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SMOOTHING SPLINE IN A CONVEX CLOSED SET OF HILBERT SPACE

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A characterisation of a smoothing spline is sought in a convex closed set C of Hilbert space: $\min\{\alpha ||Tx||_Y^2 + ||Ax - z||_Z^2, x \in C\}$, T and A are linear operators. A representation of the solution is obtained in the terms of the kernels of the above operators, of the dual operators T^* , A^* and of the dual cone C^0 . A particular case is considered when T is the differential operator and A is the operator-trace of a function.

Let X, Y, Z be Hilbert spaces with scalar products respectively $(,)_X, (,)_Y, (,)_Z$. We are given linear bounded operators

$$A: X \to Z, \quad T: X \to Y.$$

Consider the operator equation $Ax = z_0, z_0 \in Z$.

1. If $A^{-1}(z_0) \neq \emptyset$, then $\sigma \in X$ is called *an interpolating spline*, if the following minimum is reached

(1)
$$||T\sigma||_Y^2 = \min_{x \in A^{-1}(z_0)} ||Tx||_Y^2.$$

2. If $A^{-1}(z_0) = \emptyset$, we introduce a real parameter $\alpha > 0$ and construct a quadratic functional

(2)
$$\phi_{\alpha}(x) = \alpha \|Tx\|_{Y}^{2} + \|Ax - z_{0}\|_{Z}^{2}.$$

We say that $\sigma_* \in X$ is a smoothing spline, if

(3)
$$\phi_{\alpha}(\sigma_*) = \min_{x \in X} \phi_{\alpha}(x).$$

Characterisations of the solutions of problems (1) and (3) are given in [5].

A certain shape of the interpolating or smoothing spline is required in many applied problems. The characterisation of such conditions can be often described by a set $C \subset X$, which is convex and closed.

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Chui, Deutsch and Ward give a characterisation of the solution of the problem for an interpolating spline in a convex set of Hilbert space ([3, 4])

(4)
$$\min_{Ax=z_0, x\in C} ||x||^2.$$

In [2] a particular case of monotonicity is considered for both interpolating and smoothing splines. The characterisation is done from the point of view of a general optimisation problem in the terms of the Frechet-derivative and the polar cone.

We shall consider here the problem of finding a smoothing spline in a convex closed set of Hilbert space. That is, σ_* is sought so that

(5)
$$\phi_{\alpha}(\sigma_{*}) = \min_{x \in C \subset X} \phi_{\alpha}(x),$$

where $\phi_{\alpha}(x)$ is the functional in (2). This problem arises for example, if the data are corrupted by noise and one does not require exact interpolation, but a special form of spline is required.

A new linear operator L can be defined ([5]), which is acting on $F = Y \times Z$. If $f_1 = [y_1, z_1], y_1 \in Y, z_1 \in Z, f_2 = [y_2, z_2], y_2 \in Y, z_2 \in Z$, we define a scalar product in F by

$$(f_1, f_2)_F = ([y_1, z_1], [y_2, z_2])_F := \alpha(y_1, y_2)_Y + (z_1, z_2)_Z.$$

Let L be the linear bounded operator

$$L: X \to F, Lx = [Tx, Ax],$$

and let $a = [0_Y, z_0]$ be an element of F.

LEMMA 1. $\phi_{\alpha}(x) = ||Lx - a||_F^2$, where $a = [0_Y, z_0]$.

PROOF: By the definitions

$$(Lx - a, Lx - a)_F = ([Tx, Ax] - [0_Y, z_0], [Tx, Ax] - [0_Y, z_0])_F$$

= $([Tx, Ax - z_0], [Tx, Ax - z_0])_F = \alpha (Tx, Tx)_Y + (Ax - z_0, Ax - z_0)_Z$
= $\alpha ||Tx||^2 + ||Ax - z_0||^2 = \phi_{\alpha}(x).$

Therefore $\phi_{\alpha}(x) = (Lx - a, Lx - a)_F = ||Lx - a||_F^2$. Then the problem (5) is equivalent to

(6)
$$\min_{f \in K} \|f - a\|^2$$

where $K = L(C) = \{ [y, z] \in Y \times Z : y = Tx, z = Ax, x \in C \}$. Denote the kernels of T and A respectively by

$$\ker T = \{x \in X : Tx = 0_Y\}, \ \ker A = \{x \in X : Ax = 0_Z\}.$$

[2]

LEMMA 2. If T and A are linear bounded operators and ker $T \cap \ker A = \{0_X\}$, then L is a linear bounded continuous operator and ker $L = \{0_X\}$

PROOF: L is a linear bounded operator, obviously. It follows it is continuous.

Let us show that ker $L = \{0_X\}$. If $x \in \ker L$, that is, $Lx = 0_F$, then $Lx = [Tx, Ax] = [0_Y, 0_Z]$, therefore $Tx = 0_Y$, $Ax = 0_Z$, and $x \in \ker T \cap \ker A = 0_X$, or ker $L = 0_X$.

The following lemma follows from the inverse operator theorem.

LEMMA 3. If T and A are linear bounded operators, ker $T \cap \ker A = 0_X$, and L(X) is closed, then there exists $L^{-1} : L(X) \subset F \to X$ and L^{-1} is a linear bounded continuous operator, too.

Here $L(X) = \{ [y, z] : y = Tx, z = Ax, x \in X \}.$

We shall find conditions for closeness of L(X) to be closed.

LEMMA 4. L(X) is closed if and only if ker T + ker A is closed.

PROOF: L(X) is closed if and only if $L^*(F) = T^*(Y) + A^*(Z)$ is closed in X, if and only if ker $T^{\perp} + \ker A^{\perp}$ is closed if and only if ker $T + \ker A$ is closed.

LEMMA 5. If C is a closed convex subset of X, ker $T \cap \text{ker } A = \{0_X\}$, and ker T + ker A is closed, then K = L(C) is a closed and convex subset of F.

PROOF: Let $f_1, f_2 \in L(C), \lambda_1 \ge 0, \lambda_2 \ge 0, \lambda_1 + \lambda_2 = 1$. We shall show that $\lambda_1 f_1 + \lambda_2 f_2 \in L(C)$, too. There exists unique $x_1 \in C : f_1 = Lx_1$, and $x_2 \in C : f_2 = Lx_2$. We have $\lambda_1 f_1 + \lambda_2 f_2 = \lambda_1 Lx_1 + \lambda_2 Lx_2 = L(\lambda_1 x_1 + \lambda_2 x_2) \in L(C)$, because $\lambda_1 x_1 + \lambda_2 x_2 \in C$.

Let's show, that L(C) is closed. If $\{f_n\} \to f, f_n \in L(C)$, we shall show that $f \in L(C)$.

There exists unique $x_n \in C$: $f_n = Lx_n, Lx_n \to f$. Applying the inverse continuous operator L^{-1} , it follows $L^{-1}Lx_n \to L^{-1}f = x$.

So we have $x_n \to x$, but C is closed, therefore $x \in C$. It means $f = Lx \in L(C)$.

THEOREM 1. If C is a convex closed subset of X, ker $T \cap \ker A = 0_X$ and ker T + ker A is closed, then the problem (5) has the unique solution

$$\sigma_* = L^{-1} P_{L(C)}(a),$$

where $P_{L(C)}(a)$ denotes the orthogonal projection of a on L(C).

PROOF: A classical result ([5, Theorem 2.1.2]) shows, that there exist unique solution of the problem (6) $f_* \in K$, such that

$$||f_* - a||^2 = \min_{f \in K} ||f - a||^2.$$

The point $f_* \in L(C)$, in which min ||f - a|| is reached, is the orthogonal projection of a on L(C), that is, $f_* = P_{L(C)}(a)$. But $f_* = L\sigma_*$ for some σ_* , and L is converse, (in according with Lemma 3 and 4), therefore the solution has the form

(7)
$$\sigma_* = L^{-1}(P_{L(C)}(a)).$$

Further we shall omit the brackets in L(C) and denote LC := L(C). Define the dual operator A^* of A by

$$(z,Ax) = (A^*z,x)$$

for all $z \in Z, x \in X$.

We denote the dual cone of C by

$$C^{0} = \left\{ x \in X : (x, y) \leq 0, \forall y \in C \right\}.$$

It is easy to see, that

(8)
$$f_* = P_{LC}(a)$$
 if and only if $a - f_* \in (LC - f_*)^0 = (LC)^0 \cap f_*^{\perp}$.

THEOREM 2. Problem (5) has the unique solution σ_* if and only if

(9)
$$-\alpha T^*T\sigma_* + A^*(z_0 - A\sigma_*) \in C^0,$$

and

(10)
$$\Phi_{\alpha}(\sigma_{*}) = (z_0 - A\sigma_{*}, z_0).$$

PROOF: By (8), $a - f_* \in (LC)^0$ means, that

 $(a - f_*, L\sigma) \leq 0, \forall \sigma \in C.$

But $f_* \in L(C)$, therefore there exist unique σ_* such that $f_* = L\sigma_* = [T\sigma_*, A\sigma_*]$. Then

$$(a - f_*, L\sigma) = ([0_Y, z_0] - [T\sigma_*, A\sigma_*], [T\sigma, A\sigma]) = ([-T\sigma_*, z_0 - A\sigma_*], [T\sigma, A\sigma])$$

(11)
$$= -\alpha(T\sigma_*, T\sigma) + (z_0 - A\sigma_*, A\sigma) \leq 0.$$

Therefore

$$(-\alpha T^*T\sigma_*,\sigma) + (A^*(z_0 - A\sigma_*),\sigma) \leq 0, \ \sigma \in C.$$

This means, that

$$-\alpha T^*T\sigma_* + A^*(z_0 - A\sigma_*) \in C^0,$$

and (9) has been proved.

Again from (8) $a - f_* \in f_*^{\perp}$. It follows, that $(a - f_*, f_*)_F = 0$. But $f_* = (T\sigma_*, A\sigma_*)$, so

$$([0_Y, z_0] - [T\sigma_*, A\sigma_*], [T\sigma_*, A\sigma_*]) = ([-T\sigma_*, z_0 - A\sigma_*], [T\sigma_*, A\sigma_*]) = -\alpha ||T\sigma_*||^2 + (z_0 - A\sigma_*, A\sigma_*) = -\alpha ||T\sigma_*||^2 - (A\sigma_* - z_0, A\sigma_* - z_0) - (A\sigma_* - z_0, z_0) = -\Phi_\alpha(\sigma_*) - (A\sigma_* - z_0, z_0) = 0.$$

Therefore $\Phi_{\alpha}(\sigma_{*}) = (z_0 - A\sigma_{*}, z_0)$. Note that equality in (11) is reached only for the solution σ_{*} . We obtain then equation (10), and the theorem has been proved.

We shall look for the solution of problem (5) in a proper basis.

Let k_1, k_2, \ldots, k_N be linearly independent elements of X, and $A: X \to Z = Z^N$. The action of the operator A may be represented by

$$A\sigma = ((k_1, \sigma), (k_2, \sigma), \dots, (k_N, \sigma)).$$

Let K be the space of linear combinations of k_1, k_2, \ldots, k_N . The dual operator satisfies

$$A^*\lambda = \sum \lambda_i k_i.$$

The Hilbert space Y may be represented as a direct sum

$$Y = (T \ker A) \oplus (T \ker A)^{\perp}$$

For the solution $\sigma_* \in X$ there exists $y_0 \in T \ker A$, with $y_0 = Tx_0$ for some element $x_0 \in \ker A$, and there exists $y \in (T \ker A)^{\perp}$, so that

(12)
$$T\sigma_* = y_0 + y = Tx_0 + y.$$

The following equations can be proved easily.

LEMMA 6.

- (1) ker $A = K^{\perp}$.
- (2) $(TK^{\perp})^{\perp} = T^{*-1}(H)$, where $H = K \cap (\ker T)^{\perp}$.
- (3) If ker $T \cap \ker A = 0_X$ and dim ker $T = q < \infty$, then

$$\dim H = \dim K - \dim(\ker T) = N - q.$$

(4) $(T^{*-1}h)(t) = (h(x), G_+(x-t))_X$, where $h \in (\ker T)^{\perp}$.

Here $G_+(x-t)$ is the Green's function with $TG_+(x-t) = \delta_t$, and $\delta_t(v) = v(t), v \in X$. An algorithm for finding a basis in $(T \ker A)^{\perp}$ follows if we use Lemma 6.

(1) A basis for $H = K \cap (\ker T)^{\perp}$ is looked for

$$h_i = \sum_{j=1}^{N} h_{ij} k_j, i = 1, 2, \dots, N - q.$$

(2) If e_1, e_2, \ldots, e_q is a basis for ker T, then

$$0 = (h_i, e_k) = \sum_{j=1}^{N} h_{ij}(k_j, e_k), i = 1, 2, \dots, N - q, k = 1, 2, \dots, q.$$

- (3) $f_i = T^{*-1}(h_i), i = 1, 2, ..., N q$ is a basis for $T^{*-1}(H) = (TK^{\perp})^{\perp}$ $= (T \ker A)^{\perp}.$
- (4) For every $y \in (T \ker A)^{\perp}$ there exist $\lambda_1, \lambda_2, \ldots, \lambda_{N-q}$, so that

$$y = \sum_{i=1}^{N-q} \lambda_i f_i.$$

Now from (12) we have the representation

(13)
$$T\sigma_{\star} = Tx_0 + \sum \lambda_i f_i$$

Let's introduce the matrices

(14)
$$H = (h_{ij})_{i=1,\dots,N-q}^{j=1,\dots,N}, \quad F = \left((f_i, f_j) \right)_{i=1,\dots,N-q}^{j=1,\dots,N-q}$$

(15)
$$\lambda = (\lambda_1, \lambda_2, \dots, \lambda_{N-q}), \quad r = (r_1, r_2, \dots, r_N).$$

LEMMA 7. For every $y \in (T \ker A)^{\perp}$ there exists $r \in Z^N$, so that

 $T^* u = A^* r, \ r = \lambda H.$

PROOF: From the above there exist $\lambda_1, \lambda_2, \ldots, \lambda_{N-q}$, so that $y = \sum \lambda_i f_i$. Thus

$$T^*y = T^*\left(\sum \lambda_i f_i\right) = \sum \lambda_i (T^*f_i) = \sum \lambda_i h_i = \sum \lambda_i \sum h_{ij} h_{ij} k_j$$
$$= \sum \sum \lambda_i h_{ij} k_j = \sum r_j k_j = A^*r, r = (r_1, r_2, \dots, r_N).$$

Here

$$r_j = \sum_{i=1}^{N-q} \lambda_i h_{ij},$$

or, using a matrix form,

(16)

(16)
$$r = \lambda H.$$

Let us denote $v = A\sigma_* = ((k_1, \sigma_*), \dots, (k_N, \sigma_*))$. There exists a relation between λ and v .

LEMMA 8. $\lambda F = v H^T$.

PROOF: We have in (13) $T\sigma_* = Tx_0 + \sum \lambda_i f_i$. Thus

(17)
$$(T\sigma_*, f_j) = (Tx_0, f_j) + \left(\sum \lambda_i f_i, f_j\right) = (x_0, T^*f_j) + \sum \lambda_i (f_i, f_j).$$

But $(x_0, T^*f_j) = (x_0, h_j) = 0$ because of $x_0 \in \ker A = K^{\perp}, h_j \in H \subset K$. On the other hand,

(18)
$$(T\sigma_*, f_j) = (\sigma_*, h_j) = \left(\sigma_*, \sum h_{jl}k_l\right) = \sum h_{jl}(k_l, \sigma_*) = \sum h_{jl}v_l$$

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Comparing the right sides of (17) and (18), it follows

$$\lambda F = v H^T.$$

LEMMA 9. If e_1, e_2, \ldots, e_q is a basis for ker T and $B = (b_{j,k})_{j=1,2,\ldots,N}^{k=1,2,\ldots,q}$, with $b_{j,k}$ $= (k_j, e_k), \text{ then } rB = 0.$

PROOF: From (2) and (1) of the algorithm above, it follows that

$$0 = \sum \lambda_i(h_i, e_k) = \sum \lambda_i \left(\sum h_{ij}k_j, e_k\right) = \sum \lambda_i h_{ij}(k_j, e_k) = \sum r_j b_{kj}, k = 1, 2, \dots, q.$$
Therefore $r B = 0$

Therefore rB = 0.

From the representation (13) of the solution and Lemma 7 it follows that there exist $x_0 \in \ker A$ and $r \in Z^N$, so that

$$T^*T\sigma_* = T^*Tx_0 + A^*r.$$

The conditions (9) and (10) take the form

$$-\alpha T^*Tx_0 + A^*(-\alpha r + z_0 - v) \in C^0,$$

$$\alpha \left\| Tx_0 + \sum \lambda_i f_i \right\|^2 = (z_0 - v, v).$$

Let us remark

$$\left\|Tx_0 + \sum \lambda_i f_i\right\|^2 = \|Tx_0\|^2 + \sum \sum \lambda_i (f_i, f_j)\lambda_j = \|Tx_0\|^2 + \lambda F \lambda^T.$$

The following theorem is a consequence of Lemmas 7, 8 and 9.

THEOREM 3. The solution σ_* of the problem (5) may be represented in the form

$$T\sigma_* = Tx_0 + \sum \lambda_i f_i, \ x_0 \in \ker A,$$

if and only if x_0 and λ satisfy

$$-\alpha T^*Tx_0 + A^*(-\alpha r + z_0 - v) \in C^0$$

$$\alpha(||Tx_0||^2 + \lambda F\lambda^T) = (z_0 - v, v),$$

where $r \in Z^N$, $v \in Z^N$ are related to λ by

$$r = \lambda H, \quad \lambda F = v H^T, \quad rB = 0.$$

Let us consider problem (5) in the following situation. The knots

$$a = t_1 < t_2 < \ldots < t_N = b$$

[7]

(19)

and values z_1, z_2, \ldots, z_N are given in the interval [a, b]. Let $X = W_2^n[a, b]$ be the Sobolev space of functions with the usual norm

$$\|f\|_{W_2^n}^2 = \sum_{j=0}^n \|f^{(j)}\|_{L^2[a,b]}^2$$

Let $Y = L_2[a, b]$ be the space of square integrable functions. Let $T = \frac{d^n}{dt^n}$ and let $A: W_2^n \to Z = Z^N$ be the operator-trace of the function,

$$Au = (u(t_1), u(t_2), \ldots, u(t_N))$$

Let

$$C = \Big\{ \sigma \in W_2^m[a,b] : \frac{d^m \sigma}{dt^m} \ge 0 \Big\},\$$

the subset of m- convex functions in X, where $m \leq n$.

LEMMA 10.

- (1) A and T are linear bounded operators.
- (2) C is a closed convex subset of X.
- (3) If the number of the knots N is greater or equal to the order of the differentiation n, then

$$\ker\left\{\frac{d^n}{dt^n}\right\}\cap \ker A=0_X,$$

and ker T + ker A is closed in X.

PROOF: To prove (3) note ker $\frac{d^n}{dt^n} = \left\{x : \frac{d^n x}{dt^n} = 0\right\}$ consists of polynomials of order smaller then n; while ker $A = \left\{u : u(t_i) = 0, i = 1, 2, ..., N\right\}$ contains functions with zeros at these N points. The intersection of these kernels is empty, because a polynomial of degree smaller than n cannot have N > n - 1 zeros.

Obviously ker T and ker A are closed, their sum is also closed.

Lemma 10 and Theorem 3 give the following result.

THEOREM 4. The problem

(20)
$$\phi_{\alpha}(\sigma_{*}) = \min_{\sigma \in C} : \left\{ \phi(\sigma) = \alpha \left\| \frac{d^{n}\sigma}{dt^{n}} \right\|_{L_{2}}^{2} + \sum_{i=1}^{N} (\sigma(t_{i}) - z_{i})^{2} \right\}$$

has unique solution σ_* for $N \ge n$. When $\sigma^{(m)} \ge 0$,

$$\sum_{i=1}^{N} (z_i - \sigma_*(t_i)) \sigma(t_i) \leq \alpha \int \sigma^{(n)}(t) \sigma_*^{(n)}(t) dt,$$

with equality only for the solution $\sigma_*(t)$.

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PROOF: The problem has unique solution by the previous results. The condition (9) in this case is

$$-\alpha \Big(\frac{d^n}{dt^n}\Big)^*\Big(\frac{d^n\sigma_*}{dt^n}\Big) + \sum_{i=1}^n (z_i - \sigma_*(t_i))k_i \in C^0.$$

From (11) it follows, that for all σ with $\sigma^{(m)} \ge 0$ it must be performed

$$\sum_{i=1}^{N} (z_i - \sigma_*(t_i)) \sigma(t_i) \leq \alpha \int \frac{d^n \sigma_*}{dt^n} \frac{d^n \sigma}{dt^n} dt.$$

Equality is achieved only for the solution $\sigma = \sigma_*$

$$\alpha \|\sigma_*^{(n)}\|_{L^2}^2 = \sum (z_i - \sigma_*(t_i))\sigma_*(t_i).$$

In fact this equality is equivalent to the condition (10).

For $T = \frac{d^n}{dt^n}$ it is known, that

$$G_+(x-t) = \frac{(x-t)_+^{n-1}}{(n-1)!}.$$

A basis for ker T is $\{1, t, t^2, \ldots, t^{n-1}\}$, and therefore dim ker T = q = n. Then

$$h_i = \sum_{j=1}^N h_{ij} k_j,$$

so that $(h_i, t^k) = 0, i = 1, 2, ..., N - n, k = 0, 1, ..., n - 1$. It follows, that

$$\sum_{j=1}^{N} h_{ij} t_j^k = 0, i = 1, 2, \dots, N - n, k = 0, 1, \dots, n - 1.$$

We have

$$f_i(t) = \left(\sum_{j=1}^N h_{ij}k_j(x), \frac{(x-t)_+^{n-1}}{(n-1)!}\right) = \sum_{j=1}^N h_{ij}\frac{(t_j-t)_+^{n-1}}{(n-1)!}, i = 1, 2, \dots, N-n.$$

By Theorem 3 the solution of (20) has the representation

$$\sigma_*^{(n)}(t) = x_0^{(n)} + \sum \lambda_i \sum \frac{h_{ij}(t_j - t)_+^{n-1}}{(n-1)!} = x_0^{(n)} + \sum r_j \frac{(t_j - t)_+^{n-1}}{(n-1)!}$$

where the condition rB = 0 is equivalent to

(21)
$$\sum_{j=1}^{N} r_j t_j^{k-1} = 0, k = 1, 2, \dots, n.$$

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On integrating n times,

$$\sigma_*(t) = x_0(t) + \sum_{j=1}^N r_j \frac{(t_j - t)_+^{2n-1}}{(2n-1)!} + \sum_{k=0}^{n-1} c_k t^k.$$

The function

$$s(t) = \sum_{j=1}^{N} r_j \frac{(t_j - t)_+^{2n-1}}{(2n-1)!} + \sum_{k=0}^{n-1} c_k t^k$$

under the constraints (21) is a natural spline ([1]) of degree 2n-1 with knots t_1, t_2, \ldots, t_N . Since the restriction of s(t) over $(-\infty, a = t_1)$ and $(t_N = b, \infty)$ is the polynomial $\sum c_k t^k$ of degree n-1, we have the following result.

THEOREM 5. The solution of the problem (20) is a sum of a function $x_0(t)$ with zero-crossings t_1, t_2, \ldots, t_N and a natural spline s(t) of degree 2n - 1 with knots in these points.

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