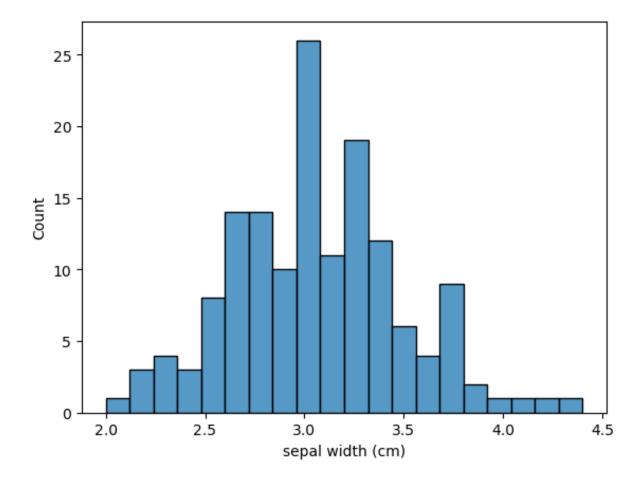
```
In [350...
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn import datasets
         iris = datasets.load_iris()
         iris data= np.array(iris['data'])
         iris_data_header = np.array(iris['feature_names'])
         iris target = np.array(iris['target'])
         iris_target_key = np.array(iris['target_names'])
         iris_target_fill = np.array(iris_target_key[iris_target])
         iris target header = np.array(['species name'])
         iris full data = np.column stack((iris data,iris target fill))
         iris_full_headers = np.append(iris_data_header,iris_target_header)
         iris_df = pd.DataFrame(iris_full_data, columns=iris_full_headers)
         iris_df = iris_df.astype({'sepal length (cm)': float, 'sepal width (cm)
         data = { "weight": [4.17, 5.58, 5.18, 6.11, 4.50, 4.61, 5.17, 4.53, 5.
         PlantGrowth = pd.DataFrame(data)
```

1a. Make a histogram of the variable Sepal.Width

```
In [351... #iris_df['sepal width (cm)']
sns.histplot(iris_df['sepal width (cm)'].astype(float), bins=20)
Out[351... <Axes: xlabel='sepal width (cm)', ylabel='Count'>
```



1b. Based on the histogram from 1a. which would you expect to be higher, the mean or the median? Why?

Answer: I would expect the median to be higher as this histogram appears to have more counted values on the left than the right

1c. Confirm my answer to 1b by finding these values

(I was wrong - the higher values on the right side pull the mean up higher than the median)

```
iris_df['sepal width (cm)'].astype(float).describe()

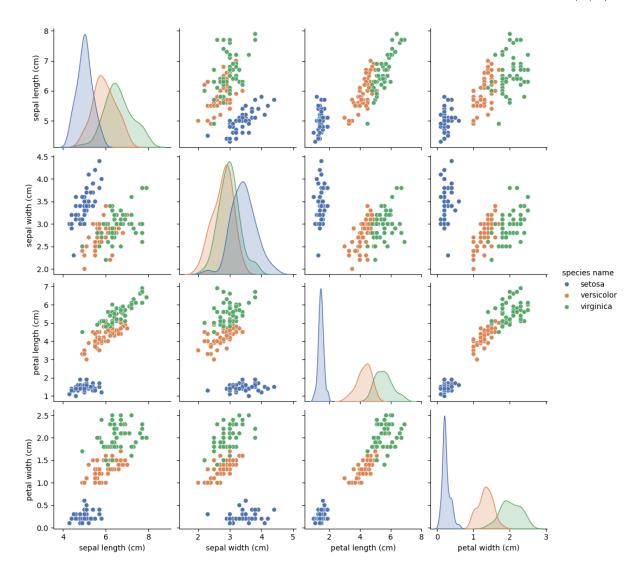
# alternate methods
# iris_df['sepal width (cm)'].mean()
# iris_df['sepal width (cm)'].median()
```

```
Out [352... count
                   150.000000
                     3.057333
          mean
          std
                     0.435866
                     2.000000
          min
          25%
                     2.800000
          50%
                     3.000000
          75%
                     3.300000
          max
                     4.400000
          Name: sepal width (cm), dtype: float64
          1d. Only 27% of the flowers have a Sepal Width higher than ____ cm.
In [353... print(f'Only 27% of the flowers have a Sepal Width higher than {iris_d
          # alternate method
          # sw_73_qt = iris_df['sepal width (cm)'].quantile(.73)
          # print(f'Only 27% of the flowers have a Sepal Width higher than {sw_7
         Only 27% of the flowers have a Sepal Width higher than 3.3 cm.
```

1e. Make scatterplots of each pair of the numerical values in iris

```
In [354... sns.pairplot(data=iris_df, hue='species name', palette='deep')
```

Out[354... <seaborn.axisgrid.PairGrid at 0x3487a4e20>

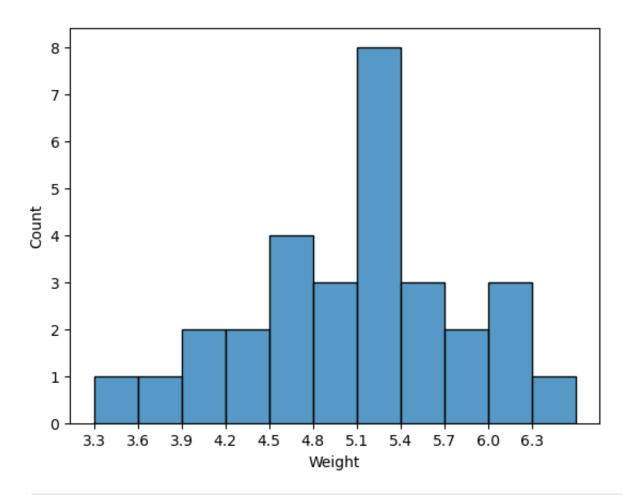


1f. Based on 1e. which two variables appear to have the strongest relationship? And which two appear to have the weakest relationship?

The strongest relationship appears to be between Petal Width and Petal Length due to the almost linear relationship. The weakest relationship appears to be between Sepal Width and Sepal Length due to there being almost no distinguishable pattern in the data.

2a. Make a histogram of the variable weight with breakpoints (bin edges) at every 0.3 units, starting at 3.3

```
In [360... ax_hist = sns.histplot(PlantGrowth['weight'], binwidth=(.3), binrange=
ax_hist.set_xticks(np.arange(3.3,6.6,.3))
ax_hist.set_xlabel('Weight')
plt.show(ax_hist)
```



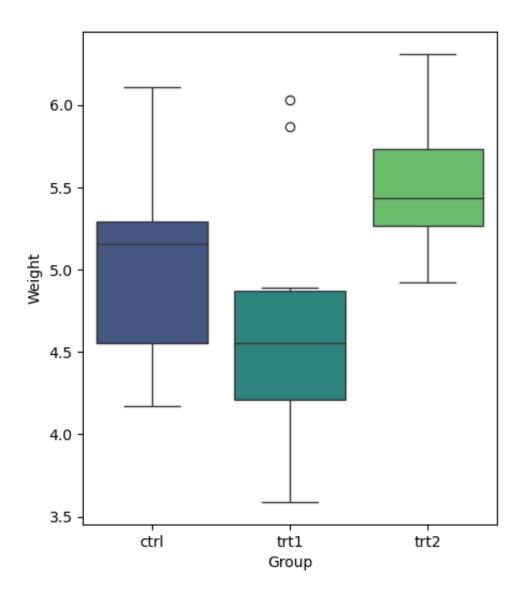
```
fig, bplots = plt.subplots(figsize=(5,6))
sns.boxplot(x=PlantGrowth['group'], y=PlantGrowth['weight'], palette='
bplots.set(xlabel='Group',ylabel='Weight')
plt.show()

# Alternate method for my benefit

#fig, bplots = plt.subplots(figsize=(5,6))

#bplots.boxplot([grouping['weight'] for x, grouping in PlantGrowth.gro
#bplots.set(xlabel='Group',ylabel='Weight')

#plt.show()
```



2c. Based on the boxplots in #2b, approximately what percentage of the "trt1" weights are below the minimum "trt2" weight?

It would appear all but two outliers of the trt1 weight values are below the minimum trt2 weight values.

2d. Find the exact percentage of the "trt1" weights that are below the minimum "trt2" weight.

```
In [363... trt2_min = PlantGrowth['weight'][PlantGrowth['group'] == 'trt2'].min()

trt1_counter = 0
for x in PlantGrowth['weight'][(PlantGrowth['group'] == 'trt1') & (Platr1_counter += 1)

trt1_total_count = len(PlantGrowth['weight'][PlantGrowth['group']=='trt1_percentage = trt1_counter/trt1_total_count*100

print(f'The exact percentage of the trt1 weights below the minimum trt
```

The exact percentage of the trt1 weights below the minimum trt12 weight is 80.0%

2e. Only including plants with a weight above 5.5, make a barplot of the variable group. Make the barplot colorful using some color palette (in R, try running? heat.colors and/or check out https://www.r-bloggers.com/palettes-in-r/).

```
In [364... min_weight = 5.5
    pg_filter = PlantGrowth[PlantGrowth['weight'] > 5.5]

    pg_gr_counts = pg_filter['group'].value_counts()
    pg_gr_label = pg_gr_counts.index.tolist()

    pg_bplot = sns.barplot(x=pg_gr_label, y=pg_gr_counts, palette='magma', pg_bplot.set(xlabel='Group', ylabel='Count')

    plt.show()
```

