

Group A

Assignment 10

Data Visualization III

Import all the required Python Libraries

```
import numpy as np
import pandas as pd
import seaborn as sns
```

Load the Dataset into pandas dataframe.

```
iris_tf=sns.load_dataset('iris')
iris_tf
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
...
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

[150 rows x 5 columns]

List down the features and their types (e.g., numeric, nominal) available in the dataset.

```
iris_tf.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   sepal_length    150 non-null   float64
1   sepal_width     150 non-null   float64
2   petal_length    150 non-null   float64
3   petal_width     150 non-null   float64
4   species         150 non-null   object
```

```
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
iris_tf.head()
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

```
iris_tf.isnull().sum()
```

```
sepal_length    0
sepal_width     0
petal_length    0
petal_width     0
species         0
dtype: int64
```

```
iris_tf.describe()
```

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.057333	3.758000	1.199333
std	0.828066	0.435866	1.765298	0.762238
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

```
iris_tf.shape
```

```
(150, 5)
```

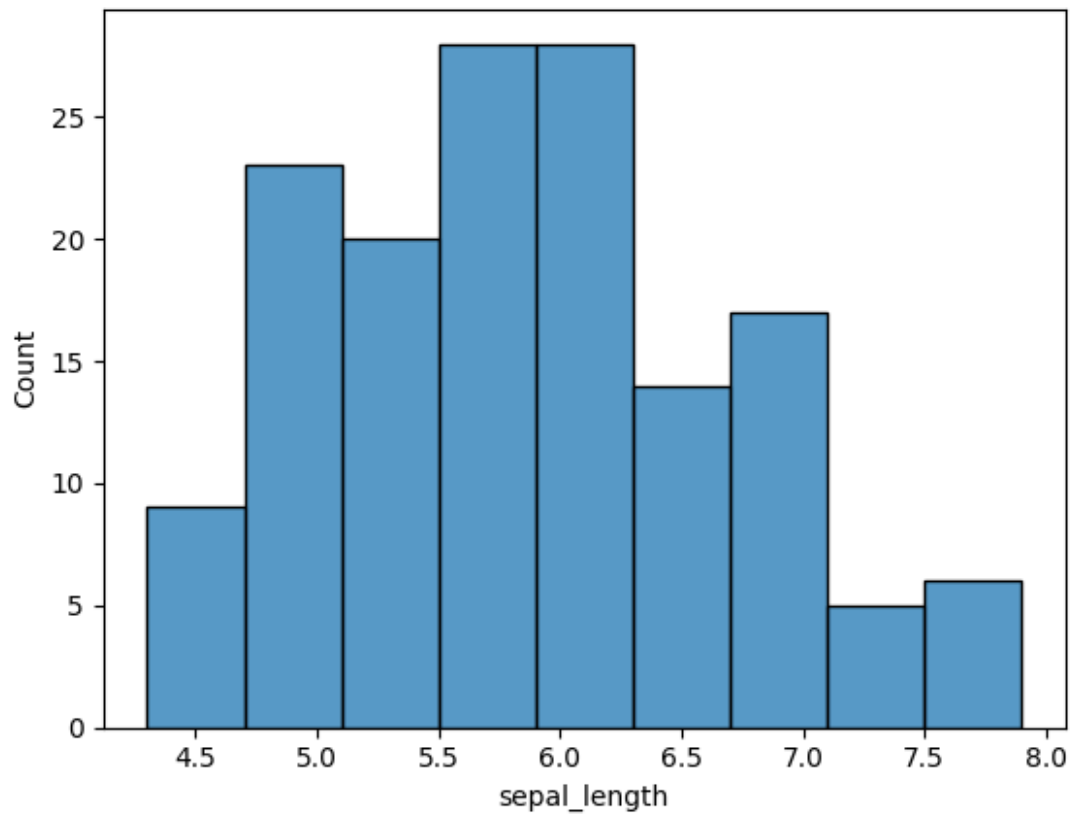
```
iris_tf.dtypes
```

```
sepal_length    float64
sepal_width     float64
petal_length     float64
petal_width     float64
species         object
dtype: object
```

Create a histogram for each feature in the dataset to illustrate the feature distributions.

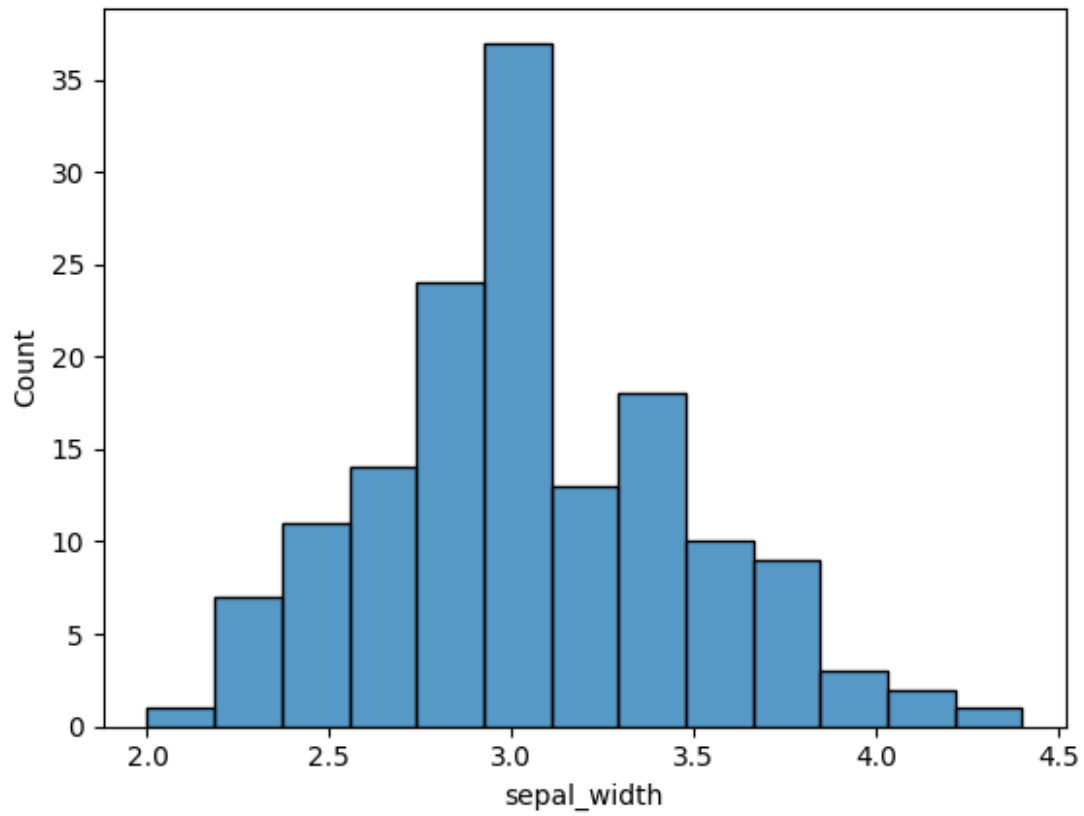
```
sns.histplot(iris_tf['sepal_length']) #Sepal length showing maximum
count between 5.5 and 6.3
```

```
<Axes: xlabel='sepal_length', ylabel='Count'>
```



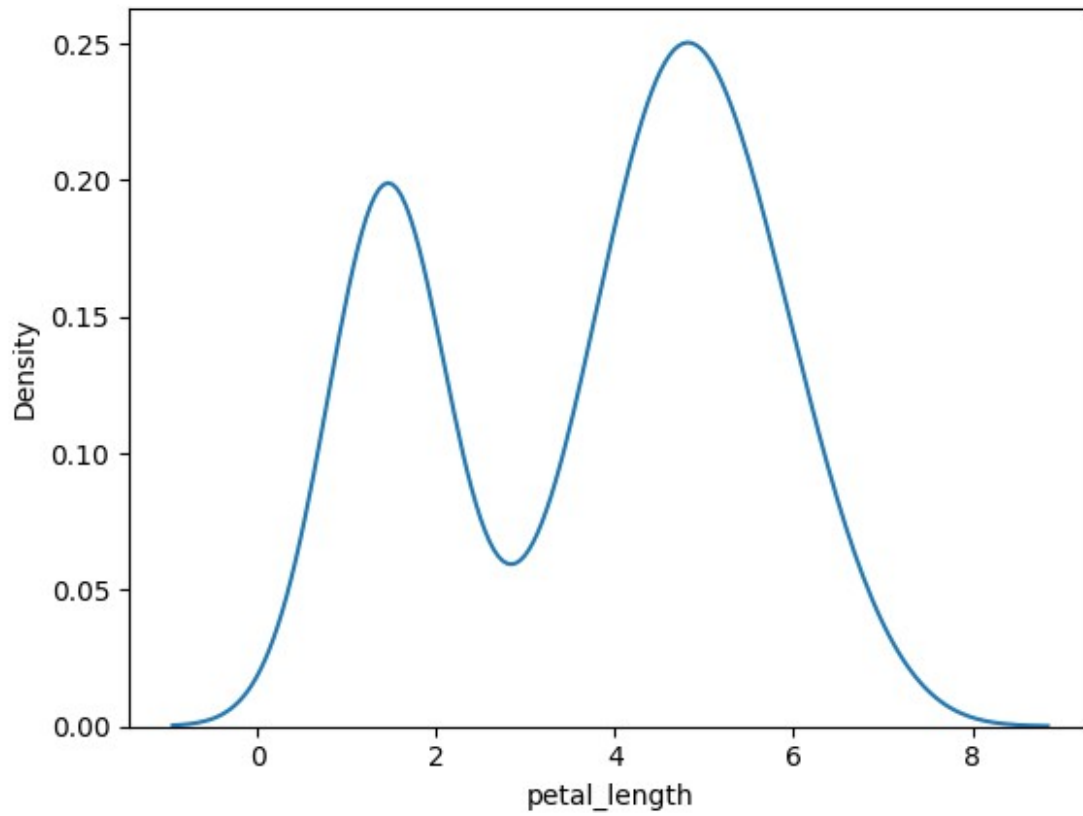
```
sns.histplot(iris_tf['sepal_width']) #Sepal width showing maximum  
count on 3
```

```
<Axes: xlabel='sepal_width', ylabel='Count'>
```

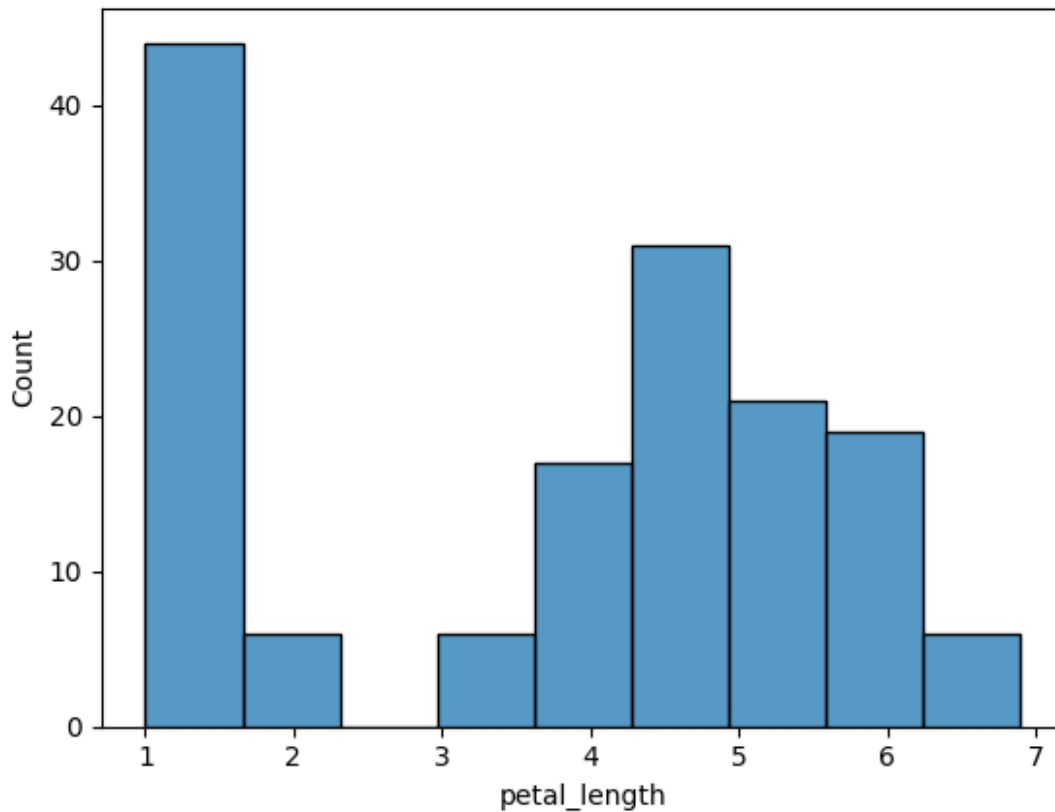


```
sns.kdeplot(iris_tf['petal_length']) #Petal length showing max density on 5
```

```
<Axes: xlabel='petal_length', ylabel='Density'>
```



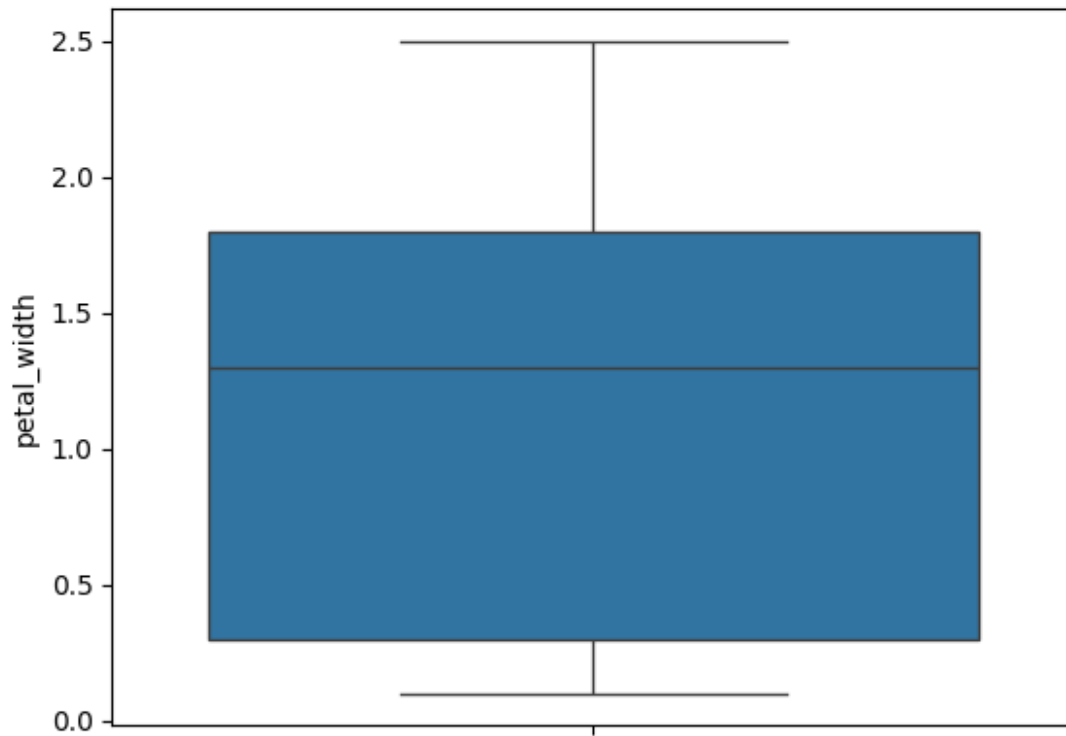
```
sns.histplot(iris_tf['petal_length']) #Petal length is max on 1  
<Axes: xlabel='petal_length', ylabel='Count'>
```



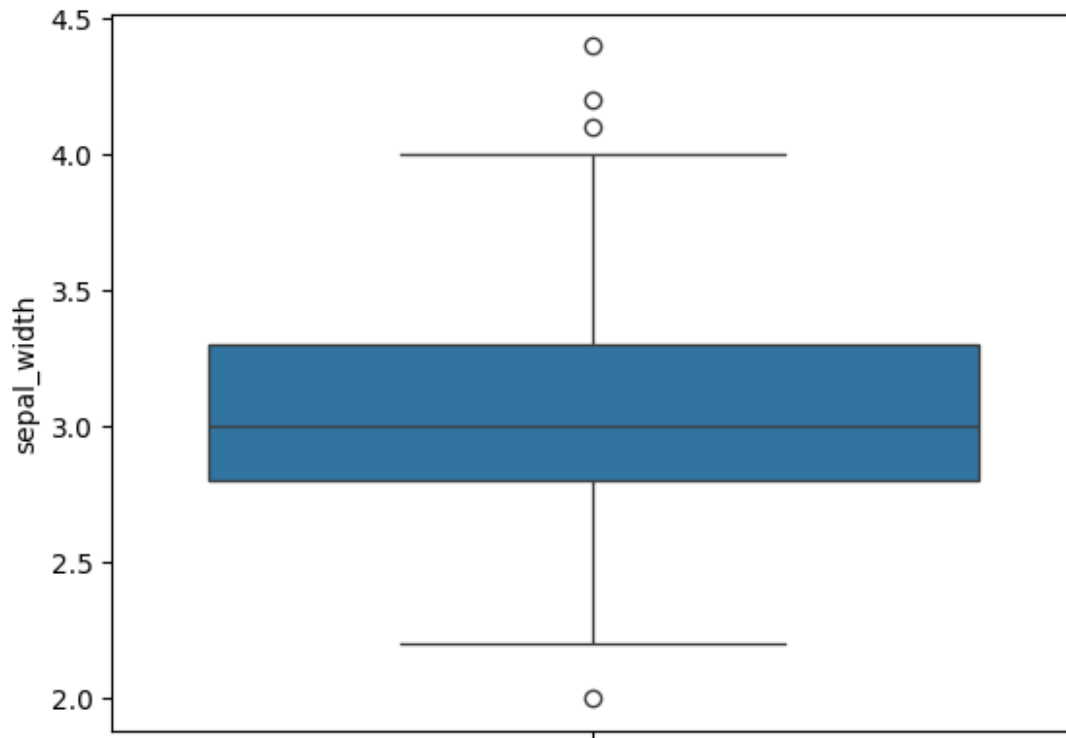
```
iris_tf['sepal_length'].skew() #Normally Distributed  
0.3149109566369728  
iris_tf['sepal_width'].skew() #Normally Distributed  
0.31896566471359966  
iris_tf['petal_length'].skew() #Normally Distributed  
-0.27488417975101276  
iris_tf['petal_width'].skew() #Normally Distributed  
-0.10296674764898116
```

Create a boxplot for each feature in the dataset

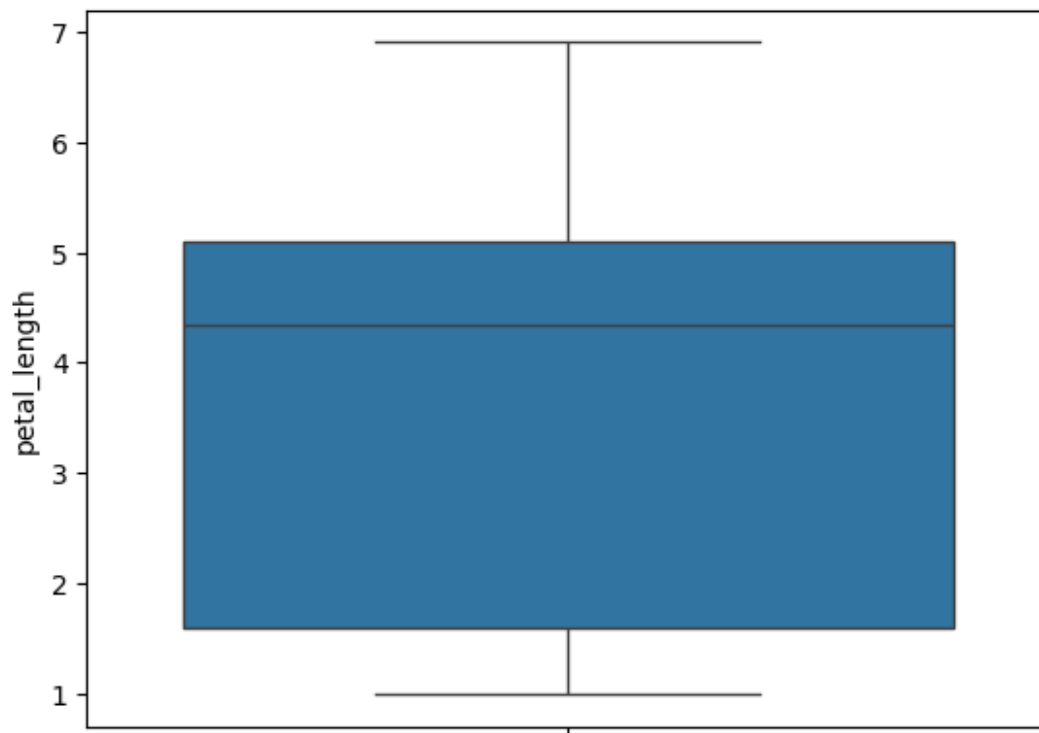
```
sns.boxplot(iris_tf['petal_width']) #Petal width has no outliers  
<Axes: ylabel='petal_width'>
```



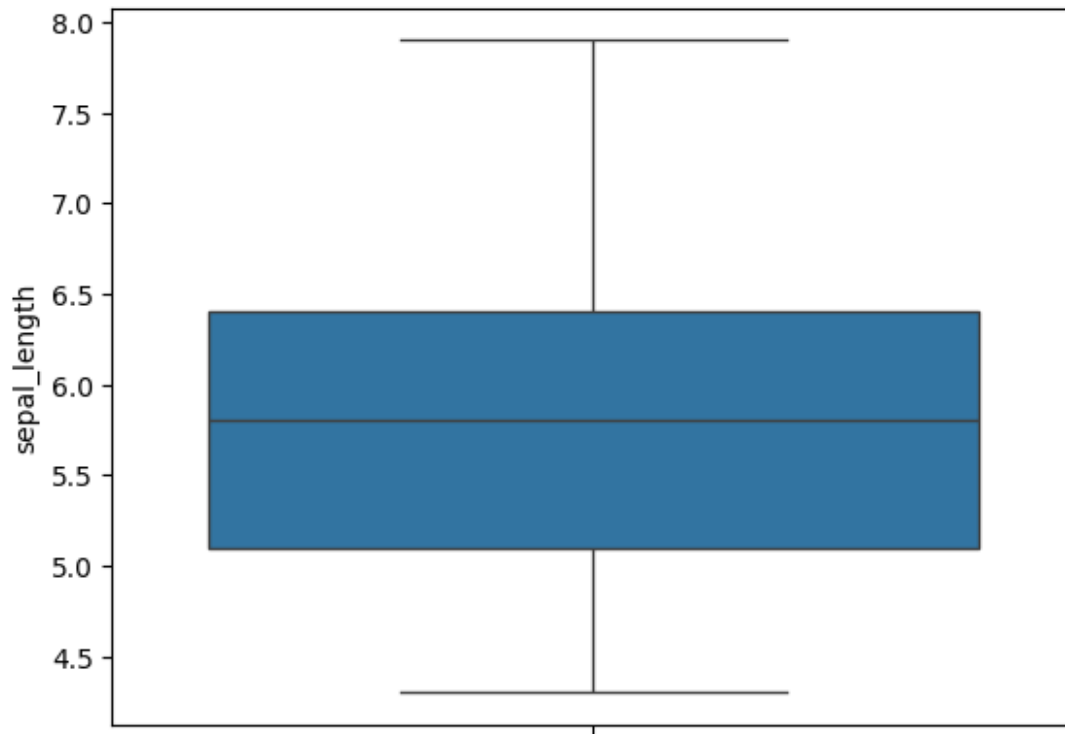
```
sns.boxplot(iris_tf['sepal_width']) #Sepal width has outliers on the  
upper fence and lower fence  
<Axes: ylabel='sepal_width'>
```



```
sns.boxplot(iris_tf['petal_length']) #Petal length has no outliers  
<Axes: ylabel='petal_length'>
```

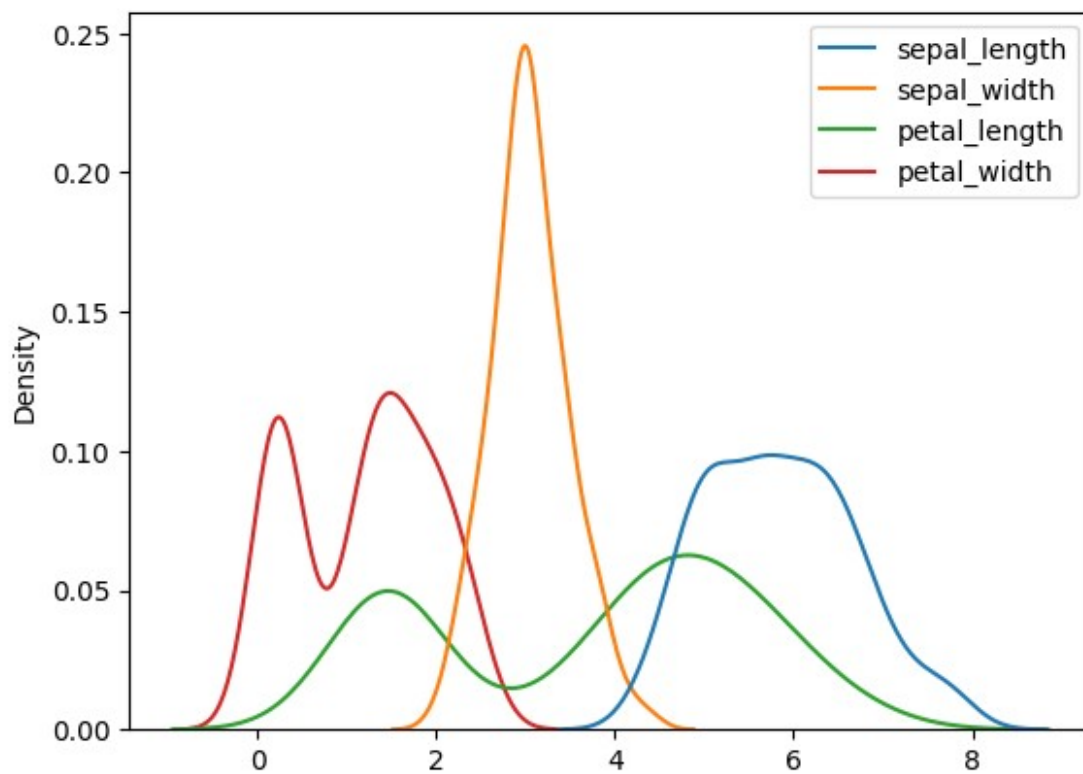



```
sns.boxplot(iris_tf['sepal_length']) #Sepal length has no outliers  
<Axes: ylabel='sepal_length'>
```



Compare distributions and identify outliers

```
sns.kdeplot(iris_tf) #Sepal width showing more density  
<Axes: ylabel='Density'>
```



```
iris_tf[(iris_tf['sepal_width']>4.0) | (iris_tf['sepal_width']<2.1)]
```

	sepal_length	sepal_width	petal_length	petal_width	species
15	5.7	4.4	1.5	0.4	setosa
32	5.2	4.1	1.5	0.1	setosa
33	5.5	4.2	1.4	0.2	setosa
60	5.0	2.0	3.5	1.0	versicolor