

Cancer_Deaths_EDA_notebook

February 12, 2023

```
[ ]: # Christine Orosco  
# Cancer Observed, Expected and Excess Deaths EDA
```

```
[1]: # Import libraries  
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
%matplotlib inline
```

0.1 Read in the flat file

```
[ ]: # Read in csv flat file in DF  
df = pd.read_csv('~DSC540/Excess_deaths.csv')  
df.shape
```

```
[3]: df.head()
```

```
[3]:
```

	Year	Cause of Death	State	State FIPS Code	HHS Region	Age Range	\
0	2005	Cancer	Alabama	AL	4	0-49	
1	2005	Cancer	Alabama	AL	4	0-49	
2	2005	Cancer	Alabama	AL	4	0-49	
3	2005	Cancer	Alabama	AL	4	0-49	
4	2005	Cancer	Alabama	AL	4	0-49	

	Benchmark	Locality	Observed Deaths	Population	Expected Deaths	\
0	2005 Fixed	All	756.0	3148377.0	451.0	
1	2005 Fixed	Metropolitan	556.0	2379871.0	341.0	
2	2005 Fixed	Nonmetropolitan	200.0	768506.0	111.0	
3	2010 Fixed	All	756.0	3148377.0	421.0	
4	2010 Fixed	Metropolitan	556.0	2379871.0	318.0	

	Potentially Excess Deaths	Percent Potentially Excess Deaths
0	305.0	40.3
1	217.0	39.0
2	89.0	44.5
3	335.0	44.3
4	238.0	42.8

```
[4]: df.tail()
```

```
[4]:      Year      Cause of Death      State State FIPS Code  HHS Region \
205915  2015  Unintentional Injury  Wyoming                WY          8
205916  2015  Unintentional Injury  Wyoming                WY          8
205917  2015  Unintentional Injury  Wyoming                WY          8
205918  2015  Unintentional Injury  Wyoming                WY          8
205919  2015  Unintentional Injury  Wyoming                WY          8

      Age Range  Benchmark      Locality  Observed Deaths  Population \
205915      0-84  2010 Fixed    Metropolitan             93.0    175787.0
205916      0-84  2010 Fixed  Nonmetropolitan            259.0    400056.0
205917      0-84   Floating              All            352.0    575843.0
205918      0-84   Floating    Metropolitan             93.0    175787.0
205919      0-84   Floating  Nonmetropolitan            259.0    400056.0

      Expected Deaths  Potentially Excess Deaths \
205915              36.0                      57.0
205916              82.0                     177.0
205917             137.0                     215.0
205918              42.0                      51.0
205919              96.0                     163.0

      Percent Potentially Excess Deaths
205915                               61.3
205916                               68.3
205917                               61.1
205918                               54.8
205919                               62.9
```

```
[5]: # Create a Subset of the data
```

```
[6]: # Select the rows where Cause of Death = Cancer and year >= 2014
df_subset = df[(df['Cause of Death'] == 'Cancer') & (df['Year'] >= 2014) &
               ↪(df['State'] != 'United States')].copy()
df_subset.head()
```

```
[6]:      Year Cause of Death      State State FIPS Code  HHS Region Age Range \
164626  2014      Cancer    Arizona                AZ          9      0-54
164896  2014      Cancer    Alabama                AL          4      0-54
165344  2014      Cancer    Alaska                 AK         10      0-59
165355  2014      Cancer    Arkansas               AR          6      0-59
165424  2014      Cancer  California               CA          9      0-64

      Benchmark Locality  Observed Deaths  Population  Expected Deaths \
164626  2005 Fixed    All             1113.0    4867353.0         1020.0
164896  2005 Fixed    All             1138.0    3475969.0          775.0
```

165344	2005 Fixed	All	252.0	625271.0	216.0
165355	2005 Fixed	All	1272.0	2325270.0	838.0
165424	2005 Fixed	All	17447.0	33809453.0	17833.0

	Potentially Excess Deaths	Percent Potentially Excess Deaths
164626	98.0	8.8
164896	363.0	31.9
165344	41.0	16.3
165355	437.0	34.4
165424	231.0	1.3

```
[7]: # Reset the index
df_subset.reset_index(drop=True, inplace=True)
```

```
[8]: pd.set_option('display.max_rows', 20)
df_subset
```

```
[8]:
```

	Year	Cause of Death	State	State FIPS Code	HHS Region	Age Range	\
0	2014	Cancer	Arizona	AZ	9	0-54	
1	2014	Cancer	Alabama	AL	4	0-54	
2	2014	Cancer	Alaska	AK	10	0-59	
3	2014	Cancer	Arkansas	AR	6	0-59	
4	2014	Cancer	California	CA	9	0-64	
...		
7339	2015	Cancer	Wyoming	WY	8	0-84	
7340	2015	Cancer	Wyoming	WY	8	0-84	
7341	2015	Cancer	Wyoming	WY	8	0-84	
7342	2015	Cancer	Wyoming	WY	8	0-84	
7343	2015	Cancer	Wyoming	WY	8	0-84	

	Benchmark	Locality	Observed Deaths	Population	\
0	2005 Fixed	All	1113.0	4867353.0	
1	2005 Fixed	All	1138.0	3475969.0	
2	2005 Fixed	All	252.0	625271.0	
3	2005 Fixed	All	1272.0	2325270.0	
4	2005 Fixed	All	17447.0	33809453.0	
...	
7339	2010 Fixed	Metropolitan	235.0	175787.0	
7340	2010 Fixed	Nonmetropolitan	533.0	400056.0	
7341	Floating	All	768.0	575843.0	
7342	Floating	Metropolitan	235.0	175787.0	
7343	Floating	Nonmetropolitan	533.0	400056.0	

	Expected Deaths	Potentially Excess Deaths	\
0	1020.0	98.0	
1	775.0	363.0	
2	216.0	41.0	

3	838.0	437.0
4	17833.0	231.0
...
7339	231.0	15.0
7340	553.0	3.0
7341	730.0	38.0
7342	215.0	27.0
7343	515.0	21.0

	Percent Potentially Excess Deaths
0	8.8
1	31.9
2	16.3
3	34.4
4	1.3
...	...
7339	6.4
7340	0.6
7341	4.9
7342	11.5
7343	3.9

[7344 rows x 13 columns]

0.2 Delete Unnecessary Data

```
[10]: # Delete unnecessary columns
df_subset.drop(columns=['HHS Region', 'Benchmark', 'Cause of Death', 'Age_
↳Range'], inplace=True)
```

```
[11]: # Display column names
df_subset.columns
```

```
[11]: Index(['Year', 'State', 'State FIPS Code', 'Locality', 'Observed Deaths',
'Population', 'Expected Deaths', 'Potentially Excess Deaths',
'Percent Potentially Excess Deaths'],
dtype='object')
```

0.3 Change Headers

```
[13]: # Change column names in the df subset
df_subset.rename(columns={
    'Observed Deaths': 'Observed_Deaths',
    'State FIPS Code': 'State_Code',
    'Potentially Excess Deaths': '
↳Potentially_Excess_Deaths',
```

```

        'Percent Potentially Excess Deaths':\
↪ 'Percent_Potentially_Excess_Deaths',
        'Expected Deaths': 'Expected_Deaths'},
    inplace=True)

```

```
[14]: df_subset.head()
```

```

[14]:   Year      State State_Code Locality  Observed_Deaths  Population \
0  2014    Arizona         AZ     All         1113.0    4867353.0
1  2014    Alabama         AL     All         1138.0    3475969.0
2  2014    Alaska         AK     All          252.0    625271.0
3  2014  Arkansas         AR     All         1272.0    2325270.0
4  2014  California        CA     All        17447.0   33809453.0

      Expected_Deaths  Potentially_Excess_Deaths \
0             1020.0                98.0
1              775.0             363.0
2              216.0              41.0
3              838.0             437.0
4            17833.0             231.0

      Percent_Potentially_Excess_Deaths
0                      8.8
1                     31.9
2                     16.3
3                     34.4
4                      1.3

```

0.4 Delete Duplicate rows in the subset

```

[16]: # drop duplicate rows
df_subset.drop_duplicates(inplace=True)
df_subset.shape
df_subset.head()

```

```

[16]:   Year      State State_Code Locality  Observed_Deaths  Population \
0  2014    Arizona         AZ     All         1113.0    4867353.0
1  2014    Alabama         AL     All         1138.0    3475969.0
2  2014    Alaska         AK     All          252.0    625271.0
3  2014  Arkansas         AR     All         1272.0    2325270.0
4  2014  California        CA     All        17447.0   33809453.0

      Expected_Deaths  Potentially_Excess_Deaths \
0             1020.0                98.0
1              775.0             363.0
2              216.0              41.0
3              838.0             437.0

```

4	17833.0	231.0
---	---------	-------

	Percent_Potentially_Excess_Deaths
0	8.8
1	31.9
2	16.3
3	34.4
4	1.3

0.5 Check Case Consistency

```
[17]: # Upper Case State_Codes
df_subset['State_upper'] = list(map(lambda x: x.isupper(),
    ↪df_subset['State_Code']))
notmet = df_subset[df_subset['State_upper'] == 0]
notmet.head()
```

```
[17]: Empty DataFrame
Columns: [Year, State, State_Code, Locality, Observed_Deaths, Population,
Expected_Deaths, Potentially_Excess_Deaths, Percent_Potentially_Excess_Deaths,
State_upper]
Index: []
```

```
[18]: # Change State Code 0 to US
# df_subset.loc[df_subset['State_Code'] == 0]
# df_subset['State_Code'] = df_subset['State_Code'].replace({0: 'US'},
    ↪inplace=True)
# Couldn't get this to work
```

```
[19]: # - Check for consistent Locality values
local_list = ['All', 'Metropolitan', 'Nonmetropolitan']
df_subset['Locality_check'] = list(map(lambda x: x in local_list,
    ↪df_subset['Locality']))
notmet = df_subset[df_subset['Locality_check'] == 0]
notmet.head()
```

```
[19]: Empty DataFrame
Columns: [Year, State, State_Code, Locality, Observed_Deaths, Population,
Expected_Deaths, Potentially_Excess_Deaths, Percent_Potentially_Excess_Deaths,
State_upper, Locality_check]
Index: []
```

```
[20]: df_subset.head()
```

```
[20]:   Year   State State_Code Locality Observed_Deaths Population \
0  2014   Arizona         AZ     All           1113.0   4867353.0
```

1	2014	Alabama	AL	All	1138.0	3475969.0
2	2014	Alaska	AK	All	252.0	625271.0
3	2014	Arkansas	AR	All	1272.0	2325270.0
4	2014	California	CA	All	17447.0	33809453.0

	Expected_Deaths	Potentially_Excess_Deaths	\
0	1020.0		98.0
1	775.0		363.0
2	216.0		41.0
3	838.0		437.0
4	17833.0		231.0

	Percent_Potentially_Excess_Deaths	State_upper	Locality_check
0	8.8	True	True
1	31.9	True	True
2	16.3	True	True
3	34.4	True	True
4	1.3	True	True

0.6 Handle Missing Values

```
[22]: # For each column count the number of missing values in the DF
df_subset.isnull().sum()
```

```
[22]: Year                0
State                  0
State_Code             0
Locality              0
Observed_Deaths       8
Population            8
Expected_Deaths       8
Potentially_Excess_Deaths 8
Percent_Potentially_Excess_Deaths 8
State_upper           0
Locality_check        0
dtype: int64
```

```
[23]: # Display the rows with missing data -
df_subset[df_subset.isna().any(axis=1)]
```

```
[23]:      Year      State State_Code      Locality \
515   2014    Delaware      DE  Nonmetropolitan
585   2014  District of\nColumbia      DC  Nonmetropolitan
2162  2014    New Jersey      NJ  Nonmetropolitan
2810  2014    Rhode Island      RI  Nonmetropolitan
4188  2015    Delaware      DE  Nonmetropolitan
4260  2015  District of\nColumbia      DC  Nonmetropolitan
```

5834	2015	New Jersey	NJ	Nonmetropolitan
6483	2015	Rhode Island	RI	Nonmetropolitan

	Observed_Deaths	Population	Expected_Deaths	Potentially_Excess_Deaths	\
515	NaN	NaN	NaN	NaN	NaN
585	NaN	NaN	NaN	NaN	NaN
2162	NaN	NaN	NaN	NaN	NaN
2810	NaN	NaN	NaN	NaN	NaN
4188	NaN	NaN	NaN	NaN	NaN
4260	NaN	NaN	NaN	NaN	NaN
5834	NaN	NaN	NaN	NaN	NaN
6483	NaN	NaN	NaN	NaN	NaN

	Percent_Potentially_Excess_Deaths	State_upper	Locality_check
515	NaN	True	True
585	NaN	True	True
2162	NaN	True	True
2810	NaN	True	True
4188	NaN	True	True
4260	NaN	True	True
5834	NaN	True	True
6483	NaN	True	True

```
[24]: # May need to change NaN to zeros or derive values from previous years.
# View previous years to
state_list = ['DE', 'DC', 'NJ', 'RI']
df_prevyears = df[(df['Cause of Death'] == 'Cancer') & (df['Year'] <= 2013) &
↳ (df['State FIPS Code'].isin(state_list))].copy()
df_prevyears
```

```
[24]:
```

	Year	Cause of Death	State	State FIPS Code	HHS Region	\
504	2005	Cancer	Delaware	DE	3	
505	2005	Cancer	Delaware	DE	3	
506	2005	Cancer	Delaware	DE	3	
507	2005	Cancer	Delaware	DE	3	
508	2005	Cancer	Delaware	DE	3	
...	
152635	2013	Cancer	Rhode Island	RI	1	
152636	2013	Cancer	Rhode Island	RI	1	
152637	2013	Cancer	Rhode Island	RI	1	
152638	2013	Cancer	Rhode Island	RI	1	
152639	2013	Cancer	Rhode Island	RI	1	

	Age Range	Benchmark	Locality	Observed Deaths	Population	\
504	0-49	2005 Fixed	All	106.0	583236.0	
505	0-49	2005 Fixed	Metropolitan	106.0	583236.0	
506	0-49	2005 Fixed	Nonmetropolitan	NaN	NaN	

507	0-49	2010 Fixed	All	106.0	583236.0
508	0-49	2010 Fixed	Metropolitan	106.0	583236.0
...
152635	0-84	2010 Fixed	Metropolitan	1810.0	1023183.0
152636	0-84	2010 Fixed	Nonmetropolitan	NaN	NaN
152637	0-84	Floating	All	1810.0	1023183.0
152638	0-84	Floating	Metropolitan	1810.0	1023183.0
152639	0-84	Floating	Nonmetropolitan	NaN	NaN

	Expected Deaths	Potentially Excess Deaths \
504	85.0	24.0
505	85.0	24.0
506	NaN	NaN
507	80.0	28.0
508	80.0	28.0
...
152635	1448.0	365.0
152636	NaN	NaN
152637	1392.0	421.0
152638	1392.0	421.0
152639	NaN	NaN

	Percent Potentially Excess Deaths
504	22.6
505	22.6
506	NaN
507	26.4
508	26.4
...	...
152635	20.2
152636	NaN
152637	23.3
152638	23.3
152639	NaN

[2592 rows x 13 columns]

```
[25]: # Check for NaN values in previous years
# Appears that these states do not have reported values for nonmetropolitan
      ↪ areas
df_prevyears.isnull().sum()
```

```
[25]: Year          0
Cause of Death     0
State              0
State FIPS Code    0
HHS Region         0
```

```

Age Range          0
Benchmark          0
Locality           0
Observed Deaths   864
Population         864
Expected Deaths   864
Potentially Excess Deaths 864
Percent Potentially Excess Deaths 864
dtype: int64

```

```

[26]: # Check the nonmetropolitan rows
nonmet = df_prevyears[df_prevyears['Locality'] == 'Nonmetropolitan']
nonmet['State FIPS Code'].unique()

```

```

[26]: array(['DE', 'DC', 'NJ', 'RI'], dtype=object)

```

```

[27]: # Because 4 States do not have values for nonmetropolitan, will check other
      ↪ data sources for values

```

0.7 Reshape the data by combining observations by State, Year and Locality

```

[76]: # For each state combine the Locality observations into three separate rows
      ↪ because not going to use the HHS region. Only using the totals for metro and
      ↪ non-metro areas
# Create pivot table to aggregate each locality number
col_list = ['Observed_Deaths', 'Population', 'Expected_Deaths',
            'Potentially_Excess_Deaths', 'Percent_Potentially_Excess_Deaths']

df_pivot = df_subset.pivot_table(index=['Year', 'State_Code'],
      ↪ columns='Locality', values=col_list, aggfunc=sum)
df_pivot

```

```

[76]:

```

		Expected_Deaths			Observed_Deaths \
Locality		All	Metropolitan	Nonmetropolitan	All
Year	State_Code				
2014	AK	9618.0	6178.0	3403.0	11037.0
	AL	76163.0	56636.0	19527.0	105675.0
	AR	46043.0	26386.0	19650.0	65784.0
	AZ	102820.0	97111.0	5711.0	106986.0
	CA	540349.0	524776.0	15576.0	548517.0
...
2015	VT	10940.0	3282.0	7660.0	12972.0
	WA	108307.0	95175.0	13128.0	117654.0
	WI	91150.0	64206.0	26942.0	102003.0
	WV	32212.0	19219.0	12993.0	47406.0
	WY	8940.0	2639.0	6303.0	8700.0

Locality		Metropolitan Nonmetropolitan	
Year	State_Code		
2014	AK	6852.0	4137.0
	AL	76464.0	29211.0
	AR	35679.0	30105.0
	AZ	101478.0	5508.0
	CA	531036.0	17481.0
...	
2015	VT	3705.0	9267.0
	WA	101973.0	15681.0
	WI	70308.0	31695.0
	WV	28209.0	19197.0
	WY	2676.0	6024.0

Locality		Percent_Potentially_Excess_Deaths	
Year	State_Code	All	Metropolitan
2014	AK	371.5	334.4
	AL	740.5	694.4
	AR	784.7	670.6
	AZ	206.8	215.1
	CA	137.5	133.3
...	
2015	VT	395.2	333.5
	WA	219.9	189.1
	WI	275.9	230.0
	WV	825.4	801.8
	WY	53.1	177.1

Locality		Population		
Year	State_Code	Nonmetropolitan	All Metropolitan	Nonmetropolitan
2014	AK	474.6	15762231.0	10738815.0
	AL	871.4	98584620.0	75517581.0
	AR	936.5	60216000.0	37603665.0
	AZ	220.7	136871601.0	129924657.0
	CA	341.2	807586629.0	791482749.0
...	
2015	VT	457.8	12369387.0	4445340.0
	WA	458.4	146927250.0	132955461.0
	WI	393.3	116414547.0	87214530.0
	WV	863.6	36394473.0	22640328.0
	WY	97.1	11976111.0	3679551.0

Potentially_Excess_Deaths

Locality		All	Metropolitan	Nonmetropolitan
Year	State_Code			
2014	AK	1578.0	837.0	797.0
	AL	29520.0	19852.0	9702.0
	AR	19829.0	9443.0	10455.0
	AZ	6592.0	6598.0	246.0
	CA	22127.0	20498.0	2036.0
...	
2015	VT	2070.0	467.0	1655.0
	WA	10804.0	8209.0	2657.0
	WI	11151.0	6514.0	4759.0
	WV	15195.0	9003.0	6204.0
	WY	158.0	186.0	139.0

[102 rows x 15 columns]

```
[30]: # Flatten the pivot_table to join the Locality category in the observation type
      ↪header names
df_pivot.columns = [' '.join(col) for col in df_pivot.columns.values]
```

```
[31]: df_pivot.columns
```

```
[31]: Index(['Expected_Deaths All', 'Expected_Deaths Metropolitan',
          'Expected_Deaths Nonmetropolitan', 'Observed_Deaths All',
          'Observed_Deaths Metropolitan', 'Observed_Deaths Nonmetropolitan',
          'Percent_Potentially_Excess_Deaths All',
          'Percent_Potentially_Excess_Deaths Metropolitan',
          'Percent_Potentially_Excess_Deaths Nonmetropolitan', 'Population All',
          'Population Metropolitan', 'Population Nonmetropolitan',
          'Potentially_Excess_Deaths All',
          'Potentially_Excess_Deaths Metropolitan',
          'Potentially_Excess_Deaths Nonmetropolitan'],
          dtype='object')
```

```
[32]: # Sort the Dataframe
df_pivot.sort_index(inplace=True)
df_pivot.query('Year == 2014')
```

```
[32]:
```

Year	State_Code	Expected_Deaths All	Expected_Deaths Metropolitan \
2014	AK	9618.0	6178.0
	AL	76163.0	56636.0
	AR	46043.0	26386.0
	AZ	102820.0	97111.0
	CA	540349.0	524776.0
...	
	VT	10845.0	3240.0

WA	106000.0	93091.0
WI	90072.0	63327.0
WV	32177.0	19185.0
WY	8859.0	2582.0

Year	State_Code	Expected_Deaths Nonmetropolitan	Observed_Deaths All \
2014	AK	3403.0	11037.0
	AL	19527.0	105675.0
	AR	19650.0	65784.0
	AZ	5711.0	106986.0
	CA	15576.0	548517.0
...	
	VT	7608.0	12753.0
	WA	12911.0	114786.0
	WI	26742.0	102246.0
	WV	12992.0	48615.0
	WY	6249.0	8979.0

Year	State_Code	Observed_Deaths Metropolitan \
2014	AK	6852.0
	AL	76464.0
	AR	35679.0
	AZ	101478.0
	CA	531036.0
...		...
	VT	3633.0
	WA	99771.0
	WI	72105.0
	WV	29391.0
	WY	2665.0

Year	State_Code	Observed_Deaths Nonmetropolitan \
2014	AK	4137.0
	AL	29211.0
	AR	30105.0
	AZ	5508.0
	CA	17481.0
...		...
	VT	9120.0
	WA	15015.0
	WI	30141.0
	WV	19224.0
	WY	6291.0

Year	State_Code	Percent_Potentially_Excess_Deaths All \
2014	AK	371.5
	AL	740.5
	AR	784.7
	AZ	206.8
	CA	137.5
...		...
	VT	429.3
	WA	234.6
	WI	289.6
	WV	886.9
	WY	101.0

Year	State_Code	Percent_Potentially_Excess_Deaths Metropolitan \
2014	AK	334.4
	AL	694.4
	AR	670.6
	AZ	215.1
	CA	133.3
...		...
	VT	382.1
	WA	208.5
	WI	291.7
	WV	922.9
	WY	199.5

Year	State_Code	Percent_Potentially_Excess_Deaths Nonmetropolitan \
2014	AK	474.6
	AL	871.4
	AR	936.5
	AZ	220.7
	CA	341.2
...		...
	VT	476.0
	WA	459.2
	WI	305.3
	WV	832.1
	WY	136.7

Year	State_Code	Population All	Population Metropolitan \
2014	AK	15762231.0	10738815.0
	AL	98584620.0	75517581.0
	AR	60216000.0	37603665.0

AZ	136871601.0	129924657.0
CA	807586629.0	791482749.0
...
VT	12437853.0	4445346.0
WA	145078080.0	131183028.0
WI	116532489.0	87141945.0
WV	36632532.0	22733298.0
WY	11971962.0	3531017.0

Year	State_Code	Population Nonmetropolitan	Potentially_Excess_Deaths All \
2014	AK	4844330.0	1578.0
	AL	23067039.0	29520.0
	AR	22612335.0	19829.0
	AZ	6946944.0	6592.0
	CA	16103880.0	22127.0
...
	VT	7992507.0	1920.0
	WA	13895052.0	9973.0
	WI	29390544.0	12357.0
	WV	13899234.0	16462.0
	WY	8310732.0	405.0

Year	State_Code	Potentially_Excess_Deaths Metropolitan \
2014	AK	837.0
	AL	19852.0
	AR	9443.0
	AZ	6598.0
	CA	20498.0
...
	VT	436.0
	WA	7796.0
	WI	8896.0
	WV	10263.0
	WY	190.0

Year	State_Code	Potentially_Excess_Deaths Nonmetropolitan
2014	AK	797.0
	AL	9702.0
	AR	10455.0
	AZ	246.0
	CA	2036.0
...
	VT	1547.0
	WA	2253.0

WI	3600.0
WV	6233.0
WY	335.0

[51 rows x 15 columns]

0.8 Find outliers or anomalies in the counts

```
[34]: # do not include the 0 State_code which is the totals row.
      # Compute Mean for Expected_Deaths_All
      df_pivot.iloc[0:51,2].mean()
```

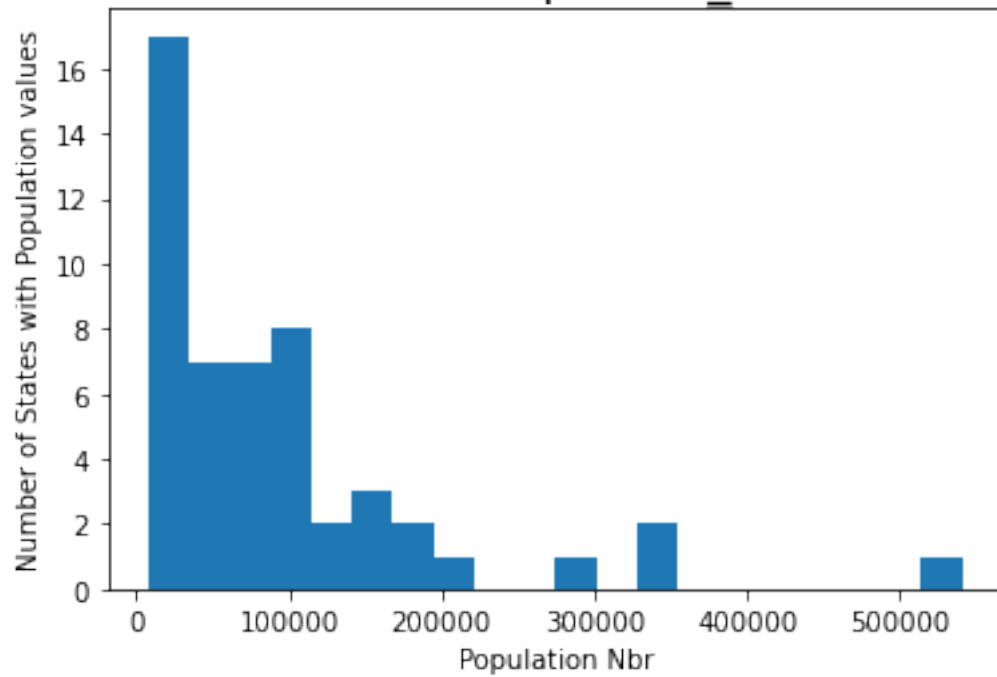
[34]: 15346.941176470587

```
[35]: df_pivot.iloc[52:,1].mean()
```

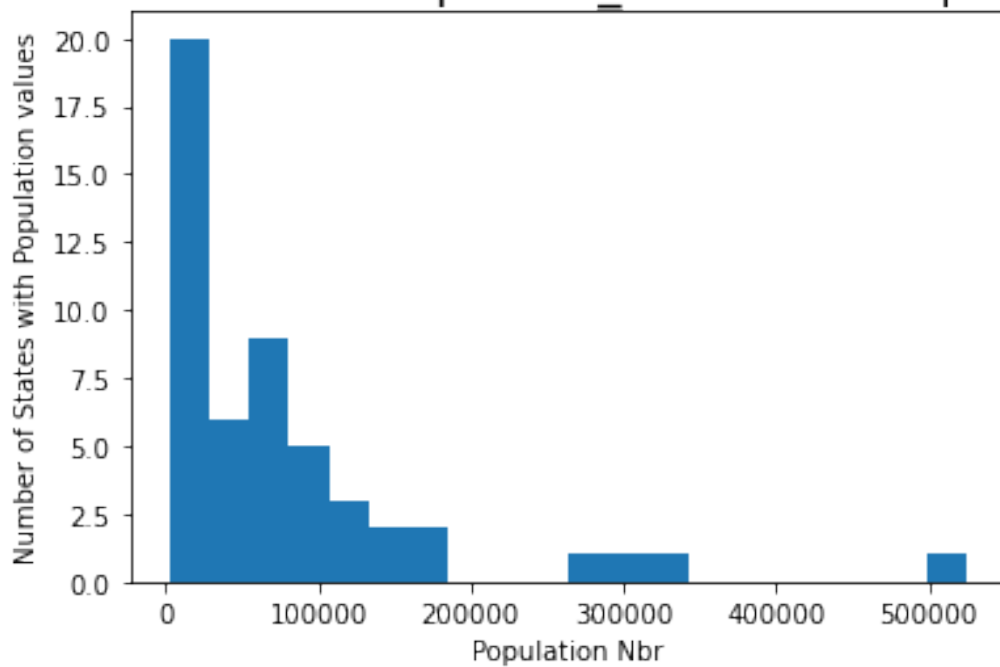
[35]: 81388.78

```
[36]: # Create histograms to visually identify outliers or anomalies
      # Create histograms for each column in the pivot df
      for x in df_pivot.columns:
          plt.title( "Plots for 2014 " +x, fontsize=18)
          plt.hist(df_pivot.iloc[0:51,0:][x], bins=20)
          plt.xlabel('Population Nbr')
          plt.ylabel("Number of States with Population values")
          plt.show()
```

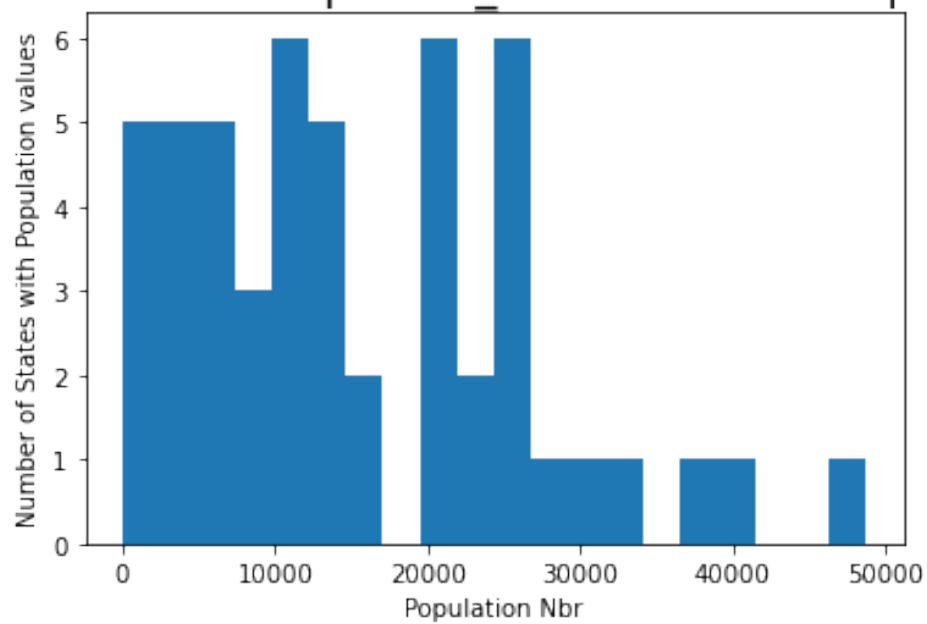

Plots for 2014 Expected_Deaths All



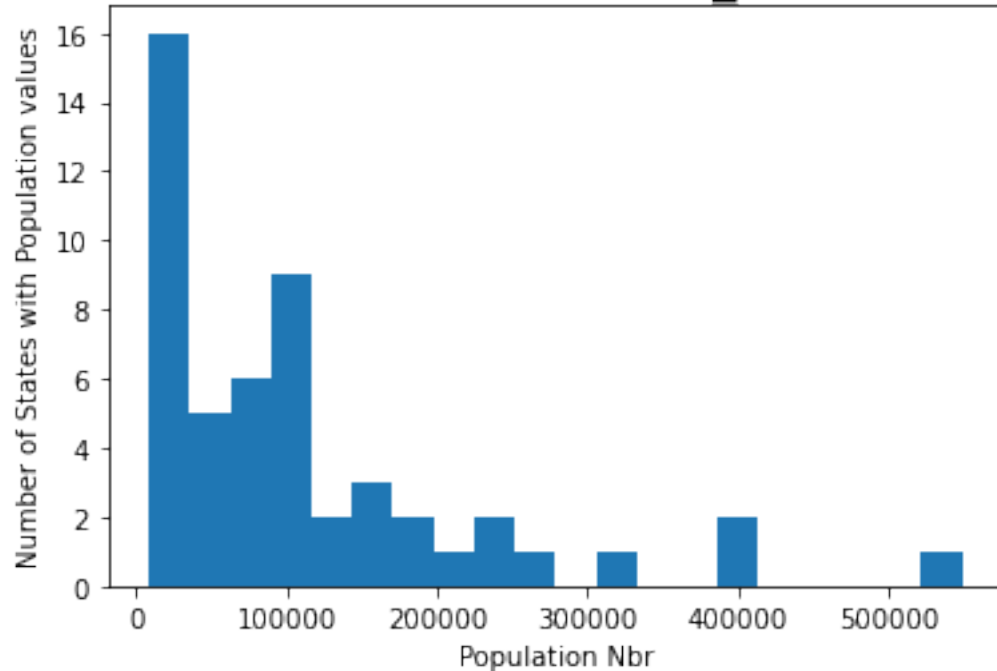
Plots for 2014 Expected_Deaths Metropolitan



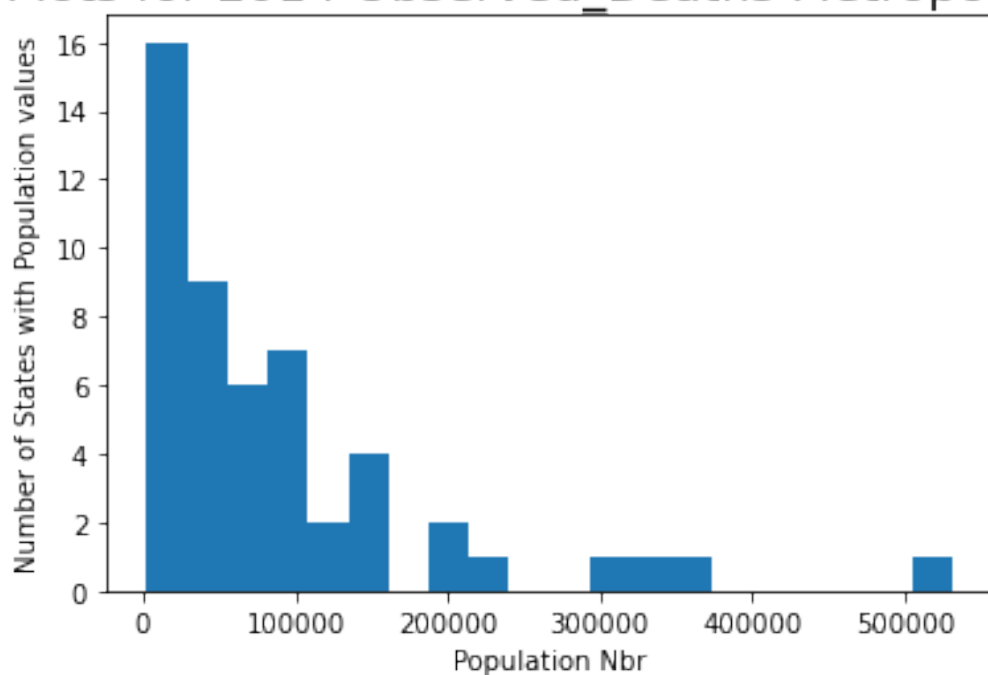
Plots for 2014 Expected_Deaths Nonmetropolitan



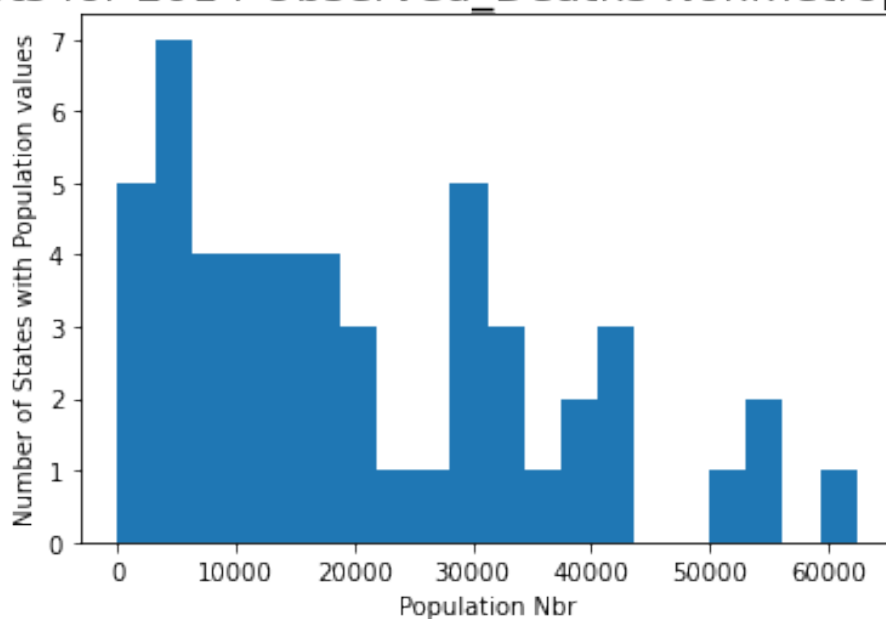
Plots for 2014 Observed_Deaths All



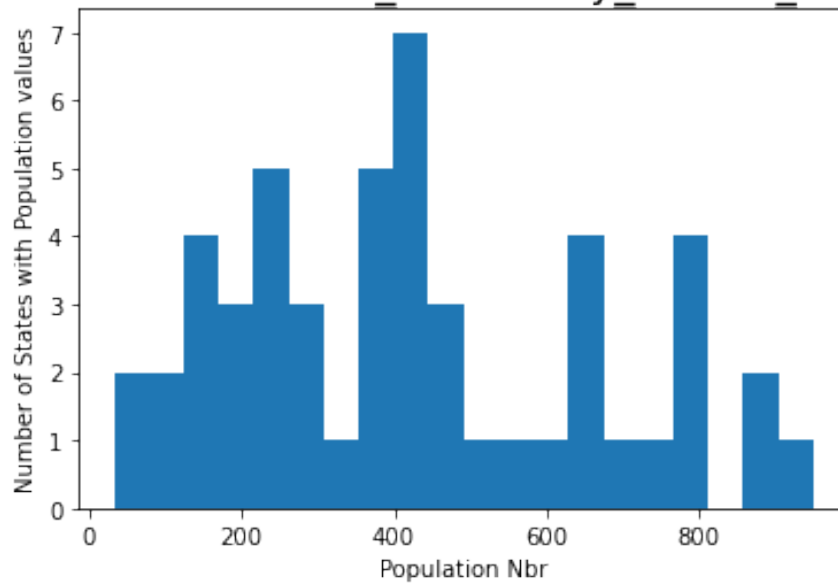
Plots for 2014 Observed_Deaths Metropolitan



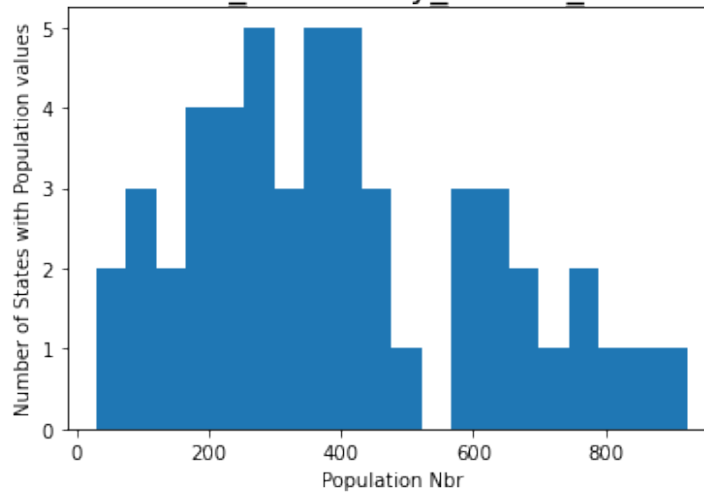
Plots for 2014 Observed_Deaths Nonmetropolitan



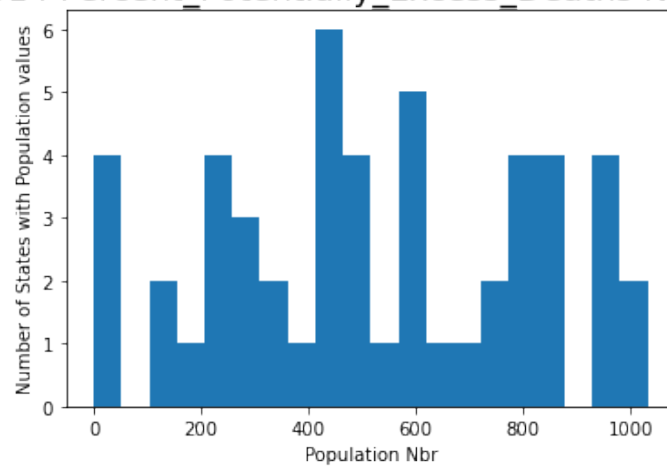
Plots for 2014 Percent_Potentially_Excess_Deaths All



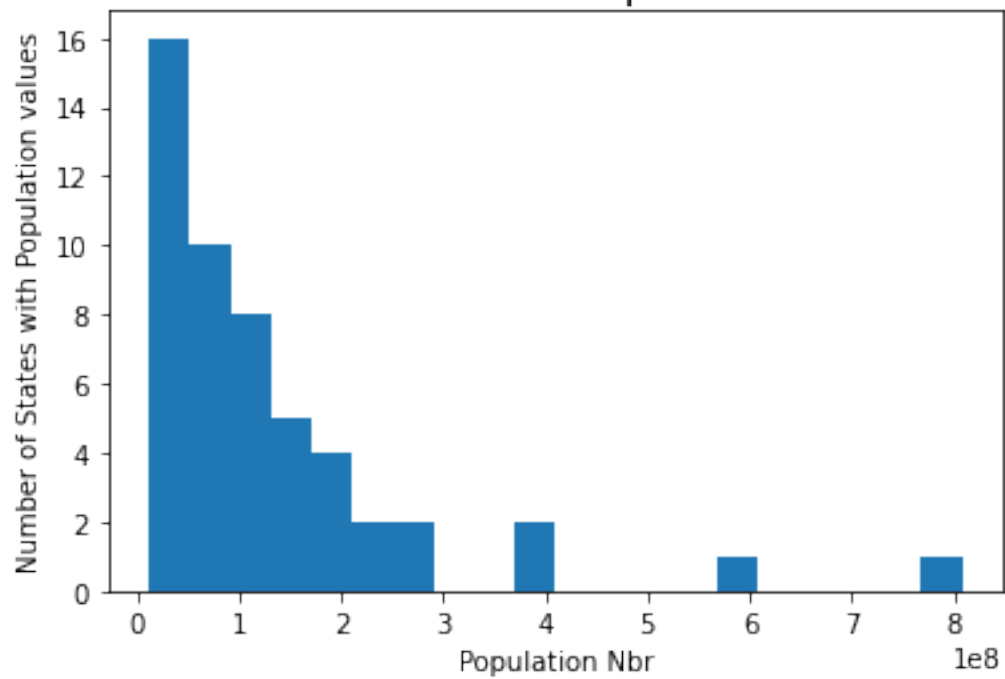
Plots for 2014 Percent_Potentially_Excess_Deaths Metropolitan



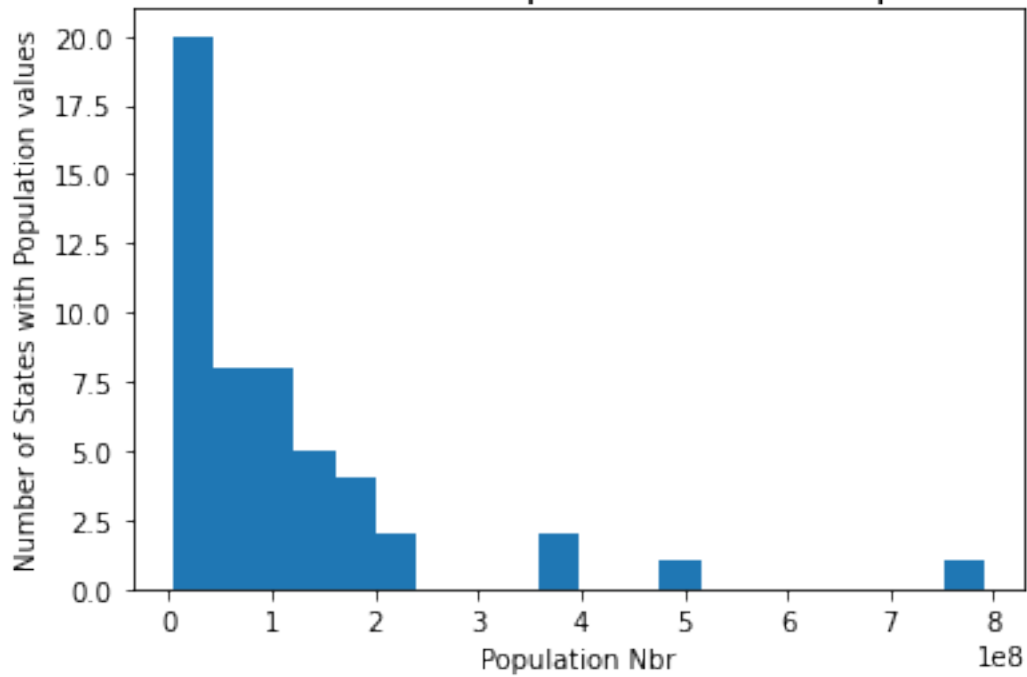
Plots for 2014 Percent_Potentially_Excess_Deaths Nonmetropolitan



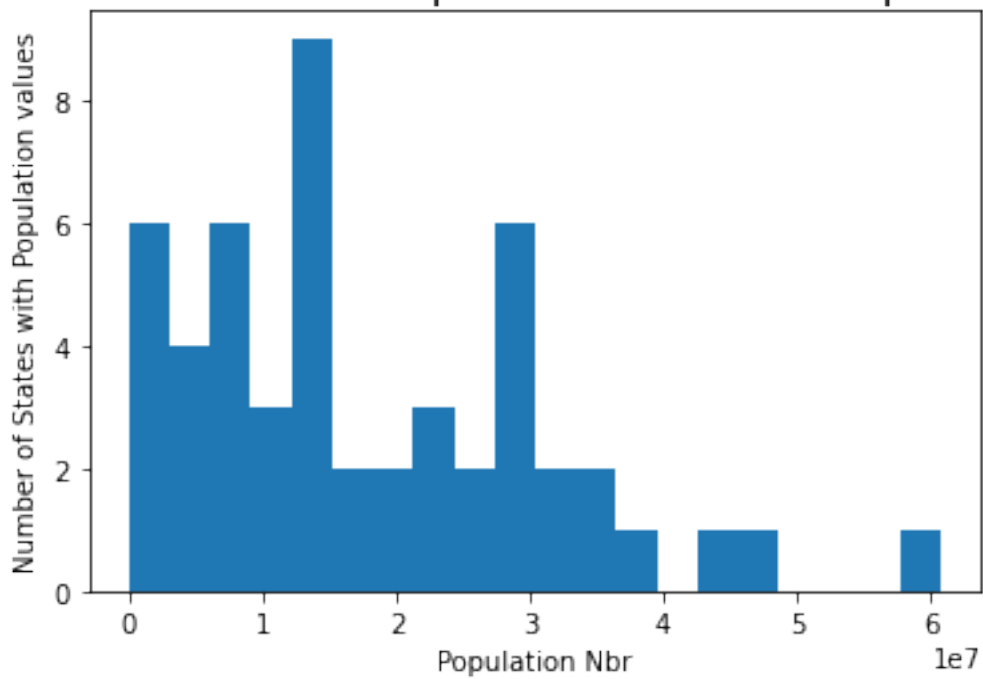
Plots for 2014 Population All



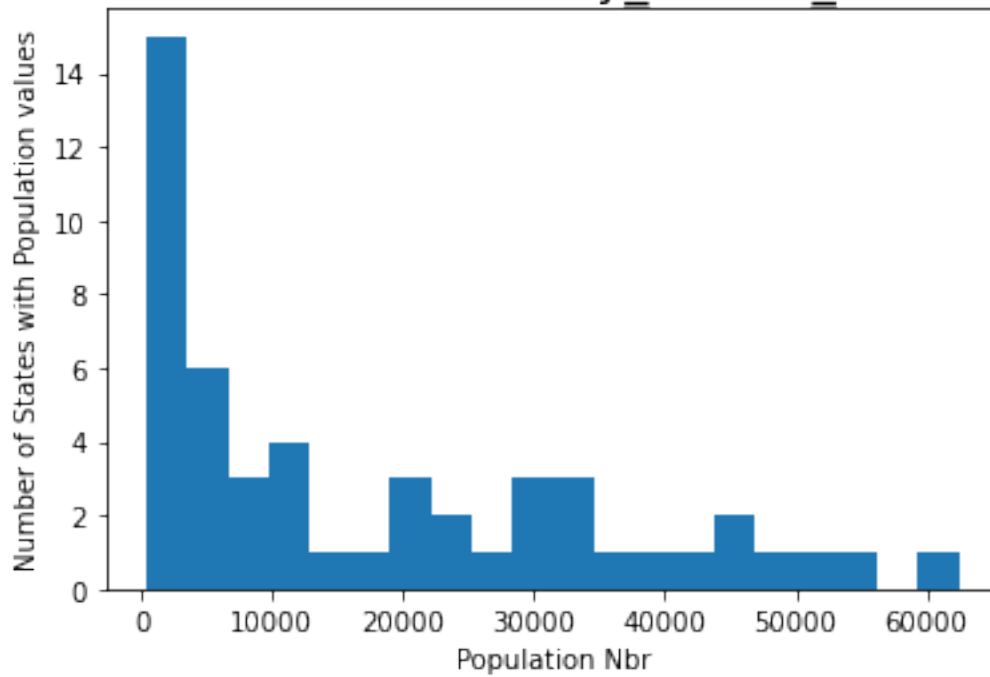
Plots for 2014 Population Metropolitan



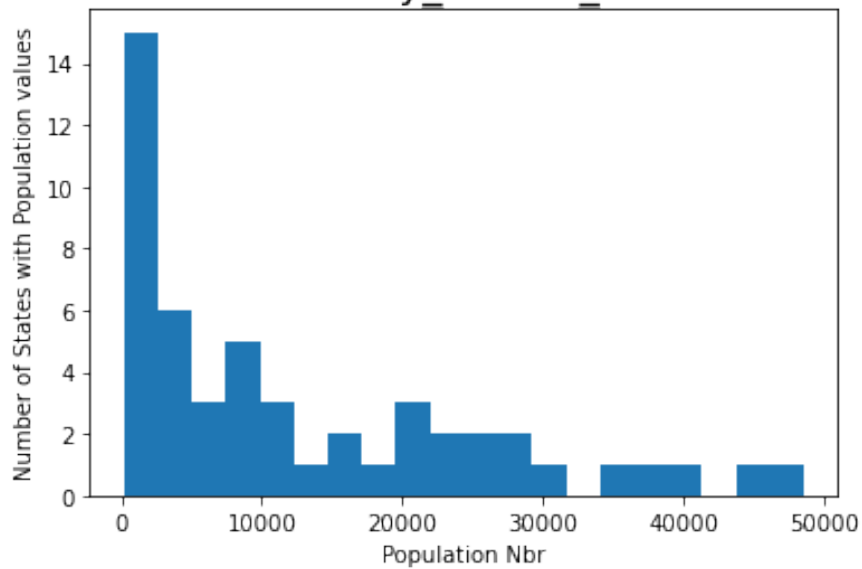
Plots for 2014 Population Nonmetropolitan



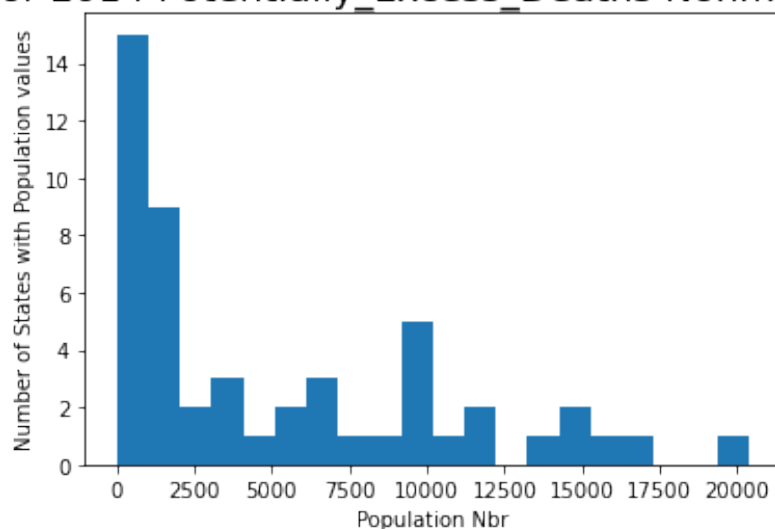
Plots for 2014 Potentially_Excess_Deaths All



Plots for 2014 Potentially_Excess_Deaths Metropolitan



Plots for 2014 Potentially_Excess_Deaths Nonmetropolitan



0.9 Compute zscores to find outliers

```
[ ]: from scipy import stats
```

```
[38]: # Computer zscores for year 2014 "Expected Deaths All"
df_2014 = df_pivot.query('Year == 2014')
df_2014
```

```
[38]:
```

Year	State_Code	Expected_Deaths All	Expected_Deaths Metropolitan \
2014	AK	9618.0	6178.0
	AL	76163.0	56636.0
	AR	46043.0	26386.0
	AZ	102820.0	97111.0
	CA	540349.0	524776.0
...
	VT	10845.0	3240.0
	WA	106000.0	93091.0
	WI	90072.0	63327.0
	WV	32177.0	19185.0
	WY	8859.0	2582.0

Year	State_Code	Expected_Deaths Nonmetropolitan	Observed_Deaths All \
2014	AK	3403.0	11037.0
	AL	19527.0	105675.0
	AR	19650.0	65784.0

	AZ	5711.0	106986.0
	CA	15576.0	548517.0
...	
	VT	7608.0	12753.0
	WA	12911.0	114786.0
	WI	26742.0	102246.0
	WV	12992.0	48615.0
	WY	6249.0	8979.0

Year	State_Code	Observed_Deaths Metropolitan \
2014	AK	6852.0
	AL	76464.0
	AR	35679.0
	AZ	101478.0
	CA	531036.0
...		...
	VT	3633.0
	WA	99771.0
	WI	72105.0
	WV	29391.0
	WY	2665.0

Year	State_Code	Observed_Deaths Nonmetropolitan \
2014	AK	4137.0
	AL	29211.0
	AR	30105.0
	AZ	5508.0
	CA	17481.0
...		...
	VT	9120.0
	WA	15015.0
	WI	30141.0
	WV	19224.0
	WY	6291.0

Year	State_Code	Percent_Potentially_Excess_Deaths All \
2014	AK	371.5
	AL	740.5
	AR	784.7
	AZ	206.8
	CA	137.5
...		...
	VT	429.3
	WA	234.6

WI	289.6
WV	886.9
WY	101.0

Year	State_Code	Percent_Potentially_Excess_Deaths Metropolitan \
2014	AK	334.4
	AL	694.4
	AR	670.6
	AZ	215.1
	CA	133.3
...		...
	VT	382.1
	WA	208.5
	WI	291.7
	WV	922.9
	WY	199.5

Year	State_Code	Percent_Potentially_Excess_Deaths Nonmetropolitan \
2014	AK	474.6
	AL	871.4
	AR	936.5
	AZ	220.7
	CA	341.2
...		...
	VT	476.0
	WA	459.2
	WI	305.3
	WV	832.1
	WY	136.7

Year	State_Code	Population All	Population Metropolitan \
2014	AK	15762231.0	10738815.0
	AL	98584620.0	75517581.0
	AR	60216000.0	37603665.0
	AZ	136871601.0	129924657.0
	CA	807586629.0	791482749.0
...	
	VT	12437853.0	4445346.0
	WA	145078080.0	131183028.0
	WI	116532489.0	87141945.0
	WV	36632532.0	22733298.0
	WY	11971962.0	3531017.0

Population Nonmetropolitan	Potentially_Excess_Deaths All \
----------------------------	---------------------------------

Year	State_Code		
2014	AK	4844330.0	1578.0
	AL	23067039.0	29520.0
	AR	22612335.0	19829.0
	AZ	6946944.0	6592.0
	CA	16103880.0	22127.0
...
	VT	7992507.0	1920.0
	WA	13895052.0	9973.0
	WI	29390544.0	12357.0
	WV	13899234.0	16462.0
	WY	8310732.0	405.0

		Potentially_Excess_Deaths Metropolitan \
Year	State_Code	
2014	AK	837.0
	AL	19852.0
	AR	9443.0
	AZ	6598.0
	CA	20498.0
...
	VT	436.0
	WA	7796.0
	WI	8896.0
	WV	10263.0
	WY	190.0

		Potentially_Excess_Deaths Nonmetropolitan
Year	State_Code	
2014	AK	797.0
	AL	9702.0
	AR	10455.0
	AZ	246.0
	CA	2036.0
...
	VT	1547.0
	WA	2253.0
	WI	3600.0
	WV	6233.0
	WY	335.0

[51 rows x 15 columns]

```
[43]: # Print zcores
zscore_2014 = stats.zscore(df_2014.iloc[:,1])
zscore_2014
```

```
[43]: array([-0.74821394, -0.22669078, -0.53934835,  0.19165023,  4.61190481,
          -0.13867121, -0.26067484, -0.72669483, -0.65440634,  2.57466941,
           0.3620977 , -0.63497506, -0.54483665, -0.65201877,  0.88114028,
          -0.04113238, -0.52921928, -0.41762378, -0.22668044,  0.24624386,
           0.08373427, -0.67304176,  0.47912982, -0.18773519, -0.11117801,
          -0.60910458, -0.75428105,  0.36402016, -0.76332487, -0.64368812,
          -0.67370325,  0.604813  , -0.59181281, -0.41980463,  2.01882109,
           0.67312222, -0.43971136, -0.28204925,  1.06229563, -0.64110418,
          -0.14810778, -0.75133535, -0.02919455,  2.33635715, -0.51047016,
           0.26600588, -0.77858049,  0.15010036, -0.15753402, -0.61377636,
          -0.78538144])
```

```
[44]: # Display zscores GT than 3 or less than -3
```

```
bad_list = []
bad_list = [x for x in zscore_2014 if x > 3 or x < -3]
print(bad_list)
```

```
[4.611904813199927]
```

```
[ ]: # The zscore shows California with a zscore over 3 for Expected Deaths All 2014.
      # Calculate 2015 zscore for comparison
```

```
[45]: # Computer zscores for year 2015 "Expected Deaths All"
df_2015 = df_pivot.query('Year == 2015')
df_2015
```

```
[45]:
```

Year	State_Code	Expected_Deaths All	Expected_Deaths Metropolitan \
2015	AK	9699.0	6239.0
	AL	76859.0	57272.0
	AR	46424.0	26799.0
	AZ	105356.0	99602.0
	CA	550846.0	535135.0
...
	VT	10940.0	3282.0
	WA	108307.0	95175.0
	WI	91150.0	64206.0
	WV	32212.0	19219.0
	WY	8940.0	2639.0

Year	State_Code	Expected_Deaths Nonmetropolitan	Observed_Deaths All \
2015	AK	3354.0	11043.0
	AL	19587.0	105708.0
	AR	19627.0	67029.0
	AZ	5755.0	109086.0

	CA	15710.0	555861.0
...	
	VT	7660.0	12972.0
	WA	13128.0	117654.0
	WI	26942.0	102003.0
	WV	12993.0	47406.0
	WY	6303.0	8700.0

		Observed_Deaths Metropolitan \
Year	State_Code	
2015	AK	7077.0
	AL	77061.0
	AR	37557.0
	AZ	103224.0
	CA	538362.0
...		...
	VT	3705.0
	WA	101973.0
	WI	70308.0
	WV	28209.0
	WY	2676.0

		Observed_Deaths Nonmetropolitan \
Year	State_Code	
2015	AK	3921.0
	AL	28647.0
	AR	29472.0
	AZ	5862.0
	CA	17499.0
...		...
	VT	9267.0
	WA	15681.0
	WI	31695.0
	WV	19197.0
	WY	6024.0

		Percent_Potentially_Excess_Deaths All \
Year	State_Code	
2015	AK	333.6
	AL	718.9
	AR	810.4
	AZ	192.2
	CA	127.4
...		...
	VT	395.2
	WA	219.9
	WI	275.9

WV	825.4
WY	53.1

Year	State_Code	Percent_Potentially_Excess_Deaths Metropolitan \
2015	AK	356.2
	AL	666.9
	AR	757.0
	AZ	185.3
	CA	124.6
...		...
	VT	333.5
	WA	189.1
	WI	230.0
	WV	801.8
	WY	177.1

Year	State_Code	Percent_Potentially_Excess_Deaths Nonmetropolitan \
2015	AK	445.9
	AL	861.4
	AR	893.8
	AZ	363.9
	CA	302.1
...		...
	VT	457.8
	WA	458.4
	WI	393.3
	WV	863.6
	WY	97.1

Year	State_Code	Population All	Population Metropolitan \
2015	AK	15752025.0	10730610.0
	AL	98453697.0	75538374.0
	AR	60298329.0	37830675.0
	AZ	138245184.0	131349090.0
	CA	812037297.0	795993207.0
...	
	VT	12369387.0	4445340.0
	WA	146927250.0	132955461.0
	WI	116414547.0	87214530.0
	WV	36394473.0	22640328.0
	WY	11976111.0	3679551.0

Year	State_Code	Population Nonmetropolitan	Potentially_Excess_Deaths All \
------	------------	----------------------------	---------------------------------

2015	AK	4667820.0	1537.0
	AL	22915323.0	28849.0
	AR	22467654.0	20624.0
	AZ	6896094.0	6485.0
	CA	16044090.0	21109.0
...	
	VT	7924047.0	2070.0
	WA	13971789.0	10804.0
	WI	29200017.0	11151.0
	WV	13754145.0	15195.0
	WY	8296560.0	158.0

Year	State_Code	Potentially_Excess_Deaths Metropolitan \
2015	AK	1015.0
	AL	19793.0
	AR	10758.0
	AZ	6083.0
	CA	19663.0
...		...
	VT	467.0
	WA	8209.0
	WI	6514.0
	WV	9003.0
	WY	186.0

Year	State_Code	Potentially_Excess_Deaths Nonmetropolitan
2015	AK	661.0
	AL	9069.0
	AR	9927.0
	AZ	505.0
	CA	1961.0
...		...
	VT	1655.0
	WA	2657.0
	WI	4759.0
	WV	6204.0
	WY	139.0

[51 rows x 15 columns]

```
[63]: # Print zcores
zscore_2015 = stats.zscore(df_2015['Expected_Deaths All'])
zscore_2015
```

```
[63]: array([-8.27748071e-01, -1.78609409e-01, -4.72780570e-01,  9.68299073e-02,
          4.40273793e+00, -1.62525925e-01, -3.68479568e-01, -8.40477607e-01,
          -7.70585980e-01,  2.44785047e+00,  4.79721196e-01, -7.09393352e-01,
          -4.55488892e-01, -6.84968495e-01,  9.22954544e-01,  4.10983182e-02,
          -5.10998174e-01, -2.51487692e-01, -2.54097392e-01,  1.00647801e-01,
          -4.40068921e-02, -6.87230235e-01,  6.15203278e-01, -1.18895611e-01,
          -3.84651178e-03, -4.86669972e-01, -7.54096543e-01,  5.74443634e-01,
          -8.18971747e-01, -6.53980726e-01, -7.01863885e-01,  4.17127063e-01,
          -6.07305762e-01, -4.95726597e-01,  1.98388450e+00,  8.61887569e-01,
          -3.61694348e-01, -2.93706836e-01,  1.10275320e+00, -7.60079522e-01,
          -1.57064887e-01, -7.95716422e-01,  7.80593999e-02,  2.57396663e+00,
          -5.90246058e-01,  2.95689032e-01, -8.15753118e-01,  1.25352960e-01,
          -4.04789646e-02, -6.10147435e-01, -8.35084227e-01])
```

```
[65]: # Display zscores GT than 3 or less than -3
```

```
bad_list = []
bad_list = [x for x in zscore_2015 if x > 3 or x < -3]
print(bad_list)
```

```
[-0.8277480711228766, -0.1786094086790072, -0.4727805699577321,
 0.09682990733741546, 4.402737928959205, -0.1625259254403741,
 -0.3684795677773897, -0.8404776068351841, -0.7705859798528789,
 2.4478504677433035, 0.47972119640905697, -0.7093933522183921,
 -0.4554888923652354, -0.6849684951606843, 0.9229545442900653,
 0.04109831818239237, -0.5109981737591476, -0.2514876921040634,
 -0.2540973919084089, 0.10064780149562463, -0.04400689210376413,
 -0.687230234991117, 0.6152032784739128, -0.11889561093364948,
 -0.003846511781335936, -0.4866699722497488, -0.7540965433113477,
 0.5744436337519682, -0.818971747336411, -0.653980726372789, -0.7018638850051138,
 0.4171270633240882, -0.607305762094328, -0.49572659712631084, 1.983884504749995,
 0.8618875688683802, -0.36169434828609137, -0.2937068356054753,
 1.1027531952546403, -0.7600795217516805, -0.15706488696091034,
 -0.7957164224132431, 0.07805939985578958, 2.5739666271770085,
 -0.590246057817773, 0.29568903242854366, -0.8157531175777182,
 0.12535295964342885, -0.04047896459048223, -0.6101474352146152,
 -0.8350842272395368]
```

```
[72]: # Check zscores for other features
```

```
zscore_2015 = stats.zscore(df_2015['Observed_Deaths All'])
zscore_2015
```

```
[72]: array([-8.90443170e-01, -4.82799333e-02, -3.92377877e-01, -1.82284104e-02,
          3.95639241e+00, -3.42309798e-01, -4.71082976e-01, -8.90149594e-01,
          -8.22947343e-01,  2.57599733e+00,  5.88780326e-01, -7.88785754e-01,
          -4.74258936e-01, -7.59294695e-01,  1.09343774e+00,  1.91545098e-01,
          -5.31559664e-01, -4.06736420e-02, -1.13613972e-01, -1.28105959e-03,
```



```
-7.79845026e-02, -7.20035556e-01, 7.76348801e-01, -2.00272316e-01,
 1.29413707e-01, -3.93285294e-01, -8.12832310e-01, 7.05410126e-01,
-8.84358137e-01, -7.11708669e-01, -7.57800125e-01, 3.03584434e-01,
-6.84139199e-01, -5.54084962e-01, 1.85721614e+00, 1.17136886e+00,
-2.53329534e-01, -3.32781917e-01, 1.28556999e+00, -8.17823105e-01,
-9.11420521e-02, -8.60391647e-01, 3.14473441e-01, 2.59177037e+00,
-7.27935424e-01, 2.98647017e-01, -8.73282310e-01, 5.79946352e-02,
-8.12405290e-02, -5.66948935e-01, -9.11287077e-01])
```

```
[73]: # Check summary statistics of Cancer - Expected Deaths All 2015
df_2015['Expected_Deaths All'].describe()
```

```
[73]: count      51.000000
      mean     95337.960784
      std     104489.655432
      min      8382.000000
      25%     26074.000000
      50%     69049.000000
      75%    107029.000000
      max     550846.000000
      Name: Expected_Deaths All, dtype: float64
```

```
[74]: # Check summary statistics of Cancer - Expected Deaths All 2014
df_2014['Expected_Deaths All'].describe()
```

```
[74]: count      51.000000
      mean     93917.647059
      std     102559.209756
      min      8260.000000
      25%     25696.000000
      50%     68254.000000
      75%    105117.000000
      max     540349.000000
      Name: Expected_Deaths All, dtype: float64
```

```
[ ]:
```