

## Observations and Insights

```

In [1]: # Dependencies and Setup
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import scipy.stats as st

# Study data files
mouse_metadata_path = "data/Mouse_metadata.csv"
study_results_path = "data/Study_results.csv"

# Read the mouse data and the study results
mouse_metadata = pd.read_csv(mouse_metadata_path)
# print(mouse_metadata)
study_results = pd.read_csv(study_results_path)
# print(study_results)

# Combine the data into a single dataset
combined_data = pd.merge(study_results, mouse_metadata, on="Mouse ID", how="left")
combined_data

# Display the data table for preview
# combined_data

# drop_duplicates()
# combined_data.sort_values("Mouse ID", inplace = True)
# combined_data.drop_duplicates(subset = "Mouse ID",
#                               keep = False, inplace = True)
# combined_data.describe()

```

Out[1]:

|      | Mouse ID | Timepoint | Tumor Volume (mm3) | Metastatic Sites | Drug Regimen | Sex    | Age_months | Weight (g) |
|------|----------|-----------|--------------------|------------------|--------------|--------|------------|------------|
| 0    | b128     | 0         | 45.000000          | 0                | Capomulin    | Female | 9          | 22         |
| 1    | f932     | 0         | 45.000000          | 0                | Ketapril     | Male   | 15         | 29         |
| 2    | g107     | 0         | 45.000000          | 0                | Ketapril     | Female | 2          | 29         |
| 3    | a457     | 0         | 45.000000          | 0                | Ketapril     | Female | 11         | 30         |
| 4    | c819     | 0         | 45.000000          | 0                | Ketapril     | Male   | 21         | 25         |
| ...  | ...      | ...       | ...                | ...              | ...          | ...    | ...        | ...        |
| 1888 | r944     | 45        | 41.581521          | 2                | Capomulin    | Male   | 12         | 25         |
| 1889 | u364     | 45        | 31.023923          | 3                | Capomulin    | Male   | 18         | 17         |
| 1890 | p438     | 45        | 61.433892          | 1                | Ceftamin     | Female | 11         | 26         |
| 1891 | x773     | 45        | 58.634971          | 4                | Placebo      | Female | 21         | 30         |
| 1892 | b879     | 45        | 72.555239          | 2                | Stelasyn     | Female | 4          | 26         |

1893 rows × 8 columns

```
In [2]: # Checking the number of mice.
number_of_mouse = len(combined_data['Mouse ID'].unique())
number_of_mouse
```

Out[2]: 249

```
In [3]: # Getting the duplicate mice by ID number that shows up for Mouse ID and Timepoint.

duplicated_ID = combined_data.loc[combined_data.duplicated(subset=["Mouse ID",
"Timepoint"]), "Mouse ID"].unique()
duplicated_ID
```

Out[3]: array(['g989'], dtype=object)

```
In [4]: # Optional: Get all the data for the duplicate mouse ID.
combined_data.loc[combined_data["Mouse ID"]=="g989"]
```

Out[4]:

|             | Mouse ID | Timepoint | Tumor Volume (mm3) | Metastatic Sites | Drug Regimen | Sex    | Age_months | Weight (g) |
|-------------|----------|-----------|--------------------|------------------|--------------|--------|------------|------------|
| <b>107</b>  | g989     | 0         | 45.000000          | 0                | Propriva     | Female | 21         | 26         |
| <b>137</b>  | g989     | 0         | 45.000000          | 0                | Propriva     | Female | 21         | 26         |
| <b>329</b>  | g989     | 5         | 48.786801          | 0                | Propriva     | Female | 21         | 26         |
| <b>360</b>  | g989     | 5         | 47.570392          | 0                | Propriva     | Female | 21         | 26         |
| <b>620</b>  | g989     | 10        | 51.745156          | 0                | Propriva     | Female | 21         | 26         |
| <b>681</b>  | g989     | 10        | 49.880528          | 0                | Propriva     | Female | 21         | 26         |
| <b>815</b>  | g989     | 15        | 51.325852          | 1                | Propriva     | Female | 21         | 26         |
| <b>869</b>  | g989     | 15        | 53.442020          | 0                | Propriva     | Female | 21         | 26         |
| <b>950</b>  | g989     | 20        | 55.326122          | 1                | Propriva     | Female | 21         | 26         |
| <b>1111</b> | g989     | 20        | 54.657650          | 1                | Propriva     | Female | 21         | 26         |
| <b>1195</b> | g989     | 25        | 56.045564          | 1                | Propriva     | Female | 21         | 26         |
| <b>1380</b> | g989     | 30        | 59.082294          | 1                | Propriva     | Female | 21         | 26         |
| <b>1592</b> | g989     | 35        | 62.570880          | 2                | Propriva     | Female | 21         | 26         |

```
In [5]: # Create a clean DataFrame by dropping the duplicate mouse by its ID.
combined_data = combined_data.loc[combined_data["Mouse ID"] != 'g989']
combined_data
```

Out[5]:

|      | Mouse ID | Timepoint | Tumor Volume (mm3) | Metastatic Sites | Drug Regimen | Sex    | Age_months | Weight (g) |
|------|----------|-----------|--------------------|------------------|--------------|--------|------------|------------|
| 0    | b128     | 0         | 45.000000          | 0                | Capomulin    | Female | 9          | 22         |
| 1    | f932     | 0         | 45.000000          | 0                | Ketapril     | Male   | 15         | 29         |
| 2    | g107     | 0         | 45.000000          | 0                | Ketapril     | Female | 2          | 29         |
| 3    | a457     | 0         | 45.000000          | 0                | Ketapril     | Female | 11         | 30         |
| 4    | c819     | 0         | 45.000000          | 0                | Ketapril     | Male   | 21         | 25         |
| ...  | ...      | ...       | ...                | ...              | ...          | ...    | ...        | ...        |
| 1888 | r944     | 45        | 41.581521          | 2                | Capomulin    | Male   | 12         | 25         |
| 1889 | u364     | 45        | 31.023923          | 3                | Capomulin    | Male   | 18         | 17         |
| 1890 | p438     | 45        | 61.433892          | 1                | Ceftamin     | Female | 11         | 26         |
| 1891 | x773     | 45        | 58.634971          | 4                | Placebo      | Female | 21         | 30         |
| 1892 | b879     | 45        | 72.555239          | 2                | Stelasyn     | Female | 4          | 26         |

1880 rows × 8 columns

```
In [6]: # Checking the number of mice in the clean DataFrame.
number_of_mice = len(combined_data["Mouse ID"].unique())
number_of_mice
```

Out[6]: 248

## Summary Statistics

```

In [7]: # Generate a summary statistics table of mean, median, variance, standard deviation, and SEM of the tumor volume for each regimen
mean = combined_data.groupby(["Drug Regimen"]).mean()["Tumor Volume (mm3)"]
mean

median = combined_data.groupby(["Drug Regimen"]).median()["Tumor Volume (mm3)"]
median

variance = combined_data.groupby(["Drug Regimen"]).var()["Tumor Volume (mm3)"]
variance

st_dev = combined_data.groupby(["Drug Regimen"]).std()["Tumor Volume (mm3)"]
st_dev

SEM = combined_data.groupby(["Drug Regimen"]).sem()["Tumor Volume (mm3)"]
SEM
# Use groupby and summary statistical methods to calculate the following properties of each drug regimen:
# mean, median, variance, standard deviation, and SEM of the tumor volume.
# Assemble the resulting series into a single summary dataframe.

statistic = pd.DataFrame({"mean": mean,
                           "median": median,
                           "variance": variance,
                           "standard deviation": st_dev,
                           "SEM": SEM})
statistic

```

Out[7]:

|              | mean      | median    | variance  | standard deviation | SEM      |
|--------------|-----------|-----------|-----------|--------------------|----------|
| Drug Regimen |           |           |           |                    |          |
| Capomulin    | 40.675741 | 41.557809 | 24.947764 | 4.994774           | 0.329346 |
| Ceftamin     | 52.591172 | 51.776157 | 39.290177 | 6.268188           | 0.469821 |
| Infubinol    | 52.884795 | 51.820584 | 43.128684 | 6.567243           | 0.492236 |
| Ketapril     | 55.235638 | 53.698743 | 68.553577 | 8.279709           | 0.603860 |
| Naftisol     | 54.331565 | 52.509285 | 66.173479 | 8.134708           | 0.596466 |
| Placebo      | 54.033581 | 52.288934 | 61.168083 | 7.821003           | 0.581331 |
| Propriva     | 52.320930 | 50.446266 | 43.852013 | 6.622085           | 0.544332 |
| Ramicane     | 40.216745 | 40.673236 | 23.486704 | 4.846308           | 0.320955 |
| Stelasyn     | 54.233149 | 52.431737 | 59.450562 | 7.710419           | 0.573111 |
| Zoniferol    | 53.236507 | 51.818479 | 48.533355 | 6.966589           | 0.516398 |

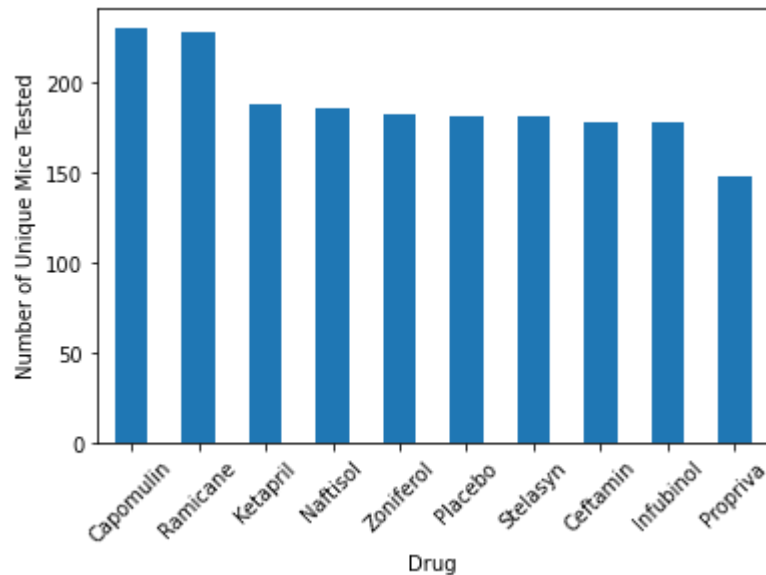
```
In [8]: # Generate a summary statistics table of mean, median, variance, standard deviation, and SEM of the tumor volume for each regimen
summary_table = combined_data.groupby(["Drug Regimen"]).agg({"Tumor Volume (mm 3)": ["mean", "median", "var", "std", "sem"]})
summary_table
# Using the aggregation method, produce the same summary statistics in a single line
```

Out[8]:

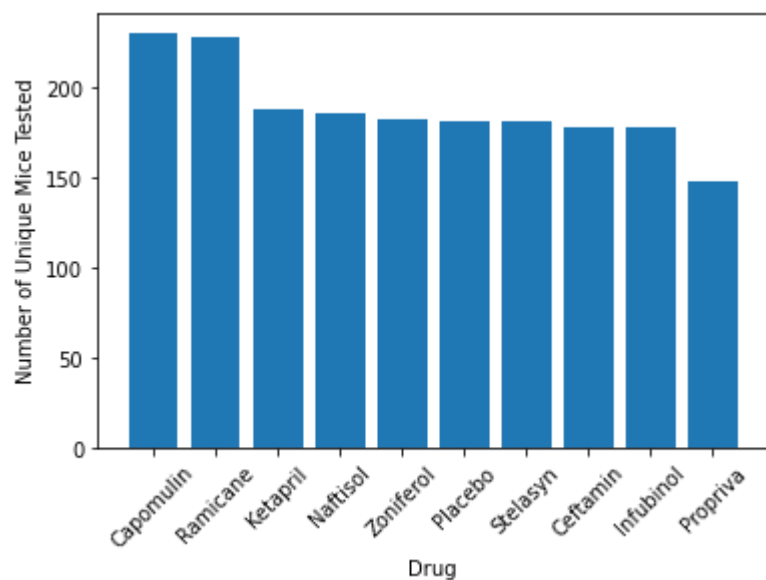
|              | Tumor Volume (mm3) |           |           |          |          |
|--------------|--------------------|-----------|-----------|----------|----------|
|              | mean               | median    | var       | std      | sem      |
| Drug Regimen |                    |           |           |          |          |
| Capomulin    | 40.675741          | 41.557809 | 24.947764 | 4.994774 | 0.329346 |
| Ceftamin     | 52.591172          | 51.776157 | 39.290177 | 6.268188 | 0.469821 |
| Infubinol    | 52.884795          | 51.820584 | 43.128684 | 6.567243 | 0.492236 |
| Ketapril     | 55.235638          | 53.698743 | 68.553577 | 8.279709 | 0.603860 |
| Naftisol     | 54.331565          | 52.509285 | 66.173479 | 8.134708 | 0.596466 |
| Placebo      | 54.033581          | 52.288934 | 61.168083 | 7.821003 | 0.581331 |
| Propriva     | 52.320930          | 50.446266 | 43.852013 | 6.622085 | 0.544332 |
| Ramicane     | 40.216745          | 40.673236 | 23.486704 | 4.846308 | 0.320955 |
| Stelasyn     | 54.233149          | 52.431737 | 59.450562 | 7.710419 | 0.573111 |
| Zoniferol    | 53.236507          | 51.818479 | 48.533355 | 6.966589 | 0.516398 |

## Bar and Pie Charts

```
In [9]: # Generate a bar plot showing the total number of unique mice tested on each drug regimen using pandas.  
drug_count = combined_data["Drug Regimen"].value_counts()  
drug_count  
drug_count.plot(kind="bar", rot=45)  
plt.xlabel("Drug")  
plt.ylabel("Number of Unique Mice Tested")  
plt.show()
```

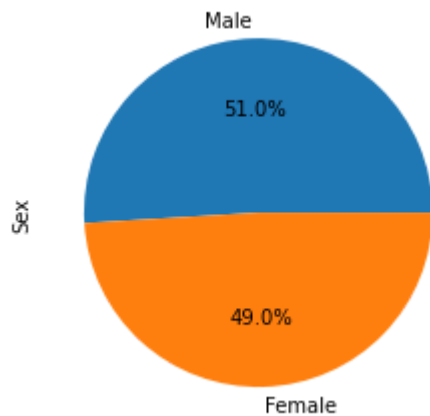


```
In [10]: # Generate a bar plot showing the total number of unique mice tested on each drug regimen using pyplot.  
plt.bar(drug_count.index.values, drug_count.values)  
plt.xticks(rotation=45)  
plt.xlabel("Drug")  
plt.ylabel("Number of Unique Mice Tested")  
plt.show()
```



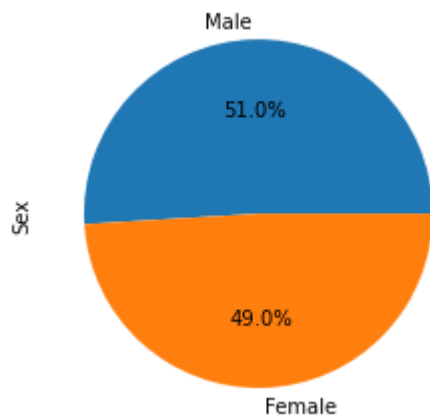
```
In [11]: # Generate a pie plot showing the distribution of female versus male mice using pandas

gender_count = combined_data["Sex"].value_counts()
gender_count
gender_count.plot(kind="pie", autopct="%1.1f%")
plt.show()
```



```
In [12]: # Generate a pie plot showing the distribution of female versus male mice using pyplot

plt.pie(gender_count.values, labels=gender_count.index.values, autopct="%1.1f%")
plt.ylabel("Sex")
plt.show()
```



## Quartiles, Outliers and Boxplots



```
In [13]: # Calculate the final tumor volume of each mouse across four of the treatment
          # regimens:
          # Capomulin, Ramicane, Infubinol, and Ceftamin

          # Start by getting the last (greatest) timepoint for each mouse
          last_tp_df = combined_data.drop_duplicates(subset=['Mouse ID'], keep='last')
          last_tp_df.head()

          # Prepare separate df for each drug
          Capomulin_df = last_tp_df[last_tp_df['Drug Regimen']=='Capomulin']
          Ramicane_df = last_tp_df[last_tp_df['Drug Regimen']=='Ramicane']
          Infubinol_df = last_tp_df[last_tp_df['Drug Regimen']=='Infubinol']
          Ceftamin_df = last_tp_df[last_tp_df['Drug Regimen']=='Ceftamin']
```

```
In [14]: # Merge this group df with the original dataframe to get the tumor volume at the last timepoint
Capomulin_merge = pd.merge(Capomulin_df,combined_data,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='left')
Capomulin_merge
Ramicane_merge = pd.merge(Capomulin_merge,Ramicane_df,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='outer')
Ramicane_merge
Infubinol_merge = pd.merge(Ramicane_merge,Infubinol_df,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='outer')
Infubinol_merge
Total_merge = pd.merge(Infubinol_merge,Ceftamin_df,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='outer')
Total_merge.head
```

```
Out[14]: <bound method NDFrame.head of
astatic Sites Drug Regimen \
0      w150      10      39.952347      0      Capomulin
1      r157      15      46.539206      0      Capomulin
2      f966      20      30.485985      0      Capomulin
3      j246      35      38.753265      1      Capomulin
4      b128      45      38.982878      2      Capomulin
..      ...      ...      ...      ...      ...
95     k210      45      68.923185      3      Ceftamin
96     x822      45      61.386660      3      Ceftamin
97     l733      45      64.299830      1      Ceftamin
98     o287      45      59.741901      4      Ceftamin
99     p438      45      61.433892      1      Ceftamin

      Sex  Age_months  Weight (g)
0     Male         23         23
1     Male         22         25
2     Male         16         17
3  Female         21         21
4  Female          9         22
..      ...      ...      ...
95    Male         15         28
96    Male          3         29
97  Female          4         30
98    Male          2         28
99  Female         11         26
```

```
[100 rows x 8 columns]>
```

```
In [15]: # Put treatments into a list for for loop (and later for plot labels)
# Create empty list to fill with tumor vol data (for plotting)
# Calculate the IQR and quantitatively determine if there are any potential outliers.

# Locate the rows which contain mice on each drug and get the tumor volume
s

# add subset

# Determine outliers using upper and lower bounds

Total_tumor_vol_data = Total_merge['Tumor Volume (mm3)']
quartiles = Total_tumor_vol_data.quantile([.25,.5,.75])
lowerq = quartiles[0.25]
upperq = quartiles[0.75]
iqr = upperq-lowerq

lower_bound = lowerq - (1.5*iqr)
upper_bound = upperq + (1.5*iqr)
print(f"The four treatments potential outliers could be values below {lower_bound} and above {upper_bound} could be outliers.")
```

The four treatments potential outliers could be values below 3.0739668737499812 and above 94.04403868375002 could be outliers.

```
In [16]: ##### CAPOMULIN
# Merge this group df with the original dataframe to get the tumor volume at the last timepoint
Capomulin_merge = pd.merge(Capomulin_df,combined_data,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='left')
Capomulin_merge.head()

# Preparing Tumor volume data for Capomulin
cap_tumor_vol_data = Capomulin_merge['Tumor Volume (mm3)']

quartiles = cap_tumor_vol_data.quantile([.25,.5,.75])
lowerq = quartiles[0.25]
upperq = quartiles[0.75]
iqr = upperq-lowerq

lower_bound = lowerq - (1.5*iqr)
upper_bound = upperq + (1.5*iqr)
print(f"Capomulin potential outliers could be values below {lower_bound} and above {upper_bound} could be outliers.")
```

Capomulin potential outliers could be values below 20.704561649999999 and above 51.83201549 could be outliers.

```
In [17]: ##### RAMICANE
# Merge this group df with the original dataframe to get the tumor volume at the last timepoint
Ramicane_merge = pd.merge(Ramicane_df,combined_data,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='left')
Ramicane_merge.head()

# Preparing Tumor volume data for Ramicane
ram_tumor_vol_data = Ramicane_merge['Tumor Volume (mm3)']

quartiles2 = ram_tumor_vol_data.quantile([.25,.5,.75])
lowerq2 = quartiles2[0.25]
upperq2 = quartiles2[0.75]
iqr2 = upperq2-lowerq2

lower_bound2 = lowerq2 - (1.5*iqr2)
upper_bound2 = upperq2 + (1.5*iqr2)
print(f"Capomulin potential outliers could be values below {lower_bound2} and above {upper_bound2} could be outliers.")
```

Capomulin potential outliers could be values below 17.912664470000003 and above 54.30681135 could be outliers.

```
In [18]: ##### INFUBINOL
# Merge this group df with the original dataframe to get the tumor volume at the last timepoint
Infubinol_merge = pd.merge(Infubinol_df,combined_data,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='left')
Infubinol_merge.head()

# Preparing Tumor volume data for infubinol
infu_tumor_vol_data = Infubinol_merge['Tumor Volume (mm3)']

quartiles3 = infu_tumor_vol_data.quantile([.25,.5,.75])
lowerq3 = quartiles3[0.25]
upperq3 = quartiles3[0.75]
iqr3 = upperq3-lowerq3

lower_bound3 = lowerq3 - (1.5*iqr3)
upper_bound3 = upperq3 + (1.5*iqr3)
print(f"Capomulin potential outliers could be values below {lower_bound3} and above {upper_bound3} could be outliers.")
```

Capomulin potential outliers could be values below 36.83290494999999 and above 82.74144559000001 could be outliers.

```
In [19]: ##### CEFTAMINE
# Merge this group df with the original dataframe to get the tumor volume at the last timepoint

Ceftamin_merge = pd.merge(Ceftamin_df,combined_data,on=["Mouse ID","Drug Regimen","Sex","Age_months","Weight (g)","Timepoint","Tumor Volume (mm3)","Metastatic Sites"],how='left')
Ceftamin_merge.head()

# Preparing Tumor volume data for Ceftmaine

ceft_tumor_vol_data = Ceftamin_merge['Tumor Volume (mm3)']

quartiles4 = ceft_tumor_vol_data.quantile([.25,.5,.75])
lowerq4 = quartiles4[0.25]
upperq4 = quartiles4[0.75]
iqr4 = upperq4-lowerq4

lower_bound4 = lowerq4 - (1.5*iqr4)
upper_bound4 = upperq4 + (1.5*iqr4)
print(f"Capomulin potential outliers could be values below {lower_bound4} and above {upper_bound4} could be outliers.")
```

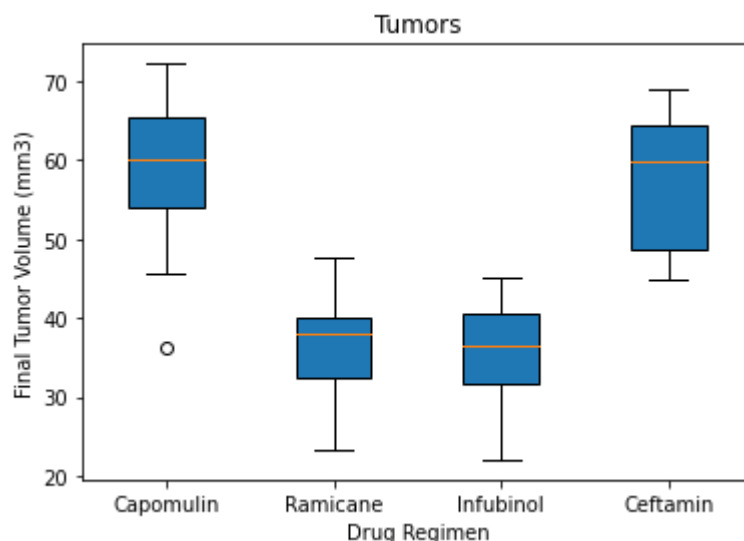
Capomulin potential outliers could be values below 25.355449580000002 and above 87.66645829999999 could be outliers.

```
In [20]: data_to_plot = [infu_tumor_vol_data,cap_tumor_vol_data,ram_tumor_vol_data,ceft
_tumor_vol_data]

fig1, ax1 = plt.subplots()
ax1.set_title('Tumors')
ax1.set_ylabel('Final Tumor Volume (mm3)')
ax1.set_xlabel('Drug Regimen')

ax1.boxplot(data_to_plot,labels=["Capomulin","Ramicane","Infubinol","Ceftamin"
], patch_artist=True, vert=True)

plt.savefig('boxplot')
plt.show()
```



## Line and Scatter Plots

```
In [21]: # Generate a line plot of tumor volume vs. time point for a mouse treated with
Capomulin
Cap_line = combined_data[combined_data["Drug Regimen"] == "Capomulin"]
Cap_line
forline_df = Cap_line.loc[Cap_line["Mouse ID"] == "b128",:]
forline_df.head()
```

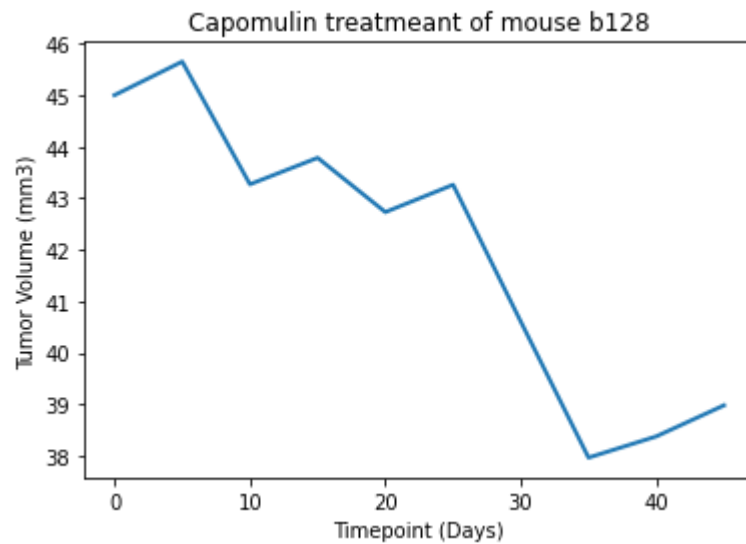
Out[21]:

|     | Mouse ID | Timepoint | Tumor Volume (mm3) | Metastatic Sites | Drug Regimen | Sex    | Age_months | Weight (g) |
|-----|----------|-----------|--------------------|------------------|--------------|--------|------------|------------|
| 0   | b128     | 0         | 45.000000          | 0                | Capomulin    | Female | 9          | 22         |
| 472 | b128     | 5         | 45.651331          | 0                | Capomulin    | Female | 9          | 22         |
| 679 | b128     | 10        | 43.270852          | 0                | Capomulin    | Female | 9          | 22         |
| 719 | b128     | 15        | 43.784893          | 0                | Capomulin    | Female | 9          | 22         |
| 933 | b128     | 20        | 42.731552          | 0                | Capomulin    | Female | 9          | 22         |

```
In [22]: x_axisTP = forline_df["Timepoint"]
tumsiz = forline_df["Tumor Volume (mm3)"]

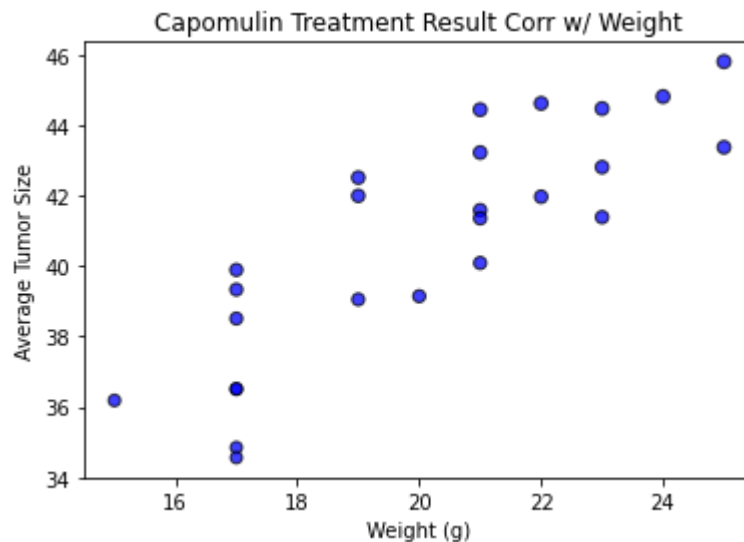
plt.title('Capomulin treatmeant of mouse b128')
plt.plot(x_axisTP, tumsiz,linewidth=2, markersize=12)
plt.xlabel('Timepoint (Days)')
plt.ylabel('Tumor Volume (mm3)')

plt.savefig('linechart')
plt.show()
```



```
In [23]: # Generate a scatter plot of average tumor volume vs. mouse weight for the Capomulin regimen
cap_regimen = Cap_line.groupby("Mouse ID").mean()
x_axis = cap_regimen["Weight (g)"]
y_axis = cap_regimen["Tumor Volume (mm3)"]

plt.scatter(x_axis, y_axis, marker="o", facecolors="blue", edgecolors="black", s=y_axis, alpha=0.75)
plt.xlabel("Weight (g)")
plt.ylabel("Average Tumor Size")
plt.title("Capomulin Treatment Result Corr w/ Weight")
plt.show()
```



## Correlation and Regression

```
In [24]: # Calculate the correlation coefficient and linear regression model

corr=round(st.pearsonr(cap_regimen['Weight (g)'],cap_regimen['Tumor Volume (mm3)'])[0],2)
print(f"The correlation between mouse weight and average tumor volume is {corr}")
```

The correlation between mouse weight and average tumor volume is 0.84

```
In [25]: model=st.linregress(cap_regimen['Weight (g)'],cap_regimen['Tumor Volume (mm3)'])
model
```

```
Out[25]: LinregressResult(slope=0.9544396890241045, intercept=21.552160532685015, rvalue=0.8419363424694718, pvalue=1.3225722434712642e-07, stderr=0.1275435903320134)
```

```
In [26]: #capavg['Weight (g)']
m_slope = 0.9544396890241045
m_intercept = 21.552160532685015
```



```

In [39]: ## for mouse weight and average tumor volume for the Capomulin regimen

x_values = cap_regimen["Weight (g)"]
y_values = cap_regimen["Tumor Volume (mm3)"]
m_slope, m_intercept, m_rvalue, m_pvalue, m_std_err = st.linregress(x_values,
y_values)
regress_values = x_values * m_slope + m_intercept
line_eq = "y = " + str(round(m_slope,2)) + "x + " + str(round(m_intercept,2))

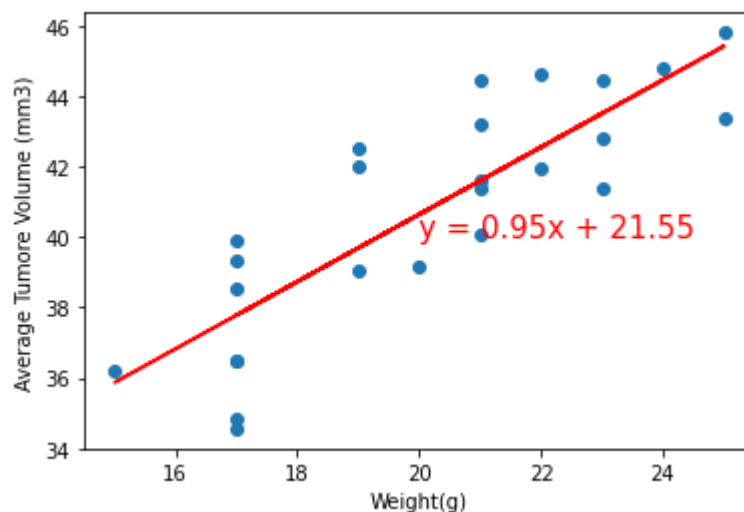
plt.scatter(x_values,y_values)
plt.plot(x_values, regress_values,"r-")
plt.annotate(line_eq,(20,40),fontsize=15,color="red")
plt.xlabel('Weight(g)')
plt.ylabel('Average Tumore Volume (mm3)')
plt.show()

## murder_rate = crime_data.iloc[:, 5]
## m_slope, m_int, m_r, m_p, m_std_err = stats.linregress(year, murder_rate)
## m_fit = m_slope * year + m_int
## plt.scatter(year,murder_rate)
## plt.plot(year,m_fit,"--")
## plt.xticks(year, rotation=90)
## plt.xlabel('Year')
## plt.ylabel('Murder Rate')
## plt.show()

# y_values = cap_regimen['Weight (g)']*m_slope+m_intercept
# plt.scatter(cap_regimen['Weight (g)'],cap_regimen['Tumor Volume (mm3)'])
# plt.plot(cap_regimen['Weight (g)'],y_values,color="red")
# plt.xlabel('Weight(g)')
# plt.ylabel('Average Tumore Volume (mm3)')

# plt.savefig('linearregression')
# plt.show()

```



In [ ]: