

# STAC33-TUT5

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## Introduction

This week's lecture we will be discussing topics from CH:9, Ch:10 and Ch:12. Chapter 9 is about **Sign test** and Chapter 10 is about **Mood Median test** and Chapter 12 is about **QQ-plot**(aka normal quantile plots). All the problems being discussed can be found on the PASIAS here

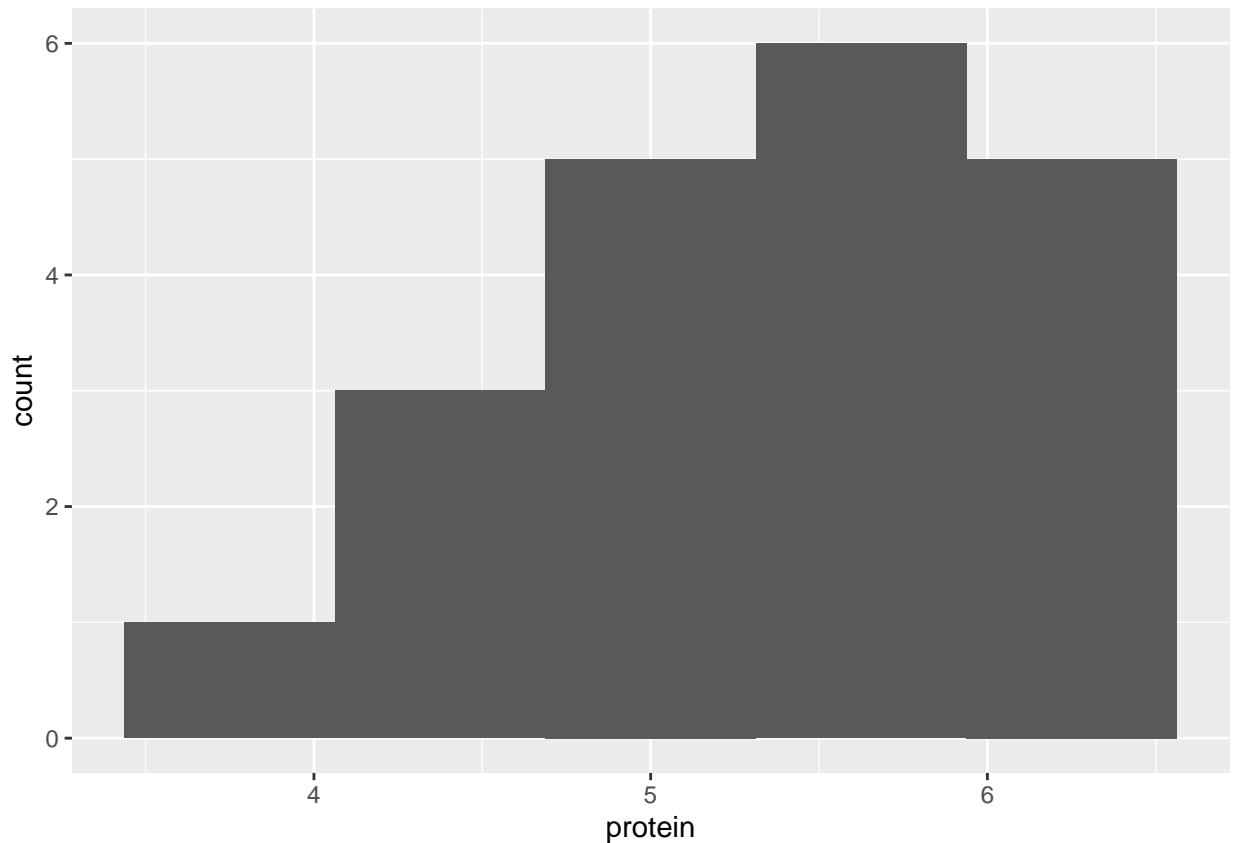
### Question 9.10

Data is about amount of protein in the package. The advertisement claims to have **6 ounces/pack**(our assumption about the package  $H_0$ ). Goal is to test this claim by sampling( $n=20$ ) and conducting some tests.

**a + b. Read the data and make suitable plot**

```
# a
my_url <- "http://ritsokiguess.site/datafiles/protein.txt"
## Data is not seperated by a space/chracter or tab so this works
meals <- read_table(my_url)

# b
## 1 quantitative variable(Can make box-plot/histogram)
## Interested in distribution rather than range so histogram
ggplot(meals, aes(x=protein)) + geom_histogram(bins = 5)
```



Note we can observe?

- (Not symmetric)
- Skewed left

**c. Why might a sign test be better than a t-test for assessing the average amount of protein per package? Explain briefly. (“Average” here means any measure of centre.)**

Assumptions for t-test:

- Normality  
Here we can see the assumption is violated(as not normal from Histogram).
- We testing  $\mu$  in t-test and here it is not a reliable measure(think about the sample not being accurate and  $\mu$ ) as it is sensitive.
- Sample size not good enough for LLN( $n=20$ ) here

Hence a sign test on Median is a better test to use.

Sign test has no dependencies/assumptions about the population parameters or the distributions

Can perform test on Median.

Median, Mode not sensitive to odd values(outliers).

**d. Run a suitable sign test for these data. What do you conclude?**

**Setup, !!!THIS IS WHAT GOOGLE WOULD SUGGEST, DON'T DO THIS!!!**

```
install.packages('smmR')
```

Follow this from lecture slides.

github source: [here](#)

```
# One time setup
install.packages('devtools')
library(usethis) # this step is specific to my machine, might not be required for you
library(devtools)
install_github("nxskok/smmr")
```

```
library(smmr)
```

This is how you run the test.

Now note we are interested in the the Median being = 6, so we state that. Furthermore think about whether it will be 2-sided or 1 sided?

- 2-sided bc the value can lie on either side.

`sign_test(data, variable, $H_{0})`  $\alpha = 0.05$

```
sign_test(meals, protein, 6)
```

```
## $above_below
## below above
##      15      5
##
## $p_values
## alternative    p_value
## 1          lower 0.02069473
## 2           upper 0.99409103
## 3      two-sided 0.04138947
```

The P-value, 0.0414, is less than  $\alpha = 0.05$ , so we reject the null hypothesis and conclude that the median is different from 6 ounces. The advertisement by the company is not accurate.

e. In your sign test, how could you have deduced that the P-value was going to be small even without looking at any of the P-values themselves?

- How is Sign test conducted ?
  - Compare the sampled values against the  $H_0$ 
    - \* What happens when values are EQUAL?  
Don't include those values as currently interested in  $>$  or  $<$

```
meals %>%
  group_by(protein > 6) %>%
  summarise(n = n())
```

```
## # A tibble: 2 x 2
##   `protein > 6`      n
##   <lgl>          <int>
## 1 FALSE           15
## 2 TRUE            5
```

- Note 15 are less than 6. Hence can reject and assume p-value will be less without conducting the test.

f. Obtain a 90% confidence interval for the population median protein content. What does this tell you about the reason for the rejection or non-rejection of the null hypothesis above?

```
ci_median(data, variable, conf.level = 1-alpha) by default  $\alpha = 0.05$ 
```

```
alpha = .1
```

```
ci_median(meals, protein, conf.level = 1-alpha)
```

```
## [1] 4.905273 5.793750
```

- Does our  $H_0$  value lie within the range at  $\alpha = 0.1$ ?
  - No! 6 not  $\in$  CI.

## 10.8 Handspans revisited

Recall the original study was to compare the handspan between males and females.

So we have 2 groups in our study.

a + b. Load data make plots to investigate and compare normality between the sample group

```
# data
```

```
my_url <- "http://ritsokiguess.site/datafiles/handspan.txt"
```

```
span <- read_delim(my_url, " ")
```

```
## Rows: 190 Columns: 2
```

```
## -- Column specification -----
```

```
## Delimiter: " "
```

```
## chr (1): sex
```

```
## dbl (1): handspan
```

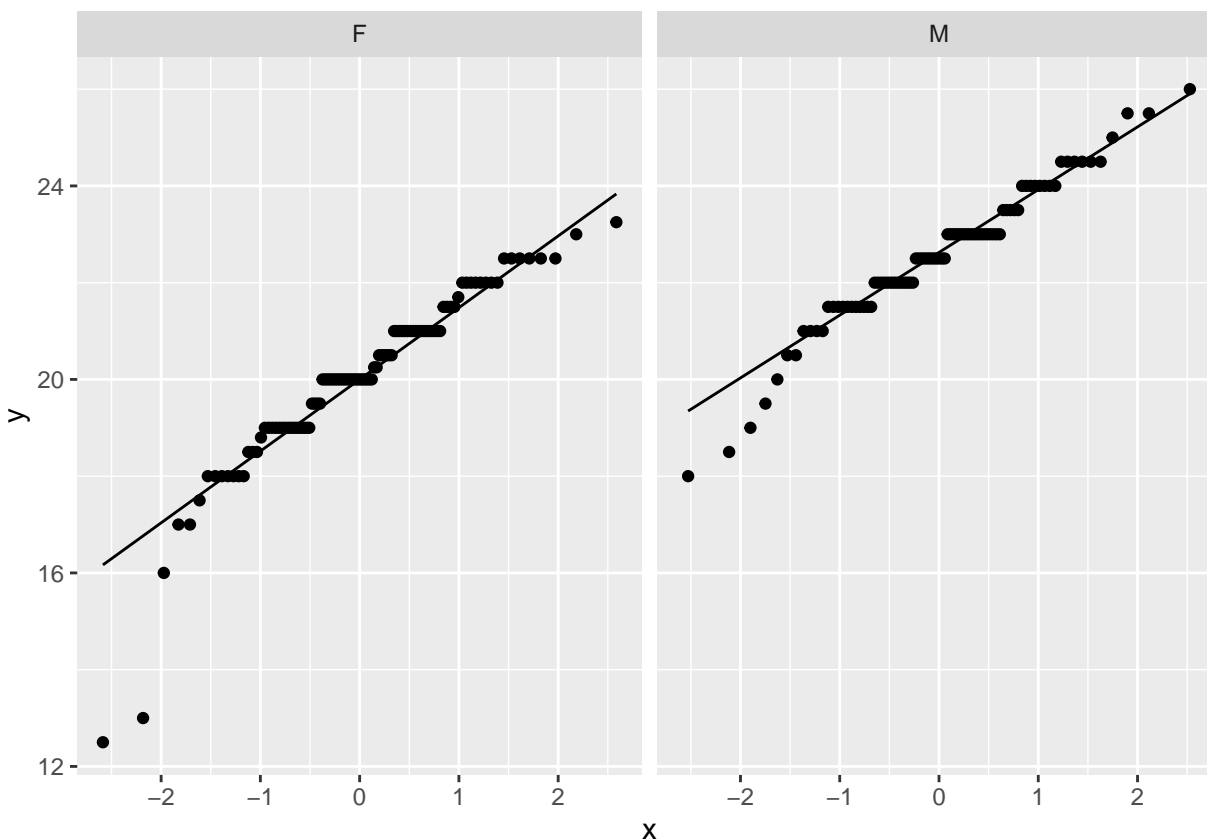
```
##
```

```
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
# QQ-Plot
```

```
ggplot(span, aes(sample=handspan)) + stat_qq() + stat_qq_line() + facet_wrap(~sex)
```



- What is QQ plot?

- What is a quantile ? Can think **similar**(but not really) to bins.

- How is it plotted?

Compare the quantiles of Normal(0,1) against the sample's quantiles

- Our plot description:

- Males:

Not normal as the values to the left are deviating away from the normal line(SLIGHTLY left skewed)

- Females:

Same as males + seem to have outliers aswell on the left side

**c. Discuss briefly whether you might prefer to use Mood's median test to compare the handspans of the male and female students, compared to a two-sample t-test.**

- T-test assumptions about the normality of the sample is not satisfied.
- Mood-Median is the move as does not rely on  $\mu$  + Normality.

**d. Run the Mood-Median test**

$$H_0 : \mu_{\frac{1}{2}} = 0 \quad \implies \mu_{\frac{1}{2}} = \text{median}$$

The median between both the groups is the same ie the difference is 0.

**Setup the call for installing the package is covered above.**

```
# library call
library(smmr)
```

To run the test. `median_test(data, variable of the median, group)`

```
median_test(span, handspan, sex)
```

```
## $table
##      above
## group above below
##      F      17      82
##      M      65      11
##
## $test
##      what      value
## 1 statistic 8.06725e+01
## 2          df 1.00000e+00
## 3    P-value 2.66404e-19
```

The P-value of  $2.66404 \times 10^{-19}$  is extremely small, so we can conclude that males and females have different median handspans. Remember that we are now comparing medians, and that this test is **two-sided**.

## 12.3 Lengths of heliconia flowers

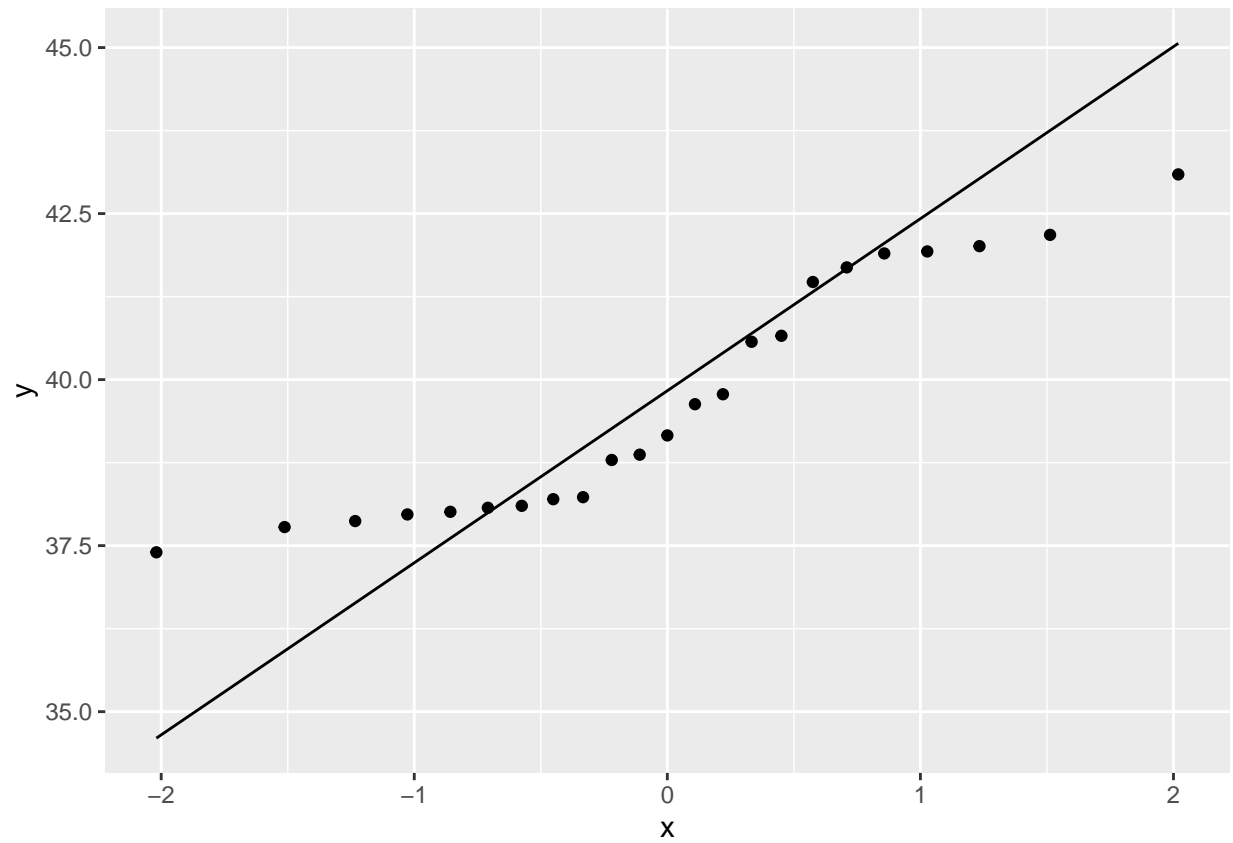
Flower length and the beak of the humming-bird has been evolved according to the specie

a,b,c,d. Read data and make QQ-Plot for each specie

```
# a
my_url <- "http://ritsokiguess.site/datafiles/heliconia.csv"
heliconia = read_csv(my_url)

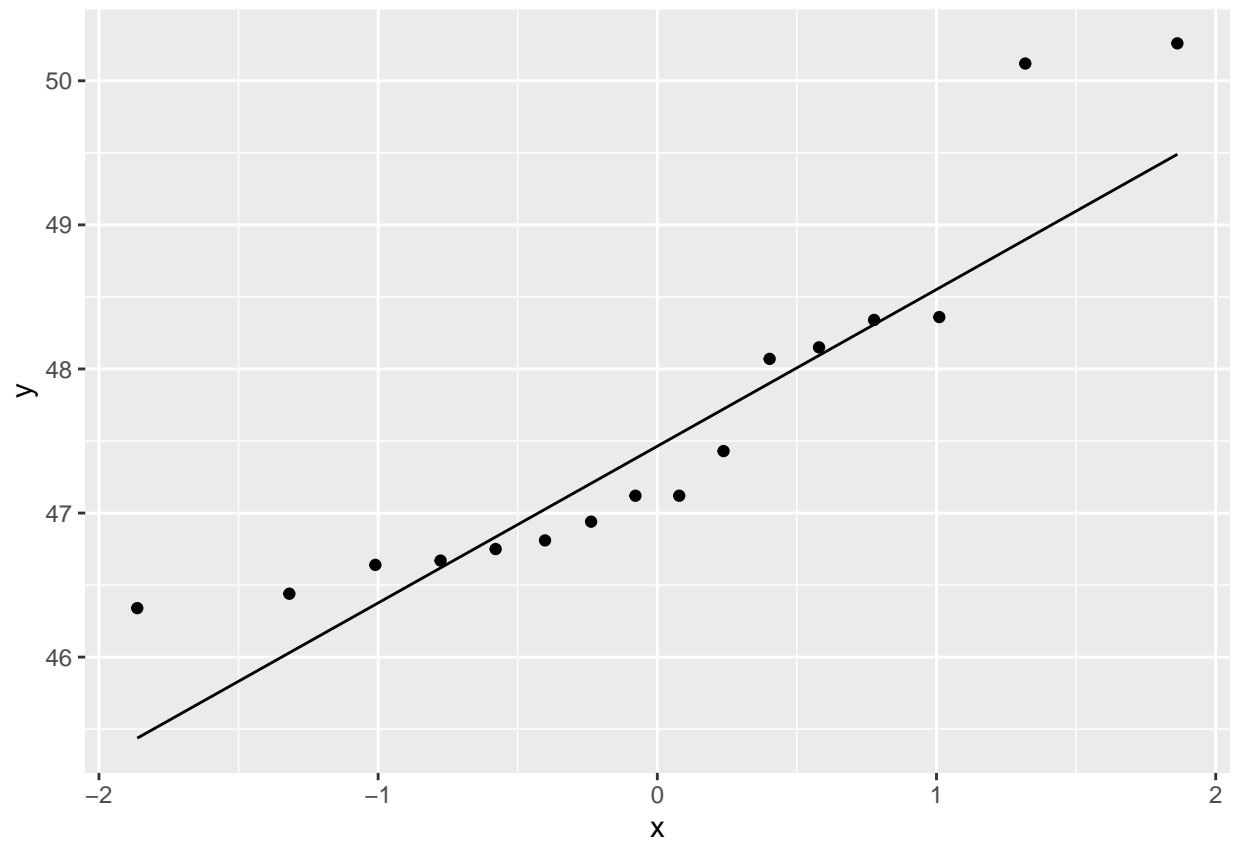
## Rows: 23 Columns: 3
## -- Column specification -----
## Delimiter: ","
## dbf (3): bihai, caribaea_red, caribaea_yellow
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

# bcd
## way 1
ggplot(heliconia, aes(sample=caribaea_red)) + stat_qq() + stat_qq_line()
```



```
## way 2
heliconia %>% ggplot(aes(sample=bihai)) + stat_qq() + stat_qq_line()

## Warning: Removed 7 rows containing non-finite values (stat_qq).
## Warning: Removed 7 rows containing non-finite values (stat_qq_line).
```

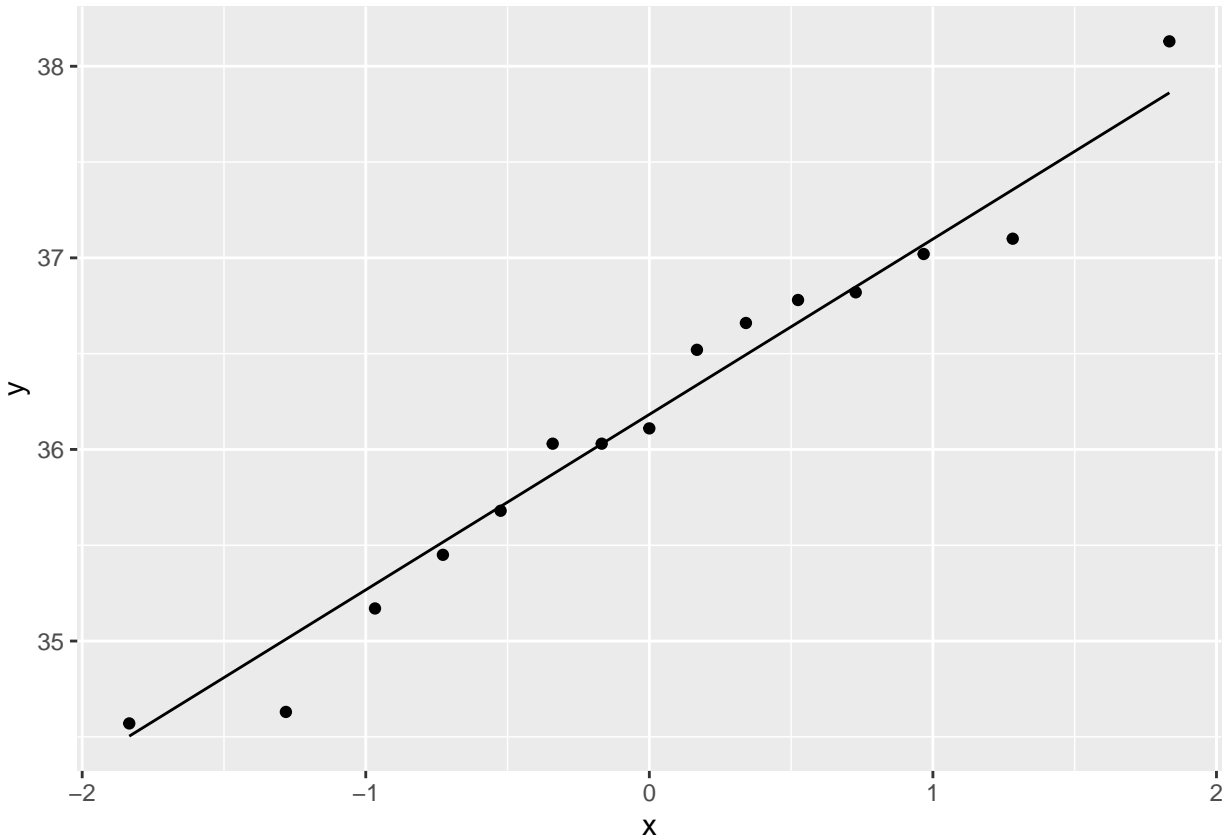


```
ggplot(heliconia,aes(sample=caribaea_yellow))+stat_qq()+stat_qq_line()
```

```
## Warning: Removed 8 rows containing non-finite values (stat_qq).
```

```
## Warning: Removed 8 rows containing non-finite values (stat_qq_line).
```

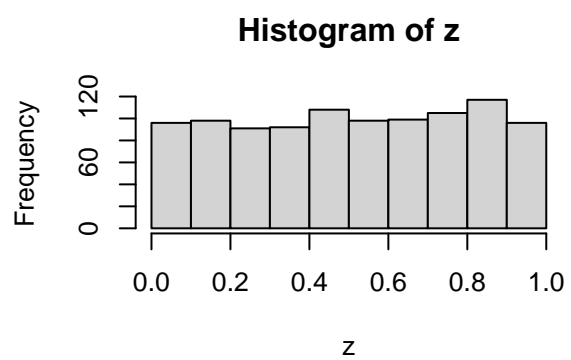
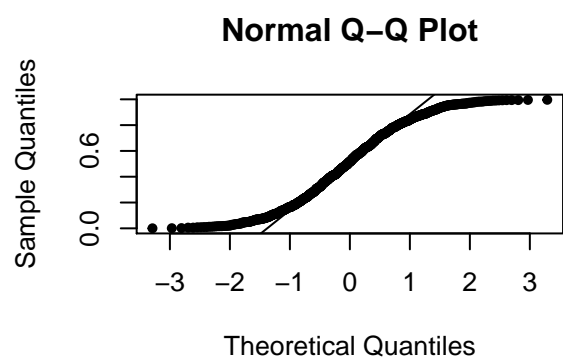
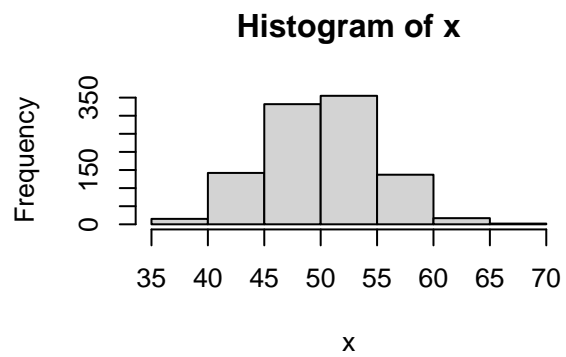
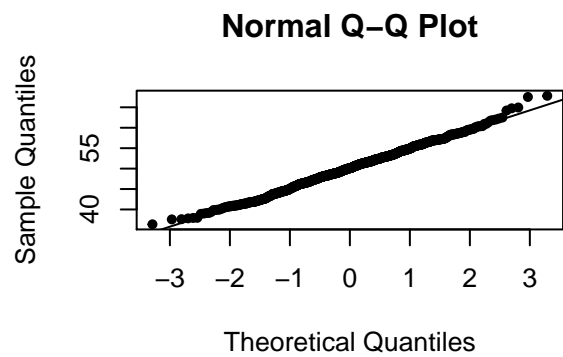




### e, f. Which one is the closest to the normal and of the other explain why they are not. - *caribaea\_yellow* closest to normal as the quantiles seems to match with Normal QQ. - *bihai* seems to have a S-shaped if we negate the outliers at the top + left skewed with removal of outliers and without the removal seems to be right skewed. - *caribaea\_red*  
Short tailed Initially points seem too bunched up on the line then they seem to tail away from the QQ-line. Looks Uniform distribution.

```
### Source: https://bioinfo.irc.ca/permutations/

x <- rnorm(1000, mean=50, sd=5) # normal distribution
z <- runif(1000) # uniform distribution
par(mfrow=c(2,2))
qqnorm(x, pch=20)
qqline(x)
hist(x)
qqnorm(z, pch=20)
qqline(z)
hist(z)
```



Check more examples of QQ x Distribution relation. [here](#)