

UNIVERSITY OF TEXAS AT EL PASO (UTEP)

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STAT 5329: Homework 3

```
In [1]: pip install NumPy
```

Requirement already satisfied: NumPy in c:\users\thomo\anaconda3\lib\site-packages (1.20.3)

Note: you may need to restart the kernel to use updated packages.

```
In [2]: import pandas as pd
import numpy as np
import statistics as st
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [14]: df=pd.read_csv("diabetes.csv",encoding = "Latin-1")
df
```

```
Out[14]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age
0	6	148	72	35	0	33.6	0.627	50
1	1	85	66	29	0	26.6	0.351	31
2	8	183	64	0	0	23.3	0.672	32
3	1	89	66	23	94	28.1	0.167	21
4	0	137	40	35	168	43.1	2.288	33
...
763	10	101	76	48	180	32.9	0.171	63
764	2	122	70	27	0	36.8	0.340	27
765	5	121	72	23	112	26.2	0.245	30
766	1	126	60	0	0	30.1	0.349	47
767	1	93	70	31	0	30.4	0.315	23

768 rows × 9 columns



The diabetes data named df has 9 variables and 768 observations.

Question(1) Calculate the mean, median,

standard deviation, IQR, and range for each variable

```
In [4]: #mean of each variable  
df.mean()
```

```
Out[4]: Pregnancies      3.845052  
Glucose      120.894531  
BloodPressure  69.105469  
SkinThickness  20.536458  
Insulin      79.799479  
BMI          31.992578  
DiabetesPedigreeFunction  0.471876  
Age          33.240885  
Outcome      0.348958  
dtype: float64
```

The mean for each variable in the dataset can be seen above. From the Output above, Glucose has the highest mean which means 768 sample of female patients of Pima Indian heritage has an average of 120.89 Glucose levels in their blood followed by an average of 79.79 insulin spike

```
In [5]: #median of each variable  
df.median()
```

```
Out[5]: Pregnancies      3.0000  
Glucose      117.0000  
BloodPressure  72.0000  
SkinThickness  23.0000  
Insulin      30.5000  
BMI          32.0000  
DiabetesPedigreeFunction  0.3725  
Age          29.0000  
Outcome      0.0000  
dtype: float64
```

The median for each variable in the dataset can be seen above. The variable Glucose has the highest median of 117.

```
In [6]: #Standard deviation of each variable in the data set  
df.std()
```

```
Out[6]: Pregnancies      3.369578  
Glucose      31.972618  
BloodPressure  19.355807  
SkinThickness  15.952218  
Insulin      115.244002  
BMI          7.884160  
DiabetesPedigreeFunction  0.331329  
Age          11.760232  
Outcome      0.476951  
dtype: float64
```

The standard deviation for each variable in the dataset can be seen above. Here, Insulin has the highest variability of 115.2 followed by Glucose.

```
In [7]: #IQR of the variables
        from scipy.stats import iqr
        iqr(df)
```

```
Out[7]: 60.53775
```

The Interquartile Range (IQR) for each variable in the dataset can be seen above

```
In [8]: # Range of each variable
        df.max() - df.min()
```

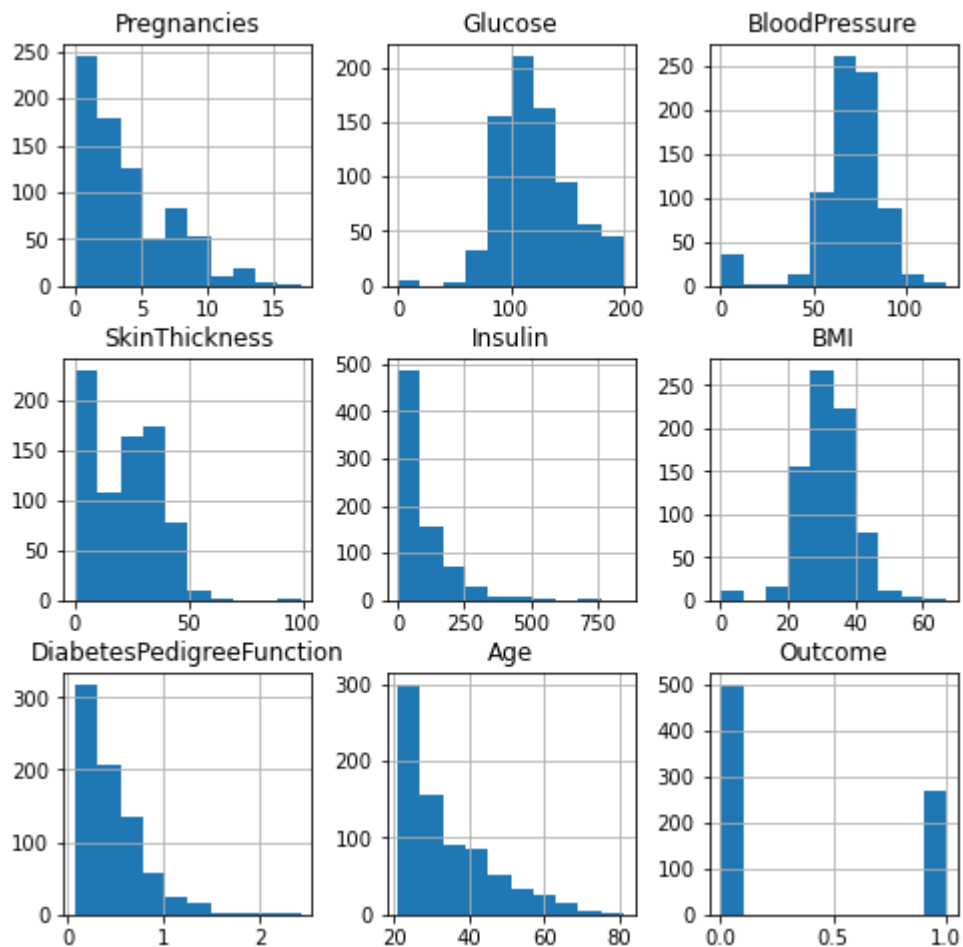
```
Out[8]: Pregnancies      17.000
        Glucose          199.000
        BloodPressure    122.000
        SkinThickness     99.000
        Insulin           846.000
        BMI               67.100
        DiabetesPedigreeFunction  2.342
        Age              60.000
        Outcome           1.000
        dtype: float64
```

The range for each variable in the dataset can be seen above

Question(2) For each variable, construct a histogram and comment on the shape of the distribution. Identify the variables that have similar histograms in terms of shape

```
In [9]: fig = plt.figure(figsize = (8,8))
        ax = fig.gca()
        df.hist(ax=ax)
        plt.show()
```

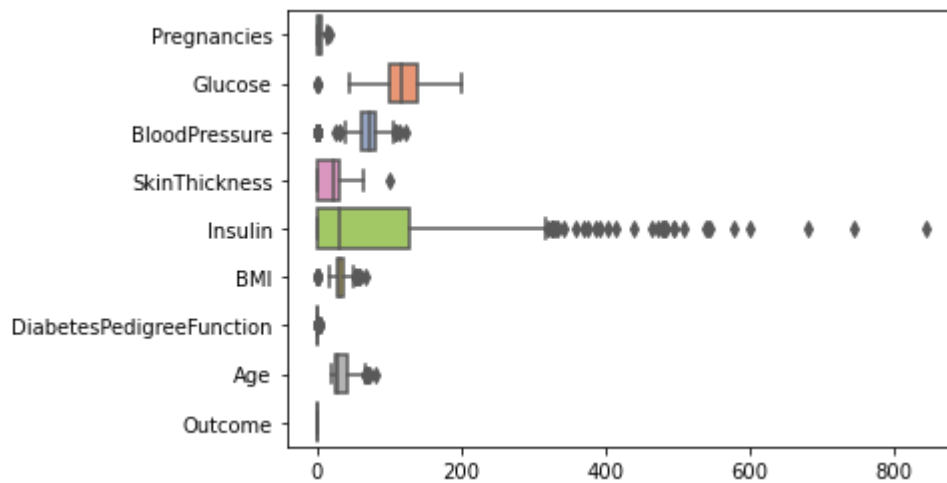
```
C:\Users\thomo\AppData\Local\Temp\ipykernel_110820\3175089013.py:3: UserWarning: To output multiple subplots, the figure containing the passed axes is being cleared
    df.hist(ax=ax)
```



It can be observed from the histograms above that Glucose, Blood Pressure and BMI are fairly normally distributed whiles Pregnancies, Skin Thickness, Insulin, Diabetes Pedigree Function and Age are skewed. We can also observe that Pregnancy, Age, Diabetes Pedigree Function and Insulin looks similar in terms of the shape of their histograms. Also, Glucose, BMI and Blood Pressure looks similar in terms of the shape of their histograms.

Question(3) Construct box-and-whisker plots for all variables

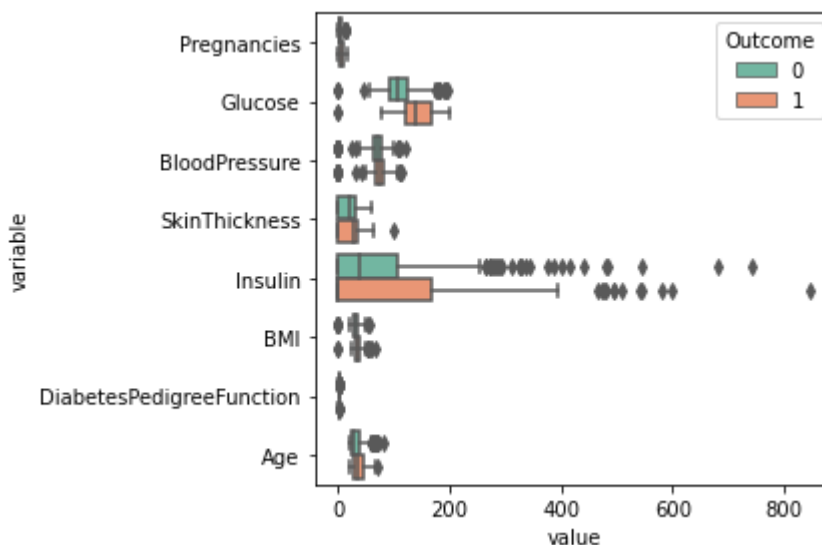
```
In [10]: ax = sns.boxplot(data=df, orient="h", palette="Set2")
```



We can observe from the box-and-whisker plot above that the variable Insulin has outliers.

Question(4) Construct side-by-side boxplots for all the variables for the Outcome=1 and Outcome=0 groups. Identify the variables that have boxplots with different shapes between the two groups? What does it mean?

```
In [11]: df_long = df.melt(id_vars=['Outcome'])
ax = sns.boxplot(data=df_long, x="value", y="variable", orient="h", palette="Set2", hue
plt.tight_layout()
plt.show()
```



It can be observed from the above boxplot that the variables; Insulin and Glucose have different shapes between the two groups and this means that; at least 21 years old female patients of Pima Indian heritage with high insulin and Glucose levels are at risk of being diabetic whiles those with low insulin and glucose levels are at no risk of being diabetic. This explanation is given on the premise that the response variable, Outcome=1 means diabetic and 0 otherwise.

