LEADING CANCER SITES

Introduction And Project Scope

In the previous project i.e. Leading Causes of Deaths in USA, we implemented data management techniques using Pandas on two datasets (1. NCHS - leading causes of deaths, containing number of deaths and its cause in USA and 2. nst -est2018-01 which had the details about the population of US over the years), to find the top 4 causes of death faced by Americans. The said top death causes found were: 1) Heart Disease 2) Cancer 3) Stroke 4) CLRD

Among these top 4 causes, Cancer being the 2nd largest cause of death in USA, the focus of this project will be to determine the leading cancer sites / cancer types diagnosed in the selected population and time period in USA.

The question we are addressing, related to the top 10 cancer sites for Americans over the years will be done in 2 sub parts:

- A) What are the over all Leading 10 Cancer Sites for Americans?
- B) What are the year wise 10 Leading Cancer Sites for Americans?

Data Collection

For the purpose of this project we collected data regarding the USA Cancer Statistics, Incidence Reports for the period 2010 to 2016 which was available on the website of "Centers for Disease Control and Prevention" in addition to the population data set provided to us previously. Thus the following datasets were used in this project:

```
1) US Cancer typewise stats, 2010-2016. (website ref: https://wonder.cdc.gov/controller/datarequest/D160)
2) nst-est2018-01
```

The first data set contains the details about yearwise and statewise leading cancer sites and count of incidence reported for the period 2010 to 2016. While the second dataset consists of details related to year wise and statewise population of US for the period 2010 to 2017.

Code, Explanation and Analysis

Preliminary Steps

```
In [1]: # Importing the packages
   import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
```

Data reading, merging and data cleaning

```
n [2]: # Reading the nst-est2018-01 dataset
```

```
file = 'C:/RIT/Spring 2021\BANA 680 - Data Management for Business Analytics/Assignment
pop_data = pd.read_excel(file)
pop_df = pd.DataFrame(pop_data)
```

```
In [3]: # Reading the US Cancer typewise stats dataset and converting it to dataframe
    Cancer_types = 'C:/RIT/Spring 2021/BANA 680 - Data Management for Business Analytics/Fi
    df_can_typ = pd.DataFrame(pd.read_csv(Cancer_types, delimiter = "\t"))
```

To check and handle missing values as well as displaying the for unique states and cancer types in the df, we execute the following commands:

```
print('Number of missing values in each column:', '\n', df_can_typ.isnull().sum())
In [4]:
        Number of missing values in each column:
         Notes
                                       7784
        States
                                        57
        States Code
                                        57
        Year
                                        57
        Year Code
                                        57
        Leading Cancer Sites
                                        57
        Leading Cancer Sites Code
                                        57
                                        57
        dtype: int64
         # Dropping column 'Notes' from the df
In [5]:
         df can typ = df can typ.drop('Notes', axis = 1)
         # dropping rows with missing values
         is_NaN = df_can_typ. isnull()
         row_has_NaN = is_NaN. any(axis=1)
         rows_with_NaN = df_can_typ[row_has_NaN]
         #print(rows with NaN)
         df cleaned = df can typ.dropna(subset = ['Year'])
         # Number of unique cancer sites and states included in cleaned df:
In [6]:
         print('Number of leading cancer sites:', '\t' , len(pd.unique(df_cleaned['Leading Cance
         print('Number of States:', '\t', len(pd.unique(df_cleaned['States'])))
                                                  22
        Number of leading cancer sites:
```

Analyzing the trend in reported cancer incidence over the years

In order to analyze the trend in cancer incidences, we need to look at the yearly reported incidence count for USA as whole and ignore the data for individual states. Also, since the population differs for each year, we need to standarized the death count for each year in order to be compared fairly.

Steps:

Number of States:

- We need the total number of deaths per year in order to analyze the death trend over the years. For this, we will apply groupby function of the df_cleaned to group it by year and aggregate the count of incidence for each year.
- We know that population differs for each year. So in order to have fair comparision of incidene
 count over the years, we need to standarized the count variable. For this, we will extract the
 population data from the second df i.e. pop_df and use it to normalize the incidence count over
 the years.
- The final goal to analyze the cancer incidence trend is to compute the number of incidence per year and then normalize it by dividing it by population for the year so that the effect of

population is ignored.

```
# Create a new df with just year wise count of cancer incidence in US
In [7]:
         df year = df cleaned.groupby(['Year']).agg({'Count': sum})
In [8]:
         #Creating a new pop us dataframe with just year and population column from the original
         #purpose of merging it in the next step.
         pop_us = pd.DataFrame({'Year' : pop_df.iloc[2, 3:], 'Population' : pop_df.iloc[3, 3:]})
         # merging the df year with pop us in order to add the population column for each year
         merge_year = pd.merge(df_year, pop_us, on = 'Year', how = 'inner')
         # adding a new column: 'Count per Million' i.e. CPM, for standarizing the incidence cou
In [9]:
         merge year['CPM'] = (merge year['Count']/merge year['Population'])*1000000
         merge year.head(2)
Out[9]:
            Year
                     Count Population
                                        CPM
        0 2010.0 1420850.0 309326085 4593.37
        1 2011.0 1453057.0 3.1158e+08 4663.51
```

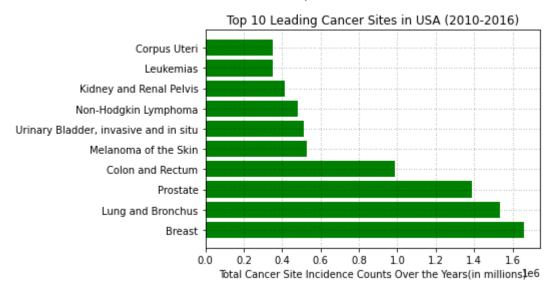
From the above output we can see that there is an increasing trend in Cancer incidences reported in US over the period.

(A) 10 Leading Cancer sites over the years in US

In order to determine the leading 10 cancer sites in US, we will used the df_cleaned created.

We will create a new df say df_topsites by grouping the df_cleaned on 'Leading cancer sites' and aggregating the count for each site. Then we will sort the count values in descending order and create a new df say df_cs_top10, which will contain the details of cancer sites having highest incidence count over the years.

```
# Groupby cancer sites and agg of count
In [10]:
          df topsites = df cleaned.groupby(['Leading Cancer Sites']).sum()['Count'].reset index()
          df_cs_top10 = df_topsites.sort_values(['Count'], ascending = False).head(10)
          df cs top10.head(2)
             Leading Cancer Sites
Out[10]:
                                   Count
                         Breast 1661438.0
          11
               Lung and Bronchus 1533872.0
In [11]:
          #Create a horizontal bar plot with descending order to show top 10 leading causes of Ca
          x = df cs top10["Leading Cancer Sites"]
          y = df cs top10["Count"]
          plt.barh(x, y, color='green', alpha=1, align="center")
          plt.title("Top 10 Leading Cancer Sites in USA (2010-2016)")
          plt.xlabel("Total Cancer Site Incidence Counts Over the Years(in millions)")
          plt.grid(linestyle='--', dashes=(1, 4), linewidth=0.5, color='black')
          plt.show()
```



(B) Year wise 10 leading cancer sites in US

```
In [12]: # Yearlwise top causes
    df_yearwise = df_cleaned.groupby(['Year', 'Leading Cancer Sites']).agg({'Count': sum})
    df_yr = df_yearwise['Count'].groupby('Year', group_keys=False)
    df = df_yr.apply(lambda x: x.sort_values(ascending=False).head(10))

# Converting to dataframe
    df_unstack = df.unstack(level = -1)
# # adding population column for each year
    df_pop = pd.merge(df_unstack, pop_us, on = 'Year', how = 'inner')
# # # pop.head(2)
```

```
In [14]: # Standarizing the incidence count by making it count per million
    df_pop.iloc[:, 1:11] = df_pop.iloc[:, 1:11].div(df_pop["Population"], axis = 0).mul(100
    df_pop.head(3)

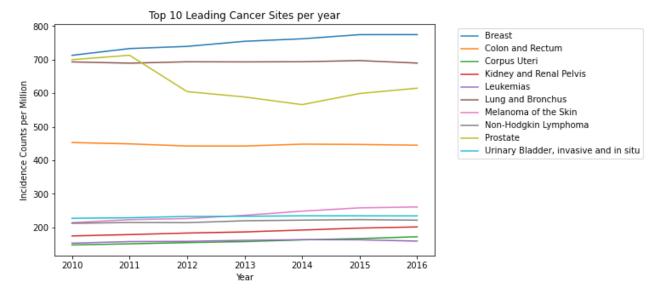
# dropping the population column from above df
    df_final = df_pop.drop('Population', axis = 1)
    df_final.head(2)
```

Out[14]:

	Year	Breast	Colon and Rectum	Corpus Uteri	Kidney and Renal Pelvis	Leukemias	Lung and Bronchus	Melanoma of the Skin	Non- Hodgkin Lymphoma	Prostate	i
0	2010.0	713.105	452.859	147.275	174.23	152.515	693.427	213.923	211.796	700.099	
1	2011.0	733.266	448.909	150.719	178.205	157.388	689.659	222.675	214.314	713.316	

```
In [15]: # plotting the yearly trend for each cancer site
plt.rcParams['figure.figsize'] = [8, 5]
df_final.plot.line(x = 'Year')
plt.title("Top 10 Leading Cancer Sites per year")
plt.xlabel("Year")
```

```
plt.ylabel("Incidence Counts per Million")
plt.legend(bbox_to_anchor=(1.05, 1.0), loc='upper left')
plt.show()
```



From the above plot we can conclude:

- Breast cancer has the highest incidence count for all the years with a gradual yearly increase
 while Leukemias and Corpus Uteri has the lowest reported incidence. Also Leukemias shows a
 declining trend over the years with incidence reported for Corpus Uteri is almost constant for all
 all years.
- Number of per million incidence reported for Prostate Cancer shows a significant decline over the years 2012 to 2014 and then again has a gradual increase from 2015.

Conclusion

- From the above analysis, it was seen that the per million incidences reported for various cancer sites / cancer types had a increasing trend over the years.
- The 10 leading cancer sites for over all year (2010 2016) and the year wise leading cancer sites
 consisted of the same 10 cancer sites.
- Breast cancer has the highest reported incidence over years with gradual increase each year while prostate cancer shows a significant decline in reported incidence for the period 2012 to 2014 with again a gradual increase from 2015.