

REGRESSION DESIGN FOR SOME EQUIVALENCE CLASSES OF KERNELS¹

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Earlier results on asymptotically optimal sequences of regression designs for autoregressive stochastic processes are extended (nearly) to the equivalence classes of such processes.

1. Introduction. Let $\{Y(t), t \in T\}$ be a stochastic process of the form

$$Y(t) = \theta f(t) + X(t)$$

where θ is an unknown constant, $f(t)$ is a known function on T , T is a closed bounded interval which we take to be $[0, 1]$, and $\{X(t), t \in T\}$ is a zero-mean Gaussian stochastic process with known continuous covariance kernel Q , $EX(t)X(t') = Q(t, t')$. The regression design problem is to choose an n -point subset (or "design") T_n ,

$$T_n = \{t_1 < t_2 < \cdots < t_n, t_i \in T\}$$

so that the variance $\sigma_{T_n}^2$ of the Gauss-Markov estimate of θ given $\{Y(t), t \in T_n\}$ is as small as possible.

This problem has been considered by Sacks and Ylvisaker, Wahba, and Hájek and Kimeldorf [3], [11], [12], [13], [14], [16] for various special cases of Q . It is known that $\sigma_{T_n}^2$ is bounded away from 0 as $\Delta = \max_i |t_{i+1} - t_i|$ tends to 0 if and only if $f \in \mathcal{H}_Q$, where \mathcal{H}_Q is the unique reproducing kernel Hilbert space (RKHS) with reproducing kernel (RK) Q , see [8]. It will be assumed that the reader is familiar with the basic properties of RKHS as given in [8], [16], see also [1].

For fixed $t \in T$, let Q_t represent the evaluation functional at t in \mathcal{H}_Q , that is

$$\langle Q_t, f \rangle_Q = f(t), \quad f \in \mathcal{H}_Q$$

and

$$Q_t(t') = Q(t, t'),$$

where $\langle \cdot, \cdot \rangle_Q$ is the inner product in \mathcal{H}_Q .

Let P_{T_n} be the projection operator in \mathcal{H}_Q onto the subspace spanned by $\{Q_t, t \in T_n\}$. It is well known that if $f \in \mathcal{H}_Q$, then $\sigma_{T_n}^{-2} = \|P_{T_n} f\|_Q^2$ and $\sigma_T^{-2} = \|f\|_Q^2$, where $\|\cdot\|_Q$ is the norm in \mathcal{H}_Q and σ_T^2 is the variance of the Gauss Markov estimate of θ , given $\{Y(t), t \in T\}$. Hence $\sigma_{T_n}^2$ is minimized by minimizing $\|f - P_{T_n} f\|_Q^2$. From this point of view, the problem becomes one of choosing

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an optimal subspace in \mathcal{H}_Q of the form $\text{span} \{Q_t, t \in T_n\}$, for the purpose of approximating the given element f . In this context, the problem has been considered by Karlin [5], [6]. The solution also has applications to the approximate solution of linear differential and integral equations, see [17], [18].

We suppose that $\{X(t), t \in T\}$ has exactly $m - 1$ quadratic mean derivatives. This entails that the functions $Q_t^{(\nu)}(\cdot)$ defined by

$$Q_t^{(\nu)}(\cdot) = \left(\frac{\partial^\nu}{\partial s^\nu} \right) Q(s, \cdot) \Big|_{s=t}$$

are all well defined and in \mathcal{H}_Q , for $t \in T$ and $\nu = 1, 2, \dots, m - 1$. Let P_{m, T_n} be the projection operator in \mathcal{H}_Q onto the subspace of \mathcal{H}_Q spanned by

$$(1.1) \quad \{Q_t^{(\nu)}, t \in T_n, \nu = 0, 1, \dots, m - 1\}.$$

The optimal experimental design problem becomes tractable if we attempt to minimize $\|f - P_{m, T_n} f\|_Q$ rather than $\|f - P_{T_n} f\|_Q$, and the results are still useful, because of the relation ([16], (1.15))

$$(1.2) \quad \inf_{T_{nm}} \|f - P_{T_{nm}} f\|_Q \leq \inf_{T_n} \|f - P_{m, T_n} f\|_Q \leq \inf_{T_n} \|f - P_{T_n} f\|_Q.$$

Further information about the role of derivatives may be found in Karlin [6], especially Theorem 3(i) and Theorem 4, and Sacks and Ylvisaker [13]. In particular, ([6], equation (13), [13], Theorem 4) if $m = 2$ and other conditions are satisfied, the right hand inequality in (1.2) becomes an equality.

Following [13], a sequence T_n^* , $n = 1, 2, \dots$ of designs is said to be asymptotically optimal (with derivatives) if

$$\lim_{n \rightarrow \infty} \frac{\|f - P_{m, T_n^*} f\|_Q}{\inf_{T_n} \|f - P_{m, T_n} f\|_Q} = 1.$$

In [16], asymptotically optimal designs (with derivatives) are found for the case where X is a stochastic process formally satisfying the stochastic differential equation

$$(1.3a) \quad (L_m X)(t) = dW(t), \quad t \in [0, 1]$$

$$(1.3b) \quad X^{(\nu)}(0) = \xi_\nu, \quad \nu = 0, 1, \dots, m - 1,$$

where L_m is defined by

$$(L_m f)(t) = \sum_{j=0}^m a_{m-j} f^{(j)}(t),$$

$\{W(t), t \in [0, 1]\}$ is a Wiener process and $\{\xi_\nu\}_{\nu=0}^{m-1}$ are m zero mean Gaussian random variables independent of $W(t)$, $t \in [0, 1]$. L_m (in [16]) is such that its null space is spanned by $\{\phi_\nu\}_{\nu=1}^m$ where $\{\phi_\nu\}_{\nu=1}^m$ is an extended, complete Tchebychev (ECT) system of continuity class C^{2m} . In [3], the conditions on L_m are relaxed to: $a_0 \neq 0$, $a_{m-j} \in C^j$, with $E\xi_\nu^2 = 0$. It is the purpose of this note to show that the results of [3] and [16] may be extended to “nearly all” stochastic processes equivalent to X of (1.3).

A sequence of designs may be conveniently described by a continuous positive density h on $T = [0, 1]$. Let $T_n = T_n(h) = \{t_{0n}, t_{1n}, \dots, t_{nn}\}$ be defined by

$$\int_{0^{i/n}}^{i/n} h(x) dx = \frac{i}{n}, \quad i = 0, 1, \dots, n.$$

(For ease of notation we are now letting T_n contain $n + 1$ points.)

We have the following result from [16] (as a consequence of Lemma 3).

PROPOSITION 1. *Let X be as in (1.3), and suppose*

$$f(t) = \int_0^1 Q(t, s)\rho(s) ds$$

where $\rho > 0$, and ρ possess a bounded first derivative. Let $T_n = T_n(h)$. Then

$$(1.4) \quad \|f - P_{m, T_n} f\|_Q^2 = \frac{1}{n^{2m}} \frac{(m!)^2}{(2m)!(2m + 1)!} \int_0^1 \frac{\rho^2(s)\alpha(s)}{h^{2m}(s)} ds + o\left(\frac{1}{n^{2m}}\right)$$

where

$$\alpha(s) = \frac{1}{a_0^2(s)}.$$

Following [11], asymptotically optimal sequences of designs are found from (1.4) by using a Hölder inequality and the fact that $\int_0^1 h(s) ds = 1$ to show that

$$\int_0^1 \frac{\rho^2(s)\alpha(s)}{h^{2m}(s)} ds \geq [\int_0^1 [\rho^2(s)\alpha(s)]^{1/2m} ds]^{2m+1}$$

with equality iff

$$h(s) = \frac{[\rho^2(s)\alpha(s)]^{1/(2m+1)}}{\int_0^1 [\rho^2(u)\alpha(u)]^{1/(2m+1)} du}.$$

Thus if

$$\int_{0^{i/n}}^{i/n} [\rho^2(s)\alpha(s)]^{1/(2m+1)} ds = \frac{i}{n} \int_0^1 [\rho^2(s)\alpha(s)]^{1/(2m+1)} ds, \quad i = 0, 1, \dots, n$$

then $T_n^* = \{t_{0n}^*, t_{1n}^*, \dots, t_{nn}^*\}$, $n = 1, 2, \dots$, is an asymptotically optimal sequence of designs with

$$\|f - P_{m, T_n^*} f\|_Q^2 = \frac{1}{n^{2m}} \frac{(m!)^2}{(2m)!(2m + 1)!} [\int_0^1 [\rho^2(s)\alpha(s)]^{1/(2m+1)} ds]^{2m+1} + o\left(\frac{1}{n^{2m}}\right).$$

The “parameter function” $\alpha(s)$, $s \in [0, 1]$, $\alpha(s) = 1/a_0^2(s)$, plays a central role in the solution. It is not hard to convince one’s self (see e.g. Hájek [2]) that two stochastic processes of the form (1.3) considered in [16] are equivalent iff their “initial value” rv’s (1.3 b) are equivalent and the leading coefficient $a_0(s)$ of the defining differential operator is the same for both processes.

Thus, a maximal generalization of Theorem 1 would appear to be to X ’s equivalent to those of the form (1.3). In the remainder of this note we show that this is, in fact, the case, modulo a regularity condition on Q which we cannot seem to get rid of.

2. Equivalence classes of kernels for X of (1.3). Let $\{X_i(t), t \in [0, 1]\}$, $i = 0, 1$ be two zero mean Gaussian stochastic processes with continuous covariances

$Q_0(s, t)$ and $Q_1(s, t)$ respectively. Now, let Q_0 and Q_1 also denote the Hilbert-Schmidt operators on $\mathcal{L}_2[0, 1]$, with Hilbert-Schmidt kernels $Q_0(s, t)$ and $Q_1(s, t)$, defined by

$$(Q_i p)(t) = \int_0^1 Q_i(t, s)p(s) ds, \quad p \in \mathcal{L}_2[0, 1], i = 0, 1.$$

A version of the Hájek-Feldman Theorem stated in Root [10] says that the measures corresponding to X_1 and X_2 are equivalent iff

$$(2.1) \quad Q_0^{-\frac{1}{2}}Q_1Q_0^{-\frac{1}{2}} = I - B$$

where $Q_i^{-\frac{1}{2}}$ is the symmetric square root of Q_i^{-1} , $i = 0, 1$, and B is a Hilbert-Schmidt operator with $I - B$ invertible. For simplicity we will say that Q_0 and Q_1 are equivalent if (2.1) holds.

Let

$$(2.2) \quad G_0(s, u)_+^{m-1} = \frac{(s-u)_+^{m-1}}{(m-1)!} c(u)$$

where

$$c(u) = 1/a_0(u)$$

and $(x)_+ = x$, $x \geq 0$, $(x)_+ = 0$ otherwise. Let

$$(2.3) \quad Q_0(s, t) = \int_0^1 G_0(s-u)G_0(t-u) du.$$

Q_0 is the covariance of X of (1.3) with $L_m = a_0 D^m$ and $E\xi_\nu^2 = 0$, $\nu = 0, 1, \dots, m-1$. The Hilbert-Schmidt operator Q_0 may be written

$$Q_0 = G_0 G_0^*$$

where G_0^* is the adjoint operator to G_0 , the Hilbert-Schmidt operator with kernel (2.2). Since

$$Q_0 = Q_0^{\frac{1}{2}}Q_0^{\frac{1}{2}} = G_0 G_0^*,$$

$Q_0^{-\frac{1}{2}}Q_1Q_0^{-\frac{1}{2}}$ is unitarily equivalent to $G_0^{-1}Q_1G_0^{*-1}$ and

$$Q_0^{-\frac{1}{2}}Q_1Q_0^{-\frac{1}{2}} = I - B$$

with B Hilbert-Schmidt and $I - B$ invertible iff

$$G_0^{-1}Q_1G_0^{*-1} = I + A$$

for A some Hilbert-Schmidt operator with $I + A$ invertible. Thus Q_0 and Q_1 are equivalent if and only if

$$Q_1 = G_0(I + A)G_0^*$$

where A is Hilbert-Schmidt and $I + A$ invertible.

We summarize these remarks as

PROPOSITION 2. *A kernel Q_1 is equivalent to Q_0 of (2.3) iff*

$$(2.4) \quad Q_1(s, t) = \int_0^1 \frac{(s-u)_+^{m-1}}{(m-1)!} \frac{(t-u)_+^{m-1}}{(m-1)!} c^2(u) du + \int_0^1 \int_0^1 \frac{(s-u)_+^{m-1}}{(m-1)!} \frac{(t-v)_+^{m-1}}{(m-1)!} c(u)A(u, v)c(v) du dv$$

where

$$\int_0^1 \int_0^1 A^2(s, t) ds dt < \infty$$

and $I + A$ is invertible, A being the operator with (symmetric) Hilbert–Schmidt kernel $A(s, t)$.

Now let \tilde{Q}_0 be

$$(2.5) \quad \tilde{Q}_0(s, t) = \sum_{j=1}^m \phi_j(s)\phi_j(t) + \int_0^1 G_0(s, u)G_0(t, u) du$$

with G_0 given by (2.2) and $\phi_j(s) = s^{j-1}/(j-1)!, j = 1, \dots, m$. A process $\{X_0(t), t \in [0, 1]\}$ with covariance (2.5) has a representation

$$X_0(t) = \sum_{\nu=0}^{m-1} X_0^{(\nu)}(0)\phi_\nu(t) + (X_0(t) - P_{m,0}X_0(t)), \quad t \in [0, 1],$$

where $P_{m,0}X_0(t) = E\{X_0(t) | X^{(\nu)}(0), \nu = 0, 1, \dots, m-1\}$ and $\{X^{(\nu)}(0)\}_{\nu=0}^{m-1}$ are i.i.d. $\mathcal{N}(0, 1)$. The process $(X_0(t) - P_{m,0}X_0(t))$ has covariance Q_0 of (2.4). For \tilde{Q}_0 to be equivalent to \tilde{Q}_0 it is necessary and sufficient that $\{X_1^{(\nu)}(0)\}_{\nu=0}^{m-1}$ exist in q.m. and have a covariance matrix of full rank, and that the process $X_1(t) - P_{m,0}X_1(t)$ have a covariance Q_1 of the form (2.4). In this case $X_1(t)$ has a representation of the form

$$X_1(t) = \sum_{\nu=0}^{m-1} X_1^{(\nu)}(0)\phi_\nu(t) + (X_1(t) - P_{m,0}X_1(t))$$

where

$$\phi_\nu(t) = \sum_{j=0}^{m-1} \sigma^{\nu j} \eta_j(t)$$

with

$$\{\sigma^{\nu j}\} = \{\sigma_{\nu j}\}^{-1}, \quad \sigma_{\nu j} = EX_1^{(\nu)}(0)X_1^{(j)}(0), \quad \nu, j = 0, 1, \dots, m-1$$

and

$$\eta_\nu(t) = EX(t)X^{(\nu)}(0) = \frac{\partial^\nu}{\partial s^\nu} \tilde{Q}_1(t, s) \Big|_{s=0}, \quad \nu = 0, 1, \dots, m-1.$$

By the properties of RKHS (see [8]), the $\{\eta_\nu\}$ must all be in $\mathcal{H}_{\tilde{Q}_1}$, and if \tilde{Q}_1 is equivalent to \tilde{Q}_0 , they must also be in $\mathcal{H}_{\tilde{Q}_0}$.

We summarize these remarks in the following

PROPOSITION 3. \tilde{Q}_1 is equivalent to \tilde{Q}_0 of (2.5) iff

$$\tilde{Q}_1(s, t) = \sum_{j=0}^{m-1} \tilde{\psi}_j(s)\tilde{\psi}_j(t) + Q_1(s, t),$$

where $\tilde{\psi}_j^{(\nu)}$ abs. cont., $\nu = 0, 1, \dots, m-1$, $\tilde{\psi}_j^{(m)} \in \mathcal{L}_2$, the $m \times m$ matrix with i th entry $\sigma_{\nu j}$,

$$\sigma_{\nu j} = \frac{\partial^{\nu+j}}{\partial s^\nu \partial t^j} \left(\sum_{i=0}^{m-1} \tilde{\psi}_i(s)\tilde{\psi}_i(t) \right) \Big|_{s=t=0}, \quad \nu, j = 0, 1, \dots, m-1$$

is of full rank, and $Q_1(s, t)$ is of the form (2.4).

Proposition 3 is a slight generalization of [15], Theorem 8; see also [4].

We have that $\mathcal{H}_{\tilde{Q}_1} = \mathcal{H}_{Q_1} \oplus \text{span}\{\tilde{\psi}_i\}_{i=1}^m$, and if T_n includes the point $t = 0$, then $\|f - P_{m, T_n} f\|_{\tilde{Q}_1}^2 = \|P_{Q_1}(f - P_{m, T_n} f)\|_{Q_1}^2$, where P_{Q_1} is the projection operator

in $\mathcal{H}_{\tilde{Q}_1}$ onto the subspace \mathcal{H}_{Q_1} . Thus we may without loss of generality consider Q_1 of the form (2.4). This remark holds, of course, whatever the rank of the matrix $\{\sigma_{\nu j}\}$.

3. Asymptotically optimal designs for \tilde{Q}_1 . The purpose of this section is to prove the following

THEOREM. *Let \tilde{Q}_1 have a representation*

$$\begin{aligned} \tilde{Q}_1(s, t) = & \sum_{i=0}^{m-1} \tilde{\phi}_i(s)\tilde{\phi}_i(t) \\ & + \int_0^1 \frac{(s-u)_+^{m-1}}{(m-1)!} \frac{(t-u)_+^{m-1}}{(m-1)!} c^2(u) du \\ & + \int_0^1 \int_0^1 \frac{(s-u)_+^{m-1}}{(m-1)!} \frac{(t-u)_+^{m-1}}{(m-1)!} c(u)A(u, v)c(v) du dv \end{aligned}$$

$$s, t \in [0, 1]$$

where

- (i) $\tilde{\phi}_i^{(\nu)}$ abs. cont., $\nu = 0, 1, \dots, m-1$, $\tilde{\phi}_i^{(m)} \in \mathcal{L}_2[0, 1]$,
- (ii) $c > 0$, c' bounded,
- (iii) $\int_0^1 \int_0^1 A^2(u, v) du dv < \infty$,
- (iv) the function γ_t given by

$$\gamma_t(s) = \int_0^s \frac{\partial}{\partial t} \frac{1}{c(t)} A(t, \eta)c(\eta) d\eta$$

is well defined and is in the RKHS \mathcal{H}_{K_1} with RK K_1 given by

$$K_1(s, t) = \int_0^{\min(s,t)} c^2(u) du + \int_0^s \int_0^t c(u)A(u, v)c(v) du dv$$

and

$$\|\gamma_t\|_{K_1} \leq M_1 < \infty$$

where $\|\cdot\|_{K_1}$ is the norm in \mathcal{H}_{K_1} .

Let

$$f(t) = \int_0^1 \tilde{Q}_1(t, s)\rho(s) ds$$

with $\rho > 0$, ρ' bounded, and let $T_n = \{t_{in}\}_{i=0}^n$ with

$$\int_0^{t_{in}} h(u) du = \frac{i}{n}, \quad i = 0, 1, \dots, n$$

where

$$\int_0^1 h(u) du = 1, \quad h > 0, h \text{ continuous.}$$

Then where $\alpha = c^2$

$$\|f - P_{m, T_n} f\|_{Q_1}^2 = \frac{1}{n^{2m}} \frac{(m!)^2}{(2m)!(2m+1)!} \int_0^1 \frac{\rho^2(s)\alpha(s)}{h^{2m}(s)} ds + o\left(\frac{1}{n^{2m}}\right).$$

REMARK. The hypotheses of the Theorem do not include $I + A$ invertible. On the other hand, if $I + A$ is invertible then condition (iv) is equivalent to $\gamma_t \in \mathcal{H}_{K_0}$, the RKHS with RK

$$K_0(s, t) = \int_0^{\min(s,t)} c^2(u) du,$$

where $\mathcal{H}_{K_0} = \{f : f(0) = 0, f \text{ abs. cont.}, f'/c \in \mathcal{L}_2\}$. Thus if $I + A$ is invertible and (ii) holds then (iv) is equivalent to

$$\int_0^1 \left(\frac{\partial}{\partial t} \frac{1}{c(t)} A(t, \eta) \right)^2 d\eta < \tilde{M} < \infty .$$

Condition (iv) is similar to a condition used in [11]. This condition is used in the proof of Lemma 1 to follow, and we see no way to eliminate it there.

PROOF. The proof below follows closely along the lines of the proof of Theorem 1 of [13], generalized with the aid of [16].

The proof begins with Lemma 1.

LEMMA 1. *Let*

$$\begin{aligned} K_0(s, t) &= \int_0^{\min(s,t)} c^2(u) du \\ K_1(s, t) &= \int_0^{\min(s,t)} c^2(u) du + \int_0^s \int_0^t c(u)A(u, v)c(v) du dv \\ f_0(t) &= \int_0^1 K_0(t, u)\rho(u) du \\ f_1(t) &= \int_0^1 K_1(t, u)\rho(u) du \end{aligned}$$

where $\int_0^1 \int_0^1 A^2(u, v) du dv < \infty$, where $c, \rho > 0$, continuous, c', ρ' bounded. Let $\mathcal{H}_{K_i}, i = 0, 1$, be the RKHS's with reproducing kernels $K_i, i = 0, 1$, and inner products $\langle \cdot, \cdot \rangle_{K_0}$ and $\langle \cdot, \cdot \rangle_{K_1}$ respectively. Suppose further that, for each t , the function γ_t defined by

$$\gamma_t(s) = \int_0^s \frac{\partial}{\partial t} \frac{1}{c(t)} A(t, \eta)c(\eta) d\eta$$

satisfies

$$(3.1) \quad \gamma_t \in \mathcal{H}_{K_1}, \quad \|\gamma_t\|_{K_1} \leq M_1 < \infty, \quad t \in [0, 1].$$

Then, there exists an ϵ independent of ρ such that, for sufficiently large n ,

$$(3.2) \quad 1 - \epsilon\Delta \leq \frac{\|f_1 - P_{T_n} f_1\|_{K_1}^2}{\|f_0 - P_{T_n} f_0\|_{K_0}^2} \leq 1 + \epsilon\Delta$$

where

$$\Delta = \max_i |t_{i+1} - t_i| .$$

Here, for $i = 0, 1, P_{T_n} f_i$ is the projection of f_i in \mathcal{H}_{K_i} onto the subspace of \mathcal{H}_{K_i} spanned by $\{K_{it}, t \in T_n\}$, where $K_{it}(t') = K_i(t, t')$.

PROOF. For $i = 0, 1$,

$$\begin{aligned} \langle f_i - P_{T_n} f_i, f_i - P_{T_n} f_i \rangle_{K_i} &= \langle f_i, f_i - P_{T_n} f_i \rangle_{K_i} \\ &= \int_0^1 \rho(u)(f_i(u) - P_{T_n} f_i(u)) du . \end{aligned}$$

Then

$$\begin{aligned} \|f_0 - P_{T_n} f_0\|_{K_0}^2 &= \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u)(f_0(u) - P_{T_n} f_0(u)) du \\ &= \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v)\rho(v) dv \end{aligned}$$

where $t_0 \equiv 0$, $t_n \equiv 1$ and, according to [16] $B_i(u, v)$ is, for $u, v \in [t_i, t_{i+1}]$, the Green's function for the differential operator $L_m^* L_m = g$ with boundary conditions $f(t_i) = f(t_{i+1}) = 0$,

$$(L_m^* L_m f)(t) = \frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} f(t).$$

Similarly,

$$(3.3) \quad \|f_1 - P_{T_n} f_1\|_{K_1}^2 = \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u) (f_1(u) - P_{T_n} f_1(u)) du.$$

Since $f_1(u) - P_{T_n} f_1(u) = 0$ for $t = t_0, t_1, \dots, t_n$, and $f_1 - P_{T_n} f_1 \in \mathcal{L}_2[0, 1]$, we may write

$$(3.4) \quad f_1(u) - P_{T_n} f_1(u) = \int_{t_i}^{t_{i+1}} B_i(u, v) \frac{d}{dv} \frac{1}{c^2(v)} \frac{d}{dv} (f_1(v) - P_{T_n} f_1(v)) dv, \quad u \in [t_i, t_{i+1}]$$

where B_i is as before. But, since

$$f_1(t) = \int_0^1 K_0(t, u) \rho(u) du + \int_0^1 \rho(u) du \int_0^u \int_0^t c(\xi) A(\xi, \eta) c(\eta) d\xi d\eta,$$

then

$$(3.5) \quad \begin{aligned} \frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} f_1(t) &= \rho(t) + \int_0^1 \rho(u) du \int_0^u \frac{\partial}{\partial t} \frac{1}{c(t)} A(t, \eta) c(\eta) d\eta \\ &= \rho(t) + \int_0^1 \rho(u) \gamma_i(u) du. \end{aligned}$$

By our assumption, $\gamma_i \in \mathcal{H}_{K_1}$, so that (3.5) becomes

$$(3.6) \quad \frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} f_1(t) = \rho(t) + \int_0^1 \rho(u) \gamma_i(u) = \rho(t) + \langle \gamma_i, f_1 \rangle_{K_1}.$$

Also

$$(P_{T_n} f_1)(t) = (K_1(t, t_1), K_1(t, t_2), \dots, K_1(t, t_n)) K_{1,n}^{-1} (f_1(t_1), f_1(t_2), \dots, f_1(t_n))'$$

where $K_{1,n}$ is the $n \times n$ matrix with ij th entry $K_1(t_i, t_j)$. Now, for $t \neq t_i$,

$$\frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} K_1(t, t_i) = \frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} [\int_0^t [\int_0^i c(\xi) A(\xi, n) c(\eta) d\xi d\eta] = \gamma_i(t_i)$$

so that, for each fixed $t \notin T_n$,

$$(3.7) \quad \begin{aligned} \frac{d}{dt} \frac{1}{c^2(t)} \frac{d}{dt} (P_{T_n} f_1(t)) &= (\gamma_i(t_1), \gamma_i(t_2), \dots, \gamma_i(t_n)) K_{1,n}^{-1} (f_1(t_1), f_1(t_2), \dots, f_1(t_n))' \\ &= \langle \gamma_i, P_{T_n} f_1 \rangle_{K_1}. \end{aligned}$$

Thus, by (3.3), (3.4), (3.6), (3.7),

$$\begin{aligned} \|f_1 - P_{T_n} f_1\|_{K_1}^2 &= \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v) \rho(v) dv \\ &\quad + \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v) \langle \gamma_v, f_1 - P_{T_n} f_1 \rangle_{K_1} dv. \end{aligned}$$

Now ρ and B_i are nonnegative, so we may write

$$\begin{aligned} & \left| \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v) \langle \gamma_v, f_1 - P_{T_n} f_1 \rangle_{K_1} dv \right| \\ & \leq \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v) dv \times M_1 \|f_1 - P_{T_n} f_1\|_{K_1}, \end{aligned}$$

where M_1 is defined in (3.1).

Now, letting

$$\hat{\xi}_0(t) = \int_0^1 K_0(t, u) du$$

it may be shown that²

$$(3.8) \quad \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \rho(u) du \int_{t_i}^{t_{i+1}} B_i(u, v) dv = \langle f_0 - P_{T_n} f_0, \hat{\xi}_0 - P_{T_n} \hat{\xi}_0 \rangle_{K_0}.$$

By (1.4),

$$\|\hat{\xi}_0 - P_{T_n} \hat{\xi}_0\|_{K_0} = M_2 \left(\frac{1}{n} \left(1 + o\left(\frac{1}{n}\right) \right) \right)$$

for appropriately chosen M_2 .

Thus

$$\|f_1 - P_{T_n} f_1\|_{K_1}^2 = \|f_0 - P_{T_n} f_0\|_{K_0}^2 + \theta \frac{M_3}{n} \|f_1 - P_{T_n} f_1\|_{K_1} \|f_0 - P_{T_n} f_0\|_{K_0}$$

for some θ with $|\theta| < 1$ and $M_3 = M_1 M_2$, and so

$$\frac{\|f_1 - P_{T_n} f_1\|_{K_1}^2}{\|f_0 - P_{T_n} f_0\|_{K_0}^2} = 1 + \theta \frac{M_3}{n} \left(1 + o\left(\frac{1}{n}\right) \right).$$

Since $1/n \leq \Delta$, the Lemma is proved.

LEMMA 2. For $m \geq 2$ let

$$Q_i(s, t) = \int_0^s \int_0^t \frac{(s-u)_+^{m-2}}{(m-2)!} \frac{(t-u)_+^{m-2}}{(m-2)!} K_i(u, v) du dv \quad i = 1, 2$$

where $K_i, i = 0, 1$ are as Lemma 1. Let

$$f_i(t) = \int_0^1 Q_i(t, u) \rho(u) du, \quad i = 0, 1.$$

Then, there exists an ε independent of ρ such that, for sufficiently large n ,

$$(3.9) \quad 1 - \varepsilon \Delta \leq \frac{\|f_1 - P_{m, T_n} f_1\|_{Q_1}^2}{\|f_0 - P_{m, T_n} f_0\|_{Q_0}^2} \leq 1 + \varepsilon \Delta.$$

Here $P_{m, T_n} f_i$ is the projection of f_i in $\mathcal{H}_{Q_i}, i = 0, 1$ onto the subspace of (1.1) with $Q = Q_i$.

The proof of this Lemma is contained within the proof of Theorem 1 of [13], page 2065 Equations (2.28)–(2.31), where it is shown that (3.2) implies (3.9).

The Theorem now follows by using the proof of Lemma 3 of [16] (where only condition ii on c is needed for $Q = Q_0$) to show that

$$\|f_0 - P_{m, T_n} f_0\|_{Q_0}^2 = \frac{1}{n^{2m}} \frac{(m!)^2}{(2m)!(2m+1)!} \int_0^1 \frac{\rho^2(s) \alpha(s)}{h^{2m}(s)} ds + o\left(\frac{1}{n^{2m}}\right).$$

² Equation (3.8) may be checked by following the argument of Lemma 1 of [16]; see equations (3.4), (3.5) and (3.22). Equation (3.4a) there should read $f(t) = EX(t) \int_0^1 X(u) \rho(u) du$.

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