



# **KLEF**

**KONERU LAKSHMAIAH EDUCATION FOUNDATION**  
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(NAAC Accredited "A" Grade University)

## **PROJECT REPORT**

**On**

### **THE DEATH OF AMERICA'S RURAL HOSPITALS**

**Submitted in partial fulfilment of the  
Requirements for the award of the Degree of  
Bachelor of Technology**

**In**

**Computer Science and Engineering**

**Under the esteemed guidance of**

**B. Manjula**

**By**

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**(DST-FIST Sponsored Department)**

**K L EDUCATION FOUNDATION**

**Green Fields, Vaddeswaram, Guntur District-522 502**

**2019-2020**

**K L EDUCATION FOUNDATION**

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**DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING**

**(DST-FIST Sponsored Department)**



**CERTIFICATE**

This is to certify that this project based lab report entitled “**THE DEATH OF AMERICA’S RURAL HOSPITALS** “ is a bonafide work done by **Maganti Sreelekha (180030523)** in the course **18CS3262S DATA MODELING & VISUALIZATION** in partial fulfilment of the requirements for the award of Degree in Bachelor of Technology in **COMPUTER SCIENCE AND ENGINEERING** during the Even Semester of Academic year 2020-2021

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ENGINEERING

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**DECLARATION**

We hereby declare that this project based lab report entitled “**THE DEATH OF AMERICA’S RURAL HOSPITALS**” has been prepared by us in the course 18CS3262S DATA MODELING & VISUALIZATION in partial fulfilment of the requirement for the award of degree bachelor of technology in COMPUTER SCIENCE AND ENGINEERING during the Even Semester of the academic year 2020-2021. We also declare that this project-based lab report is of our own effort and it has not been submitted to any other university for the award of any degree.

Date: 3/11/2020

Place: VIJAYAWADA

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## **ABSTRACT:**

The project is totally based on nonmetropolitan and metropolitan areas have significant numbers of deaths from the five leading causes of death. These figures accompany this report by presenting information about deaths in nonmetropolitan and metropolitan areas at the state level i.e rural states of America. They also add additional years of data and options for selecting different age ranges and benchmarks.

The dataset presents number of deaths and age-adjusted death rates for the 5 leading causes of death i.e cancer,unintentional injury,stroke,respiratory and heart diseases in the United States and by rural states for 2005-2013. This real time death dataset i.e National Centre for Health Statistics by simply showing a dataset that contains 165125 more or less people may confuse while seeing a patient records In these project we have applied data preprocessing,dataanalysis and data visualization.techniques

## INTRODUCTION:

This Project is mainly about visualizing the number of deaths in Rural States of America.

In the United States, there is a rural-urban gap in age-adjusted death rates and potentially excess deaths from the five leading causes of death. Rural communities experience higher age-adjusted death rates and a higher number of potentially excess deaths from the five leading causes compared with urban areas. Higher death rates and potentially excess deaths are often associated with various interconnected societal, geographic, behavioral, and structural factors. Although rural communities are at higher risk for death from the five leading causes of death, funding to address risk factors is allocated on a population basis often resulting in underfunded rural programs. An increased emphasis on need and epidemiologic burden of disease as major factors in targeting future allocation of public health and prevention funding might contribute, among other factors, to bridging the mortality gap from the five leading causes of death between rural and urban areas in the United States

## PROBLEM AND DEFINITION:

This Project is about visualizing the cause of deaths in Rural states of America. It is used to demonstrate and visualize the no of deaths in rural states of America using tableau and python, here we represented the no of deaths from various aspects and the ways to cure them. major no of deaths are causing to the people

For many life-threatening diseases, the common risk factors include overweight, smoking, drinking, stressful lifestyle etc like symptoms' Cancer, stroke, respiratory and heart diseases are some very serious diseases across the states that have resulted in a lot of deaths. Causes of deaths can be life-threatening and requires immediate emergency

### 1.3-SCOPE:

Compared with metropolitan areas, nonmetropolitan areas have higher age-adjusted death rates and greater percentages of potentially excess deaths from the five leading causes of death, nationally and across public health regions.

Not all **potentially excess deaths** can be prevented; some areas might have characteristics that predispose them to higher rates of death. However, many potentially excess deaths might represent deaths that could be prevented through improved public health programs that support healthier behaviors and neighborhoods or better access to health care services.

**PURPOSE:**

To demonstrate and visualize the cause of deaths in rural states of America using tableau and python, here we represented the no of deaths from various rural states of America and the ways to cure them. Major no of deaths are causing due to cancer,unintentional injury,heart and repository diseases

**PROBLEM AND EXISTING TECHNOLOGY:**

To better understand digital visualization challenges, first let us understand what exactly data visualization is. As defined by Oracle, "Data visualization is the presentation of abstract information in graphical form. Data visualization allows us to spot patterns, trends, and correlations that otherwise might go unnoticed in traditional reports, tables, or spreadsheets." In the early days of data sciences, businesses were suffering to calculate margins around the globe. They used to maintain data by Excel, CSV, MS Access Application, Flat files etc. It was very tough to maintain & visualize data. Especially when connecting to legacy systems, fetching data takes so much time & estimating the time to extract is not constant every time. Even after development, we must check whether the data visualization is similar on all the browser platforms or not. After completing development, we need to host manually on either the individual server or cloud servers. This takes huge time to publish the report. After publishing the report, we need to provide authentication & authorization for security purpose. Authorization is to validate the user and ensure that user is having permission to view or not.

**PROPOSED SYSTEM:**

Mortality data for U.S. residents from the National Vital Statistics System were used to calculate age-adjusted death rates and potentially excess deaths for nonmetropolitan and metropolitan areas for the five leading causes of death. Age-adjusted death rates included all ages and were adjusted to the 2005 U.S. standard population by the direct method. Potentially excess deaths are defined as deaths among persons aged <80 years that exceed the numbers that would be expected if the death rates of states with the lowest rates (i.e., benchmark states) occurred across all states. (Benchmark states were the three states with the lowest rates for each cause during 2008–2010.) Potentially excess deaths were calculated separately for nonmetropolitan and metropolitan areas. Data are presented for the United States and the 10 U.S. Department of Health and Human Services public health regions.



## **REQUIREMENTS AND ANALYSIS:**

### **PLATFORM REQUIREMENTS:**

OS WINDOWS 7 OR LATER/MAC

TOOL:PYTHON JUPITER NOTEBOOK,TABLEAU

LANGUAGE:PYTHON

RAM:1GB/MORE

HARDDISK:6GB/MORE

PROCESSOR:ANY INTEL PROCESSOR

### **MODULE DESCRIPTION:**

This module makes us understand the statistics of the american states death rate. We have analysed the data and visualized to clearly understand the death rate and how it got affected with each attribute provided in the dataset

## **ALGORITHM:**

### **IN JUPYTER:**

Step 1 : Open jupyter notebook.

Step 2 : import the packages what ever we need for this project.

Step 3 : read input data in csv

Step 4 : Use pandas dataframe for reading csv file

Step 5 : Display the data present in the input file

Step 6 : Now visualize the data that represents the opioid deaths in Connecticut.

### **IN TABLEAU:**

Step 1 : Open Tableau Desktop

Step 2 : Click on the text file or Microsoft excel based on the data

Step 3 : Import dataset

Step 4 : Perform the visualization based on problem statement and drag the values into rows and columns whatever we need.

Step 5 : Apply different colours, shapes to the visualization for understanding easily

Step 6 : Rename the sheets Step

7: Save the book.

## PSEUDO CODE:

**#180030523**

```
import pandas as pd import numpy as np from  
  
sklearn.model_selection import train_test_split  
  
import seaborn as sns import matplotlib.pyplot as  
  
plt import missingno as msno  
  
%matplotlib inline import warnings warnings.filterwarnings("ignore")  
  
import chart_studio.plotly as pt import plotly.graph_objs as go from  
  
plotly.offline import download_plotlyjs,init_notebook_mode,plot,iplot  
  
init_notebook_mode(connected = True)  
  
df.describe()  
df.info() df.isnull().sum() dfcod=df['Cause of Death'].value_counts() dfcod  
names=['Heart Disease','Stroke','Cancer','Respiratory disease','injury'] fig =  
plt.figure(figsize =(10, 7))  
  
plt.pie(dfcod,colors=['yellow','lightblue','lightgreen','pink','hotpink'],labels=names)  
plt.show() dfl=df['Locality'].value_counts()  
  
dfl  
  
dfmet=df[df['Locality']=='Metropolitan']
```

```

dfem=dfmet['Expected Deaths'].sum()
dfpedm=dfmet['Potentially Excess Deaths'].sum()
dfnon=df[df['Locality']=='Nonmetropolitan']
dfen=dfnon['Expected Deaths'].sum()

dfpedn=dfnon['Potentially Excess Deaths'].sum()
locality=['Metropolitan','Nonmetropolitan']

Expected=[dfem,dfen] Ped=[dfpedm,dfpedn]

dfplt=pd.DataFrame({'Locality':locality,'Expected
Deaths':Expected,'PED':Ped}) width=0.35

fig = plt.figure(figsize=(5,5)) p1=plt.bar(dfplt['Locality'],dfplt['Expected
Deaths'],width,color='crimson')

p2=plt.bar(dfplt['Locality'],dfplt['PED'],width)

plt.yticks(np.arange(10000000,50000000,10000000))

plt.legend((p1,p2),('Excess Deaths','Expected Deaths'))

dfstate=df['State'].value_counts() dfstate

msno.heatmap(df) msno.bar(df,color='green')

plt.figure(figsize=(8,6))

sns.heatmap(df.corr(),annot=True)

df.isnull().sum()

df.hist(bins=10,figsize=(20,20)) plt.show()

```

# SCREENSHOTS:

Output:1

Out[4]:

	Year	Cause of Death	State	State FIPS Code	HHS Region	Age Range	Benchmark	Locality	Observed Deaths	Population	Expected Deaths	Potentially Excess Deaths	Potentially Excess Deaths	Percent Excess Deaths
0	2005.0	Cancer	Alabama	AL	4.0	0-49	2005 Fixed	All	756.0	3148377.0	451.0	305.0		40.3
1	2005.0	Cancer	Alabama	AL	4.0	0-49	2005 Fixed	Metropolitan	556.0	2379871.0	341.0	217.0		39.0
2	2005.0	Cancer	Alabama	AL	4.0	0-49	2005 Fixed	Nonmetropolitan	200.0	768506.0	111.0	89.0		44.5
3	2005.0	Cancer	Alabama	AL	4.0	0-49	2010 Fixed	All	756.0	3148377.0	421.0	335.0		44.3
4	2005.0	Cancer	Alabama	AL	4.0	0-49	2010 Fixed	Metropolitan	556.0	2379871.0	318.0	238.0		42.8

Output 2

Out[4]:

	Year	HHS Region	Observed Deaths	Population	Expected Deaths	Potentially Excess Deaths	Percent Potentially Excess Deaths
count	165124.000000	165124.000000	156824.000000	1.608820e+05	156824.000000	156824.000000	156824.000000
mean	2008.93405	5.237216	2964.486303	6.965942e+06	2095.181898	879.752015	35.618019
std	2.55710	2.822844	17159.402910	2.839582e+07	13138.396045	4472.576572	18.537710
min	2005.000000	0.000000	10.000000	5.666500e+04	2.000000	0.000000	0.000000
25%	2007.000000	3.000000	153.000000	6.819730e+05	91.000000	41.000000	21.100000
50%	2009.000000	5.000000	503.000000	1.608654e+06	296.000000	158.000000	35.600000
75%	2011.000000	8.000000	1571.000000	4.642670e+06	1011.000000	539.000000	49.800000
max	2041.000000	45.000000	486080.000000	3.100880e+08	442680.000000	175703.000000	84.300000

Output 3

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 165125 entries, 0 to 165124
Data columns (total 13 columns):
#   Column                                Non-Null Count  Dtype
---  ---                                -
0   Year                                165124 non-null float64
1   Cause of Death                     165124 non-null object
2   State                              165124 non-null object
3   State FIPS Code                    165124 non-null object
4   HHS Region                         165124 non-null float64
5   Age Range                         165124 non-null object
6   Benchmark                         165124 non-null object
7   Locality                          165124 non-null object
8   Observed Deaths                   156824 non-null float64
9   Population                        160882 non-null float64
10  Expected Deaths                   156824 non-null float64
11  Potentially Excess Deaths         156824 non-null float64
12  Percent Potentially Excess Deaths 156824 non-null float64
dtypes: float64(7), object(6)
memory usage: 16.44 MB
```

Output 4

```
In [6]: df.isnull().sum()

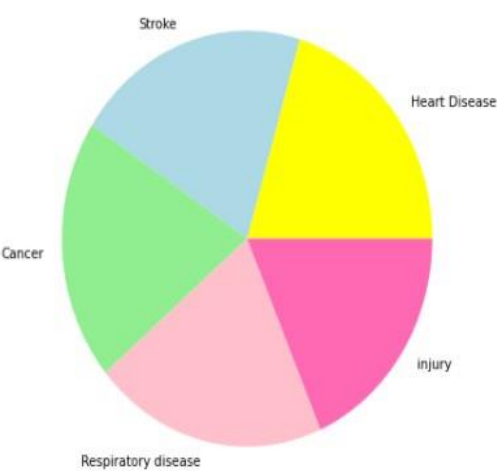
Out[6]: Year          1
Cause of Death      1
State               1
State FIPS Code     1
HHS Region         1
Age Range          1
Benchmark          1
Locality           1
Observed Deaths    8301
Population         4243
Expected Deaths    8301
Potentially Excess Deaths 8301
Percent Potentially Excess Deaths 8301
dtype: int64
```

Output:5

```
0]: Cancer          33700
Stroke             33649
Chronic Lower Respiratory Disease 33625
Heart Disease      33581
Unintentional Injury 30569
Name: Cause of Death, dtype: int64

1: Name: Cause of Death, dtype: int64
```

Output 6

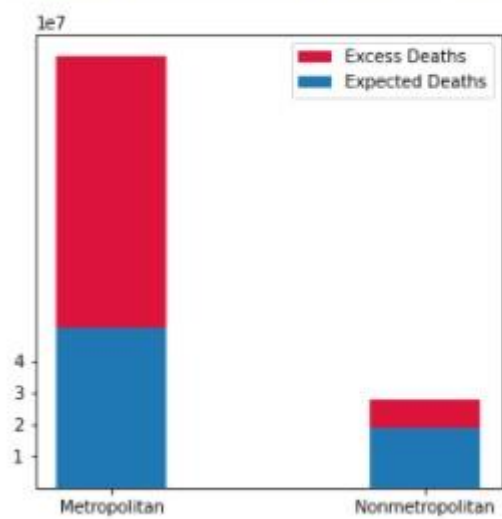


## Output:7

```
: Nonmetropolitan    55061  
  All                55034  
  Metropolitan      55029  
  Name: Locality, dtype: int64
```

## Output 8

```
: <matplotlib.legend.Legend at 0x243fcbd5dc0>
```

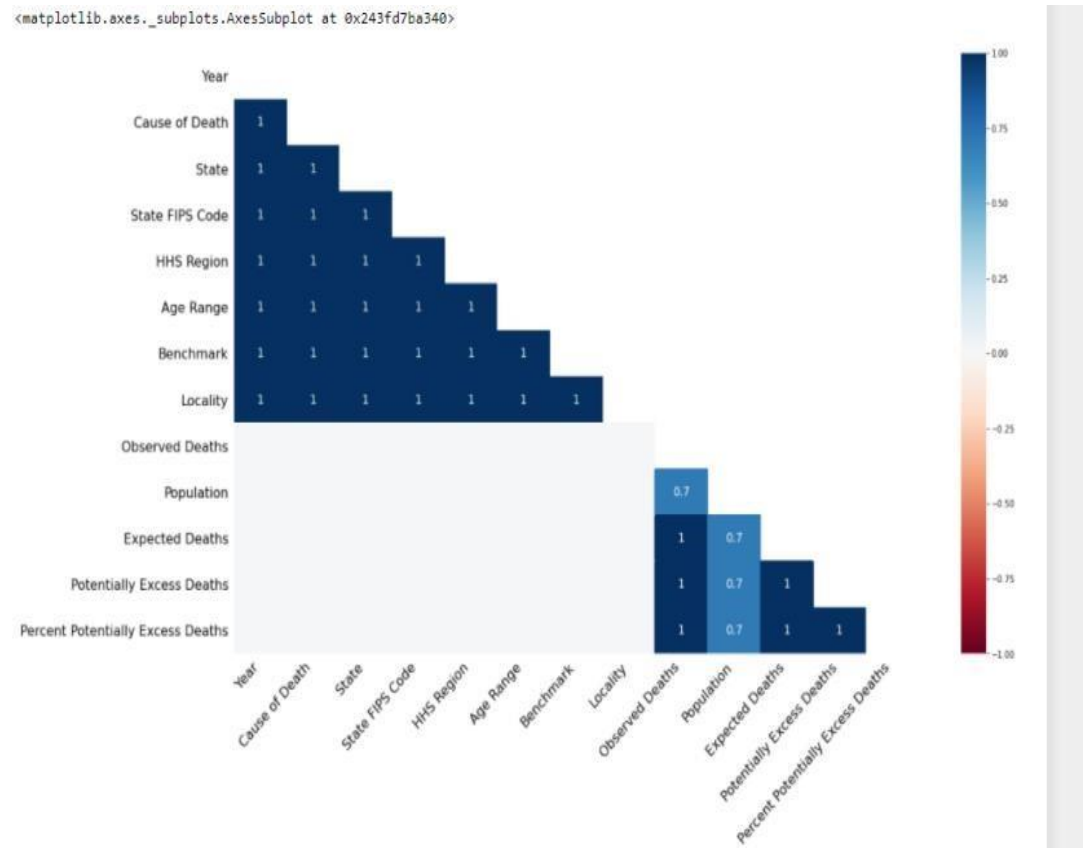


# Output 9

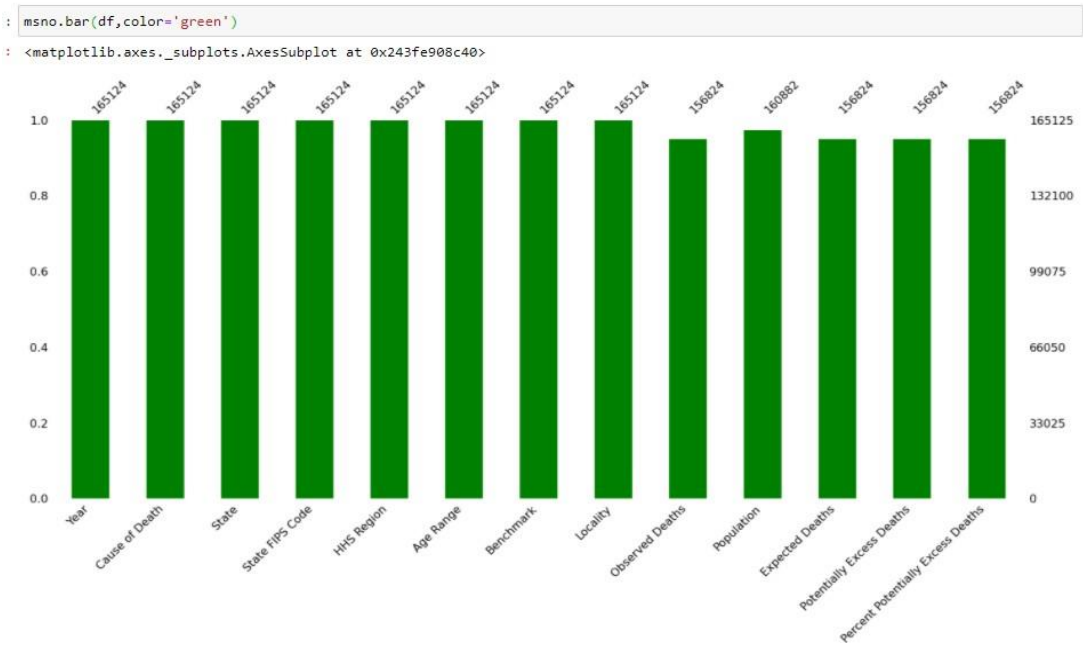
Alabama		3241
Alaska		3241
Arizona		3241
Arkansas		3241
Colorado		3240
California		3240
Connecticut		3238
Delaware		3202
District of\nColumbia		3184
Rhode Island		3169
New Jersey		3169
Nebraska		3168
South Carolina		3168
Indiana		3168
Oklahoma		3168
Texas		3168
Michigan		3168
Tennessee		3168
Ohio		3168
Utah		3168
Louisiana		3168
Pennsylvania		3168
Mississippi		3168
Virginia		3168
Illinois		3168
Kentucky		3168
North Dakota		3168
Oregon		3168
Minnesota		3168
North Carolina		3168
New Hampshire		3168
South Dakota		3168
West Virginia		3168
Washington		3168
New Mexico		3168
New York		3168
United States		3168
Vermont		3168
Wisconsin		3167
Maine		3167
Missouri		3164
Kansas		3161
Massachusetts		3161
Georgia		3160
Nevada		3156
Florida		3156
Hawaii		3154
Wyoming		3152
Iowa		3152
Idaho		3150
Maryland		3144
Montana		3138
Name: State, dtype: int64		



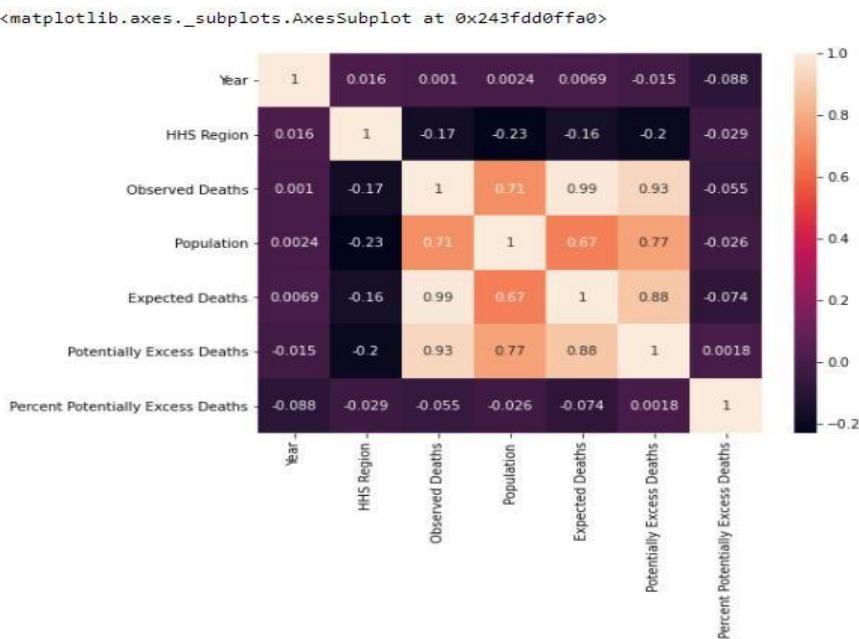
Output 10



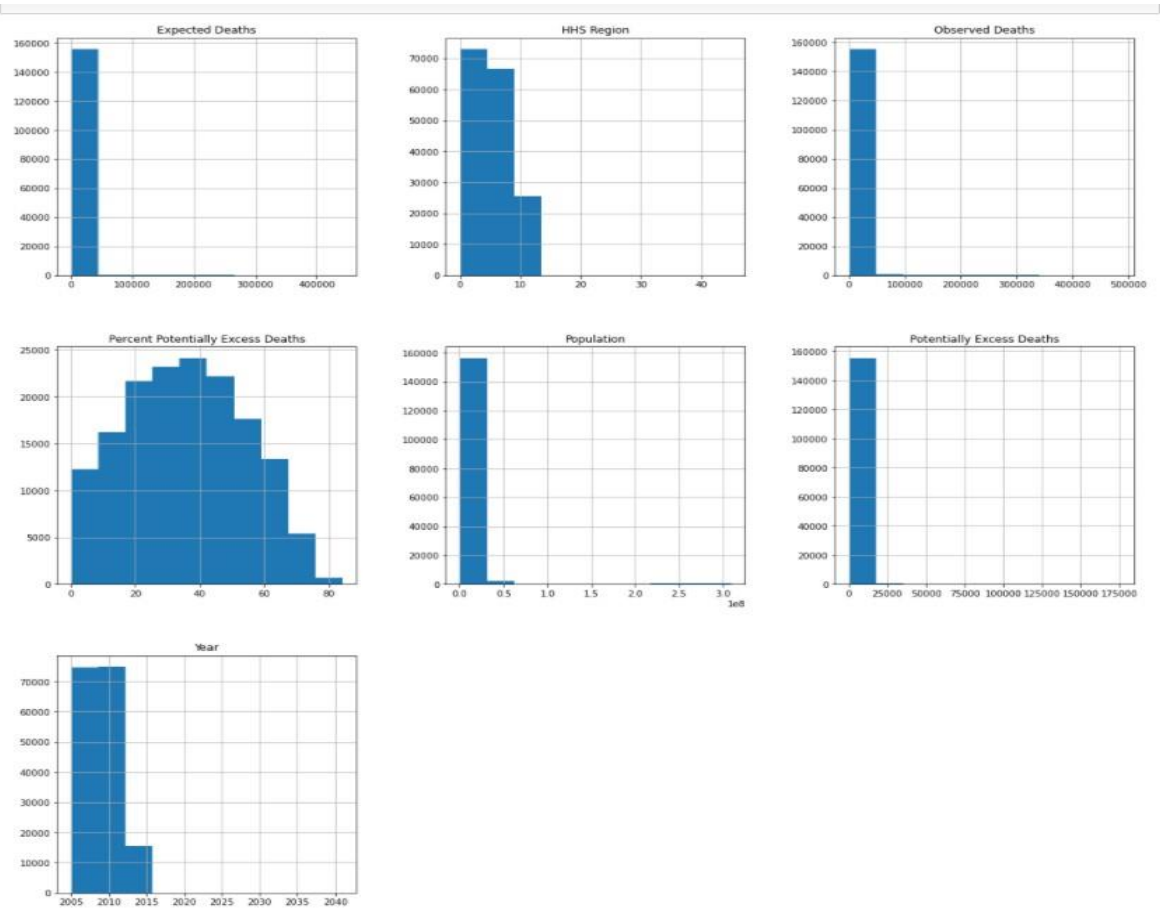
Output 11



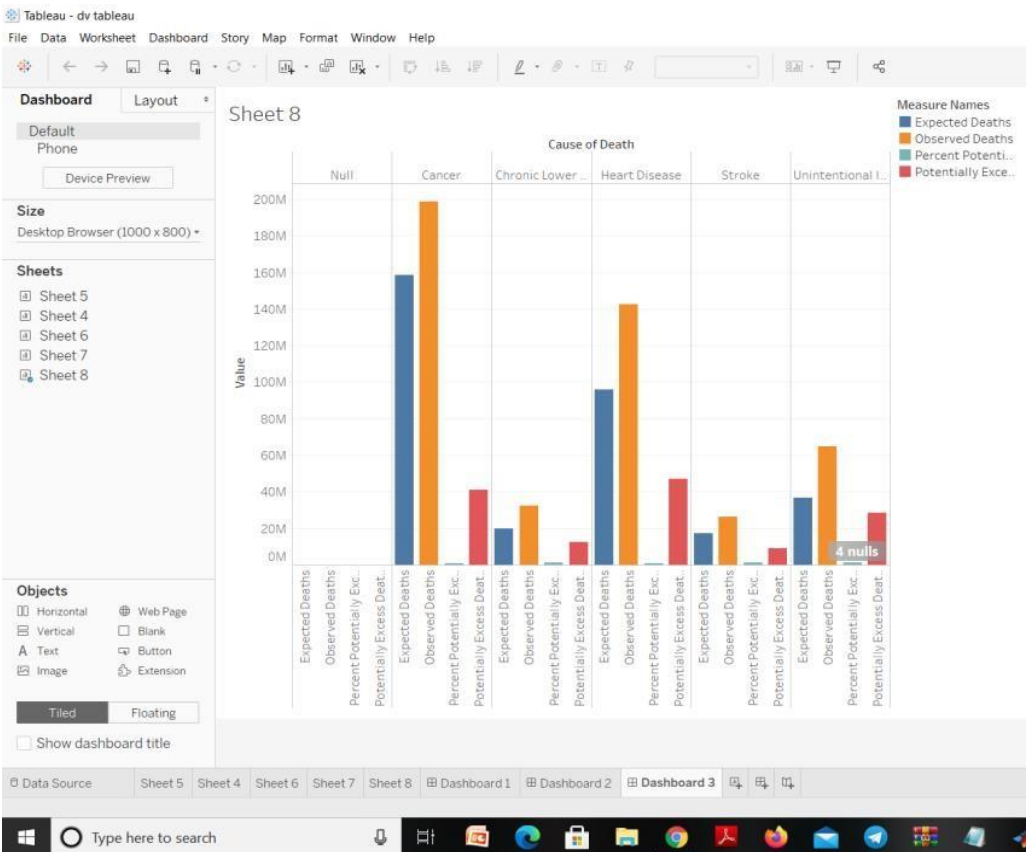
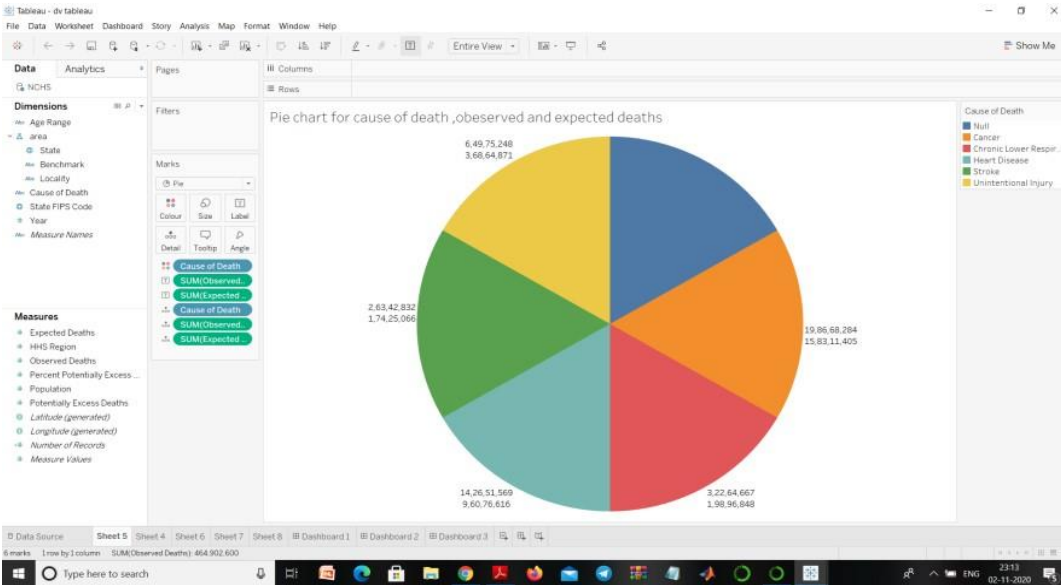
Output 12

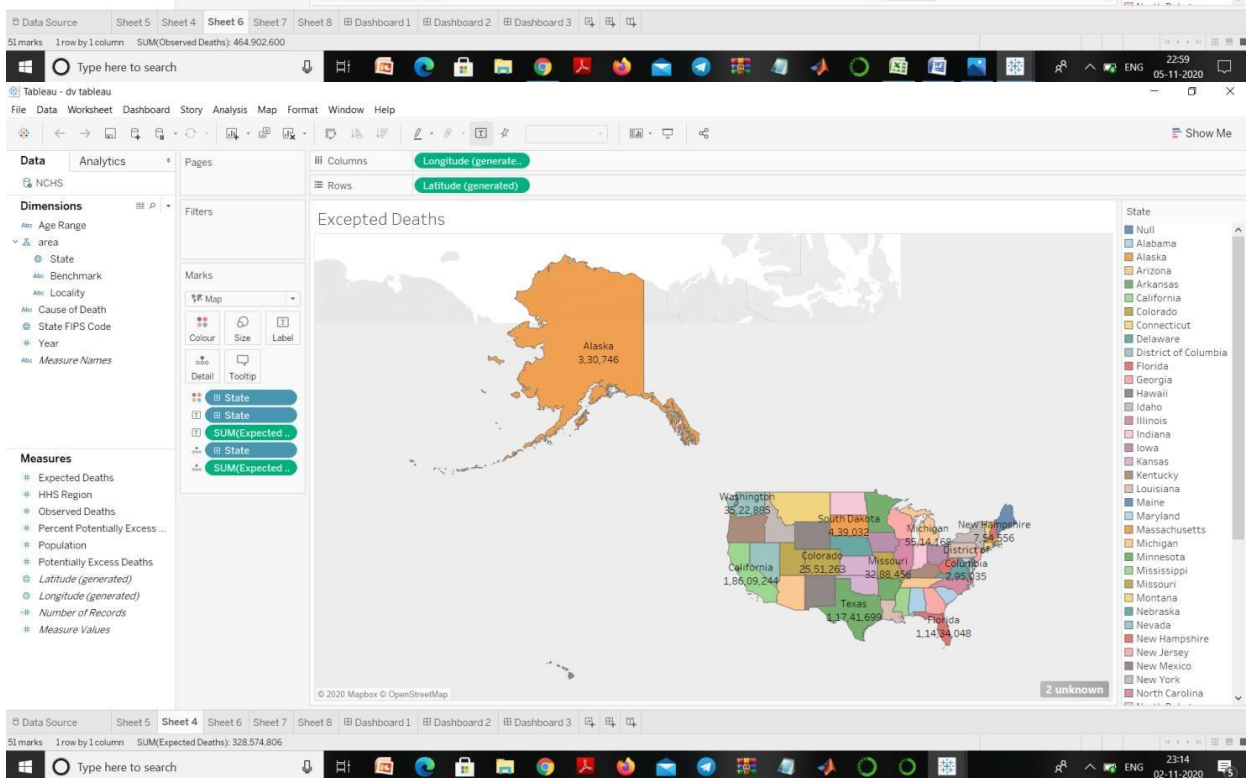
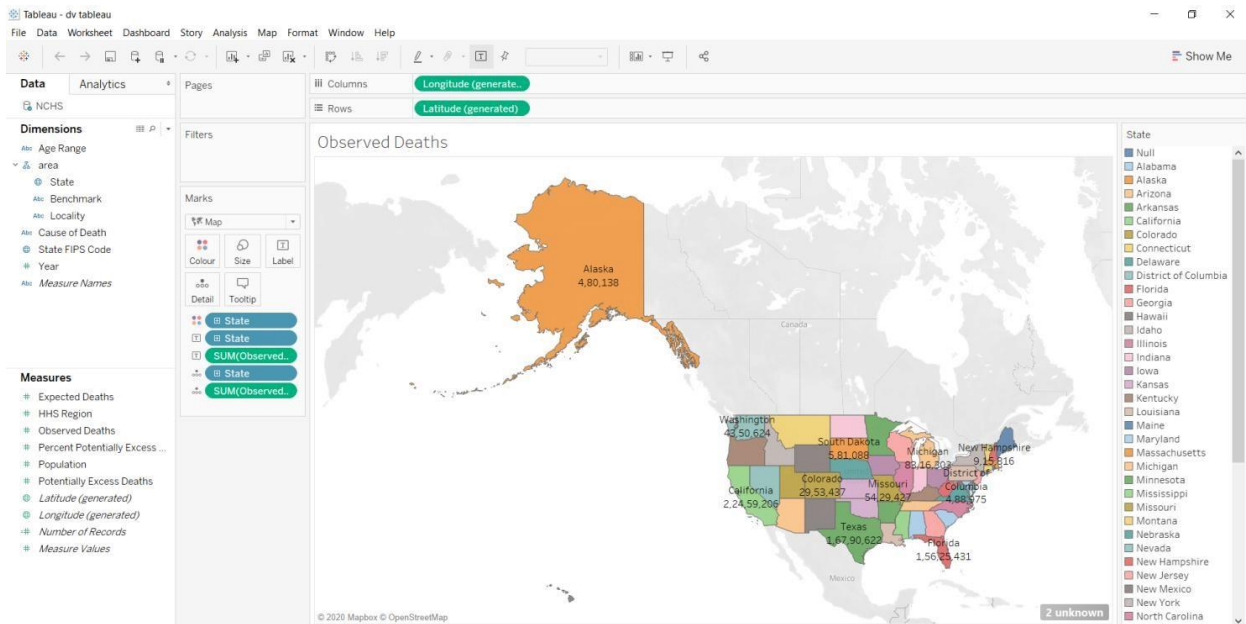


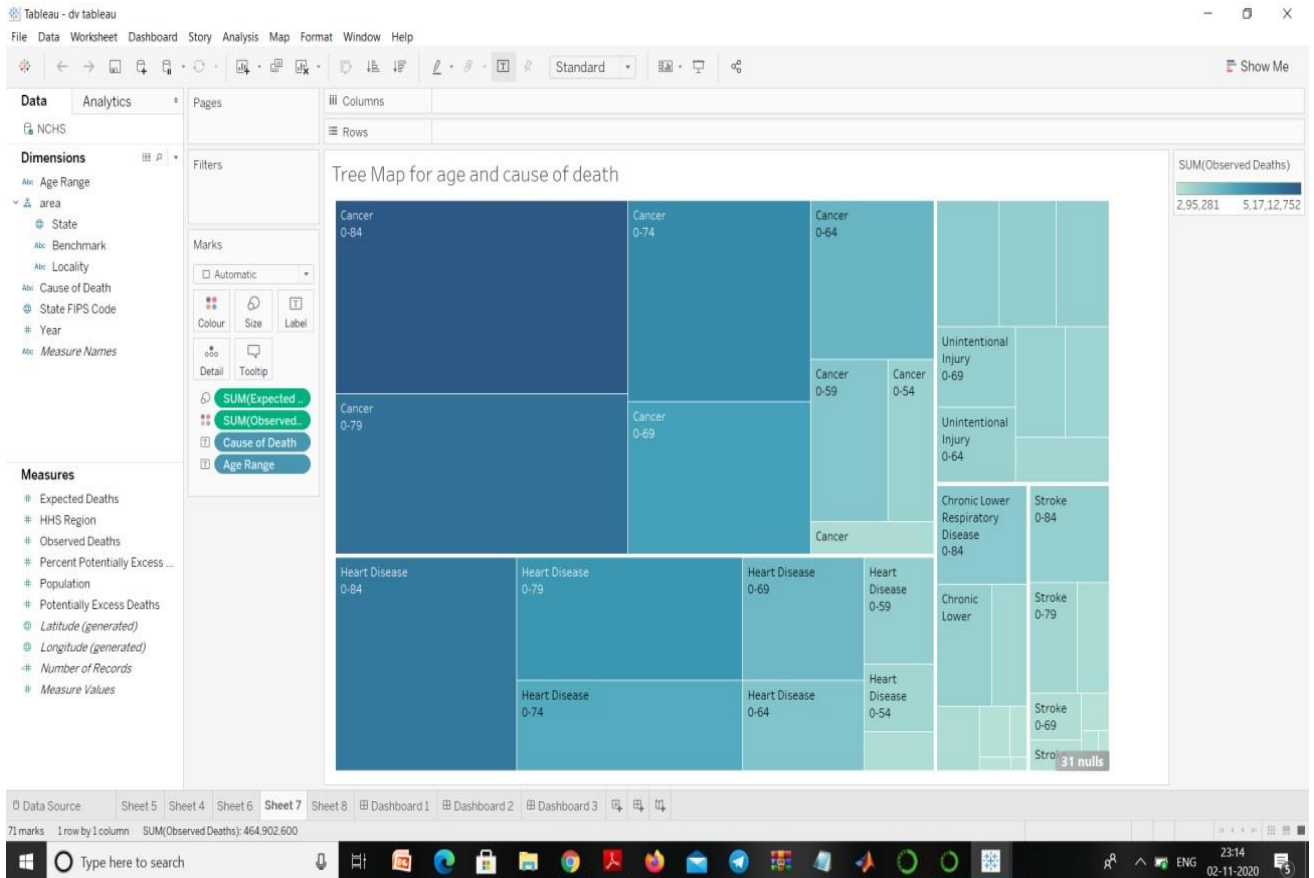
Output 13



## Tableau







## CONCLUSION:

Nonmetropolitan areas have higher age-adjusted death rates and greater percentages of potentially excess deaths from the five leading causes of death. Routine tracking of potentially excess deaths from the five leading causes of death in nonmetropolitan and metropolitan areas might help public health officials monitor important rural health disparities and select effective programs and policies to improve the health of residents of rural areas. Additional information on potentially excess deaths might be used to evaluate the success of public health interventions and to help determine where to allocate resources in areas with the greatest need. State and local public health officials in rural areas might seek advice from officials in rural areas with fewer potentially excess deaths for ways to reduce mortality in their jurisdictions or increase coordination with urban areas to ensure rural residents have timely access to specialized services

REFERENCES:

<https://www.cdc.gov/nchs/data-visualization/potentially-excess-deaths/index.htm>

<https://www.cdc.gov/mmwr/volumes/66/ss/ss6602a1.htmxxi>