STICHTING MATHEMATISCH CENTRUM

2e BOERHAAVESTRAAT 49 AMSTERDAM

ZW 1958 - 019

Voordracht in de serie "Actualiteiten"

29 november 1958

A class of tests for the hypothesis that k parameters θ_1,\dots,θ_k satisfy the inequalities $\theta_1{\le}\dots{\le}\theta_k$

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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 θ_1,\ldots,θ_k satisfy the inequalities

$$(1;1) \qquad \qquad \Theta_1 \leq \ldots \leq \Theta_k$$

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

In the chapters 1-3 of my thesis a related problem is treated namely the problem of estimating k unknown parameters Θ_1,\ldots,Θ_k , known to satisfy

(1.2)
$$\begin{cases} \text{1. inequalities of the type: } \mathcal{P}_{i}\left(\Theta_{i}\right) \leq \mathcal{P}_{j}\left(\Theta_{j}\right), \\ \text{2. inequalities of the type: } c_{i} \leq \mathcal{P}_{i}\left(\Theta_{i}\right) \leq d_{i}, \\ \text{where, for each i=1,...,k, } \mathcal{P}_{i}\left(\Theta_{i}\right) \text{ is a given function of } \Theta_{i}, \\ \text{whereas } c_{i} \text{ and } d_{i} \text{ are given numbers. A special case of this } \\ \text{problem is e.g. the estimation of k parameters } \Theta_{1}, \ldots, \Theta_{k}, \\ \text{known to satisfy the equalities } \Theta_{1} \leq \ldots \leq \Theta_{k}. \end{cases}$$

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 0, 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.

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

In this paper no proofs will be given; these may be found in $\begin{bmatrix} 4 \end{bmatrix}$.

2. Description of the tests

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

The hypothesis

$$(2;1) \qquad \qquad \text{H}_{\text{o}} : \text{e}_{1} \leqslant \ldots \leqslant \text{e}_{k}$$

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 $i=1,\ldots,k-1$, $T_{\hat{1}}$ denote a test for the hypothesis

$$(2;3) \quad H_{0,i} : \Theta_i \leq \Theta_{i+1}$$

against the alternative hypothesis

$$(2;4)$$
 $H_{\hat{1}}: \Theta_{\hat{1}} > \Theta_{\hat{1}+1}.$

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 x_{i+

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

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

¹⁾ Random variables are distinguished from numbers (e.g. from the values they take in an experiment) by underlining their symbols.

(2;5)
$$\begin{cases} \propto_{i} \frac{\det f}{t_{i}} P\{\underline{t_{i}} \in Z_{i} | \theta_{i} = \theta_{i+1}\},^{2} \\ N_{i} \frac{\det f}{t_{i}} n_{i} + n_{i+1} \end{cases}$$

and let, for each i=1,...,k-1, the limit N $_{\hat{1}} \longrightarrow \infty$ be taken under the conditions

(2;6)
$$\begin{cases} \lim_{N_{i} \to \infty} n_{i} = \infty, \\ \lim_{N_{i} \to \infty} n_{i+1} = \infty, \end{cases}$$

then we suppose that, for each

$$\begin{cases} 1. & P\{\underline{t}_{\hat{1}} \in Z_{\hat{1}} | \Theta_{\hat{1}} < \Theta_{\hat{1}+1} \} \leq \alpha_{i}, \\ 2. & \lim_{N_{i} \to \infty} P\{\underline{t}_{\hat{1}} \in Z_{\hat{1}} | \Theta_{\hat{1}} < \Theta_{\hat{1}+1} \} = 0, \\ 3. & \lim_{N_{i} \to \infty} P\{\underline{t}_{\hat{1}} \in Z_{\hat{1}} | \Theta_{\hat{1}} > \Theta_{\hat{1}+1} \} = 1. \end{cases}$$

Now it may easily be proved (cf.[4]) that the test for the hypothesis $\mathbf{H}_{\mathbf{0}}$ possesses the following properties. Let $\mathbf{x}_{\mathbf{0}}$ denote the size of the critical region of the test for H $_{
m o}$ (i.e. let $m imes_{
m o}$ denote the probability, if H_0 is true, of rejecting H_0), let

$$(2;8) n \stackrel{\text{def}}{=} \sum_{i=1}^{k} n_i$$

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

(2;9)
$$\lim_{n\to\infty} n = \infty \quad \text{for each } i=1,\ldots,k ,$$

then we have

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

$$(2;11) \ P\{\underline{t}_{\hat{\mathbf{1}}} \in Z_{\hat{\mathbf{1}}} \text{ and } \underline{t}_{\hat{\mathbf{j}}} \in Z_{\hat{\mathbf{j}}} | e_{\hat{\mathbf{1}}} = e_{\hat{\mathbf{1}}+1}, \ e_{\hat{\mathbf{j}}} = e_{\hat{\mathbf{j}}+1}\} \leqslant$$

$$\leqslant P\{\underline{t}_{\hat{\mathbf{1}}} \in Z_{\hat{\mathbf{1}}} | e_{\hat{\mathbf{1}}} = e_{\hat{\mathbf{1}}+1}\} \cdot P\{\underline{t}_{\hat{\mathbf{j}}} \in Z_{\hat{\mathbf{j}}} | e_{\hat{\mathbf{j}}} = e_{\hat{\mathbf{j}}+1}\} ,$$

²⁾ $P\{A\}$ denotes the probability of event A.

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

(2;12)
$$\begin{cases} \text{the probability of rejecting H}_{0}, \text{ under the hypothesis} \\ e_{1} = \dots = e_{k}, \text{ is } \geq \sum_{i=1}^{k-1} \alpha_{i} - \frac{1}{2} \left\{ \sum_{i=1}^{k-1} \alpha_{i} \right\}^{2}. \end{cases}$$

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

- 1. the probability of rejecting $H_{\rm o}$, if $H_{\rm o}$ is true, is \leqslant 0,05,
- 2. the probability of rejecting H_0 , under the hypothesis $\theta_1 = ... = \theta_k$, is $\ge 0.05 \frac{1}{2}(0.05)^2 = 0.04875$.

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 Θ_{i}

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

(3.1;1)
$$P\{\underline{x}_{\hat{1}} \leq x\} = 1 - e^{-\theta_{\hat{1}}x} \quad (x \geq 0)$$
.

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

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

(3.1;3)
$$t_{1} = \frac{x_{1+1}}{\overline{x}_{1}}$$

and for Z_i we take a critical region of the form t_i \geqslant t_i, \bowtie , where (cf.(2;5)) t_i, \bowtie , satisfies

Now (3.1;1) entails that, for each i=1,...,k, $2e_i n_i \overline{x}_i$ possesses a χ^2 -distribution with $2n_i$ degrees of freedom, thus \underline{t}_i possesses, for each i=1,...,k-1, under the hypothesis $e_i = e_{i+1}$ an F-distribution with $2n_{i+1}$ and $2n_i$ degrees of freedom. Thus the critical values $\underline{t}_{i,\alpha}$ may be found from a table of the F-distribution.

It may easily be proved (cf. [4]) that these tests T_1, \dots, T_{k-1} satisfy the conditions (2;7) and (2;11).

3.2 A normal distribution with variance 0;

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

$$\begin{cases}
\overline{x}_{i} \stackrel{\text{def}}{=} \sum_{j=1}^{n_{i}} x_{i,j}, \\
s_{i} \stackrel{\text{def}}{=} \frac{1}{n_{i-1}} \sum_{j=1}^{n_{i}} (x_{i,j} - \overline{x}_{i})^{2},
\end{cases}$$
(1=1,...,k)

we take, as a test statistic for the hypothesis Ho.i

(3.2;2)
$$t_{1} = \frac{s_{i}^{2}}{s_{i+1}^{2}} \quad (i=1,...,k-1).$$

Now $\frac{(n_i-1)\frac{s}{2}^2}{\Theta_i}$ possesses, for each i=1,...,k, a χ^2 -distribution with n_i -1 degrees of freedom; thus, for each i=1,...,k-1, \underline{t}_i possesses, under the hypothesis $\underline{\theta}_{i+1}$ an F-distribution with n_i -1 and n_{i+1} -1 degrees of freedom. We again take critical regions of the form $t_i \geq t_i$, where t_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 μ_i is known then \underline{s}_i^2 is replaced by $\underline{s}_i'^2 \stackrel{\text{def}}{=} \frac{1}{n_i} \sum_{\ell=1}^{n_i} (\underline{x}_{i,\ell} - \mu_i)^2$, where $\frac{n_i \underline{s}_i'^2}{9_i}$ possesses a χ^2 -distribution with n_i degrees of freedom.

3.3 A normal distribution with mean Θ_{i} and known variance

We now consider the case that, for each i=1,...,k, \underline{x}_i possesses a normal distribution with mean θ_i and known variance σ_i^2 . Let, for each i=1,...,k,

$$(3.3;1) \qquad \overline{x}_{\underline{i}} \stackrel{\text{def}}{=} \frac{1}{n_{i}} \sum_{\chi=1}^{n_{i}} \times_{i,\gamma}$$

then we take

(3.3;2)
$$t_{i} = \bar{x}_{i+1} - \bar{x}_{i} \quad (i=1,...,k-1) .$$

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

(3.3;3)
$$\sigma^{2}(\underline{t}_{i}|e_{i} = e_{i+1}) = \frac{\sigma_{i}^{2}}{n_{i}} + \frac{\sigma_{i+1}^{2}}{n_{i+1}} \quad (i=1,...,k-1)$$

We take a critical region of the form
$$t_i \ge t_i, x_i$$
; then
$$(3.3;4) \qquad \qquad t_{i,x_i} = \sum_{\alpha_i} \sqrt{\frac{\sigma_i^2}{n_i} + \frac{\sigma_{i+1}^2}{n_{i+1}}},$$

where ξ is defined by

(3.3;5)
$$\frac{1}{\sqrt{2\pi}} \int_{x}^{\infty} e^{-\frac{1}{2}x^{2}} dx = \infty .$$

 $\frac{1}{\sqrt{2\pi}} \int_{0}^{\infty} e^{-\frac{1}{2}x^2} dx = \infty .$ Thus t_{i,α_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}_i 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}_i possess a two-dimensional normal distribution with negative correlation coefficient and it may easily be proved (cf. $\lceil 2 \rceil$) that (2;11) holds in this case.

3.4 A rectangular distribution between 0 and 0,

Finally, let, for each i=1,...,k, x_i possess a rectangular distribution between 0 and $\theta_i > 0$. Let, for each i=1,...,k,

$$z_{i} \stackrel{\text{def}}{=} \max_{1 \leq i \leq n} x_{i,j},$$

then (cf. $\begin{bmatrix} 4 \end{bmatrix}$, chapter 2) z_i is the maximum likelihood estimate of Θ_{i} . In this case we take, for i=1,...,k-1,

$$(3.4;2)$$
 $t_{\hat{1}} = \frac{z_{\hat{1}}}{z_{\hat{1}+1}}$

with critical regions of the form $t > t_{i, \propto}$. Now we have (cf.[4])

$$(3.4;3) \qquad t_{i,\alpha_{i}} = \begin{cases} \left(\frac{n_{1}}{N_{1}\alpha_{1}}\right)^{\frac{1}{n_{1+1}}} & \text{if } \alpha_{i} \leq \frac{n_{1}}{N_{1}}, \\ \left\{\frac{N_{1}}{n_{1+1}} \left(1-\alpha_{i}\right)^{n_{1}}\right\}^{n_{1}} & \text{if } \alpha_{i} \geq \frac{n_{1}}{N_{1}}. \end{cases}$$

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,...,k, $x_1,1,\ldots,x_n$ be independent observations of \underline{x}_i and let (cf.[1])

$$(3.5;1) \qquad \qquad W_{\hat{1}} \stackrel{\text{def}}{=} \sum_{\lambda=1}^{n_{\hat{1}}} \sum_{\lambda=1}^{n_{\hat{1}+1}} \operatorname{sgn}(x_{\hat{1},\gamma} - x_{\hat{1}+1,\lambda}), \qquad \qquad 3)$$

where

In the sequel of this section a test will be described for the hypothesis H' 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' denote the hypothesis that \underline{x}_i and \underline{x}_{i+1} possess the same probability distribution and let Z' denote a critical region of the form $W_i \geqslant W_i, \alpha,$ where

$$(3.5;3) P\{\underline{W}_{i} \in Z_{i} | H'_{0,i}\} = P\{\underline{W}_{i} \geqslant W_{i,\alpha_{i}} | H'_{0,i}\} = \alpha_{i}.$$

Then the hypothesis H' 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 $W_{i, \propto i}$ may be found from a table of the exact probability distribution of \underline{W}_i under the hypothesis $H_{0,i}^!$ (cf. e.g. [6] and [7]). For large values of n_i and $n_{i+1}\underline{W}_i$ is, under the hypothesis $H_{0,i}^!$, approximately normally distributed with zero mean and variance

(3.5;4)
$$\sigma^{2}(\underline{W}_{1}|H'_{0,1}) = \frac{1}{3} n_{1}n_{1+1}(N_{1}+1).$$

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

Now let α_o denote the size of the critical region of the test for H_o^i , i.e. let

(3.5;5)
$$\alpha_0 \stackrel{\text{def}}{=} P\{\underline{W}_i \in Z_i' \text{ for at least one value of } i \mid H_0'\}$$

³⁾ If U_i is the test statistic of WILCOXON's two sample test, according to H.B. MANN and D.R. WHITNEY $\begin{bmatrix} 6 \end{bmatrix}$, then W_i = 2U_i-n_in_{i+1}.

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

(3.5;6)
$$\sum_{i=1}^{k-1} \alpha_i -\frac{1}{2} \left\{ \sum_{i=1}^{k-1} \alpha_i \right\}^2 \leq \alpha_i \leq \sum_{i=1}^{k-1} \alpha_i .$$

Further the test for the hypothesis H_{\circ}^{r} possesses the following properties. Let

(3.5;7)
$$e_{i}^{!} \stackrel{\text{def}}{=} P\{\underline{x}_{i} > \underline{x}_{i+1}\} \quad (i=1,...,k-1),$$

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

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

$$\begin{cases}
1. & H'_1: \text{ for each value of i}: \theta'_1 < \frac{1}{2}, \\
2. & H'_2: \text{ at least one value of i} \text{ exists with } \theta'_1 > \frac{1}{2}, \\
3. & H'_3: \begin{cases}
\text{ for each value of i}: \theta'_1 \leqslant \frac{1}{2}, \\
\text{ at least one value of i} \text{ exists with } \theta'_1 = \frac{1}{2}.
\end{cases}$$
Then we have, $(\text{cf}, [47])$, for $n \to \infty$

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

(3.5;10) 1. the probability of rejecting H' under the hypothesis H', tends to zero,

2. the probability of rejecting H' under the hypothesis H', tends to 1,

3. if \(\alpha_i \) is sufficiently small for each value of i with $\Theta_1^i = \frac{1}{2}$, the probability of rejecting H' under the hypothesis H' tends to a limit <1.

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