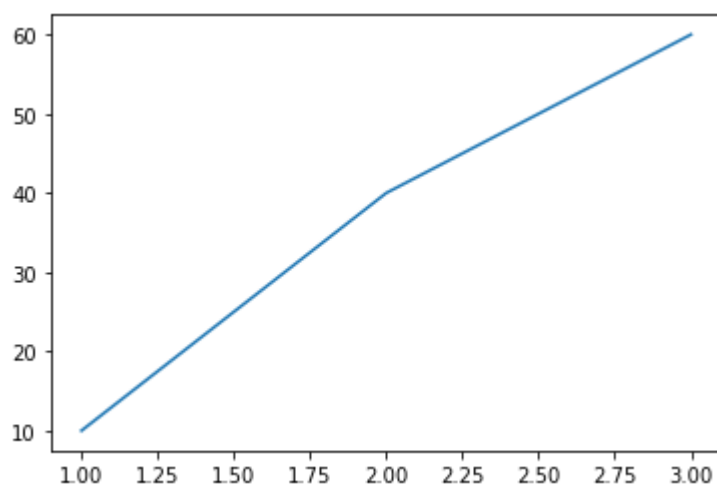
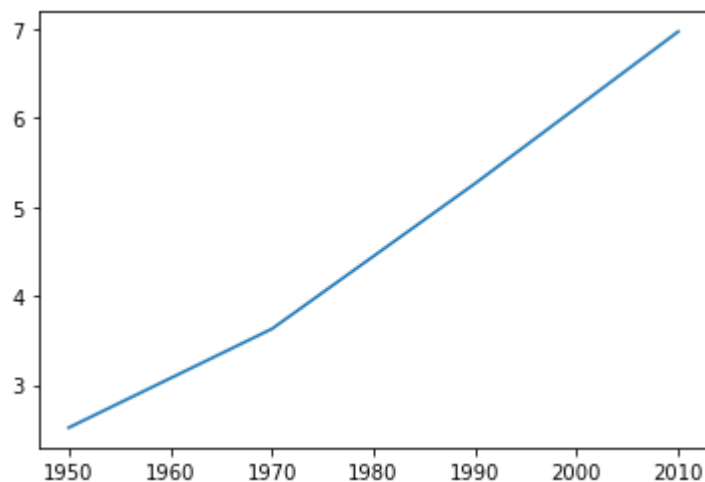


```
In [9]: ▶ import matplotlib.pyplot as plt
%matplotlib inline

x = [1,2,3]
y = [10,40,60]
plt.plot(x,y)
plt.show()
```



```
In [13]: ▶ import matplotlib.pyplot as plt
%matplotlib inline
year = [1950,1970,1990,2010]
pop = [2.519, 3.629, 5.263, 6.97]
plt.plot(year, pop)
plt.show()
```

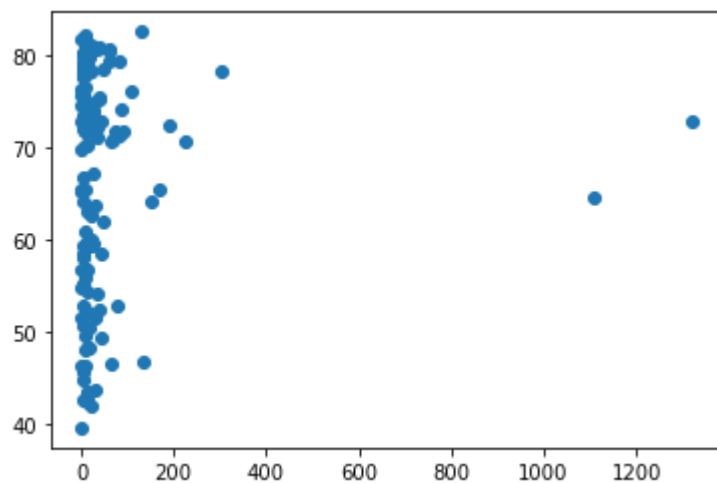


A scatter plot showing the relationship between the number of nodes in the largest component (y-axis) and the total number of nodes in the network (x-axis). The x-axis is on a logarithmic scale, ranging from approximately 500 to 50,000, with major ticks at  $10^3$  and  $10^4$ . The y-axis ranges from 40 to 80. The data points are blue circles. The plot shows a general upward trend, indicating that as the total number of nodes increases, the size of the largest component also tends to increase. There is a noticeable change in the slope of the trend around  $10^4$  nodes.

## ScatterPlot

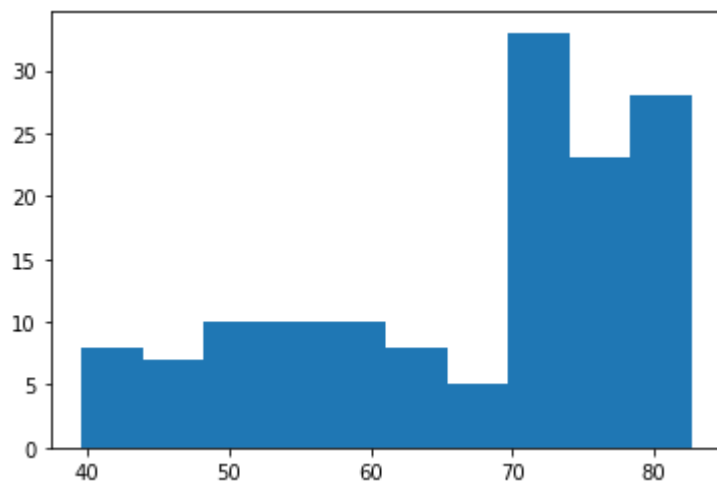
```
In [28]: ▶ # gdp_cap = [974.5803384, 5937.029525999999, 6223.367465, 4797.231267, 127.
life_exp = [43.828, 76.423, 72.301, 42.731, 75.32, 81.235, 79.829, 75.635,
pop = [31.889923, 3.600523, 33.333216, 12.420476, 40.301927, 20.434176, 8.

import matplotlib.pyplot as plt
%matplotlib inline
plt.scatter(pop, life_exp)
#plt.xscale('log')
plt.show()
```



## Histogram of life\_exp

```
In [31]: ▶ plt.hist(life_exp)
plt.show()
```



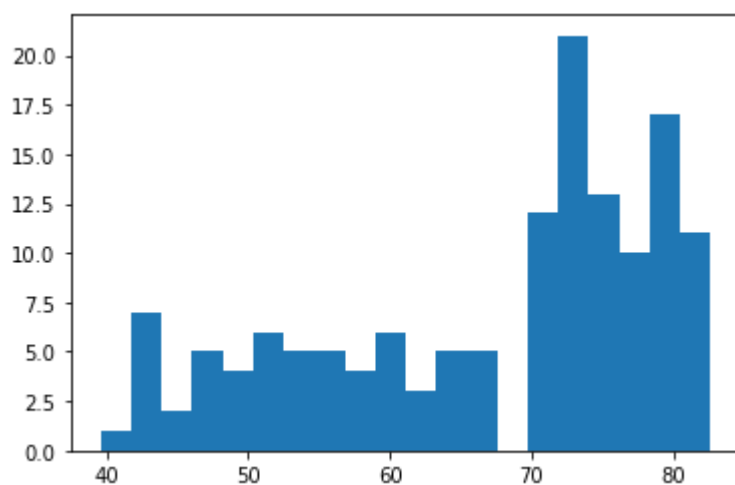
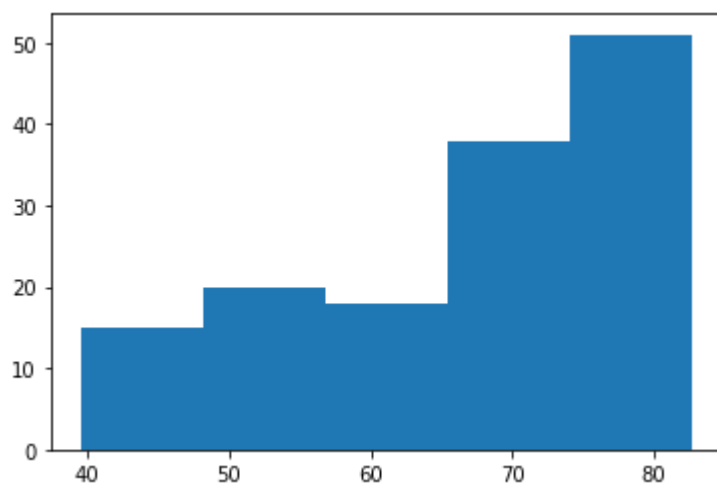
## Ex2

## Histogram of Life exp

In [40]:

```
plt.hist(life_exp, bins = 5)
plt.show()

plt.hist(life_exp, bins = 20)
plt.show()
```



the bins 5 histogram is not easy to understand.

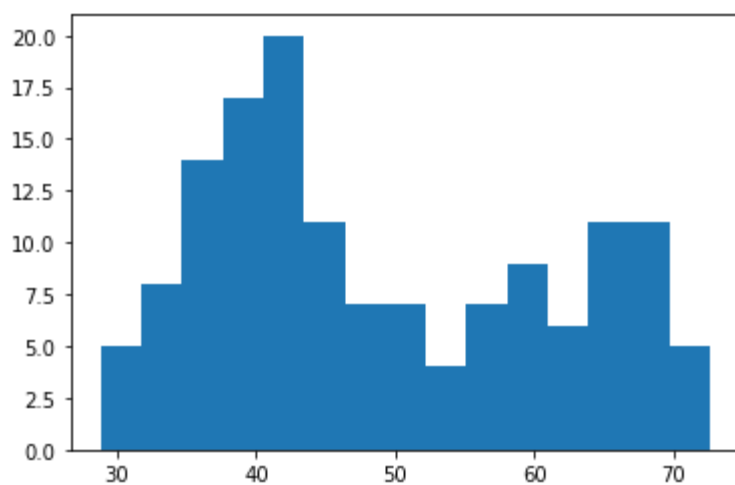
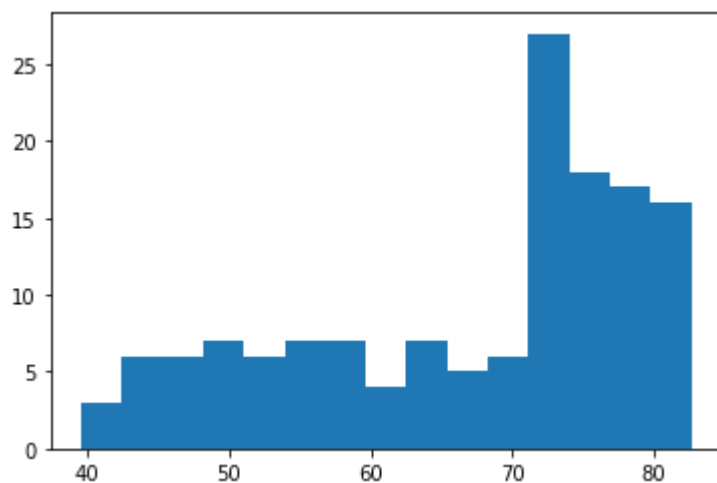
this bins 20 histogram is better because we can clearly see which timeline has more or less than the other.

## Ex3

```
In [45]: ▶ life_exp1950 = [28.8, 55.23, 43.08, 30.02, 62.48, 69.12, 66.8, 50.94, 37.48]

plt.hist(life_exp, bins = 15)
plt.show()

plt.hist(life_exp1950, bins = 15)
plt.show()
```



I observe that they have different frequency even if the bins are the same.

life\_exp 2007 has the highest frequency.

## Ex4

I will use Histogram because the grades on your exam follow a particular distribution

## Ex5

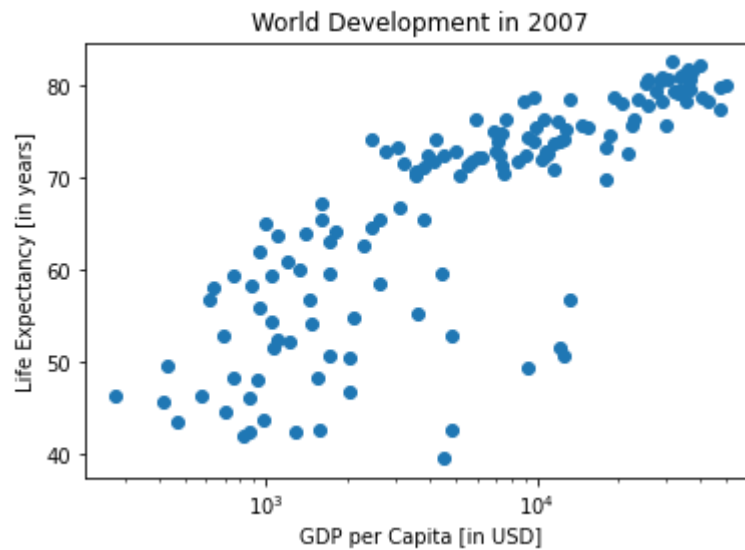
I will use ScatterPlot because longer answers on exam questions lead to higher grades

## Ex6

```
In [51]: ▶ plt.scatter(gdp_cap, life_exp)
plt.xscale('log')

xlab = 'GDP per Capita [in USD]'
ylab = 'Life Expectancy [in years]'
title = 'World Development in 2007'

plt.xlabel(xlab)
plt.ylabel(ylab)
plt.title(title)
plt.show()
```



## Ex7

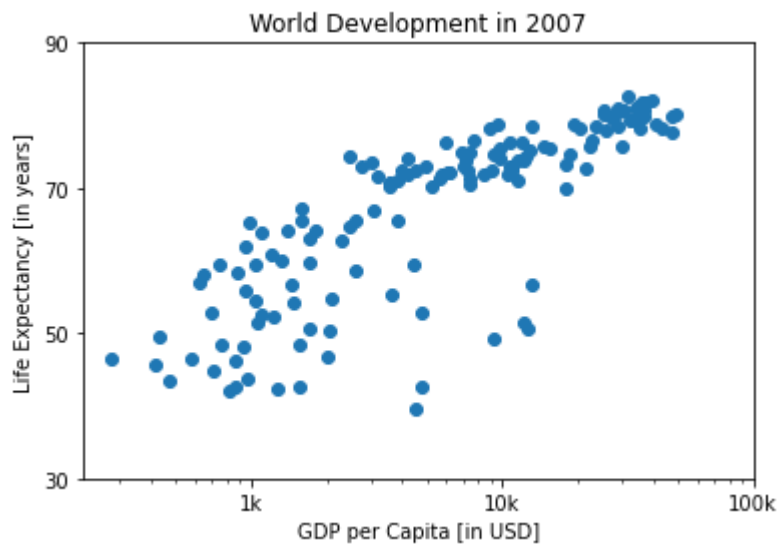
```
In [54]: ▶ # Scatter plot
plt.scatter(gdp_cap, life_exp)
# Previous customizations
plt.xscale('log')
plt.xlabel('GDP per Capita [in USD]')
plt.ylabel('Life Expectancy [in years]')
plt.title('World Development in 2007')

# Definition of tick_val and tick_lab
tick_val = [1000, 10000, 100000]
tick_lab = ['1k', '10k', '100k']

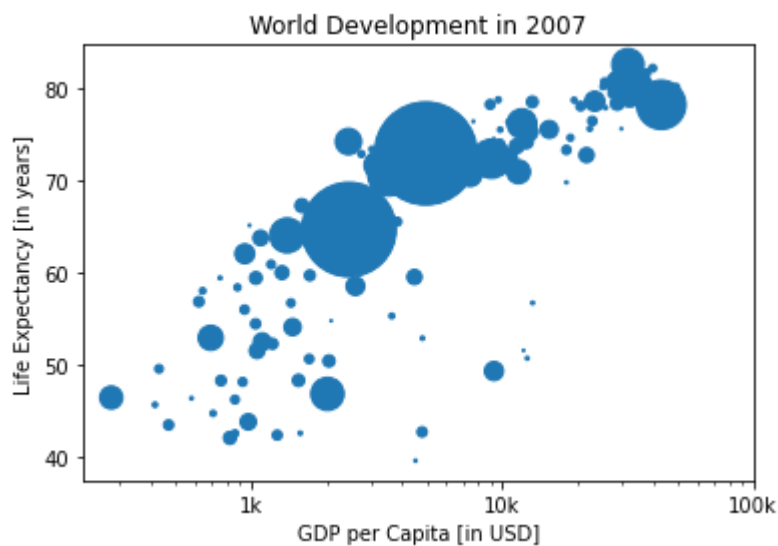
tick_yval = [30, 50, 70, 90]
tick_ylab = ['30', '50', '70', '90']

# Adapt the ticks on the x-axis
plt.xticks(tick_val, tick_lab)
plt.yticks(tick_yval, tick_ylab)

# After customizing, display the plot
plt.show()
```



```
In [59]: # Import numpy as np  
import numpy as np  
  
# Store pop as a numpy array: np_pop  
np_pop = np.array(pop)  
  
# Double np_pop  
np_pop = np_pop * 2  
  
# Update: set s argument to np_pop  
plt.scatter(gdp_cap, life_exp, s = np_pop)  
  
# Previous customizations  
plt.xscale('log')  
plt.xlabel('GDP per Capita [in USD]')  
plt.ylabel('Life Expectancy [in years]')  
plt.title('World Development in 2007')  
plt.xticks([1000, 10000, 100000],['1k', '10k', '100k'])  
  
# Display the plot  
plt.show()
```

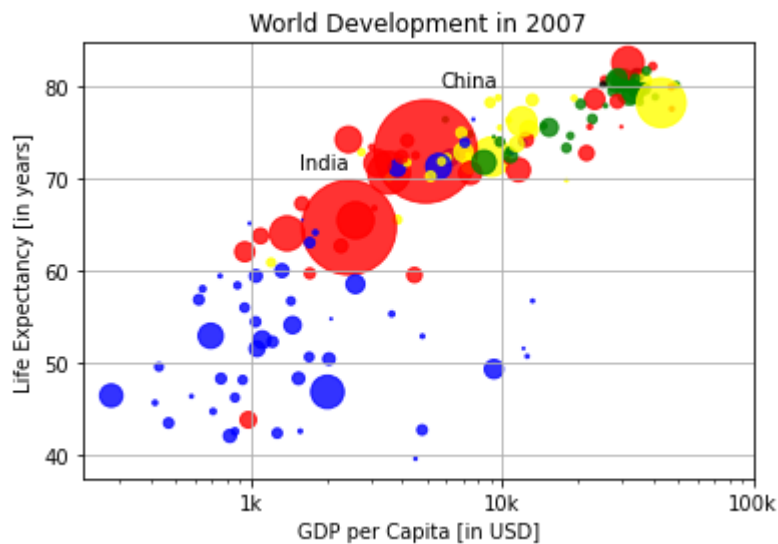




In [68]:

```
col = ['red', 'green', 'blue', 'blue', 'yellow', 'black', 'green', 'red',

# Scatter plot
plt.scatter(x = gdp_cap, y = life_exp, s = np.array(pop) * 2, c = col, alpa
# Previous customizations
plt.xscale('log')
plt.xlabel('GDP per Capita [in USD]')
plt.ylabel('Life Expectancy [in years]')
plt.title('World Development in 2007')
plt.xticks([1000,10000,100000], ['1k', '10k', '100k'])
# Additional customizations
plt.text(1550, 71, 'India')
plt.text(5700, 80, 'China')
# Add grid() call
plt.grid(1)
# Show the plot
plt.show()
```

**Ex8**

```
In [70]: ▶ # Scatter plot
plt.scatter(gdp_cap, life_exp)
# Previous customizations
plt.xscale('log')
plt.xlabel('GDP per Capita [in USD]')
plt.ylabel('Life Expectancy [in years]')
plt.title('World Development in 2007')

# Definition of tick_val and tick_lab
tick_val = [1000, 10000, 100000]
tick_lab = ['1k', '10k', '100k']

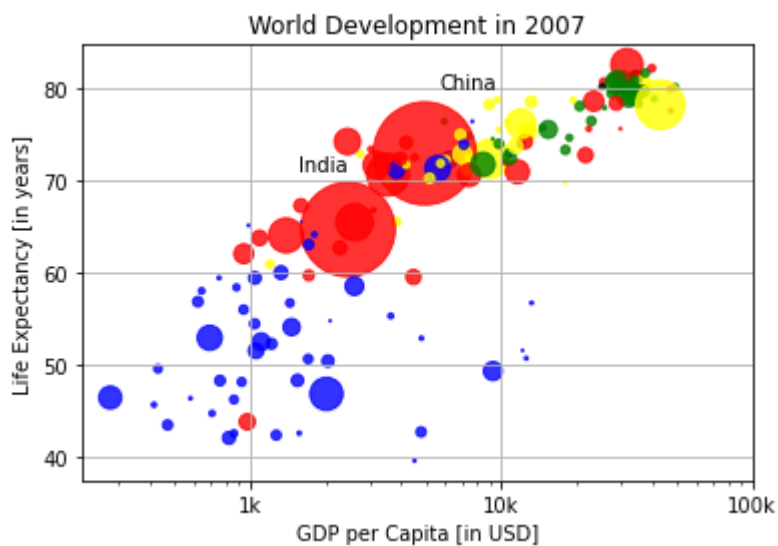
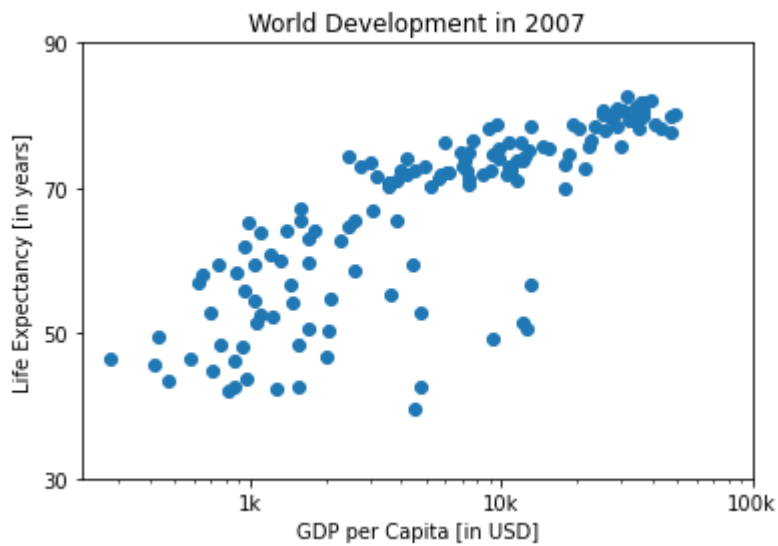
tick_yval = [30, 50, 70, 90]
tick_ylab = ['30', '50', '70', '90']

# Adapt the ticks on the x-axis
plt.xticks(tick_val, tick_lab)
plt.yticks(tick_yval, tick_ylab)

# After customizing, display the plot
plt.show()

col = ['red', 'green', 'blue', 'blue', 'yellow', 'black', 'green', 'red',

# Scatter plot
plt.scatter(x = gdp_cap, y = life_exp, s = np.array(pop) * 2, c = col, alp
# Previous customizations
plt.xscale('log')
plt.xlabel('GDP per Capita [in USD]')
plt.ylabel('Life Expectancy [in years]')
plt.title('World Development in 2007')
plt.xticks([1000, 10000, 100000], ['1k', '10k', '100k'])
# Additional customizations
plt.text(1550, 71, 'India')
plt.text(5700, 80, 'China')
# Add grid() call
plt.grid(1)
# Show the plot
plt.show()
```



the first one is true. (o The countries in blue, corresponding to Africa, have both low life expectancy and a low GDP per capita.)

In [72]: ▶

In [ ]: ▶