## **Exercise1**

Question. Write a function temp\_tester that takes a definition of normal body temperature and returns a function that returns True if its argument is within 1 degree of normal temperature and False if not.

```
In [30]: ##Creating a function to assess whether a given body temperature is normal

def temp_tester(temp):

    def inner_function(x):
        if x <= temp + 1 and x >= temp -1:
            return True
        else:
            return False

    return inner_function
```

```
In [31]: ##Normal Body Temperature Values in %
human_tester = temp_tester(37)
chicken_tester = temp_tester(41.1)
```

Test the function:

chicken\_tester(42) # True -- i.e. not a fever for a chicken human\_tester(42) # False -- this would be a severe fever for a human chicken\_tester(43) # False

human\_tester(35) # False -- too low

human\_tester(98.6) # False -- normal in degrees F but our reference temp was in degrees C

```
In [32]: ## Testing the function

print(chicken_tester(42))
print(human_tester(42))
print(chicken_tester(43))
print(human_tester(35))
print(human_tester(98.6))
True
```

False False False False

## **Exercise 2**

Question. Download the sqlite3 database from hw1-population.db. Examine data. What columns

the mean, standard deviation, minimum, maximum. (2 points) Plot a histogram of the distribution with an appropriate number of bins for the size of the dataset (describe in your readme the role of the number of bins). (3 points) Comment on any outliers or patterns you notice in the distribution of ages. (1 point)

Repeat the above for the distribution of weights. (3 points)

Make a scatterplot of the weights vs the ages. (3 points) Describe the general relationship between the two variables (3 points). You should notice at least one outlier that does not follow the general relationship. What is the name of the person? (3 points) Be sure to explain your process for identifying the person whose values don't follow the usual relationship in the readme. (3 points)

Below is the code to load the dataset.

```
In [33]: ## Importing the ibraries pandas and sqlite3 and loading the dataset.

import pandas as pd
import sqlite3
with sqlite3.connect("/Users/mahimakaur/Downloads/hwl-population.db") as db
    data = pd.read_sql_query("SELECT * FROM population", db)
In [34]: ##Overview of the dataset
```

In [34]: ##Overview of the dataset
 data

#### Out[34]:

	name	age	weight	eyecolor
0	Edna Phelps	88.895690	67.122450	brown
1	Cara Yasso	9.274597	29.251244	brown
2	Gail Rave	18.345613	55.347903	brown
3	Richard Adams	16.367545	70.352184	brown
4	Krista Slater	49.971604	70.563859	brown
152356	John Fowler	23.930833	71.532569	blue
152357	Diana Shuffler	21.884819	67.936753	brown
152358	Kevin Cuningham	87.705907	60.074646	brown
152359	James Libengood	21.727666	81.774985	brown
152360	Cathleen Ballance	10.062236	34.327767	brown

152361 rows × 4 columns

Part1: Examine data. What columns does it have? How many rows does it have?

Below is the code to find the columns and number of rows in the dataset.

```
In [35]: data.info()
        print("The columns in the data include:", ", ".join(list(data.columns)))
        print("There are", data['name'].count(), "rows in the data.")
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 152361 entries, 0 to 152360
         Data columns (total 4 columns):
             Column
                       Non-Null Count
                                        Dtype
                     152361 non-null object
          0
            name
                       152361 non-null float64
          1
             age
          2
             weight
                      152361 non-null float64
             eyecolor 152361 non-null object
         dtypes: float64(2), object(2)
         memory usage: 4.6+ MB
         The columns in the data include: name, age, weight, eyecolor
         There are 152361 rows in the data.
```

**Response**: The dataset has four columns namely : *name*, *age*, *weight*, *eyecolor*. There are *152361* rows (people) in the dataset.

**Part2**: Examine the distribution of the ages in the dataset. In particular, be sure to have your code report the mean, standard deviation, minimum, maximum. Plot a histogram of the distribution with an appropriate number of bins for the size of the dataset (describe in your readme the role of the number of bins). Comment on any outliers or patterns you notice in the distribution of ages.

Below is the code to examine the distribution of age in the dataset.

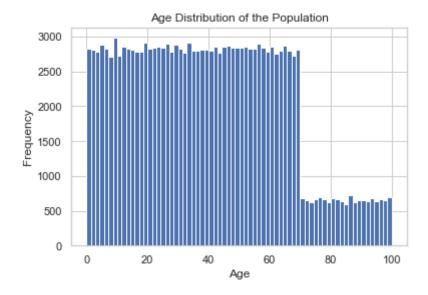
```
In [36]: ##To examine the distribution of the age in the dataset.
         data['age'].describe()
Out[36]: count
                  152361.000000
         mean
                      39.510528
         std
                      24.152760
         min
                      0.000748
         25%
                     19.296458
         50%
                      38.468955
         75%
                      57.623245
                      99.991547
         max
         Name: age, dtype: float64
```

Below is the code for Age distribution.

```
In [37]: | ## import libraries matplotlib and numpy
         import matplotlib.pyplot as plt
         import numpy as np
         age = data['age'] #Creating a variable age
         ## Calculating bins for the histogram using Freedman-Diaconis rule:
         q1 = age.quantile(0.25)
         q3 = age.quantile(0.75)
         iqr = q3 - q1
         bin_width = (2 * iqr) / (len(age) ** (1 / 3))
         bin_count = int(np.ceil((age.max() - age.min()) / bin_width))
         print("Freedman—Diaconis number of bins:", bin_count)
         ##Histogram of the Age Distribution
         plt.hist(age, bins=bin_count);
         plt.ylabel('Frequency')
         plt.xlabel('Age')
         plt.title('Age Distribution of the Population')
```

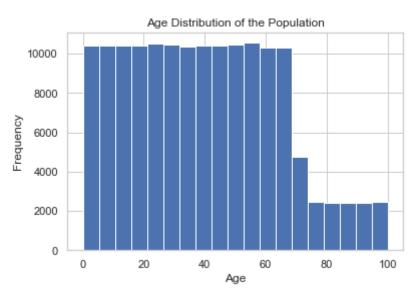
Freedman—Diaconis number of bins: 70

Out[37]: Text(0.5, 1.0, 'Age Distribution of the Population')



Sturges Rule number of bins: 19

Out[38]: Text(0.5, 1.0, 'Age Distribution of the Population')



**Response:** The mean age is 39.5 years, and the maximum age included in the dataset is 99 years. It is important to have appropriate number of bins as if we take less bins, the histogram doesn't portray the data accurately. If large number of bins are used, the graph does not give a sense of the distribution of the dataset. I have calculated the width of bins using Freedman–Diaconis rule. Freedman–Diaconis rule not only considers the sample size but also considers the spread of the sample. Another method through which the bin size can be calculated is Sturge's rule. Sturges rule takes into account the size of the data to decide on the number of bins.

From the graph it seems that the age distribution is skewed right.

**Part3**: Repeat the above for the distribution of weights.

Below is the code to examine the age distribution.

```
In [39]: ##To examine the distribution of the weight in the dataset.
data['weight'].describe()
```

```
Out[39]: count
                152361.000000
        mean
                    60.884134
        std
                    18.411824
        min
                    3.382084
        25%
                    58.300135
        50%
                    68.000000
        75%
                    71.529860
                   100.435793
        max
        Name: weight, dtype: float64
```

Below is the code to plot histogram to depict the distribution of weights.

```
In [40]: ##To depict the distribution of weight

weight = data['weight'] #Creating a variable weight

## Calculating bins for the histogram using Freedman—Diaconis rule:

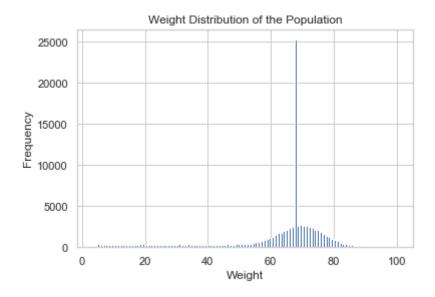
q1 = weight.quantile(0.25)
q3 = weight.quantile(0.75)
iqr = q3 - q1
bin_width = (2 * iqr) / (len(weight) ** (1 / 3))
bin_count_weight = int(np.ceil((weight.max() - weight.min()) / bin_width))
print("Freedman—Diaconis number of bins:", bin_count_weight)

## Histogram

plt.hist(weight, bins= bin_count_weight);
plt.ylabel('Frequency')
plt.xlabel('Weight')
plt.title('Weight Distribution of the Population')
```

Freedman-Diaconis number of bins: 196

Out[40]: Text(0.5, 1.0, 'Weight Distribution of the Population')



```
In [41]: ## Calculating bins for the histogram using Sturge's rule:

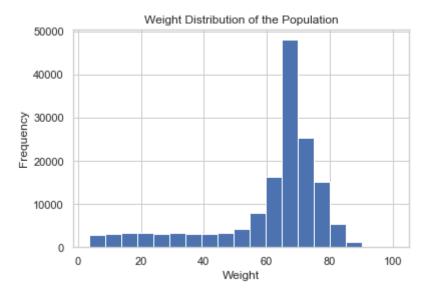
SR_bin_count_weight = int(np.ceil(np.log2(len(weight))) + 1)
print("Sturges Rule number of bins:", SR_bin_count_weight)

## Histogram depicting the weight distribution

plt.hist(weight, bins= SR_bin_count_weight);
plt.ylabel('Frequency')
plt.xlabel('Weight')
plt.title('Weight Distribution of the Population')
```

Sturges Rule number of bins: 19

Out[41]: Text(0.5, 1.0, 'Weight Distribution of the Population')

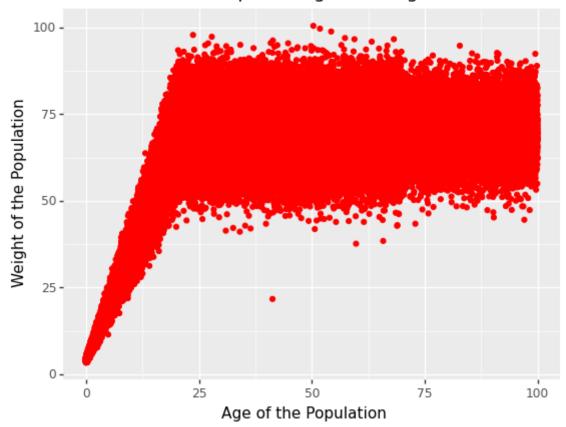


**Response**: The mean weight of the population is 60.88kgs and the max weight is 100 kgs. From the histogram distribution of weight it can be inferred that most of the people in the dataset, weighed in the range of 65-75 kg and then there is a decline in the count. The graph is skewed-left.

general relationship. What is the name of the person? (a points) be sure to explain your process for identifying the person whose values don't follow the usual relationship in the readme. (3 points)

Below is the code to plot the scatterplot between age v/s weight.

# Scatterplot of age v/s weight



Out[42]: <ggplot: (8775671177177)>

Below is the code to identify the outier that does not follow the general relationship.

```
In [43]: outlier_age = data.loc[data['age']>30]
    outlier_weight = outlier_age.loc[data['weight']<25]
    outlier_name = "".join(outlier_weight['name'])
    print("The name of the person who does not follow the general age v/s weigh</pre>
```

The name of the person who does not follow the general age v/s weight relationship is **Anthony Freeman** .

**Response**: From the graph it can be observed that as the age of the person increases the weight also increases. The relationship between both the variables is linear. But, after the age of around 20 years, the relationship between the age and weight is stable.

The person who does not follow the general age v/s weight relationship is Anthony Freeman. Identification of the outlier was made using the scatterplot. In the scatterplot, we can see one point, i.e., between 25-50 years, having a weight less than 25kgs, which is as an outlier. So, to accurately identify the name, I created a data variable *outlier\_age* of individuals older than 30 years. Then, I narrowed down the dataset to include individuals with weight less than 25kgs. From the dataset's dataframe the name of the individual was identified.

### **Exercise 3**

Download historical data for COVID-19 cases by state from The New York Times's GitHub at <a href="https://raw.githubusercontent.com/nytimes/covid-19-data/master/us-states.csv">https://raw.githubusercontent.com/nytimes/covid-19-data/master/us-states.csv</a>). (The full repository including licensing terms is at github.com/nytimes/covid-19-data). As this is an ongoing public health crisis, include in your readme the date you downloaded the data (2 points). Since you are using data from an external source, be sure to credit The New York Times as your data source in your readme as well (2 points).

The data was downloaded on September 13, 2022 at 10:50 AM.

Citation: The New York Times. (2021). Coronavirus (Covid-19) Data in the United States. Retrieved [September 13, 2022], from <a href="https://github.com/nytimes/covid-19-data."/">https://github.com/nytimes/covid-19-data."/</a>

Below is the code to load the datset

```
In [44]: ## Importing the libraries and loading the dataset
    import pandas as pd
    import seaborn as sns
    from datetime import datetime
    import matplotlib.pyplot as plt
    import numpy as np
    import warnings
    warnings.filterwarnings("ignore")

data = pd.read_csv("/Users/mahimakaur/Desktop/us-states.csv")
```

In [45]: data ##To view how to the dataframe looks like

#### Out[45]:

	date	state	fips	cases	deaths
0	2020-01-21	Washington	53	1	0
1	2020-01-22	Washington	53	1	0
2	2020-01-23	Washington	53	1	0
3	2020-01-24	Illinois	17	1	0
4	2020-01-24	Washington	53	1	0
51185	2022-09-12	Virginia	51	2062984	21610
51186	2022-09-12	Washington	53	1796343	14195
51187	2022-09-12	West Virginia	54	592324	7334
51188	2022-09-12	Wisconsin	55	1846100	15138
51189	2022-09-12	Wyoming	56	175290	1884

51190 rows × 5 columns

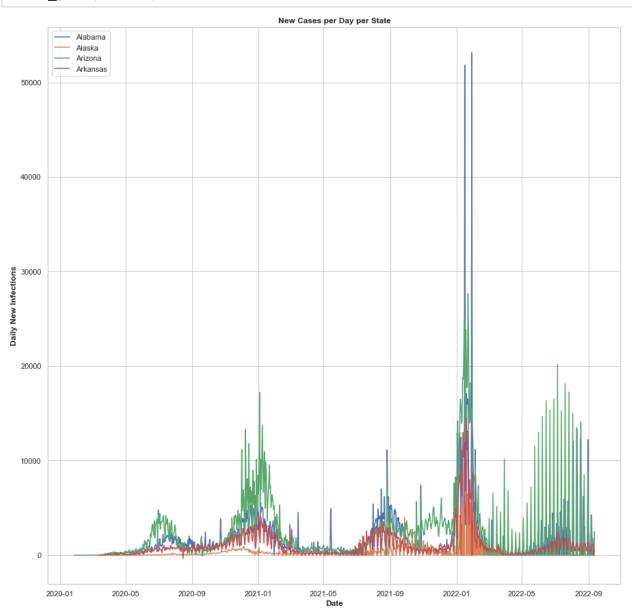
**Part1**: Make a function that takes a list of state names and plots their new cases vs date using overlaid line graphs, one for each selected state. (Note: the data file shows running totals, so you'll have to process it to get new case counts.) Be sure to provide a way to tell which line corresponds to what state (one possibility: using colors and a legend). If your approach has any specific limitations, explain them in your readme. (4 points)

Below is the code for the function that takes a list of state names and plots their new cases vs date using overlaid line graph.

```
In [46]: | ## Changing the dates format
         data['date'] = pd.to_datetime(data['date'], format='%Y-%m-%d')
         ## Creating a Function to plot the graph
         def state_plot(states):
         ## using matplotlib.pyplot and seaborn to draw the overlaid line graphs, on
             plt.figure(figsize = (15,15))
             for state in states:
                 data1 = data[data['state'] == state]
                 data1['newcases'] = data1['cases'].diff() ##Calculating new case c
                 plt.plot(data1['date'], data1['newcases'], label = state)
             plt.legend(loc = 'upper left')
             plt.title('New Cases per Day per State', loc = 'center', weight='bold')
             plt.xlabel('Date', weight='bold')
             plt.ylabel('Daily New Infections', weight='bold')
             plt.show()
             sns.set()
             sns.set_style('whitegrid')
```

Part2: Test the above function and provide examples of it in use. (4 points)

```
In [47]: ## Testing the above function
    states = ['Alabama', 'Alaska', 'Arizona', 'Arkansas']
    state_plot(states)
```



**About the graph**: In the graph, the x-axis indicates the dates, y-axis indicates the daily new cases tested positive for COVID-19 infection of each selected state. Colors of lines indicate different states.

**Part3**: Make a function that takes the name of a state and returns the date of its highest number of new cases. (4 points)

Below is the function that takes the name of a state and returns the date of its highest number of new cases.

```
In [48]: ## Creating a function that takes the name of a state and returns the date

def max_case_date(state):
    data1 = data[data['state'] == state]
    data1['newcases'] = data1['cases'].diff()
    max_case = data1['newcases'].max()
    date = data1[data1['newcases'] == max_case]['date']
    return (date.iloc[0].date())

print(max_case_date('California'))
print(max_case_date('Arizona'))

2022-01-10
2022-01-22
```

**Part4**: Make a function that takes the names of two states and reports which one had its highest number of daily new cases first and how many days separate that one's peak from the other one's peak. (5 points)

Below is the function that takes the names of two states and reports which one had its highest number of daily new cases first and how many days separate that one's peak from the other one's peak.

```
In [49]: he function that takes the names of two states and reports which one had it first and how many days separate that one's peak from the other one's peak el,state2):

max_case_date(state1))

max_case_date(state2))

> date2:
(state2,'had highest number of daily new cases first by', abs((date1 - date 1 < date2:
t(state1,'had highest number of daily new cases first by',abs((date2 - date (State1, "and", State2, "have highest number of cases on the same day.")
```

Part5: Test the above function and provide examples of it in use. (4 points)

Below is the code to test the above function.

```
In [50]: ## Testing the above function with examples.

peak('California', 'Arizona')
peak('New York', 'Arizona')
peak('Connecticut', 'New York')
peak('Alabama', 'Arizona')
```

California had highest number of daily new cases first by 12 days than Ar izona.

New York had highest number of daily new cases first by 14 days than Ariz ona .

New York had highest number of daily new cases first by 2 days than Conne

Arizona had highest number of daily new cases first by 7 days than Alabam a .

## **Excerise 4**

Question: Download the MeSH data desc2022.xml from

https://nlmpubs.nlm.nih.gov/projects/mesh/MESH\_FILES/xmlmesh/

(https://nlmpubs.nlm.nih.gov/projects/mesh/MESH\_FILES/xmlmesh/) (Links to an external site.) (A guide to MeSH XML is available at: https://www.nlm.nih.gov/mesh/xmlmesh.html

(https://www.nlm.nih.gov/mesh/xmlmesh.html) (Links to an external site.)) You'll probably want to look at a snippet of the file to get a sense of how it's written.

```
In [51]: ## importing element tree under the alias of ET
import xml.etree.ElementTree as ET
```

**Part1**: Write Python code that reads the XML and reports: the DescriptorName associated with DescriptorUI D007154 (the text of the name is nested inside a String tag).

```
In [54]: ## the below function reads the XML file and reports the DescriptorName ass

def DescriptorName(DescriptorUI):
    lst = []
    i = -1
    while True:
        i += 1
        try:
        lst.append(root[i][0].text)
    except:
        break
    length = len(lst)
    for i in range(length):
        if(root[i][0].text == DescriptorUI):
            parent_index = i
    return root[parent_index][1].find("String").text
```

```
In [55]: ##Testing the function
DescriptorName("D007154")
```

Out[55]: 'Immune System Diseases'

**Part2**: Write Python code that reads the XML and reports: the DescriptorUI (MeSH Unique ID) associated with DescriptorName "Nervous System Diseases".

```
In [56]: inction reads the XML file and reports the DescriptorUI which starts with DO

I(DescriptorName):

.append(root[i][1].find("String").text)

ak
in(lst)
inge(length):
[i][1].find("String").text == DescriptorName):
ent_index = i
ak
.[parent_index][0].text
```

```
In [57]: ## Testing the function

Descriptor_UI = DescriptorUI("Nervous System Diseases")
print(Descriptor_UI)
```

D009422

**Part3**: Write Python code that reads the XML and reports: The DescriptorNames of items in the MeSH hierarchy that are descendants of both "Nervous System Diseases" and D007154. (That is, each item is a subtype of both, as defined by its TreeNumber(s).)

```
In [58]: he treeNumber associated with the respective DescriptorName or DescriptorUI
        orUI(ui or name) ##Using the previous function
        t == ui or name:
        scendants.iter('TreeNumberList'): #the iter method, which will search for m
        t[0].text
         XML Records'
        he descendants of the given DescriptorNames of items in the MeSH hierarchy
        Names(name1, name2):
        ame1)
         ame2)
        dants.iter('TreeNumberList'): #the iter method, which will search for match
         concept:
          in treeNumber.text:
        umber in concept:
        me2Tree2 in treeNumber.text:
         Descedants Name.append(descendants[1][0].text)
         cedants Name) ## to remove the duplicates
```

# In [59]: ##Testing the function print(descendants\_of\_DescriptorNames('Nervous System Diseases', 'D007154'))

{'AIDS Dementia Complex', 'Multiple Sclerosis, Chronic Progressive', 'Ker nicterus', 'Giant Cell Arteritis', 'Diffuse Cerebral Sclerosis of Schilde r', 'Anti-N-Methyl-D-Aspartate Receptor Encephalitis', 'Lupus Vasculitis, Central Nervous System', 'Lambert-Eaton Myasthenic Syndrome', 'Nervous Sy stem Autoimmune Disease, Experimental', 'Demyelinating Autoimmune Disease s, CNS', 'Multiple Sclerosis', 'Guillain-Barre Syndrome', 'Myasthenia Gra vis', 'Myasthenia Gravis, Autoimmune, Experimental', 'Encephalomyelitis, Autoimmune, Experimental', 'Encephalomyelitis, Acute Disseminated', 'Neur omyelitis Optica', 'Autoimmune Hypophysitis', 'Leukoencephalitis, Acute H emorrhagic', 'Autoimmune Diseases of the Nervous System', 'Ataxia Telangi ectasia', 'Polyradiculoneuropathy, Chronic Inflammatory Demyelinating', 'AIDS Arteritis, Central Nervous System', 'Microscopic Polyangiitis', 'Mi ller Fisher Syndrome', 'Mevalonate Kinase Deficiency', 'Vasculitis, Centr al Nervous System', 'Polyradiculoneuropathy', 'Uveomeningoencephalitic Sy ndrome', 'Neuritis, Autoimmune, Experimental', 'Myasthenia Gravis, Neonat al', 'Multiple Sclerosis, Relapsing-Remitting', 'Myelitis, Transverse', 'Stiff-Person Syndrome', 'POEMS Syndrome'}

Part4: Explain briefly in terms of biology/medicine what the above search has found.

representing a category, and the rest are made up of numbers. One of more tree numbers can describe each MeSH heading to reflect its hierarchy in the tree structure and relationships with other MeSH headings. Every three digits represent a hierarchy in the tree structure. The highest level of the MeSH tree structure consists of 16 broad categories. A MeSH term can be part of one or more hierarchies. To illustrate, we will consider the disease 'POEMS syndrome,' a rare blood disorder that damages your nerves and affects other body parts. That means it is an immune system disorder and a disease affecting the nervous system. Therefore, it is MeSh term under two hierarchies: Nervous System Diseases and Immune System Diseases. When using MeSH terms in Pubmed, it searches for the subject heading and any subject headings underneath that term in the MeSH tree. This process leads to expanding the search results and including all the available resources. For example, if you were searching for Nervous System Diseases, you'll also see information about Neuromuscular Diseases(it will consist of more diseases), Peripheral Nervous System Diseases, Polyneuropathies, etc.

In short, the function gave us all the diseases classified under the hierarchy/category of Nervous System Disease and Immune System Disorder.