# Nonparametric statistic

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## November 2024

## 1 Introduction

# 2 Sign Test

**Null Hypothesis:** 

$$H_0: M_0 = m_0$$

$$H_1: M_0 \neq m_0$$

$$H_0: P(\text{observing "} + ") = P(\text{observing "} - ")$$

### Test Statistic:

Let T be the number of positive signs in the sample.

$$T = \sum_{i=1}^{n} I(X_i < m_0)$$

where I is the indicator function. The null distribution of T is Bin (n,  $p = \frac{1}{2}$ ).

$$n = n' - \text{numbers of "tie"}$$

where n' is the sample size.

Let

$$Y \sim \operatorname{Bin}(n, p = \frac{1}{2})$$

To find r

$$P(Y \le r) = \alpha_1 \approx \frac{\alpha}{2}$$

If T is less than or equal to r or is greater than or equal to n-r, we will reject  $H_0$ .

when n > 25:

$$r = \frac{1}{2} \left( n + z_{\alpha/2} \cdot \sqrt{n} \right)$$

# 3 Wilcoxon Signed-Rank Test

**Null Hypothesis:** 

$$H_0: M_0 = m_0$$

$$H_1: M_0 \neq m_0$$

## Procedure:

## 1. Calculate the Differences

For each observation  $X_i$ , calculate the absolute difference from  $m_0$ :

$$Z_i = |X_i - m_0|$$

### 2. Exclude Zero Differences

Remove any pairs where  $Z_i = 0$  to avoid ties.

### 3. Rank the Absolute Differences

Rank the non-zero values  $Z_i$  from smallest to largest, assigning each  $Z_i$  a rank  $R_i$ . For tied ranks, assign the average rank to each tied value.

### 4. Assign Signs to Ranks

Assign a positive sign to the ranks of positive differences  $X_i > m_0$  and a negative sign to the ranks of negative differences  $X_i < m_0$ .

## 5. Compute the Test Statistic $\boldsymbol{W}$

Let W be the sum of the ranks with positive signs:

$$W = \sum_{X_i > m_0} R_i$$

### 6. Determine the Critical Value

Use the Wilcoxon signed-rank test critical value table or approximate the critical value for large n with the normal distribution.

#### **Decision Rule:**

Reject  $H_0$  if W is less than or equal to the critical value (or, for a two-tailed test, if W is outside the acceptance region).

# 4 Kolmogorov-Smirnov Test

### **Null Hypothesis:**

The sample distribution does not significantly differ from a specified theoretical distribution:

$$H_0: F(x) = F_0(x)$$

$$H_1: F(x) \neq F_0(x)$$

where F(x) is the cumulative distribution function (CDF) of the sample, and  $F_0(x)$  is the CDF of the theoretical distribution.

### Test Statistic:

Calculate the empirical cumulative distribution function S(x):

$$S(x) = \frac{\text{number of } X_i \text{'s that are } \le x}{n}, \quad x \in R$$

The test statistic D is then defined as:

$$D = \max |F_0(x) - S(x)|$$

### **Decision Rule:**

Use the observed D value and compare it to the critical value obtained from the Kolmogorov-Smirnov table. Reject  $H_0$  if  $D_{\rm obs}$  exceeds the  $(1-\alpha)$ th quantile of the distribution of D at the desired significance level  $\alpha$ .

# 5 Randomness (Run Test)

## **Null Hypothesis:**

The sequence of observations is random, meaning that the number of runs is consistent with the expected distribution of runs under random conditions:

 $H_0$ : The sequence is random

## Procedure:

- 1. **Set a Threshold:** Choose a threshold value (e.g., mean or median) to categorize the continuous variable into two groups.
- 2. Binarize the Data: Label each observation as "success" (above threshold) or "failure" (below threshold).
- 3. Count the Runs: Calculate the number of runs R in the binary sequence.
  - U: The number of changes between successes and failures in the sequence.
  - $n_1$ : Number of successes.
  - $n_2$ : Number of failures.
- 4. **Determine Critical Values:** Find the lower and upper critical values  $r_1$  and  $r_2$  from the table for a significance level  $\alpha = 0.05$ .

## Test Statistic:

U: The number of changes between successes and failures in the sequence.

### **Decision Rule:**

Find the lower and upper critical values  $r_1$  and  $r_2$  from the table for a significance level  $\alpha = 0.05$ . Reject  $H_0$  if  $U \leq r_1$  or  $U \geq r_2$ .

When either  $n_1$  or  $n_2$  is greater than 20, use the test statistic:

$$Z = \frac{U - \frac{2n_1n_2}{n_1 + n_2 + 1}}{\sqrt{\frac{2n_1n_2(2n_1n_2 - n_1 - n_2)}{(n_1 + n_2)^2(n_1 + n_2 - 1)}}} \sim N(0, 1).$$

# 6 Trend Test (Cox and Stuart Test)

## **Null Hypothesis:**

 $H_0$ : There is no trend in the data.

 $H_1$ : There is either an upward or a downward trend.

## Procedure

Divide the data into two halves. For each pair of observations  $(X_i, X_{i+c})$ , where  $c = \frac{n'}{2}$  if n' is even, and  $c = \frac{n'+1}{2}$  if n' is odd. Assign to each  $(X_i, X_{i+c})$  a "+" symbol if  $X_i < X_{i+c}$ , a "-" symbol if  $X_i > X_{i+c}$ , and a "tie" symbol otherwise. Let n be the number of non-tied pairs of data.

### **Test Statistic:**

$$T = \text{total number of +'s}$$

and the null distribution of T is Bin (n,  $p = \frac{1}{2}$ ). The test hypothesis and the rest of the procedure is identical to that of the sign test.

### **Decision Rule:**

Use the binomial test to determine the significance of the count of upward trends. Reject  $H_0$  if the number of upward or downward trends is significantly different from what would be expected by chance.