

Exercise 4

A paper published by *Quality Engineering* reported a dataset that consists of loading weights (in grams) of insecticide tanks. Data are reported in the file `ESE7_ex3.csv`.

1. Design a control chart for the mean. Which conclusions can we draw about the process?
2. Consider the sample means as individual measurements: design a control chart for the mean.
3. Compare the results at point 1 and 2. What can we conclude?

```
In [ ]: # Import the necessary libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from scipy import stats
import qda

# Import the dataset
data = pd.read_csv('ESE7_ex3.csv')

# Inspect the dataset
data.head()
```

```
Out[ ]:    x1  x2  x3  x4
0  456  458  439  448
1  459  462  495  500
2  443  453  457  458
3  470  450  478  470
4  457  456  460  457
```

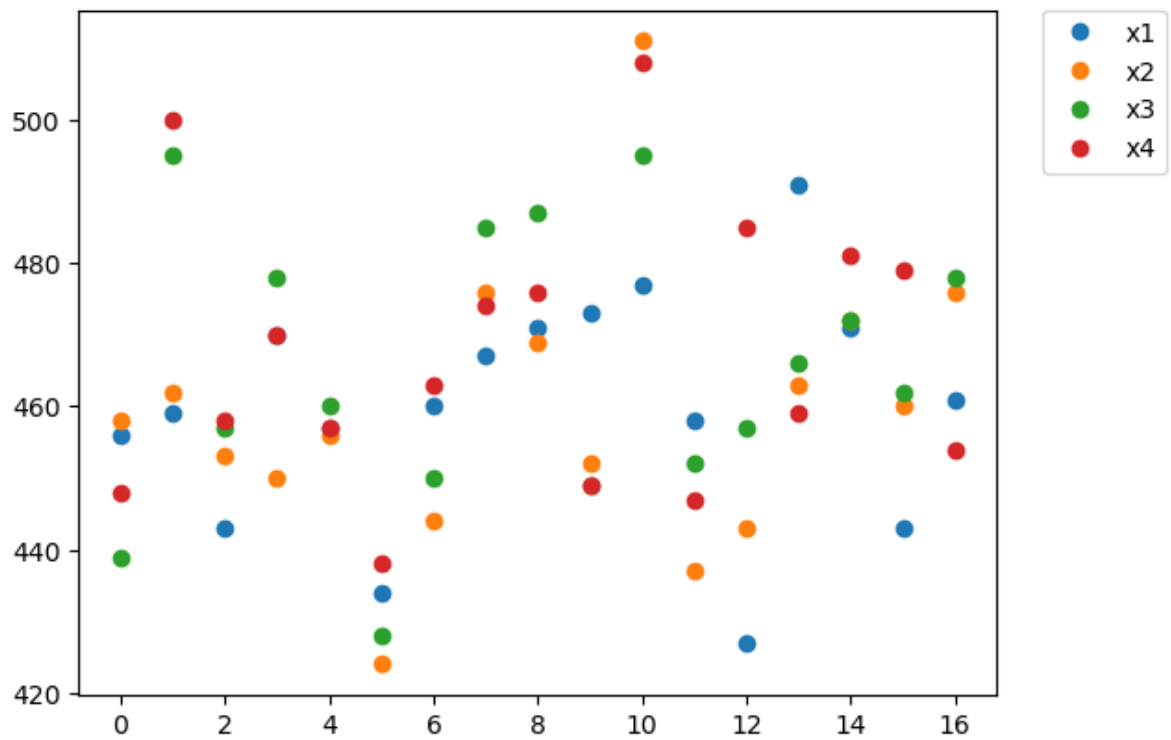
Point 1

Design a control chart for the mean. Which conclusions can we draw about the process?

Solution

Inspect the data by plotting the individual datapoints.

```
In [ ]: # Make a scatter plot of all the columns against the index
plt.plot(data['x1'], linestyle='none', marker='o', label = 'x1')
plt.plot(data['x2'], linestyle='none', marker='o', label = 'x2')
plt.plot(data['x3'], linestyle='none', marker='o', label = 'x3')
plt.plot(data['x4'], linestyle='none', marker='o', label = 'x4')
# place the legend outside the plot
plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
plt.show()
```



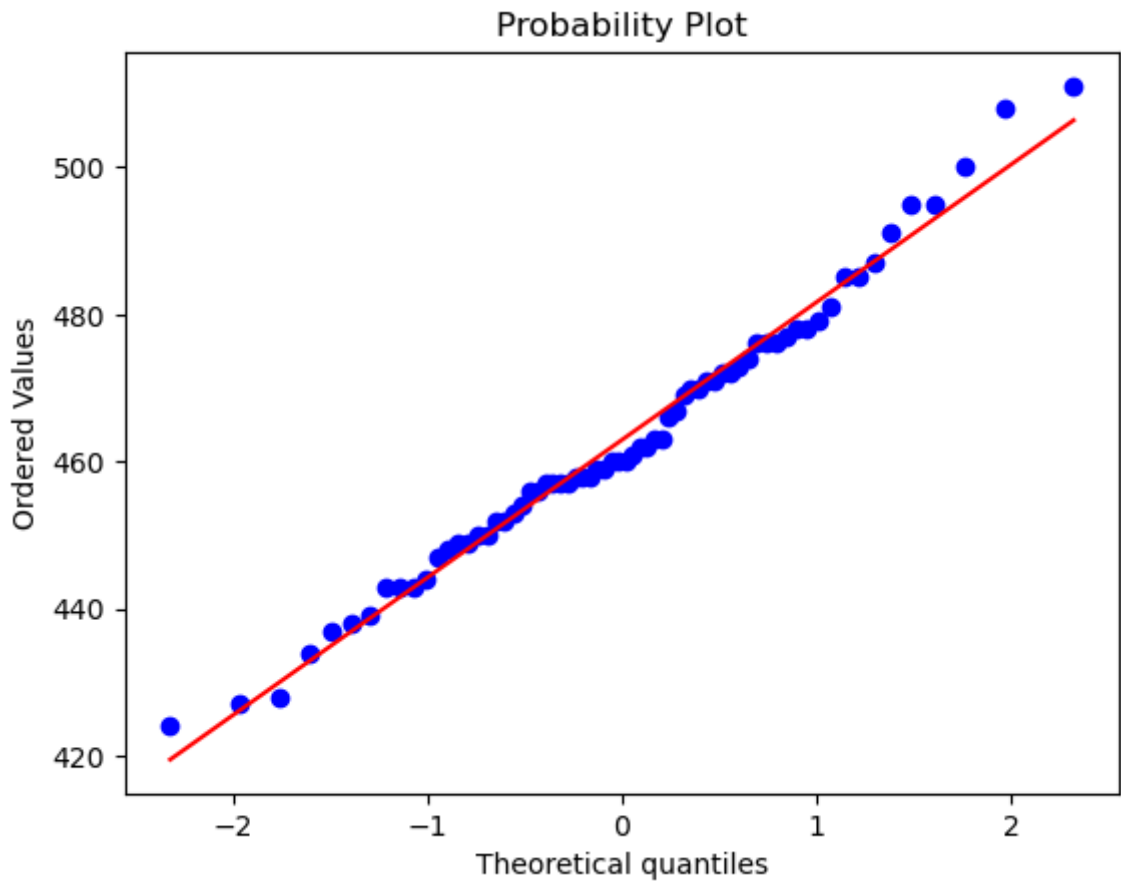
Check if the data are normal.

```
In [ ]: # Stack the columns on top of each other
data_stack = data.stack()

# Perform the Shapiro-Wilk test
_, pval_SW = stats.shapiro(data_stack)
print('Shapiro-Wilk test p-value = %.3f' % pval_SW)

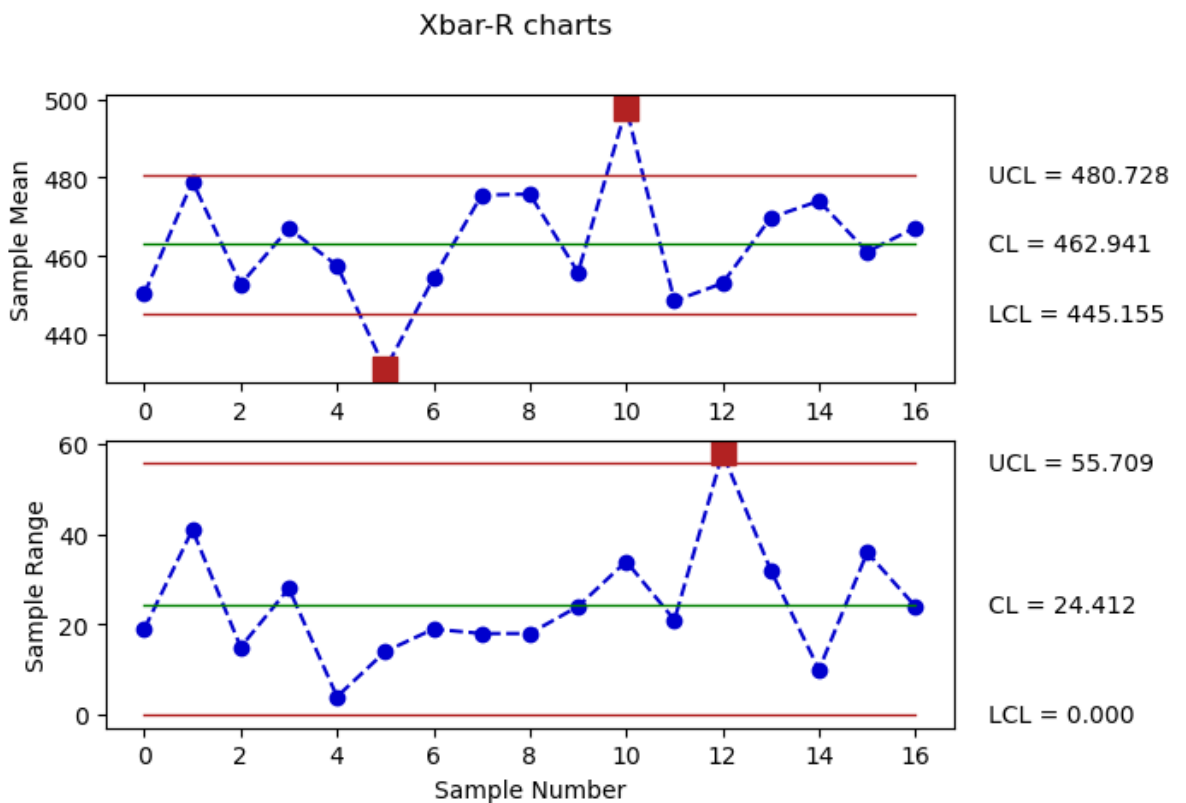
# Plot the qqplot
stats.probplot(data_stack, dist="norm", plot=plt)
plt.show()
```

Shapiro-Wilk test p-value = 0.593



Let's design the Xbar-R chart.

```
In [ ]: data_XR = qda.ControlCharts.XbarR(data)
```



The control limits of the Xbar chart look to narrow with respect to the natural variability of the statistic. This can be caused by a violation of assumptions (independence) within the

sample. Thus, the Xbar-R control chart may be not appropriate to monitor these data.

Point 2

Consider the sample means as individual measurements: design a control chart for the mean.

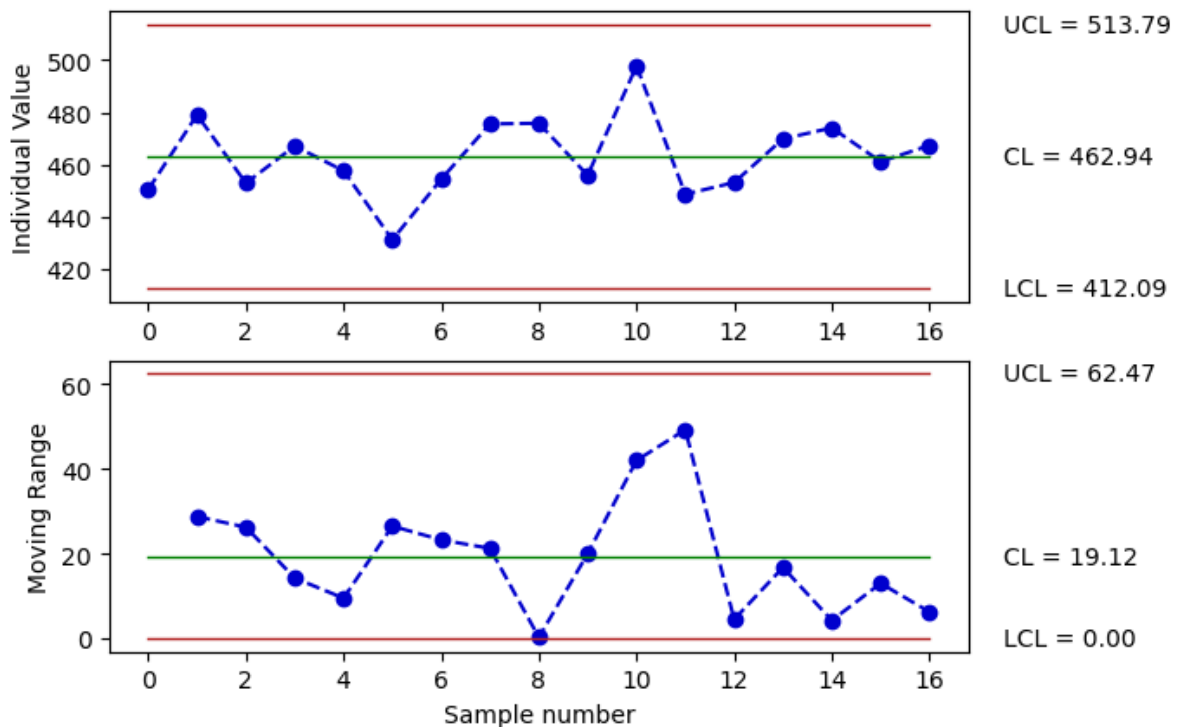
Solution

Design a "Between groups" control chart, i.e., a chart that assumes all the samples to be individual measurements.

```
In [ ]: # Create a new dataframe that stores the mean of all the samples
data_Xbar = pd.DataFrame(data_XR['sample_mean'])

# Build the IMR chart using this new dataframe
data_Xbar = qda.ControlCharts.IMR(data_Xbar, 'sample_mean')
```

I-MR charts of sample_mean



With the I chart we can get rid of the violation of the independence assumption within the sample. The MR chart allows monitoring the between sample variability, while the R chart designed before can still be used to monitor the within sample variability. A control charting scheme quite effective in this case is the so-called I-MR-R control chart (or I-MR-S, if the S chart is used in place of the R chart), where: 1) the I chart allows monitoring the mean of the process treating sample means as individual observations; 2) the MR chart allows monitoring the between sample variability; 3) the MR chart allows monitoring the within sample variability.

```
In [ ]: # Design a I-MR-R control chart

# Build the IMR chart using this new dataframe
```

```

data_Xbar = qda.ControlCharts.IMR(data_Xbar, 'sample_mean')

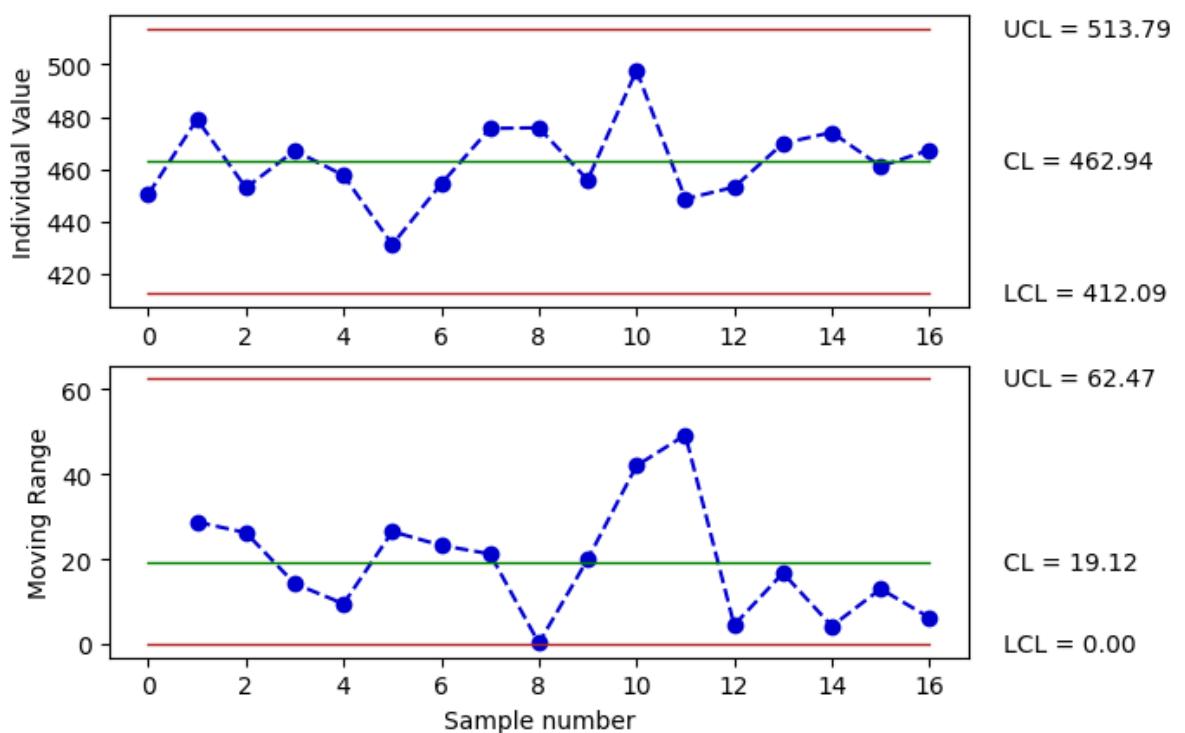
# Plot the R chart as well
plt.title('R chart')
plt.plot(data_XR['sample_range'], color='b', linestyle='--', marker='o')
plt.plot(data_XR['R_UCL'], color='r')
plt.plot(data_XR['R_CL'], color='g')
plt.plot(data_XR['R_LCL'], color='r')
plt.ylabel('Sample range')
plt.xlabel('Sample number')

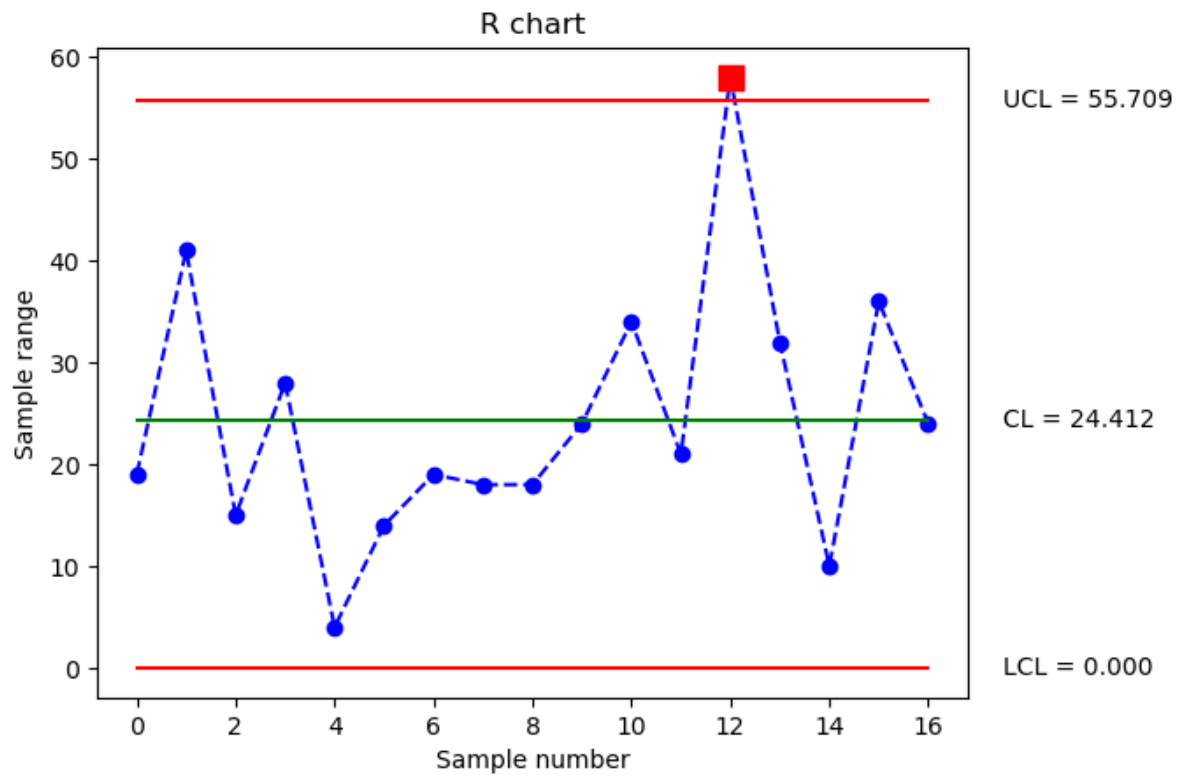
# add the values of the control limits on the right side of the plot
plt.text(len(data_XR)+.5, data_XR['R_UCL'].iloc[0], 'UCL = {:.3f}'.format(data_XR['R_UCL'].iloc[0]))
plt.text(len(data_XR)+.5, data_XR['R_CL'].iloc[0], 'CL = {:.3f}'.format(data_XR['R_CL'].iloc[0]))
plt.text(len(data_XR)+.5, data_XR['R_LCL'].iloc[0], 'LCL = {:.3f}'.format(data_XR['R_LCL'].iloc[0]))

# highlight the points that violate the alarm rules
plt.plot(data_XR['R_TEST1'], linestyle='none', marker='s', color='r', markersize=10)
plt.show()

```

I-MR charts of sample_mean





There is an alarm in the 13th sample. A search for assignable causes shall be performed. In the absence of information about possible assignable causes, the alarm can be labelled as a false alarm, and the design phase of the control chart is over.