

```
In [1]: #I think this data set will be much more interesting than the plane crashes one (also more nicely formatted)
#This data set contains data for all of the cities and states in the union, and many of their population demographic features
#over the last several years

#There is no missing data
#The Data is a combination of strings (State Name and County Name), and intergers (everything else)
#http://www.census.gov/popest/data/counties/totals/2014/files/CO-EST2014-alldata.pdf (link to descriptions of column headers)
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In [85]: import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
import vincent #rapid mapping tool

%matplotlib inline
```

```
In [95]: data = pd.read_table("population_data_2014.csv", sep = ',', parse_dates = True)
state_only = pd.DataFrame(columns = data.columns)
counties_only = pd.DataFrame(columns = data.columns)
for index, row in data.iterrows():
    if row.STNAME == row.CTYNAME and row.STNAME != "District of Columbia":
        #print "True!"
        state_only = state_only.append(row)
    elif row.STNAME != row.CTYNAME and row.STNAME != "District of Columbia":
        counties_only = counties_only.append(row)
```

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In [138]: counties_only
```

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Out[138]:
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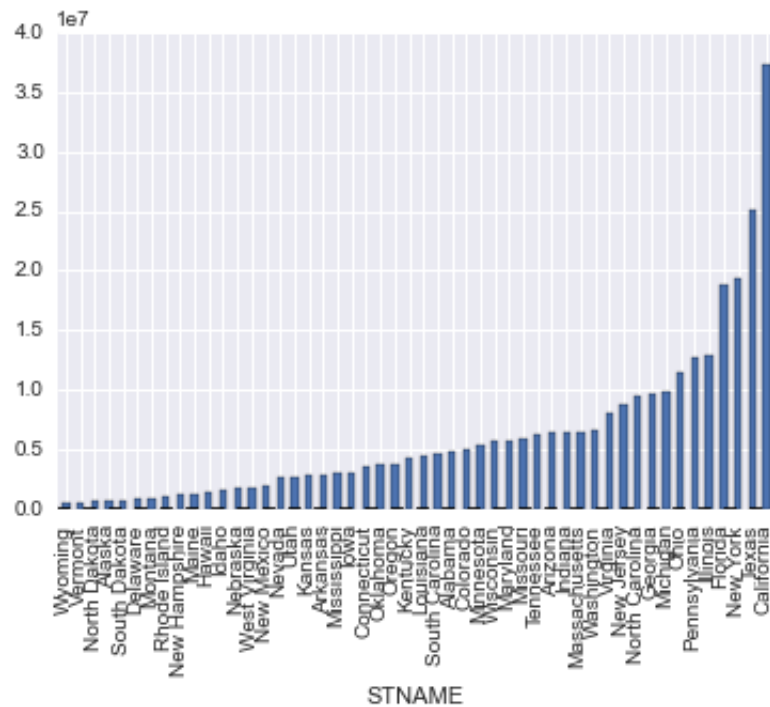
	index	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	CENSUS2010POP	ESTIMATESBASE
0	1	50	3	6	1	1	Alabama	Autauga	54571	54571
1	2	50	3	6	1	3	Alabama	Baldwin	182265	182265
2	3	50	3	6	1	5	Alabama	Barbour	27457	27457
3	4	50	3	6	1	7	Alabama	Bibb	22915	22919

4	5	50	3	6	1	9	Alabama	Blount	57322	57322
5	6	50	3	6	1	11	Alabama	Bullock	10914	10915
6	7	50	3	6	1	13	Alabama	Butler	20947	20946
7	8	50	3	6	1	15	Alabama	alhoun	118572	118586
8	9	50	3	6	1	17	Alabama	hambers	34215	34170
9	10	50	3	6	1	19	Alabama	herokee	25989	25986
10	11	50	3	6	1	21	Alabama	hilton	43643	43631
11	12	50	3	6	1	23	Alabama	hocktaw	13859	13858
12	13	50	3	6	1	25	Alabama	larke	25833	25840
13	14	50	3	6	1	27	Alabama	lay	13932	13932
14	15	50	3	6	1	29	Alabama	leburne	14972	14972
15	16	50	3	6	1	31	Alabama	ffee	49948	49948
16	17	50	3	6	1	33	Alabama	lbert	54428	54428
17	18	50	3	6	1	35	Alabama	ecuh	13228	13228
18	19	50	3	6	1	37	Alabama	sa	11539	11758
19	20	50	3	6	1	39	Alabama	vington	37765	37765
20	21	50	3	6	1	41	Alabama	renshaw	13906	13906
21	22	50	3	6	1	43	Alabama	llman	80406	80410
22	23	50	3	6	1	45	Alabama	Dale	50251	50251
23	24	50	3	6	1	47	Alabama	Dallas	43820	43820
24	25	50	3	6	1	49	Alabama	DeKalb	71109	71115
25	26	50	3	6	1	51	Alabama	Elmore	79303	79296
26	27	50	3	6	1	53	Alabama	Escambia	38319	38319
27	28	50	3	6	1	55	Alabama	Etowah	104430	104427
28	29	50	3	6	1	57	Alabama	Fayette	17241	17241
29	30	50	3	6	1	59	Alabama	Franklin	31704	31709

...	...	...	...	...	...	...	...	...	...
<b>3111</b>	3162	50	2	3	55	129	Wisconsin	Washburn	15911
<b>3112</b>	3163	50	2	3	55	131	Wisconsin	Washington	131887

```
In [135]: state_only = state_only.set_index(state_only.STNAME).sort(['CENSUS2010POP'])
state_only.CENSUS2010POP.plot(kind = 'bar')
#the most populus state is California, by quite a large margin
```

```
Out[135]: <matplotlib.axes._subplots.AxesSubplot at 0x11016eb50>
```



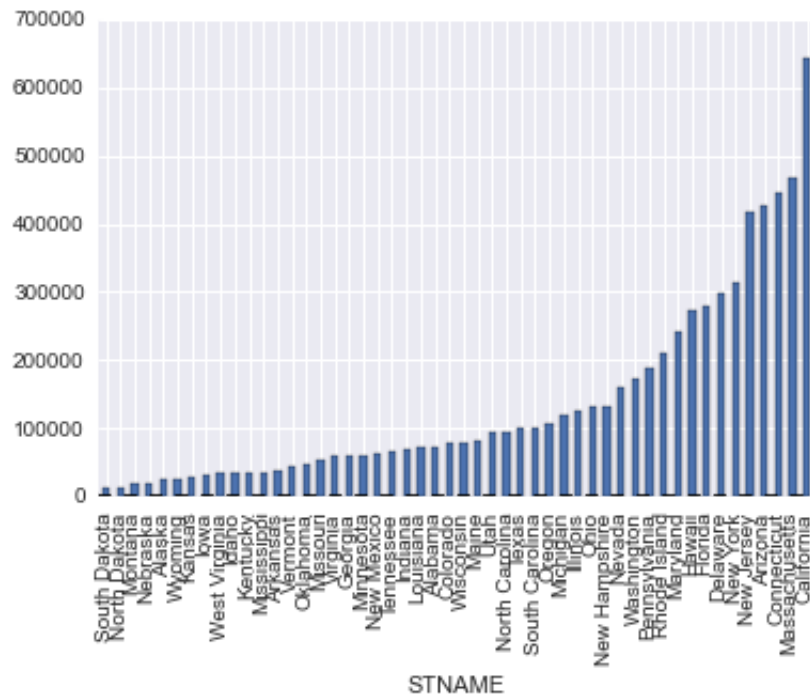
```
In [96]: counties_only = counties_only.reset_index(drop=False)
counties_only.CTYNAME = counties_only.CTYNAME.map(lambda x: x.strip('County'))
```

```
In [137]: counties_only.columns
```

```
Out[137]: Index([u'index', u'SUMLEV', u'REGION', u'DIVISION', u'STATE', u'COUNTRY', u'STNAME', u'CTYNAME', u'CENSUS2010POP', u'ESTIMATESBASE2010', u'POPESTIMATE2010', u'POPESTIMATE2011', u'POPESTIMATE2012', u'POPESTIMATE2013', u'POPESTIMATE2014', u'NPOPCHG_2010', u'NPOPCHG_2011', u'NPOPCHG_2012', u'NPOPCHG_2013', u'NPOPCHG_2014', u'BIRTHS2010', u'BIRTHS2011', u'BIRTHS2012', u'BIRTHS2013', u'BIRTHS2014', u'DEATHS2010', u'DEATHS2011', u'DEATHS2012', u'DEATHS2013', u'DEATHS2014', u'NATURALINC2010', u'NATURALINC2011', u'NATURALINC2012', u'NATURALINC2013', u'NATURALINC2014', u'INTERNATIONALMIG2010', u'INTERNATIONALMIG2011', u'INTERNATIONALMIG2012', u'INTERNATIONALMIG2013', u'INTERNATIONALMIG2014', u'DOMESTICMIG2010', u'DOMESTICMIG2011', u'DOMESTICMIG2012', u'DOMESTICMIG2013', u'DOMESTICMIG2014', u'NETMIG2010', u'NETMIG2011', u'NETMIG2012', u'NETMIG2013', u'NETMIG2014', u'RESIDUAL2010', u'RESIDUAL2011', u'RESIDUAL2012', u'RESIDUAL2013', u'RESIDUAL2014', u'GQESTIMATESBASE2010', u'GQESTIMATES2010', u'GQESTIMATES2011', u'GQESTIMATES2012', u'GQESTIMATES2013', u'GQESTIMATES2014', u'RBIRTH2011', u'RBIRTH2012', u'RBIRTH2013', u'RBIRTH2014', u'RDEATH2011', u'RDEATH2012', u'RDEATH2013', u'RDEATH2014', u'RNATURALINC2011', u'RNATURALINC2012', u'RNATURALINC2013', u'RNATURALINC2014', u'RINTERNATIONALMIG2011', u'RINTERNATIONALMIG2012', u'RINTERNATIONALMIG2013', u'RINTERNATIONALMIG2014', u'RDOMESTICMIG2011', u'RDOMESTICMIG2012', u'RDOMESTICMIG2013', u'RDOMESTICMIG2014', u'RNETMIG2011', u'RNETMIG2012', u'RNETMIG2013', u'RNETMIG2014'], dtype='object')
```

```
In [136]: counties_only.groupby(counties_only.STNAME).mean().sort(['CENSUS2010POP']).CENSUS2010POP.plot(kind='bar')  
#california also has, by far, the most populous county(ies) in the union, on average
```

Out[136]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1169082d0>



In [101]:

```
In []: #the .pdf in the folder has all the more informative headers for these columns
#questions to ask:
#where should you invest in real-estate based on people moving in?
#Where are populations growing the fastest?
#does immigration have a great affect on population increase?
#which state has the most people moving around within the state?
#In which state/county is the death rate outpacing the birth rate? Is the population declining
or are there enough
#immigrants to support the population?
```

In []: