

```
import numpy as np
import pandas as pd

df = pd.read_csv("IRIS.csv")
df
```

sepal_length

sepal_width

petal_length

petal_width

species

0

5.1

3.5

1.4

0.2

Iris-setosa

1

4.9

3.0

1.4

0.2

Iris-setosa

2

4.7

3.2

1.3

0.2

Iris-setosa

3

4.6

3.1

1.5

0.2

Iris-setosa

4

5.0

3.6

1.4

0.2

Iris-setosa

...

...

...

...

...

...

145

6.7

3.0

5.2

2.3

Iris-virginica

146

6.3

2.5

5.0

1.9

Iris-virginica

147

6.5

3.0

5.2

2.0

Iris-virginica

148

6.2

3.4

5.4

2.3

Iris-virginica

149

5.9

3.0

5.1

1.8

Iris-virginica

150 rows × 5 columns

```
sub_df =  
df[["sepal_length", "sepal_width", "petal_length", "petal_width", "species"  
]]  
sub_df.head()
```

sepal_length

sepal_width

petal_length

petal_width

species

0

5.1

3.5

1.4

0.2

Iris-setosa

1

4.9

3.0

1.4

0.2

Iris-setosa

2

4.7

3.2

1.3

0.2

Iris-setosa

3

4.6

3.1

1.5

0.2

Iris-setosa

4

5.0

3.6

1.4

0.2

Iris-setosa

#Q1. Central Tendency Without directly calculating the mean, determine which sepal measurement (sepal_length or sepal_width) has a higher average by observing the dataset.

#Explanation = sepal_length has higher average because sepal_length has higher mean than the sepal_width.

```
sub_df.describe()
```

sepal_length

sepal_width

petal_length

petal_width

count

150.000000

150.000000

150.000000

150.000000

mean

5.843333

3.054000

3.758667

1.198667

std

0.828066

0.433594

1.764420

0.763161

min

4.300000

2.000000

1.000000

0.100000

25%

5.100000

2.800000

1.600000

0.300000

50%

5.800000

3.000000

4.350000

1.300000

75%

6.400000

3.300000

5.100000

1.800000

max

7.900000

4.400000

6.900000

2.500000

#Q2. Dispersion Compare the variability of petal measurements by checking the difference between the ranges of petal_length and petal_width. Which feature shows more spread?

#Explanation = range of petal_length has more variability

#range of petal_length

```
r1 = sub_df['petal_length'].max() - sub_df['petal_length'].min()  
r1
```

5.9

```
#range of petal_width
r2 = sub_df['petal_width'].max() - sub_df['petal_width'].min()
r2
```

2.4

#Q3. Quartiles Identify which feature, sepal_length or sepal_width, has a higher interquartile range (IQR), indicating greater variability in the middle 50% of the data.

#Explanation = IQR 1 has the greater variability. so, sepal_length has more variability.

```
#IQR of sepal_length
IQR1 = sub_df['sepal_length'].quantile(0.75) -
sub_df['sepal_length'].quantile(0.25)
IQR1
```

1.3000000000000007

```
#IQR of sepal_width
IQR2= sub_df['sepal_width'].quantile(0.75) -
sub_df['sepal_width'].quantile(0.25)
IQR2
```

0.5

#Q4. Standard Deviation Insight Based on the computed standard deviations of sepal_length and sepal_width, determine which attribute exhibits more variability and discuss what this indicates about the consistency of the two features.

#Explanation = sepal_length has more standard deviation which means sepal_length data set has more variability but less consistency and sepal_width has less standard deviation which means sepal_width has less consistency.

```
#Standard Deviation of sepal_length
sd1 = sub_df["sepal_length"].std()
sd1
```

0.8280661279778629

```
#Standard Deviation of sepal_width
sd2 = sub_df["sepal_width"].std()
sd2
```

0.4335943113621737

#Q5. Shape of Distribution Without plotting, assess whether the petal_length distribution is skewed. Use a descriptive statistic that measures skewness and infer the distribution shape.

#Explanation = The data of petal_length is left skewed because mean of petal length is less than the median

#Mean

```
mean= sub_df["petal_length"].mean()  
mean
```

3.7586666666666666

#Median

```
median= sub_df["petal_length"].median()  
median
```

4.35

#Skew

```
skew = sub_df["petal_length"].skew()  
skew
```

-0.27446425247378287

#Q6. Symmetry For which attribute, sepal_length or sepal_width, is the skewness closer to zero? What does this imply about the symmetry of its distribution?

#Explanation = Both sepal_length and sepal_width are right skewed data and both attributes has not skewness which is closer to zero

#Mean of sepal_length

```
mean_l=sub_df["sepal_length"].mean()  
mean_l
```

5.843333333333334

#Median of sepal_length

```
median_l=sub_df["sepal_length"].median()  
median_l
```

5.8

#Skew

```
skew_l = sub_df["sepal_length"].skew()  
skew_l
```

0.3149109566369728

#Mean of sepal_width

```
mean_w=sub_df["sepal_width"].mean()  
mean_w
```

3.0540000000000003


```
#Median of sepal_width
median_w=sub_df["sepal_width"].median()
median_w
```

3.0

```
#Skew
skew_w=sub_df["sepal_width"].skew()
skew_w
```

0.3340526621720866

#Q7. Coefficient of Variation Calculate the coefficient of variation (CV) for sepal_length and petal_length. Which feature has greater relative variability, and what does this tell you about the dataset?

#Explanation = A higher CV indicates greater disparity or variation in the data set

```
mean = sub_df["sepal_length"].mean()
mean
```

5.843333333333334

```
sd = sub_df["sepal_length"].std()
sd
```

0.8280661279778629

```
cv_s = sd/mean*100
```

```
mean = sub_df["petal_length"].mean()
mean
```

3.7586666666666666

```
sd = sub_df["petal_length"].std()
sd
```

1.7644204199522617

```
cv_p = sd/mean*100
```

*#Q8. Outlier Detection using IQR Identify the outliers in the sepal_length feature using the IQR method (define outliers as values that fall below $Q1 - 1.5 * IQR$ or above $Q3 + 1.5 * IQR$). How many outliers are present?*

#Explanation = there is no outliers

```
#Outliers of sepal_length
lower_bound = sub_df['sepal_length'].quantile(0.25) - 1.5 * IQR1
lower_bound
```

3.1499999999999986

```
upper_bound = sub_df['sepal_length'].quantile(0.75) + 1.5 * IQR1  
upper_bound
```

8.350000000000001

```
outliers = sub_df[(sub_df['sepal_length'] < lower_bound) |  
(sub_df['sepal_length'] > upper_bound)]  
print("Outliers Rows:")  
print(outliers)
```

Outliers Rows:

Empty DataFrame

Columns: [sepal_length, sepal_width, petal_length, petal_width,
species]

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