# MATHEMATISCH CENTRUM 2º BOERHAAVESTRAAT 49 A M S T E R D A M STATISTISCHE AFDELING

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Maximum likelihood estimation of ordered probabilities

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#### 1. Introduction

The problem considered in this report concerns k ( $k \ge 2$ ) independent series of independent trials, each trial resulting in a success or a failure. The i-th series consists of  $n_i$  trials with  $\underline{\alpha}_i^{(1)}$  successes and  $\underline{b}_i = n_i - \underline{\alpha}_i$  failures;  $\pi_i$  is the (unknown) probability of a success for each trial of the i-th series ( $i=1,2,\ldots,k$ ) and  $\pi_i,\pi_2,\ldots,\pi_k$  satisfy the inequalities

$$(1.1) \qquad \qquad \pi_1 \leq \pi_2 \leq \ldots \leq \pi_k.$$

In section 2 a method will be described by means of which the maximum likelihood estimates may be found; in section 3 a generalization of the problem will be considered.

#### 2. The maximum likelihood estimates of $\pi_1, \pi_2, \dots, \pi_k$

## 2.1. The likelihood function

The maximum likelihood estimates of  $\pi_1, \pi_2, \dots, \pi_k$  are those values of  $\beta_1, \beta_2, \dots, \beta_k$  which maximize

(2.1.1) 
$$L = L(p_1, p_2, ..., p_k) \stackrel{\text{def}}{=} \sum_{i=1}^{k} \{a_i \log p_i + (n_i - a_i) \log q_i\}$$
 (q:=1-pi)

in the domain

(2.1.2) 
$$D: \begin{cases} p_1 \leq p_2 \leq \ldots \leq p_k, \\ 0 \leq p_i \leq 1 \quad (i = 1, 2, \ldots, k). \end{cases}$$

In this section  $\$  will, unless explicitely stated otherwise, only be considered in this domain  $\$ ; the maximum likelihood estimates will be denoted by  $\$ V,  $\$ V $_{k}$ , and

(2.1.3) 
$$L_i = L_i(p_i) \stackrel{\text{def}}{=} a_i \lg p_i + (m_i - a_i) \lg q_i \quad (i = 1, 2, ..., k).$$

2.2. The estimates for the case that  $\frac{\alpha_i}{n_i} \le \frac{\alpha_{i+1}}{n_{i+1}}$  for each i = 1, 2, ..., k-1

Theorem I: If  $\frac{a_i}{n_i} \le \frac{a_{i+1}}{n_{i+1}}$  for each i = 1, 2, ..., k-1 then  $(2.2.1) \qquad \forall_i = \frac{a_i}{n_i} \quad (i = 1, 2, ..., k).$ 

<u>Proof:</u> This follows immediately from the fact that the maximum of L in D coincides with the maximum of L in the domain:  $0 \le p_i \le 1$  (i = 1, 2, ..., k) if  $\frac{\alpha_i}{m_i} \le \frac{\alpha_{i+1}}{m_{i+1}}$  for each i = 1, 2, ..., k-1.

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

# 2.3. The estimates for the case that $\frac{\alpha_i}{m} > \frac{\alpha_{i+1}}{m}$ for at least one value of i = 1, 2, ..., k-1

In this section the following theorem will be proved.

### Theorem II:

(2.3.1) 
$$V_{i} = V_{i+1} \quad \text{for each i with } \frac{\alpha_{i}}{n_{i}} > \frac{\alpha_{i+1}}{n_{i+1}}.$$

Further a method will be described by means of which the estimates may be found.

For the proofs we need the following lemma and theorem.

#### Lemma I:

(2.3.2) 
$$L_{i}(p_{i}) > L_{i}(p_{i})$$

# if (þ., þ.) is a pair of values satisfying

$$(2.3.3) o \leq p_i < p_i' \leq \frac{a_i}{n_i} or \frac{a_i}{n_i} \leq p_i' < p_i \leq 1.$$

#### Proof:

From (2.1.3) follows

$$\frac{dL_i}{dp_i} = \frac{a_i - n_i p_i}{p_i q_i}.$$

Therefore

(2.3.5) 
$$\frac{dL_{i}}{d\rho_{i}} \begin{cases} > 0 & \text{if } \rho_{i} < \frac{\alpha_{i}}{m_{i}}, \\ = 0 & \text{if } \rho_{i} = \frac{\alpha_{i}}{m_{i}}, \\ < 0 & \text{if } \rho_{i} > \frac{\alpha_{i}}{m_{i}}. \end{cases}$$

and lemma I follows from (2.3.5).

Theorem III: If  $\frac{\alpha_i}{n_i} > \frac{\alpha_{i+1}}{n_{i+1}}$  for any i and if  $p_i, p_2, \dots, p_K$  is any set in  $\mathbb D$  with

$$(2.3.6)$$
  $p_i < p_{i+1}$ 

then a number b exists with

(2.3.7) 
$$p_i \le p \le p_{i+1}$$

which, substituted into  $L(p_1, p_2, \ldots, p_k)$  for  $p_i$  and  $p_{i+1}$ , increases L.

#### Proof:

A number  $\beta$  which, substituted for  $\beta_i$  and  $\beta_{i+i}$  in L , increases L must satisfy the relation

(2.3.8) 
$$L_{i}(p) + L_{i+1}(p) > L_{i}(p_{i}) + L_{i+1}(p_{i+1}).$$

Further the following cases may be distinguished

1.  $p_i < p_{i+1} \le \frac{\alpha_i}{n_i}$ ; in that case we take  $p = p_{i+1}$ , satisfying (2.3.7).

According to lemma I we then have

$$(2.3.9) \qquad L_{i}(p) > L_{i}(p_{i})$$

and b being equal to bit

(2.3.10) 
$$L_{i+1}(p) = L_{i+1}(p_{i+1}).$$

(2.3.8) then follows from (2.3.9) and (2.3.10)

- 2.  $\frac{\alpha_i}{n_i} \le p_i < p_{i+1}$ ; in that case take  $p = p_i$ . In the same way as in case 1 it may be proved that this number p satisfies (2.3.7) and (2.3.8).
- $\frac{3}{n_i} < \frac{\alpha_i}{n_i} < p_{i+1}$ ; then if we take  $p = \frac{\alpha_i}{n_i}$ , p satisfies (2.3.7) and

(2.3.11) 
$$p_i$$

From lemma I and (2.3.11) then follows

(2.3.12) 
$$L_{i}(p) > L_{i}(p_{i}).$$

Further **b** satisfies

and from lemma I and (2.3.13) follows

$$(2.3.14) \qquad \qquad L_{i+1}(p) > L_{i+1}(p_{i+1}).$$

(2.3.8) then follows from (2.3.12) and (2.3.14).

Further it will be clear that if  $p_1, p_2, \ldots, p_k$  is a set in D and p a number satisfying (2.3.7) then  $p_1, \ldots, p_{i-1}, p_i, p_i, p_{i+2}, \ldots, p_k$  is also a set in D . Therefore from theorem III follows

Theorem TV: If  $\frac{a_i}{m_i} > \frac{\alpha_{i+1}}{m_{i+1}}$  for i=i, then the maximum likelihood estimates of  $\pi_1, \ldots, \pi_{i_1}, \pi_{i_1+2}, \ldots, \pi_k$  are those values of  $p_1, \ldots, p_{i_1}, p_{i_1+2}, \ldots, p_k$  which maximize

where

$$\begin{array}{c}
\alpha_{i}' = \alpha_{i} \\
(2.3.16)
\end{array}$$

$$\begin{array}{c}
\alpha_{i_{1}}' = \alpha_{i_{1}} + \alpha_{i_{1}+1} \\
\alpha_{i_{2}}' = \alpha_{i_{1}} + \alpha_{i_{1}+1}
\end{array}$$

$$\begin{array}{c}
\alpha_{i_{1}}' = \alpha_{i_{1}} + \alpha_{i_{1}+1} \\
\alpha_{i_{2}}' = \alpha_{i_{1}} + \alpha_{i_{2}+1}
\end{array}$$

# in the domain

(2.3.17) 
$$D': \begin{cases} p_1 \leq \ldots \leq p_{i_1} \leq p_{i_1+2} \leq \ldots \leq p_{k_1}, \\ 0 \leq p_i \leq 1 \quad (i = 1, \ldots, i_1, i_1 + 2, \ldots, k). \end{cases}$$

In this way the problem is reduced to the case of k-1 series of trials and may then be solved by means of theorem I or reduced to the case of k-2 series of trials by means of theorem IV. This procedure is necessarily finite, k being finite, Therefore it leads to a unique maximum for L.

Theorem II then follows from this uniqueness and the foregoing theorems.

## 2.4. Example

The procedure described in section 2.3 may be illustrated by means of the following example.

Suppose k=4 and

Suppose 
$$k=4$$
 and

$$\begin{pmatrix}
i & 1 & 2 & 3 & 4 \\
a_i & 4 & 3 & 10 & 8 \\
m_i & 10 & 5 & 30 & 15 \\
\frac{a_i}{m_i} & 0.4 & 0.6 & 0.33 & 0.53.
\end{pmatrix}$$

From (2.4.1) and theorem II follows

$$(2.4.2)$$
  $V_2 = V_3.$ 

The problem is then reduced to the case of k-1=3 series of trials with (cf. theorem IV):

From (2.4.3) and theorem II follows

$$(2.4.4)$$
  $V_1 = V_2$ ,

which reduces the problem to the case k-2=2 series of trials with

$$\begin{cases}
\dot{a} & 1 & 4 \\
a_{i}^{"} & 17 & 8 \\
n_{i}^{"} & 45 & 15 \\
\frac{a_{i}^{"}}{n_{i}^{"}} & 0.38 & 0.53.
\end{cases}$$
There form the same T and (0.45) for 3.7

Then from theorem I and (2.4.5) follows

$$(2.4.6)$$
  $V_{1} = 0.58$  ,  $V_{4} = 0.53$ 

and from (2.4.2), (2.4.4) and (2.4.6)

$$(2.4.7)$$
  $V_1 = V_2 = V_3 = 0.38$ ,  $V_4 = 0.53$ .

# 3. A generalization of the problem

The problem treated in the foregoing sections may be generalized as follows:

Suppose the probabilities  $\pi_1, \pi_2, \dots, \pi_k$  satisfy the inequalities

(3.2) 
$$\begin{cases} \alpha_{i,j} = -\alpha_{j,i}, \\ \alpha_{i,j} = 0 \end{cases}$$
 for  $m_0$  pairs of values  $(i,j)$  with  $i < j$ , 
$$\alpha_{i,j} = 1$$
 for  $m_0$  pairs of values  $(i,j)$  with  $i < j$ .

$$(3.3) m_0 + m_1 = {k \choose 2}$$

and, if i < l < j then

$$(3.4) \qquad \alpha_{i,j} = 1 \quad \text{if} \qquad \alpha_{i,\ell} = \alpha_{\ell,j} = 1.$$

If  $m_{1}=0$  then no restriction is imposed on  $\pi_{1},\pi_{2},\ldots,\pi_{k}$  and it is well known that in this case the maximum likelihood estimate of  $\pi_i$  is:  $\frac{\alpha_i}{m_i}$  (i=1,2,...k). Further, if  $m_0=0$  then (3.1) is identical with:  $\pi_{\text{\tiny I}} \leq \pi_{\text{\tiny 2}} \leq \ldots \leq \pi_{\text{\tiny K}}$  and this case has been considered in the foregoing sections. Therefore we suppose

$$(3.5) \qquad \begin{cases} m_1 \ge 1, \\ m_2 \ge 1. \end{cases}$$

Then from (3.3) and (3.5) it follows that

$$(3.6) k \ge 3.$$

In this report only the case k=3 will be considered; the maximum likelihood estimates will be denoted by  $\vee_1$ ,  $\vee_2$ ,  $\vee_3$  and the domain

(3.7) 
$$\begin{cases} \alpha_{i,j} (p_i - p_j) \leq 0 \\ 0 \leq p_i \leq 1 \end{cases}$$

will be denoted by D ..

The following cases may be distinguished (cf. (3.3) and (3.5)).

(3.8) 
$$\begin{cases} 1. & m_1 = 1, m_2 = 2, \\ 2. & m_1 = 2, m_2 = 1. \end{cases}$$

In case (3.8.1) we may suppose, without any loss of generality

$$(3.9) \alpha_{1,2} = \alpha_{1,3} = 0, \ \alpha_{2,3} = 1.$$

It will be clear that in this case

$$(3.10) \qquad \qquad V_1 = \frac{\alpha_1}{m_1}$$

and that the estimates of  $\pi_2$  and  $\pi_3$  may be found by means of the procedure described in section 2.

In the case (3.8.2) we may suppose without any loss of generality

$$(3.11) \qquad \alpha_{1,2} = \alpha_{1,3} = 1 , \quad \alpha_{2,3} = 0$$

and

$$\frac{\alpha_{\lambda}}{n_{\lambda}} \leq \frac{\alpha_{\lambda}}{n_{\lambda}}.$$

Theorem V: If k = 3 and (3.11) and (3.12) are satisfied and if þ., þ., þ. is a set in D. with

$$(3.13)$$
  $p_2 > p_3$ 

then a number 
$$p$$
 exists with

$$\begin{cases}
1. & p_2 \ge p \ge p_3, \\
2. & L_2(p) + L_3(p) > L_2(p_2) + L_3(p_3).
\end{cases}$$

Proof: The proof is analogous to the proof of theorem IV. Here the following cases may be distinguished

1. 
$$p_2 > p_3 \ge \frac{\alpha_2}{n_3}$$
; then take  $p = p_3$ .

2. 
$$\frac{\alpha_2}{n_2} \ge p_2 > p_3$$
; then take  $p = p_2$ ,

2. 
$$\frac{\alpha_2}{m_2} \ge p_2 > p_3$$
; then take  $p = p_2$ ,  
3.  $p_2 > \frac{\alpha_2}{m_2} > p_3$ ; then take  $p = \frac{\alpha_2}{m_2}$ .

Further it will be clear that if  $p_1, p_2, p_3$  is a set in  $\mathfrak{D}$ , with  $p_2 > p_2$  then, for each number p satisfying (3.14.1),  $p_1, p_2 > p_3$ also a set in  $\mathfrak{D}_{\text{\tiny L}}$  . Therefore it follows from theorem V that

Theorem VI: If k=3 and (3.11) and (3.12) are satisfied then the maximum likelihood estimates of  $\pi_1, \pi_2, \pi_3$  are the values of  $p_1, p_2, p_3$  which maximize L in the domain

$$(3.15)$$
  $p_1 \le p_2 \le p_3.$ 

In this way the problem may, for k=3, be reduced to the case treated in section 2.

This may be illustrated by means of the following example.

Suppose k=3.

$$\begin{pmatrix}
\dot{a} & 1 & 2 & 3 \\
a_{i} & 13 & 12 & 6 \\
n_{i} & 20 & 25 & 15 \\
\frac{a_{i}}{m_{i}} & 0,65 & 0,48 & 0,4
\end{pmatrix}$$

$$(3.17)$$
  $\alpha_{1,3} = \alpha_{2,5} = 1$  ,  $\alpha_{1,2} = 0$ .

(3.18) 
$$\begin{cases} \pi_{1}^{'} \stackrel{\text{def}}{=} 1 - \pi_{3}, \\ \pi_{2}^{'} \stackrel{\text{def}}{=} 1 - \pi_{1}, \\ \pi_{3}^{'} \stackrel{\text{def}}{=} 1 - \pi_{2}, \end{cases}$$

then the problem is reduced to the case of 3 series of trials

$$(3.19) \begin{cases} \dot{i} & 1 & 2 & 3 \\ \dot{a_i} & 9 & 7 & 13 \\ \dot{m_i} & 15 & 20 & 25 \\ \frac{\dot{a_i}}{\dot{m_i}} & 0.6 & 0.35 & 0.52 \end{cases}$$
and

(3.20) 
$$\alpha'_{1,2} = \alpha'_{1,3} = 1$$
,  $\alpha'_{2,3} = 0$ .

For these three series of trials (3.11) and (3.12) are satisfied and therefore the estimates of  $\pi_i', \pi_2', \pi_3'$  (denoted by  $v_i', v_2', v_3'$ ) may be found by means of theorem VI. This leads to

$$(3.21)$$
  $V'_1 = V'_2 = 0,46$ ,  $V'_3 = 0,52$ 

and from (3.18) and (3.21) follows

$$(3.22) V_1 = V_2 = 0,54, V_2 = 0,48.$$

The investigation of cases with k>3 is in progress.