

Project: Investigate Healthcare Spending vs Health

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Project: Investigate a Dataset (Replace this with something more specific!)

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Introduction

Investigate a Dataset Project by Z. McLaughlin

In recent news there's been a lot of talk about the fact that the US pays a lot more for healthcare than other developed countries for no better results in healthcare. I decided to use the gapminder data to investigate how the cost of healthcare impacts the health of the people across a set of developed countries: Canada, United States, United Kingdom, France, Germany, Australia, Japan, Norway.

Questions:

- Is there a correlation between healthcare spending and health?
- Is there a correlation between healthcare spending and productivity and mental health in older age?

```
In [1]: # import statements for all of the packages needed

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
% matplotlib inline
```

Data Wrangling

Went through the gapminder data (<https://www.gapminder.org/data/> (<https://www.gapminder.org/data/>)) and decided which datasets to use to investigate the cost of healthcare vs health.

Settled on comparing the cost of healthcare per capital in US dollars vs:

1. Infant mortality (infant health)
2. Working rates (general health)
3. Suicide rates (mental health)

Found the following data files:

Measure Spending: Government spending on healthcare per capita

Sources - WHO

Country by year

Gapminder_healthcare_gov_spending_per_capita.csv

Total spending on healthcare per capita

Sources - WHO

Country by year

Gapminder_healthcare_total_spending_per_capita.csv

Measure Health:

Infant mortality rate (per 1,000 live births)

Sources - various gapminder

Country by year

Gapminder_infant_mortality.csv

Total 15-64 labour to population (%)

Sources - International Labour Organization

Country by year

Gapminder_15_64_labor_to_population_percentage.csv

Total 65+ labour to population (%)

Sources - International Labour Organization

Country by year

Gapminder_65_plus_labor_to_population_percentage.csv

Suicide 0-14 per 100,000

Sources - WHO

Country by year

Gapminder_0_14_suicide.csv

Suicide 15-29 per 100,000

Sources - WHO

Country by year

Gapminder_15_29_suicide.csv

Suicide 30-44 per 100,000

Sources - WHO

Country by year

Gapminder_30_44_suicide.csv

Suicide 45-59 per 100,000

Sources - WHO

Country by year

Gapminder_45_59_suicide.csv

Suicide 60 plus per 100,000

Sources - WHO

Country by year

Gapminder_60_plus_suicide.csv

Setup arrays to hold the countries and data files I wanted to look at and a variable to hold the year of data I wanted to look at.

Did some preliminary reads of the csv files and settled on the year 2004 because that was the last year that consistently had the data I was looking for. Because of the nature of the data, didn't feel comfortable trying to use mean data to fill-in information for future dateds.

```
In [2]: # Setting up data files and countries and years that will be analyzed

countries = ['Canada', 'United States', 'United Kingdom', 'France', 'Germany', 'Australia', 'Japan', 'Norway'] # selection of first world countries
just_one = ['United States']
year = '2004'

# Data files used below
data_files = ['./data/gapminder_healthcare_total_spending_per_capita.csv',
               './data/gapminder_healthcare_gov_spending_per_capita.csv',
               './data/gapminder_15_64_labor_to_population_percentage.csv',
               './data/gapminder_65_plus_labor_to_population_percentage.csv',
               './data/gapminder_infant_mortality.csv',
               './data/gapminder_suicide_0_14.csv',
               './data/gapminder_suicide_15_29.csv',
               './data/gapminder_suicide_30_44.csv',
               './data/gapminder_suicide_60_plus.csv',
               './data/gapminder_suicide_45_59.csv']
```

Loading Data and Data Cleaning

Steps taken to clean the data:

1. Read in each csv file and filtered down to the year 2004 and the desired countries.
2. Joined all the data into one dataframe.
3. Used numpy to add in the year to the dataframe for the data.

Note: Adding the year would make it easier in the future to expand the program to get data on several years and add them to the same dataframe.

```
In [3]: def gather_data(temp_df):  
        '''  
        This function takes a dataset filters down to just one year and a specific set of countries.  
        '''  
  
        data_name = list(temp_df)[0]  
        temp_df.rename(index=str, columns={data_name : "country"},inplace=True)  
        temp_df.rename(index=str, columns={year : data_name},inplace=True)  
        temp_df=temp_df[['country',data_name]]  
        temp_df = temp_df[temp_df['country'].isin(countries)]  
        temp_df.set_index('country',inplace=True)  
  
        return temp_df
```

```
In [4]: def add_year(temp_df,year):  
        '''  
        This function takes a dataset and a year and adds a column with the year values to the dataset using numpy.  
        In the end decided not to play around with different years, but this could be used for generating a larger  
        dataframe that contains more than one year's worth of data.  
        '''  
  
        rows=temp_df.shape[0]  
        year_array=np.repeat(year, rows)  
        #print(year_array)  
        temp_df['year']=year_array  
        #print(temp_df.head())  
        return temp_df
```

In [5]: # Generating the clean dataframe

```
def read_files(data_files):
    '''
    This function takes a setup of data files and consolidates them into
    one dataframe using gather_data and add_year
    '''

    df_names = []
    data_names = []

    # Reading in all the data and joining them together into one datafra
me
for file in data_files:
    name = file[:-4]
    name = name[10:]
    data_names=np.append(data_names, name)
    df_names=np.append(df_names, file)

    x=0
    for name_csv in reversed(df_names):
        reading_df = pd.read_csv(name_csv,encoding='latin1')
        new_reading_df=gather_data(reading_df)
        if x==0:
            old_reading_df = new_reading_df.copy()
        if x!=0:
            new_reading_df = new_reading_df.join(old_reading_df, how='ou
ter')
            old_reading_df = new_reading_df
        x=x+1

    # Adding an additional column that calcuates private expenditures on
    health based on total and gov spending.
    new_column = 'Per capita private expenditure on health at average ex
change rate (US$)'
    test = new_reading_df['Per capita total expenditure on health at ave
rage exchange rate (US$)'] - new_reading_df['Per capita government expen
diture on health at average exchange rate (US$)']

    new_reading_df.insert(2,new_columnn, test)

    # Adds a column to the dataframe with values for the year that the
    data was filtered to.
    new_reading_df = add_year(new_reading_df,year)

    return new_reading_df

cleaned_data=read_files(data_files)

cleaned_data.head(12)
```

Out[5]:

	Per capita total expenditure on health at average exchange rate (US\$)	Per capita government expenditure on health at average exchange rate (US\$)	Per capita private expenditure on health at average exchange rate (US\$)	Total 15- 64 labour to population (%)	Total 65+ labour to population (%)	Infant mortality rate	Suic 0-14 age
country							
Australia	2871.998165	1915.153068	956.845097	74.599998	6.700000	4.9	0.18
Canada	3036.640797	2132.403387	904.237410	78.199997	7.600000	5.2	0.43
France	3630.427195	2861.502206	768.924989	69.500000	1.200000	3.9	0.17
Germany	3527.979257	2706.498338	821.480919	72.900002	2.700000	4.0	0.17
Japan	2912.805483	2352.091729	560.713754	72.500000	19.700001	2.9	0.26
Norway	5436.267867	4542.518377	893.749490	78.500000	13.800000	3.4	0.31
United Kingdom	2940.673101	2392.048021	548.625080	76.000000	6.000000	5.2	0.04
United States	6330.932631	2785.602950	3545.329681	74.800003	14.400000	6.9	0.44

Basic chart of data has been created. See above

Did a few point checks in original csv files to confirm that the chart data matches the original data files.

Saving the cleaned data off to a csv ready for analysis

Checked the resulting data using head() above and then saved off the cleaned data to clean_data.csv

```
In [6]: # Saving the cleaned data to a csv file.

cleaned_data.to_csv('clean_data.csv')
```

```
In [7]: cd=pd.read_csv('clean_data.csv')
cd.head()
```

Out[7]:

	country	Per capita total expenditure on health at average exchange rate (US\$)	Per capita government expenditure on health at average exchange rate (US\$)	Per capita private expenditure on health at average exchange rate (US\$)	Total 15-64 labour to population (%)	Total 65+ labour to population (%)	Infant mortality rate	S
0	Australia	2871.998165	1915.153068	956.845097	74.599998	6.700000	4.9	0.1
1	Canada	3036.640797	2132.403387	904.237410	78.199997	7.600000	5.2	0.4
2	France	3630.427195	2861.502206	768.924989	69.500000	1.200000	3.9	0.1
3	Germany	3527.979257	2706.498338	821.480919	72.900002	2.700000	4.0	0.1
4	Japan	2912.805483	2352.091729	560.713754	72.500000	19.700001	2.9	0.2

Exploratory Data Analysis

Is there a correlation between healthcare spending and health?

Is there a correlation between healthcare spending and productivity and mental health in older age?


```
In [8]: cd['country'].loc[cd['Suicide 30-44 all age adj'].idxmax()]
values=list(cd.columns.values)
values.remove('country')
print("Max Values: " + year)
print('-----')
for item in values:
    print(item)
    print(' ' + cd['country'].loc[cd[item].idxmax()] + ' -- ' + str(cd
[item].max()))
print("")
print("Min Values: " + year)
print('-----')
for item in values:
    print(item)
    print(' ' + cd['country'].loc[cd[item].idxmin()] + ' -- ' + str(cd
[item].min()))
```

Max Values: 2004

Per capita total expenditure on health at average exchange rate (US\$)
United States -- 6330.932631
Per capita government expenditure on health at average exchange rate (US\$)
Norway -- 4542.518377
Per capita private expenditure on health at average exchange rate (US\$)
United States -- 3545.329681
Total 15-64 labour to population (%)
Norway -- 78.5
Total 65+ labour to population (%)
Japan -- 19.70000076
Infant mortality rate
United States -- 6.9
Suicide 0-14 all age adj
United States -- 0.4436988
Suicide 15-29 all age adj
Japan -- 14.95609
Suicide 30-44 all age adj
Japan -- 24.35557
Suicide 60+ all age adj
Japan -- 31.43446
Suicide 45-59 all age adj
Japan -- 36.55532
year
Australia -- 2004

Min Values: 2004

Per capita total expenditure on health at average exchange rate (US\$)
Australia -- 2871.998165
Per capita government expenditure on health at average exchange rate (US\$)
Australia -- 1915.153068
Per capita private expenditure on health at average exchange rate (US\$)
United Kingdom -- 548.62508
Total 15-64 labour to population (%)
France -- 69.5
Total 65+ labour to population (%)
France -- 1.200000048
Infant mortality rate
Japan -- 2.9
Suicide 0-14 all age adj
United Kingdom -- 0.0425276
Suicide 15-29 all age adj
United Kingdom -- 6.338408
Suicide 30-44 all age adj
United Kingdom -- 10.73936
Suicide 60+ all age adj
United Kingdom -- 6.788108
Suicide 45-59 all age adj
United Kingdom -- 10.01023
year
Australia -- 2004

```
In [9]: cd = pd.read_csv('clean_data.csv',encoding='latin1')
cd.head(10)
```

Out[9]:

	country	Per capita total expenditure on health at average exchange rate (US\$)	Per capita government expenditure on health at average exchange rate (US\$)	Per capita private expenditure on health at average exchange rate (US\$)	Total 15- 64 labour to population (%)	Total 65+ labour to population (%)	Infant mortality rate	5 (%
0	Australia	2871.998165	1915.153068	956.845097	74.599998	6.700000	4.9	0.7
1	Canada	3036.640797	2132.403387	904.237410	78.199997	7.600000	5.2	0.4
2	France	3630.427195	2861.502206	768.924989	69.500000	1.200000	3.9	0.7
3	Germany	3527.979257	2706.498338	821.480919	72.900002	2.700000	4.0	0.7
4	Japan	2912.805483	2352.091729	560.713754	72.500000	19.700001	2.9	0.4
5	Norway	5436.267867	4542.518377	893.749490	78.500000	13.800000	3.4	0.3
6	United Kingdom	2940.673101	2392.048021	548.625080	76.000000	6.000000	5.2	0.6
7	United States	6330.932631	2785.602950	3545.329681	74.800003	14.400000	6.9	0.4

```
In [10]: cd.describe()
```

Out[10]:

	Per capita total expenditure on health at average exchange rate (US\$)	Per capita government expenditure on health at average exchange rate (US\$)	Per capita private expenditure on health at average exchange rate (US\$)	Total 15- 64 labour to population (%)	Total 65+ labour to population (%)	Infant mortality rate	Suicid 0-14 a age ac
count	8.000000	8.000000	8.000000	8.000000	8.000000	8.000000	8.00000
mean	3835.965562	2710.977259	1124.988303	74.625000	9.012500	4.550000	0.25539
std	1317.186182	809.001875	989.806981	3.011288	6.363610	1.268295	0.13838
min	2871.998165	1915.153068	548.625080	69.500000	1.200000	2.900000	0.04252
25%	2933.706197	2297.169644	716.872180	72.800001	5.175000	3.775000	0.17516
50%	3282.310027	2549.273179	857.615205	74.700001	7.150000	4.450000	0.22919
75%	4081.887363	2804.577764	917.389332	76.549999	13.950000	5.200000	0.34520
max	6330.932631	4542.518377	3545.329681	78.500000	19.700001	6.900000	0.44369

```
In [11]: cd.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8 entries, 0 to 7
Data columns (total 13 columns):
country
      8 non-null object
Per capita total expenditure on health at average exchange rate (US$)
      8 non-null float64
Per capita government expenditure on health at average exchange rate (U
S$)      8 non-null float64
Per capita private expenditure on health at average exchange rate (US$)
      8 non-null float64
Total 15-64 labour to population (%)
      8 non-null float64
Total 65+ labour to population (%)
      8 non-null float64
Infant mortality rate
      8 non-null float64
Suicide 0-14 all age adj
      8 non-null float64
Suicide 15-29 all age adj
      8 non-null float64
Suicide 30-44 all age adj
      8 non-null float64
Suicide 60+ all age adj
      8 non-null float64
Suicide 45-59 all age adj
      8 non-null float64
year
      8 non-null int64
dtypes: float64(11), int64(1), object(1)
memory usage: 912.0+ bytes
```

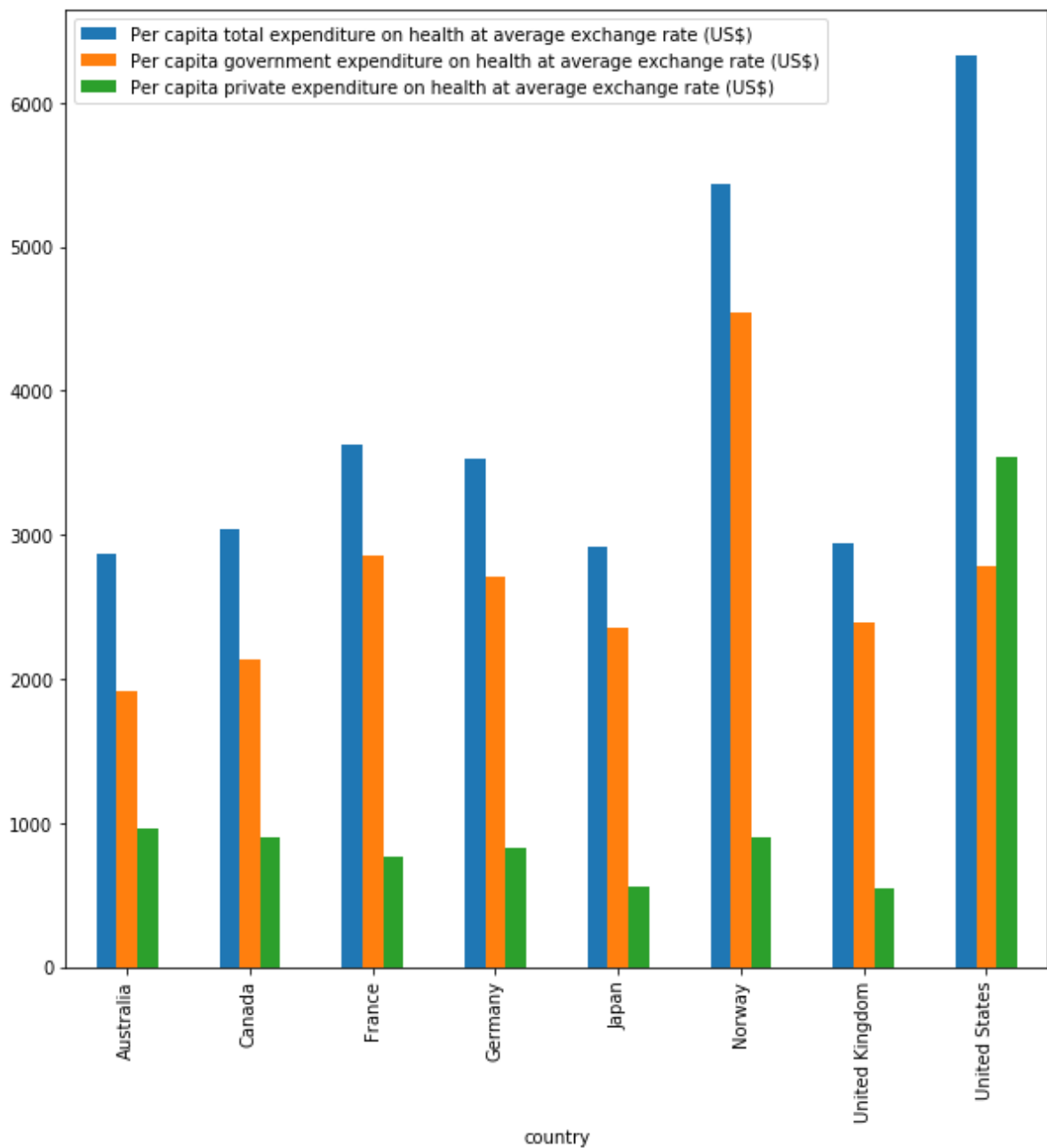
Observations on above data:

- The United States spends the maximum number of dollars on healthcare compared to the other countries.
- Norway spends more government dollars on healthcare than the United States.
- The United states has the highest infant mortality rate.
- Japan had the highest suicide rates.
- France had surprisingly high suicide rates for people 30 and up. -- This last figure in particular seemed questionable which made me wonder if I could trust the data, but found: -- <https://www.thelocal.fr/20130910/why-france-has-such-a-high-suicide-rate> (<https://www.thelocal.fr/20130910/why-france-has-such-a-high-suicide-rate>) -- Which says that France's suicide rate is quite high and -- <https://www.statista.com/statistics/767585/evolution-rate-death-suicide-la-france/> (<https://www.statista.com/statistics/767585/evolution-rate-death-suicide-la-france/>) -- that showed it is decreasing explaining why the numbers in the above article are far lower than those shown in the WHO data.

```
In [12]: # Setting up some arrays to make it easier to define charts
```

```
costs = ['Per capita total expenditure on health at average exchange rate (US$)', 'Per capita government expenditure on health at average exchange rate (US$)', 'Per capita private expenditure on health at average exchange rate (US$)']  
infant_mortality = 'Infant mortality rate'  
working_rates = ['Total 15-64 labour to population (%)', 'Total 65+ labour to population (%)']  
suicide_rates = ['Suicide 0-14 all age adj', 'Suicide 15-29 all age adj', 'Suicide 30-44 all age adj', 'Suicide 45-59 all age adj', 'Suicide 60+ all age adj']
```

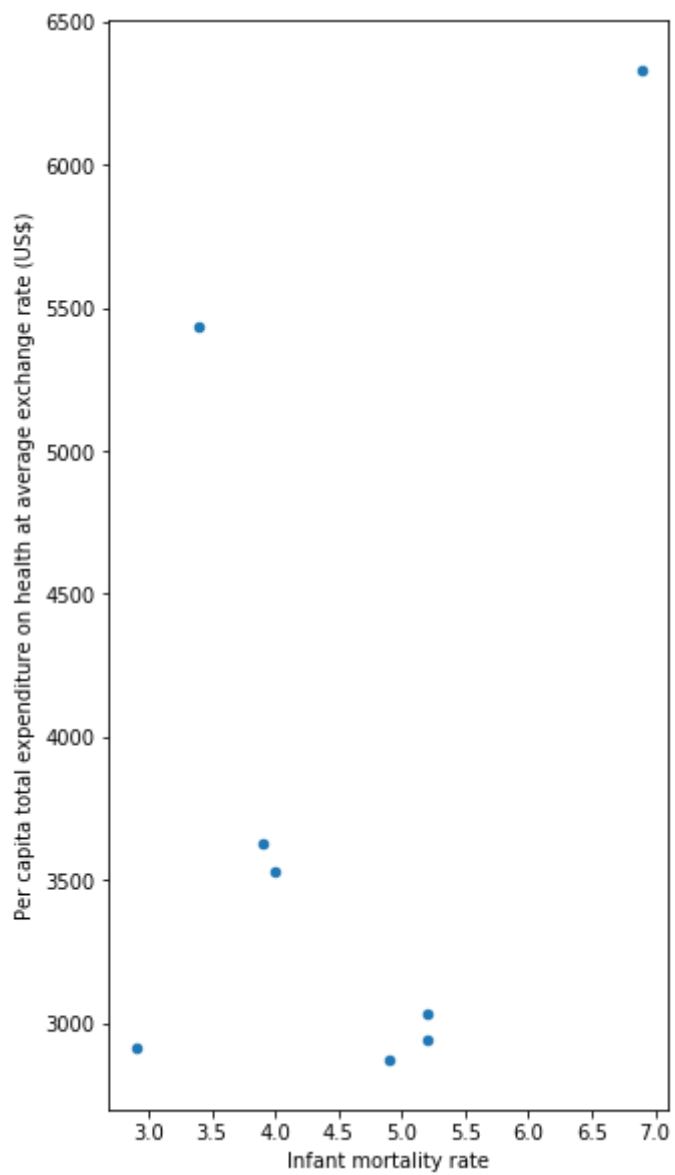
```
In [13]: cd.plot('country', costs, kind='bar', figsize=(10,10));
```

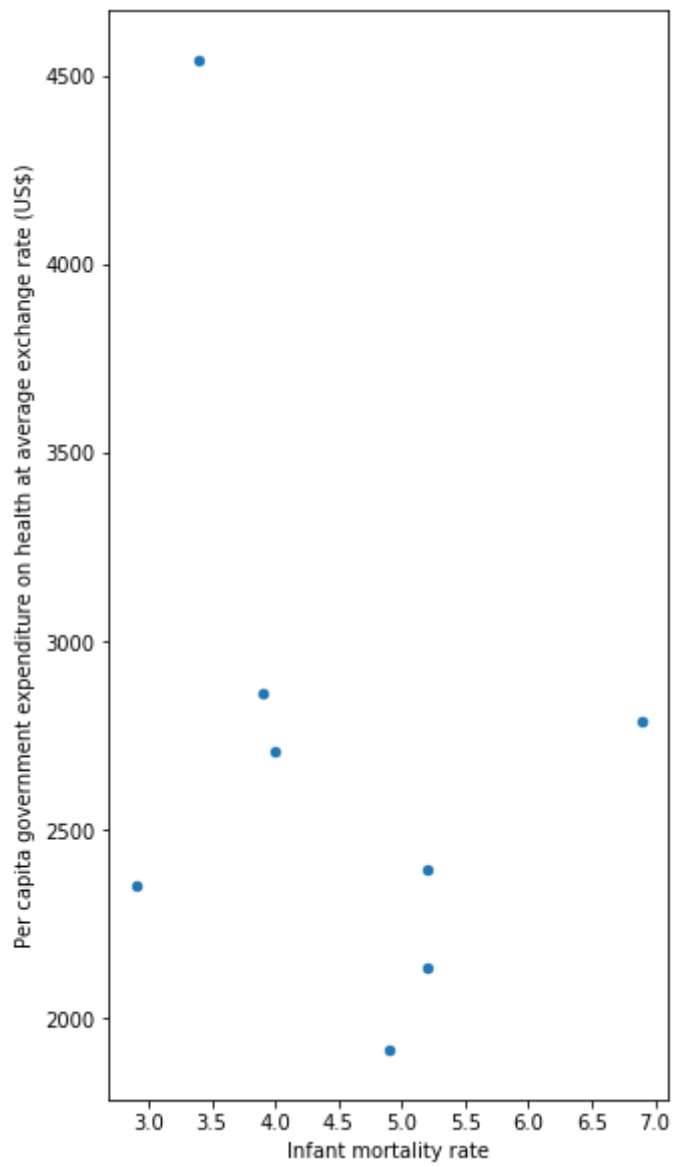


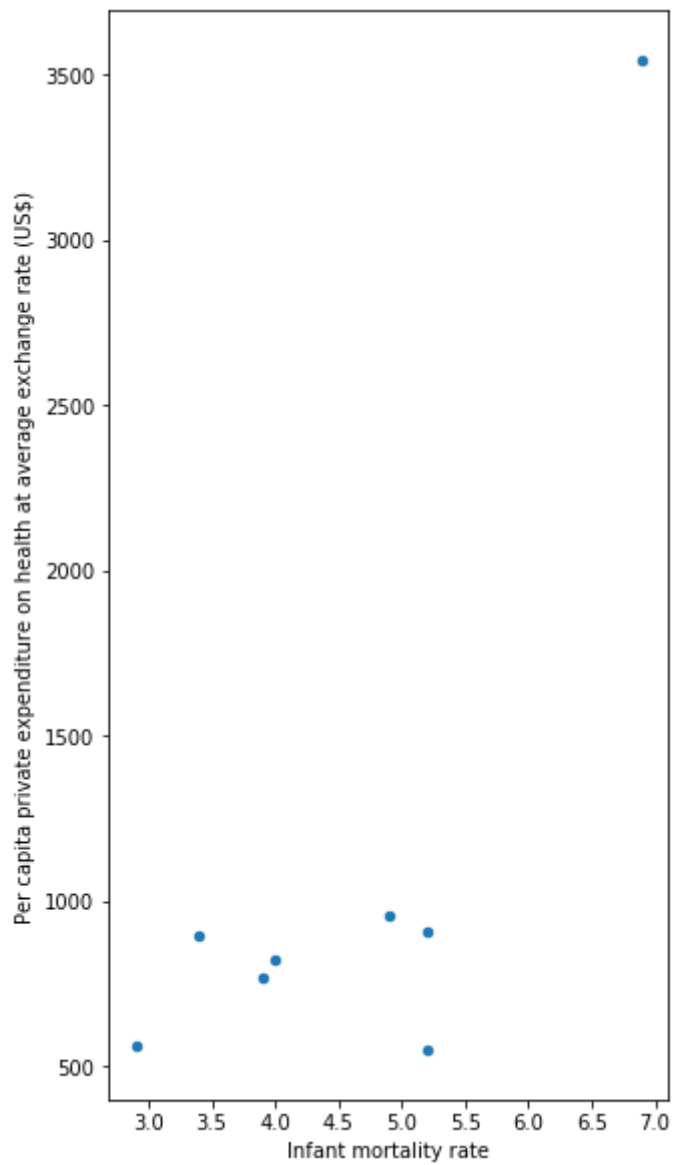
Observations on the above chart:

- The United States spends substantially more on healthcare than other countries except for Norway that also spends a lot on healthcare.
- The government contribution for most countries, except Norway is about the same.
- The US spends far more private dollars on healthcare than the other countries.

```
In [14]: for cost in costs:  
         cd.plot(infant_mortality,cost,kind='scatter',figsize=(5,10));
```





Observations on the above charts on infant mortality:

- For developed countries it doesn't appear that the amount of money spent on healthcare substantially changes the infant mortality rates.
- Maybe there is some other factor impacting infant mortality.

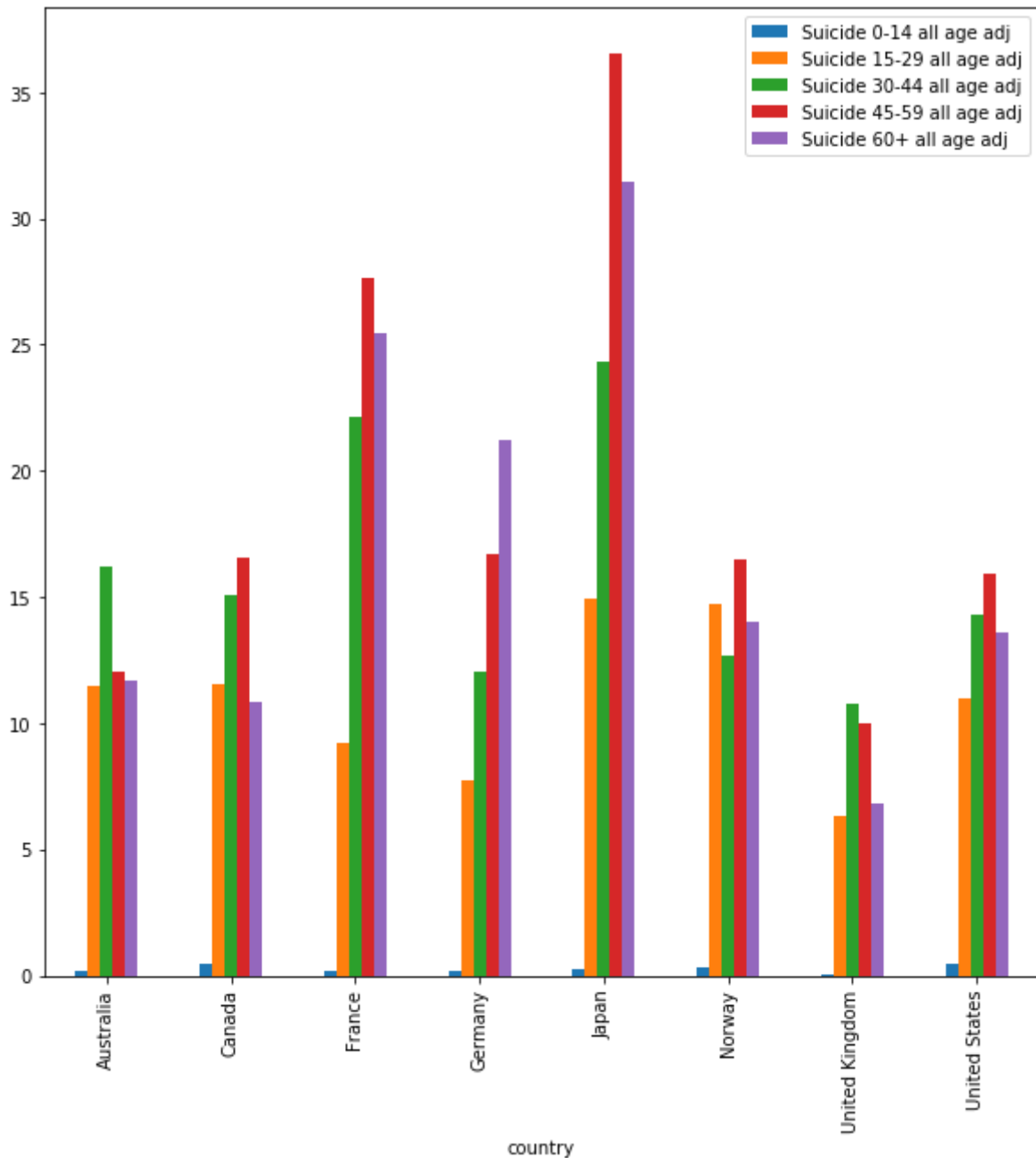
```
In [15]: cd.head(12)
```

Out[15]:

	country	Per capita total expenditure on health at average exchange rate (US\$)	Per capita government expenditure on health at average exchange rate (US\$)	Per capita private expenditure on health at average exchange rate (US\$)	Total 15-64 labour to population (%)	Total 65+ labour to population (%)	Infant mortality rate	\$ (€)
0	Australia	2871.998165	1915.153068	956.845097	74.599998	6.700000	4.9	0.7
1	Canada	3036.640797	2132.403387	904.237410	78.199997	7.600000	5.2	0.4
2	France	3630.427195	2861.502206	768.924989	69.500000	1.200000	3.9	0.7
3	Germany	3527.979257	2706.498338	821.480919	72.900002	2.700000	4.0	0.7
4	Japan	2912.805483	2352.091729	560.713754	72.500000	19.700001	2.9	0.7
5	Norway	5436.267867	4542.518377	893.749490	78.500000	13.800000	3.4	0.7
6	United Kingdom	2940.673101	2392.048021	548.625080	76.000000	6.000000	5.2	0.6
7	United States	6330.932631	2785.602950	3545.329681	74.800003	14.400000	6.9	0.4



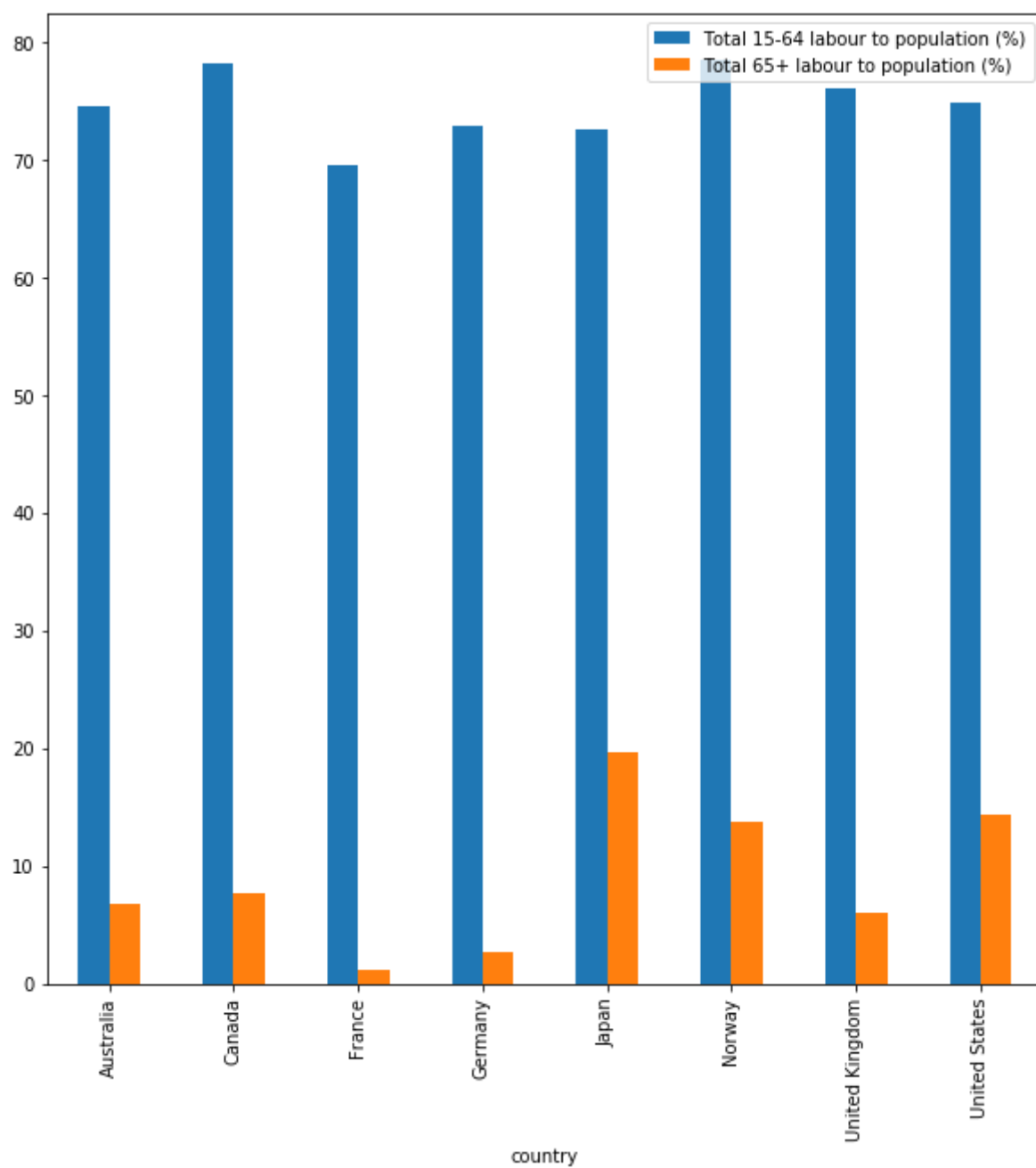
```
In [16]: cd.plot('country',suicide_rates,kind='bar',figsize=(10,10));
```



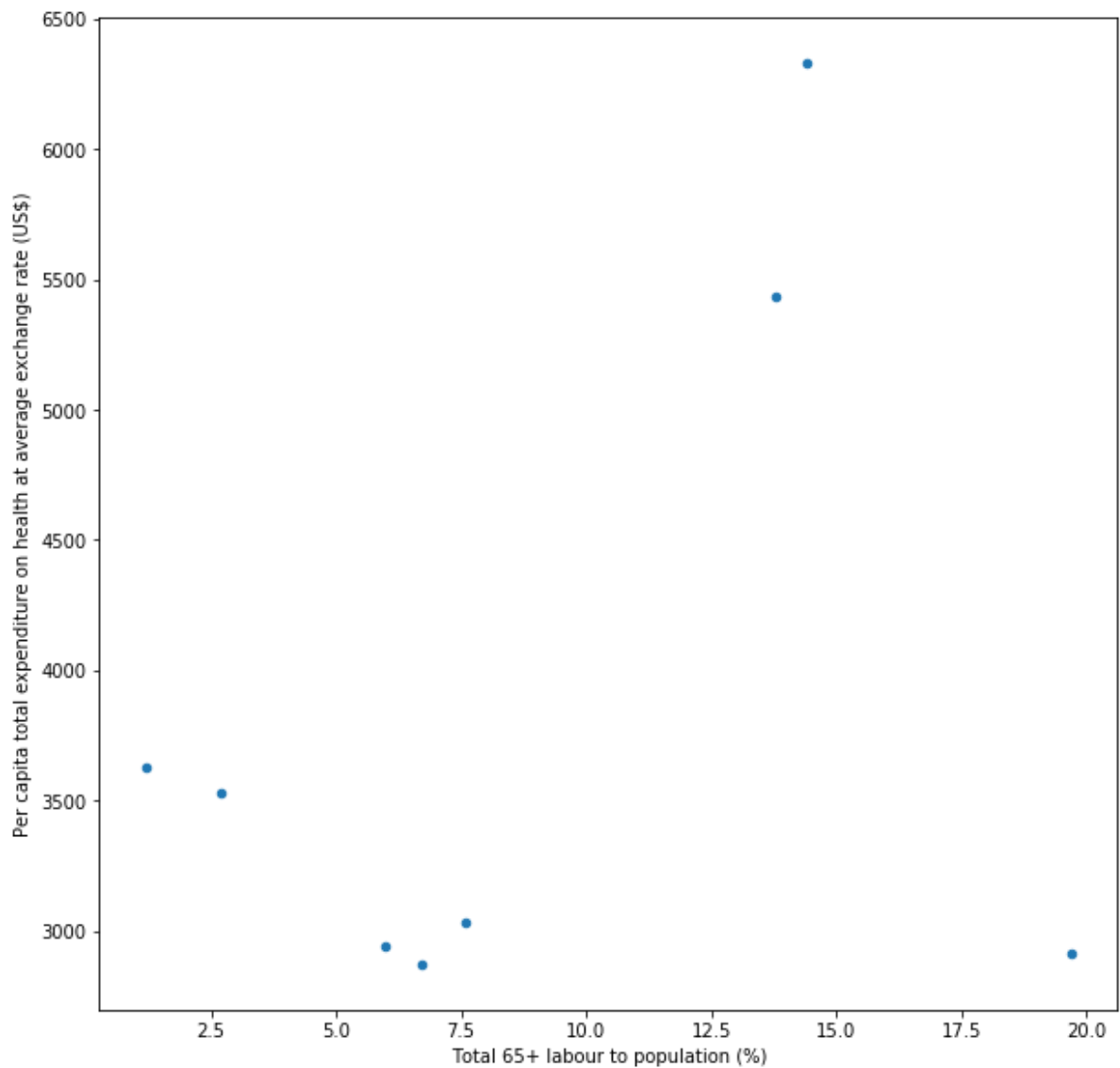
Observations on the above chart:

- Similar to infant mortality there doesn't seem to be a connection between suicide and healthcare spending.
- It's possible that healthcare spending does not impact mental health in a meaningful way and that choosing mental health as a measure may not have been the right decision.

```
In [17]: cd.plot('country',working_rates,kind='bar',figsize=(10,10));
```



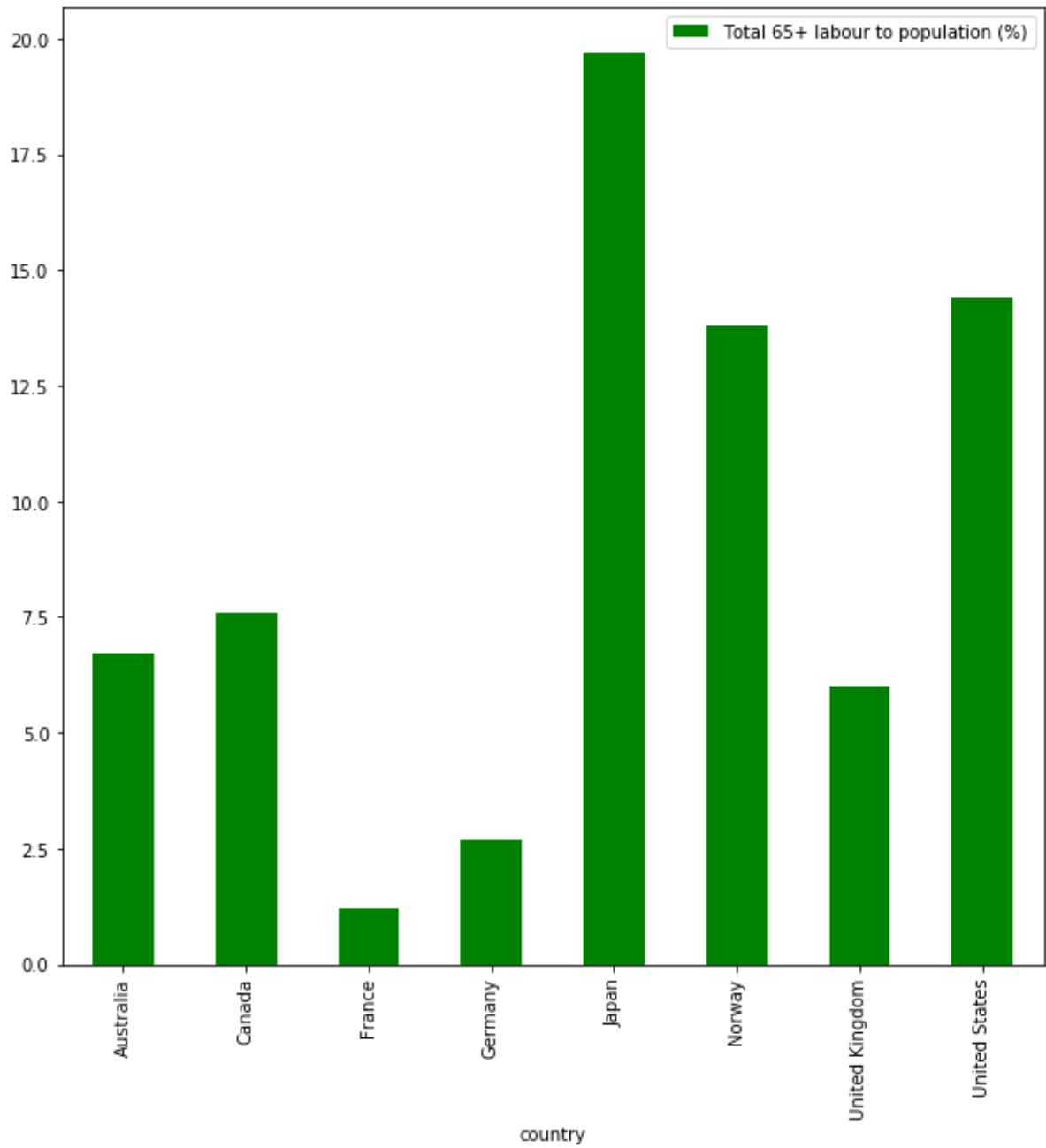
```
In [18]: cd.plot('Total 65+ labour to population (%)', 'Per capita total expenditure on health at average exchange rate (US$)', kind='scatter', figsize=(10, 10));
```



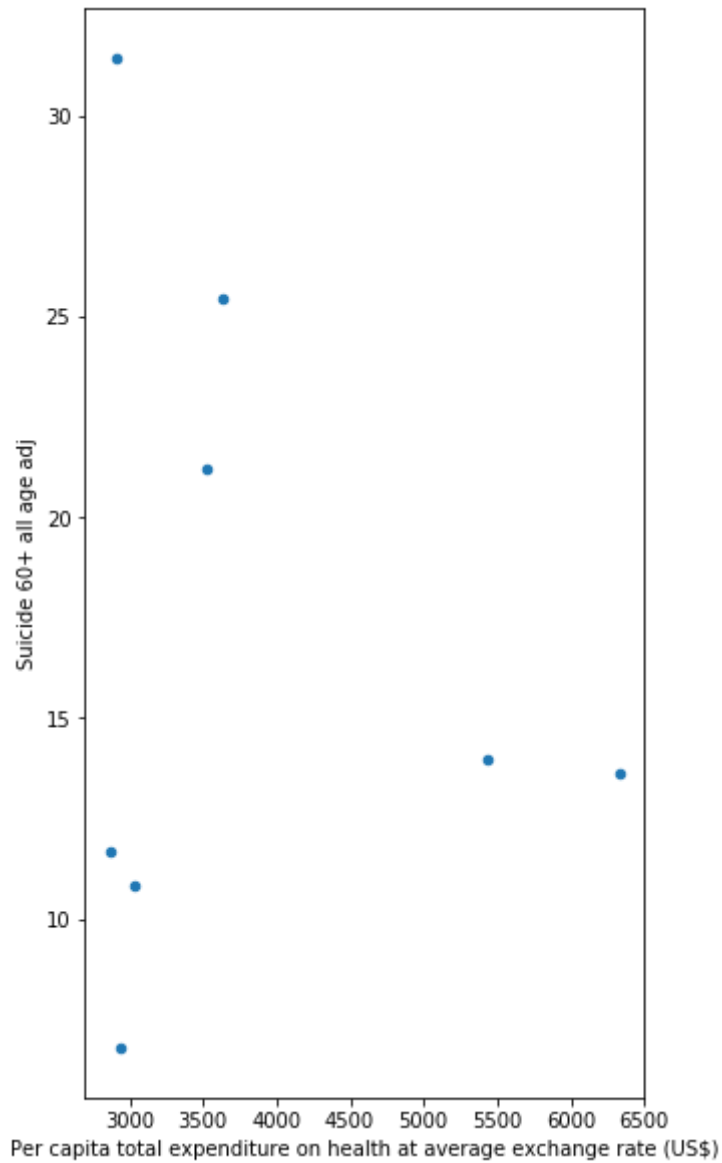
Observations on the above charts:

- It does appear that spending a lot more on healthcare may result in more productivity 65+.

```
In [19]: cd.plot('country',working_rates[1],kind='bar',figsize=(10,10),color='green');
```




```
In [20]: cd.plot(costs[0],suicide_rates[4],kind='scatter',figsize=(5,10));
```



Observations on the above chart:

- Working Rates are very similar 16-64 with no spikes for spending extra money.
- Countries that pay more for healthcare have more folks working 65+

Conclusions

Given that the data from this study was from 2004 did not expect to see that the United States was spending significantly more than other countries, but the data confirmed that this was also the case 14 years ago. The amount of money spent on healthcare did not appear to correlate to the variables that I chose to use to measure health. There was only one set of data that hinted at a correlation: Total spent on healthcare vs Working 65+ - more money spent resulted in more working years. This might mean that people are healthier, but it could also mean that people are living longer or that people are poorer and have to work more years.

There were a couple of interesting points that would be worth more investigation:

1. Once a country is spending a certain amount on healthcare the infant mortality rate does not seem to correlate to the amounts spent, but there is variation. It would be interesting to try and find out where that variation comes from. Example, why does the US have a higher infant mortality rate than other countries.
2. It would be interesting to see how spending on healthcare varies over the years and whether the three variables change overtime.

Overall it would be interesting to get a more specific breakdown of where the money goes to see if there are still differences. Example, are the insurance company fees a factor, are the cost of drugs a factor,

Keeping for reference:

To export the report to the workspace, you should run the code cell below. If it worked correctly, you should get a return code of 0, and you should see the generated .html file in the workspace directory (click on the jupyter icon in the upper left). Alternatively, you can download the html report via the **File > Download as** submenu and then manually upload it to the workspace directory. Once you've done this, you can submit your project by clicking on the "Submit Project" button in the lower right. Congratulations!

```
In [21]: from subprocess import call
        call(['python', '-m', 'nbconvert', 'Investigate_a_Dataset.ipynb'])
```

```
Out[21]: 0
```