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Maximum likelihood estimation of partially
or completely ordered parameters

by

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1. Introduction

The problem treated in this report concerns the maximum likelihood estimation of partially or completely ordered parameters of probability distributions. A special case of this problem, the maximum likelihood estimation of ordered probabilities, has been treated in [2].

The problem will be formulated in section 2; in section 4 and 5 methods will be given by means of which the estimates may be found. For the proofs of the theorems we need some lemma's which will be proved in section 3 and in section 6 some examples will be given.

2. The problem

Consider k independent random variables x_1, x_2, \dots, x_k ¹⁾ and n_i independent observations $x_{i,1}, x_{i,2}, \dots, x_{i,n_i}$ of x_i ($i = 1, 2, \dots, k$). The distribution of x_i contains one unknown parameter θ_i ($i = 1, 2, \dots, k$) and its distribution function is

$$(2.1) \quad F_i(x_i | \theta_i) \stackrel{\text{def}}{=} P[x_i \leq x_i | \theta_i] \quad (i = 1, 2, \dots, k).$$

Two types of restrictions are imposed on the parameters $\theta_1, \theta_2, \dots, \theta_k$. First let \mathcal{U}_i be a closed interval such that $F_i(x_i | y_i)$ is a distribution function for each value of $y_i \in \mathcal{U}_i$ ($i = 1, 2, \dots, k$). By means of the choice of \mathcal{U}_i restrictions of the type $c_i \leq \theta_i \leq d_i$ may be imposed. The second type of restrictions consists of a partial or complete ordering of the parameters, which may be described as follows. Let $\alpha_{i,j}$ ($i, j = 1, 2, \dots, k$) be numbers satisfying the conditions

$$(2.2) \quad \begin{cases} 1. \alpha_{i,j} = -\alpha_{j,i}, \\ 2. \alpha_{i,j} = 0 \text{ if the intersection } \mathcal{U}_i \cap \mathcal{U}_j \text{ contains at most} \\ \quad \text{one point,} \\ 3. \alpha_{i,j} = 0, +1 \text{ or } -1 \text{ in all other cases} \end{cases}$$

and

$$(2.3) \quad \alpha_{i,j} = 1 \text{ if } \alpha_{i,h} = \alpha_{h,j} = 1 \text{ for any } h.$$

The restrictions imposed on $\theta_1, \theta_2, \dots, \theta_k$ are then

$$(2.4) \quad \begin{cases} 1. \alpha_{i,j} (\theta_i - \theta_j) \leq 0 \\ 2. \theta_i \in \mathcal{U}_i \end{cases} \quad (i, j = 1, 2, \dots, k).$$

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

and it will be supposed that the parameters $\theta_1, \theta_2, \dots, \theta_k$ are numbered in such a way that

$$(2.5) \quad \alpha_{i,j} \geq 0 \text{ for each pair of values } (i,j).$$

No other restrictions on $\theta_1, \theta_2, \dots, \theta_k$ are admitted, such that all points y_1, y_2, \dots, y_k of the Cartesian product

$$(2.6) \quad G \stackrel{\text{def}}{=} \prod_{i=1}^k \mathcal{Y}_i,$$

satisfying

$$(2.7) \quad \alpha_{i,j} (y_i - y_j) \leq 0 \quad (i,j = 1, 2, \dots, k)$$

belong to the parameterspace, which thus is a convex subdomain of G . This subdomain will be denoted by D .

Let

$$(2.8) \quad \begin{cases} 1. \alpha_{i,j} = 0 \text{ for } \nu_0 \text{ pairs of values } (i,j) \text{ with } i < j, \\ 2. \alpha_{i,j} = 1 \text{ for } \nu_1 \text{ pairs of values } (i,j) \text{ with } i < j, \end{cases}$$

then

$$(2.9) \quad \nu_0 + \nu_1 = \binom{k}{2}.$$

Let further $f_i(x_i | \theta_i)$ denote the density function of x_i if x_i possesses a continuous probability distribution and $P[x_i = x_i | \theta_i]$ if x_i possesses a discrete probability distribution and let

$$(2.10) \quad \begin{cases} 1. L_i = L_i(y_i) \stackrel{\text{def}}{=} \sum_{x_i=1}^{m_i} \lg f_i(x_i | y_i) \quad (i=1, 2, \dots, k), \\ 2. L = L(y_1, y_2, \dots, y_k) \stackrel{\text{def}}{=} \sum_{i=1}^k L_i(y_i). \end{cases}$$

Then the maximum likelihood estimates of $\theta_1, \theta_2, \dots, \theta_k$ are the values of y_1, y_2, \dots, y_k which maximize L in the domain D .

Unless explicitly stated otherwise L will only be considered in this domain D ; the maximum likelihood estimates will throughout this paper be denoted by t_1, t_2, \dots, t_k .

Further the restrictions $\theta_i \leq \theta_j$ (i.e. $\alpha_{i,j} = 1$) satisfying

$$(2.11) \quad \alpha_{i,h} \cdot \alpha_{h,j} = 0 \text{ for each } h \text{ between } i \text{ and } j$$

will be denoted by R_1, R_2, \dots, R_s . Each R_x thus corresponds with one pair (i,j) ; this pair will be denoted by (i_x, j_x) .

Because of the transitivity relations (2.3) the system R_1, R_2, \dots, R_s is equivalent to (2.4.1) and uniquely determined by (2.4.1).

The restrictions R_1, R_2, \dots, R_s will be called the essential restrictions.

Remark 1: H.D. BRUNK [1] described a method by means of which the estimates of $\theta_1, \theta_2, \dots, \theta_k$ may be found if the distribution of x_i belongs to the "exponential family" ($i=1, 2, \dots, k$) and if moreover \mathcal{Y}_i is the set of all values of y_i for which $F_i(x_i | y_i)$ is a distribution function ($i=1, 2, \dots, k$). His method however leads to much more complicated computations than ours.

3. Lemma's

Definition: A function $\varphi(y)$ of a variable y will be called strictly unimodal in an interval \mathcal{Y} if there exists a value $y^* \in \mathcal{Y}$ such that

$$(3.1) \quad \varphi(y) < \varphi(z) < \varphi(y^*)$$

for each pair of values $(y, z) \in \mathcal{Y}$ with

$$(3.2) \quad y < z < y^*$$

and for each pair of values $(y, z) \in \mathcal{Y}$ with

$$(3.3) \quad y^* < z < y.$$

It follows at once from this definition that a strictly unimodal function $\varphi(y)$ is bounded in every closed subdomain of \mathcal{Y} not containing y^* .

Now let $\varphi_\kappa(y_\kappa)$ be a strictly unimodal function of y_κ in the interval \mathcal{Y}_κ ($\kappa=1, 2, \dots, k$) and let further

$$(3.4) \quad \Phi(y_1, y_2, \dots, y_k) \stackrel{\text{def}}{=} \sum_{\kappa=1}^k \varphi_\kappa(y_\kappa),$$

then

Lemma I: $\Phi(y_1, y_2, \dots, y_k)$ possesses a unique maximum in

$$(3.5) \quad \Gamma \stackrel{\text{def}}{=} \prod_{\kappa=1}^k \mathcal{Y}_\kappa.$$

Proof: Let $\varphi_\kappa(y_\kappa)$ attain its maximum in \mathcal{Y}_κ for $y_\kappa = y_\kappa^*$ ($\kappa=1, 2, \dots, k$). Then it follows from the fact that $\Phi(y_1, y_2, \dots, y_k)$ is the sum of the k functions $\varphi_\kappa(y_\kappa)$ and that Γ is the Cartesian product of the k intervals \mathcal{Y}_κ , that $\Phi(y_1, y_2, \dots, y_k)$ possesses a unique maximum in Γ and attain this maximum for $y_\kappa = y_\kappa^*$ ($\kappa=1, 2, \dots, k$).

We now define a function V as follows.

Let $y_1^0, y_2^0, \dots, y_k^0$ be a given point in Γ with $y_\kappa^0 \neq y_\kappa^*$ for at least one value of κ and let

$$(3.6) \quad \begin{cases} y_\kappa(\beta) \stackrel{\text{def}}{=} (1-\beta)y_\kappa^0 + \beta y_\kappa^* & (\kappa=1, 2, \dots, k), \\ 0 \leq \beta \leq 1. \end{cases}$$

Then $\{y_1(\beta), y_2(\beta), \dots, y_k(\beta)\}$ is a point in Γ and V is defined by

$$(3.7) \quad V(\beta) \stackrel{\text{def}}{=} \Phi\{y_1(\beta), y_2(\beta), \dots, y_k(\beta)\}.$$

Lemma II: $V(\beta)$ is a monotone increasing function of β in the interval $0 \leq \beta \leq 1$.

Proof: Consider a value of κ with

$$(3.8) \quad y_\kappa^0 = y_\kappa^*$$

then

$$(3.9) \quad y_\kappa(\beta) = y_\kappa^* \quad \text{for each } \beta \text{ with } 0 \leq \beta \leq 1.$$

Thus in this case we have

$$(3.10) \quad \varphi_\kappa(y_\kappa^0) = \varphi_\kappa\{y_\kappa(\beta)\} = \varphi_\kappa(y_\kappa^*) \quad \text{for each } \beta \text{ with } 0 \leq \beta \leq 1.$$

Now consider a value of κ with

$$(3.11) \quad y_\kappa^0 \neq y_\kappa^*,$$

then it follows from the fact that $\varphi_\kappa(y_\kappa)$ is, in the interval \mathcal{Y}_κ , a strictly unimodal function of y_κ and attain its maximum in \mathcal{Y}_κ for $y_\kappa = y_\kappa^*$ that

$$(3.12) \quad \varphi_\kappa(y_\kappa^0) < \varphi_\kappa\{y_\kappa(\beta_1)\} < \varphi_\kappa\{y_\kappa(\beta_2)\} < \varphi_\kappa(y_\kappa^*)$$

for each pair of values (β_1, β_2) with $0 < \beta_1 < \beta_2 < 1$.

From (3.4) and the fact that there exists at least one value of κ with (3.11) it follows then that

$$(3.13) \quad V(0) < V(\beta_1) < V(\beta_2) < V(1)$$

for each pair of values (β_1, β_2) with $0 < \beta_1 < \beta_2 < 1$.

Lemma III: If C is a closed convex subdomain of Γ , not containing the point $(y_1^*, y_2^*, \dots, y_k^*)$, then $\Phi(y_1, y_2, \dots, y_k)$ attains its maximum in C only in one or more points on its border.

Proof: Consider any inner point $y_1^0, y_2^0, \dots, y_k^0$ of C and let $y_\kappa(\beta)$ be defined by (3.6) ($\kappa=1, 2, \dots, k$). Then, C being a closed convex domain not containing the point $(y_1^*, y_2^*, \dots, y_k^*)$ there exists a value of β in the interval $0 < \beta < 1$, say β_0 ,

such that $\{y_1(\beta_0), y_2(\beta_0), \dots, y_k(\beta_0)\}$ is a border point of C . Further it follows from Lemma II that

$$(3.14) \quad \Phi\{y_1(\beta_0), y_2(\beta_0), \dots, y_k(\beta_0)\} > \Phi(y_1^0, y_2^0, \dots, y_k^0).$$

Thus for each inner point $(y_1^0, y_2^0, \dots, y_k^0)$ of C there exists a border point (y_1, y_2, \dots, y_k) of C with a larger value of Φ . Moreover Φ is bounded in C , because the point $(y_1^*, y_2^*, \dots, y_k^*)$ is not contained in C . Thus Φ has a maximum in C , which can evidently only be attained in border points.

4. The maximum likelihood estimates of $\theta_1, \theta_2, \dots, \theta_k$

Let M be a subset of the numbers $1, 2, \dots, k$; let further

$$(4.1) \quad \mathcal{Y}_M \stackrel{\text{def}}{=} \bigcap_{i \in M} \mathcal{Y}_i$$

and if $\mathcal{Y}_M \neq \emptyset$

$$(4.2) \quad L_M(z) \stackrel{\text{def}}{=} \sum_{i \in M} L_i(z) \quad z \in \mathcal{Y}_M.$$

Throughout this report it will be supposed that the following condition is satisfied

(4.3) Condition: For each M with $\mathcal{Y}_M \neq \emptyset$ the function $L_M(z)$ is strictly unimodal in the interval \mathcal{Y}_M .

Now let M_ν ($\nu = 1, 2, \dots, N$) be subsets of the numbers $1, 2, \dots, k$ with

$$(4.4) \quad \begin{cases} 1. \bigcup_{\nu=1}^N M_\nu = \{1, 2, \dots, k\}, \\ 2. M_{\nu_1} \cap M_{\nu_2} = \emptyset \text{ for each pair of values } \nu_1, \nu_2 = 1, 2, \dots, N \\ \quad \text{with } \nu_1 \neq \nu_2, \\ 3. \mathcal{Y}_{M_\nu} \neq \emptyset \text{ for each } \nu = 1, 2, \dots, N, \end{cases}$$

where

$$(4.5) \quad \mathcal{Y}_{M_\nu} \stackrel{\text{def}}{=} \bigcap_{i \in M_\nu} \mathcal{Y}_i \quad (\nu = 1, 2, \dots, N).$$

Let further

$$(4.6) \quad G_N \stackrel{\text{def}}{=} \prod_{\nu=1}^N \mathcal{Y}_{M_\nu}$$

and

$$(4.7) \quad L_{M_v}(z_v) \stackrel{\text{def}}{=} \sum_{i \in M_v} L_i(z_v) \quad z_v \in \mathcal{Y}_{M_v} \quad (v=1, 2, \dots, N).$$

Then for all points in G_N $L(y_1, y_2, \dots, y_N)$ reduces to a function of N variables z_1, z_2, \dots, z_N ; we denote this function by $L'(z_1, z_2, \dots, z_N)$ and thus have

$$(4.8) \quad L'(z_1, z_2, \dots, z_N) = \sum_{v=1}^N L_{M_v}(z_v),$$

which is according to (4.3), a sum of strictly unimodal functions.

Theorem I: L possesses a unique maximum in D

Proof: This theorem will be proved by induction.

Let M_1, M_2, \dots, M_N be an arbitrary set of subsets of the numbers $1, 2, \dots, k$ satisfying (4.4) and let

$$(4.9) \quad D_{N,s} \stackrel{\text{def}}{=} D \cap G_N,$$

where s denotes the number of essential restrictions defining D and where G_N is defined by (4.6). Then $D_{N,s}$ is convex and:

for $N=k$ we have $\mathcal{Y}_{M_v} = \mathcal{Y}_v$ ($v=1, 2, \dots, N$), thus $G_N=G$ and $D_{N,s}=D$
 for $s=0$ we have $D=G$ thus $D_{N,0} = G_N$.

We shall say that the function $L'(z_1, z_2, \dots, z_N)$ can be monotonously traced to its maximum in $D_{N,s}$ if

$$(4.10) \quad \left\{ \begin{array}{l} 1. L'(z_1, z_2, \dots, z_N) \text{ possesses a unique maximum in } D_{N,s}, \\ 2. \text{ every point of } D_{N,s} \text{ can be connected with the point in } D_{N,s} \text{ where } L' \text{ assumes its maximum by means of a line in } D_{N,s} \text{ such that } L' \text{ increases monotonously along this line. (Such a line will be called a trace)} \end{array} \right.$$

For $s=0$ $L'(z_1, z_2, \dots, z_N)$ has this property for every set M_1, M_2, \dots, M_N satisfying (4.4) and every N . This follows from the fact that L' is the sum of strictly unimodal functions and that $D_{N,0}$ is the Cartesian product of the intervals \mathcal{Y}_{M_v} ($v=1, 2, \dots, N$), so that the Lemma's I and II may be applied.

Let us now suppose that it has been proved that L' can be monotonously traced to its maximum for all values of $s \leq s_0$ for every set M_1, M_2, \dots, M_N satisfying (4.4) and for every N . We then prove that the same holds for s_0+1 essential restrictions.

Consider, for a given set M_1, M_2, \dots, M_N , satisfying

(4.4), a domain D_{N, s_0+1} and the domain D_{N, s_0} which is obtained by omitting one of the essential restrictions defining D_{N, s_0+1} . Let this be the restriction $R_\lambda : z_{i_\lambda} \leq z_{j_\lambda}$. Then clearly

$$(4.11) \quad D_{N, s_0+1} \subset D_{N, s_0}.$$

Now L' has a unique maximum in D_{N, s_0} , attained in (say) the point T . We first consider the case that T is outside D_{N, s_0+1} . Then an arbitrary point P of D_{N, s_0+1} with $z_{i_\lambda} < z_{j_\lambda}$ can be connected with T by means of a trace in D_{N, s_0} and this trace must contain at least one border point of D_{N, s_0+1} with $z_{i_\lambda} = z_{j_\lambda}$, because within D_{N, s_0+1} we have: $z_{i_\lambda} < z_{j_\lambda}$ and outside $D_{N, s_0+1} : z_{i_\lambda} > z_{j_\lambda}$. The first of these points when following the trace be denoted by U ; then L' assumes a larger value in U than in P . Now U lies in a domain D_{N', s'_0} , where $N' = N - 1$ and $s'_0 \leq s_0$ and L' can thus monotonously be traced from U to its unique maximum in D_{N', s'_0} by means of a trace within D_{N', s'_0} . The trace from P to U in D_{N, s_0+1} and from U to the maximum of L' in D_{N', s'_0} together form a trace from P to the maximum of L' in D_{N, s_0+1} .

Consider next the case where T is a point of D_{N, s_0+1} . Then L' attains a unique maximum in D_{N, s_0+1} in T . If T is the maximum of L' in G_N then, according to Lemma II, L' can be monotonously traced to its maximum from every point of D_{N, s_0+1} by means of a straight line, connecting this point with T . If T is not the maximum of L' in G_N then it follows from Lemma III that T is a border point of D_{N, s_0+1} where at least two z_i from z_1, z_2, \dots, z_N corresponding to an essential restriction for D_{N, s_0+1} are equal. Let this pair be

$$(4.12) \quad z_{i_\mu} = z_{j_\mu},$$

then we consider the domain D'_{N, s_0} which is obtained from D_{N, s_0+1} by omitting the restriction $R_\mu : z_{i_\mu} \leq z_{j_\mu}$ from the essential restrictions defining D_{N, s_0+1} . The maximum of L' in D'_{N, s_0} then exists and the point where it is attained is a point of D'_{N, s_0} with $z_{i_\mu} \geq z_{j_\mu}$. The rest of the proof for this case is then the same as for the first case considered.

Thus L' can be monotonously traced to its maximum in every $D_{N, s}$, one of which is D .

Remark 2: For $s = 0$ and $N = k$ we have $D_{N, s} = G$. Thus L attains a unique maximum in G in a point which will be denoted by v_1, v_2, \dots, v_k .

Theorem II: If t'_1, t'_2, \dots, t'_k are the values of y_1, y_2, \dots, y_k which maximise L in G and under the restrictions $R_1, \dots, R_{\lambda-1}, R_{\lambda+1}, \dots, R_s$ then

$$(4.13) \quad \begin{cases} 1. t_i = t'_i & (i = 1, 2, \dots, k) \text{ if } t'_{i_\lambda} \leq t'_{j_\lambda}, \\ 2. t_{i_\lambda} = t'_{j_\lambda} & \text{if } t'_{i_\lambda} > t'_{j_\lambda}. \end{cases}$$

Proof: The R_λ have not been arranged in a special order, thus we may take without any loss of generality $\lambda = s$. First consider the case that $t'_{i_s} \leq t'_{j_s}$; then t'_1, t'_2, \dots, t'_k satisfy all restrictions R_1, R_2, \dots, R_s ; thus in this case we have

$$(4.14) \quad t_i = t'_i \quad (i = 1, 2, \dots, k).$$

If $t'_{i_s} > t'_{j_s}$ then (4.13.2) may be proved as follows. The domain defined by the essential restrictions R_1, R_2, \dots, R_{s-1} will be denoted by D' . Then for each point (y_1, y_2, \dots, y_k) in D with $y_{i_s} < y_{j_s}$ there exists a trace in D' from the point (y_1, y_2, \dots, y_k) to the point $(t'_1, t'_2, \dots, t'_k)$ and this trace contains a point $(y'_1, y'_2, \dots, y'_k)$ with

$$(4.15) \quad \begin{cases} 1. y'_{i_s} = y'_{j_s}, \\ 2. L(y'_1, y'_2, \dots, y'_k) > L(y_1, y_2, \dots, y_k). \end{cases}$$

Thus, if $t'_{i_s} > t'_{j_s}$, then $L(y_1, y_2, \dots, y_k)$ attains its maximum in D for $y_{i_s} = y_{j_s}$; (4.13.2) then follows from the uniqueness of this maximum

Remark 3:

If

$$(4.16) \quad P[x_i = 1] = \theta_i, \quad P[x_i = 0] = 1 - \theta_i \quad (i = 1, 2, \dots, k)$$

and

$$(4.17) \quad a_i \stackrel{\text{def}}{=} \sum_{x=1}^{m_i} x_{i,x}, \quad b_i \stackrel{\text{def}}{=} n_i - a_i \quad (i = 1, 2, \dots, k)$$

then

$$(4.18) \quad L(y_1, y_2, \dots, y_k) = \sum_{i=1}^k \{ a_i \lg y_i + b_i \lg (1 - y_i) \}.$$

In [2] it has been proved that, if \mathcal{U}_i is the interval $(0, 1)$, this function L satisfies the following condition.

(4.19) Condition: If (y_1, y_2, \dots, y_k) and (z_1, z_2, \dots, z_k) are any two points in G with $y_i \neq z_i$ for at least one value of i and if

$$y_i(\beta) \stackrel{\text{def}}{=} (1-\beta)y_i + \beta z_i \quad (i=1,2,\dots,k)$$

then $L\{y_1(\beta), y_2(\beta), \dots, y_k(\beta)\}$ is a strictly unimodal function of β in the interval $0 \leq \beta \leq 1$.

This condition is stronger than condition (4.3) and the theorems I and II of this report have been proved in [2] by using condition (4.19).

Further if condition (4.19) is satisfied then theorem I of this report may be proved in a more simple way than we did in [2] as follows. Consider any two points

(y_1, y_2, \dots, y_k) and (z_1, z_2, \dots, z_k) in D with $y_i \neq z_i$ for at least one value of i and

$$(4.20) \quad L(y_1, y_2, \dots, y_k) = L(z_1, z_2, \dots, z_k).$$

Then it follows from condition (4.19) that there exists a point (y_1, y_2, \dots, y_k) in D with

$$(4.21) \quad L(y_1, y_2, \dots, y_k) > L(z_1, z_2, \dots, z_k).$$

Thus L possesses a unique maximum in D .

The maximum likelihood estimates of $\theta_1, \theta_2, \dots, \theta_k$ may always be found by repeatedly applying theorem II. This follows from the fact that $L(z_1, z_2, \dots, z_N)$ is a sum of strictly unimodal functions and that $D_{N,S}$ is a convex subdomain of the Cartesian product of the intervals \mathcal{M}_v ($v=1,2,\dots,N$) for each set M_1, M_2, \dots, M_N and each N .

This leads however to a rather complicated procedure which may often be simplified by using one of the theorems of the following section.

5. Some special theorems

The theorems III-VI in this section may be proved in precisely the same way as the theorems II-V in [2].

Theorem III: If $\alpha_{i,j}(v_i - v_j) \leq 0$ for each pair of values (i,j) then

$$(5.1) \quad t_i = v_i \quad (i=1,2,\dots,k).$$

Theorem IV: If l_1, l_2, \dots, l_m is a set of values satisfying

$$(5.2) \quad \alpha_{i,l_1} = \alpha_{i,l_2} = \dots = \alpha_{i,l_m} = 0 \quad \text{for each } i \neq l_1, l_2, \dots, l_m$$

then the maximum likelihood estimates of $\theta_{l_1}, \theta_{l_2}, \dots, \theta_{l_m}$ are the values of $y_{l_1}, y_{l_2}, \dots, y_{l_m}$ which maximize $L_{l_1} + L_{l_2} + \dots + L_{l_m}$ in the domain

$$(5.3) \quad D, \begin{cases} \alpha_{i,j}(y_i - y_j) \leq 0 \\ y_i \in G_i \end{cases} \quad (i, j = 1, 2, \dots, k_m).$$

Theorem V: If for some pair of values (i, j) with $i < j$

$$(5.4) \quad \alpha_{i,j}(v_i - v_j) > 0$$

and

$$(5.5) \quad \begin{cases} 1. \alpha_{i,h} = \alpha_{h,j} = 0 \text{ for each } h \text{ between } i \text{ and } j, \\ 2. \alpha_{h,i} = \alpha_{h,j} \text{ for each } h < i, \\ 3. \alpha_{i,h} = \alpha_{j,h} \text{ for each } h > j, \end{cases}$$

then

$$(5.6) \quad t_i = t_j.$$

Theorem VI: If (i, j) is a pair of values satisfying

$$(5.7) \quad v_i \leq v_j$$

and

$$(5.8) \quad \begin{cases} 1. \alpha_{i,j} = 0, \\ 2. \alpha_{h,i} \leq \alpha_{h,j} \text{ for each } h < i, \\ 3. \alpha_{i,h} \geq \alpha_{j,h} \text{ for each } h > j, \end{cases}$$

then

$$(5.9) \quad t_i \leq t_j.$$

Theorem VII: If (i, j) is a pair of values with

$$(5.10) \quad \alpha_{i,j} = 0,$$

if D' is the subdomain of D where $y_i \leq y_j$ and if $(t'_1, t'_2, \dots, t'_k)$ is the point where L assume its maximum in D' then

$$(5.11) \quad \begin{cases} 1. t_i = t'_i, t_2 = t'_2, \dots, t_k = t'_k & t'_i < t'_j, \\ 2. t_i \geq t_j & t'_i = t'_j. \end{cases}$$

Proof: The proof of this theorem differs from the one given for theorem VI in [2] only in the form of the trace from a point in D' to the maximum in D . This trace which is a straight line in [2], need not be straight now (cf. the proof of theorem II of the present report).

6. Examples

In this section the pooled samples of x_i and x_j will be denoted by $x'_{i,y}$ ($y = 1, 2, \dots, n_i$), where $n'_i = n_i + n_j$.

6.1 x_i possesses a normal distribution with mean θ_i and known variance ($i = 1, 2, \dots, k$).

Without any loss of generality we may suppose that $\sigma^2\{x_i\} = 1$ for all i ; then

$$(6.1.1) \quad L_i(y_i) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} \sum_{\gamma=1}^{n_i} (x_{i,\gamma} - y_i)^2 \quad (i = 1, 2, \dots, k).$$

From (6.1.1) it follows that

$$(6.1.2) \quad \frac{dL_i(y_i)}{dy_i} = \sum_{\gamma=1}^{n_i} (x_{i,\gamma} - y_i) \quad (i = 1, 2, \dots, k),$$

thus, if

$$(6.1.3) \quad m_i \stackrel{\text{def}}{=} \frac{1}{n_i} \sum_{\gamma=1}^{n_i} x_{i,\gamma} \quad (i = 1, 2, \dots, k),$$

then

$$(6.1.4) \quad \frac{dL_i(y_i)}{dy_i} \begin{cases} > 0 & \text{if } y_i < m_i, \\ = 0 & \text{if } y_i = m_i, \\ < 0 & \text{if } y_i > m_i. \end{cases} \quad (i = 1, 2, \dots, k).$$

From (6.1.4) it follows that $L_i(y_i)$ is a strictly unimodal function of y_i in the interval $(-\infty, +\infty)$, thus $L_i(y_i)$ is a strictly unimodal function of y_i in each closed subinterval Y_i of the interval $(-\infty, +\infty)$ ($i = 1, 2, \dots, k$).

Further if $y_i = y_j$ then $L_i(y_i) + L_j(y_j)$ reduces to one term of the form

$$(6.1.5) \quad L_i(y_i) + L_j(y_j) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} \sum_{\gamma=1}^{n_i} (x'_{i,\gamma} - y_i)^2$$

and analogous relations hold if more than two of the y_i are equal. Thus L satisfies condition (4.3).

From (6.1.5) it follows further that if L attain its maximum for $y_i = y_j$ then the two samples of x_i and x_j are to be pooled.

The procedure will now be illustrated by means of the following example.

Suppose $k = 4$, $n_0 = 2$, $n_1 = 4$ and

$$(6.1.6) \quad \alpha_{1,2} = \alpha_{1,3} = \alpha_{3,4} = 1.$$

Let further

(6.1.7)

1	1	2	3	4
$x_{i,j}$	-0,40	1,43	-0,70	0,29
	2,56	1,86	2,61	0
	0,25	0,06	0,79	1,31
	2,87	0,07	0,86	0,15
		1,14	0,14	2,53
		0,29		1,86
		2,57		
		0,85		
	1,21			
$n_i m_i$	5,28	9,48	3,70	6,14
n_i	4	9	5	6
m_i	1,32	1,05	0,74	1,02

and let $\mathcal{Y}_1, \mathcal{Y}_2, \mathcal{Y}_3, \mathcal{Y}_4$ be the intervals

(6.1.8)

i	1	2	3	4
\mathcal{Y}_i	$(-\infty, 1)$	$(-\infty, +\infty)$	$(0,5, \infty)$	$(-\infty, +\infty)$

Then it follows from (6.1.7) and (6.1.8) that the coordinates of the maximum in G are

(6.1.9)

i	1	2	3	4
v_i	1	1,05	0,74	1,02

From (6.1.6) and (6.1.9) it then follows that the pairs $i=3, j=2$ and $i=4, j=2$ satisfy (5.7) and (5.8). Thus according to theorem VI L attains its maximum in D for

(6.1.10) $y_1 \cong y_3 \cong y_4 \cong y_2.$

From (6.1.9), (6.1.10) and theorem V then follows

(6.1.11) $t_1 = t_3.$

In this way the problem is reduced to the case of 3 samples with $\alpha'_0 = 0.$

(6.1.12)

i	1(+3)	4	2
$x_{i,y}$	-0,40	0,29	1,43
	2,56	0	1,86
	0,25	1,31	0,06
	2,87	0,15	0,07
	-0,70	2,53	1,14
	2,61	1,86	0,29
	0,79		2,57
	0,86		0,85
	0,14		1,21
$n_i m_i$	8,98	6,14	9,48
n_i	9	6	9
m_i	0,998	1,02	1,05
γ_i	(0,5, 1)	($-\infty, +\infty$)	($-\infty, +\infty$)
v_i	0,998	1,02	1,05

and

(6.1.13) $\alpha_{1,4} = \alpha_{4,2} = 1.$

From (6.1.11), (6.1.12) and (6.1.13) then follows

(6.1.14) $t_1 = t_3 = 0,998, t_2 = 1,05, t_4 = 1,02.$

6.2. x_i possesses a normal distribution with known mean and variance θ_i ($i = 1, 2, \dots, k$).

We suppose without loss of generality $\mathcal{L} x_i = 0$ ($i = 1, 2, \dots, k$); then

(6.2.1) $L_i(y_i) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} n_i \lg y_i - \frac{1}{2} \frac{\sum_{y=1}^{m_i} x_{i,y}^2}{y_i} \quad (i = 1, 2, \dots, k).$

From (6.2.1) it follows, if

(6.2.2) $S_i^2 \stackrel{\text{def}}{=} \frac{1}{n_i} \sum_{y=1}^{m_i} x_{i,y}^2 \quad (i = 1, 2, \dots, k),$

that

(6.2.3) $\frac{dL_i(y_i)}{dy_i} \begin{cases} > 0 & \text{if } 0 \leq y_i < S_i^2, \\ = 0 & \text{if } y_i = S_i^2 \\ < 0 & \text{if } y_i > S_i^2 \end{cases} \quad (i = 1, 2, \dots, k);$

thus $L_i(y_i)$ is a strictly unimodal function of y_i in the interval $(0, \infty)$.

Further if $y_i = y_j$ then $L_i(y_i) + L_j(y_j)$ reduces to

$$(6.2.4) \quad L_i(y_i) + L_j(y_i) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} n_i \lg y_i - \frac{1}{2} \frac{\sum_{\gamma=1}^{n_i} x_{i,\gamma}^2}{y_i}$$

and analogously for more than two of the y_i equal. Thus L satisfies condition (4.3) and if L attains its maximum for $y_i = y_j$ then the two samples of x_i and x_j are to be pooled. Numerically the method is thus precisely the same as in 6.1, with s_i^2 in stead of m_i .

6.3 x_i possesses a Poisson distribution with parameter θ_i ($i=1,2,\dots,k$)

In this case we have

$$(6.3.1) \quad L_i(y_i) = -n_i y_i + \sum_{\gamma=1}^{n_i} x_{i,\gamma} \lg y_i - \sum_{\gamma=1}^{n_i} \lg x_{i,\gamma}! \quad (i=1,2,\dots,k).$$

From (6.3.1) it follows that, if

$$(6.3.2) \quad m_i \stackrel{\text{def}}{=} \frac{1}{n_i} \sum_{\gamma=1}^{n_i} x_{i,\gamma} \quad (i=1,2,\dots,k),$$

then

$$(6.3.3) \quad \frac{dL_i(y_i)}{dy_i} \begin{cases} > 0 & \text{if } 0 \leq y_i < m_i, \\ = 0 & \text{if } y_i = m_i, \\ < 0 & \text{if } y_i > m_i; \end{cases} \quad (i=1,2,\dots,k)$$

thus $L_i(y_i)$ is a strictly unimodal function of y_i the interval $(0, \infty)$ ($i=1,2,\dots,k$).

Further if $y_i = y_j$ then $L_i(y_i) + L_j(y_j)$ reduces to

$$(6.3.4) \quad L_i(y_i) + L_j(y_i) = -n_i y_i + \sum_{\gamma=1}^{n_i} x'_{i,\gamma} \lg y_i - \sum_{\gamma=1}^{n_i} x'_{i,\gamma}!;$$

thus L satisfies condition (4.3) and if L attains its maximum for $y_i = y_j$ then the two samples of x_i and x_j are to be pooled.

The theorems of the foregoing sections may e.g. also be applied in the following case.

6.4. x_i possesses a normal distribution with mean θ_i and known variance for $i = l_1, l_2, \dots, l_g$ and a Poisson distribution with parameter θ_i for $i \neq l_1, l_2, \dots, l_g$.

Taking $\sigma^2\{x_i\} = 1$ for $i = l_1, l_2, \dots, l_g$ we have

$$(6.4.1) \quad \begin{cases} L_i(y_i) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} \sum_{\gamma=1}^{n_i} (x_{i,\gamma} - y_i)^2 & (i = l_1, l_2, \dots, l_g), \\ L_i(y_i) = -n_i y_i + \sum_{\gamma=1}^{n_i} x_{i,\gamma} \lg y_i - \sum_{\gamma=1}^{n_i} \lg x_{i,\gamma}! & (i \neq l_1, l_2, \dots, l_g). \end{cases}$$

From the sections 6.1 and 6.3 it follows that $L_i(y_i)$ is a strictly unimodal function of y_i in the interval $(-\infty, +\infty)$ for $i = l_1, l_2, \dots, l_g$ and in the interval $(0, \infty)$ for $i \neq l_1, l_2, \dots, l_g$. Further, if $y_i = y_j$, where x_i possesses a normal and x_j a Poisson distribution then $L_i(y_i) + L_j(y_j)$ reduces to

$$(6.4.3) \quad L_i(y_i) + L_j(y_j) = -\frac{1}{2} n_i \lg 2\pi - \frac{1}{2} \sum_{x=1}^{n_i} (x_{i,x} - y_i)^2 - n_j y_i + \sum_{x=1}^{n_j} x_{j,x} \lg y_i - \sum_{x=1}^{n_j} \lg x_{j,x}!$$

It may be proved as follows that $L_{i,j}(y_i) \stackrel{\text{def}}{=} L_i(y_i) + L_j(y_i)$ is a strictly unimodal function of y_i in the interval $(0, \infty)$. We have

$$(6.4.4) \quad \frac{dL_{i,j}(y_i)}{dy_i} = n_i(m_i - y_i) - n_j + \frac{n_j m_j}{y_i}.$$

Thus if $m_i - \frac{n_j}{n_i} \leq 0$ and $m_j = 0$ then

$$(6.4.5) \quad \frac{dL_{i,j}(y_i)}{dy_i} < 0 \text{ for each } y_i > 0$$

and in all other cases

$$(6.4.6) \quad \frac{dL_{i,j}(y_i)}{dy_i} \begin{cases} > 0 & \text{if } 0 \leq y_i < m'_i \stackrel{\text{def}}{=} \frac{1}{2} \left\{ m_i - \frac{n_j}{n_i} + \sqrt{\left(m_i - \frac{n_j}{n_i} \right)^2 + 4 \frac{n_j m_j}{n_i}} \right\}, \\ = 0 & \text{if } y_i = m'_i, \\ < 0 & \text{if } y_i > m'_i. \end{cases}$$

Analogous relations hold if more than two of the y_i are equal. Thus L satisfies condition (4.3).

This case will be illustrated by means of the following example.

Suppose $k = 4$, $r_0 = r_1 = 3$,

$$(6.4.7) \quad \alpha_{1,2} = \alpha_{1,4} = \alpha_{3,4} = 1$$

and $l_1 = 1$, $l_2 = 2$, $g = 2$. Further

(6.4.8)

i	1	2	3	4
$x_{i,j}$	5,38	4,84	4	2
	3,88	3,56	5	7
	4,14	4,40	3	5
	5,36	4,77	3	4
	5,48		4	
$m_i m_i$	24,24	17,57	19	18
m_i	5	4	5	4
m_i	4,85	4,39	3,8	4,5
\mathcal{Y}_i	$(-\infty, 5)$	$(-\infty, +\infty)$	$(0, \infty)$	$(0, 4)$
v_i	4,85	4,39	3,8	4

Then the pairs $i=3, j=2; i=4, j=2$ and $i=3, j=1$ satisfy (5.7) and (5.8). Thus the problem is reduced to the case of the 4 samples (6.4.8) with $\alpha'_0 = 0$ and

(6.4.9) $\alpha'_{3,1} = \alpha'_{1,4} = \alpha'_{4,2} = 1.$

From (6.4.3), (6.4.9) and theorem V then follows

(6.4.10) $t_1 = t_4.$

In this way the problem is reduced to the problem of maximizing the function

(6.4.11) $L'(y_1, y_2, y_3) \stackrel{\text{def}}{=} L(y_1, y_2, y_3, y_1)$

in the domain

(6.4.12) $D' \begin{cases} 0 \leq y_3 \leq y_1 \leq y_2, \\ y_1 \leq 4. \end{cases}$

From (6.4.5) and (6.4.6) it follows that

(6.4.13)

i	3	1	2
m'_i	3,8	4,8	4,39
\mathcal{Y}'_i	$(0, \infty)$	$(0, 4)$	$(-\infty, +\infty)$
v'_i	3,8	4	4,39

Thus

(6.4.14) $t_1 = t_4 = 4, t_2 = 4,39, t_3 = 3,8.$

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