# Non-parametric tests

12.2, 12.4 of Ross Textbook

#### Non-parametric tests

- We make no assumptions of the form of the distribution function unlike previous cases where we assume Normal or Binomial.
- Generically denote distribution as F(X). Note form of F(X) is not known
- Possible hypothesis that can be tested in such cases
  - What is the median of F(X)?
    - Sign test
  - Is the distribution around the median similar
    - Sign rank test
  - Given samples of two distributions: Are they likely to be from the same or different distributions?
    - Two sample test

## Sign test

Let  $X_1, ..., X_n$  denote a sample from a continuous distribution F and suppose that we are interested in testing the hypothesis that the median of F, call it m, is equal to a specified value  $m_0$ . That is, consider a test of

$$H_0: m = m_0$$
 versus  $H_1: m \neq m_0$ 

where *m* is such that F(m) = .5.

There is such that 
$$F(m) = .3$$
.

$$T = \# \text{ af } \chi_{i} - s \text{ less than } m_{0} \qquad \widehat{T} = \text{ median } \text{ af } (\chi_{i} - ... \chi_{n})$$

$$\widehat{T}(x) = \frac{1}{n} \sum_{i=1}^{n} \mathbb{E}(\chi_{i} \leq \chi_{i}) - \mathbb{E}(\chi_{i}) = \mathbb{E}(\chi_{i}) = \mathbb{E}(\chi_{i})$$

$$\widehat{T}(x) = \frac{1}{n} \sum_{i=1}^{n} \mathbb{E}(\chi_{i}) = \mathbb{E}(\chi_{i}) = \mathbb{E}(\chi_{i})$$

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#### T-statistic

• T = sum of the sign of  $m_0 - X_i$ 

• 
$$T = \sum_{i} I_{\underline{i}}$$
  $I_{\underline{i}} = \begin{cases} 1 & \text{if } X_{i} < m_{0} \\ 0 & \text{if } X_{i} \ge m_{0} \end{cases}$   $P(I_{\underline{i}}) \sim \text{Burwilli} \left(\frac{1}{2}\right)$ 

What is the distribution of T under the null hypothesis?

$$\frac{P_{H_0}(T) \sim \text{Binonmal}(n, \frac{1}{2})}{\sum_{j=1}^{m_{12}} \left( \frac{x_{1}}{x_{2}} - \frac{x_{1}}{x_{1}} + \frac{x_{1}}{x_{1}} + \frac{x_{1}}{x_{1}} - \frac{x_{1}}{x_{1}} \right)}$$

$$T = \frac{2}{J=1} \left( \frac{x_{1}}{x_{2}} - \frac{x_{1}}{x_{1}} + \frac{x_{1}}{x_{1}} - \frac{x_{1}}{x_{1}} - \frac{x_{1}}{x_{1}} \right) \qquad P(T)$$

Example 12.2.b. A financial institution has decided to open an office in a certain community if it can be established that the median annual income of families in the community is greater than \$90,000. To obtain information, a random sample of 80 families was chosen, and the family incomes deter- n = 80mined. If 28 of these families had annual incomes below and 52 had annual incomes above \$90,000, is this significant enough to establish, say, at the 5 percent level of significance, that the median annual income in the community is greater than \$90,000?

mo = 90K

**Solution**. We need to see if the data are sufficient to enable us to reject the null hypothesis when testing

$$H_0: m \leq 90$$
 versu

 $H_1: m > 90$ versus

$$p$$
-value =  $P(Bin(80, 1/2) \le 28) = pbinom(28, 80, 1/2) = 0.004841425$ 

and so the null hypothesis that the median income is less than or equal to \$90,000 is rejected.

### Signed rank test

Given n sample  $X_1, \dots, X_n$  from unknown distribution F, we are interested in the hypothesis that F is symmetric about a given median  $m_0$ , that is,

•  $H_0$ :  $P(X > m_0 + a) = P(X < m_0 - a)$ , for all a

Let  $Y_i = X_i - m_0$ , i = 1, ..., n and rank (that is, order) the absolute values  $|Y_1|, |Y_2|, ..., |Y_n|$ . Set, for j = 1, ..., n,

$$I_{j} = \begin{cases} 1 & \text{if the } j \text{th smallest value comes from a data value that is smaller} \\ & \text{than } m_{0} \\ 0 & \text{otherwise} \end{cases}$$
**lest statistic**

$$T = \sum_{j=1}^{n} j I_j$$

**Example 12.3.a.** If n = 4,  $m_0 = 2$ , and the data values are  $X_1 = 4.2$ ,  $X_2 = 1.8$ ,  $X_3 = 5.3$ ,  $X_4 = 1.7$ , then the rankings of  $|X_i - 2|$  are .2, .3, 2.2, 3.3. Since the first of these values — namely, .2 — comes from the data point  $X_2$ , which is less than 2, it follows that  $I_1 = 1$ . Similarly,  $I_2 = 1$ , and  $I_3$  and  $I_4$  equal 0. Hence, the value of the test statistic is T = 1 + 2 = 3.

$$7i$$
  $9.2$   $1.8$   $5.3$   $1.7$ 
 $-2$   $-2$   $-2$   $-2$ 
 $1i$   $2.2$   $-0.2$   $3.3$   $-0.3$ 
 $14:1$   $6.2$   $0.3$   $2.2$   $3.3$ 
 $1$ 

Distribution of test statistic  $P_{H_0}(T)$  under the null hypothesis?

Expected value and variance of T under  $H_0$  probability that the jth absolute difference is from an  $P\{I_j=1\}=\frac{1}{2}=P\{I_j=0\}, \quad j=1,\ldots,n$   $E[I_j]=\frac{1}{2}, \quad Var(I_j)=\frac{1}{4}$ 

$$P\{I_j = 1\} = \frac{1}{2} = P\{I_j = 0\}, \quad j = 1, \dots, n$$

Hence, we can conclude that under  $H_0$ ,

$$E[T] = E\left[\sum_{j=1}^{n} jI_{j}\right]$$

$$= \sum_{j=1}^{n} \frac{j}{2} = \frac{n(n+1)}{4}$$

$$Var(T) = Var\left(\sum_{j=1}^{n} jI_{j}\right)$$

$$= \sum_{j=1}^{n} j^{2} Var(I_{j})$$

$$= \sum_{j=1}^{n} \frac{j^{2}}{4} = \frac{n(n+1)(2n+1)}{24}$$

 $P_{H_0}(T)$  = approximately normal for large n with mean and variance as above. But we can do better..

An exact computation of probability  $P_{H_0}(T)$  recursively

$$\begin{split} \underline{P_{k}(i)} &= P_{H_{0}} \left\{ \sum_{j=1}^{k} j I_{j} \leq i \right\} \\ &= P_{H_{0}} \left\{ \sum_{j=1}^{k} j I_{j} \leq i | I_{k} = 1 \right\} \\ &+ P_{H_{0}} \left\{ \sum_{j=1}^{k} j I_{j} \leq i | I_{k} = 0 \right\} \\ &+ P_{H_{0}} \left\{ \sum_{j=1}^{k} j I_{j} \leq i | I_{k} = 0 \right\} \\ &= P_{H_{0}} \left\{ \sum_{j=1}^{k-1} j I_{j} \leq i - k | I_{k} = 1 \right\} \\ &+ P_{H_{0}} \left\{ \sum_{j=1}^{k-1} j I_{j} \leq i | I_{k} = 0 \right\} \\ &+ P_{H_{0}} \left\{ \sum_{j=1}^{k-1} j I_{j} \leq i | I_{k} = 0 \right\} \\ &= P_{H_{0}} \left\{ \sum_{j=1}^{k-1} j I_{j} \leq i - k \right\} P_{H_{0}} \{I_{k} = 1\} + P_{H_{0}} \left\{ \sum_{j=1}^{k-1} j I_{j} \leq i \right\} P_{H_{0}} \{I_{k} = 0\} \end{split}$$

#### Continued..

$$P_{H_0}\{I_k = 1\} = P_{H_0}\{I_k = 0\} = \frac{1}{2}$$

we see that

$$P_k(i) = \frac{1}{2}P_{k-1}(i-k) + \frac{1}{2}P_{k-1}(i)$$

$$P_{\mu}(i) = P_{k-1}(i-k)P\{I_{k}=1\}$$
+  $P_{\mu-1}(i)P\{I_{k}=1\}$ 

Base Case: 
$$P_{\underline{1}}(i) = \begin{cases} \frac{0}{2} & i < 0 \\ \frac{1}{2} & i = 0 \\ 1 & i \ge 1 \end{cases} \qquad P(\underline{I}, \langle 0 \rangle) = 0$$

$$P_{\underline{1}}(i) = P(\underline{I}, \langle 0 \rangle) = 1$$

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# Example

Compute: 
$$P_4(3)$$
 =  $\frac{1}{2}P_3(-1) + \frac{1}{2}P_3(3)$   
 $= P_{H_0}(\frac{4}{2}i\Gamma_i \le 3)$  =  $0 + \frac{1}{2}[P_2(0) + P_2(3)]$   
 $\frac{1}{2}[P_1(-2) + P_1(0) + P_1(1) + P_2(3)]$ 

#### HW

• How to extend paired-t-test to the non-parametric case?

#### Are two distributions equal?

- Let F and G be two continuous distributions of unknown form
- Given
  - n samples  $X_1, \dots, X_n$  from F  $\sim$
  - m samples  $Y_1, \dots, Y_m$  from G
- Null hypothesis:  $H_0$ : F = G
- Test is called: Rank-sum test, Mann-Whitney test, Wilcoxon test

Rank order the n+m items.

$$R_i$$
 = rank of the data value  $X_i$ 

Test statistic:

$$T = \sum_{i=1}^{n} R_i$$

**Example 12.4.a.** An experiment designed to compare two treatments against corrosion yielded the following data in pieces of wire subjected to the two treatments.

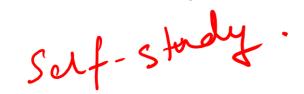
Treatment 1 65.2, 67.1, 69.4, 78.2, 74, 80.3 Treatment 2 59.4, 72.1, 68, 66.2, 58.5

(The data represent the maximum depth of pits in units of one thousandth of an inch.) The ordered values are 58.5, 59.4, 65.2\*, 66.2, 67.1\*, 68, 69.4\*, 72.1, 74\*, 78.2\*, 80.3\* with an asterisk noting that the data value was from sample 1. Hence, the value of the test statistic is T = 3 + 5 + 7 + 9 + 10 + 11 = 45.

## Distribution of test-statistic under the null hypothesis $P_{H_0}(T)$

• Again we will compute recursively.

• Let 
$$P(n, m, t) = P_{H_0}(T \le t)$$



Either the last item in the rank is one of the N  $X_i s$ , or it is one of the M  $Y_i s$ . Under the null hypothesis, this probability:

$$P(N, M, K) = \frac{N}{N+M} P(N-1, M, K-N-M) + \frac{M}{N+M} P(N, M-1, K)$$

Starting with the boundary condition

$$P(1,0,K) = \begin{cases} 0 & K \le 0 \\ 1 & K > 0 \end{cases}, \qquad P(0,1,K) = \begin{cases} 0 & K < 0 \\ 1 & K \ge 0 \end{cases}$$

**Example 12.4.b.** Suppose we wanted to determine P(2, 1, 3). We use Equation (12.4.3) as follows:

$$P(2, 1, 3) = \frac{2}{3}P(1, 1, 0) + \frac{1}{3}P(2, 0, 3)$$

and

$$P(1, 1, 0) = \frac{1}{2}P(0, 1, -2) + \frac{1}{2}P(1, 0, 0) = 0$$

$$P(2, 0, 3) = P(1, 0, 1)$$

$$= P(0, 0, 0) = 1$$

**Example 12.4.a.** An experiment designed to compare two treatments against corrosion yielded the following data in pieces of wire subjected to the two treatments.

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(The data represent the maximum depth of pits in units of one thousandth of an inch.) The ordered values are 58.5, 59.4, 65.2\*, 66.2, 67.1\*, 68, 69.4\*, 72.1, 74\*, 78.2\*, 80.3\* with an asterisk noting that the data value was from sample 1. Hence, the value of the test statistic is T = 3 + 5 + 7 + 9 + 10 + 11 = 45.

$$P(6,5,45) = \frac{6}{11}P(5,5,34) + \frac{5}{11}P(6,4,45) = \cdots$$

#### Wilcoxon rank sum test

data: x and yW = 24, p-value = 0.1255

#### Errors in Hypothesis testing

- Type-I error: Rejecting  $H_0$  even when  $H_0$  is true.
  - ullet The probability with which it happens is called significant level lpha
- Type-II error: Accepting  $H_0$  when it is false

#### Summary of hypothesis testing

#### Follow this framework:

- Formulate null and alternative hypothesis
- Collect data
- Decide on test statistic
- Identify distribution of test statistic under null hypothesis
- Apply p-value or critical region test to accept or reject null hypothesis
- We applied this framework on
  - Mean of Gaussian with unknown variance is  $\mu_0$
  - Are means of two normal distributions with shared unknown variance same?
  - Difference in means of two normal with unknown variance from paired observations

#### Summary...

- Parameter p of Bernoulli is  $p_0$
- Non-parametric tests
  - Median is a given value
  - Distribution is symmetric around a median
  - Are two distributions equal

# Topics not covered.

- Goodness of fit tests
- Test on sequences