

ON THE DISTRIBUTION OF THE NUMBER OF SUCCESSSES IN INDEPENDENT TRIALS¹

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Let S be the number of successes in n independent Bernoulli trials, where p_j is the probability of success on the j th trial. Let $\mathbf{p} = (p_1, p_2, \dots, p_n)$, and for any integer c , $0 \leq c \leq n$, let $H(c | \mathbf{p}) = P\{S \leq c\}$. Let $\mathbf{p}^{(1)}$ be one possible choice of \mathbf{p} for which $E(S) = \lambda$. For any $n \times n$ doubly stochastic matrix Π , let $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$. Then in the present paper it is shown that $H(c | \mathbf{p}^{(1)}) \leq H(c | \mathbf{p}^{(2)})$ for $0 \leq c \leq [\lambda - 2]$, and $H(c | \mathbf{p}^{(1)}) \geq H(c | \mathbf{p}^{(2)})$ for $[\lambda + 2] \leq c \leq n$. These results provide a refinement of inequalities for $H(c | \mathbf{p})$ obtained by Hoeffding [3]. Their derivation is achieved by applying consequences of the partial ordering of majorization.

1. Introduction and summary. Let S be the number of successes in n independent Bernoulli trials, where p_j is the probability of success on the j th trial, $0 \leq p_j \leq 1$. Let

$$(1.1) \quad \mathbf{p} = (p_1, p_2, \dots, p_n),$$

and for any integer c , $0 \leq c \leq n$, let

$$(1.2) \quad H(c | \mathbf{p}) = P\{S \leq c\}.$$

For fixed c , we are interested in the relationship between $H(c | \mathbf{p}^{(1)})$ and $H(c | \mathbf{p}^{(2)})$, where $\mathbf{p}^{(1)}$ and $\mathbf{p}^{(2)}$ each belong to the region

$$(1.3) \quad D_\lambda = \{\mathbf{p} : 0 \leq p_i \leq 1, i = 1, 2, \dots, n; \sum_{i=1}^n p_i = \lambda\}.$$

That is, $\mathbf{p}^{(1)}$ and $\mathbf{p}^{(2)}$ are sequences of probabilities for the independent Bernoulli trials each of which result in an expected number of successes, $E(S)$, equal to λ .

Hoeffding ([3] Theorem 4) has shown that for all $\mathbf{p} \in D_\lambda$,

$$(1.4) \quad 0 \leq H(c | \mathbf{p}) \leq H(c | n^{-1}(\lambda, \lambda, \dots, \lambda)) \quad \text{if } 0 \leq c \leq [\lambda - 2],$$

$$(1.5) \quad H(c | n^{-1}(\lambda, \lambda, \dots, \lambda)) \leq H(c | \mathbf{p}) \leq 1 \quad \text{if } [\lambda + 2] \leq c \leq n,$$

where

$$(1.6) \quad H(c | n^{-1}(\lambda, \lambda, \dots, \lambda)) = \sum_{k=0}^c \binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k},$$

and $[x]$ denotes the greatest integer $\leq x$. Hoeffding ([3] Theorem 4) also

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obtained bounds on $H(c | \mathbf{p})$ for $c = [\lambda - 1]$, $[\lambda]$, and $[\lambda + 1]$. These will be discussed at the end of Section 3.

To motivate the major result of the present paper, let

$$(1.7) \quad \mathbf{p}^*(\lambda) = n^{-1}(\lambda, \lambda, \dots, \lambda)$$

and

$$(1.8) \quad \hat{\mathbf{p}}(\lambda) = (1, 1, \dots, 1, \lambda - [\lambda], 0, 0, \dots, 0),$$

where in $\hat{\mathbf{p}}(\lambda)$ there are $[\lambda]$ ones and $n - [\lambda] - 1$ zeroes. Note that both $\mathbf{p}^*(\lambda)$ and $\hat{\mathbf{p}}(\lambda)$ are elements of D_λ . We have already noted the role $\mathbf{p}^*(\lambda)$ plays in Hoeffding's bounds (1.4) and (1.5), giving the upper bound to $H(c | \mathbf{p})$ for $0 \leq c \leq [\lambda - 2]$ in (1.4) and the lower bound to $H(c | \mathbf{p})$ for $[\lambda + 2] \leq c \leq n$ in (1.5). On the other hand, we have $H(c | \hat{\mathbf{p}}(\lambda)) = 0$ for $0 \leq c \leq [\lambda - 2]$ and $H(c | \hat{\mathbf{p}}(\lambda)) = 1$ for $[\lambda + 2] \leq c \leq n$; these, of course, are the lower and upper bounds to $H(c | \mathbf{p})$ in (1.4) and (1.5) respectively.

Now note that for any $\mathbf{p} \in D_\lambda$, we can write

$$(1.9) \quad \mathbf{p}^*(\lambda) = \mathbf{p}\Pi^*,$$

where Π^* is an $n \times n$ doubly stochastic matrix, all of whose elements are n^{-1} . Also, for any $\mathbf{p} \in D_\lambda$, we can write

$$(1.10) \quad \mathbf{p} = \hat{\mathbf{p}}(\lambda)\Pi(\mathbf{p}),$$

where $\Pi(\mathbf{p})$ is an $n \times n$ doubly stochastic matrix. [This fact follows directly from Lemma 2.1 of Section 2.] It is thus apparent that proof of the following theorem would yield Hoeffding's inequalities (1.4) and (1.5) as corollaries, and would provide a more detailed picture of the behavior of $H(c | \mathbf{p})$ as a function of \mathbf{p} .

THEOREM 1.1. *Let $\mathbf{p}^{(1)} \in D_\lambda$ and suppose that there exists a doubly stochastic $n \times n$ matrix Π for which*

$$(1.11) \quad \mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi.$$

Then $\mathbf{p}^{(2)} \in D_\lambda$ and

$$(1.12) \quad H(c | \mathbf{p}^{(1)}) \leq H(c | \mathbf{p}^{(2)}) \quad \text{if } 0 \leq c \leq [\lambda - 2],$$

and

$$(1.13) \quad H(c | \mathbf{p}^{(1)}) \geq H(c | \mathbf{p}^{(2)}) \quad \text{if } [\lambda + 2] \leq c \leq n.$$

In Section 2, we apply Ostrowski's ([1] Theorem 15) fundamental theorem on majorization to the problem of ordering, over various choices of $\mathbf{p} \in D_\lambda$, the expected values $Eg(S)$ of any function $g(k)$ on $0, 1, 2, \dots, n$. The results obtained in Section 2 are then used in Section 3 to prove Theorem 1.1.

2. Majorization. A $1 \times n$ vector \mathbf{x} is said to *majorize* a $1 \times n$ vector \mathbf{y} if $x_{[1]} \geq y_{[1]}$, $x_{[1]} + x_{[2]} \geq y_{[1]} + y_{[2]}$, \dots , $\sum_{i=1}^{n-1} x_{[i]} \geq \sum_{i=1}^{n-1} y_{[i]}$, and $\sum_{i=1}^n x_{[i]} = \sum_{i=1}^n y_{[i]}$, where the $x_{[i]}$'s and $y_{[i]}$'s are the components of \mathbf{x} and \mathbf{y} , respectively,

arranged in descending order ($x_{[1]} \geq x_{[2]} \geq \dots \geq x_{[n]}$, and similarly for the $y_{[i]}$'s). The relation of majorization to doubly stochastic matrices is given by the following result of Hardy, Littlewood, and Pólya ([2] page 49). (Note: In [1] this result is incorrectly attributed to Karamata.)

LEMMA 2.1. *The vector \mathbf{x} majorizes the vector \mathbf{y} if and only if there exists an $n \times n$ doubly stochastic matrix Π such that $\mathbf{y} = \mathbf{x}\Pi$.*

The following result, originally due to Ostrowski (see [1] pages 30–33), relates majorization to the ordering of the values of functions $F(\mathbf{z})$ over regions of n -dimensional Euclidean space.

LEMMA 2.2. *Let $F(\mathbf{z})$ be a permutation-symmetric function defined on n -dimensional vectors $\mathbf{z} = (z_1, z_2, \dots, z_n)$. For any $i, j, i \neq j$, and all \mathbf{z} in a permutation-symmetric region D , suppose that*

$$(2.1) \quad (z_i - z_j) \left(\frac{\partial F}{\partial z_i} - \frac{\partial F}{\partial z_j} \right) \geq 0.$$

If $\mathbf{x}, \mathbf{y} \in D$, and if \mathbf{x} majorizes \mathbf{y} , then

$$(2.2) \quad F(\mathbf{x}) \geq F(\mathbf{y}).$$

A permutation-symmetric function satisfying (2.1) over a region D is said to satisfy a *Schur condition on D* . It should be remarked that the condition that F be permutation-symmetric (i.e. $F(z_1, z_2, \dots, z_n) = F(z_{\sigma(1)}, z_{\sigma(2)}, \dots, z_{\sigma(n)})$ for all permutations σ , all z_1, \dots, z_n) is incorrectly omitted from the statement of Ostrowski's lemma in [1].

Let $g(k)$ be any function on $0, 1, \dots, n$, and let S be the number of successes in n independent Bernoulli trials, where p_j is the probability of success on the j th trial. Let

$$(2.3) \quad h(\mathbf{p}) = Eg(S),$$

for $\mathbf{p} = (p_1, p_2, \dots, p_n)$. Then $h(\mathbf{p})$ is a permutation-symmetric function defined over the region

$$D = \{\mathbf{z} : 0 \leq z_i \leq 1, i = 1, 2, \dots, n\}.$$

LEMMA 2.3. *For any two components p_i and $p_j, i < j$, of \mathbf{p} ,*

$$(2.4) \quad (p_i - p_j) \left(\frac{\partial h(\mathbf{p})}{\partial p_i} - \frac{\partial h(\mathbf{p})}{\partial p_j} \right) = -(p_i - p_j)^2 \sum_{k=0}^{n-2} f(k | \mathbf{p}^{ij}) \Delta g(k),$$

where for any function $s(k)$ defined on the nonnegative integers

$$(2.5) \quad \Delta s(k) \equiv s(k+2) - 2s(k+1) + s(k)$$

is the second difference of $s(k)$, where \mathbf{p}^{ij} is the $1 \times (n-2)$ vector formed by deleting the i th and j th components of \mathbf{p} , and where

$$(2.6) \quad f(k | \mathbf{p}^{ij}) = \text{probability of } k \text{ successes in the } n-2 \text{ trials} \\ \text{other than trials } i \text{ and } j;$$

for $k = 0, 1, \dots, n-2$.

PROOF. We adopt the convention that $f(k | \mathbf{p}^{ij}) = 0$ for $k < 0$ or $k > n - 2$. Under this convention,

$$(2.7) \quad P\{S = k\} = (1 - p_i)(1 - p_j)f(k | \mathbf{p}^{ij}) + (p_i + p_j - 2p_i p_j)f(k - 1 | \mathbf{p}^{ij}) + p_i p_j f(k - 2 | \mathbf{p}^{ij}).$$

Using (2.7), we find that the left-hand side of (2.4) is

$$(2.8) \quad (p_i - p_j) \left(\frac{\partial h(\mathbf{p})}{\partial p_i} - \frac{\partial h(\mathbf{p})}{\partial p_j} \right) = \sum_{k=0}^n g(k) \Delta f(k - 2 | \mathbf{p}^{ij}).$$

It is now easily shown that the right-hand sides of (2.4) and (2.8) are equal. \square

As a corollary of Lemma 2.3, we can prove a result earlier obtained by Karlin and Novikoff [5].

COROLLARY 2.1. *Suppose that $g(k)$ is convex on $0, 1, \dots, n - 2$, in the sense that $\Delta g(k) \geq 0, k = 0, 1, \dots, n - 2$. If $\mathbf{p}^{(1)} \in D_\lambda$ and if $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$, where Π is any $n \times n$ doubly stochastic matrix, then $\mathbf{p}^{(2)} \in D_\lambda$ and*

$$(2.9) \quad h(\mathbf{p}^{(1)}) \leq h(\mathbf{p}^{(2)}).$$

PROOF. Since $\Delta g(k) \geq 0, k = 0, 1, \dots, n - 2, -h(\mathbf{p})$ satisfies a Schur condition, as can be seen from (2.4). Hence, Lemmas 2.1 and 2.2 imply that $-h(\mathbf{p}^{(1)}) \geq -h(\mathbf{p}^{(2)})$, from which (2.9) immediately follows. \square

Karlin and Novikoff [5] proved Corollary 2.1 in a somewhat different way. Their proof, however, embodies the ideas underlying the usual proof of Lemma 2.2.

From Corollary 2.1 and the arguments in Section 1 relating any $\mathbf{p} \in D_\lambda$ by doubly stochastic matrices to $\mathbf{p}^*(\lambda)$ and $\hat{\mathbf{p}}(\lambda)$, it follows that for any $g(k)$ convex on $0, 1, \dots, n - 2$, and any $\mathbf{p} \in D_\lambda$,

$$(2.10) \quad (1 - \delta)g([\lambda]) + \delta g([\lambda + 1]) \leq h(\mathbf{p}) = Eg(S) \leq \sum_{k=0}^n g(k) \binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k},$$

where $\delta \equiv \lambda - [\lambda]$.

The result (2.10) implies that $E|S - b|^a$, for any $a \geq 1$ and any real number b , is highest over D_λ when S has a binomial distribution with parameters n and $n^{-1}\lambda$ (i.e., $\mathbf{p} = \mathbf{p}^*(\lambda)$), and lowest when

$$\begin{aligned} S &= [\lambda + 1], & \text{with probability } \delta, \\ &= [\lambda], & \text{with probability } 1 - \delta \end{aligned}$$

(i.e., $\mathbf{p} = \hat{\mathbf{p}}(\lambda)$). The upper bound in (2.10) was first obtained (using a different method) by Hoeffding [3]. The lower bound in (2.10) can also be obtained by the methods in Hoeffding's [3] paper.

3. Proof of Theorem 1.1. For fixed integer $c, 0 \leq c \leq n$, let

$$(3.1) \quad \begin{aligned} g_c(k) &= 1 & \text{if } 0 \leq k \leq c \\ &= 0 & \text{if } c + 1 \leq k \leq n. \end{aligned}$$

Then

$$(3.2) \quad H(c | \mathbf{p}) = E(g_c(S)).$$

Note that $g_c(k)$ is not convex on $0, 1, \dots, n - 2$ when $c \leq n - 1$, so that we cannot directly use Corollary 2.1 to prove Theorem 2.1. Instead, we make use of Lemmas 2.1 to 2.3.

First note that for $c \leq n - 1$

$$(3.3) \quad \begin{aligned} \Delta g_c(k) &= 1, & k = c, \\ &= -1, & k = c - 1, \\ &= 0, & \text{otherwise.} \end{aligned}$$

Thus, from Lemma 2.3,

$$(3.4) \quad (p_i - p_j) \left(\frac{\partial H(c | \mathbf{p})}{\partial p_i} - \frac{\partial H(c | \mathbf{p})}{\partial p_j} \right) = -(p_i - p_j)^2 (f(c | \mathbf{p}^{ij}) - f(c - 1 | \mathbf{p}^{ij})).$$

Now, Samuels [7] has shown (using a well-known inequality attributed to Newton) that if $f(k)$ is the probability of k successes in m independent Bernoulli trials, and if $\sum_{k=0}^m k f(k) = \tau$, then $f(k)$ is increasing in k for $k \leq [\tau]$ and decreasing in k for $k \geq [\tau + 1]$. Hence, using the characterization of $f(k | \mathbf{p}^{ij})$ given in (2.6), and noting that $\sum_{k=0}^{n-2} k f(k | \mathbf{p}^{ij}) = \lambda - p_i - p_j$, we have that (3.4) is nonnegative for $c \geq [\lambda - p_i - p_j + 2]$ and nonpositive for $c \leq [\lambda - p_i - p_j]$. Since $0 \leq p_i + p_j \leq 2$, all $i \neq j$, this result means that for all $\mathbf{p} \in D_\lambda$, (3.4) is ≤ 0 for $c \leq [\lambda - 2]$ and ≥ 0 for $c \geq [\lambda + 2]$. Thus, the bounds (1.12) and (1.13) in Theorem 1.1 follow by a direct application of Lemmas 2.1 and 2.2. \square

REMARK 1. Hoeffding ([3] Theorem 4) also showed that for all $\mathbf{p} \in D_\lambda$,

$$(3.5) \quad 0 \leq H([\lambda - 1] | \mathbf{p}) \leq H([\lambda - 1] | n^{-1}(\lambda, \lambda, \dots, \lambda)),$$

and

$$(3.6) \quad H([\lambda + 1] | n^{-1}(\lambda, \lambda, \dots, \lambda)) \leq H([\lambda + 1] | \mathbf{p}) \leq 1.$$

It might be thought that more detailed results for the cases $c = [\lambda - 1]$, $c = [\lambda + 1]$, similar to the results in Theorem 1.1, can be obtained. That is, we might suspect that $\mathbf{p}^{(1)} \in D_\lambda$, $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$ for doubly stochastic Π , implies that

$$(3.7) \quad H([\lambda - 1] | \mathbf{p}^{(1)}) \leq H([\lambda - 1] | \mathbf{p}^{(2)}),$$

$$(3.8) \quad H([\lambda + 1] | \mathbf{p}^{(1)}) \geq H([\lambda + 1] | \mathbf{p}^{(2)}).$$

The inequalities (3.7) and (3.8) do not, however, always hold. Inequality (3.7) holds if $\mathbf{p}^{(1)}$ is restricted to belong to the subset

$$D_\lambda^0 = \{\mathbf{p} : \mathbf{p} \in D_\lambda; [\lambda - 1] \leq [\lambda - p_i - p_j] \leq [\lambda + 1], \text{ all } i \neq j\}$$

of D_λ , as can be seen from the proof of Theorem 1.1. (Note: if $\mathbf{p}^{(1)} \in D_\lambda^0$ and $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$, Π doubly stochastic, then $\mathbf{p}^{(2)} \in D_\lambda^0$.) Similarly, inequality (3.8)

holds if $\mathbf{p}^{(1)}$ is restricted to the subset

$$D_\lambda^1 = \{\mathbf{p} : \mathbf{p} \in D_\lambda; [\lambda] \leq [\lambda - p_i - p_j] \leq [\lambda + 1], \text{ all } i \neq j\}$$

of D_λ .

That (3.7) does not hold in general can be seen by letting $n = 4$, $\mathbf{p}^{(1)} = (1, \frac{1}{2}, \frac{1}{4}, \frac{1}{4})$, and

$$(3.9) \quad \Pi = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

Here, $\lambda = 2$, $\mathbf{p}^{(2)} = (\frac{3}{4}, \frac{3}{4}, \frac{1}{4}, \frac{1}{4})$, $[\lambda - 1] = 1$, and

$$H([\lambda - 1] | \mathbf{p}^{(1)}) = \frac{9}{32} > \frac{6 \cdot 9}{2 \cdot 5 \cdot 6} = H([\lambda - 1] | \mathbf{p}^{(2)}).$$

That (3.8) does not hold in general can be seen by letting $n = 4$, $\mathbf{p}^{(1)} = (\frac{1}{4}, 0, \frac{3}{4}, \frac{3}{4})$, and Π be as in (3.9). Here $\lambda = \frac{7}{4}$, $\mathbf{p}^{(2)} = (\frac{1}{8}, \frac{1}{8}, \frac{3}{4}, \frac{3}{4})$, $[\lambda + 1] = 2$ and

$$H([\lambda + 1] | \mathbf{p}^{(1)}) = \frac{5 \cdot 5}{8 \cdot 4} < \frac{8 \cdot 8 \cdot 3}{1 \cdot 0 \cdot 2 \cdot 4} = H([\lambda + 1] | \mathbf{p}^{(2)}).$$

Since when λ is not an integer, Theorem 4 of [3] does not even show that $H([\lambda] | \mathbf{p})$ is bounded by the values of $H([\lambda] | \mathbf{p})$ for $\mathbf{p} = \mathbf{p}^*(\lambda)$ and $\mathbf{p} = \hat{\mathbf{p}}(\lambda)$, it is unlikely that an ordering between $H([\lambda] | \mathbf{p}^{(1)})$ and $H([\lambda] | \mathbf{p}^{(2)})$, for $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$ that always goes in the same direction for all $\mathbf{p}^{(1)} \in D_\lambda$, all doubly stochastic Π , can be demonstrated. Indeed, it is easy to find examples in which $H([\lambda] | \mathbf{p}^{(1)}) < H([\lambda] | \mathbf{p}^{(2)})$ for one choice of $\mathbf{p}^{(1)} \in D_\lambda$ and a doubly-stochastic matrix Π , and in which $H([\lambda] | \mathbf{p}^{(1)}) > H([\lambda] | \mathbf{p}^{(2)})$ for another choice of Π and $\mathbf{p}^{(1)} \in D_\lambda$. The case when λ is an integer is more interesting, since in this case Theorem 4 of [3] states that

$$H(\lambda | n^{-1}(\lambda, \lambda, \dots, \lambda)) \leq H(\lambda | \mathbf{p}) \leq 1$$

for all $\mathbf{p} \in D_\lambda$. Thus, for any $\mathbf{p}^{(1)} \in D_\lambda$ it is possible to find a doubly stochastic matrix Π mapping $\mathbf{p}^{(1)}$ into $\mathbf{p}^{(2)} = n^{-1}(\lambda, \lambda, \dots, \lambda)$ for which

$$H(\lambda | \mathbf{p}^{(2)}) \leq H(\lambda | \mathbf{p}^{(1)}),$$

and unless $\mathbf{p}^{(1)} = n^{-1}(\lambda, \lambda, \dots, \lambda)$, this inequality will be strict (see [3]). On the other hand, even in this special case it is unfortunately true that we can find a $\mathbf{p}^{(1)} \in D_\lambda$ and a doubly stochastic matrix Π such that

$$H(\lambda | \mathbf{p}^{(1)}\Pi) > H(\lambda | \mathbf{p}^{(1)}).$$

For example, let $\mathbf{p}^{(1)} = (\frac{1}{2}, 0, \frac{3}{4}, \frac{3}{4})$, $n = 4$, and Π be given by (3.9). Then $\lambda = 2$, $\mathbf{p}^{(2)} = (\frac{1}{4}, \frac{1}{4}, \frac{3}{4}, \frac{3}{4})$, and

$$H(2 | \mathbf{p}^{(2)}) = \frac{1}{2} \frac{8 \cdot 7}{5 \cdot 6} > \frac{1}{2} \frac{8 \cdot 4}{8 \cdot 8} = H(2 | \mathbf{p}^{(1)}).$$

REMARK 2. Hoeffding ([3] Theorem 5) also showed that if $0 \leq b \leq \lambda \leq c \leq n$, then for all $\mathbf{p} \in D_\lambda$,

$$(3.10) \quad \begin{aligned} H(c | \mathbf{p}^*(\lambda)) - H(b - 1 | \mathbf{p}^*(\lambda)) \\ \leq P\{b \leq S \leq c\} = H(c | \mathbf{p}) - H(b - 1 | \mathbf{p}) \\ \leq 1. \end{aligned}$$

Correspondingly, as a corollary to Theorem 1.1, we can establish the following result.

THEOREM 3.1. *Suppose $0 \leq b \leq [\lambda - 1]$ and $[\lambda + 2] \leq c \leq n$. Let $\mathbf{p}^{(1)} \in D_\lambda$ and let $\mathbf{p}^{(2)} = \mathbf{p}^{(1)}\Pi$, where Π is an $n \times n$ doubly stochastic matrix. Then*

$$(3.11) \quad H(c | \mathbf{p}^{(2)}) - H(b - 1 | \mathbf{p}^{(2)}) \leq H(c | \mathbf{p}^{(1)}) - H(b - 1 | \mathbf{p}^{(1)}).$$

REMARK 3. For the possible statistical applications of the results obtained in this paper, the reader is urged to read Section 5 of [3], and also the comments in [4] and [7].

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