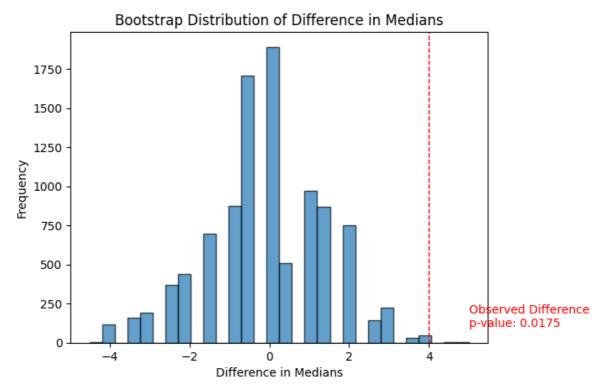
Hariharan Kalimuthu, 2020115015

19.02.2024, In Class Assignment

Question 2 (a)

```
In [ ]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        # Assuming your data is already loaded into a DataFrame called df
        df = pd.read excel('data.xlsx', sheet name=0)
        group1 median = df[df['Group'] == 'Beer']['No. of Mosquitoes'].median()
        group2_median = df[df['Group'] == 'Water']['No. of Mosquitoes'].median()
        observed_difference = group1_median - group2_median
        # Pool the data from both groups
        pooled_data = df['No. of Mosquitoes'].values
        # Perform permutation test
        perm_diffs = []
        for in range(10000):
            np.random.shuffle(pooled_data)
            perm_group1_median = np.median(pooled_data[:len(df[df['Group'] == 'Be
            perm_group2_median = np.median(pooled_data[len(df[df['Group'] == 'Bee')
            perm_diffs.append(perm_group1_median - perm_group2_median)
        # Calculate p-value
        p_value = np.mean(np.abs(perm_diffs) >= np.abs(observed_difference))
        # Plot histogram
        plt.hist(perm_diffs, bins=30, edgecolor='k', alpha=0.7)
        plt.axvline(observed_difference, color='red', linestyle='dashed', linewid
        plt.xlabel('Difference in Medians')
        plt.ylabel('Frequency')
        plt.title('Bootstrap Distribution of Difference in Medians')
        plt.text(observed_difference + 1, 100, f'Observed Difference\np-value: {p
        # plot the two medians as well
        plt.show()
        print("Observed Difference in Medians:", observed_difference)
        print("p-value:", p_value)
```

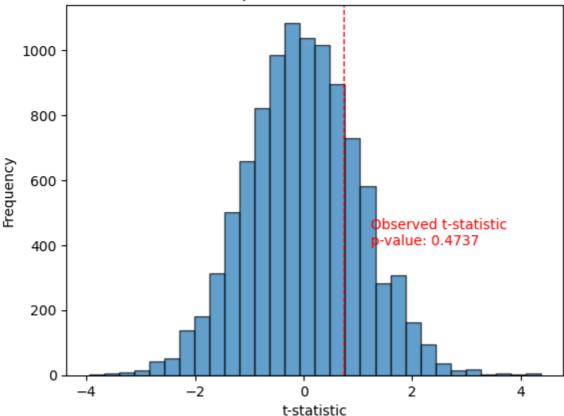


Observed Difference in Medians: 4.0 p-value: 0.0175

Question 2 (b)

```
In [ ]: # Calculate the observed t-statistic
        group1_data = df[df['Group'] == 'Beer']['No. of Mosquitoes']
        group2_data = df[df['Group'] == 'Water']['No. of Mosquitoes']
        observed_t_statistic, p_value = ttest_ind(group1_data, group2_data)
        # Perform permutation test
        perm_t_stats = []
        for _ in range(10000):
            np.random.shuffle(pooled_data)
            perm_group1_data = pooled_data[:len(group1_data)]
            perm_group2_data = pooled_data[len(group1_data):]
            perm_t_statistic, _ = ttest_ind(perm_group1_data, perm_group2_data)
            perm_t_stats.append(perm_t_statistic)
        # Calculate p-value
        p_value = np.mean(np.abs(perm_t_stats) >= np.abs(observed_t_statistic))
        # Plot histogram
        plt.hist(perm_t_stats, bins=30, edgecolor='k', alpha=0.7)
        plt.axvline(observed_t_statistic, color='red', linestyle='dashed', linewi
        plt.xlabel('t-statistic')
        plt.ylabel('Frequency')
        plt.title('Bootstrap Distribution of t-statistic')
        plt.text(observed_t_statistic + 0.5, 400, f'Observed t-statistic\np-value
        plt.show()
        print("Observed t-statistic:", observed_t_statistic)
        print("p-value:", p_value)
```

Bootstrap Distribution of t-statistic



Observed t-statistic: 0.743632120058524

p-value: 0.4737

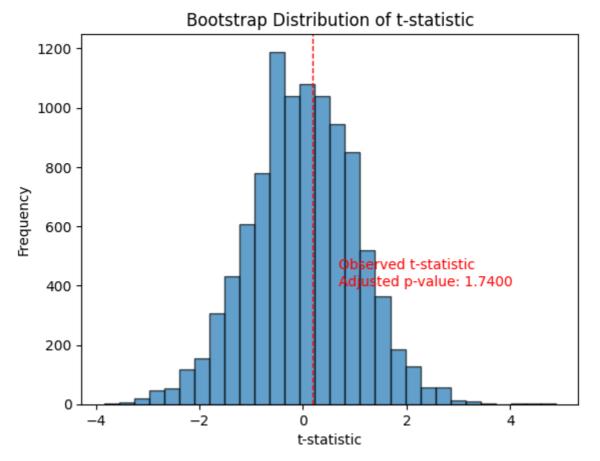
Question 2(c)

To calculate the significance values for a non-directional alternative hypothesis (suggesting that there will be a difference in groups without specifying the direction), you need to consider both tails of the distribution. This means doubling the probability obtained from the permutation test. Here's how you can adjust the code to calculate the new significance values:

```
In [ ]: # Calculate the observed t-statistic
        group1_data = df[df['Group'] == 'Beer']['No. of Mosquitoes']
        group2_data = df[df['Group'] == 'Water']['No. of Mosquitoes']
        observed_t_statistic, p_value = ttest_ind(group1_data, group2_data)
        # Perform permutation test
        perm_t_stats = []
        for _ in range(10000):
            np.random.shuffle(pooled_data)
            perm_group1_data = pooled_data[:len(group1_data)]
            perm_group2_data = pooled_data[len(group1_data):]
            perm_t_statistic, _ = ttest_ind(perm_group1_data, perm_group2_data)
            perm_t_stats.append(perm_t_statistic)
        # Calculate p-value
        p_value = np.mean(np.abs(perm_t_stats) >= np.abs(observed_t_statistic))
        # Adjust p-value for non-directional hypothesis
        p_value *= 2
```

```
# Plot histogram
plt.hist(perm_t_stats, bins=30, edgecolor='k', alpha=0.7)
plt.axvline(observed_t_statistic, color='red', linestyle='dashed', linewi
plt.xlabel('t-statistic')
plt.ylabel('Frequency')
plt.title('Bootstrap Distribution of t-statistic')
plt.text(observed_t_statistic + 0.5, 400, f'Observed t-statistic\nAdjuste
plt.show()

print("Observed t-statistic:", observed_t_statistic)
print("Adjusted p-value:", p_value)
```



Observed t-statistic: 0.19229903954222835

Adjusted p-value: 1.74

Question 3

```
In []: import numpy as np
import pandas as pd

df = pd.read_excel('data2.xlsx', sheet_name=0)

# Calculate the observed correlation coefficient
observed_corr = df['IQ'].corr(df['TestScore'])

#display first 5 rows of the dataframe
print(df.head())

# Perform permutation test
perm_corrs = []
```

```
for _ in range(10000):
    np.random.shuffle(df['TestScore'].values)
    perm_corr = df['IQ'].corr(df['TestScore'])
    perm_corrs.append(perm_corr)

# Calculate p-value
p_value = np.mean(np.abs(perm_corrs) >= np.abs(observed_corr))

print("Observed Correlation Coefficient:", observed_corr)
print("p-value:", p_value if p_value > 1e-10 else "< 1e-10")

if p_value < 0.05:
    print("Reject the null hypothesis that there exists no correlation be else:
    print("Accept the null hypothesis that there exists no correlation be</pre>
```

```
IQ TestScore
0 111 67
1 107 43
2 100 52
3 107 66
4 114 58
```

Observed Correlation Coefficient: 0.4931479069133802

p-value: < 1e-10

Reject the null hypothesis that there exists no correlation between IQ and testscores.