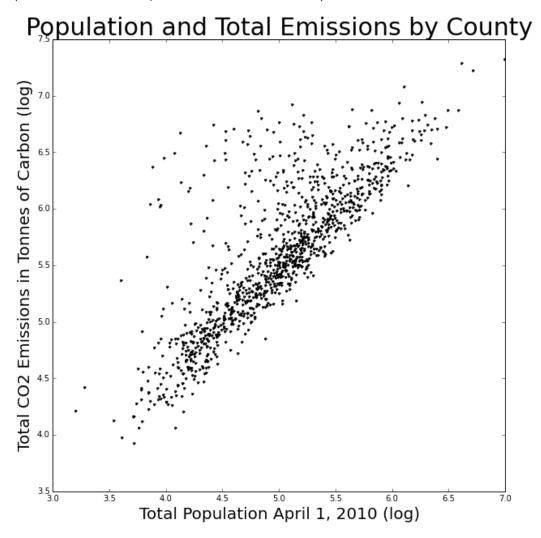
```
In [20]: | from __future__ import print_function
          _Author__ = 'dp1618'
         import pylab as pl
         import numpy as np
         import pandas as pd
         import statsmodels.formula.api as sm
         import statsmodels.api as sml
         import csv
         from scipy import stats
         %pylab inline
         Populating the interactive namespace from numpy and matplotlib
 In [2]: #Import Files
         dfpop = pd.read csv('Population by FIPS County.csv')
         dfpop['FIPS'] = dfpop['County FIPS']
         dfpop.columns
         #print (dfpop)
 Out[2]: Index([u'Area Name-Legal/Statistical Area Description', u'Qualifying Name',
                u'Area (Land)', u'Area (Water)', u'Summary Level',
                u'Geographic Component', u'Region', u'Division', u'County FIPS',
                u'State (FIPS)', u'County', u'Total Population', u'Area Total',
                u'Area Total: Area (Land)', u'Area Total: Area (Water)', u'FIPS'],
               dtype='object')
 In [3]: dfemit = pd.read csv('CountiesBySector.csv', skiprows = 14, header = True)
         dfemit.columns
         #print (dfemit)
 Out[3]: Index([u'State', u' County', u'FIPS', u' Total', u'Unnamed: 4', u'Commercial',
                u'Industrial', u'Residential', u'Electricity Prod', u'Onroad',
                u'Cement', u'Aircraft', u'Airborne', u'Nonroad'],
               dtype='object')
 In [4]: dfMSA = pd.read csv('CountiesMSACodes.csv', skiprows = 2)
         dfMSA.columns
         #print (dfMSA)
 Out[4]: Index([u'CBSA Code', u'Metro Division Code', u'CSA Code', u'CBSA Title',
                u'Level of CBSA', u'Status, 1=metro 2=micro',
                u'Metropolitan Division Title', u'CSA Title', u'Component Name',
                u'State', u'FIPS'],
               dtype='object')
 In [5]: #Join the counties by sector and the counties with MSA Codes.
         #Join over FIPS Column
         merge1 = dfemit.merge(dfMSA, on='FIPS')
         #merge1
 In [6]: #Merge population per CBSA Code
         EmitandPop = pd.merge(merge1, dfpop, on = 'FIPS')
         #EmitandPop = dfpop.merge(merge1, on = 'CBSA Code')
         #print(EmitandPop)
         EmitandPop[' Total'] = 1000000*EmitandPop[' Total']
 In [7]: EmitandPop = EmitandPop.drop(EmitandPop[EmitandPop['Status, 1=metro 2=micro'] == 2].index)
```

```
In [32]: #Plot - Total CO2 Emissions and Population
    pl.figure(figsize = (10,10))
    pl.plot(np.log10(EmitandPop['Total Population']), np.log10(EmitandPop['Total']), 'k.')
    pl.xlabel('Total Population April 1, 2010 (log)', fontsize = 20)
    pl.ylabel('Total CO2 Emissions in Tonnes of Carbon (log)', fontsize = 20)
    pl.title ('Population and Total Emissions by County', fontsize = 30)

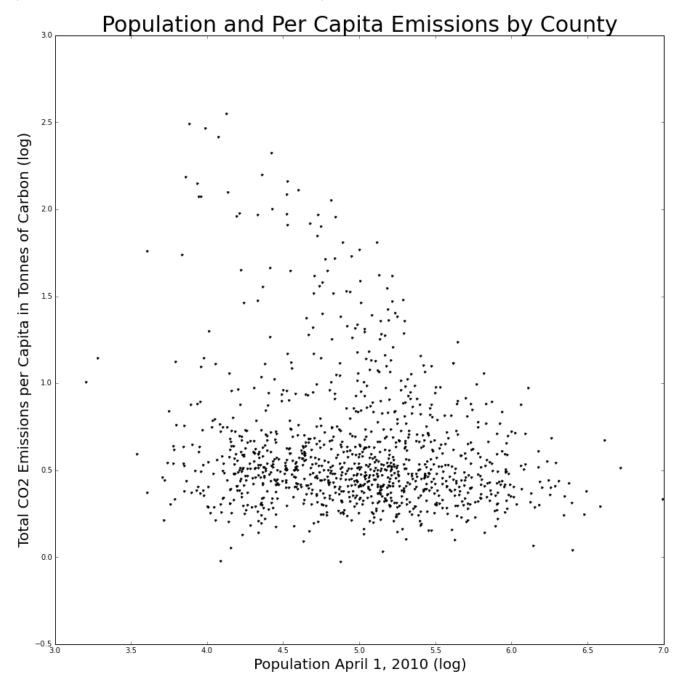
#Correlation Analysis:
    cor = stats.pearsonr(EmitandPop['Total Population'], EmitandPop['Total'])
    print (cor)

#Strong positive correlation. Pearsons Value of .755, where zero indicates no correlation and
#1 and -1 indicate strong correlation
```

(0.75541073023113547, 5.7151620561426247e-202)

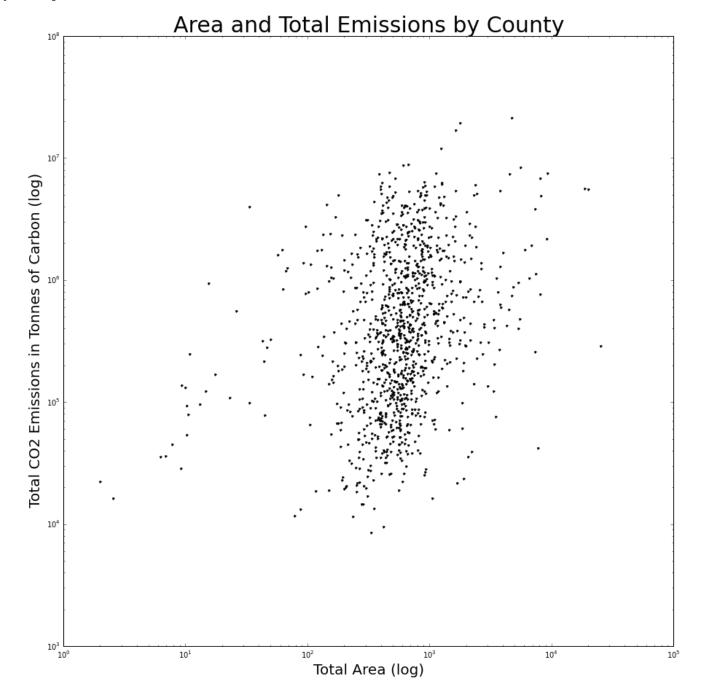


(-0.081107116374212807, 0.0073819052325340339)

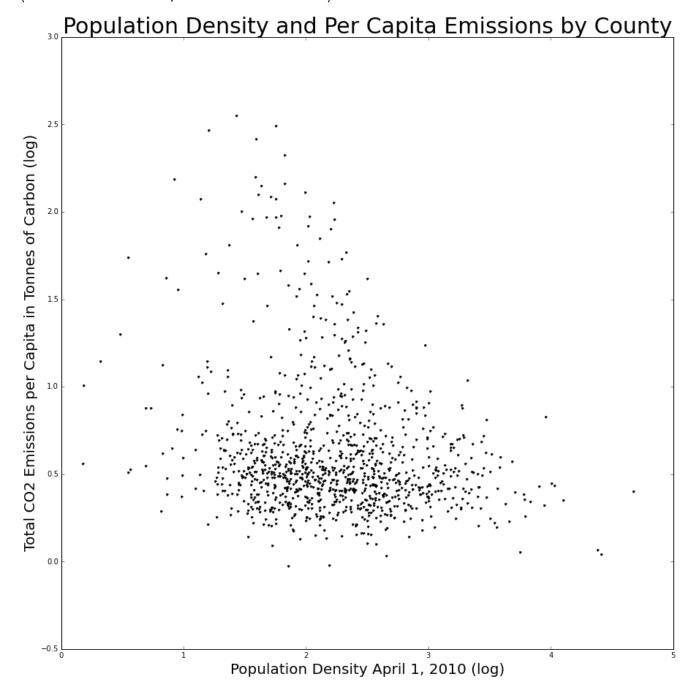


```
In [10]: pl.figure(figsize = (15,15))
    pl.loglog(EmitandPop['Area Total'], EmitandPop[' Total'], 'k.')
    pl.xlabel('Total Area (log)', fontsize = 20)
    pl.ylabel('Total CO2 Emissions in Tonnes of Carbon (log)', fontsize = 20)
    pl.title ('Area and Total Emissions by County', fontsize = 30)
```

Out[10]: <matplotlib.text.Text at 0x10b22f650>



(-0.081107116374212807, 0.0073819052325340339)



In [12]: EmitandPop.columns

Out[13]:

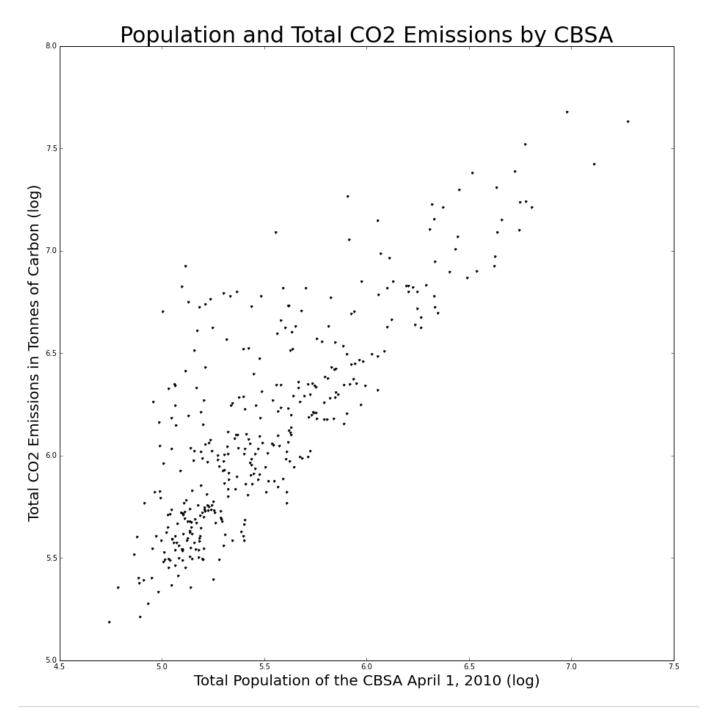
| | State_x | County | FIPS | Total | CBSA Code | CBSA Title | Total Population | Area Total | CO2perCapita_Tonnes | PopD |
|------|---------|------------------|-------|-------------|--------------|--|---------------------|-------------|---------------------|-------|
| 1461 | TX | Coryell | 48099 | 71437.166 | 28660 | Killeen- Temple-Fort Hood, TX | 75388 | 1056.756000 | 0.947593 | 71.33 |
| 1650 | VA | Poquoson | 51735 | 11658.060 | 47260 | Virginia Beach- Norfolk- Newport News, VA- NC | 12150 | 78.425670 | 0.959511 | 154.9 |
| 300 | GA | Paulding | 13223 | 154374.859 | 12060 | Atlanta- Sandy Springs- Marietta, GA | 142324 | 314.341700 | 1.084672 | 452.7 |
| 1014 | NY | Kings | 36047 | 2762535.626 | 35620 | New York- Northern New Jersey- Long Island, NY-N | 2504700 | 96.917300 | 1.102941 | 25843 |
| 1645 | VA | Manassas Park | 51685 | 16285.060 | 47900 | Washington- Arlington- Alexandria, DC-VA-MD- WV | 14273 | 2.534767 | 1.140970 | 5630. |

Out[14]:

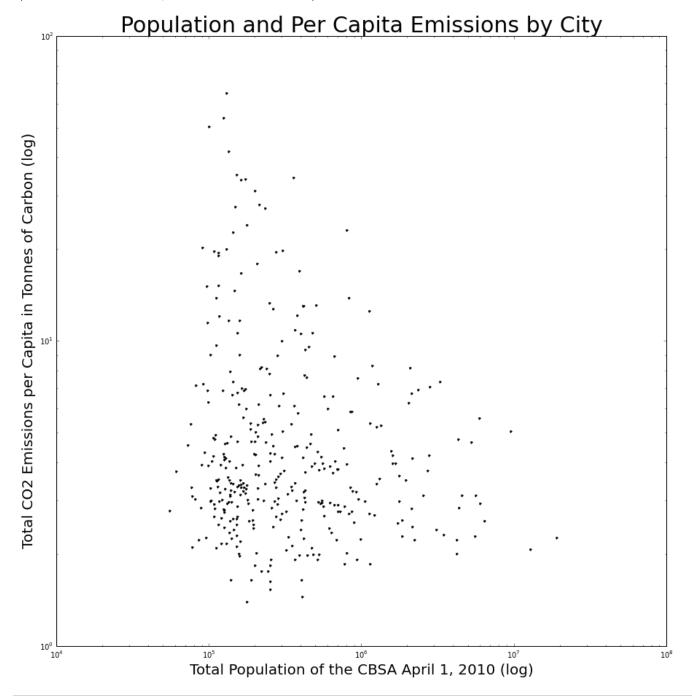
| | | Total | Total Population | Area Total |
|-----------------------------|-----------|-------------|------------------|------------|
| CBSA Title | CBSA Code | | | |
| Abilene, TX | 10180 | 649929.762 | 165252 | 2757.7055 |
| Akron, OH | 10420 | 2048256.942 | 703200 | 924.1192 |
| Albany, GA | 10500 | 977521.540 | 157308 | 1957.9622 |
| Albany-Schenectady-Troy, NY | 10580 | 2816506.060 | 870716 | 2878.2170 |
| Albuquerque, NM | 10740 | 2265460.755 | 887077 | 9297.1590 |

```
In [37]: #plot grouped by CBSA
pl.figure(figsize = (15,15))
pl.plot(log10(EmitandPopbyCBSA['Total Population']), log10(EmitandPopbyCBSA[' Total']), 'k.')
pl.xlabel('Total Population of the CBSA April 1, 2010 (log)', fontsize = 20)
pl.ylabel('Total CO2 Emissions in Tonnes of Carbon (log)', fontsize = 20)
pl.title('Population and Total CO2 Emissions by CBSA', fontsize = 30)
#Correlation Analysis:
cor = stats.pearsonr(EmitandPopbyCBSA['Total Population'], EmitandPopbyCBSA[' Total'])
print (cor)
```

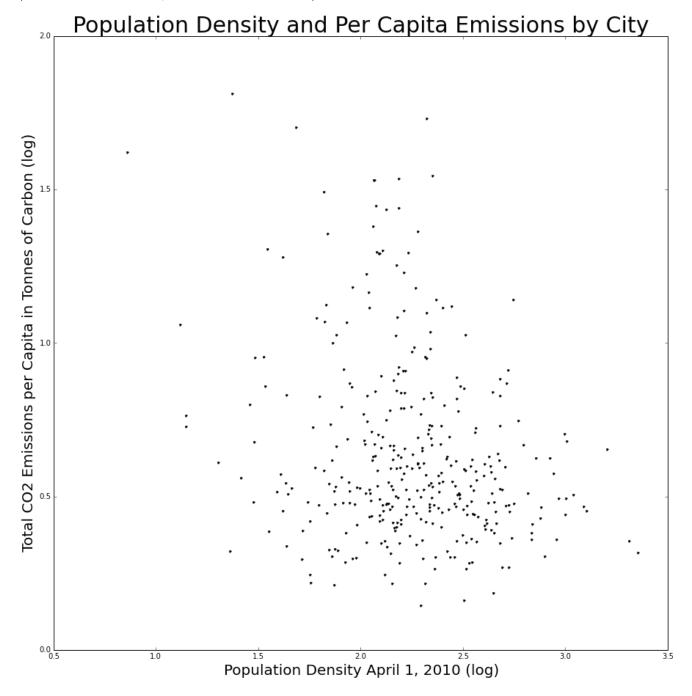
(0.85144115032796652, 6.1150188819454323e-103)



(-0.10455157895091345, 0.046833759702735837)



(-0.10455157895091345, 0.046833759702735837)



In [17]: Sorted2 = EmitandPopbyCBSA.sort(columns = 'CO2perCapita_Tonnes')
 Sorted2.head()

Out[17]:

| | | Total | Total Population | Area Total | CO2perCapita_Tonnes |
|---------------------------|-----------|------------|------------------|------------|---------------------|
| CBSA Title | CBSA Code | | | | |
| Jacksonville, NC | 27340 | 248844.833 | 177772 | 905.9130 | 1.399798 |
| Brownsville-Harlingen, TX | 15180 | 589178.470 | 406220 | 1276.4580 | 1.450393 |
| Bremerton-Silverdale, WA | 14740 | 385786.439 | 251133 | 565.9188 | 1.536184 |
| Laredo, TX | 29700 | 407319.151 | 250304 | 3375.5900 | 1.627298 |
| Greenville, NC | 24780 | 312057.515 | 189510 | 921.2242 | 1.646655 |

In []: