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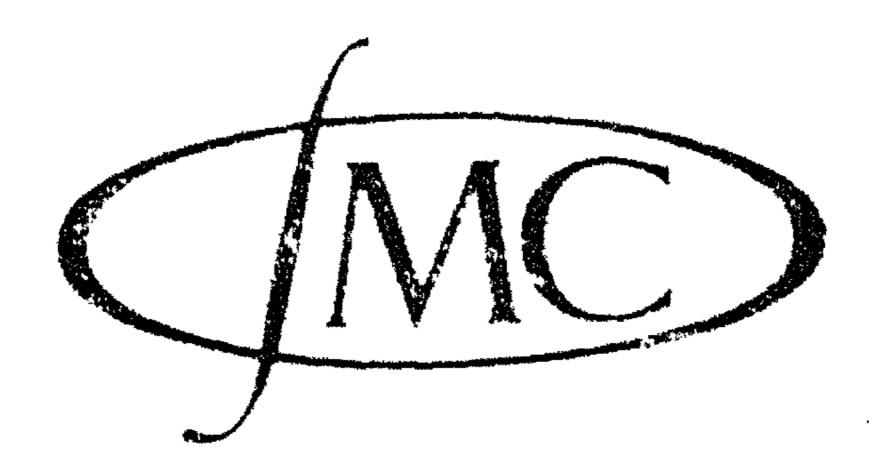
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A class of tests for the hypothesis that K parameters  $\theta_1,\ldots,\theta_k$  satisfy the inequalities  $\theta_1,\ldots,\xi_k$ .

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# A class of tests for the hypothesis that K parameters $\theta_1, ..., \theta_k$ satisfy the inequalities $\theta_1 \leqslant ... \leqslant \theta_k$

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## A CLASS OF TESTS FOR THE HYPOTHESIS THAT K PARAMETERS

 $\theta_1, ..., \theta_k$  SATISFY THE INEQUALITIES  $\theta_1 \leqslant ... \leqslant \theta_k$  (\*)

by

Dr. Constance van EEDEN

#### 1. - INTRODUCTION

In this paper a description will be given of a class of tests treated in chapter 4 of my thesis [4]. By means of these tests the hypothesis  $H_0$  that k parameters  $\theta_1, \ldots, \theta_k$  satisfy the inequalities

$$(1.1) \theta_1 \leq \ldots \leq \theta_k$$

may be tested against the alternative hypothesis that at least one value of i exists with  $\theta_i > \theta_{i+1}$ .

In the chapters 1-3 of my thesis a related problem is treated namely the problem of estimating k unknown parameters  $\theta_1, \ldots, \theta_k$ , known to satisfy

(1.2)
$$\begin{cases}
1. \text{ inequalities of the type} : \varphi_i(\theta_i) \leq \varphi_j(\theta_j), \\
2. \text{ inequalities of the type} : c_i \leq \varphi_i(\theta_i) \leq d_i,
\end{cases}$$

where, for each i = 1, ..., k,  $\varphi_i(\theta_i)$  is a given function of  $\theta_i$ , whereas  $c_i$  and  $d_i$  are given numbers. A special case of this problem is e.g. the estimation of k parameters  $\theta_1, ..., \theta_k$ , known to satisfy the equalities  $\theta_1 \leqslant ... \leqslant \theta_k$ .

A description of this estimation problem and its solution has been given by J. HEMELRIJK [5]. The proofs may be found in [4].

A description of the class of tests for the hypothesis (1.1) will be given in this paper in section 2. Section 3 contains the special cases where  $\theta_i$  is

- 1. the parameter of an exponential distribution,
- 2. the variance of a normal distribution,
- 3. the mean of a normal distribution with known variance,
- 4. the length of the interval of a rectangular distribution.
- (\*) Report SP 65 of the Statistical Department of the Mathematical Centre, Amsterdam.

Further an analogous distributionfree test, based on WILCOXON's two sample test, will be described.

In this paper no proofs will be given; these may be found in [4].

#### 2. — DESCRIPTION OF THE TESTS

The situation to be considered may be described as follows. Let  $\underline{x}_1, \ldots, \underline{x}_k^{-1}$ ) be k independent random variables and let, for each  $i = 1, \ldots, k$ ,  $x_{i,\gamma}(\gamma = 1, \ldots, n_i)$  be  $n_i$  independent observations of  $\underline{x}_i$ . Let further, for each  $i = 1, \ldots, k$ ,  $\theta_i$  denote an unknown parameter of the distribution of  $\underline{x}_i$ . The hypothesis

$$(2.1) H_0: \theta_1 \leqslant \ldots \leqslant \theta_k$$

will be tested against the alternative hypothesis

(2.2) 
$$H$$
: at least one value of i exists with  $\theta_i > \theta_{i+1}$ .

This test is performed as follows. Let, for each  $i = 1, ..., k - 1, T_i$  denote a test for the hypothesis

$$(2.3) H_{0,i}: \theta_i \leqslant \theta_{i+1}$$

against the alternative hypothesis

$$(2.4) H_i: \theta_i > \theta_{i+1}.$$

Let, for each i = 1, ..., k-1,  $t_i$  denote the test statistic and  $Z_i$  the critical region of this test. Then  $t_i$  is a function of  $x_{i,1}, ..., x_{i,n_i}, x_{i+1,1}, ..., x_{i+1,n_{i+1}}$  and  $H_{0,i}$  is rejected if and only if  $t_i \in Z_i$ .

The test for the hypothesis  $H_0$  then consists of rejecting  $H_0$  if and only if a value of i exists with  $t_i \in Z_i$ .

Now suppose that the tests  $T_1, ..., T_{k-1}$  possess the following properties. Let

(2.5) 
$$\begin{cases} \alpha_{i} \stackrel{\text{def}}{=} P\{\underline{t}_{i} \in Z_{i} \mid \theta_{i} = \theta_{i+1}\},^{2}\} \\ N_{i} \stackrel{\text{def}}{=} n_{i} + n_{i+1} \end{cases}$$

and let, for each i=1,...,k-1, the limit  $N_i\to\infty$  be taken under the conditions

$$\begin{cases}
\lim_{N_i \to \infty} n_i = \infty, \\
N_i \to \infty
\end{cases}$$

$$\lim_{N_i \to \infty} n_{i+1} = \infty,$$

then we suppose that, for each i = 1, ..., k-1,

- 1) Random variables are distinguished from numbers (e.g. from the values they take in an experiment) by underlining their symbols.
- <sup>2</sup>)  $P\{A\}$  denotes the probability of event A.

$$\left\{\begin{array}{ll} 1. & \mathrm{P}\{\underline{t}_{i} \ \varepsilon \ Z_{i} \ | \ \theta_{i} < \theta_{i+1}\} \leqslant \alpha_{i}, \\ \\ 2. & \lim_{N_{i} \to \infty} \mathrm{P}\{\underline{t}_{i} \ \varepsilon \ Z_{i} \ | \ \theta_{i} < \theta_{i+1}\} = 0, \\ \\ 3. & \lim_{N_{i} \to \infty} \mathrm{P}\{\underline{t}_{i} \ \varepsilon \ Z_{i} \ | \ \theta_{i} > \theta_{i+1}\} = 1. \end{array}\right.$$

Now it may easily be proved (cf. [4]) that the test for the hypothesis  $H_0$  possesses the following properties. Let  $\alpha_0$  denote the size of the critical region of the test for  $H_0$  (i.e. let  $\alpha_0$  denote the probability, if  $H_0$  is true, of rejecting  $H_0$ ), let

$$n \stackrel{\text{def}}{=} \sum_{i=1}^{k} n_i$$

and let the limit  $n \rightarrow \infty$  be taken under the conditions

(2.9) 
$$\lim_{n\to\infty} n_i = \infty \text{ for each } i=1, ..., k,$$

then we have

$$(2.10) \begin{cases} 1. & \alpha_0 \leqslant \sum\limits_{i=1}^{k-1} \alpha_i, \\ 2. & \text{the probability of rejecting $H_0$, under the hypothesis} \\ \theta_1 < \ldots < \theta_k, \text{ tends to zero for $n \to \infty$,} \end{cases}$$

$$(3. & \text{the probability of rejecting $H_0$, under the hypothesis $H$, tends to 1 for $n \to \infty$.}$$

If, moreover, we suppose that, for each pair of values (i,j) with i < j

$$(2.11) \quad \mathbf{P}\{\underline{t}_{i} \ \varepsilon \ Z_{i} \ \text{and} \ \underline{t}_{j} \ \varepsilon \ Z_{j} \ | \ \theta_{i} = \theta_{i+1}, \ \theta_{j} = \theta_{j+1}\} \leqslant \\ \leqslant \mathbf{P}\{\underline{t}_{i} \ \varepsilon \ Z_{i} \ | \ \theta_{i} = \theta_{i+1}\} \cdot \mathbf{P}\{\underline{t}_{j} \ \varepsilon \ Z_{j} \ | \ \theta_{j} = \theta_{j+1}\},$$

then we have also (cf. [3] and [4])

$$\left\{ \begin{array}{l} \text{the probability of rejecting $H_0$, under the hypothesis} \\ \theta_1 = \ldots = \theta_k \text{, is } \geqslant \sum\limits_{i=1}^{k-1} \alpha_i - \frac{1}{2} \, \big\{ \sum\limits_{i=1}^{k-1} \alpha_i \big\}^2. \end{array} \right.$$

Thus if we take e.g.  $\sum_{i=1}^{k-1} \alpha_i = 0.05$  then we have

- 1. the probability of rejecting  $H_0$ , if  $H_0$  is true, is  $\leq 0.05$ ,
- 2. the probability of rejecting  $H_0$ , under the hypothesis

$$\theta_1 = \dots = \theta_k$$
, is  $\geq 0.05 - \frac{1}{2}(0.05)^2 = 0.04875$ .

Tests  $T_i$  satisfying the conditions (2.7) and (2.11) will be described in section 3.

#### 3. — EXAMPLES

#### 3.1. An exponential distribution with parameter \theta\_i

We first consider the case that  $\underline{x}_i$  possesses, for each i = 1, ..., k, an exponential distribution with parameter  $\theta_i$ , i. e.

(3.1.1) 
$$P\{z_{i} \leq x\} = 1 - e^{-a_{i}x} \quad (x \geq 0).$$

Now let, for each i = 1, ..., k,

then we take, for each i=1,...,k-1, as a test statistic for the hypothesis  $H_{0,i}$ 

$$(3.1.3) t_i = \frac{\bar{x}_{i+1}}{\bar{x}_i}$$

and for  $Z_i$  we take a critical region of the form  $t_i \ge t_{i,\alpha_i}$  where [cf. (2.5)]  $t_{i,\alpha_i}$  satisfies

$$(3.1.4) P\{\underline{t}_i \geq \underline{t}_{i,\alpha_i} \mid \theta_i = \theta_{i+1}\} = \alpha_i.$$

Now (3.1.1) entails that for each i=1, ..., k,  $2\theta_i n_i x_i$  possesses a  $\chi^2$ -distribution with  $2n_i$  degrees of freedom, thus  $\underline{t}_i$  possesses, for each i=1, ..., k-1, under the hypothesis  $\theta_i = \theta_{i+1}$ , an F-distribution with  $2n_{i+1}$ , and  $2n_i$  degrees of freedom. Thus the critical values  $t_{i,\alpha_i}$  may be found from a table of the F-distribution.

It may easily be proved (cf. [4]) that these tests  $T_1, ..., T_{k-1}$  satisfy the conditions (2.7) and 2.11).

#### 3.2. — A normal distribution with variance $\theta_i$

Now let, for each  $i = 1, ..., k, \underline{x}_i$  possess a normal distribution with unknown mean  $\mu_i$  and variance  $\theta_i$ . Then, if

$$(3.2.1)$$

$$\begin{cases}
\bar{x}_i \stackrel{\text{def}}{=} \frac{\sum_{\gamma=1}^n x_{i,\gamma}}{\sum_{\gamma=1}^n (x_{i,\gamma} - \bar{x}_i)^2}, \\
s_i^2 \stackrel{\text{def}}{=} \frac{1}{n_i - 1} \sum_{\gamma=1}^n (x_{i,\gamma} - \bar{x}_i)^2,
\end{cases}$$

we take, as a test statistic for the hypothesis  $H_{0,i}$ 

(3.2.2) 
$$t_i = \frac{s_i^2}{s_{i+1}^2} (i = 1, ..., k-1).$$

Now  $\frac{(n_i-1)\underline{s}_i^2}{\theta_i}$  possesses, for each i=1,...,k, a  $\chi^2$ -distribution with  $n_i-1$  degrees of freedom; thus, for each i=1,...,k-1,  $\underline{t}_i$  possesses, under the

hypothesis  $\theta_i = \theta_{i+1}$ , an F-distribution with  $n_i - 1$  and  $n_{i+1} - 1$  degrees of freedom. We again take critical regions of the form  $t_i \ge t_{i,\alpha_i}$ , where  $t_{i,\alpha_i}$  may be found from a table of the F-distribution.

The proofs of (2.7) and (2.11) are identical with those of the foregoing example.

Remark (3.2.1)

If 
$$\mu_i$$
 is known then  $\underline{s}_i^2$  is replaced by  $\underline{s}_i'^2 \stackrel{\text{def}}{=} \frac{1}{n_i} \sum_{\gamma=1}^{n_i} (\underline{x}_{i,\gamma} - \mu_i)^2$ , where  $\frac{n_i \underline{s}_i'^2}{\theta_i}$  possesses a  $\chi^2$ -distribution with  $n_i$  degrees of freedom.

3.3. — A normal distribution with mean  $\theta_i$  and known variance

We now consider the case that, for each  $i = 1, ..., k, x_i$  possesses a normal distribution with mean  $\theta_i$  and known variance  $\sigma_i^2$ . Let, for each i = 1, ..., k,

then we take

$$(3.3.2) t_i = \bar{x}_{i+1} - \bar{x}_i (i = 1, ..., k-1).$$

The statistic  $t_i$  possesses, under the hypothesis  $\theta_i = \theta_{i+1}$ , a normal distribution with zero mean and variance

(3.3.3) 
$$\sigma^{2}(\underline{t}_{i} \mid \theta_{i} = \theta_{i+1}) = \frac{\sigma_{i}^{2}}{n_{i}} + \frac{\sigma_{i+1}^{2}}{n_{i+1}} \quad (i = 1, ..., k-1).$$

We take a critical region of the form  $t_i \ge t_{i,a}$ ; then

$$(3.3.4) t_{i,a} = \xi_{a} \sqrt{\frac{\sigma_{i}^{2}}{n_{i}}} + \frac{\sigma_{i+1}^{2}}{n_{i+1}},$$

where  $\xi_a$  is defined by

$$\frac{1}{\sqrt{2\pi}} \int_{\epsilon}^{\infty} e^{-\frac{1}{2}x^2} dx = \alpha.$$

Thus  $t_{i,a_i}$  may be found by means of a table of the normal distribution. It may easily be seen that this test satisfies (2.7). Further  $\underline{t}_i$  and  $\underline{t}_j$  are, for j > i + 1, independently distributed, i.e. (2.11) holds for each pair of values (i,j) with j > i + 1. For j = i + 1,  $\underline{t}_i$  and  $\underline{t}_j$  possess a two-dimensional normal distribution with negative correlationcoefficient and it may easily be proved (cf. [2]) that (2.11) holds in this case.

#### 3.4. — A rectangular distribution between 0 and \theta\_i

Finally, let, for each i = 1, ..., k,  $\underline{x}_i$  possess a rectangular distribution between 0 and  $\theta_i > 0$ . Let, for each i = 1, ..., k,

$$z_i \stackrel{\text{def}}{=} \max_{1 \leq \gamma \leq n_i} x_{i,\gamma},$$

then (cf. [4], chapter 2)  $z_i$  is the maximum likelihood estimate of  $\theta_i$ . In this case we take, for i = 1, ..., k-1,

$$t_i = \frac{z_i}{z_{i+1}}$$

with critical regions of the form  $t_i \ge t_{i,\alpha_i}$ .

Now we have (cf. [4])

$$(3.4.3) t_{i,\alpha_i} = \begin{cases} \left(\frac{n_i}{N_i\alpha_i}\right)^{\frac{1}{n_{i+1}}} & \text{if } \alpha_i \leqslant \frac{n_i}{N_i}, \\ \left(\frac{N_i}{n_{i+1}}(1-\alpha_i)\right)^{\frac{1}{n_i}} & \text{if } \alpha_i \leqslant \frac{n_i}{N_i}. \end{cases}$$

The proof of (2.7) and (2.11) may be found in [4].

#### 3.5. — An analogous distributionfree test

In this section an analogous distributionfree test based on WILCOXON's two sample test will be described. Let  $\underline{x}_1, \ldots, \underline{x}_k$  be independent random variables, possessing continuous probability distributions. Let further, for each  $i = 1, \ldots, k, x_{i,1}, \ldots, x_{i,n_i}$  be independent observations of  $\underline{x}_i$  and let (cf. [1])

(3.5.1) 
$$W_{i} \stackrel{\text{def}}{=} \Sigma \sum_{\gamma=1}^{n} sgn(x_{i,\gamma} - x_{i+1,\lambda}), ^{3})$$

where

(3.5.2) 
$$sgn z \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } z > 0, \\ 0 & \text{if } z = 0, \\ -1 & \text{if } z < 0. \end{cases}$$

In the sequel of this section a test will be described for the hypothesis  $H'_{0}$  that  $\underline{x}_{1}, \ldots, \underline{x}_{k}$  possess the same probability distribution. This test is based on  $W_{1}, \ldots, W_{k+1}$  and is performed as follows. Let, for  $i = 1, \ldots, k-1, H'_{0,i}$  denote the hypothesis that  $\underline{x}_{i}$  and  $\underline{x}_{i+1}$  possess the same probability distribution and let  $Z'_{i}$  denote a critical region of the form  $W_{i} \geqslant W_{i,a}$  where

$$(3.5.3) P\{\underline{W}_i \in Z_i' | H_{0,i}'\} = P\{\underline{W}_i \geqslant W_{i,\alpha_i} | H_{0,i}'\} = \alpha_i.$$

If  $U_i$  is the test statistic of WILCOXON's two sample test, according to H.B. MANN and D.R. WHITNEY [6] then  $W_i = 2U_i - n_i n_{i+1}$ .

Then the hypothesis  $H'_0$  is rejected if and only if a value of i exists with  $W_i \in Z'_i$ .

For small values of  $n_i$  and  $n_{i+1}$  the critical values  $W_{i,\alpha_i}$  may be found from a table of the exact probability distribution of  $\underline{W}_i$  under the hypothesis  $H'_{o,i}$  (cf. e.g. [6] and [7]). For large values of  $n_i$  and  $n_{i+1}$   $\underline{W}_i$  is under the hypothesis  $H'_{o,i}$  approximately normally distributed with zero mean and variance

(3.5.4) 
$$\sigma^{2}(\underline{W}_{i} | H'_{0,i}) = \frac{1}{3} n_{i} n_{i+1} (N_{i} + 1).$$

Thus in this case an approximation to  $W_{i,\alpha_i}$  may be found from a table of the normal distribution.

Now let  $\alpha_0$  denote the size of the critical region of the test for  $H'_0$ , i.e. let

(3.5.5) 
$$\alpha_0 \stackrel{\text{def}}{=} P\{\underline{W}_i \varepsilon Z_i' \text{ for at least on value of } i \mid H_o'\}$$

then it may be proved (cf. [4]) that

(3.5.6) 
$$\sum_{i=1}^{k-1} \alpha_i - \frac{1}{2} \left\{ \sum_{i=1}^{k-1} \alpha_i \right\}^2 \leqslant \alpha_0 \leqslant \sum_{i=1}^{k-1} \alpha_i.$$

Further the test for the hypothesis  $H'_{\circ}$  possesses the following properties.

(3.5.7) 
$$\theta_i' \stackrel{\text{def}}{=} \mathbf{P}\{\underline{x}_i > \underline{x}_{i+1}\}$$
  $(i = 1, ..., k-1),$ 

let the limit  $n \rightarrow \infty$  be taken under the conditions

(3.5.8) 
$$\lim_{n\to\infty} n_i = \infty \qquad \text{for each } i = 1, ..., k$$

and let  $H'_1$ ,  $H'_2$  and  $H'_3$  denote the hypotheses

(3.5.9) 
$$\begin{cases} 1. & H'_1 : \text{ for each value of } i : \theta'_i < \frac{1}{2}, \\ 2. & H'_2 : \text{ at least one value of } i \text{ exists with } \theta'_i > \frac{1}{2}, \\ 3. & H'_3 : \left( \text{ for each values of } i : \theta'_i \leqslant \frac{1}{2}, \\ \text{ at least one values of } i \text{ exists with } \theta'_i = \frac{1}{2}. \end{cases}$$

Then we have, (cf. [4]), for  $n \to \infty$ 

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#### RESUME

Des tests pour l'hypothèse  $\theta_1 \leqslant ... \leqslant \theta_k$  concernant k paramètres  $\theta_1, ..., \theta_k$  inconnus.

Soient  $\theta_1$ ,  $\theta_k$  des paramètres inconnus de k lois de distributions. Le problème, dont une solution est donnée ici, est de tester l'hypothèse

$$\theta_1 \leqslant \ldots \leqslant \theta_k$$

contre les hypothèses alternatives qu'il y a au moins un pair  $(\theta_i, \theta_j)$  avec i < j et

$$\theta_i > \theta_j$$
.

Le test se compose d'une série de tests de deux échantillons pour l'hypothèse  $\theta_i \leq \theta_{i+1}$   $(i=1,\ldots,k-1)$ . Le type de ces tests pour deux échantillons dépend de l'information disponible sur la forme des lois de distribution dont les échantillons ont été prélevés.