

GOVERNMENT POLYTECHNIC

NAGAMANGALA

DEPARTMENT OF COMPUTER SCIENCE AND  
ENGINEERING

Vth Semester Diploma

Artificial Intelligence and Machine Learning  
(20CS51)

Assignment: 02

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AIML (20CS51)



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## ASSIGNMENT – WEEK 03

1. Download any two datasets from the internet and perform the following operations.
  - a) Analyze the univariate dataset Ex- Mean, Mode, Median, Range, Std, and Variance and perform Univariate tests for the dataset.
  - b) Analyze the multivariate of the dataset Ex- co-variance, co-relation.
  - c) Visualize the univariate and multivariate with various plots.
  - d) Push the code to your GitHub Repository.

### Important Note:

1. Last Date for Submission: **31-07-2024**.
2. Everyone must perform the above operation using **datasets (Previously used)**.
3. Submit the report to the email [aimlgptn@gmail.com](mailto:aimlgptn@gmail.com)



## 1. Download any two datasets from the internet and perform the following operations.

a) Analyze the univariate dataset Ex- Mean, Mode, Median, Range, Std, and Variance and perform Univariate tests for the dataset.

```
import pandas as pd
```

```
data=pd.read_csv('/content/people (1).csv')
```

```
df=pd.read_csv('/content/Country-data.csv')
```

### Mean:

```
data.mean(numeric_only=True)
```

```
df.mean(numeric_only=True)
```

### Output:

```
Height(cm) 166.15
```

```
Weight(kg) 74.35
```

```
dtype: float64
```

```
child_mort 38.270060
```

```
exports 41.108976
```

```
health 6.815689
```

```
imports 46.890215
```

```
income 17144.688623
```

```
inflation 7.781832
```

```
life_expec 70.555689
```

```
total_fer 2.947964
```

```
gdpp 12964.155689
```

```
dtype: float64
```

### Median:

```
data.median(numeric_only=True)
```

```
df.median(numeric_only=True)
```



Output:

Height(cm) 164.0

Weight(kg) 64.0

dtype: float64

child\_mort 19.30

exports 35.00

health 6.32

imports 43.30

income 9960.00

inflation 5.39

life\_expec 73.10

total\_fer 2.41

gdpp 4660.00

dtype: float64

Mode:

data.mode(numeric\_only=True)

df.mode(numeric\_only=True)

Output:

Height(cm) Weight(kg)

0 160 64

child_mort	exports	health	imports	income	inflation	life_expec	total_fer	gdpp	
0	4.5	23.8	5.07	19.4	1390.0	16.6	70.4	1.57	1310.0
1	NaN	35.0	5.20	23.8	12700.0	NaN	74.5	1.92	NaN
2	NaN	NaN	NaN	26.8	15300.0	NaN	76.3	NaN	NaN
3	NaN	NaN	NaN	28.0	15400.0	NaN	76.4	NaN	NaN
4	NaN	NaN	NaN	28.5	20100.0	NaN	79.5	NaN	NaN
5	NaN	NaN	NaN	37.4	21100.0	NaN	80.4	NaN	NaN
6	NaN	NaN	NaN	42.1	28300.0	NaN	NaN	NaN	NaN
7	NaN	NaN	NaN	43.0	28700.0	NaN	NaN	NaN	NaN
8	NaN	NaN	NaN	43.3	29600.0	NaN	NaN	NaN	NaN
9	NaN	NaN	NaN	46.2	36200.0	NaN	NaN	NaN	NaN
10	NaN	NaN	NaN	49.6	41100.0	NaN	NaN	NaN	NaN
11	NaN	NaN	NaN	51.3	NaN	NaN	NaN	NaN	NaN
12	NaN	NaN	NaN	57.5	NaN	NaN	NaN	NaN	NaN
13	NaN	NaN	NaN	58.9	NaN	NaN	NaN	NaN	NaN
14	NaN	NaN	NaN	62.9	NaN	NaN	NaN	NaN	NaN
15	NaN	NaN							
	NaN	63.6	NaN	NaN	NaN	NaN	NaN		



## Range

```
data.max(numeric_only=True)-data.min(numeric_only=True)
```

```
df.max(numeric_only=True)-df.min(numeric_only=True)
```

Output:

```
Height(cm) 46
```

```
Weight(kg) 91
```

```
dtype: int64
```

```
child_mort 205.4000
```

```
exports 199.8910
```

```
health 16.0900
```

```
imports 173.9341
```

```
income 124391.0000
```

```
inflation 108.2100
```

```
life_expec 50.7000
```

```
total_fer 6.3400
```

```
gdpp 104769.0000
```

```
dtype: float64
```

## Variance:

```
data.var(numeric_only=True)
```

```
df.var(numeric_only=True)
```

Output:

```
Height(cm) 144.450000
```

```
Weight(kg) 883.081579
```

```
dtype: float64
```

```
child_mort 1.626423e+03
```

```
exports 7.514183e+02
```

```
health 7.545116e+00
```

```
imports 5.861042e+02
```

```
income 3.716439e+08
```

```
inflation 1.117398e+02
```

```
life_expec 7.908851e+01
```

```
total_fer 2.291734e+00
```

```
gdpp 3.359414e+08
```

```
dtype: float64
```



Standard deviation:

```
data.std(numeric_only=True)
```

```
df.std(numeric_only=True)
```

Output:

```
Height(cm) 12.018735
```

```
Weight(kg) 29.716689
```

```
dtype: float64
```

```
child_mort 40.328931
```

```
exports 27.412010
```

```
health 2.746837
```

```
imports 24.209589
```

```
income 19278.067698
```

```
inflation 10.570704
```

```
life_expec 8.893172
```

```
total_fer 1.513848
```

```
gdpp 18328.704809
```

```
dtype: float64
```

## Univariate tests

T-test:

```
import pandas as pd
```

```
import scipy.stats as stats
```

```
df = pd.read_csv('/content/Country-data.csv')
```

```
df.head()
```

```
imports1 = df[df['imports'] == 'imports1']['child_mort']
```

```
income2 = df[df['income'] == 'income2']['child_mort']
```

```
t_stat, p_value = stats.ttest_ind(imports1, income2)
```

```
print("T-test Results:")
```

```
print("T-statistic:", t_stat)
```

Output:

```
T-test Results:
```

```
T-statistic: nan
```

Chi-square test:

```
observed_values = pd.crosstab(df['child_mort'], df['life_expec'])
```



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```
chi2,p,dof,expected = stats.chi2_contingency(observed_values)
```

```
print("Chi-square Test Results:")
```

```
print("Chi-square statistic:",chi2)
```

Output:

Chi-square Test Results:

Chi-square statistic: 18031.36111111111

Anova test:

```
healths = [df[df[health] == group][income] for group in df[health].unique()]
```

```
f_stat, p_value = stats.f_oneway(*healths)
```

```
print("ANOVA Test Results:")
```

```
print("F-statistic:", f_stat)
```

Output:

ANOVA Test Results:

F-statistic: 2.7949074494831665

b) Analyze the multivariate of the dataset Ex- co-variance,

co-relation.

Co-variance:

```
data.cov(numeric_only=True)
```

```
df. cov(numeric_only=True)
```

Output:

```
Height(cm)  Weight(kg)
Height(cm)  144.450000  231.260526
Weight(kg)  231.260526  883.081579
```

	child_mort	exports	health	imports	income	inflation	life_expec	total_fer	gdpp
child_mort	1626.422713	-351.651128		-22.199943	-124.201982		-4.076360e+05	122.893627	-318.008262
51.801162	-3.570463e+05								
exports	-351.651128	751.418298	-8.614534	489.350622	2.730946e+05		-31.090078	77.110598	-13.279671
									2.103785e+05
health	-22.199943	-8.614534	7.545116	6.365141	6.861669e+03	-7.415093	5.146808	-0.817828	1.741797e+04
imports	-124.201982	489.350622	6.365141	586.104198	5.712872e+04		-63.208898	11.710284	-5.829066
									5.125005e+04
income	-407635.982270	273094.598023		6861.669071		57128.721588		3.716439e+08	-30110.122438
104916.785517	-14645.727927		3.164430e+08						
inflation	122.893627	-31.090078	-7.415093	-63.208898	-3.011012e+04	111.739781	-22.533965	5.071509	-4.294042e+04



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life_expec	-318.008262	77.110598	5.146808	11.710284	1.049168e+05	-22.533965	79.088507	-10.243585	9.781472e+04
total_fer	51.801162	-13.279671	-0.817828	-5.829066	-1.464573e+04	5.071509	-10.243585	2.291734	-1.262233e+04
gdp	-357046.306154	210378.470377	17417.971217	51250.050217	3.164430e+08	-42940.421636			
	97814.722603	-12622.333657	3.359414e+08						

Co-relation:

data.corr(numeric\_only=True)

data.corr(numeric\_only=True)

Output:

	Height(cm)	Weight(kg)
Height(cm)	1.000000	0.647504
Weight(kg)	0.647504	1.000000

	child_mort	exports	health	imports	income	inflation	life_expec	total_fer	gdp
child_mort	1.000000	-0.318093	-0.200402	-0.127211	-0.524315	0.288276	-0.886676	0.848478	-0.483032
exports	-0.318093	1.000000	-0.114408	0.737381	0.516784	-0.107294	0.316313	-0.320011	0.418725
health	-0.200402	-0.114408	1.000000	0.095717	0.129579	-0.255376	0.210692	-0.196674	0.345966
imports	-0.127211	0.737381	0.095717	1.000000	0.122406	-0.246994	0.054391	-0.159048	0.115498
income	-0.524315	0.516784	0.129579	0.122406	1.000000	-0.147756	0.611962	-0.501840	0.895571
inflation	0.288276	-0.107294	-0.255376	-0.246994	-0.147756	1.000000	-0.239705	0.316921	-0.221631
life_expec	-0.886676	0.316313	0.210692	0.054391	0.611962	-0.239705	1.000000	-0.760875	0.600089
total_fer	0.848478	-0.320011	-0.196674	-0.159048	-0.501840	0.316921	-0.760875	1.000000	-0.454910
gdp	-0.483032	0.418725	0.345966	0.115498	0.895571	-0.221631	0.600089	-0.454910	1.000000

c) Visualize the univariate and multivariate with various plots.

Univariate plots:

import pandas as pd

df=pd.read\_csv('/content/people (1).csv')

import matplotlib.pyplot as plt

import seaborn as sns

plt.figure(figsize=(20, 15))

<Figure size 2000x1500 with 0 Axes>

<Figure size 2000x1500 with 0 Axes>



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### Bar plot:

```
plt.subplot(5, 5, 6)

plt.bar(df['Height(cm)', df['Weight(kg)'])

plt.title("bar plot")

plt.xlabel("Height(cm)")

plt.ylabel("Weight(kg)")

plt.show()
```

### Box plot:

```
plt.subplot(5, 5, 6)

plt.boxplot(df['Height(cm)'])

plt.title("box plot")

plt.xlabel("Height(cm)")

plt.ylabel("Weight(kg)")

plt.show()
```

### Histogram:

```
plt.subplot(5, 5, 6)

plt.hist(df['Height(cm)'], color='green')

plt.title("histogram")

plt.xlabel("Height(cm)")

plt.ylabel("Weight(kg)")

plt.show()
```

### Violin plot:

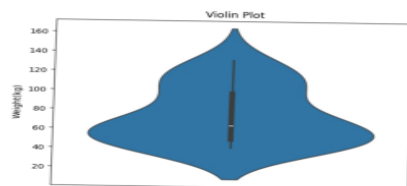
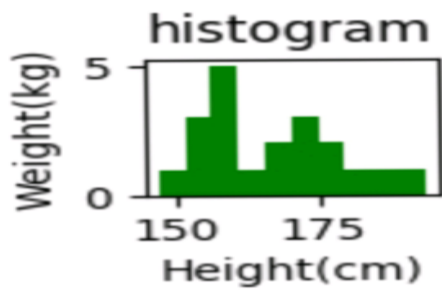
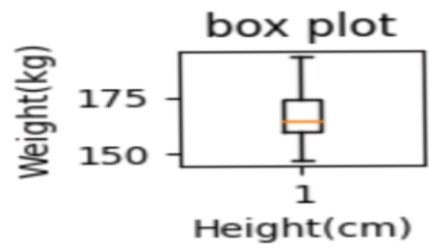
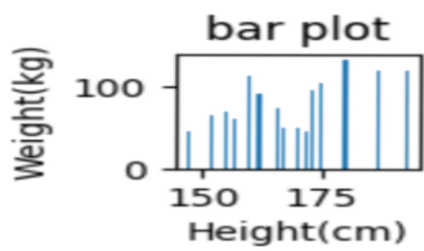
```
sns.violinplot(y=df['Weight(kg)'])

plt.title("Violin Plot")

plt.show()
```

### Output:





```
import pandas as pd

df=pd.read_csv('/content/Country-data.csv')

import matplotlib.pyplot as plt

import seaborn as sns

plt.figure(figsize=(20, 15))
```

<Figure size 2000x1500 with 0 Axes>

<Figure size 2000x1500 with 0 Axes>

**Bar plot:**

```
plt.subplot(2, 3, 2)
```



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```
plt.bar(df['child_mort'], df['health'])

plt.title("bar plot")

plt.xlabel('chiL_mort')

plt.ylabel('health')

plt.show()
```

#### Box plot:

```
plt.subplot(2, 3, 6)

plt.boxplot(df['exports'])

plt.title("box plot")

plt.xlabel('exports')

plt.ylabel('imports')

plt.show()
```

#### Histogram:

```
plt.subplot(2, 3, 6)

plt.hist(df['exports'], color='red')

plt.title("histogram")

plt.xlabel('exports')

plt.ylabel('health')

plt.show()
```

#### Violin plot:

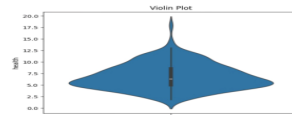
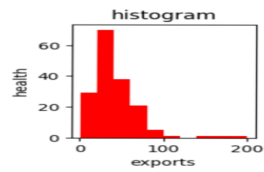
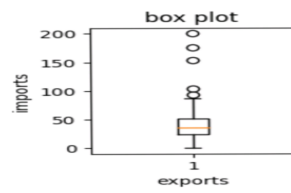
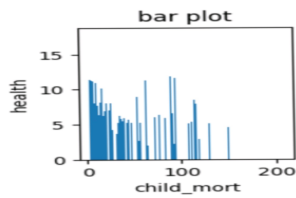
```
sns.violinplot(y=df['export'])

plt.title('Violin Plot')

plt.show()
```

#### Output:





### Multivariate plots:

```
import pandas as pd

df=pd.read_csv("/content/people (1).csv")

import matplotlib.pyplot as plt

import seaborn as sns

plt.figure(figsize=(20, 15))
```

<Figure size 2000x1500 with 0 Axes>

<Figure size 2000x1500 with 0 Axes>

### Box plot:

```
sns.boxplot(x="Height(cm)", y="Weight(kg)", data=df)

plt.show()
```

### Joint plot:

```
sns.jointplot(x="Weight(kg)", y="Height(cm)", data=df)

plt.show()
```

### Bar plot:

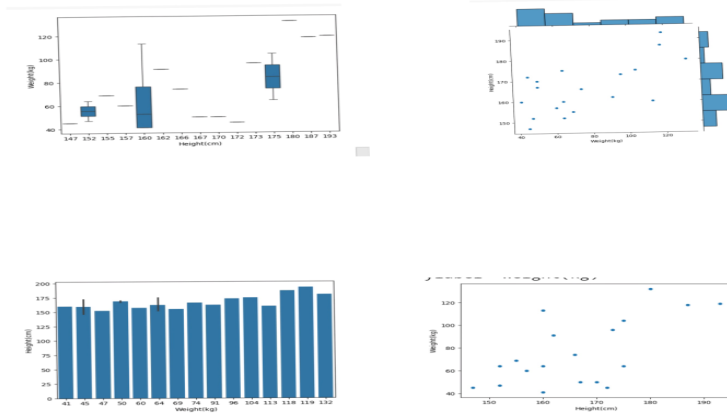
```
sns.barplot(x = 'Weight(kg)', y = 'Height(cm)', data = df)
```



Scatter plot:

```
sns.scatterplot(x='Height(cm)',y='Weight(kg)',data = df)
```

Output:



```
import pandas as pd
```

```
data=pd.read_csv('/content/Country-data.csv')
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
plt.figure(figsize=(20, 15))
```

<Figure size 2000x1500 with 0 Axes>

<Figure size 2000x1500 with 0 Axes>

Box plot:

```
sns.boxplot(x="imports", y="health",data=data)
```

```
plt.show()
```

Joint plot:

```
sns.jointplot(x="child_mort", y="imports", data=data)
```

```
plt.show()
```

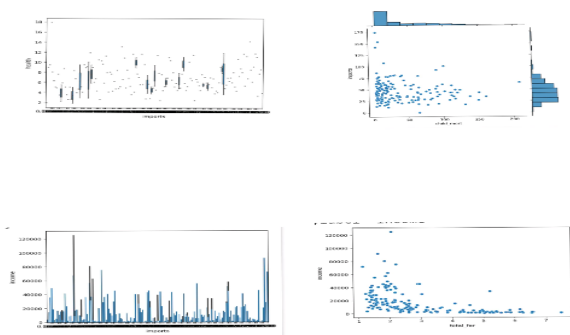
Bar plot:

```
sns.barplot(x='imports', y='income', data = data)
```

Scatter plot:

```
sns.scatterplot(x='total_fer', y='income', data = data)
```

Output:



d) Push the code to your GitHub Repository.

e)perform any probability calculation

```
import pandas as pd
```

```
file_path = ('/content/people (1).csv')
```

```
if 'Height(cm)' in data.columns:
```

```
people_above_30 = data[data['Height(cm)'] > 30]

probability_above_30 = len(people_above_30) / len(data)

print(f"Probability of a person being older than 30: {probability_above_30:.4f}")

else:

    print("The 'Height(cm)' column is not found in the dataset.")
```

#### Output:

Probability of a person being older than 30: 1.0000

2)

```
import pandas as pd

file_path = '/content/Country-data.csv'

if 'child_mort' in data.columns:

    Country_above_30 = data[data['child_mort'] > 30]

    probability_above_30 = len(Country_above_30) / len(data)

    print(f"Probability of a person being older than 30: {probability_above_30:.4f}")

else:

    print("The 'child_mort' column is not found in the dataset.")
```

#### Outputs :

The 'child\_mort' column is not found in the dataset.

