

FISHER INFORMATION AND SPLINE INTERPOLATION¹

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It is shown that among all cumulative distribution functions passing through $k \geq 2$ given points there is a unique one with minimal Fisher information; it is obtained by a curious type of spline interpolation. This answers some questions raised by D. G. Kendall and J. W. Tukey.

Problem. Estimate Fisher information $I(F) = \int (f'/f)^2 f dx$ from $k \geq 2$ given points of the cumulative distribution function F .

This problem, which was posed by J. W. Tukey in [3], has a distinguished "minimal" solution—the smallest possible Fisher information $I(F_0)$ for distributions passing through the given points. The minimizing distribution function F_0 is obtained by a curious and rather non-trivial type of spline interpolation, as follows. At the same time, an existence problem, mentioned in [1], page 33 f., is solved.

Assume that the given values are $F(\xi_i) = t_i$ with $-\infty < \xi_1 < \xi_2 < \dots < \xi_k < \infty$, $t_1 < t_2$, $t_{k-1} < t_k$. For convenience, we put $\xi_0 = -\infty$, $t_0 = 0$, $\xi_{k+1} = \infty$, $t_{k+1} = 1$. Then the solution F_0 can be described as follows.

- (i) $F_0(\xi_i) = t_i$, $i = 0, \dots, k + 1$;
- (ii) F_0 is two times continuously differentiable;
- (iii) the density $f_0 = F_0'$ is strictly positive, except that it vanishes on those intervals $[\xi_i, \xi_{i+1}]$ for which $t_i = t_{i+1}$;
- (iv) on each interval (ξ_i, ξ_{i+1}) the function $f_0^{3/2} / f_0^{1/2}$ is constant $= \lambda_i$, i.e.

$$\begin{aligned} (f_0(x))^{1/2} &= a_i e^{\lambda_i x} + b_i e^{-\lambda_i x}, & \text{if } \lambda_i > 0 \\ &= a_i x + b_i, & \text{if } \lambda_i = 0 \\ &= a_i \cos |\lambda_i| x + b_i \sin |\lambda_i| x, & \text{if } \lambda_i < 0. \end{aligned}$$

(v) There is one and only one F_0 satisfying (i) to (iv); it is the unique F_0 minimizing $I(F)$ subject to $F(\xi_i) = t_i$, $i = 1, \dots, k$. The value of the minimum is $I(F_0) = -4 \sum (t_{i+1} - t_i) \lambda_i$.

The proof is somewhat involved, but as very similar proofs already occur in [2] it is hardly necessary to present all the details.

Let \mathcal{F} be the set of all monotone functions satisfying

$$0 \leq F(\xi_i - 0) = F(\xi_i) \leq t_i \leq F(\xi_i + 0) \leq 1, \quad i = 1, \dots, k.$$

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Define

$$(1) \quad I(F) = \sup \frac{(\int \phi' dF)^2}{\int \phi^2 dF}$$

where ϕ ranges over the continuously differentiable functions with compact support. Note that $I(F)$ is the $L_2(F)$ -norm of the functional $\phi \rightarrow -\int \phi' dF$, and that

$$(2) \quad I(F) = \int (f'/f)^2 f dx$$

whenever one of the two sides of (2) is finite (the density $f = F'$ then must be absolutely continuous; see [2]).

LEMMA 1. *There is an $F_0 \in \mathcal{F}$ which minimizes $I(F)$.*

PROOF. As $I(\cdot)$ is the supremum of a family of vaguely continuous functions, it is lower semicontinuous, and hence attains its minimum on the vaguely compact set \mathcal{F} .

LEMMA 2. *If F_0 minimizes $I(F)$ and if $f_0 > 0$ except on the intervals $[\xi_i, \xi_{i+1}]$ where $t_i = t_{i+1}$, then F_0 is the unique member of \mathcal{F} minimizing $I(F)$.*

PROOF. Assume that $f_0 > 0$ for all x (the general case is treated analogously). Let $F_1 \in \mathcal{F}$ be any distribution having finite Fisher information; put

$$F_\epsilon = (1 - \epsilon)F_0 + \epsilon F_1, \quad 0 \leq \epsilon \leq 1.$$

We may assume without loss of generality that $f_1 > 0$ (otherwise replace F_1 by, say, $(F_0 + F_1)/2$, which has this property). As $I(\cdot)$ is convex (see [2]), monotone convergence and Fatou's lemma yield, respectively,

$$(3) \quad \frac{d}{d\epsilon} I(F_\epsilon) = \int \left[2 \frac{f'_\epsilon}{f_\epsilon} (f'_1 - f'_0) - \left(\frac{f'_\epsilon}{f_\epsilon} \right)^2 (f_1 - f_0) \right] dx$$

and

$$(4) \quad \frac{d^2}{d\epsilon^2} I(F_\epsilon) \geq 2 \int \left(\frac{f'_1}{f_1} - \frac{f'_0}{f_0} \right)^2 \frac{f_0^2 f_1^2}{f_\epsilon^3} dx$$

(primes always denote differentiation with respect to x).

If also F_1 minimizes Fisher information, then, by convexity, $I(F_\epsilon) = I(F_0)$ for all $\epsilon \in [0, 1]$ and (4) implies

$$(5) \quad \frac{f'_1}{f_1} = \frac{f'_0}{f_0} \quad \text{a.e.}$$

We integrate this and obtain

$$(6) \quad f_1 = c f_0$$

for some constant c (it was overlooked in [2], that this is false unless $f_0 > 0$). Since

$$I(F_0) = I(F_1) = \int (f'_1/f_1)^2 f_1 dx = \int (f'_0/f_0)^2 c f_0 dx = c I(F_0),$$

we must have $c = 1$.

Now consider the following auxiliary problems.

Problem A. Let $-\infty < \xi < \eta < \infty$ and assume that $F(\xi), F(\eta), f(\xi), f(\eta)$ are given and fixed. The problem is to minimize the convex functional

$$(7) \quad I_{\xi\eta}(F) = \int_{\xi}^{\eta} (f'/f)^2 f \, dx .$$

If this problem has a solution F_0 , then (3) implies that

$$(8) \quad \left[\frac{d}{d\varepsilon} I_{\xi\eta}(F_{\varepsilon}) \right]_{\varepsilon=0} = \int_{\xi}^{\eta} \left\{ 2 \frac{f_0'}{f_0} (f_1' - f_0') - \left(\frac{f_0'}{f_0} \right)^2 (f_1 - f_0) \right\} dx \geq 0$$

for all F_1 with finite $I_{\xi\eta}(F_1)$ and satisfying the side conditions. Conversely, if (8) holds for all these F_1 , then F_0 minimizes (7).

If $f_0 > 0$ on $[\xi, \eta]$, then (8) can be integrated by parts to give

$$(9) \quad \begin{aligned} \left[\frac{d}{d\varepsilon} I_{\xi\eta}(F_{\varepsilon}) \right]_{\varepsilon=0} &= - \int_{\xi}^{\eta} \left\{ 2 \left(\frac{f_0'}{f_0} \right)' + \left(\frac{f_0'}{f_0} \right)^2 \right\} (f_1 - f_0) \, dx \\ &= -4 \int_{\xi}^{\eta} \frac{f_0^{\frac{1}{2}''}}{f_0^{\frac{1}{2}}} (f_1 - f_0) \, dx . \end{aligned}$$

Provided $f_0 > 0$ on $[\xi, \eta]$, it follows that F_0 minimizes (7) iff f_0 satisfies the differential equation

$$(10) \quad \frac{f_0^{\frac{1}{2}''}}{f_0^{\frac{1}{2}}} = \lambda = \text{const. on } [\xi, \eta] .$$

It is fairly easy to see that the side conditions can always be met with a solution of (10) which is strictly positive in (ξ, η) . If $f(\xi) = f(\eta) = 0$, this is trivial to show. Assume therefore that $f(\xi)$ and $f(\eta)$ are not both 0. For $\lambda > 0$, the solution of (10) is

$$(11) \quad (f_0(x))^{\frac{1}{2}} = ae^{\lambda x} + be^{-\lambda x}$$

with a, b not both being negative. If $ab \geq 0$, then $f_0(x) > 0$ for all x ; if $ab < 0$, then $f_0(x)$ is monotone and hence > 0 in (ξ, η) . For each value of λ in the range $(-\pi/(\eta - \xi), \infty)$ there is a unique positive solution to (10) which takes the given values of f_0 at ξ and η , and as λ decreases from ∞ to $-\pi/(\eta - \xi)$, this solution increases from 0 to ∞ , hence it is also possible to obtain the given value of $F_0(\eta) - F_0(\xi)$.

Alternatively, one can also show that a solution of (10) which has an isolated zero in (ξ, η) cannot correspond to a minimum of (7); see below (end of Problem B).

Problem B. Let $-\infty < \xi < \eta < \zeta < \infty$ and assume that $F(\xi), F(\eta), F(\zeta), f(\xi)$ and $f(\zeta)$ are given and fixed. The problem is to minimize $I_{\xi\zeta}(\cdot)$. It is evident from (2) that f_0 must be continuous at η , and it must satisfy (10) in each of the intervals $(\xi, \eta), (\eta, \zeta)$. I assert now that f_0' must be continuous at η .

PROOF. Assume first that $f_0(\eta) > 0$. Then a discontinuity in f_0' corresponds to a Dirac δ in $f_0^{\frac{1}{2}''}/f_0^{\frac{1}{2}}$ at η , and it is easy to verify that by choosing f_1 such that $f_1 - f_0$ is symmetric around η and nonzero at η it is possible to achieve a strictly

negative value for (9). If $f_0(\eta) = 0$, then a glance at the solutions of (10) shows that $f_0'(\eta) = 0$ and f_0' is trivially continuous.

Furthermore, $f_0(\eta) = 0$ can happen only when either $F(\xi) = F(\eta)$ or $F(\eta) = F(\zeta)$. The proof is based on the following idea: if $f_0(\eta) = 0$ and f_0 satisfies (10), then $f_0'(x)/f_0(x) \sim 2/(x - \eta)$ near η , and the integral (8) diverges to $-\infty$ if f_1 is smooth and nonzero at η . This allows to show that F_0 then does not correspond to a minimum.

Problem C. Minimization of $I_{-\infty, \xi_1}(\cdot)$. The unbounded intervals $(-\infty, \xi_1)$ and (ξ_k, ∞) need a special treatment; it suffices to consider one of them. To exclude trivialities, assume $0 < t_1 < t_2$ (here $k \geq 2$ is essential!). If $f_0(\xi_1) = 0$, then F_0 does not correspond to a minimum (see the preceding paragraph); if $f_0(\xi_1) > 0$, the density must be of the form

$$(12) \quad f_0(x) = a^2 e^{-2\lambda|x|}$$

on $(-\infty, \xi_1)$, with $\lambda > 0$. Hence (9) reads

$$(13) \quad \left[\frac{d}{d\varepsilon} I_{-\infty, \xi_1}(F_\varepsilon) \right]_{\varepsilon=0} = -4\lambda \int_{-\infty}^{\xi_1} (f_1 - f_0) dx.$$

In order that (13) is always ≥ 0 , we must make the total mass $F_0(\xi_1) - F_0(-\infty)$ as large as possible, that is $F_0(-\infty) = 0$, and similarly $F_0(+\infty) = 1$; hence F_0 is a genuine probability distribution.

We are now ready to collect and put together these many pieces of evidence. Take a distribution F_0 minimizing $I(\cdot)$; its existence is asserted in Lemma 1. The auxiliary problems A, B, C show that F_0 satisfies (10) in each of the intervals (ξ_i, ξ_{i+1}) and that it satisfies the assumptions of Lemma 2; hence F_0 is the unique distribution minimizing $I(\cdot)$. Moreover, it satisfies (i) to (iv). On the other hand, any F_0 satisfying (i) to (iv) has the property that

$$\left[\frac{d}{d\varepsilon} I(F_\varepsilon) \right]_{\varepsilon=0} = -4 \int \frac{f_0^{3/2}''}{f_0^{3/2}} (f_1 - f_0) dx = 0;$$

hence $I(\cdot)$ is stationary at F_0 ; since $I(\cdot)$ is convex, F_0 corresponds to a minimum. This terminates the proof of the main assertion of this paper.

When the grid (ξ_1, \dots, ξ_k) is refined, then $I(F_0)$ converges to the true value $I(F)$. This follows at once from the remark that $I(\cdot)$ is lower semicontinuous and that F_0 converges weakly to F , thus $\liminf I(F_0) \geq I(F)$, but $I(F_0) \leq I(F)$. I have not yet been able to find the rate of convergence.

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