

Observations and Insights

- 1) The study had a near identical number of male and female mice
- 2) Based on final tumor volume both Capomulin and Ramicane appeared to be equally effective out of all four drugs
- 3) Out Of these four drugs only Infubinol had one outlier data point - the rest were "well behaved" with the datasets not having any outliers.
- 4) Based on data for Campomulin, the correlation between mouse weight and the average (at that weight) tumor volume is very strong.
- 5) All drugs were tested on almost same number of mice
- 6) Looking at the data for drug Capomulin on mouse s185, shows that over the time tumor volume had reduced indicating Capomulin can be effective drug.

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In [1]: # Dependencies and Setup

```
import matplotlib.pyplot as plt
import pandas as pd
import scipy.stats as st
import numpy as np

# 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)
study_results = pd.read_csv(study_results_path)

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

# Display the data table for preview
combined_data.head()
```

Out[1]:

	Mouse ID	Drug Regimen	Sex	Age_months	Weight (g)	Timepoint	Tumor Volume (mm3)	Metastatic Sites
0	k403	Ramicane	Male	21	16	0	45.000000	0
1	k403	Ramicane	Male	21	16	5	38.825898	0
2	k403	Ramicane	Male	21	16	10	35.014271	1
3	k403	Ramicane	Male	21	16	15	34.223992	1
4	k403	Ramicane	Male	21	16	20	32.997729	1

In [2]: # Checking the number of mice.

```
# Get the total number of records
total_records = combined_data["Mouse ID"].count()

# Get the total number of unique mice
unique_mice = combined_data["Mouse ID"].nunique()

# Display values
pd.DataFrame({"Count of all Records" :[total_records],
               "Count of unique mice" : [unique_mice]})
```

Out[2]:

	Count of all Records	Count of unique mice
0	1893	249

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

```
duplicates = combined_data[combined_data.duplicated(["Mouse ID", "Timepoint"])]
duplicates
```

Out[3]:

Mouse ID	Drug Regimen	Sex	Age_months	Weight (g)	Timepoint	Tumor Volume (mm3)	Metastatic Sites	
909	g989	Propriva	Female	21	26	0	45.000000	0
911	g989	Propriva	Female	21	26	5	47.570392	0
913	g989	Propriva	Female	21	26	10	49.880528	0
915	g989	Propriva	Female	21	26	15	53.442020	0
917	g989	Propriva	Female	21	26	20	54.657650	1

```
In [4]: # Optional: Get all the data for the duplicate mouse ID.
# Set the index of combined_data df to mouse_id
combined_data = combined_data.set_index("Mouse ID")

# Get all the data for mouse id got from above step 3
mouse = combined_data.loc["g989", :]

# Display data retrieved
mouse
```

Out[4]:

Drug Regimen	Sex	Age_months	Weight (g)	Timepoint	Tumor Volume (mm3)	Metastatic Sites
Mouse ID						
g989	Propriiva	Female	21	26	0	45.000000
g989	Propriiva	Female	21	26	0	45.000000
g989	Propriiva	Female	21	26	5	48.786801
g989	Propriiva	Female	21	26	5	47.570392
g989	Propriiva	Female	21	26	10	51.745156
g989	Propriiva	Female	21	26	10	49.880528
g989	Propriiva	Female	21	26	15	51.325852
g989	Propriiva	Female	21	26	15	53.442020
g989	Propriiva	Female	21	26	20	55.326122
g989	Propriiva	Female	21	26	20	54.657650
g989	Propriiva	Female	21	26	25	56.045564
g989	Propriiva	Female	21	26	30	59.082294
g989	Propriiva	Female	21	26	35	62.570880

```
In [5]: # Create a clean DataFrame by dropping the duplicate mouse by its ID.  
clean_mice_data = combined_data.drop("g989")  
  
# Reset the index of data frame  
clean_data = clean_mice_data.reset_index()  
  
# Display dataframe  
clean_data
```

Out[5]:

	Mouse ID	Drug Regimen	Sex	Age_months	Weight (g)	Timepoint	Tumor Volume (mm3)	Metastatic Sites
0	k403	Ramicane	Male	21	16	0	45.000000	0
1	k403	Ramicane	Male	21	16	5	38.825898	0
2	k403	Ramicane	Male	21	16	10	35.014271	1
3	k403	Ramicane	Male	21	16	15	34.223992	1
4	k403	Ramicane	Male	21	16	20	32.997729	1
...
1875	z969	Naftisol	Male	9	30	25	63.145652	2
1876	z969	Naftisol	Male	9	30	30	65.841013	3
1877	z969	Naftisol	Male	9	30	35	69.176246	4
1878	z969	Naftisol	Male	9	30	40	70.314904	4
1879	z969	Naftisol	Male	9	30	45	73.867845	4

1880 rows × 8 columns

```
In [6]: # Checking the number of mice in the clean DataFrame.  
# Get the total number of records (rows)  
new_total_records = clean_data["Mouse ID"].count()  
  
# Get the number of unique Mouse ID's  
new_unique_mice = clean_data["Mouse ID"].nunique()  
  
# Display retrieved values as data frame  
pd.DataFrame({"Count of all Records(New)": [new_total_records],  
              "Count of unique mice(New)": [new_unique_mice]})
```

Out[6]:

	Count of all Records(New)	Count of unique mice(New)
0	1880	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

```
# 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.

# Create a group by object by grouping on "Drug Regimen"
drug_grouped = clean_data.groupby("Drug Regimen")

# Perform statistical methods to calculate mean, median, variance, standard deviation, and SEM on tumor volume.
Tumor_Vol_Mean = round(drug_grouped["Tumor Volume (mm3)"].mean(),3)
Tumor_Vol_Median = round(drug_grouped["Tumor Volume (mm3)"].median(),3)
Tumor_Vol_Variance = round(drug_grouped["Tumor Volume (mm3)"].var(),3)
Tumor_Vol_StDev = round(drug_grouped["Tumor Volume (mm3)"].std(),3)
Tumor_Vol_SEM = round(drug_grouped["Tumor Volume (mm3)"].sem(),3)

# Create a data frame with values obtained above
summary_stat_df = pd.DataFrame({ "Tumor_Vol_Mean" : Tumor_Vol_Mean,
                                  "Tumor_Vol_Median" : Tumor_Vol_Median,
                                  "Tumor_Vol_Variance" : Tumor_Vol_Variance,
                                  "Tumor_Vol_StDev" : Tumor_Vol_StDev,
                                  "Tumor_Vol_SEM" : Tumor_Vol_SEM
                                })

# Reset the index of summary_df
summary_stat_df=summary_stat_df.reset_index()

# Display data of summary_df
summary_stat_df
```

Out[7]:

	Drug Regimen	Tumor_Vol_Mean	Tumor_Vol_Median	Tumor_Vol_Variance	Tumor_Vol_StDev	Tumor_Vol_SEM
0	Capomulin	40.676	41.558	24.948	4.995	0.329
1	Ceftamin	52.591	51.776	39.290	6.268	0.470
2	Infubinol	52.885	51.821	43.129	6.567	0.492
3	Ketapril	55.236	53.699	68.554	8.280	0.604
4	Naftisol	54.332	52.509	66.173	8.135	0.596
5	Placebo	54.034	52.289	61.168	7.821	0.581
6	Propriva	52.321	50.446	43.852	6.622	0.544

	Drug Regimen	Tumor_Vol_Mean	Tumor_Vol_Median	Tumor_Vol_Variance	Tumor_Vol_StDev	Tumor_Vol_SEM
7	Ramicane	40.217	40.673	23.487	4.846	0.321
8	Stelasyn	54.233	52.432	59.451	7.710	0.573
9	Zoniferol	53.237	51.818	48.533	6.967	0.516

```
In [8]: └─ # Generate a summary statistics table of mean, median, variance, standard deviation, and SEM of the tumor volume for each regimen
# Create a group by object by grouping on "Drug Regimen" and use aggregation method, produce the same summary statistics
drug_agg_grouped = clean_data.groupby("Drug Regimen").agg(
    Tumor_Vol_Mean=("Tumor Volume (mm3)", np.mean),
    Tumor_Vol_Median=("Tumor Volume (mm3)", np.median),
    Tumor_Vol_Variance=("Tumor Volume (mm3)", np.var),
    Tumor_Vol_StDev=("Tumor Volume (mm3)", np.std),
    Tumor_Vol_SEM=("Tumor Volume (mm3)", st.sem)
).round(3).reset_index()

# display values obtained above
drug_agg_grouped
```

Out[8]:

	Drug Regimen	Tumor_Vol_Mean	Tumor_Vol_Median	Tumor_Vol_Variance	Tumor_Vol_StDev	Tumor_Vol_SEM
0	Capomulin	40.676	41.558	24.948	4.995	0.329
1	Ceftamin	52.591	51.776	39.290	6.268	0.470
2	Infubinol	52.885	51.821	43.129	6.567	0.492
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6	Propriva	52.321	50.446	43.852	6.622	0.544
7	Ramicane	40.217	40.673	23.487	4.846	0.321
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Bar and Pie Charts

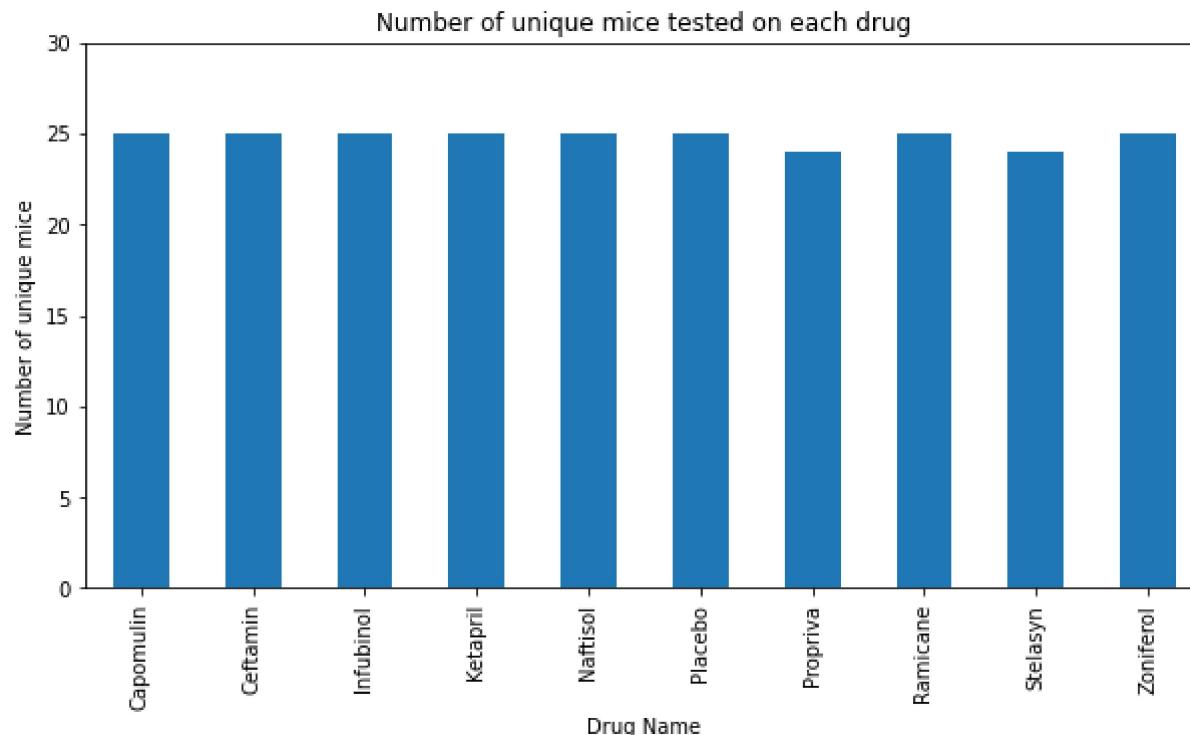
In [9]: # Generate a bar plot showing the total number of unique mice tested on each drug regimen using pandas.

```
# Groupby "Drug Regimen" on clean_data and get the unique mouse ID count per drug
drug_mice = clean_data.groupby("Drug Regimen")["Mouse ID"].nunique()
drug_mice = drug_mice.reset_index()

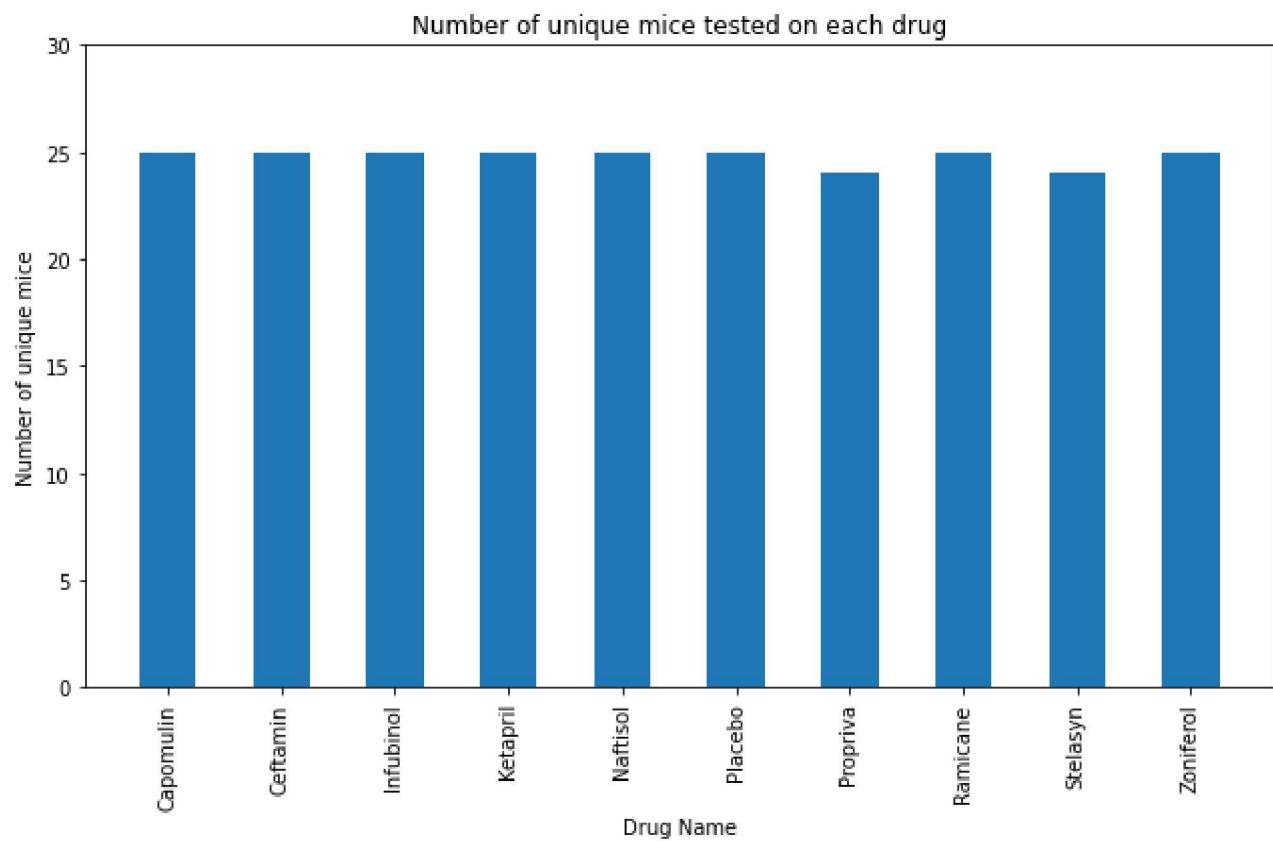
# plot bar graph on df obtained above
drug_mice_count_plot = drug_mice.plot(kind="bar", figsize=(10,5), title="Number of unique mice tested on each drug",
                                       xlabel = "Drug Name", ylabel = "Number of unique mice",
                                       ylim = (0, drug_mice["Mouse ID"].max()+5), legend=False)

# set xticks rotation to 90
drug_mice_count_plot.set_xticklabels(drug_mice["Drug Regimen"], rotation=90)

# Save plot
drug_mice_count_plot.get_figure().savefig("Plot_Images/Drug_Mice_Bar_df.png", bbox_inches="tight")
```



```
In [10]: # Generate a bar plot showing the total number of unique mice tested on each drug regimen using pyplot.  
# Set the figure size  
plt.figure(figsize=(9,6))  
  
# plot bar graph with drug on x-axis and Count on unique mose ID's on y-axis, with width set to 0.5  
plt.bar(drug_mice["Drug Regimen"], drug_mice["Mouse ID"], width=0.5)  
  
# Set x-ticks rotated 90 degree  
plt.xticks(drug_mice["Drug Regimen"], rotation=90)  
  
# Set title, x & y labels, y limits  
plt.title("Number of unique mice tested on each drug")  
plt.xlabel("Drug Name")  
plt.ylabel("Number of unique mice")  
plt.ylim(0, drug_mice["Mouse ID"].max()+5)  
plt.tight_layout()  
  
# Save plot image  
plt.savefig("Plot_Images/drug_mice_pyplot_Bar.png")  
  
# Display Plot  
plt.show()
```



In [11]: # Generate a pie plot showing the distribution of female versus male mice using pandas
First create a groupby object by grouping on the 'Sex' column
gender_stat = clean_data.groupby("Sex")

Create a dataframe counting the number of data points for each gender
gender_count = pd.DataFrame(gender_stat["Mouse ID"].nunique())

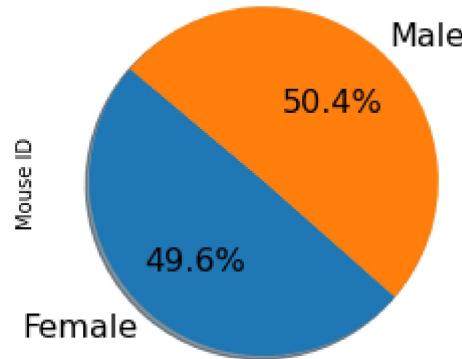
#Plot the graph

gender_mouse_plot = gender_count.plot(kind="pie", y="Mouse ID", autopct="%1.1f%%", shadow=True, startangle=140, legend=False, fontsize=16, title="Distribution of female versus male mice")

Save plot

gender_mouse_plot.get_figure().savefig("Plot_Images/Male_Female_Pie_df.png", bbox_inches="tight")

Distribution of female versus male mice



In [12]:

```
# Generate a pie plot showing the distribution of female versus male mice using pyplot
# Get the list of genders from the index values
genders = list(gender_count.index.values)

# Get the unique mouse id on gender
gen_count = gender_stat["Mouse ID"].nunique()

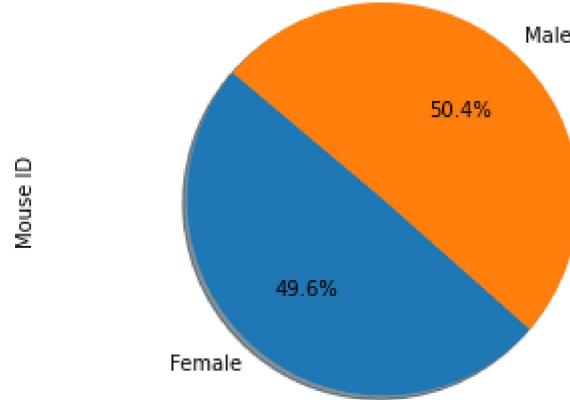
#Plot the graph
plt.pie(gen_count, labels = genders, autopct="%1.1f%%", shadow=True, startangle=140)

# Set the font size, title & ylabel
plt.rcParams['font.size'] = 16
plt.title("Distribution of female versus male mice")
plt.ylabel("Mouse ID")
plt.axis("equal")

# Save plot image
plt.savefig("Plot_Images/Male_Female_Pie_pyplot.png")

# Show plot
plt.show()
```

Distribution of female versus male mice



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
# Get the data for above drugs to another df
drug_list = ["Capomulin", "Ramicane", "Infubinol", "Ceftamin"]
four_drug_data = clean_data[clean_data["Drug Regimen"].isin(drug_list)]

# Start by getting the last (greatest) timepoint for each mouse
four_drug_data = four_drug_data.groupby(["Drug Regimen", "Mouse ID"])["Timepoint"].max()
four_drug_data = four_drug_data.reset_index()

# Merge this group df with the original dataframe to get the tumor volume at the last timepoint
top_drug = pd.merge(four_drug_data, clean_data[["Mouse ID", "Timepoint", "Tumor Volume (mm3)"]],
                     on=["Mouse ID", "Timepoint"], how="left")

# Display data frame created above
top_drug
```

Out[13]:

	Drug Regimen	Mouse ID	Timepoint	Tumor Volume (mm3)
0	Capomulin	b128	45	38.982878
1	Capomulin	b742	45	38.939633
2	Capomulin	f966	20	30.485985
3	Capomulin	g288	45	37.074024
4	Capomulin	g316	45	40.159220
...
95	Ramicane	s508	45	30.276232
96	Ramicane	u196	45	40.667713
97	Ramicane	w678	5	43.166373
98	Ramicane	y449	15	44.183451
99	Ramicane	z578	45	30.638696

100 rows × 4 columns

In [14]: # Put treatments into a list for for loop (and later for plot labels)

```
# Create list of empty list to fill with tumor vol data (for plotting)
# drug list declared above is as below with their indices:
# 0 - Capomulin , so add tumor volume with this drug at index 0 of Tumor Vol List tumor_vol
# 1 - Ramicane , so add tumor volume with this drug at index 1 of Tumor Vol List tumor_vol
# 2 - Infubinol , so add tumor volume with this drug at index 2 of Tumor Vol List tumor_vol
# 3 - Ceftamin , so add tumor volume with this drug at index 3 of Tumor Vol List tumor_vol

tumor_vol = [[] for i in range(len(drug_list))]

# Iterate through a datafram created before whcih contains data for four drungs
for i in range (len(top_drug)):
    # With iloc methd get the drug name & tumor volume from each row
    drug = top_drug.iloc[i,:]["Drug Regimen"]
    t_vol = round(top_drug.iloc[i,:]["Tumor Volume (mm3)"],3)

    # Iterate through a list of four drugs with index
    for j in range(len(drug_list)):

        # If drug name from list matches with the drug from data frame,
        # add tumor volume to list with same index inside a list
        if drug == drug_list[j]:
            tumor_vol[j].append(t_vol)

# 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 volumes
# add subset
# Determine outliers using upper and lower bounds

counter = 0 # counter is declared to track the first iteration

# Iterate through List of four drugs
for drug in drug_list:
    # From drug df , if drug name matches the one from list, add tumor volume to df
    drug_df = top_drug.loc[top_drug["Drug Regimen"]== drug,"Tumor Volume (mm3)"]

    # Get the quartiles, Q1, Q3 & IQR of tumor volume for specific drug
    # Calculate lowe & upper bound to get potential outliers
    quartiles = drug_df.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 the data received. Counter is used to print the dotted line only on top & increment counter
if counter == 0:
    print("-----")
    counter +=1
print(f"The lower quartile of {drug} tumor volume is: {lowerq}")
print(f"The upper quartile of {drug} tumor volume is: {upperq}")
print(f"The interquartuile range of {drug} tumor volume is: {iqr}")
print(f"The median of {drug} tumor volume is: {quartiles[0.5]}")
print(f"Values below {lower_bound} could be outliers for {drug} tumor volume")
print(f"Values above {upper_bound} could be outliers for {drug} tumor volume")
print("-----")
```

```
-----
The lower quartile of Capomulin tumor volume is: 32.37735684
The upper quartile of Capomulin tumor volume is: 40.1592203
The interquartuile range of Capomulin tumor volume is: 7.781863460000004
The median of Capomulin tumor volume is: 38.12516439999996
Values below 20.70456164999999 could be outliers for Capomulin tumor volume
Values above 51.83201549 could be outliers for Capomulin tumor volume
-----
```

```
The lower quartile of Ramicane tumor volume is: 31.56046955
The upper quartile of Ramicane tumor volume is: 40.65900627
The interquartuile range of Ramicane tumor volume is: 9.09853671999998
The median of Ramicane tumor volume is: 36.56165229
Values below 17.91266447000003 could be outliers for Ramicane tumor volume
Values above 54.30681135 could be outliers for Ramicane tumor volume
-----
```

```
The lower quartile of Infubinol tumor volume is: 54.04860769
The upper quartile of Infubinol tumor volume is: 65.52574285
The interquartuile range of Infubinol tumor volume is: 11.47713516000003
The median of Infubinol tumor volume is: 60.16518046
Values below 36.83290494999999 could be outliers for Infubinol tumor volume
Values above 82.74144559000001 could be outliers for Infubinol tumor volume
-----
```

```
The lower quartile of Ceftamin tumor volume is: 48.72207785
The upper quartile of Ceftamin tumor volume is: 64.29983003
The interquartuile range of Ceftamin tumor volume is: 15.57775217999997
The median of Ceftamin tumor volume is: 59.85195552
Values below 25.35544958000002 could be outliers for Ceftamin tumor volume
Values above 87.66645829999999 could be outliers for Ceftamin tumor volume
-----
```

```
In [15]: # Generate a box plot of the final tumor volume of each mouse across four regimens of interest
# To plot boxplots for all four drugs in same figure, declare axes as array with one row and four columns,
fig1,axs = plt.subplots(nrows=1, ncols=4, sharey=True, figsize=(10,6))

#Set title for the figure
fig1.suptitle('Final Tumor volume across four regimens', fontsize=18, fontweight='bold')

# Set Y Label
axs[0].set_ylabel('Volume (mm3)')

# Set the title, draw boxplots with drug & tumor volume List and set xticks to empty list
# ax[0] is for drug Capomulin
axs[0].set_title(drug_list[0])
axs[0].boxplot(tumor_vol[0])
axs[0].set_xticks([])

# ax[1] is for drug Ramicane
axs[1].set_title(drug_list[1])
axs[1].boxplot(tumor_vol[1])
axs[1].set_xticks([])

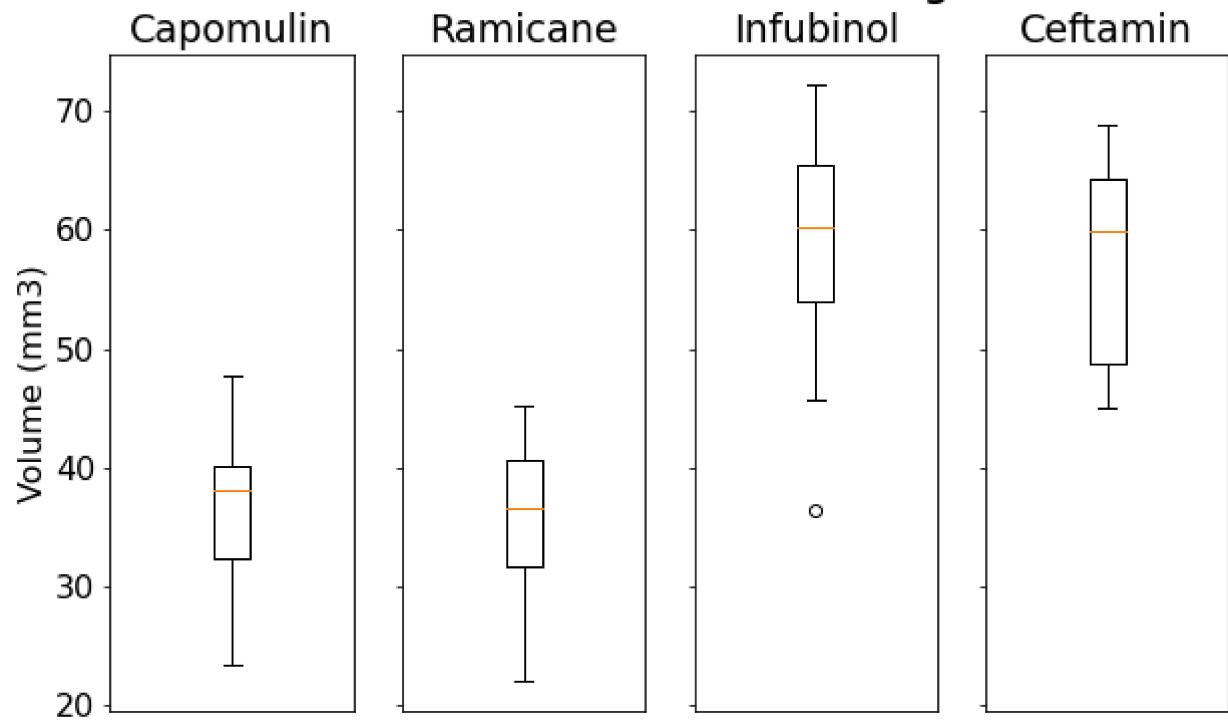
# ax[2] is for drug Infubinol
axs[2].set_title(drug_list[2])
axs[2].boxplot(tumor_vol[2])
axs[2].set_xticks([])

# ax[3] is for drug Ceftamin
axs[3].set_title(drug_list[3])
axs[3].boxplot(tumor_vol[3])
axs[3].set_xticks([])

# Save Plot Image
fig1.savefig("Plot_Images/Boxplot.png")

# Display plot
plt.show()
```

Final Tumor volume across four regimens



Line and Scatter Plots

In [16]: # Generate a Line plot of tumor volume vs. time point for a mouse treated with Capomulin
get the data for drug Capomulin from clean_data
Capomulin_data = clean_data.loc[clean_data["Drug Regimen"] == "Capomulin",:]

Display A DataFrame & get any mouse id
Capomulin_data

Get all the data for Mouse ID using Loc method from Capomulin data
single_mouse_data = Capomulin_data.loc[Capomulin_data["Mouse ID"] == "s185",:]

Set the figure size
plt.figure(figsize=(10,5))

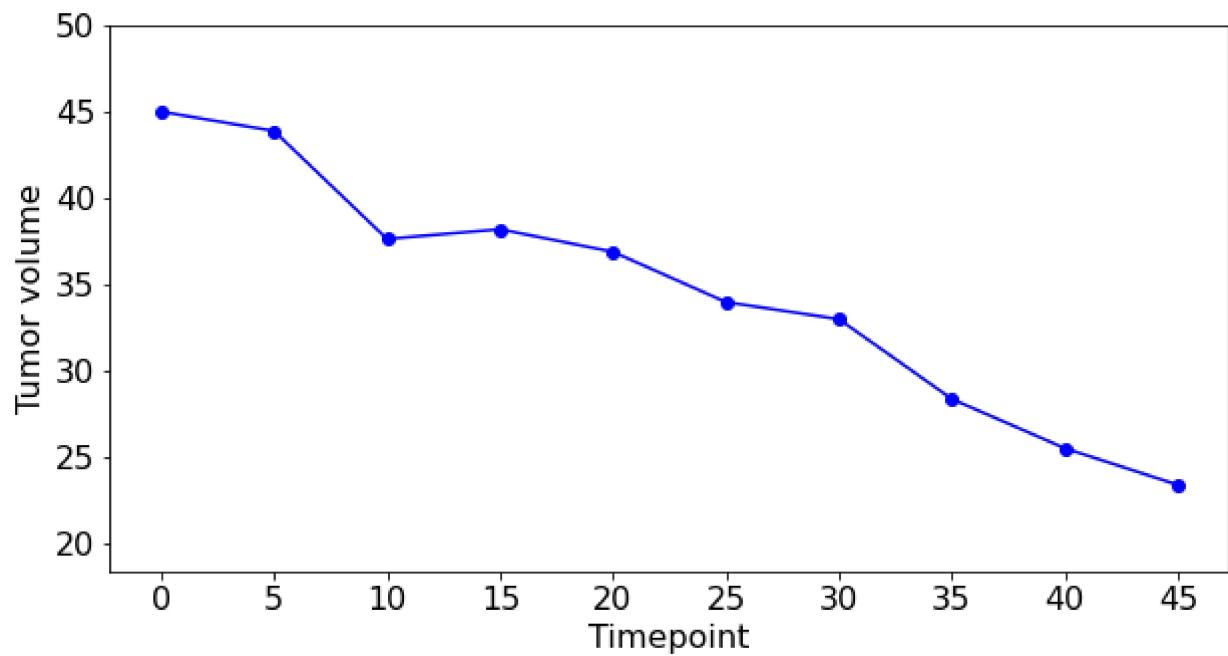
Plot the line graph with Timepoint as x axis & Tumor Volume as Y axis from Signe Mouse data
plt.plot(single_mouse_data["Timepoint"], single_mouse_data["Tumor Volume (mm3)"], marker = 'o', color='blue')

Set the title , x & T Label, ylim & xticks
plt.title("Tumor Volume vs. Time Point for mouse s185", pad=25)
plt.xlabel("Timepoint")
plt.ylabel("Tumor volume")
plt.ylim((single_mouse_data["Tumor Volume (mm3)"].min()) - 5,(single_mouse_data["Tumor Volume (mm3)"].max()) + 5)
plt.xticks(np.arange(0,50,5))

Save plot image
plt.savefig("Plot_Images/sigle_mouse_line_graph.png")

Display plot
plt.show()

Tumor Volume vs. Time Point for mouse s185



```
In [17]: # Generate a scatter plot of average tumor volume vs. mouse weight for the Capomulin regimen
# Perform a groupby operation on capomulin drug data on Mouse ID & weight and get the average tumor volume
weight_vol_data = Capomulin_data.groupby(["Mouse ID", "Weight (g)"]).agg(Avg_Tumor_Vol=("Tumor Volume (mm3)", np.mean))
weight_vol_data = weight_vol_data.reset_index()

# set the figure size
plt.figure(figsize=(10,5))

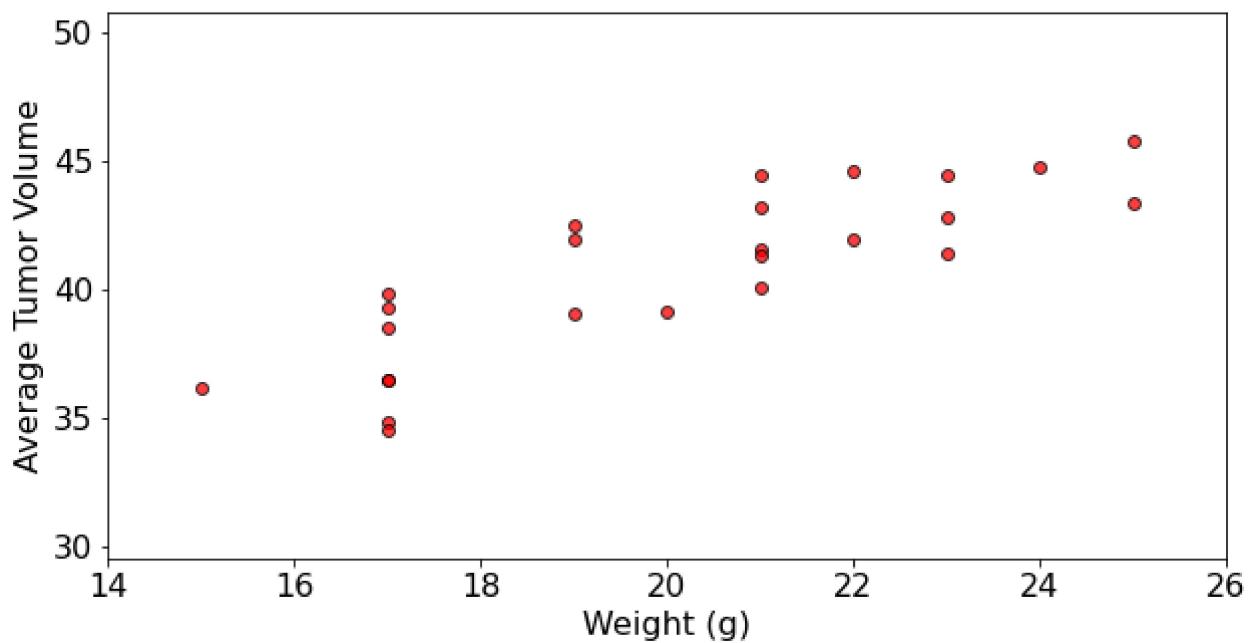
# plot scatter plot with weight as x axis & avg tumor volume as y-axis.
plt.scatter(weight_vol_data["Weight (g)"], weight_vol_data["Avg_Tumor_Vol"], marker="o", facecolors="red",
            edgecolors="black", alpha=0.75)

# Set the title, xlabel, ylabel, xlim & ylim
plt.title("Mouse Weight vs Average Tumor Volume", pad=25)
plt.xlabel("Weight (g)")
plt.ylabel("Average Tumor Volume")
plt.ylim((weight_vol_data["Avg_Tumor_Vol"].min()) - 5,(weight_vol_data["Avg_Tumor_Vol"].max()) + 5)
plt.xlim(14,26)

# Save plot image
plt.savefig("Plot_Images/weight_Avg_tumorvol_scatter.png")

# Display plot
plt.show()
```

Mouse Weight vs Average Tumor Volume



Correlation and Regression

```
In [18]: # Calculate the correlation coefficient and Linear regression model
# for mouse weight and average tumor volume for the Capomulin regimen

# Get the linear equation
(slope, intercept, rvalue, pvalue, stderr) = st.linregress(weight_vol_data["Weight (g)"], weight_vol_data["Avg_Tumor_Vol"])
regress_values = weight_vol_data["Weight (g)"] * slope + intercept
line_eq = "y = " + str(round(slope,2)) + "x + " + str(round(intercept,2))

# Plot the scatter plot
plt.scatter(weight_vol_data["Weight (g)"],weight_vol_data["Avg_Tumor_Vol"])
plt.plot(weight_vol_data["Weight (g)"],regress_values,"-y")

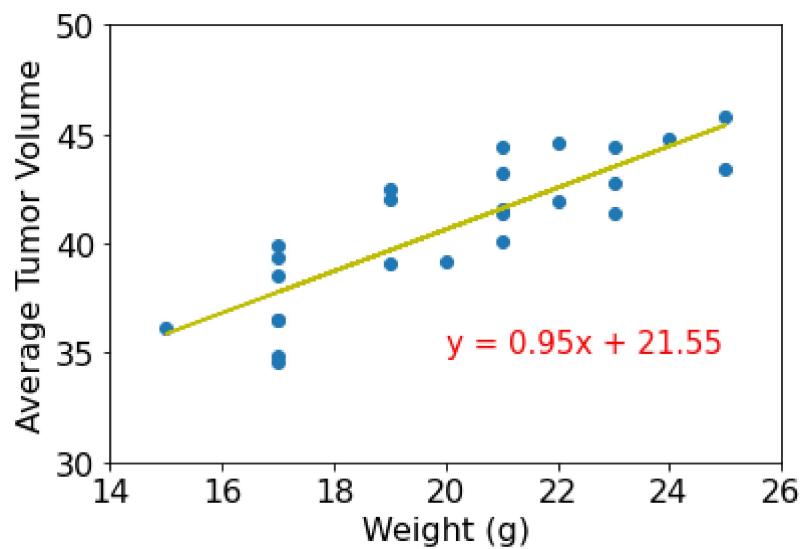
# Annotate the line_equation on graph
plt.annotate(line_eq,(20,35),fontsize=15,color="red")

# Set label & limits for x & y axis
plt.xlabel("Weight (g)")
plt.ylabel("Average Tumor Volume")
plt.xlim(14,26)
plt.ylim(30,50)

# save plot image
plt.savefig("Plot_Images/linear_equation_plot.png")

# Display a plot
plt.show()

# Calculate correlation & display it
correlation = st.pearsonr(weight_vol_data["Weight (g)"], weight_vol_data["Avg_Tumor_Vol"])
print(f'r squared value is : {rvalue ** 2}')
print(f"The correlation coefficient between weight and average final tumor volume is {round(correlation[0],2)}")
```



r squared value is : 0.7088568047708717

The correlation coefficient between weight and average final tumor volume is 0.84

In []: