

SUFFICIENT EQUIVALENCES IN A SAMPLING SPACE

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Abstract. For a sampling design of an ordered sample on a finite population we define the induced sampling space, and the notion of consistency of a data with an unknown parameter $\mathbf{Y} = (\mathbf{Y}_1, \mathbf{Y}_2, \dots, \mathbf{Y}_N)$, which is some characteristic of the units in the population. We examine the sufficient statistics for \mathbf{Y} . The sufficiency of a statistic is connected with the kernel of the statistic. We introduce sufficient equivalences on the sampling space and give a complete description of a maximal elements in the set of all sufficient equivalences.

1. SAMPLING DESIGN OF AN ORDERED SAMPLE

Let $B = \{b_1, b_2, \dots, b_N\}$ be a finite set (called population) and let $U(B)$ be the free semigroup generated by B . Any finite sequence $\sigma = (b_1, b_2, \dots, b_n)$ with elements (units) in B is called an ordered sample on B . The number $n = L(\sigma)$ is called the length, the set $\{b_1, b_2, \dots, b_n\} = C(\sigma)$ the content of σ and the number $n(\sigma) = |C(\sigma)|$ the effective size of σ . If $b \in C(\sigma)$ we will write $b \in \sigma$ and say that the unit b is in the sample σ . The set of all ordered samples on B is $U(B)$, which will be denoted only by U if B is a given population. To simplify the notation the first N positive integers will be used as notations for B , and we will use the word sample instead of ordered sample.

Definition 1.1. *Sampling design is an ordered triple $S = (B, U, p)$, where $p : U \rightarrow \mathbb{R}$ is a mapping from U to the set of real numbers (\mathbb{R}) such that*

$$i) \quad p(\sigma) \geq 0, \forall \sigma \in U;$$

$$ii) \quad \sum_{\sigma \in U} p(\sigma) = 1$$

In *ii)* we have in mind that U is an infinite countable set. The subset $U_p = \{\sigma | p(\sigma) > 0\}$ of U is called the support of the design S . The mapping p is called a *probability function* of the design S , and $p(\sigma)$ is said to be the probability of σ in the design S .

2000 Mathematics Subject Classification. 62B05, 94A20.

Key words and phrases. sample, sampling design, sampling space, statistics, sufficient statistic, sufficient equivalence.

1.1.⁰ The probability function of the design, p , induces a probability measure on the set of all subsets of U , $B(U)$, defined by

$$P(A) = \sum_{\sigma \in A} p(\sigma), \quad \forall A \subseteq U. \quad \square$$

In this way, to any design $S = (B, U, p)$ corresponds a probability space $(U, B(U), P)$. From the fact that U is countable set, and the σ -algebra is $B(U)$, we have that any real function $X : U \rightarrow \mathbb{R}$ is a random variable on $(U, B(U), P)$.

2. SAMPLING SPACE INDUCED BY A SAMPLING DESIGN

For a given population $B = \{1, 2, \dots, N\}$ and a real function $\mathbf{Y} : B \rightarrow \mathbb{R}$, we denote $\mathbf{Y}(i)$ by Y_i , for $i \in B$. The variable \mathbf{Y} is a characteristic of the population. Let the vector $\mathbf{Y} = (Y_1, Y_2, \dots, Y_N)$ be the unknown *parameter* of the population. Let $S(B, U, p)$ be a given sampling design, $\mathbf{Y} \in \mathbb{R}^N$, and let the components of \mathbf{Y} be fixed, i.e. if a unit from B is available, then the corresponding component of \mathbf{Y} is completely determined. So, the randomness in the model is introduced only by the definition of the sampling design. Since $\mathbf{Y} \in \mathbb{R}^N$ can be considered as a mapping $\mathbf{Y} : i \rightarrow Y_i$ from B to \mathbb{R} and $\sigma \in U$ as a mapping $\sigma : i \rightarrow \sigma(i)$ from $\mathbb{N}_n = \{1, 2, \dots, n\}$ to B , where $n = L(\sigma)$, we have that $\mathbf{Y}\sigma : i \rightarrow Y_{\sigma(i)}$ is a mapping from \mathbb{N}_n to \mathbb{R} . Then $\mathbf{y} = \mathbf{Y}\sigma \in \mathbb{R}^{L(\sigma)}$.

One of the main notions in the theory of sampling design is the notion of *data* obtained by a given sample.

We define the sets Δ' and Δ'^+ by

$$\Delta' = \{(\sigma, \mathbf{Y}\sigma) | \sigma \in U, \mathbf{Y} \in \mathbb{R}^N\} \text{ and } \Delta'^+ = \{(\sigma, \mathbf{Y}\sigma) | \sigma \in U_p, \mathbf{Y} \in \mathbb{R}^N\}$$

The pair $(\sigma, \mathbf{y}) = ((s_1, s_2, \dots, s_n), (y_1, y_2, \dots, y_n))$ can be observed as a sequence of pairs $((s_1, y_1), \dots, (s_n, y_n))$.

Using the previous definitions the following proposition holds:

2.1⁰

$$\begin{aligned} \Delta' &= \{(\sigma, \mathbf{y}) | \sigma \in U, \mathbf{y} \in \mathbb{R}^{L(\sigma)}, \exists \mathbf{Y} \in \mathbb{R}^N \text{ for which } \mathbf{y} = \mathbf{Y}\sigma\} \\ &= \{(\sigma, \mathbf{y}) | \sigma \in U, \mathbf{y} \in \mathbb{R}^{L(\sigma)}, \ker \sigma \subseteq \ker \mathbf{y}\}, \end{aligned}$$

where in the second notation we consider σ and \mathbf{y} as $\sigma : \mathbb{N}_n \rightarrow B$ and $\mathbf{y} : \mathbb{N}_n \rightarrow \mathbb{R}$. \square

The set Δ' is a subset of the set $\Delta = \{(\sigma, \mathbf{y}) | \sigma \in U, \mathbf{y} \in \mathbb{R}^{L(\sigma)}\}$. We will call the set Δ a *sampling space*. We also define $\Delta^+ = \{(\sigma, \mathbf{y}) | \sigma \in U_p, \mathbf{y} \in \mathbb{R}^{L(\sigma)}\}$.

Definition 2.1. We say that the data $\mathbf{d} = (\sigma, \mathbf{y}) \in \Delta$ is consistent with the parameter $\mathbf{Y} \in \mathbb{R}^N$ if and only if (iff) $\mathbf{y} = \mathbf{Y}\sigma$ and we write $\mathbf{d} \sim \mathbf{Y}$.

Definition 2.2. For $\forall \mathbf{Y} \in \mathbb{R}^N$ we define a mapping $p_{\mathbf{Y}} : \Delta \rightarrow \mathbb{R}$ by:

$$p_{\mathbf{Y}}((\sigma, \mathbf{y})) = \begin{cases} p(\sigma) & \text{if } \mathbf{y} = \mathbf{Y}\sigma \\ 0 & \text{otherwise} \end{cases}$$

Then the following proposition holds:

2.2⁰ The triple $(\Delta, B(\Delta), P_{\mathbf{Y}})$ is a probability space, where $\forall A \subseteq \Delta$,

$$P_{\mathbf{Y}}(A) = \sum_{(\sigma, \mathbf{y}) \in A} p_{\mathbf{Y}}((\sigma, \mathbf{y})). \quad \square$$

3. SUFFICIENT STATISTIC

Let $\Omega \subseteq \mathbb{R}^m$ and $f : \Delta \rightarrow \Omega$ be a statistic.

Definition 3.1. The statistic $f : \Delta \rightarrow \Omega$ is called a sufficient statistic for the parameter $\mathbf{Y} \in \mathbb{R}^N$ if for given $\mathbf{w}_0 \in \Omega$ and $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$ the following equation holds

$$P_{\mathbf{Y}'}(\mathbf{D} = \mathbf{d} | f(\mathbf{D}) = \mathbf{w}_0) = P_{\mathbf{Y}''}(\mathbf{D} = \mathbf{d} | f(\mathbf{D}) = \mathbf{w}_0).$$

From the definition of a conditional probability we have

$$P_{\mathbf{Y}}(\mathbf{D} = \mathbf{d} | f(\mathbf{d}) = \mathbf{w}_0) = \frac{P_{\mathbf{Y}}(\{\mathbf{d}\} \cap f^{-1}(\{\mathbf{w}_0\}))}{P_{\mathbf{Y}}(f^{-1}(\{\mathbf{w}_0\}))} \quad (1)$$

when there exists $\mathbf{d}^* = (\sigma^*, \mathbf{y}^*) \in f^{-1}(\{\mathbf{w}_0\})$ for which $\mathbf{Y}\sigma^* = \mathbf{y}^*$ and $p(\sigma^*) = p_{\mathbf{Y}}(\mathbf{d}^*) > 0$. Otherwise, $P_{\mathbf{Y}}(\mathbf{D} = \mathbf{d} | f(\mathbf{d}) = \mathbf{w}_0) = 0$ when $f(\mathbf{d}) \neq \mathbf{w}_0$ or $p_{\mathbf{Y}}(\mathbf{d}) = 0$ for each \mathbf{d} such that $f(\mathbf{d}) = \mathbf{w}_0$.

Theorem 3.1. If $f : \Delta \rightarrow \Omega$ is a statistic, the following statements are equivalent:

- i) f is sufficient for \mathbf{Y} ,
- ii) If $\mathbf{d}', \mathbf{d}'' \in \Delta'^+$, $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$ are such that $f(\mathbf{d}') = f(\mathbf{d}'')$, $\mathbf{d}' \sim \mathbf{Y}'$, $\mathbf{d}'' \sim \mathbf{Y}''$, then $\mathbf{d}' \sim \mathbf{Y}''$, $\mathbf{d}'' \sim \mathbf{Y}'$,
- iii) If $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$, $\mathbf{w}_0 \in \Omega$ are such that $P_{\mathbf{Y}'}(f^{-1}(\{\mathbf{w}_0\})) > 0$, $P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\})) > 0$, then $p_{\mathbf{Y}'}(\mathbf{d}) = p_{\mathbf{Y}''}(\mathbf{d})$ for each $\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\})$,
- iv) Under the conditions as in iii), $p_{\mathbf{Y}'}(\mathbf{d}) = p_{\mathbf{Y}''}(\mathbf{d})$, for $\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\}) \cap \Delta'^+$.

Proof. (1) Let f be a sufficient for \mathbf{Y} and the conditions in ii) hold. Then from $f(\mathbf{d}') = f(\mathbf{d}'') = \mathbf{w}_0$, $p_{\mathbf{Y}'}(\mathbf{d}') > 0$ and $p_{\mathbf{Y}''}(\mathbf{d}'') > 0$ it follows that

$$P_{\mathbf{Y}'}(f^{-1}(\{\mathbf{w}_0\})) > 0, \quad P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\})) > 0, \quad \text{and}$$

$$P_{\mathbf{Y}'}(\mathbf{D} = \mathbf{d}' | f(\mathbf{D}) = \mathbf{w}_0) = P_{\mathbf{Y}''}(\mathbf{D} = \mathbf{d}'' | f(\mathbf{D}) = \mathbf{w}_0) \text{ i.e.}$$

$$\frac{p_{\mathbf{Y}'}(\mathbf{d}')}{P_{\mathbf{Y}'}(f^{-1}(\{\mathbf{w}_0\}))} = \frac{p_{\mathbf{Y}''}(\mathbf{d}'')}{P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\}))}$$

Since $\mathbf{d}' \sim \mathbf{Y}'$ it follows that $p_{\mathbf{Y}'}(\mathbf{d}') > 0$, which means that $p_{\mathbf{Y}''}(\mathbf{d}'') > 0$, i. e. $\mathbf{d}'' \sim \mathbf{Y}''$. By symmetry it follows that $\mathbf{d}'' \sim \mathbf{Y}'$. So we have that i) \Rightarrow ii).

(2) Let ii) holds and the conditions from iii) are satisfied. Let $\mathbf{d} \in \Delta$ be such that $\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\})$, or $f(\mathbf{d}) = \mathbf{w}_0$. If $\mathbf{d} \in \Delta \setminus \Delta'^+$, then $p_{\mathbf{Y}'}(\mathbf{d}) = 0 = p_{\mathbf{Y}''}(\mathbf{d})$. So let $\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\}) \cap \Delta'^+$. If $p_{\mathbf{Y}'}(\mathbf{d}) = 0$ and $p_{\mathbf{Y}''}(\mathbf{d}) = 0$ then they are equal. So let $p_{\mathbf{Y}'}(\mathbf{d}) > 0$. Then $\mathbf{d} \sim \mathbf{Y}'$. Since $P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\})) > 0$, there exists $\mathbf{d}'' \in f^{-1}(\{\mathbf{w}_0\}) \cap \Delta'^+$ for which $p_{\mathbf{Y}''}(\mathbf{d}'') > 0$, so $\mathbf{d}'' \sim \mathbf{Y}''$. The above discussion

shows that for $\mathbf{d} = \mathbf{d}'$, \mathbf{d}'' and \mathbf{Y}' , $\mathbf{Y}'' \in \mathbb{R}^N$, the conditions from ii) holds, which implies that $\mathbf{d} \sim \mathbf{Y}''$. Then the definition of $p_{\mathbf{Y}}$ implies that $p_{\mathbf{Y}''}(\mathbf{d}) = p_{\mathbf{Y}'}(\mathbf{d}) > 0$. So we proved that ii) \Rightarrow iii).

(3) It is obvious that iii) \Leftrightarrow iv), so we have to prove that iii) \Rightarrow i).

(4) Let iii) holds and let $\mathbf{d}_0 \in \Delta$, $\mathbf{w}_0 \in \Omega$ and \mathbf{Y}' , $\mathbf{Y}'' \in \mathbb{R}^N$ are such that $P_{\mathbf{Y}'}(f^{-1}(\{\mathbf{w}_0\})) > 0$, and $P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\})) > 0$. Then, for each $\mathbf{d} \in (f^{-1}(\{\mathbf{w}_0\}))$ according to iii), we have that $p_{\mathbf{Y}''}(\mathbf{d}) = p_{\mathbf{Y}'}(\mathbf{d})$. We have to prove that if

$\lambda = P_{\mathbf{Y}'}(f^{-1}(\{\mathbf{w}_0\})) > 0$, and $\mu = P_{\mathbf{Y}''}(f^{-1}(\{\mathbf{w}_0\})) > 0$, then $\lambda p_{\mathbf{Y}''}(\mathbf{d}_0) = \mu p_{\mathbf{Y}'}(\mathbf{d}_0)$.

If $\mathbf{d}_0 \notin f^{-1}(\{\mathbf{w}_0\})$, then $p_{\mathbf{Y}''}(\mathbf{d}_0) = p_{\mathbf{Y}'}(\mathbf{d}_0) = 0$ and the equality holds. So, let $\mathbf{d}_0 \in f^{-1}(\{\mathbf{w}_0\})$. Then iii) implies that $p_{\mathbf{Y}''}(\mathbf{d}_0) = p_{\mathbf{Y}'}(\mathbf{d}_0)$.

So $\lambda = \sum_{\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\})} p_{\mathbf{Y}'}(\mathbf{d}) = \sum_{\mathbf{d} \in f^{-1}(\{\mathbf{w}_0\})} p_{\mathbf{Y}''}(\mathbf{d}) = \mu$ □

Corollary 3.1.1. *If a statistic $f : \Delta \rightarrow \Omega$ is injection, then f is sufficient for \mathbf{Y} .*

Theorem 3.2. *Let $f : \Delta \rightarrow \Omega_1$, $g : \Delta \rightarrow \Omega_2$ be such that $\ker f \subseteq \ker g$. If g is sufficient for \mathbf{Y} then f is sufficient for \mathbf{Y} .*

Proof. Let g be sufficient for \mathbf{Y} . By the theorem 3.1 it is sufficient to show that f satisfies the condition ii). Let $\mathbf{d}' = (\sigma', \mathbf{y}')$, $\mathbf{d}'' = (\sigma'', \mathbf{y}'') \in \Delta^+$ and \mathbf{Y}' , $\mathbf{Y}'' \in \mathbb{R}^N$, are such that $f(\mathbf{d}') = f(\mathbf{d}'')$, $\mathbf{d}' \sim \mathbf{Y}'$, $\mathbf{d}'' \sim \mathbf{Y}''$. Then, $\ker f \subseteq \ker g$ implies that $g(\mathbf{d}') = g(\mathbf{d}'')$. Since g is sufficient, ii) holds for g , which means that $\mathbf{d}' \sim \mathbf{Y}''$, $\mathbf{d}'' \sim \mathbf{Y}'$. So, we proved that condition ii) from Theorem 3.1 is satisfied for f , which implies that f is sufficient for \mathbf{Y} . □

Corollary 3.2.1. *Let $f : \Delta \rightarrow \Omega_1$, $g : \Delta \rightarrow \Omega_2$ be such that $\ker f = \ker g$. Then f is sufficient for \mathbf{Y} iff g is sufficient for \mathbf{Y} .*

Theorem 3.3. *Let $f : \Delta \rightarrow \Omega_1$, $g : \Delta \rightarrow \Omega_2$ be statistics such that $\ker f_{\Delta^+} \subseteq \ker g_{\Delta^+}$. If g is sufficient for \mathbf{Y} , then f is sufficient for \mathbf{Y} .*

4. SUFFICIENT EQUIVALENCE IN Δ

In the previous section we have seen that for a statistic $f : \Delta \rightarrow \Omega$, its sufficiency depends of the kernel of f , which is an equivalence relation on Δ . Let $\ker f = \alpha$. If we take Ω to be the factor set Δ/α and define $\bar{f}(d) = d^\alpha$ we have an onto mapping \bar{f} from Δ to Δ/α and $\ker f = \ker \bar{f}$. From the previous discussion it follows that if f is sufficient so is \bar{f} .

This construction is possible for any equivalence ω on Δ . Let $\text{nat } \omega : \Delta \rightarrow \Delta/\omega$ be the natural mapping, i.e. $\text{nat } \omega(d) = d^\omega$. The previous discussion suggests the definition of a sufficient equivalence on Δ .

Definition 4.1. *We say that an equivalence ω on Δ is sufficient iff the natural mapping $\text{nat } \omega : \Delta \rightarrow \Delta/\omega$ is a sufficient statistic.*

Theorem 4.1. *An equivalence ω on Δ is sufficient iff the following holds:*

If \mathbf{d}' , $\mathbf{d}'' \in \Delta^+$ and \mathbf{Y}' , $\mathbf{Y}'' \in \mathbb{R}^n$ are such that $\mathbf{d}' \sim \mathbf{Y}'$, $\mathbf{d}'' \sim \mathbf{Y}''$ and $\mathbf{d}'^\omega = \mathbf{d}''^\omega$, then $\mathbf{d}' \sim \mathbf{Y}''$, $\mathbf{d}'' \sim \mathbf{Y}'$. □

Let $Eq(\Delta)$ be the set of all equivalences on Δ , and $SEq(\Delta)$ be the set of all sufficient equivalences on Δ .

In the family of all equivalences on Δ , the smallest element is θ , and the largest element is ε , where $\mathbf{d}^\theta = \{\mathbf{d}\}$ and $\mathbf{d}^\varepsilon = \Delta$, for $\mathbf{d} \in \Delta$.

As a direct interpretation of the results in previous section we get the following results.

4.2⁰ For any design S , θ is sufficient equivalence.

Proof: For given $\mathbf{d}' = (\sigma', \mathbf{y}')$ and $\mathbf{d}'' = (\sigma'', \mathbf{y}'')$ in Δ , $P_{\mathbf{Y}}(\mathbf{d} = \mathbf{d}' \mid \text{nat } \omega = \{\mathbf{d}''\})$ exists iff $0 < P_{\mathbf{Y}}((\text{nat } \omega)^{-1}\{\mathbf{d}''\}) = p_{\mathbf{Y}}(\mathbf{d}'')$. The latest is true iff $p(\sigma'') > 0$ and $\mathbf{Y}\sigma'' = \mathbf{y}''$. Then

$$P_{\mathbf{Y}}(\mathbf{d} = \mathbf{d}' \mid \text{nat } \omega(\mathbf{d}) = \{\mathbf{d}''\}) = \begin{cases} 0 & \text{if } \mathbf{d}' \neq \mathbf{d}'' \\ 1 & \text{if } \mathbf{d}' = \mathbf{d}'' \end{cases}$$

which means that θ is sufficient equivalence. \square

4.3⁰ The equivalence ε is not sufficient.

Proof. Since, for each $\mathbf{d} \in \Delta$, $\mathbf{d}^\omega = \Delta$, it follows that

$$P_{\mathbf{Y}}(\mathbf{d} = (\sigma^0, \mathbf{y}^0) \mid \text{nat } \omega(\mathbf{d}) = \Delta) = p_{\mathbf{Y}}(\sigma^0, \mathbf{y}^0) = \begin{cases} p(\sigma^0) & \text{for } \mathbf{Y}\sigma^0 = \mathbf{y}^0 \\ 0 & \text{otherwise} \end{cases}$$

To show that the condition for sufficiency is not satisfied, we have to show that there are $(\sigma^0, \mathbf{y}^0) \in \Delta$ and $\mathbf{Y}', \mathbf{Y}'' \in \mathbf{R}^n$ such that $\mathbf{Y}'\sigma^0 = \mathbf{y}^0$, $\mathbf{Y}''\sigma^0 \neq \mathbf{y}^0$ and $p(\sigma^0) > 0$. If we choose $\mathbf{d}^0 = (\sigma^0, \mathbf{y}^0)$ such that $\ker \sigma^0 \subseteq \ker \mathbf{y}^0$ and $p(\sigma^0) > 0$, it is enough to choose \mathbf{Y}' such that for each $i \in \mathbb{N}_n$ and \mathbf{Y}'' such that $\mathbf{Y}''_{\sigma^0(i)} \neq \mathbf{y}^0_i$ for at least one $i \in \mathbb{N}_n$. \square

Theorem 4.4 Let $\alpha, \beta \in Eq(\Delta)$ be such that $\alpha \subseteq \beta$. If $\beta \in SEq(\Delta)$, then $\alpha \in SEq(\Delta)$. \square

Theorem 4.5 If $\alpha, \beta \in Eq(\Delta)$ are such that $\alpha_{/\Delta'+} \subseteq \beta_{/\Delta'+}$ then:

$$\alpha \in SEq(\Delta) \text{ if and only if } \beta \in SEq(\Delta). \quad \square$$

It is well known that the set $Eq(\Delta)$ is a complete lattice, (i.e., for each subset $\Gamma \subseteq Eq(\Delta)$, $\sup \Gamma$ and $\inf \Gamma$ are elements in $Eq(\Delta)$). In this lattice $\inf \Gamma$ is the intersection of the relations on Γ , and $\sup \Gamma$ is the transitive product of the relations in Γ . As a consequence of Theorem 4.5 we obtain the following

Corollary 4.5.1 If $\Gamma \subseteq SEq(\Delta)$ then $\inf \Gamma \in SEq(\Delta)$. (More generally, if $\Gamma \cap SEq(\Delta) \neq \emptyset$ then $\inf \Gamma \in SEq(\Delta)$). \square

It is natural to ask the question if an analogous result holds for $\sup \Gamma$. To consider this question we have to define the notion of chain in $Eq(\Delta)$. Namely,

we say that Γ is a *chain* in $Eq(\Delta)$ iff Γ is a subset of $Eq(\Delta)$ such that for each $\alpha, \beta \in \Gamma$, $\alpha \subseteq \beta$ or $\beta \subseteq \alpha$. In a similar way we define a chain in $SEq(\Delta)$.

Theorem 4.6 *If Γ is a chain in $SEq(\Delta)$ then $\sup \Gamma$, denoted by γ , is a union of the relations in Γ and $\gamma \in SEq(\Delta)$.*

Proof. We will show that $\sup \Gamma$ is γ , a union of the relations in Γ . Let $\gamma = \bigcup_{\beta \in \Gamma} \beta$.

- 1) γ is reflexive iff at least one relation of Γ is reflexive;
- 2) Since all elements in Γ are symmetric, so is γ ;
- 3) Let $(\mathbf{d}_1, \mathbf{d}_2), (\mathbf{d}_2, \mathbf{d}_3) \in \gamma$. Then, there exist $\alpha, \beta \in \Gamma$, such that $(\mathbf{d}_1, \mathbf{d}_2) \in \alpha$ and $(\mathbf{d}_2, \mathbf{d}_3) \in \beta$. But, $\alpha \subseteq \beta$ or $\beta \subseteq \alpha$. Let $\alpha \subseteq \beta$. Then $(\mathbf{d}_1, \mathbf{d}_2), (\mathbf{d}_2, \mathbf{d}_3) \in \beta$, and since β is transitive, $(\mathbf{d}_1, \mathbf{d}_3) \in \beta$. So $(\mathbf{d}_1, \mathbf{d}_3) \in \gamma$, which means that γ is transitive.

From 1), 2) and 3) it follows that γ is an equivalence and for each $\mathbf{d} \in \Delta$,

$$\mathbf{d}^\gamma = \bigcup_{\beta \in \Gamma} \beta = \mathbf{d}^\beta \quad (2)$$

Using the last equation and Theorem 4.1 we will show that $\gamma \in SEq(\Delta)$.

Let $\mathbf{d}', \mathbf{d}'' \in \Delta$ and $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$ be such that $\mathbf{d}' \sim \mathbf{Y}'$, $\mathbf{d}'' \sim \mathbf{Y}''$ and $\mathbf{d}'^\gamma = \mathbf{d}''^\gamma$. From $\mathbf{d}'^\gamma = \mathbf{d}''^\gamma$ it follows that $\mathbf{d}' \gamma \mathbf{d}''$, which means that there exists $\beta \in \Gamma$, such that $\mathbf{d}' \beta \mathbf{d}''$, i.e., $\mathbf{d}'^\beta = \mathbf{d}''^\beta$. Since $\beta \in SEq(\Delta)$, the equation (2) and Theorem 4.1 imply that $\mathbf{d}' \sim \mathbf{Y}''$, $\mathbf{d}'' \sim \mathbf{Y}'$, and (again by Theorem 4.1) that $\gamma \in SEq(\Delta)$. \square

As a consequence of theorem 4.7 and the lemma of Zorn, we get the following result:

Theorem 4.7 *For any sufficient equivalence α there is a maximal sufficient equivalence γ , such that $\alpha \subseteq \gamma$. In other words, for $\forall \alpha \in SEq(\Delta)$ there is a maximal element $\gamma \in SEq(\Delta)$ such that $\alpha \subseteq \gamma$. \square*

At the end we will define a sufficient equivalence κ which will enable a complete description of $SEq(\Delta)$, and by that the class of all sufficient statistics.

For each $\mathbf{d} = (\sigma, \mathbf{y}) \in \Delta$ define a set $A(\sigma, \mathbf{y}) = A(\mathbf{d})$ by

$x \in A(\sigma, \mathbf{y})$ iff $x = (s_i, y_i)$ for some $i \in \mathbb{N}_n$,

where $(\sigma, \mathbf{y}) = ((s_1, \dots, s_n), (y_1, \dots, y_n))$, i.e. $\mathbf{d} = ((s_1, y_1), \dots, (s_n, y_n))$.

Let $[\Delta] = \{A(\mathbf{d}) | \mathbf{d} \in \Delta\}$. Define a statistic $k : \Delta \rightarrow [\Delta]$ by

$$k(\mathbf{d}) = A(\mathbf{d}). \quad (3)$$

Theorem 4.8 *The statistic k is sufficient statistic.*

Proof. Suppose that the conditions from theorem 4.1 (ii) hold, i.e.: $\mathbf{d}' = (\sigma, \mathbf{y})$, $\mathbf{d}'' = (\tau, \mathbf{z})$ and $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$ are such that $k(\mathbf{d}') = k(\mathbf{d}'')$, $\mathbf{Y}'\sigma = \mathbf{y}$ and $\mathbf{Y}''\tau = \mathbf{z}$. We have to show that $\mathbf{Y}''\sigma = \mathbf{y}$ and $\mathbf{Y}'\tau = \mathbf{z}$.

From the equation $A(\mathbf{d}') = A(\mathbf{d}'')$ it follows that for each $i \in \mathbb{N}_n$ there is $i \in \mathbb{N}_m$ such that $\sigma_i = \tau_j$ and $y_i = z_j$. Then from $\mathbf{Y}'\sigma = \mathbf{y}$ and $\mathbf{Y}''\tau = \mathbf{z}$, it follows that $\mathbf{Y}''_{\tau_j} = z_j = y_i = \mathbf{Y}'_{\sigma_i}$. So, $\mathbf{Y}''_{\sigma_i} = \mathbf{Y}''_{\tau_j} = z_j = y_i$, which means that $\mathbf{Y}''\sigma = \mathbf{y}$. By symmetry it follows that $\mathbf{Y}'\tau = \mathbf{z}$. \square

Theorem 4.9 *A statistic $f : \Delta \rightarrow \Omega$ is sufficient for the parameter \mathbf{Y} iff :*

$$\ker f_{\Delta'+} \subseteq \ker k_{\Delta'+}. \quad (4)$$

Proof. If the relation (4) holds, theorem 3.3 implies that f is sufficient statistic.

Suppose that (4) is not satisfied. This means that there are $\mathbf{d}' = (\sigma, \mathbf{y})$; $\mathbf{d}'' = (\tau, \mathbf{z}) \in \Delta'^+$ such that $f(\mathbf{d}') = f(\mathbf{d}'')$, but $k(\mathbf{d}') \neq k(\mathbf{d}'')$. From the inequality $k(\mathbf{d}') \neq k(\mathbf{d}'')$, by symmetry, we can suppose that there is $i \in \mathbb{N}_n$, such that $(\sigma_i, y_i) \neq (\tau_j, z_j)$ for each $j \in \mathbb{N}_m$. Two cases are possible: 1) $\sigma_i \in \tau$ and 2) $\sigma_i \notin \tau$.

Let $\mathbf{Y}', \mathbf{Y}'' \in \mathbb{R}^N$ be such that $\mathbf{d}' \sim \mathbf{Y}'$ and $\mathbf{d}'' \sim \mathbf{Y}''$, i.e. let $\mathbf{Y}'\sigma = \mathbf{y}$ and $\mathbf{Y}''\tau = \mathbf{z}$.

In case 1), there is $j \in \mathbb{N}_m$ such that $\sigma_i = \tau_j$, but since $(\sigma_i, y_i) \neq (\tau_j, z_j)$, $y_i \neq z_j$. But then $y_i \neq z_j = \mathbf{Y}''_{\tau_j} = \mathbf{Y}''_{\sigma_i}$. This means that $\mathbf{Y}''\sigma \neq \mathbf{y}$, or \mathbf{d} is not consistent with \mathbf{Y}'' .

In case 2), $\sigma_i \neq \tau_j$ for each $j \in \mathbb{N}_m$. Let $\mathbf{Y}^* \in \mathbb{R}^N$ be such that $\mathbf{Y}^*_k = \mathbf{Y}''_k$ for all k except for $k = \sigma_i$, namely $\mathbf{Y}^*_{\sigma_i} \neq y_i$ and all other components of \mathbf{Y}^* and \mathbf{Y}'' are equal. Then $\mathbf{Y}^*\tau = \mathbf{z}$, but $\mathbf{Y}^*\sigma \neq \mathbf{y}$, i.e. \mathbf{d}'' is consistent with \mathbf{Y}^* , but \mathbf{d}' is not consistent with \mathbf{Y}^* .

In both cases theorem 4.1 (ii) implies that f is not sufficient statistic for \mathbf{Y} . \square

The Theorem 4.9 can be formulated in the following manner, where $\kappa = \ker k$.

Theorem 4.9' *The equivalence $\alpha \in Eq(\Delta)$ is sufficient iff $\alpha_{\Delta'+}$. \square*

Let us note that the theorems 4.7 and 4.8 are consequences of theorem 4.9 or 4.9'. At the end we will describe the maximal elements in $Seq(\Delta)$.

Theorem 4.10 *If ω is maximal sufficient equivalence, then $\omega_{\Delta'+} = \kappa_{\Delta'+}$.*

Proof. Let ω be maximal and sufficient equivalence and let $\mathbf{d}', \mathbf{d}'' \in \Delta'^+$ be such that $\mathbf{d}'\kappa\mathbf{d}''$, but $(\mathbf{d}', \mathbf{d}'') \notin \omega$. Let $\tilde{\omega}$ be the equivalence generated by $\omega \cup (\mathbf{d}', \mathbf{d}'')$. Then $\mathbf{d}'\tilde{\omega} \supseteq \mathbf{d}'\omega \cup \mathbf{d}''\omega$, and $\tilde{\omega}$ is defined by:

$$\begin{aligned} \text{if } \mathbf{d} \notin \mathbf{d}'\omega \cup \mathbf{d}''\omega \text{ then } \mathbf{d}\tilde{\omega} &= \mathbf{d}\omega \text{ and} \\ \text{if } \mathbf{d} \in \mathbf{d}'\omega \cup \mathbf{d}''\omega \text{ then } \mathbf{d}\tilde{\omega} &= \mathbf{d}'\omega \cup \mathbf{d}''\omega. \end{aligned}$$

By the definition of $\tilde{\omega}$ it follows that $\omega \subset \tilde{\omega}$ and that $\tilde{\omega}_{\Delta'+} \subseteq \kappa_{\Delta'+}$. Theorem 4.9' implies that $\tilde{\omega}$ is sufficient equivalence for \mathbf{Y} . But this together with $\omega \subset \tilde{\omega}$ contradicts the fact that ω is a maximal equivalence. \square

The condition $\omega_{\Delta'^+} = \kappa_{\Delta'^+}$ is not a sufficient condition for maximality of ω . This is shown by the following example. Let ρ be the equivalence defined by :

for $\mathbf{d} \in \Delta'^+$, $\mathbf{d}^\rho = \mathbf{d}''^{\kappa} \cap \Delta'^+$ and for $\mathbf{d} \notin \Delta'^+$, $\mathbf{d}^\rho = \Delta \setminus \Delta'^+$.

Then $\rho \in \text{Seq}(\Delta)$ and $\rho_{\Delta'^+} = \kappa_{\Delta'^+}$. But ρ is not a maximal element, since it is a proper subset of $\bar{\rho} \in \text{Seq}(\Delta)$ defined as follows. Fix $\mathbf{d}' \in \Delta'^+$ and let $\mathbf{d}'^{\bar{\rho}} = (\Delta \setminus \Delta'^+) \cup \mathbf{d}'^\rho$, and for all other $\mathbf{d} \in \Delta$, let $\mathbf{d}'^{\bar{\rho}} = \mathbf{d}'^\rho$. Since, $\bar{\rho}_{\Delta'^+} = \kappa_{\Delta'^+}$, $\bar{\rho} \in \text{Seq}(\Delta)$ and $\rho \subset \bar{\rho}$.

The previous discussion gives in fact the description of maximal elements of $\text{Seq}(\Delta)$, given by following theorem:

Theorem 4.11 *An equivalence $\omega \in \text{Seq}(\Delta)$ is a maximal element in $\text{Seq}(\Delta)$ if and only if the following conditions hold*

- i) $\omega_{\Delta'^+} = \kappa_{\Delta'^+}$ and
- ii) for each $\mathbf{d} \in \Delta$, there is $\mathbf{d}' \in \Delta'^+$, such that $\mathbf{d}^\omega = \mathbf{d}'^\omega$. \square

REFERENCES

- [1] Cassel, Claes-Magnus, *Foundations of inference in survey sampling*, 1977, John & Sons, Inc.
- [2] A.S.Hedajat, B.K.Sinha, *Design and inference in a finite population sampling*, Wiley series in probability and mathematical statistics, 1990.
- [3] Ž. Popeska, *Algebarsko kombinatorni metodi vo teorija na primerok i eksperimentalen dizajn* - doktorska disertacija, Skopje, 1998.
- [4] Ž. Popeska, *Quotient sampling design*, Prilozi, Odd.mat.teh.nauki, MANU, **15,1**, (1994), 37-48.
- [5] Ž. Popeska, *Sample as an element of a semigroup*, Bulletin of the International Statistical Institute, **Tom LVII, Book 3**, Helsinki, 1999, 139-140.
- [6] Ž. Popeska, *Sampling Space and Statistics on a Sample as an Element of a Semigroup*, Bulletin of the International Statistical Institute, 53rd Session, **Tom LIX -Book 2**, Seoul, 2001, 375-377.

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