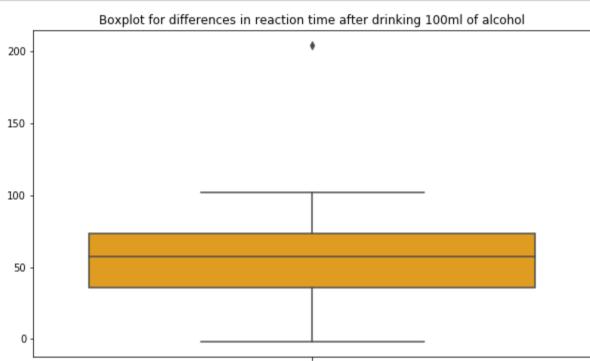
```
In [48]: import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
          from scipy import stats
          import math
In [49]: data = pd.read_csv("L2_data_group3.txt", delimiter="\t")
In [50]: data.describe()
Out[50]:
                    Before
                               After
                 50.000000
                           50.000000
           count
           mean 198.082600 253.613200
             std 26.693944 44.167379
            min 132.040000 163.320000
            25% 177.687500 227.572500
            50% 196.485000 247.600000
            75% 222.032500 267.550000
            max 250.970000 443.140000
In [51]: data.head()
Out[51]:
             Before After
           0 171.14 242.95
           1 191.22 216.94
           2 207.76 265.78
           3 165.44 163.32
           4 205.87 213.53
```

#### Simple plot to visualize outliers

```
In [52]: data['differences'] = data["After"]-data['Before']
         plt.figure(figsize=(10,6))
         sns.boxplot(y=differences, color='orange')
         plt.title("Boxplot for differences in reaction time after drinking 100ml of alcohol")
         plt.show()
```



Based on this boxplot we can see that outlied sample has got difference in reaction time in around ~200 ms. Now we can perform Tukey test to detect this and maybe other outlied samples.

### **Tukey test**

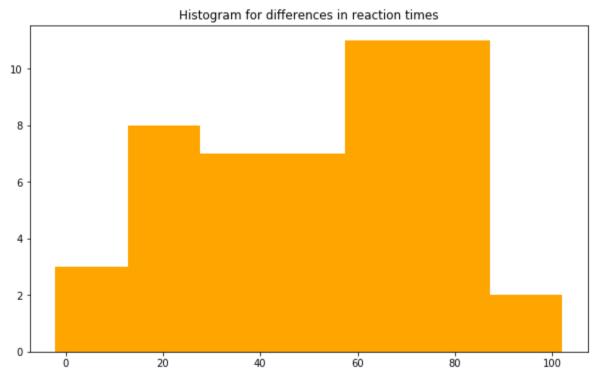
```
In [55]: def tukey_outliers(df, column):
           Q1 = np.percentile(df[column], 25)
           Q3 = np.percentile(df[column], 75)
           IQR = Q3 - Q1
           outlier_step = 1.5 * IQR
           outliers_index = df[(df[column] < Q1 - outlier_step) | (df[column] > Q3 + outlier_step)].i
In [36]: outliers_index = tukey_outliers(data, 'differences')
In [56]: len(outliers_index)
Out[56]: 1
```

Tukey test found only one outlier in considered dataset. Now we can drop this outlied sample.

```
In [58]: data.drop(outliers_index, inplace=True)
```

## Histogram

```
In [59]: plt.figure(figsize=(10,6))
         x = len(data["differences"])
         binsizes = math.sqrt(x)
         plt.hist(data["differences"], bins = int(binsizes),color='orange')
         plt.title('Histogram for differences in reaction times')
         plt.show()
```



The data is not normally distributed.

# Hypothesis and paired sampled t-test

- H0: There is no difference in reaction time before and after drinking 100ml of alcohol in population of 50 drivers.
- H0: u1 == u2
- Ha: There is a difference in reaction time before and after drinking 100ml of alcohol in population of 50 drivers.
- Ha: u1 != u2

```
In [46]: | ttest, pval = stats.ttest_rel(data['Before'], data['After'])
         print(pval)
         if pval<0.05:
             print("The p value is lower than p < 0.05, so we can reject the null hypothesis.")
             print("The p value is not lower than p < 0.05, so we can not reject the null hypothesi
         1.94533467005807e-19
         The p value is lower than p < 0.05, so we can reject the null hypothesis.
```

consuming 100ml of alcohol for the population under study.

**Conclusion** Based on p value obtained in paired sampled t-test, we can assume that there is a difference in reaction times after