

You are currently looking at **version 1.0** 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.

Merging Dataframes

In [1]:

```
import pandas as pd

df = pd.DataFrame([{'Name': 'Chris', 'Item Purchased': 'Sponge', 'Cost': 22.50},
                   {'Name': 'Kevyn', 'Item Purchased': 'Kitty Litter', 'Cost': 2.50},
                   {'Name': 'Filip', 'Item Purchased': 'Spoon', 'Cost': 5.00}],
                  index=['Store 1', 'Store 1', 'Store 2'])

df
```

Out[1]:

	Cost	Item Purchased	Name
Store 1	22.5	Sponge	Chris
Store 1	2.5	Kitty Litter	Kevyn
Store 2	5.0	Spoon	Filip

In [2]:

```
df['Date'] = ['December 1', 'January 1', 'mid-May']
df
```

Out[2]:

	Cost	Item Purchased	Name	Date
Store 1	22.5	Sponge	Chris	December 1
Store 1	2.5	Kitty Litter	Kevyn	January 1
Store 2	5.0	Spoon	Filip	mid-May

In [3]:

```
df['Delivered'] = True
df
```

Out[3]:

	Cost	Item Purchased	Name	Date	Delivered
Store 1	22.5	Sponge	Chris	December 1	True
Store 1	2.5	Kitty Litter	Kevyn	January 1	True
Store 2	5.0	Spoon	Filip	mid-May	True

In [4]:

```
df['Feedback'] = ['Positive', None, 'Negative']
df
```

Out[4]:

	Cost	Item Purchased	Name	Date	Delivered	Feedback
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	Cost	Item Purchased	Name	Date	Delivered	Feedback
Store 1	22.5	Sponge	Chris	December 1	True	Positive
Store 1	2.5	Kitty Litter	Kevyn	January 1	True	None
Store 2	5.0	Spoon	Filip	mid-May	True	Negative

In [6]:

```
adf = df.reset_index()
adf['Date'] = pd.Series({0: 'December 1', 2: 'mid-May'})
adf
```

Out[6]:

	index	Cost	Item Purchased	Name	Date	Delivered	Feedback
0	Store 1	22.5	Sponge	Chris	December 1	True	Positive
1	Store 1	2.5	Kitty Litter	Kevyn	NaN	True	None
2	Store 2	5.0	Spoon	Filip	mid-May	True	Negative

In [7]:

```
staff_df = pd.DataFrame([{'Name': 'Kelly', 'Role': 'Director of HR'},
                          {'Name': 'Sally', 'Role': 'Course liasion'},
                          {'Name': 'James', 'Role': 'Grader'}])
staff_df = staff_df.set_index('Name')
student_df = pd.DataFrame([{'Name': 'James', 'School': 'Business'},
                            {'Name': 'Mike', 'School': 'Law'},
                            {'Name': 'Sally', 'School': 'Engineering'}])
student_df = student_df.set_index('Name')
print(staff_df.head())
print()
print(student_df.head())
```

	Role
Name	
Kelly	Director of HR
Sally	Course liasion
James	Grader

	School
Name	
James	Business
Mike	Law
Sally	Engineering

In [8]:

```
pd.merge(staff_df, student_df, how='outer', left_index=True, right_index=True)
```

Out[8]:

	Role	School
Name		
James	Grader	Business
Kelly	Director of HR	NaN
Mike	NaN	Law
Sally	Course liasion	Engineering

In [9]:

```
pd.merge(staff_df, student_df, how='inner', left_index=True, right_index=True)
```

Out[9]:

	Role	School
Name		
Sally	Course liasion	Engineering
James	Grader	Business

In [10]:

```
pd.merge(staff_df, student_df, how='left', left_index=True, right_index=True)
```

Out[10]:

	Role	School
Name		
Kelly	Director of HR	NaN
Sally	Course liasion	Engineering
James	Grader	Business

In [11]:

```
pd.merge(staff_df, student_df, how='right', left_index=True, right_index=True)
```

Out[11]:

	Role	School
Name		
James	Grader	Business
Mike	NaN	Law
Sally	Course liasion	Engineering

In [12]:

```
staff_df = staff_df.reset_index()
student_df = student_df.reset_index()
pd.merge(staff_df, student_df, how='left', left_on='Name', right_on='Name')
```

Out[12]:

	Name	Role	School
0	Kelly	Director of HR	NaN
1	Sally	Course liasion	Engineering
2	James	Grader	Business

In [13]:

```
staff_df = pd.DataFrame([{'Name': 'Kelly', 'Role': 'Director of HR', 'Location': 'State Street'},
                          {'Name': 'Sally', 'Role': 'Course liasion', 'Location': 'Washin gton Avenue'},
                          {'Name': 'James', 'Role': 'Grader', 'Location': 'Washington Ave nue'}])
student_df = pd.DataFrame([{'Name': 'James', 'School': 'Business', 'Location': '1024 Bil liard Avenue'},
                             {'Name': 'Mike', 'School': 'Law', 'Location': 'Fraternity Hou se #22'},
                             {'Name': 'Sally', 'School': 'Engineering', 'Location': '512 W ilson Crescent'}])
pd.merge(staff_df, student_df, how='left', left_on='Name', right_on='Name')
```

Out[13]:

	Location	Name	Role	Location	School
--	----------	------	------	----------	--------

	Location_x	Name	Role	Location_y	School
0	State Street	Kelly	Director of HR	NaN	NaN
1	Washington Avenue	Sally	Course liasion	512 Wilson Crescent	Engineering
2	Washington Avenue	James	Grader	1024 Billiard Avenue	Business

In [14]:

```
staff_df = pd.DataFrame([{'First Name': 'Kelly', 'Last Name': 'Desjardins', 'Role': 'Director of HR'},
                          {'First Name': 'Sally', 'Last Name': 'Brooks', 'Role': 'Course liasion'},
                          {'First Name': 'James', 'Last Name': 'Wilde', 'Role': 'Grader'}])
student_df = pd.DataFrame([{'First Name': 'James', 'Last Name': 'Hammond', 'School': 'Business'},
                             {'First Name': 'Mike', 'Last Name': 'Smith', 'School': 'Law'},
                             {'First Name': 'Sally', 'Last Name': 'Brooks', 'School': 'Engineering'}])
pd.merge(staff_df, student_df, how='inner', left_on=['First Name', 'Last Name'], right_on=['First Name', 'Last Name'])
```

Out[14]:

	First Name	Last Name	Role	School
0	Sally	Brooks	Course liasion	Engineering

Idiomatic Pandas: Making Code Pandorable

In [20]:

```
import pandas as pd
df = pd.read_csv('census.csv')
df
```

Out[20]:

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPE
0	40	3	6	1	0	Alabama	Alabama	4779736	4780127	
1	50	3	6	1	1	Alabama	Autauga County	54571	54571	
2	50	3	6	1	3	Alabama	Baldwin County	182265	182265	
3	50	3	6	1	5	Alabama	Barbour County	27457	27457	
4	50	3	6	1	7	Alabama	Bibb County	22915	22919	
5	50	3	6	1	9	Alabama	Blount County	57322	57322	
6	50	3	6	1	11	Alabama	Bullock County	10914	10915	
7	50	3	6	1	13	Alabama	Butler County	20947	20946	
8	50	3	6	1	15	Alabama	Calhoun County	118572	118586	
9	50	3	6	1	17	Alabama	Chambers County	34215	34170	
10	50	3	6	1	19	Alabama	Cherokee County	25989	25986	

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPE
11	50	3	6	1	21	Alabama	Chilton County	43643	43631	
12	50	3	6	1	23	Alabama	Choctaw County	13859	13858	
13	50	3	6	1	25	Alabama	Clarke County	25833	25840	
14	50	3	6	1	27	Alabama	Clay County	13932	13932	
15	50	3	6	1	29	Alabama	Cleburne County	14972	14972	
16	50	3	6	1	31	Alabama	Coffee County	49948	49948	
17	50	3	6	1	33	Alabama	Colbert County	54428	54428	
18	50	3	6	1	35	Alabama	Conecuh County	13228	13228	
19	50	3	6	1	37	Alabama	Coosa County	11539	11758	
20	50	3	6	1	39	Alabama	Covington County	37765	37765	
21	50	3	6	1	41	Alabama	Crenshaw County	13906	13906	
22	50	3	6	1	43	Alabama	Cullman County	80406	80410	
23	50	3	6	1	45	Alabama	Dale County	50251	50251	
24	50	3	6	1	47	Alabama	Dallas County	43820	43820	
25	50	3	6	1	49	Alabama	DeKalb County	71109	71115	
26	50	3	6	1	51	Alabama	Elmore County	79303	79296	
27	50	3	6	1	53	Alabama	Escambia County	38319	38319	
28	50	3	6	1	55	Alabama	Etowah County	104430	104427	
29	50	3	6	1	57	Alabama	Fayette County	17241	17241	
...
3163	50	2	3	55	131	Wisconsin	Washington County	131887	131885	
3164	50	2	3	55	133	Wisconsin	Waukesha County	389891	389938	
3165	50	2	3	55	135	Wisconsin	Waupaca County	52410	52410	
3166	50	2	3	55	137	Wisconsin	Waushara County	24496	24496	
3167	50	2	3	55	139	Wisconsin	Winnebago County	166994	166994	
3168	50	2	3	55	141	Wisconsin	Wood County	74749	74749	
3169	40	4	8	56	0	Wyoming	Wyoming	563626	563767	
3170	50	4	8	56	1	Wyoming	Albany County	36299	36299	
3171	50	4	8	56	3	Wyoming	Big Horn County	11668	11668	

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPE
3172	50	4	8	56	5	Wyoming	Campbell County	46133	46133	
3173	50	4	8	56	7	Wyoming	Carbon County	15885	15885	
3174	50	4	8	56	9	Wyoming	Converse County	13833	13833	
3175	50	4	8	56	11	Wyoming	Crook County	7083	7083	
3176	50	4	8	56	13	Wyoming	Fremont County	40123	40123	
3177	50	4	8	56	15	Wyoming	Goshen County	13249	13247	
3178	50	4	8	56	17	Wyoming	Hot Springs County	4812	4812	
3179	50	4	8	56	19	Wyoming	Johnson County	8569	8569	
3180	50	4	8	56	21	Wyoming	Laramie County	91738	91881	
3181	50	4	8	56	23	Wyoming	Lincoln County	18106	18106	
3182	50	4	8	56	25	Wyoming	Natrona County	75450	75450	
3183	50	4	8	56	27	Wyoming	Niobrara County	2484	2484	
3184	50	4	8	56	29	Wyoming	Park County	28205	28205	
3185	50	4	8	56	31	Wyoming	Platte County	8667	8667	
3186	50	4	8	56	33	Wyoming	Sheridan County	29116	29116	
3187	50	4	8	56	35	Wyoming	Sublette County	10247	10247	
3188	50	4	8	56	37	Wyoming	Sweetwater County	43806	43806	
3189	50	4	8	56	39	Wyoming	Teton County	21294	21294	
3190	50	4	8	56	41	Wyoming	Uinta County	21118	21118	
3191	50	4	8	56	43	Wyoming	Washakie County	8533	8533	
3192	50	4	8	56	45	Wyoming	Weston County	7208	7208	

3193 rows x 100 columns



In [16]:

```
(df.where(df['SUMLEV']==50)
    .dropna()
    .set_index(['STNAME','CTYNAME'])
    .rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'}))
```

Out[16]:

	SUMLEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	Estimates Base 2010	POPESTIMATE2010	PO
	STNAME	CTYNAME							

Alabama	County	SUM	LEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010	POP	Estimates Base 2010	POPESTIMATE2010	PO
		50.0	3.0	6.0	1.0	1.0	54571.0	54571.0	54660.0	PO		
STNAME	CTYNAME	50.0	3.0	6.0	1.0	3.0	182265.0	182265.0	183193.0			
	Barbour County	50.0	3.0	6.0	1.0	5.0	27457.0	27457.0	27341.0			
	Bibb County	50.0	3.0	6.0	1.0	7.0	22915.0	22919.0	22861.0			
	Blount County	50.0	3.0	6.0	1.0	9.0	57322.0	57322.0	57373.0			
	Bullock County	50.0	3.0	6.0	1.0	11.0	10914.0	10915.0	10887.0			
	Butler County	50.0	3.0	6.0	1.0	13.0	20947.0	20946.0	20944.0			
	Calhoun County	50.0	3.0	6.0	1.0	15.0	118572.0	118586.0	118437.0			
	Chambers County	50.0	3.0	6.0	1.0	17.0	34215.0	34170.0	34098.0			
	Cherokee County	50.0	3.0	6.0	1.0	19.0	25989.0	25986.0	25976.0			
	Chilton County	50.0	3.0	6.0	1.0	21.0	43643.0	43631.0	43665.0			
	Choctaw County	50.0	3.0	6.0	1.0	23.0	13859.0	13858.0	13841.0			
	Clarke County	50.0	3.0	6.0	1.0	25.0	25833.0	25840.0	25767.0			
	Clay County	50.0	3.0	6.0	1.0	27.0	13932.0	13932.0	13880.0			
	Cleburne County	50.0	3.0	6.0	1.0	29.0	14972.0	14972.0	14973.0			
	Coffee County	50.0	3.0	6.0	1.0	31.0	49948.0	49948.0	50177.0			
	Colbert County	50.0	3.0	6.0	1.0	33.0	54428.0	54428.0	54514.0			
	Conecuh County	50.0	3.0	6.0	1.0	35.0	13228.0	13228.0	13208.0			
	Coosa County	50.0	3.0	6.0	1.0	37.0	11539.0	11758.0	11758.0			
	Covington County	50.0	3.0	6.0	1.0	39.0	37765.0	37765.0	37796.0			
	Crenshaw County	50.0	3.0	6.0	1.0	41.0	13906.0	13906.0	13853.0			
	Cullman County	50.0	3.0	6.0	1.0	43.0	80406.0	80410.0	80473.0			
	Dale County	50.0	3.0	6.0	1.0	45.0	50251.0	50251.0	50358.0			
	Dallas County	50.0	3.0	6.0	1.0	47.0	43820.0	43820.0	43803.0			
	DeKalb County	50.0	3.0	6.0	1.0	49.0	71109.0	71115.0	71142.0			
	Elmore County	50.0	3.0	6.0	1.0	51.0	79303.0	79296.0	79465.0			
	Escambia County	50.0	3.0	6.0	1.0	53.0	38319.0	38319.0	38309.0			
	Etowah County	50.0	3.0	6.0	1.0	55.0	104430.0	104427.0	104442.0			
	Fayette	50.0	3.0	6.0	1.0	57.0	15011.0	15011.0	15001.0			

STNAME	County	50.0	3.0	6.0	1.0	57.0	17241.0	17241.0	17231.0
	Franklin County	SUMLEV 50.0	REGION 3.0	DIVISION 6.0	STATE 1.0	COUNTY 59.0	CENSUS2010POP 31704.0	Estimates Base 2010709.0	POPESTIMATE2010 31734.0
	CTYNAME
Wisconsin	Washburn County	50.0	2.0	3.0	55.0	129.0	15911.0	15911.0	15930.0
	Washington County	50.0	2.0	3.0	55.0	131.0	131887.0	131885.0	131967.0
	Waukesha County	50.0	2.0	3.0	55.0	133.0	389891.0	389938.0	390076.0
	Waupaca County	50.0	2.0	3.0	55.0	135.0	52410.0	52410.0	52422.0
	Waushara County	50.0	2.0	3.0	55.0	137.0	24496.0	24496.0	24506.0
	Winnebago County	50.0	2.0	3.0	55.0	139.0	166994.0	166994.0	167059.0
	Wood County	50.0	2.0	3.0	55.0	141.0	74749.0	74749.0	74807.0
Wyoming	Albany County	50.0	4.0	8.0	56.0	1.0	36299.0	36299.0	36428.0
	Big Horn County	50.0	4.0	8.0	56.0	3.0	11668.0	11668.0	11672.0
	Campbell County	50.0	4.0	8.0	56.0	5.0	46133.0	46133.0	46244.0
	Carbon County	50.0	4.0	8.0	56.0	7.0	15885.0	15885.0	15837.0
	Converse County	50.0	4.0	8.0	56.0	9.0	13833.0	13833.0	13826.0
	Crook County	50.0	4.0	8.0	56.0	11.0	7083.0	7083.0	7114.0
	Fremont County	50.0	4.0	8.0	56.0	13.0	40123.0	40123.0	40222.0
	Goshen County	50.0	4.0	8.0	56.0	15.0	13249.0	13247.0	13408.0
	Hot Springs County	50.0	4.0	8.0	56.0	17.0	4812.0	4812.0	4813.0
	Johnson County	50.0	4.0	8.0	56.0	19.0	8569.0	8569.0	8581.0
	Laramie County	50.0	4.0	8.0	56.0	21.0	91738.0	91881.0	92271.0
	Lincoln County	50.0	4.0	8.0	56.0	23.0	18106.0	18106.0	18091.0
	Natrona County	50.0	4.0	8.0	56.0	25.0	75450.0	75450.0	75472.0
	Niobrara County	50.0	4.0	8.0	56.0	27.0	2484.0	2484.0	2492.0
	Park County	50.0	4.0	8.0	56.0	29.0	28205.0	28205.0	28259.0
	Platte County	50.0	4.0	8.0	56.0	31.0	8667.0	8667.0	8678.0
	Sheridan County	50.0	4.0	8.0	56.0	33.0	29116.0	29116.0	29146.0
	Sublette County	50.0	4.0	8.0	56.0	35.0	10247.0	10247.0	10244.0
	Sweetwater County	50.0	4.0	8.0	56.0	37.0	43806.0	43806.0	43593.0

STNAME	CTYNAME	Teton County	SUMLEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	Estimates Base 2010	POPESTIMATE2010	PO
		50.0	4.0	8.0	56.0	39.0	21294.0	Estimates Base 2010	21294.0	21297.0	
		Uinta County	50.0	4.0	8.0	56.0	41.0	21118.0	21118.0	21102.0	
		Washakie County	50.0	4.0	8.0	56.0	43.0	8533.0	8533.0	8545.0	
		Weston County	50.0	4.0	8.0	56.0	45.0	7208.0	7208.0	7181.0	

3142 rows x 98 columns



In [17]:

```
#print(df.drop(df[df['Quantity'] == 0].index).rename(columns={'Weight': 'Weight (oz.)'}))
```

In [21]:

```
df = df[df['SUMLEV']==50]
df.set_index(['STNAME','CTYNAME'], inplace=True)
df.rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'})
```

Out[21]:

STNAME	CTYNAME	SUMLEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	Estimates Base 2010	POPESTIMATE2010	PO
		50	3	6	1	1	54571	54571	54660	
Alabama	Autauga County	50	3	6	1	1	54571	54571	54660	
	Baldwin County	50	3	6	1	3	182265	182265	183193	
	Barbour County	50	3	6	1	5	27457	27457	27341	
	Bibb County	50	3	6	1	7	22915	22919	22861	
	Blount County	50	3	6	1	9	57322	57322	57373	
	Bullock County	50	3	6	1	11	10914	10915	10887	
	Butler County	50	3	6	1	13	20947	20946	20944	
	Calhoun County	50	3	6	1	15	118572	118586	118437	
	Chambers County	50	3	6	1	17	34215	34170	34098	
	Cherokee County	50	3	6	1	19	25989	25986	25976	
	Chilton County	50	3	6	1	21	43643	43631	43665	
	Choctaw County	50	3	6	1	23	13859	13858	13841	
	Clarke County	50	3	6	1	25	25833	25840	25767	
	Clay County	50	3	6	1	27	13932	13932	13880	
	Cleburne County	50	3	6	1	29	14972	14972	14973	
	Coffee County	50	3	6	1	31	49948	49948	50177	

STNAME	CTNAME	50	3	6	1	33	54428	Estimates	54514	PO
		SUMLEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	Base 2010	POESTIMATE2010	
	Colbert County	50	3	6	1	33	54428	54428	54514	PO
	Conecuh County	50	3	6	1	35	13228	13228	13208	
	Coosa County	50	3	6	1	37	11539	11758	11758	
	Covington County	50	3	6	1	39	37765	37765	37796	
	Crenshaw County	50	3	6	1	41	13906	13906	13853	
	Cullman County	50	3	6	1	43	80406	80410	80473	
	Dale County	50	3	6	1	45	50251	50251	50358	
	Dallas County	50	3	6	1	47	43820	43820	43803	
	DeKalb County	50	3	6	1	49	71109	71115	71142	
	Elmore County	50	3	6	1	51	79303	79296	79465	
	Escambia County	50	3	6	1	53	38319	38319	38309	
	Etowah County	50	3	6	1	55	104430	104427	104442	
	Fayette County	50	3	6	1	57	17241	17241	17231	
	Franklin County	50	3	6	1	59	31704	31709	31734	

Wisconsin	Washburn County	50	2	3	55	129	15911	15911	15930	
	Washington County	50	2	3	55	131	131887	131885	131967	
	Waukesha County	50	2	3	55	133	389891	389938	390076	
	Waupaca County	50	2	3	55	135	52410	52410	52422	
	Waushara County	50	2	3	55	137	24496	24496	24506	
	Winnebago County	50	2	3	55	139	166994	166994	167059	
	Wood County	50	2	3	55	141	74749	74749	74807	
Wyoming	Albany County	50	4	8	56	1	36299	36299	36428	
	Big Horn County	50	4	8	56	3	11668	11668	11672	
	Campbell County	50	4	8	56	5	46133	46133	46244	
	Carbon County	50	4	8	56	7	15885	15885	15837	
	Converse County	50	4	8	56	9	13833	13833	13826	
	Crook County	50	4	8	56	11	7083	7083	7114	
	Fremont County	50	4	8	56	13	40123	40123	40222	

	Goshen County	SUMLEV50	REGION4	DIVISION8	STATE56	COUNTY15	CENSUS2010POP13249	Estimates Base13247	POPESTIMATE201013408	PO
STNAME	Hot Springs County	50	4	8	56	17	4812	4812	4813	
	Johnson County	50	4	8	56	19	8569	8569	8581	
	Laramie County	50	4	8	56	21	91738	91881	92271	
	Lincoln County	50	4	8	56	23	18106	18106	18091	
	Natrona County	50	4	8	56	25	75450	75450	75472	
	Niobrara County	50	4	8	56	27	2484	2484	2492	
	Park County	50	4	8	56	29	28205	28205	28259	
	Platte County	50	4	8	56	31	8667	8667	8678	
	Sheridan County	50	4	8	56	33	29116	29116	29146	
	Sublette County	50	4	8	56	35	10247	10247	10244	
	Sweetwater County	50	4	8	56	37	43806	43806	43593	
	Teton County	50	4	8	56	39	21294	21294	21297	
	Uinta County	50	4	8	56	41	21118	21118	21102	
	Washakie County	50	4	8	56	43	8533	8533	8545	
	Weston County	50	4	8	56	45	7208	7208	7181	

3142 rows x 98 columns

In [22]:

```
import numpy as np
def min_max(row):
    data = row[['POPESTIMATE2010',
                  'POPESTIMATE2011',
                  'POPESTIMATE2012',
                  'POPESTIMATE2013',
                  'POPESTIMATE2014',
                  'POPESTIMATE2015']]
    return pd.Series({'min': np.min(data), 'max': np.max(data)})
```

In [23]:

```
df.apply(min_max, axis=1)
```

Out[23]:

		min	max
STNAME	CTYNAME		
Alabama	Autauga County	54660.0	55347.0
	Baldwin County	183193.0	203709.0
	Barbour County	26489.0	27341.0
	Bell County	22510.0	22221.0

		Bibb County	22512.0	22861.0
		min	max	
STNAME		Blount County	57373.0	57776.0
		CTYNAME		
		Bullock County	10006.0	10007.0
		Butler County	20154.0	20944.0
		Calhoun County	115620.0	118437.0
		Chambers County	33993.0	34153.0
		Cherokee County	25859.0	26084.0
		Chilton County	43665.0	43943.0
		Choctaw County	13170.0	13841.0
		Clarke County	24675.0	25767.0
		Clay County	13456.0	13880.0
		Cleburne County	14921.0	15072.0
		Coffee County	50177.0	51211.0
		Colbert County	54354.0	54514.0
		Conecuh County	12662.0	13208.0
		Coosa County	10724.0	11758.0
		Covington County	37796.0	38060.0
		Crenshaw County	13853.0	13963.0
		Cullman County	80374.0	82005.0
		Dale County	49501.0	50358.0
		Dallas County	41131.0	43803.0
		DeKalb County	70869.0	71387.0
		Elmore County	79465.0	81468.0
		Escambia County	37784.0	38309.0
		Etowah County	103057.0	104442.0
		Fayette County	16759.0	17231.0
		Franklin County	31507.0	31734.0

Wisconsin		Washburn County	15552.0	15930.0
		Washington County	131967.0	133674.0
		Waukesha County	390076.0	396488.0
		Waupaca County	51945.0	52422.0
		Waushara County	24033.0	24581.0
		Winnebago County	167059.0	169639.0
		Wood County	73435.0	74807.0
Wyoming		Albany County	36428.0	37956.0
		Big Horn County	11672.0	12022.0
		Campbell County	46244.0	49220.0
		Carbon County	15559.0	15856.0
		Converse County	13728.0	14343.0
		Crook County	7114.0	7444.0
		Fremont County	40222.0	41129.0
		Goshen County	13383.0	13666.0
		Hot Springs County	4741.0	4846.0

STNAME	Johnson County	8552.0	8636.0
	Laramie County	92271.0	97121.0
CTYNAME	Lincoln County	17943.0	18722.0
	Natrona County	75472.0	82178.0
	Niobrara County	2475.0	2548.0
	Park County	28259.0	29237.0
	Platte County	8678.0	8812.0
	Sheridan County	29146.0	30020.0
	Sublette County	9899.0	10418.0
	Sweetwater County	43593.0	45162.0
	Teton County	21297.0	23125.0
	Uinta County	20822.0	21102.0
	Washakie County	8316.0	8545.0
	Weston County	7065.0	7234.0

3142 rows × 2 columns

In [24]:

```
import numpy as np
def min_max(row):
    data = row[['POPESTIMATE2010',
                  'POPESTIMATE2011',
                  'POPESTIMATE2012',
                  'POPESTIMATE2013',
                  'POPESTIMATE2014',
                  'POPESTIMATE2015']]
    row['max'] = np.max(data)
    row['min'] = np.min(data)
    return row
df.apply(min_max, axis=1)
```

Out[24]:

SUMLEV		REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	ESTIMATESBASE2010	POPESTIMA
STNAME	CTYNAME							
Alabama	Autauga County	50.0	3.0	6.0	1.0	1.0	54571.0	54571.0
	Baldwin County	50.0	3.0	6.0	1.0	3.0	182265.0	182265.0
	Barbour County	50.0	3.0	6.0	1.0	5.0	27457.0	27457.0
	Bibb County	50.0	3.0	6.0	1.0	7.0	22915.0	22919.0
	Blount County	50.0	3.0	6.0	1.0	9.0	57322.0	57322.0
	Bullock County	50.0	3.0	6.0	1.0	11.0	10914.0	10915.0
	Butler County	50.0	3.0	6.0	1.0	13.0	20947.0	20946.0
	Calhoun County	50.0	3.0	6.0	1.0	15.0	118572.0	118586.0
	Chambers County	50.0	3.0	6.0	1.0	17.0	34215.0	34170.0
	Cherokee County	50.0	3.0	6.0	1.0	19.0	25989.0	25986.0

STNAME	CTYNAME	SUMLEV	REGION	DIVISION	STATE	COUNTY	CENSUS2010POP	ESTIMATESBASE2010	POPESTIMATE2010
		50.0	3.0	6.0	1.0	23.0	13859.0	13858.0	
Alabama	Chilton County	50.0	3.0	6.0	1.0	23.0	13859.0	13858.0	
	Choctaw County	50.0	3.0	6.0	1.0	25.0	25833.0	25840.0	
	Clarke County	50.0	3.0	6.0	1.0	27.0	13932.0	13932.0	
	Clay County	50.0	3.0	6.0	1.0	29.0	14972.0	14972.0	
	Cleburne County	50.0	3.0	6.0	1.0	31.0	49948.0	49948.0	
	Coffee County	50.0	3.0	6.0	1.0	33.0	54428.0	54428.0	
	Colbert County	50.0	3.0	6.0	1.0	35.0	13228.0	13228.0	
	Conecuh County	50.0	3.0	6.0	1.0	37.0	11539.0	11758.0	
	Coosa County	50.0	3.0	6.0	1.0	39.0	37765.0	37765.0	
	Covington County	50.0	3.0	6.0	1.0	41.0	13906.0	13906.0	
	Crenshaw County	50.0	3.0	6.0	1.0	43.0	80406.0	80410.0	
	Cullman County	50.0	3.0	6.0	1.0	45.0	50251.0	50251.0	
	Dale County	50.0	3.0	6.0	1.0	47.0	43820.0	43820.0	
	Dallas County	50.0	3.0	6.0	1.0	49.0	71109.0	71115.0	
	DeKalb County	50.0	3.0	6.0	1.0	51.0	79303.0	79296.0	
	Elmore County	50.0	3.0	6.0	1.0	53.0	38319.0	38319.0	
	Escambia County	50.0	3.0	6.0	1.0	55.0	104430.0	104427.0	1
	Etowah County	50.0	3.0	6.0	1.0	57.0	17241.0	17241.0	
	Fayette County	50.0	3.0	6.0	1.0	59.0	31704.0	31709.0	
...	
Wisconsin	Washburn County	50.0	2.0	3.0	55.0	129.0	15911.0	15911.0	
	Washington County	50.0	2.0	3.0	55.0	131.0	131887.0	131885.0	1
	Waukesha County	50.0	2.0	3.0	55.0	133.0	389891.0	389938.0	3
	Waupaca County	50.0	2.0	3.0	55.0	135.0	52410.0	52410.0	
	Waushara County	50.0	2.0	3.0	55.0	137.0	24496.0	24496.0	
	Winnebago County	50.0	2.0	3.0	55.0	139.0	166994.0	166994.0	1
	Wood County	50.0	2.0	3.0	55.0	141.0	74749.0	74749.0	
Wyoming	Albany County	50.0	4.0	8.0	56.0	1.0	36299.0	36299.0	

STNAME	Big Horn County CTYNAME	SUMLEV 50.0	REGION 4.0	DIVISION 8.0	STATE 56.0	COUNTY 3.0	CENSUS2010POP 11668.0	ESTIMATESBASE2010 11668.0	POPESTIMA
	Campbell County	50.0	4.0	8.0	56.0	5.0	46133.0	46133.0	
	Carbon County	50.0	4.0	8.0	56.0	7.0	15885.0	15885.0	
	Converse County	50.0	4.0	8.0	56.0	9.0	13833.0	13833.0	
	Crook County	50.0	4.0	8.0	56.0	11.0	7083.0	7083.0	
	Fremont County	50.0	4.0	8.0	56.0	13.0	40123.0	40123.0	
	Goshen County	50.0	4.0	8.0	56.0	15.0	13249.0	13247.0	
	Hot Springs County	50.0	4.0	8.0	56.0	17.0	4812.0	4812.0	
	Johnson County	50.0	4.0	8.0	56.0	19.0	8569.0	8569.0	
	Laramie County	50.0	4.0	8.0	56.0	21.0	91738.0	91881.0	
	Lincoln County	50.0	4.0	8.0	56.0	23.0	18106.0	18106.0	
	Natrona County	50.0	4.0	8.0	56.0	25.0	75450.0	75450.0	
	Niobrara County	50.0	4.0	8.0	56.0	27.0	2484.0	2484.0	
	Park County	50.0	4.0	8.0	56.0	29.0	28205.0	28205.0	
	Platte County	50.0	4.0	8.0	56.0	31.0	8667.0	8667.0	
	Sheridan County	50.0	4.0	8.0	56.0	33.0	29116.0	29116.0	
	Sublette County	50.0	4.0	8.0	56.0	35.0	10247.0	10247.0	
	Sweetwater County	50.0	4.0	8.0	56.0	37.0	43806.0	43806.0	
	Teton County	50.0	4.0	8.0	56.0	39.0	21294.0	21294.0	
	Uinta County	50.0	4.0	8.0	56.0	41.0	21118.0	21118.0	
	Washakie County	50.0	4.0	8.0	56.0	43.0	8533.0	8533.0	
	Weston County	50.0	4.0	8.0	56.0	45.0	7208.0	7208.0	

3142 rows x 100 columns



In [25]:

```
rows = ['POPESTIMATE2010',
        'POPESTIMATE2011',
        'POPESTIMATE2012',
        'POPESTIMATE2013',
        'POPESTIMATE2014',
        'POPESTIMATE2015']
df.apply(lambda x: np.max(x[rows]), axis=1)
```

Out[25]:

STNAME CTYNAME

STATE	COUNTY	POPULATION	
Alabama	Autauga County	55347.0	
	Baldwin County	203709.0	
	Barbour County	27341.0	
	Bibb County	22861.0	
	Blount County	57776.0	
	Bullock County	10887.0	
	Butler County	20944.0	
	Calhoun County	118437.0	
	Chambers County	34153.0	
	Cherokee County	26084.0	
	Chilton County	43943.0	
	Choctaw County	13841.0	
	Clarke County	25767.0	
	Clay County	13880.0	
	Cleburne County	15072.0	
	Coffee County	51211.0	
	Colbert County	54514.0	
	Conecuh County	13208.0	
	Coosa County	11758.0	
	Covington County	38060.0	
	Crenshaw County	13963.0	
	Cullman County	82005.0	
	Dale County	50358.0	
	Dallas County	43803.0	
	DeKalb County	71387.0	
	Elmore County	81468.0	
	Escambia County	38309.0	
	Etowah County	104442.0	
	Fayette County	17231.0	
	Franklin County	31734.0	
		...	
	Wisconsin	Washburn County	15930.0
Washington County		133674.0	
Waukesha County		396488.0	
Waupaca County		52422.0	
Waushara County		24581.0	
Winnebago County		169639.0	
Wyoming	Wood County	74807.0	
	Albany County	37956.0	
	Big Horn County	12022.0	
	Campbell County	49220.0	
	Carbon County	15856.0	
	Converse County	14343.0	
	Crook County	7444.0	
	Fremont County	41129.0	
	Goshen County	13666.0	
	Hot Springs County	4846.0	
	Johnson County	8636.0	
	Laramie County	97121.0	
	Lincoln County	18722.0	
	Natrona County	82178.0	
	Niobrara County	2548.0	
	Park County	29237.0	
	Platte County	8812.0	
	Sheridan County	30020.0	
	Sublette County	10418.0	
	Sweetwater County	45162.0	
Teton County	23125.0		
Uinta County	21102.0		
Washakie County	8545.0		
Weston County	7234.0		

Length: 3142, dtype: float64

Group by

In [26]:

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



```
df = pd.read_csv('census.csv')
df = df[df['SUMLEV']==50]
df
```

Out[26]:

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPE
1	50	3	6	1	1	Alabama	Autauga County	54571	54571	
2	50	3	6	1	3	Alabama	Baldwin County	182265	182265	
3	50	3	6	1	5	Alabama	Barbour County	27457	27457	
4	50	3	6	1	7	Alabama	Bibb County	22915	22919	
5	50	3	6	1	9	Alabama	Blount County	57322	57322	
6	50	3	6	1	11	Alabama	Bullock County	10914	10915	
7	50	3	6	1	13	Alabama	Butler County	20947	20946	
8	50	3	6	1	15	Alabama	Calhoun County	118572	118586	
9	50	3	6	1	17	Alabama	Chambers County	34215	34170	
10	50	3	6	1	19	Alabama	Cherokee County	25989	25986	
11	50	3	6	1	21	Alabama	Chilton County	43643	43631	
12	50	3	6	1	23	Alabama	Choctaw County	13859	13858	
13	50	3	6	1	25	Alabama	Clarke County	25833	25840	
14	50	3	6	1	27	Alabama	Clay County	13932	13932	
15	50	3	6	1	29	Alabama	Cleburne County	14972	14972	
16	50	3	6	1	31	Alabama	Coffee County	49948	49948	
17	50	3	6	1	33	Alabama	Colbert County	54428	54428	
18	50	3	6	1	35	Alabama	Conecuh County	13228	13228	
19	50	3	6	1	37	Alabama	Coosa County	11539	11758	
20	50	3	6	1	39	Alabama	Covington County	37765	37765	
21	50	3	6	1	41	Alabama	Crenshaw County	13906	13906	
22	50	3	6	1	43	Alabama	Cullman County	80406	80410	
23	50	3	6	1	45	Alabama	Dale County	50251	50251	
24	50	3	6	1	47	Alabama	Dallas County	43820	43820	
25	50	3	6	1	49	Alabama	DeKalb County	71109	71115	
						Alabama	Elmore County			

3186	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPE
3187	50	4	8	56	35	Wyoming	Sublette County	10247		10247
3188	50	4	8	56	37	Wyoming	Sweetwater County	43806		43806
3189	50	4	8	56	39	Wyoming	Teton County	21294		21294
3190	50	4	8	56	41	Wyoming	Uinta County	21118		21118
3191	50	4	8	56	43	Wyoming	Washakie County	8533		8533
3192	50	4	8	56	45	Wyoming	Weston County	7208		7208

3142 rows x 100 columns

In []:

```
%%timeit -n 10
for state in df['STNAME'].unique():
    avg = np.average(df.where(df['STNAME']==state).dropna()['CENSUS2010POP'])
    print('Counties in state ' + state + ' have an average population of ' + str(avg))
```

In [27]:

```
##%timeit -n 10
for group, frame in df.groupby('STNAME'):
    avg = np.average(frame['CENSUS2010POP'])
    print('Counties in state ' + group + ' have an average population of ' + str(avg))
```

```
Counties in state Alabama have an average population of 71339.34328358209
Counties in state Alaska have an average population of 24490.724137931036
Counties in state Arizona have an average population of 426134.4666666667
Counties in state Arkansas have an average population of 38878.90666666667
Counties in state California have an average population of 642309.5862068966
Counties in state Colorado have an average population of 78581.1875
Counties in state Connecticut have an average population of 446762.125
Counties in state Delaware have an average population of 299311.3333333333
Counties in state District of Columbia have an average population of 601723.0
Counties in state Florida have an average population of 280616.5671641791
Counties in state Georgia have an average population of 60928.63522012578
Counties in state Hawaii have an average population of 272060.2
Counties in state Idaho have an average population of 35626.86363636364
Counties in state Illinois have an average population of 125790.50980392157
Counties in state Indiana have an average population of 70476.10869565218
Counties in state Iowa have an average population of 30771.262626262625
Counties in state Kansas have an average population of 27172.55238095238
Counties in state Kentucky have an average population of 36161.39166666667
Counties in state Louisiana have an average population of 70833.9375
Counties in state Maine have an average population of 83022.5625
Counties in state Maryland have an average population of 240564.66666666666
Counties in state Massachusetts have an average population of 467687.78571428574
Counties in state Michigan have an average population of 119080.0
Counties in state Minnesota have an average population of 60964.65517241379
Counties in state Mississippi have an average population of 36186.54878048781
Counties in state Missouri have an average population of 52077.62608695652
Counties in state Montana have an average population of 17668.125
Counties in state Nebraska have an average population of 19638.075268817203
Counties in state Nevada have an average population of 158855.9411764706
Counties in state New Hampshire have an average population of 131647.0
Counties in state New Jersey have an average population of 418661.61904761905
Counties in state New Mexico have an average population of 62399.36363636364
Counties in state New York have an average population of 312550.03225806454
Counties in state North Carolina have an average population of 95354.83
Counties in state North Dakota have an average population of 12690.396226415094
Counties in state Ohio have an average population of 131096.63636363635
Counties in state Oklahoma have an average population of 40710.04115044115
```

Counties in state Oklahoma have an average population of 48718.844133844133
Counties in state Oregon have an average population of 106418.72222222222
Counties in state Pennsylvania have an average population of 189587.74626865672
Counties in state Rhode Island have an average population of 210513.4
Counties in state South Carolina have an average population of 100551.39130434782
Counties in state South Dakota have an average population of 12336.060606060606
Counties in state Tennessee have an average population of 66801.1052631579
Counties in state Texas have an average population of 98998.27165354331
Counties in state Utah have an average population of 95306.37931034483
Counties in state Vermont have an average population of 44695.78571428572
Counties in state Virginia have an average population of 60111.29323308271
Counties in state Washington have an average population of 172424.10256410256
Counties in state West Virginia have an average population of 33690.8
Counties in state Wisconsin have an average population of 78985.91666666667
Counties in state Wyoming have an average population of 24505.478260869564

In [28]:

```
df.head()
```

Out[28]:

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE2010	POPESTIM/
1	50	3	6	1	1	Alabama	Autauga County	54571	54571	
2	50	3	6	1	3	Alabama	Baldwin County	182265	182265	
3	50	3	6	1	5	Alabama	Barbour County	27457	27457	
4	50	3	6	1	7	Alabama	Bibb County	22915	22919	
5	50	3	6	1	9	Alabama	Blount County	57322	57322	

5 rows x 100 columns



In [29]:

```
df = df.set_index('STNAME')

def fun(item):
    if item[0]<'M':
        return 0
    if item[0]<'Q':
        return 1
    return 2

for group, frame in df.groupby(fun):
    print('There are ' + str(len(frame)) + ' records in group ' + str(group) + ' for processing.')
```

There are 1177 records in group 0 for processing.
There are 1134 records in group 1 for processing.
There are 831 records in group 2 for processing.

In [30]:

```
df = pd.read_csv('census.csv')
df = df[df['SUMLEV']==50]
```

In [31]:

```
df.groupby('STNAME').agg({'CENSUS2010POP': np.average})
```

Out[31]:

STNAME	CENSUS2010POP
Alabama	71339.343284
Alaska	24490.724138
Arizona	426134.466667
Arkansas	38878.906667
California	642309.586207
Colorado	78581.187500
Connecticut	446762.125000
Delaware	299311.333333
District of Columbia	601723.000000
Florida	280616.567164
Georgia	60928.635220
Hawaii	272060.200000
Idaho	35626.863636
Illinois	125790.509804
Indiana	70476.108696
Iowa	30771.262626
Kansas	27172.552381
Kentucky	36161.391667
Louisiana	70833.937500
Maine	83022.562500
Maryland	240564.666667
Massachusetts	467687.785714
Michigan	119080.000000
Minnesota	60964.655172
Mississippi	36186.548780
Missouri	52077.626087
Montana	17668.125000
Nebraska	19638.075269
Nevada	158855.941176
New Hampshire	131647.000000
New Jersey	418661.619048
New Mexico	62399.363636
New York	312550.032258
North Carolina	95354.830000
North Dakota	12690.396226
Ohio	131096.636364
Oklahoma	48718.844156
Oregon	106418.722222
Pennsylvania	189587.746269
Rhode Island	210513.400000
South Carolina	100551.391304
South Dakota	12336.060606
Tennessee	66801.105263
Texas	98998.271654

STNAME	CENSUS2010POP
Utah	95306.379310
Vermont	44695.785714
Virginia	60111.293233
Washington	172424.102564
West Virginia	33690.800000
Wisconsin	78985.916667
Wyoming	24505.478261

In [32]:

```
#print(df.groupby('Category').apply(lambda df,a,b: sum(df[a] * df[b]), 'Weight (oz.)', 'Q
uantity'))

# Or alternatively without using a lambda:
# def totalweight(df, w, q):
#     return sum(df[w] * df[q])
#
# print(df.groupby('Category').apply(totalweight, 'Weight (oz.)', 'Quantity'))
```

In [33]:

```
print(type(df.groupby(level=0) ['POPESTIMATE2010', 'POPESTIMATE2011']))
print(type(df.groupby(level=0) ['POPESTIMATE2010']))
```

<class 'pandas.core.groupby.groupby.DataFrameGroupBy'>
<class 'pandas.core.groupby.groupby.SeriesGroupBy'>

In [34]:

```
(df.set_index('STNAME').groupby(level=0) ['CENSUS2010POP']
    .agg({'avg': np.average, 'sum': np.sum}))
```

C:\Users\Parth\Anaconda3\lib\site-packages\ipykernel_launcher.py:2: FutureWarning: using a dict on a Series for aggregation is deprecated and will be removed in a future version

Out[34]:

	avg	sum
STNAME		
Alabama	71339.343284	4779736
Alaska	24490.724138	710231
Arizona	426134.466667	6392017
Arkansas	38878.906667	2915918
California	642309.586207	37253956
Colorado	78581.187500	5029196
Connecticut	446762.125000	3574097
Delaware	299311.333333	897934
District of Columbia	601723.000000	601723
Florida	280616.567164	18801310
Georgia	60928.635220	9687653
Hawaii	272060.200000	1360301
Idaho	35626.863636	1567582
Illinois	125790.509804	12830632
Indiana	70476.108696	6483802

Iowa	27172.552381	2853118
STNAME		
Kentucky	36161.391667	4339367
Louisiana	70833.937500	4533372
Maine	83022.562500	1328361
Maryland	240564.666667	5773552
Massachusetts	467687.785714	6547629
Michigan	119080.000000	9883640
Minnesota	60964.655172	5303925
Mississippi	36186.548780	2967297
Missouri	52077.626087	5988927
Montana	17668.125000	989415
Nebraska	19638.075269	1826341
Nevada	158855.941176	2700551
New Hampshire	131647.000000	1316470
New Jersey	418661.619048	8791894
New Mexico	62399.363636	2059179
New York	312550.032258	19378102
North Carolina	95354.830000	9535483
North Dakota	12690.396226	672591
Ohio	131096.636364	11536504
Oklahoma	48718.844156	3751351
Oregon	106418.722222	3831074
Pennsylvania	189587.746269	12702379
Rhode Island	210513.400000	1052567
South Carolina	100551.391304	4625364
South Dakota	12336.060606	814180
Tennessee	66801.105263	6346105
Texas	98998.271654	25145561
Utah	95306.379310	2763885
Vermont	44695.785714	625741
Virginia	60111.293233	7994802
Washington	172424.102564	6724540
West Virginia	33690.800000	1852994
Wisconsin	78985.916667	5686986
Wyoming	24505.478261	563626

In [35]:

```
(df.set_index('STNAME').groupby(level=0) ['POPESTIMATE2010', 'POPESTIMATE2011']
    .agg({'avg': np.average, 'sum': np.sum}))
```

C:\Users\Parth\Anaconda3\lib\site-packages\pandas\core\groupby\groupby.py:4656: FutureWarning: using a dict with renaming is deprecated and will be removed in a future version
return super(DataFrameGroupBy, self).aggregate(arg, *args, **kwargs)

Out[35]:

avg		sum	
POPESTIMATE2010	POPESTIMATE2011	POPESTIMATE2010	POPESTIMATE2011

STNAME	POPESTIMATE2010	POPESTIMATE2011	POPESTIMATE2010	POPESTIMATE2011
	avg	sum		
Alabama	71420.313433	71658.328358	4785161	4801108
Alaska	24621.413793	24921.379310	714021	722720
Arizona	427213.866667	431248.800000	6408208	6468732
Arkansas	38965.253333	39180.506667	2922394	2938538
California	643691.017241	650000.586207	37334079	37700034
Colorado	78878.968750	79991.875000	5048254	5119480
Connecticut	447464.625000	448719.875000	3579717	3589759
Delaware	299930.333333	302638.666667	899791	907916
District of Columbia	605126.000000	620472.000000	605126	620472
Florida	281341.641791	285157.208955	18849890	19105533
Georgia	61090.905660	61712.452830	9713454	9812280
Hawaii	272796.000000	275645.400000	1363980	1378227
Idaho	35704.227273	36003.045455	1570986	1584134
Illinois	125894.598039	126096.882353	12841249	12861882
Indiana	70549.891304	70835.271739	6490590	6516845
Iowa	30815.090909	30963.525253	3050694	3065389
Kansas	27226.895238	27332.542857	2858824	2869917
Kentucky	36232.808333	36399.016667	4347937	4367882
Louisiana	71014.859375	71490.328125	4544951	4575381
Maine	82980.937500	83016.062500	1327695	1328257
Maryland	241183.708333	243507.125000	5788409	5844171
Massachusetts	468931.142857	472271.214286	6565036	6611797
Michigan	119004.445783	118995.048193	9877369	9876589
Minnesota	61044.862069	61472.632184	5310903	5348119
Mississippi	36223.365854	36317.060976	2970316	2977999
Missouri	52139.582609	52265.973913	5996052	6010587
Montana	17690.053571	17816.892857	990643	997746
Nebraska	19677.688172	19810.569892	1830025	1842383
Nevada	159025.882353	159930.529412	2703440	2718819
New Hampshire	131670.800000	131834.400000	1316708	1318344
New Jersey	419232.428571	421092.095238	8803881	8842934
New Mexico	62567.909091	62976.545455	2064741	2078226
New York	312950.322581	314890.354839	19402920	19523202
North Carolina	95589.790000	96510.250000	9558979	9651025
North Dakota	12726.981132	12930.679245	674530	685326
Ohio	131145.068182	131198.204545	11540766	11545442
Oklahoma	48825.922078	49176.961039	3759596	3786626
Oregon	106610.333333	107458.583333	3837972	3868509
Pennsylvania	189731.552239	190226.895522	12712014	12745202
Rhode Island	210643.800000	210371.200000	1053219	1051856
South Carolina	100780.304348	101581.152174	4635894	4672733
South Dakota	12368.166667	12489.227273	816299	824289
Tennessee	66911.421053	67351.663158	6356585	6398408
Texas	90297.255006	90400.926772	95244262	95654464

	avg	sum		
Utah	95704.344828	97118.620690	2775426	2816440
POPESTIMATE2010		POPESTIMATE2011	POPESTIMATE2010	POPESTIMATE2011
Vermont	44713.142857	44763.357143	625984	626687
STNAME				
Virginia	60344.263158	60983.330827	8025787	8110783
Washington	172898.974359	174954.589744	6743060	6823229
West Virginia	33713.181818	33726.327273	1854225	1854948
Wisconsin	79030.611111	79301.666667	5690204	5709720
Wyoming	24544.173913	24685.565217	564516	567768

In [36]:

```
(df.set_index('STNAME').groupby(level=0) ['POPESTIMATE2010', 'POPESTIMATE2011']
    .agg({'POPESTIMATE2010': np.average, 'POPESTIMATE2011': np.sum}))
```

Out[36]:

	POPESTIMATE2010	POPESTIMATE2011
STNAME		
Alabama	71420.313433	4801108
Alaska	24621.413793	722720
Arizona	427213.866667	6468732
Arkansas	38965.253333	2938538
California	643691.017241	37700034
Colorado	78878.968750	5119480
Connecticut	447464.625000	3589759
Delaware	299930.333333	907916
District of Columbia	605126.000000	620472
Florida	281341.641791	19105533
Georgia	61090.905660	9812280
Hawaii	272796.000000	1378227
Idaho	35704.227273	1584134
Illinois	125894.598039	12861882
Indiana	70549.891304	6516845
Iowa	30815.090909	3065389
Kansas	27226.895238	2869917
Kentucky	36232.808333	4367882
Louisiana	71014.859375	4575381
Maine	82980.937500	1328257
Maryland	241183.708333	5844171
Massachusetts	468931.142857	6611797
Michigan	119004.445783	9876589
Minnesota	61044.862069	5348119
Mississippi	36223.365854	2977999
Missouri	52139.582609	6010587
Montana	17690.053571	997746
Nebraska	19677.688172	1842383
Nevada	159025.882353	2718819
New Hampshire	131670.800000	1318344

New Jersey	419232.428571	8842934
New Mexico	62567.909091	2078226
New York	312950.322581	19523202
North Carolina	95589.790000	9651025
North Dakota	12726.981132	685326
Ohio	131145.068182	11545442
Oklahoma	48825.922078	3786626
Oregon	106610.333333	3868509
Pennsylvania	189731.552239	12745202
Rhode Island	210643.800000	1051856
South Carolina	100780.304348	4672733
South Dakota	12368.166667	824289
Tennessee	66911.421053	6398408
Texas	99387.255906	25654464
Utah	95704.344828	2816440
Vermont	44713.142857	626687
Virginia	60344.263158	8110783
Washington	172898.974359	6823229
West Virginia	33713.181818	1854948
Wisconsin	79030.611111	5709720
Wyoming	24544.173913	567768

Scales

In [37]:

```
df = pd.DataFrame(['A+', 'A', 'A-', 'B+', 'B', 'B-', 'C+', 'C', 'C-', 'D+', 'D'],
                  index=['excellent', 'excellent', 'excellent', 'good', 'good', 'good',
                        'ok', 'ok', 'ok', 'poor', 'poor'])
df.rename(columns={0: 'Grades'}, inplace=True)
df
```

Out[37]:

Grades	
excellent	A+
excellent	A
excellent	A-
good	B+
good	B
good	B-
ok	C+
ok	C
ok	C-
poor	D+
poor	D

In [38]:

```
df['Grades'].astype('category').head()
```

Out[38]:

```
excellent    A+
excellent    A
excellent    A-
good         B+
good         B
Name: Grades, dtype: category
Categories (11, object): [A, A+, A-, B, ..., C+, C-, D, D+]
```

In [39]:

```
grades = df['Grades'].astype('category',
                             categories=['D', 'D+', 'C-', 'C', 'C+', 'B-', 'B', 'B+', 'A-', 'A', 'A+'],
                             ordered=True)
grades.head()
```

C:\Users\Parth\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: FutureWarning: specifying 'categories' or 'ordered' in .astype() is deprecated; pass a CategoricalDtype instead

This is separate from the ipykernel package so we can avoid doing imports until

Out[39]:

```
excellent    A+
excellent    A
excellent    A-
good         B+
good         B
Name: Grades, dtype: category
Categories (11, object): [D < D+ < C- < C ... B+ < A- < A < A+]
```

In [40]:

```
grades > 'C'
```

Out[40]:

```
excellent    True
excellent    True
excellent    True
good         True
good         True
good         True
ok           True
ok           False
ok           False
poor         False
poor         False
Name: Grades, dtype: bool
```

In [41]:

```
df = pd.read_csv('census.csv')
df = df[df['SUMLEV']==50]
df = df.set_index('STNAME').groupby(level=0)['CENSUS2010POP'].agg({'avg': np.average})
pd.cut(df['avg'],10)
```

C:\Users\Parth\Anaconda3\lib\site-packages\ipykernel_launcher.py:3: FutureWarning: using a dict on a Series for aggregation is deprecated and will be removed in a future version

This is separate from the ipykernel package so we can avoid doing imports until

Out[41]:

```
STNAME
Alabama      (11706.087, 75333.413]
Alaska       (11706.087, 75333.413]
Arizona      (390320.176, 453317.529]
Arkansas     (11706.087, 75333.413]
California   (579312.234, 642309.586]
Colorado     (75333.413, 138330.766]
Connecticut  (200220.176, 453317.529]
```

```
Connecticut (390320.176, 453317.529]
Delaware (264325.471, 327322.823]
District of Columbia (579312.234, 642309.586]
Florida (264325.471, 327322.823]
Georgia (11706.087, 75333.413]
Hawaii (264325.471, 327322.823]
Idaho (11706.087, 75333.413]
Illinois (75333.413, 138330.766]
Indiana (11706.087, 75333.413]
Iowa (11706.087, 75333.413]
Kansas (11706.087, 75333.413]
Kentucky (11706.087, 75333.413]
Louisiana (11706.087, 75333.413]
Maine (75333.413, 138330.766]
Maryland (201328.118, 264325.471]
Massachusetts (453317.529, 516314.881]
Michigan (75333.413, 138330.766]
Minnesota (11706.087, 75333.413]
Mississippi (11706.087, 75333.413]
Missouri (11706.087, 75333.413]
Montana (11706.087, 75333.413]
Nebraska (11706.087, 75333.413]
Nevada (138330.766, 201328.118]
New Hampshire (75333.413, 138330.766]
New Jersey (390320.176, 453317.529]
New Mexico (11706.087, 75333.413]
New York (264325.471, 327322.823]
North Carolina (75333.413, 138330.766]
North Dakota (11706.087, 75333.413]
Ohio (75333.413, 138330.766]
Oklahoma (11706.087, 75333.413]
Oregon (75333.413, 138330.766]
Pennsylvania (138330.766, 201328.118]
Rhode Island (201328.118, 264325.471]
South Carolina (75333.413, 138330.766]
South Dakota (11706.087, 75333.413]
Tennessee (11706.087, 75333.413]
Texas (75333.413, 138330.766]
Utah (75333.413, 138330.766]
Vermont (11706.087, 75333.413]
Virginia (11706.087, 75333.413]
Washington (138330.766, 201328.118]
West Virginia (11706.087, 75333.413]
Wisconsin (75333.413, 138330.766]
Wyoming (11706.087, 75333.413]
Name: avg, dtype: category
Categories (10, interval[float64]): [(11706.087, 75333.413] < (75333.413, 138330.766] < (138330.766, 201328.118] < (201328.118, 264325.471] ... (390320.176, 453317.529] < (453317.529, 516314.881] < (516314.881, 579312.234] < (579312.234, 642309.586]]
```

Pivot Tables

In [42]:

```
#http://open.canada.ca/data/en/dataset/98f1a129-f628-4ce4-b24d-6f16bf24dd64
df = pd.read_csv('cars.csv')
```

In [43]:

```
df.head()
```

Out[43]:

	YEAR	Make	Model	Size	(kW)	Unnamed: 5	TYPE	CITY (kWh/100 km)	HWY (kWh/100 km)	COMB (kWh/100 km)	CITY (Le/100 km)	HWY (Le/100 km)
0	2012	MINI	Cooper S	SUBCOMPACT	49	A1	B	16.9	21.4	18.7	1.9	2.4
1	2012	NISSAN	LEAF	MID-SIZE	80	A1	B	19.3	23.0	21.1	2.2	2.6

2	2013	FORD	ELECTRIC	COMPACT	107	A1	B	CITY	HWY	COMB	CITY	HWY
3	2013	MITSUBISHI	i-MiEV	SUBCOMPACT	49	A1	B	(kWh/100	(kWh/100	(kWh/100	(Le/100	(Le/100

4	2013	NISSAN	LEAF	MID-SIZE	80	A1	B	19.3	23.0	21.1	2.2	2.6
---	------	--------	------	----------	----	----	---	------	------	------	-----	-----

In [44]:

```
df.pivot_table(values='(kW)', index='YEAR', columns='Make', aggfunc=np.mean)
```

Out[44]:

Make	BMW	CHEVROLET	FORD	KIA	MITSUBISHI	NISSAN	SMART	TESLA
YEAR								
2012	NaN	NaN	NaN	NaN	49.0	80.0	NaN	NaN
2013	NaN	NaN	107.0	NaN	49.0	80.0	35.0	280.000000
2014	NaN	104.0	107.0	NaN	49.0	80.0	35.0	268.333333
2015	125.0	104.0	107.0	81.0	49.0	80.0	35.0	320.666667
2016	125.0	104.0	107.0	81.0	49.0	80.0	35.0	409.700000

In [45]:

```
df.pivot_table(values='(kW)', index='YEAR', columns='Make', aggfunc=[np.mean,np.min], margins=True)
```

Out[45]:

mean										amin		
Make	BMW	CHEVROLET	FORD	KIA	MITSUBISHI	NISSAN	SMART	TESLA	All	BMW	CHEVROLET	FORD
YEAR												
2012	NaN	NaN	NaN	NaN	49.0	80.0	NaN	NaN	64.500000	NaN	NaN	NaN
2013	NaN	NaN	107.0	NaN	49.0	80.0	35.0	280.000000	158.444444	NaN	NaN	107.0
2014	NaN	104.0	107.0	NaN	49.0	80.0	35.0	268.333333	135.000000	NaN	104.0	107.0
2015	125.0	104.0	107.0	81.0	49.0	80.0	35.0	320.666667	181.428571	125.0	104.0	107.0
2016	125.0	104.0	107.0	81.0	49.0	80.0	35.0	409.700000	252.263158	125.0	104.0	107.0
All	125.0	104.0	107.0	81.0	49.0	80.0	35.0	345.478261	190.622642	125.0	104.0	107.0

Date Functionality in Pandas

In [2]:

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

Timestamp

In [3]:

```
pd.Timestamp('9/1/2016 10:05AM')
```

Out[3]:

Timestamp('2016-09-01 10:05:00')

Period

In [4]:

```
pd.Period('1/2016')
```

Out[4]:

```
Period('2016-01', 'M')
```

In [5]:

```
pd.Period('3/5/2016')
```

Out[5]:

```
Period('2016-03-05', 'D')
```

DatetimeIndex

In [6]:

```
t1 = pd.Series(list('abc'), [pd.Timestamp('2016-09-01'), pd.Timestamp('2016-09-02'), pd.Timestamp('2016-09-03')])
t1
```

Out[6]:

```
2016-09-01    a
2016-09-02    b
2016-09-03    c
dtype: object
```

In [7]:

```
type(t1.index)
```

Out[7]:

```
pandas.tseries.index.DatetimeIndex
```

PeriodIndex

In [8]:

```
t2 = pd.Series(list('def'), [pd.Period('2016-09'), pd.Period('2016-10'), pd.Period('2016-11')])
t2
```

Out[8]:

```
2016-09    d
2016-10    e
2016-11    f
Freq: M, dtype: object
```

In [9]:

```
type(t2.index)
```

Out[9]:

```
pandas.tseries.period.PeriodIndex
```

Converting to Datetime

In [10]:

```
d1 = ['2 June 2013', 'Aug 29, 2014', '2015-06-26', '7/12/16']
ts3 = pd.DataFrame(np.random.randint(10, 100, (4,2)), index=d1, columns=list('ab'))
ts3
```

```
Out[10]:
```

	a	b
2 June 2013	16	46
Aug 29, 2014	14	66
2015-06-26	59	99
7/12/16	27	17

```
In [11]:
```

```
ts3.index = pd.to_datetime(ts3.index)
ts3
```

```
Out[11]:
```

	a	b
2013-06-02	16	46
2014-08-29	14	66
2015-06-26	59	99
2016-07-12	27	17

```
In [12]:
```

```
pd.to_datetime('4.7.12', dayfirst=True)
```

```
Out[12]:
```

```
Timestamp('2012-07-04 00:00:00')
```

Timedeltas

```
In [13]:
```

```
pd.Timestamp('9/3/2016') - pd.Timestamp('9/1/2016')
```

```
Out[13]:
```

```
Timedelta('2 days 00:00:00')
```

```
In [14]:
```

```
pd.Timestamp('9/2/2016 8:10AM') + pd.Timedelta('12D 3H')
```

```
Out[14]:
```

```
Timestamp('2016-09-14 11:10:00')
```

Working with Dates in a Dataframe

```
In [15]:
```

```
dates = pd.date_range('10-01-2016', periods=9, freq='2W-SUN')
dates
```

```
Out[15]:
```

```
DatetimeIndex(['2016-10-02', '2016-10-16', '2016-10-30', '2016-11-13',
               '2016-11-27', '2016-12-11', '2016-12-25', '2017-01-08',
               '2017-01-22'],
              dtype='datetime64[ns]', freq='2W-SUN')
```

```
In [16]:
```

```
df = pd.DataFrame({'Count 1': 100 + np.random.randint(-5, 10, 9).cumsum(),
```

```
'Count 2': 120 + np.random.randint(-5, 10, 9)}, index=dates)
df
```

Out[16]:

	Count 1	Count 2
2016-10-02	104	125
2016-10-16	109	122
2016-10-30	111	127
2016-11-13	117	126
2016-11-27	114	126
2016-12-11	109	121
2016-12-25	105	126
2017-01-08	105	125
2017-01-22	101	123

In [17]:

```
df.index.weekday_name
```

Out[17]:

```
array(['Sunday', 'Sunday', 'Sunday', 'Sunday', 'Sunday', 'Sunday',
       'Sunday', 'Sunday', 'Sunday'], dtype=object)
```

In [18]:

```
df.diff()
```

Out[18]:

	Count 1	Count 2
2016-10-02	NaN	NaN
2016-10-16	5.0	-3.0
2016-10-30	2.0	5.0
2016-11-13	6.0	-1.0
2016-11-27	-3.0	0.0
2016-12-11	-5.0	-5.0
2016-12-25	-4.0	5.0
2017-01-08	0.0	-1.0
2017-01-22	-4.0	-2.0

In [19]:

```
df.resample('M').mean()
```

Out[19]:

	Count 1	Count 2
2016-10-31	108.0	124.666667
2016-11-30	115.5	126.000000
2016-12-31	107.0	123.500000
2017-01-31	103.0	124.000000

In [20]:

```
df['2017']
```

Out[20]:

	Count 1	Count 2
2017-01-08	105	125
2017-01-22	101	123

In [21]:

```
df['2016-12']
```

Out[21]:

	Count 1	Count 2
2016-12-11	109	121
2016-12-25	105	126

In [22]:

```
df['2016-12':]
```

Out[22]:

	Count 1	Count 2
2016-12-11	109	121
2016-12-25	105	126
2017-01-08	105	125
2017-01-22	101	123

In []:

```
df.asfreq('W', method='ffill')
```

In []:

```
import matplotlib.pyplot as plt
%matplotlib inline

df.plot()
```