

In [2]:

```
import pandas as pd
import numpy as np
```

In [3]:

```
#Load the data with pandas
df = pd.read_csv('../gapminder.csv', low_memory=False)
df.head()
```

Out[3]:

	country	incomeperperson	alcoholconsumption	armedforcesrate	breastcancerper100th	co2emissions	femaleemployrate
0	Afghanistan		.03	.5696534	26.8	75944000	25.60000038146
1	Albania	1914.99655094922	7.29	1.0247361	57.4	223747333.333333	42.09999847412
2	Algeria	2231.99333515006	.69	2.306817	23.5	2932108666.66667	31.70000076293
3	Andorra	21943.3398976022	10.17				
4	Angola	1381.00426770244	5.57	1.4613288	23.1	248358000	69.40000152587

In [4]:

```
# Convert columns from strings to numeric
columns = ['incomeperperson', 'alcoholconsumption', 'armedforcesrate', 'breastcancerper100th',
           'co2emissions', 'femaleemployrate', 'hivrate', 'internetuserate', 'lifeexpectancy',
           'oilperperson', 'polityscore', 'relectricperperson', 'suicideper100th', 'employrate', 'urbanrate']
data = df[columns].apply(pd.to_numeric, errors='coerce')
data['country'] = df['country']
data
```

Out[4]:

	incomeperperson	alcoholconsumption	armedforcesrate	breastcancerper100th	co2emissions	femaleemployrate	hivrate	internetuserate
0	NaN	0.03	0.569653	26.8	7.594400e+07	25.600000	NaN	
1	1914.996551	7.29	1.024736	57.4	2.237473e+08	42.099998	NaN	
2	2231.993335	0.69	2.306817	23.5	2.932109e+09	31.700001	0.1	
3	21943.339898	10.17	NaN	NaN	NaN	NaN	NaN	
4	1381.004268	5.57	1.461329	23.1	2.483580e+08	69.400002	2.0	
...	...	...	...	...	...	...	...	...
208	722.807559	3.91	1.085367	16.2	1.425435e+09	67.599998	0.4	
209	NaN	NaN	5.936085	NaN	1.424133e+07	11.300000	NaN	
210	610.357367	0.20	2.316235	35.1	2.348647e+08	20.299999	NaN	
211	432.226337	3.56	0.341335	13.0	1.320257e+08	53.500000	13.5	
212	320.771890	4.96	1.032785	19.0	5.902197e+08	58.099998	14.3	

213 rows x 16 columns

## Null Values

In [18]:

```
#Remove null values
data= data.replace(0, np.NaN)
data = data.dropna()
```

## Limit Data

In [23]:

```
#Countries with lowest incomes
lowestData = data[data['incomeperperson'] < 749]
```

In [24]:

```
#Countries with the hieghest incomes
highestData = data[data['incomeperperson'] > 9379]
```

## Calculate frequencies

In [26]:

```
print("Values for incomeperperson:")
incomeperperson_freq = pd.concat(dict(counts = data["incomeperperson"].value_counts(sort=False, bins=10, dropna=False), percentages = data["incomeperperson"].value_counts(sort=False, bins=10, dropna=False, normalize=True)), axis=1)
print(incomeperperson_freq)
```

Values for incomeperperson:

	counts	percentages
(518.648, 4499.492]	22	0.392857
(4499.492, 8440.921]	9	0.160714
(8440.921, 12382.35]	3	0.053571
(12382.35, 16323.779]	3	0.053571
(16323.779, 20265.208]	2	0.035714
(20265.208, 24206.637]	1	0.017857
(24206.637, 28148.066]	8	0.142857
(28148.066, 32089.495]	1	0.017857
(32089.495, 36030.924]	3	0.053571
(36030.924, 39972.353]	4	0.071429

In [32]:

```
print("Values for lifeexpectancy:")
lifeexpectancy_freq = pd.concat(dict(counts = data["lifeexpectancy"].value_counts(sort=False, bins=10, dropna=False), percentages = data["lifeexpectancy"].value_counts(sort=False, bins=10, dropna=False, normalize=True)), axis=1)
print(lifeexpectancy_freq)
```

Values for lifeexpectancy:

	counts	percentages
(52.765, 55.857]	1	0.017857
(55.857, 58.916]	0	0.000000
(58.916, 61.976]	0	0.000000
(61.976, 65.036]	0	0.000000
(65.036, 68.095]	3	0.053571
(68.095, 71.155]	8	0.142857
(71.155, 74.215]	12	0.214286
(74.215, 77.275]	7	0.125000
(77.275, 80.334]	9	0.160714
(80.334, 83.394]	16	0.285714

In [34]:

```
print("Values for internetuserate:")
internetuserate_freq = pd.concat(dict(counts = data["internetuserate"].value_counts(sort=False, bins=10, dropna=False), percentages = data["internetuserate"].value_counts(sort=False, bins=10, dropna=False, normalize=True)), axis=1)
print(internetuserate_freq)
```

Values for internetuserate:

values for internetuserate:

	counts	percentages
(3.609, 12.658]	5	0.089286
(12.658, 21.616]	4	0.071429
(21.616, 30.573]	3	0.053571
(30.573, 39.531]	7	0.125000
(39.531, 48.489]	9	0.160714
(48.489, 57.447]	3	0.053571
(57.447, 66.404]	4	0.071429
(66.404, 75.362]	5	0.089286
(75.362, 84.32]	10	0.178571
(84.32, 93.278]	6	0.107143

In [35]:

```
print("Values for breastcancerper100th:")
breastcancerper100th_freq = pd.concat(dict(counts = data["breastcancerper100th"].value_counts(sort=False, bins=10, dropna=False), percentages = data["breastcancerper100th"].value_counts(sort=False, bins=10, dropna=False, normalize=True)), axis=1)
print(breastcancerper100th_freq)
```

Values for breastcancerper100th:

	counts	percentages
(16.514999999999997, 25.05]	11	0.196429
(25.05, 33.5]	7	0.125000
(33.5, 41.95]	7	0.125000
(41.95, 50.4]	9	0.160714
(50.4, 58.85]	4	0.071429
(58.85, 67.3]	1	0.017857
(67.3, 75.75]	5	0.089286
(75.75, 84.2]	3	0.053571
(84.2, 92.65]	8	0.142857
(92.65, 101.1]	1	0.017857

## New Categorical Variables

In [39]:

```
#Categorical variable to label incomeperperson
data['incomelabel'] =pd.cut(data.incomeperperson,4,labels=['low','medium','high','very high'])
income_freq = pd.concat(dict(counts = data["incomelabel"].value_counts(sort=False, dropna=False),
                             percentages = data["incomelabel"].value_counts(sort=False, dropna=False, normalize=True)),
                        axis=1)
```

In [ ]: