WDI Data Analysis

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Loading Packages and Data for Analysis

```
#importing packages
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# loading data
wdi = pd.read_csv("wdi.csv")
```

Running Analysis on the Code

```
# exploratory analysis
print(wdi.describe())
# focusing on adult_literacy_rate
# income_inequality
# gdp_per_capita
```

	inflation_rate	exports_gdp_share	gdp_growth_rate	gdp_per_capita	\
count	169.00	169.00	202.00	203.00	
mean	12.49	46.17	4.37	20345.71	
std	19.68	34.00	6.63	31308.94	
min	-6.69	1.57	-28.76	259.03	
25%	5.52	24.53	2.44	2570.56	
50%	7.97	40.22	4.20	7587.59	
75%	11.67	55.46	6.20	25982.63	

max	171.21	211.	28 63.	44 24	10862.18	
	adult_literacy_ra	te primary sc	hool enrolment ra	te \		
count	49.		114.			
mean	79.		100.87			
std	19.		12.04			
min	27.		64.			
25%	72.		94.			
50%	83.		100.			
75%	95.		105.			
max	100.	00	138.	19		
	education_expendi	ture_gdp_share	measles_immunis	ation_rate	\	
count		105.00		193.00		
mean		4.23		83.85		
std	2.07 16.00					
min	1.03 0.00					
25%	2.90 76.00					
50%	3.89 90.00					
75%		5.16		96.00		
max		16.58		99.00		
	health_expenditur	e_gdp_share i	ncome_inequality	unemployme	ent_rate \	
count	_	20.00	28.00		186.00	
mean		9.04	38.33		7.27	
std		2.70	7.72		5.83	
min		5.10	26.40		0.13	
25%		7.26	32.90		3.50	
50%		8.93	38.10		5.54	
75%		10.63	43.12		9.46	
max		16.57	54.80		37.85	
	life_expectancy	total_populati	on			
count	209.00	217.	00			
mean	72.42	36536447.				
std	7.71	141058261.				
min	53.00	11312.				
25%	66.78	808726.				
50%	73.51	6465097.				
75%	78.47	26069416.				
max	85.38	1417173173.	00			

The GDP per capita variable is very interesting as it shows the massive disparities in wealth

between countries, providing context for a lot of the other variables (min: 259, max: 240862). Adult literacy rate is interesting since there are only 20 entries and a lot of missing data. Likewise, income inequality is missing a significant amount of data.

Plotting the Data

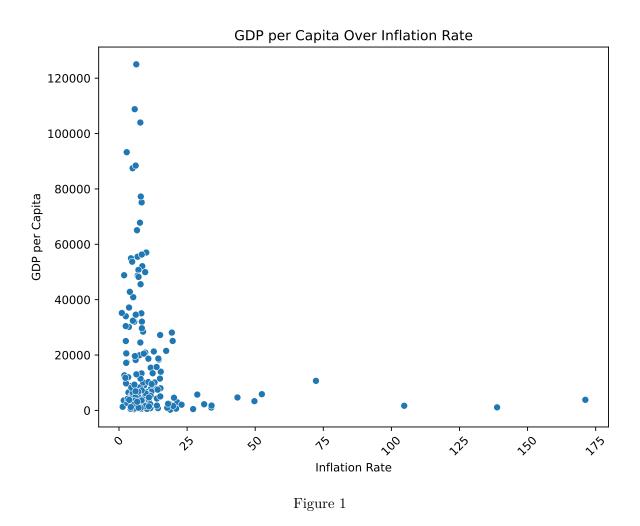


Figure 1: Scatter plot showing the relationship between GDP per capita and total population.

Source: World Development Indicators dataset.

According to Figure 1 countries with high inflation rates generally seem to have low GDP per capita values. Further analysis is needed to derive a correlation, especially regarding the lower inflation countries.

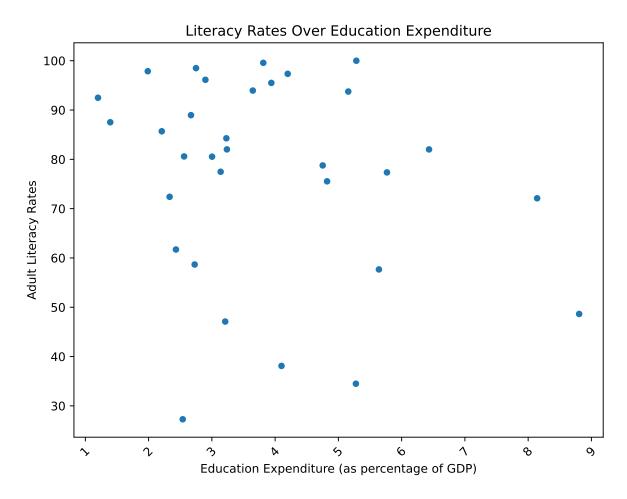


Figure 2

Figure 2: Scatter plot showing the relationship between Literacy Rates and education expenditure.

Source: World Development Indicators dataset.

Figure 2 is very sparse and has very few data points. This is an example of a plot from which conclusions cannot be drawn. We should be very cautious of plots like these as they are missing lots of data!

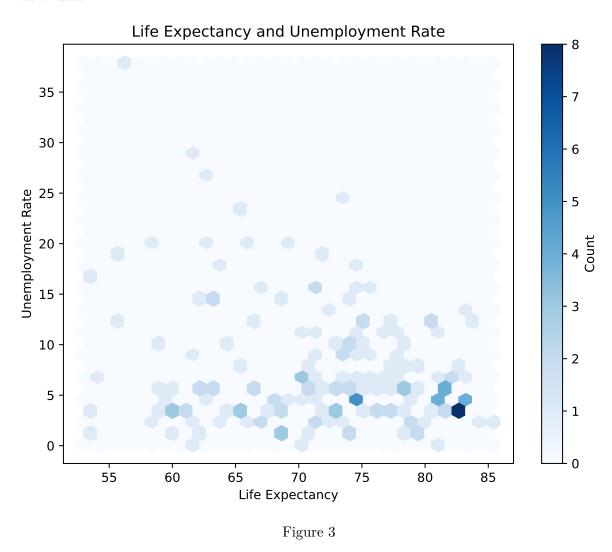


Figure 3: Scatter plot showing the relationship between Life Expectancy and Unemployment Rate.

Source: World Development Indicators dataset.

According to Figure 3 countries with high life expectancies generally seem to have low unemployment rates. This is likely the result of powerful countries having stronger economies and healthcare systems.

Key Statistics

The table below highlights the variables used

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шеап	min	max
20345.71	259.03	240862.18
12.49	-6.69	171.21
79.57	27.28	100.00
4.23	1.03	16.58
72.42	53.00	85.38
7.27	0.13	37.85
	12.49 79.57 4.23 72.42	20345.71 259.03 12.49 -6.69 79.57 27.28 4.23 1.03 72.42 53.00