A QUANTILE DOMAIN PERSPECTIVE ON THE RELATIONSHIPS BETWEEN OPTIMAL GROUPING, SPACING AND STRATIFICATION PROBLEMS

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Running head: Relationships Between Grouping, Spacing and Stratification Problems

Abstract. The relationships between two distributions having the same solutions for problems of optimal spacing selection for the asymptotically best linear unbiased estimator of a location or scale parameter or for problems of optimal stratification for estimation of a population mean are investigated. Easily checked necessary and sufficient conditions under which two distributions have identical solutions to these problems are given in terms of their quantile and density-quantile functions. As an application of these results a quantile domain analog of a theorem due to Adatia and Chan (1981, Scand. Actuar. J., 193-202) on the equivalence of optimal grouping, spacing and stratification problems is obtained.

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Key Words and Phrases: Approximation, density-quantile function, quantile function, optimal grouping, optimal spacing, optimal stratification.
1. **Introduction.** For a particular absolutely continuous distribution $F$, assumed to depend on a location or scale parameter $\theta$, the insightful work of Adatia and Chan (1981) has focused on the equivalence of problems of i) optimal grouping for maximum likelihood estimation of $\theta$, ii) optimal quantile (optimal spacing) selection for the asymptotically best linear unbiased estimator (ABLUE) of $\theta$ and iii) optimal stratification for estimation of a population mean. In this paper the more general question of when different distributions have equivalent solutions for these problems is considered from a quantile domain perspective. We focus, initially, on the latter two problems which can be stated as follows:

Problem 1. Select percentile points $0 = u_0 < u_1 < \cdots < u_{k+1} = 1$, (frequently called a spacing) corresponding to sample quantiles which maximize the asymptotic relative Fisher efficiency of the ABLUE for $\theta$ (cf. Sarhan and Greenberg (1962, Chap. 5)).

Problem 2. Given strata boundaries $a = x_0 < x_1 < \cdots < x_{k+1} = b$ (where $a$ and $b$ are possibly infinite values which bound the support of $F$), a stratified random sample of size $n$ is to be selected using proportional allocation, i.e., the "number" of sample elements taken from $(x_{i-1}, x_i]$ is $n[F(x_i) - F(x_{i-1})]$. If $\theta$ is the mean for $F$, the usual estimator of $\theta$ is $\hat{\theta} = \frac{1}{k+1}\sum_{i=1}^{k+1}[F(x_i) - F(x_{i-1})]\bar{x}_i$ where $\bar{x}_i$ is the sample mean from the $i$th stratum. The problem is to select the boundaries to minimize the variance of $\hat{\theta}$ (cf. Dalenius (1950)). Observe that when $F$ is a normal distribution $\theta$ is a location parameter whereas for the exponential distribution $\theta$ corresponds to a scale parameter.

Both Problems 1 and 2 are nonlinear in nature so that, as a rule, their solutions must be tabulated numerically. However, in some instances it has been possible to exploit relationships between different types of distri-
butions to obtain, for example, optimal spacings for one distribution
in terms of those for another which have already been computed
(cf. Kulldorff (1973)). When applicable, this approach can save con-
siderable time, effort and expense. The question that arises, of course,
is when and under what conditions can such tactics be expected to work or
collapse. Thus, we would like easily checked conditions regarding the equi-
valence (or non-equivalence) of optimal spacing or stratification problems
for different distribution types. Motivated by such considerations we
will study the relationship between two distributions having the same
solutions for either of Problems 1 or 2. Our major results in this regard
(Theorems 1-3) are stated and discussed in the next section. It will be
seen that necessary and sufficient conditions for the solutions of any of
these problems to coincide for two distributions can be succinctly sum-
marized in terms of relationships between their quantile and density-quantile
functions. Proofs are given in Section 3 with Section 4 devoted to the
application of results in Section 2 to the optimal grouping problem.

2. Optimal Spacing and Stratification. Let \( F_1(x; \theta_1) \) and \( F_2(x; \theta_2) \)
be two strictly monotone, continuously differentiable, distribution functions
(d.f.'s) which depend on parameters \( \theta_1 \) and \( \theta_2 \) of either the location or
scale variety. The standardized forms of these d.f.'s, corresponding
to \( \theta_i = 0 \) or \( i = 1, 2 \) contingent on whether \( \theta_i \) is a location or scale para-
meter, will be denoted as \( H_1 \) and \( H_2 \), respectively, with their associated
continuous densities written as \( h_1 \) and \( h_2 \). Thus, for example, \( F_1(x; \theta) \) can be
expressed as \( H_1(x - \theta_1) \), if \( \theta_1 \) is a location parameter, or \( H_1(x/\theta_1) \), if \( \theta_1 \)
is a scale parameter. Also, define the standardized quantile functions
(q.f.'s) and density-quantile functions (d.q.f.'s)
\[ Q_i(u) = H_i^{-1}(u) = \inf\{x : H_i(x) \geq u\}, \quad 0 < u < 1, \quad i=1,2, \quad (1) \]
and
\[ d_i(u) = h_i(Q_i(u)), \quad 0 \leq u \leq 1, \quad i=1,2. \quad (2) \]

Our principal result regarding two distributions having the same optimal spacings is provided by the following theorem.

**Theorem 1.** Let \( g_i, i=1,2 \), denote either \( d_i \) or the product of \( d_i \) and \( Q_i \), \( d_i \cdot Q_i \), depending on whether \( \theta_i \) is a location or scale parameter. Assume that \( g_i \) is either concave or convex, vanishes at 0 and 1 and is twice continuously differentiable on \((0,1)\) with \((g_i')^2\) and \( |g_i''|^{2/3} \) integrable.

Under these assumptions \( F_1 \) and \( F_2 \) will have the same optimal spacings for the estimation of \( \theta_1 \) and \( \theta_2 \) for all \( k \) if and only if there exists constants \( \alpha, \beta (\beta \neq 0) \) such that

\[ g_1'(u) = \alpha + \beta g_2'(u), \quad u \in (0,1). \quad (3) \]

To exemplify the use of Theorem 1 consider the Weibull distribution

\[ F_1(x; \theta_1) = 1 - \exp\{- (x/\theta_1)^\nu\}, \quad x, \nu > 0, \]
for which \( H_1(x) = 1 - \exp\{-x^\nu\}, \quad Q_1(u) = (\ln(1-u))^{1/\nu} \) and
\[ d_1(u) = \nu(1-u)(-\ln(1-u))^{1-1/\nu}. \]
Since \( \theta_1 \) is a scale parameter we use
\[ g_1(u) = d_1(u)Q_1(u) = -\nu(1-u)\ln(1-u). \]
A special case of the Weibull is the exponential distribution which corresponds to \( \nu = 1 \). The optimal spacings for the exponential have been tabulated and may be found, for instance, in Sarhan, Greenberg and Ogawa (1963). Taking \( g(u) = -(1-u)\ln(1-u) \) it follows from (3) that these spacings are also optimal for the Weibull when \( \nu \neq 1 \). This relationship has also been noted by Kulldorff (1973). Other results obtained by Kulldorff (1973) also follow similarly from Theorem 1.
As another example consider the logistic distribution with

\[ F_1(x; \theta_1) = \left[ 1 + \exp\left( -\pi(x - \theta_1) / \sqrt{3} \right) \right]^{-1}, \quad -\infty < x < \infty, \]

so that \( H_1(x) = \left[ 1 + \exp(-\pi x / \sqrt{3}) \right]^{-1}, \) \( Q_1(u) = \frac{\sqrt{3}}{\pi} \ln(u/(1-u)) \) and \( g_1(u) = d_1(u) = \pi/\sqrt{3} \ u(1-u). \) By choosing \( g_2(u) = \phi^{-1}(u) \), where \( \phi \) and \( \phi^{-1} \) are the standard normal distribution and density function, respectively, it is seen that the optimal spacings for location parameter estimation for the logistic and normal distributions cannot be identical for all \( k. \)

If instead we choose \( F_2(x; \theta_2) = 1 - (1 + x / \theta_2)^{-\nu}, \) \( x, \nu > 0, \) it follows that the optimal spacings for location parameter estimation for the logistic are the same as those for scale parameter estimation in the Pareto distribution if and only if \( \nu = 1. \)

Two distributions will be said to have the same optimal solutions for Problem 2 if, for any set of optimal strata boundaries \( \{x_{i1}\}_{i=0}^{k+1} \) for \( F_1, \) there is a corresponding set \( \{x_{i2}\}_{i=0}^{k+1} \) of optimal boundaries for \( F_2 \) which satisfies

\[ F_1(x_{i1}; \theta_1) = F_2(x_{i2}; \theta_2), \quad i = 0, \ldots, k+1. \] (4)

Such a definition is natural since it considers equivalence in terms of percentage points that are not influenced by departures in the values of \( x_{i1} \) and \( x_{i2} \) due merely to factors of location or scale. The next theorem has the consequence that a distribution is essentially determined by its optimal strata boundaries.

Theorem 2. If, for \( i=1,2, \) \( Q_i \) is square integrable and \( Q_i' = 1/d_i \) is monotone and continuous on \((0,1)\) with \( d_i^{-2/3} \) integrable, then \( F_1 \) and \( F_2 \) have the same solution for Problem 2 (in the sense of (4)) for all \( k \) if and only if there exists constants \( a, b, (b \neq 0) \) such that

\[ a \leq F_1(x; \theta_1) - F_2(x; \theta_2) \leq b, \quad -\infty < x < \infty, \]
Theorem 1 has the implication that distributions with the same optimal strata boundaries must be members of the same family which differ by at most factors of location and/or scale. While the direct implication of this condition appears obvious the converse, although intuitive, is somewhat less transparent.

It is also reasonable to ask under what conditions the optimal spacings for one distribution can be obtained in terms of the optimal strata boundaries for another in the sense that, if \( \{x_i\}^{k+1}_{i=0} \) is a set of optimal boundaries for \( F_2 \), an optimal spacing for \( F_1 \) is provided by

\[
u_{i1} = F_2(x_{i2};\theta_2), \quad i=0,...,k+1.
\]  

Such conditions are provided by the following theorem.

**Theorem 3.** Let \( g_1 \) denote either \( d_1 \) or \( d_1Q_1 \), depending on whether \( \theta_1 \) is a location or scale parameter, and assume that \( g_1 \) and \( Q_2 \) satisfy the hypotheses of Theorem 1 and 2 respectively. Then Problem 1 for \( F_1 \) is equivalent to Problem 2 for \( F_2 \) (in the sense of (6)) for all \( k \) if and only if there exists constants \( a, \beta \neq 0 \) such that

\[
g_1'(u) = a + \beta Q_2(u), \quad u \in (0,1).
\]  

As an illustration, note that it follows from (7) that optimal spacing selection for location parameter estimation for the logistic is equivalent to the problem of optimal stratification for the estimation of the mean of a uniform distribution on any finite interval \([a,b]\).
3. Proofs. In this section Theorems 1-3 will be proven. The proofs are accomplished by a series of three lemmas.

**Lemma 1.** Let \( d \) and \( Q \) denote the standardized \( d.q.f. \) and \( q.f. \) for a distribution which depends on either a location or scale parameter, \( \theta \). Define \( g \) as \( d \), for \( \theta \) a location parameter, or \( d.Q \), for \( \theta \) a scale parameter, and assume that \( g \) vanishes at 0 and 1 and is absolutely continuous with square integrable derivative \( g' \). Then, the problem of optimal spacing selection for the ABLUE of \( \theta \) is equivalent to the selection of a best set of breakpoints for \( L_2[0,1] \) approximation of \( g' \) by piecewise constants.

**Proof.** See Eubank, Smith and Smith (1981).

**Lemma 2.** Let \( Q \) denote the standardized \( q.f. \) for an absolutely continuous \( d.f. \: F(x; \theta) \) where \( \theta \) is either a location or scale parameter. Assume that \( F \) has a finite second moment and mean proportional to \( \theta \) and that \( Q \) is continuous and strictly monotone on \((0,1)\). Then, Problem 2 for \( F \) is equivalent to selecting optimal breakpoints for \( L_2[0,1] \) approximation of \( Q \) by piecewise constants.

**Proof.** First observe that, since \( F \) has a finite second moment, \( Q \in L_2[0,1] \).

For the case of \( \theta \) a location parameter, \( F \) may be expressed as \( H(x- \theta) \), where \( H \) is the distribution function corresponding to \( Q \). Now, for any set of strata boundaries \( a = x_0 < x_1 \leq \ldots < x_{k+1} = b \) it has been shown by Dalenius (1950) that the variance of \( \hat{\theta} \) is

\[
V(\hat{\theta}) = n^{-1} \sum_{i=1}^{k+1} [H(x_i - \theta) - H(x_{i-1} - \theta)]^2 \sigma_i^2
\] (8)
where

\[
[H(x_i-\theta)-H(x_{i-1}-\theta)]^2 = \int_{x_{i-1}}^{x_i} \left( \frac{2}{2} \int_{x_{i-1}}^{x_i} dH(x-\theta) - [H(x_i-\theta)-H(x_{i-1}-\theta)] \right) \cdot \left( \int_{x_{i-1}}^{x_i} dH(x-\theta) \right)^2.
\]

Making the change of variable \(x-\theta = Q(u)\) in (9), letting \(u_i = H(x_i-\theta)\) and simplifying gives

\[
(u_i-u_{i-1})^2 = \int_{u_{i-1}}^{u_i} Q(u)^2 du - (u_i-u_{i-1})^{-1} \left[ \int_0^1 Q(u) I(u_{i-1}, u_i)(u) du \right]^2
\]

where \(I(u_{i-1}, u_i)\) is the indicator function for \((u_{i-1}, u_i)\). Thus, \(nV(\theta)\) is now recognized as the squared \(L_2[0,1]\) error for the approximation of \(Q\) by piecewise constants with breakpoints at \(0 = u_0 < u_1 < \cdots < u_{k+1} = 1\).

Observing that the \(u_i\) and \(x_i\) are uniquely defined by \(\theta + Q(u_i) = x_i\) it follows that minimization of (8) with respect to the \(u_i\)’s or the \(x_i\)’s are equivalent problems. The case of scale parameter estimation is proven similarly.

As a result of Lemmas 1 and 2 questions concerning the equivalence of Problems 1 and 2 can now be viewed as questions regarding the equivalence of breakpoint selection problems for \(L_2[0,1]\) approximation by piecewise constants. This subject is treated by the next lemma.

**Lemma 3.** Let \(m_1\) and \(m_2\) be square integrable functions and assume that, for \(i=1,2\), \(m'_i\) is continuous, monotone and of one sign on \((0,1)\) with \(|m'_i|^{2/3}\) integrable. Then \(m_1\) and \(m_2\) will have the same optimal breakpoints for \(L_2[0,1]\) approximation by piecewise constants for all \(k\) if and only if there exists constants \(\alpha, \beta (\beta \neq 0)\) such that

\[
m_1(u) = \alpha + \beta m_2(u), \quad u \in (0,1).
\]
Proof. The sufficiency of (10) follows immediately upon noting that, in this event, the $L_2[0,1]$ errors for approximation of $m_1$ and $m_2$ are proportional with proportionality factor $|\beta|$. To establish its necessity let $(U_k)$ be a sequence of sets of optimal breakpoints for $m_1$ where $U_k = \{u_0^k, \ldots, u_{k+1}^k\}$ and $0 = u_0^k < \ldots < u_{k+1}^k = 1$. By hypothesis each $U_k$ is also optimal for $m_2$. Now, as in Barrow and Smith (1978), define piecewise linear functions $s_k$ with $s_k(u_i) = i/k+1$, $i=0, \ldots, k+1$. Then, using Theorem 1.1 of Burchard and Hale (1975) in conjunction with the proof of Theorem 3 in Barrow and Smith (1978), it is seen that

$$\lim_{k \to \infty} s_k(\tau) = \int_0^\tau |m_1'(u)|^{2/3} du / \int_0^1 |m_1'(t)|^{2/3} dt.$$  

(11)

However, as the $U_k$ are also optimal for $m_2$, it must be that

$$\lim_{k \to \infty} s_k(\tau) = \int_0^\tau |m_2'(u)|^{2/3} du / \int_0^1 |m_2'(t)|^{2/3} dt.$$  

(12)

The lemma now follows by equating (11) and (12) and differentiating.

To prove Theorem 1 take $m_1 = g_1', m_2 = g_2'$ in Lemma 3 and observe that the convexity or concavity of $g_1'$ is equivalent to $g_1''$ being of one sign on $(0,1)$. Theorems 2 and 3 can be obtained similarly by taking $m_1 = Q_1$, $m_2 = Q_2$ and $m_1 = g_1', m_2 = Q_2$.

Remark. Lemmas 1 and 2 have the consequence that problems of optimal spacing and stratification, when viewed in the quantile domain, have simple geometric interpretations as piecewise constant approximation problems. Relationships between the solutions to these problems for different distributions can, therefore, often be detected by merely graphing the appropriate functions.
4. **Extension to Optimal Grouping.** The results of Section 2 can be extended to include the following optimal grouping problem considered by Kulldorff (1961).

**Problem 3.** Choose group boundaries, \( a = x_0 < x_1 < \cdots < x_{k+1} = b \), that minimize the asymptotic variance of the maximum likelihood estimator of a location or scale parameter, \( \theta \), obtained using only the group boundaries and the proportion of sample elements within each group.

Theorems 1 and 3 can also be shown to apply to Problem 3 through the use of Theorem 4 of Adatia and Chan (1981). Their result states that for a given distribution Problems 1 and 3 are always equivalent (in the sense of (6)), provided the distribution satisfies certain regularity conditions specified by Kulldorff (1961, pg. 20). Using this fact and restricting attention to the special case of \( F_1 = F_2 = F \), \( \theta_1 = \theta_2 = \theta \), \( d_1 = d_2 = d \) and \( Q_1 = Q_2 = Q \), a quantile domain version of Theorem 5 of Adatia and Chan (1981) follows from Theorem 3.

**Corollary.** Let \( g \) represent either \( d \) or \( d'Q \), depending on whether \( \theta \) is a location or scale parameter, and assume that \( g \) and \( Q \) satisfy the hypotheses of Theorem 1 and 2 respectively. Then Problems 1-3 are equivalent for all \( k \) if and only if there exists constants \( a, \beta \) (\( \beta \neq 0 \)) such that

\[
g'(u) = a + \beta Q(u) , \quad u \in (0,1).
\]  

(13)

In comparing the Corollary with the results of Adatia and Chan observe that our approach dispenses with conditions requiring the existence of a sequence of strata boundaries, \( \{ b_k \} \), for which the corresponding estimators satisfy \( \lim_{k \to \infty} \sqrt{k} V(\hat{\theta}_k) = 0 \). We note that it follows immediately from (13) that Problems 1-3 are equivalent for either a normal or gamma distribution.
REFERENCES


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17. DISTRIBUTION STATEMENT (of the ABSTRACT entered in Block 20, if different from Report)

18. SUPPLEMENTARY NOTES

19. KEY WORDS (Continued on reverse side if necessary and identify by block number)

Approximation; Density-Quantile Function; Quantile Function; Optimal Grouping; Optimal Spacing; Optimal Stratification

20. ABSTRACT (Continued on reverse side if necessary and identify by block number)

The relationships between two distributions having the same solutions for problems of optimal spacing selection for the asymptotically best linear unbiased estimator of a location or scale parameter or for problems of optimal stratification for estimation of a population mean are investigated. Easily checked necessary and sufficient conditions under which two distributions have identical solutions to these problems are given in terms of their quantile and density-quantile functions. As an application of these results a quantile domain analog of a theorem due to Adatia and Chan (1981, Scand. Actuar. J.,...
(20. Continued) 193-202) on the equivalence of optimal grouping, spacing and stratification problems is obtained.