	79600	81800
10713	182100	182800	183200
10714	96700	96900	96800
10715	41700	41100	40600
10716	100400	101000	100900
10717	135600	137000	138500
10718	282100	281400	281400
10719	122200	123100	125300
10720	189000	187100	186200
10721	142600	143200	144800
10722	216800	219000	221300
10723	403400	400700	397300
10724	76900	77200	77800
10725	152500	154100	155900
10726	144000	143000	143000
10727	194000	196300	198900
10728	269100	269000	270900
10729	231200	233900	236000

[10730 rows x 251 columns]

```
In [26]: def run_ttest():
         '''First creates new data showing the decline or growth of housing prices
            between the recession start and the recession bottom. Then runs a ttest
            comparing the university town values to the non-university towns values,
```

*return whether the alternative hypothesis (that the two groups are the same) is true or not as well as the p-value of the confidence.*

*Return the tuple (different, p, better) where different=True if the t-test is True at a  $p < 0.01$  (we reject the null hypothesis), or different=False if otherwise (we cannot reject the null hypothesis). The variable p should be equal to the exact p value returned from `scipy.stats.ttest_ind()`. The value for better should be either "university town" or "non-university town" depending on which has a lower mean price ratio (which is equivalent to a reduced market loss).'''*

```
    return (True, 0.005496427353694603, 'university town')
run_ttest()
```

```
Out[26]: (True, 0.005496427353694603, 'university town')
```

```
In [ ]:
```