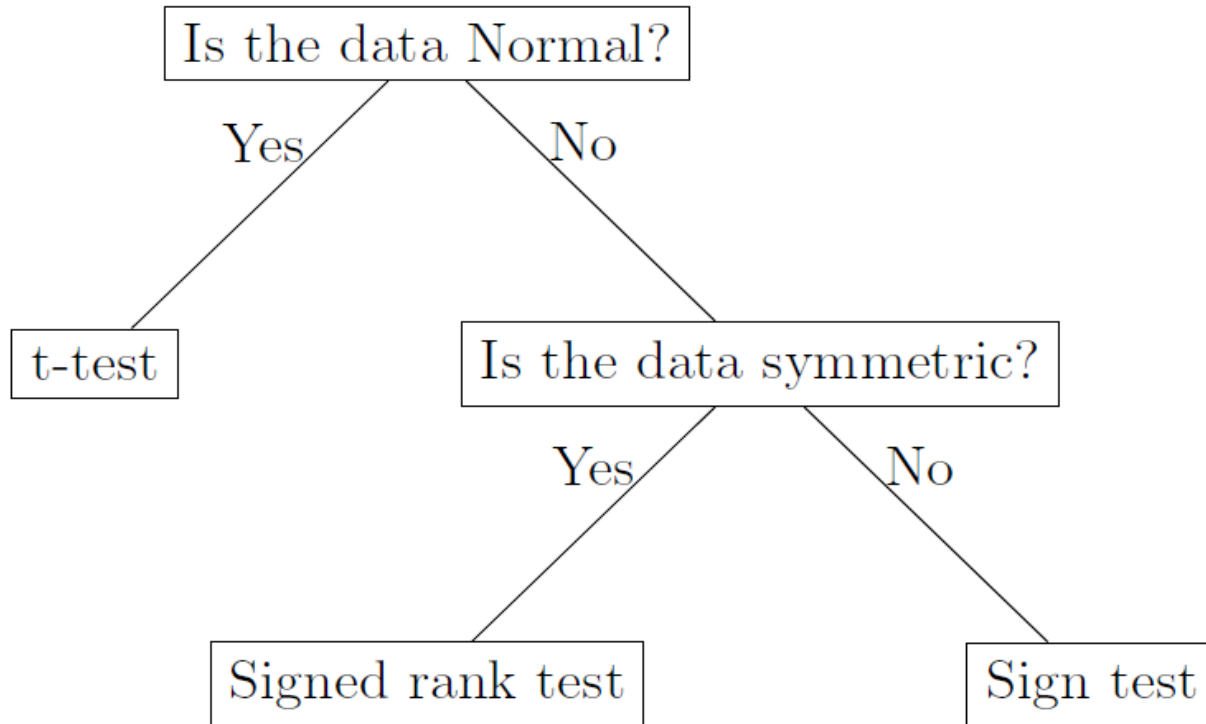


# **One-sample and two-sample inferential test**

# One-sample location test



## Inferential test on location parameter

- Two-sided test:

$$H_0: \mu = m_0 \text{ vs. } H_1: \mu \neq m_0$$

( $m_0$  is a number)

- One-sided test:

$$H_0: \mu = m_0 \text{ vs. } H_1: \mu > m_0 \\ \text{(or } \mu < m_0 \text{)}$$

# One-sample location test

- If data is normally distributed, we test MEAN value with t-test statistic (**parametric test**)

$$t = \frac{\bar{x} - m_0}{\hat{\sigma} / \sqrt{n}},$$

where  $n$  is sample size,  $\bar{x}$  is sample mean, and  $\hat{\sigma}$  is sample standard deviation.

- If data is not normally distributed, we test MEDIAN with **non-parametric test** (Signed rank or sign test)

# Example: Water data

The data:

- 61 data points from towns in England
- **Mortal**: Mortality rate per 100,000 males (averaged over 1958-1964)
- **Hardness**: Calcium concentration (higher = harder water) in ppm in the town's drinking water
- **Location**: Indicator for Southern or Northern town

# Example: Water data

- One-sample location test
  - Take  $m_0=1500$  for **mortal** and  $m_0=45$  for **hardness** and perform two-sided test
1. Specify null and alternative hypotheses
  2. Data exploration and Normality check
  3. Choose which test to use
  4. Make a conclusion

# Example: Water data

For testing Mortality rate ( $H_0: m=1500$  vs.  $H_1: m \neq 1500$ )

First check normality through visualization and shapiro-wilk test

```
# one-sample t-test
t.test(water$mortal, mu=1500)

##
## One Sample t-test
##
## data: water$mortal
## t = 1.005, df = 60, p-value = 0.319
## alternative hypothesis: true mean is not equal to 1500
## 95 percent confidence interval:
## 1476.083 1572.212
## sample estimates:
## mean of x
## 1524.148
```

$H_0$ : mean of mortality = 1500  
 $H_a$ : mean of mortality  $\neq$  1500

# Example: Water data

For testing hardness ( $H_0: m=45$  vs.  $H_1: m \neq 45$ )

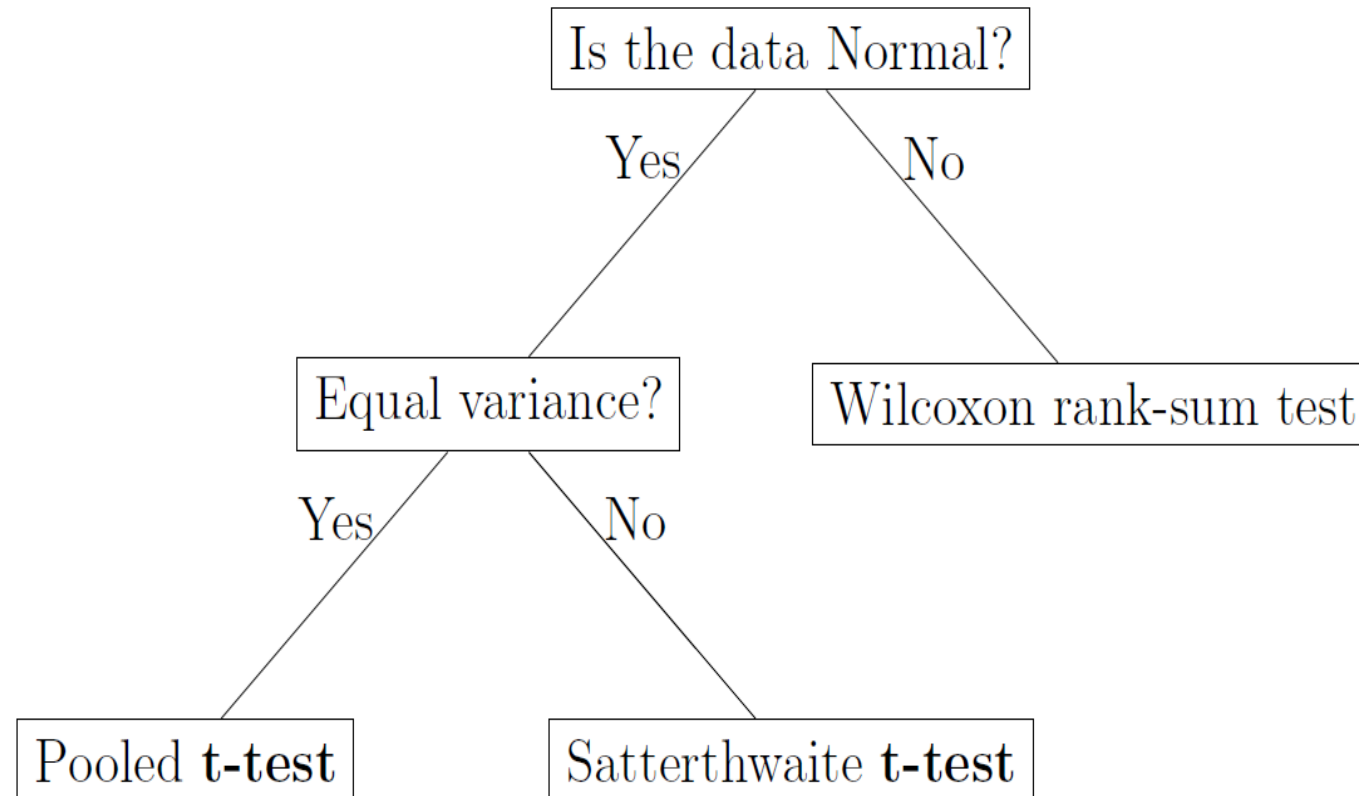
First check normality through visualization and shapiro-wilk test. If normality assumption does not hold, check its symmetry

```
SIGN.test(water$hardness, md=45)
```

```
##  
## One-sample Sign-Test  
##  
## data: water$hardness  
## s = 27, p-value = 0.4426  
## alternative hypothesis: true median is not equal to 45  
## 95 percent confidence interval:  
## 18.63777 58.36223  
## sample estimates:  
## median of x  
## 39
```

$H_0$ : median of hardness = 45  
 $H_a$ : median of hardness  $\neq$  45

# Two-sample test of population difference



## T- test

(parametric test) -> **comparing two mean values**

- Two-sided test:

$$H_0: \mu_1 = \mu_2 \quad \text{vs.} \quad H_1: \mu_1 \neq \mu_2$$

- One-sided test:

$$H_0: \mu = m_0 \quad \text{vs.} \quad H_1: \mu > m_0$$

$$t_{pooled} = \frac{\bar{x}_1 - \bar{x}_2}{\hat{\sigma}_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}, \quad \text{or} \quad t_{Satt} = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\hat{\sigma}_1^2}{n_1} + \frac{\hat{\sigma}_2^2}{n_2}}}$$

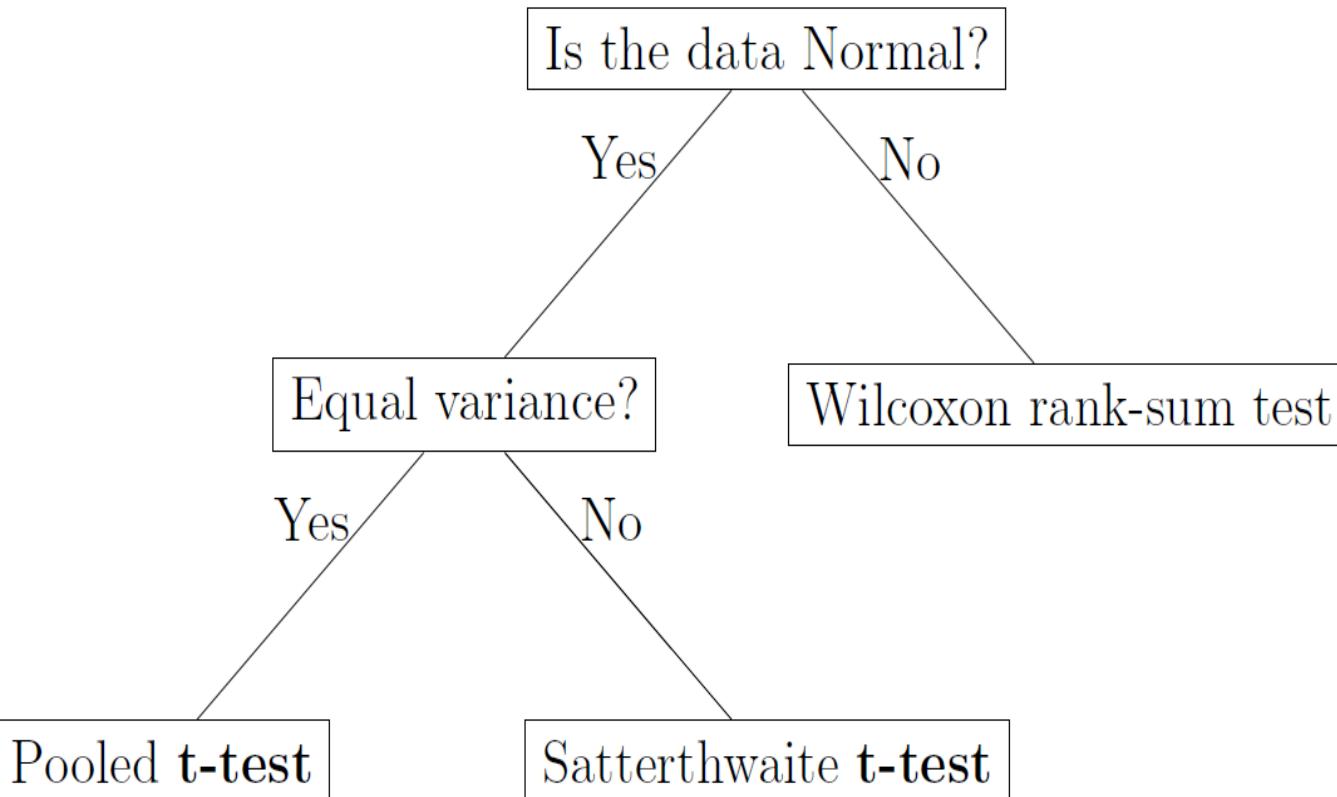


# Two-sample test of population difference

## Wilcoxon rank-sum test

(non-parametric test) -> **not about comparison of mean values**

<https://www.stat.auckland.ac.nz/~wild/ChanceEnc/Ch10.wilcoxon.pdf>



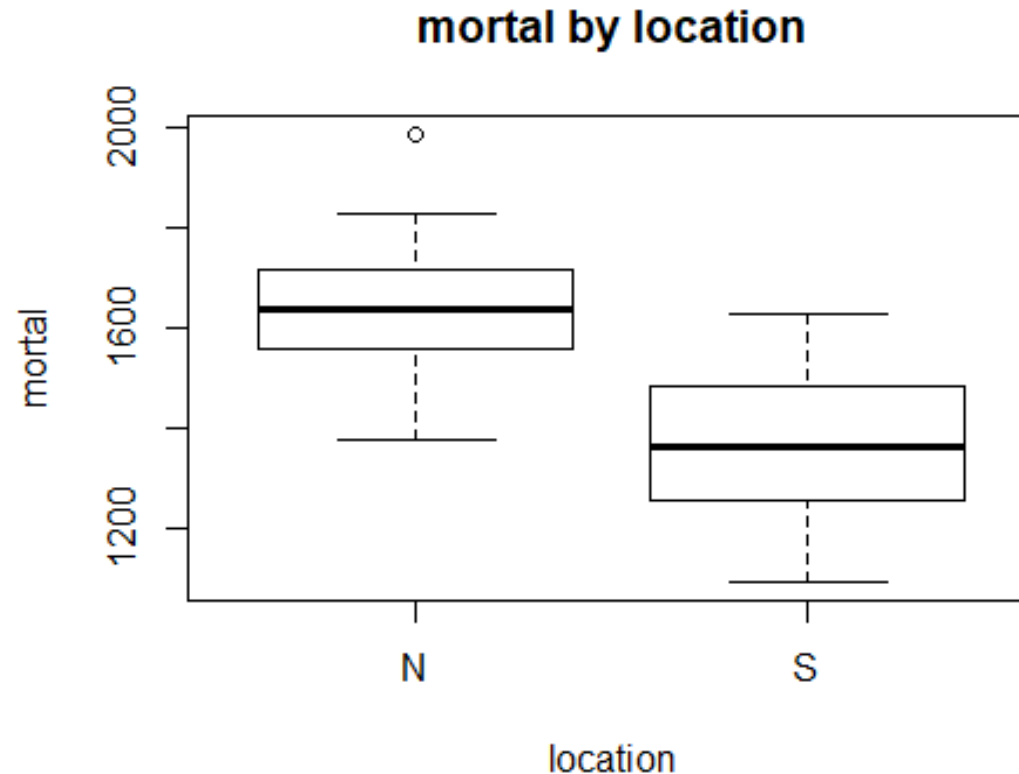
- Two-sided test:  
H0: Two populations come from the same distribution  
H1: One of the populations tends to have larger values (either population 1 or 2)
- One-sided test:  
H0: Two population come from the same distribution  
H1: Population 1 tends to have larger values than Population 2

# Example: Water data

- Test for a significant difference between the mortality rates in the north and south
  - Do the same for the water hardness values
  - What are our conclusions?
1. Specify null and alternative hypotheses
  2. Data exploration and Normality check by location
  3. Choose which test to use
    - I. If both follow normal distribution, perform equal variance test
    - II. Depending on variance test result choose the proper one
  4. Make a conclusion

# Example: Water data

- For two-sample test for Mortality rate comparison between South and North, first check if **BOTH** are normally distributed



# Example: Water data

- For two-sample test for Mortality rate comparison between South and North, first check if **BOTH** are normally distributed

```
shapiro.test(water$mortal[water$location=="S"]) # Mortal of SOUTH
```

```
##
```

```
## Shapiro-Wilk normality test
```

```
##
```

```
## data: water$mortal[water$location == "S"]
```

```
## W = 0.96579, p-value = 0.518
```

H0: Data follow normal distribution  
Ha: Data does not follow normal distribution

```
shapiro.test(water$mortal[water$location=="N"]) # Mortal of NORTH
```

```
##
```

```
## Shapiro-Wilk normality test
```

```
##
```

```
## data: water$mortal[water$location == "N"]
```

```
## W = 0.97554, p-value = 0.6117
```

# Example: Water data

- If **BOTH** follow Normal distribution, we perform two-sample t-test. To do this, check equal variance to choose between pooled t-test (equal variance case) and Satterthwaite t-test (unequal variance case)

```
var.test(mortal ~ location, water,  
         alternative = "two.sided")
```

```
##
```

```
## F test to compare two variances
```

```
##
```

```
## data: mortal by location
```

```
## F = 0.95305, num df = 34, denom df = 25, p-value = 0.883
```

```
## alternative hypothesis: true ratio of variances is not equal to  
1
```

```
## 95 percent confidence interval:
```

```
## 0.4428321 1.9655085
```

```
## sample estimates:
```

```
## ratio of variances
```

```
## 0.9530519
```

H0: two groups have the same variance

Ha: two groups have different variances

P-value is calculated as 0.883 – not reject H0

# Example: Water data

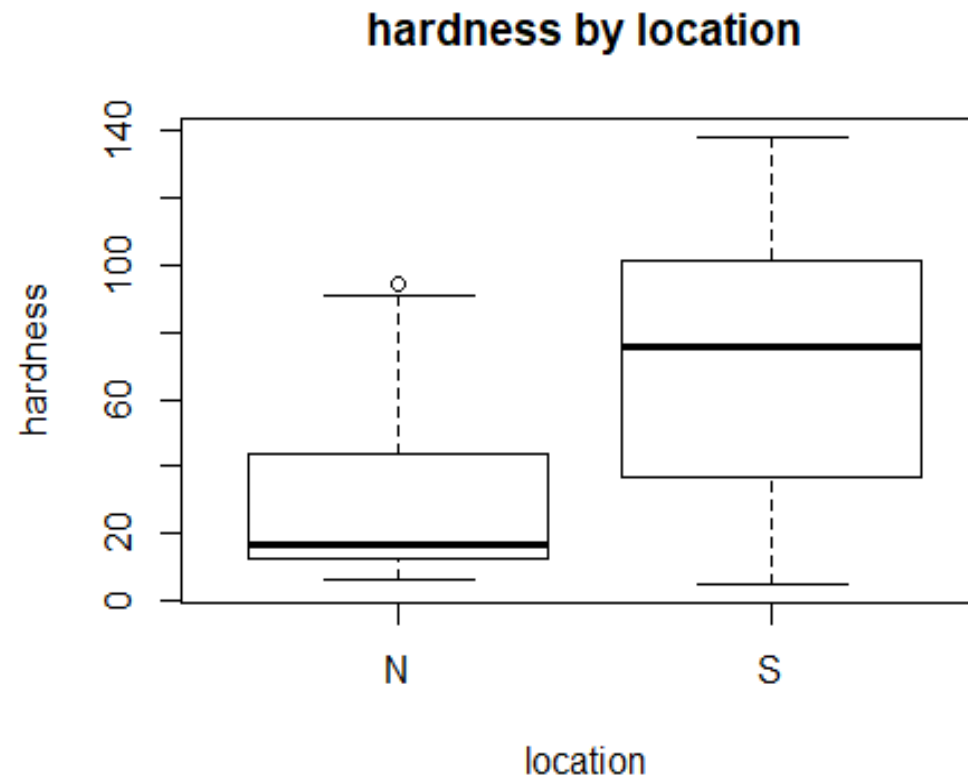
```
t.test(mortal ~ location, water, alternative = "two.sided", var.
equal=TRUE)      # NOTE: if unequal variance => var.equal=FALSE

##
##  Two Sample t-test
##
## data: mortal by location
## t = 7.1686, df = 59, p-value = 1.402e-09
## alternative hypothesis: true difference in means is not equal
## to 0
## 95 percent confidence interval:
##  185.1125  328.4721
## sample estimates:
## mean in group N mean in group S
##      1633.600      1376.808
```

H0: mean of South = mean of North  
Ha: mean of South != mean of North

# Example: Water data

- For two-sample test for hardness comparison between South and North, first check if **BOTH** are normally distributed



# Example: Water data

- For two-sample test for hardness comparison between South and North, first check if **BOTH** are normally distributed

```
shapiro.test(water$hardness[water$location=="S"]) # Mortal of South
```

```
##
```

```
## Shapiro-Wilk normality test
```

```
##
```

```
## data: water$hardness[water$location == "S"]
```

```
## W = 0.95562, p-value = 0.3127
```

```
shapiro.test(water$hardness[water$location=="N"]) # Mortal of North
```

```
##
```

```
## Shapiro-Wilk normality test
```

```
##
```

```
## data: water$hardness[water$location == "N"]
```

```
## W = 0.81139, p-value = 3.439e-05
```



# Example: Water data

- If at least one does NOT follow Normal distribution, we perform nonparametric Wilcoxon test

```
wilcox.test(hardness ~ location, data=water, exact=FALSE)
##
##  Wilcoxon rank sum test with continuity correction
##
## data:  hardness by location
## W = 202.5, p-value = 0.0002363
## alternative hypothesis: true location shift is not equal to 0
```

H0: Two groups are from the same distribution (same median)

Ha: One group (either South or North) tends to have larger values  
(One group has larger median value than the other group)

# Other alternative for non-normal data

- Non-parametric test requires more computation
- Software may not include functions for the implementation
- Can we take advantages of parametric test for non-normal data?
  - Transformation (e.g., **log** or square-root transformation)
    - Box-cox transformation
    - [http://support.sas.com/documentation/cdl/en/statug/63033/HTML/default/viewer.htm#statug\\_tran\\_sreg\\_sect015.htm](http://support.sas.com/documentation/cdl/en/statug/63033/HTML/default/viewer.htm#statug_tran_sreg_sect015.htm)