NUMERICAL APPROXIMATION FOR THE INFINITE-DIMENSIONAL
DISCRETE-TIME OPTIMAL LINEAR-QUADRATIC REGULATOR PROBLEM

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NUMERICAL APPROXIMATION FOR THE INFINITE-DIMENSIONAL DISCRETE-TIME
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ABSTRACT

An abstract approximation framework is developed for the finite and
infinite time horizon discrete-time linear-quadratic regulator problem for
systems whose state dynamics are described by a linear semigroup of operators
on an infinite dimensional Hilbert space. The schemes included in the
framework yield finite dimensional approximations to the linear state feedback
gains which determine the optimal control law. Convergence arguments are
given. Examples involving hereditary and parabolic systems and the vibration
of a flexible beam are considered. Spline-based finite element schemes for
these classes of problems, together with numerical results, are presented and
discussed.

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1. Introduction

Recent advances in micro-processor technology have led to increased interest in digital or discrete-time control systems. In addition, because many current application areas involve complex systems which are most appropriately modelled using functional and/or partial differential equations, it has become important to study digital control techniques in the context of infinite dimensional or distributed systems.

A great deal of attention has been given to the continuous-time infinite-dimensional linear-quadratic regulator problem. The general theory and characterization of the linear state feedback form of the optimal control are discussed in [5], [6], [8], [9], [21] and [22], while its application to hereditary, parabolic and hyperbolic systems with emphasis on approximation is treated in [2], [3], [7], [10], [11], [14] and [17] to mention just some of the work that has been done.

On the other hand, relatively little can be found in the literature concerning the corresponding discrete-time problem. The major contributions in this area can be found in the papers by Lee, Chow and Barr [20] and Zabczyk [28]. In these studies the Riccati difference equations that characterize the linear feedback form of the optimal control for the finite time problem are given and limiting properties as the length of the time horizon tends to infinity are discussed. However, the issue of approximation is not considered.

In the present paper, we develop numerical approximation schemes that yield finite dimensional approximations to the feedback gain operators which determine the discrete-time optimal control law. We consider control systems whose dynamics can be described in terms of a linear semigroup of operators on an infinite dimensional Hilbert space. The basis of our approach is the construction of a sequence of finite dimensional (presumably finite element
based) state approximations which in turn leads to a sequence of finite dimensional discrete-time linear-quadratic regulator problems each of which can be solved using standard techniques.

Under appropriate assumptions on the nature of the original problem and the convergence of the state approximation, we are able to prove that the approximating optimal controls and feedback gains converge to the true optimal control sequences and feedback laws for the original infinite dimensional system. Depending upon the convergence properties of the state approximation, we are able to establish strong or uniform norm convergence of the approximating gain operators and the corresponding weak or strong convergence of the approximating feedback kernels which are used in the implementation of the optimal control. We treat both the finite and infinite-time horizon problems.

We have tested our schemes on a wide variety of examples. This paper includes numerical results for problems with state dynamics given by hereditary and parabolic (heat/diffusion) differential equations and a hybrid system of partial and ordinary differential equations for the vibration of an Euler-Bernoulli beam connected to a rigid body and a lumped mass. We implemented and tested the methods on an IBM Personal Computer.

We give a brief outline of the remainder of the paper. In section 2 we briefly outline previous results concerning the characterization of the optimal control and feedback gains for both the finite and infinite time horizon discrete-time regulator problem for distributed systems. The Riccati difference and algebraic equations whose solutions determine the optimal feedback control law are discussed. In section 3 we develop the abstract approximation framework and convergence arguments. Section 4 contains a discussion of particular schemes for the classes of problems mentioned above.
together with the results of our numerical studies. Some concluding remarks are given in Section 5.

We employ standard notation throughout. For an interval \((a, b)\), we denote by \(H^k(a, b)\) the usual Sobolev spaces of real-valued functions defined on \((a, b)\) whose \((k-1)\)st derivatives are absolutely continuous and whose \(k\)th derivatives are \(L_2\). The standard Sobolev inner product on \(H^k(a, b)\) is denoted by \(\langle \cdot , \cdot \rangle_k\). For \(X\) and \(Y\) normed linear spaces we denote by \(L(X, Y)\) the space of bounded linear operators from \(X\) into \(Y\). When \(Y = X\), we use the shorthand notation \(L(X)\).

2. The Optimal Control Problem

2.1 Optimal Control on a Finite Interval

Let \(Z\) and \(U\) be Hilbert spaces with inner products \(\langle \cdot , \cdot \rangle_Z\) and \(\langle \cdot , \cdot \rangle_U\), respectively, with \(U\) finite dimensional. For \(\{H, \langle \cdot , \cdot \rangle_H\}\) a Hilbert space, let \(L^2(t_0, t_f; H)\) denote the usual Hilbert space of sequences \(x = \{x(t)\}_{t=t_0}^{t_f}\) with \(x(t) \in H\) together with the inner product

\[
\langle x, y \rangle_{L^2} = \sum_{t=t_0}^{t_f} \langle x(t), y(t) \rangle_H.
\]

The discrete-time linear quadratic regulator problem on the finite time interval \([t_0, t_f]\) is

(P1) Choose \(u \in L^2(t_0, t_f; U)\) to minimize the quadratic performance index
\[ J(G; t_0, t_f, z(t_0), u) = \]
\[ \sum_{t=t_0}^{t_f-1} \left[ \langle Qz(t), z(t) \rangle_Z + \langle Ru(t), u(t) \rangle_U \right] + \langle Gz(t_f), z(t_f) \rangle_Z \]

subject to the discrete-time control system

\[ z(t+1) = Tz(t) + Bu(t), \quad t > t_0 \]
\[ z(t_0) \in Z, \]

where \( T \) and \( B \) are bounded linear operators from \( Z \) into \( Z \) and \( U \) into \( Z \), respectively, \( Q \) and \( G \) are bounded, nonnegative self-adjoint operators on \( Z \), and \( R \) is a positive definite self-adjoint operator on \( U \).

Of primary concern to us will be applications where (2.3) is the sampled form of the continuous-time control system

\[ \dot{z}(s) = Az(s) + Bu(s) \]

where \( A \) is the infinitesimal generator of a \( C_0 \)-semigroup of bounded linear operators \( T(s), s \geq 0 \), on \( Z \), and \( B \) is a possibly unbounded linear operator from \( U \) into \( Z \). In this case we have

\[ T = T(\tau) \quad \text{and} \quad B = \int_0^T T(s)B \, ds, \]

where \( \tau \) is the sampling interval. If, as in our subsequent example discussed in Section 4.1 where \( u \) is a boundary control in a heat equation, \( B \) is unbounded (more precisely, \( B \) maps \( U \) not into \( Z \) but into some larger space), then the integral in (2.5) is not interpreted literally.
The solution to Problem (P1) has been given for infinite dimensional control systems in [20],[28], and the equations representing the solution have the same form as in the finite dimensional case. We will give now the version of the solution that is most useful for our purposes.

For given \( z(t_0) \), \( J(G; t_0, t_f, z(t_0), u) \) is a bounded linear-quadratic functional on \( L^2(t_0, t_f; U) \) with coercive quadratic part. Therefore, for each \( z(t_0) \), there exists a unique optimal control sequence in \( L^2(t_0, t_f; U) \). Also, the minimum value of the performance index is a quadratic functional of \( z(t_0) \), so that there exists a unique nonnegative, self-adjoint \( \Pi(t_0) \in L(Z) \) such that

\[
(2.6) \quad J_* = \min \ J(G; t_0, t_f, z(t_0), u) = \langle \Pi(t_0)z(t_0), z(t_0) \rangle_z.
\]

Application of the principle of dynamic optimality establishes that the optimal control has the feedback form

\[
(2.7) \quad u_*(t) = -F(t)z_*(t), \quad t_0 < t < t_f - 1
\]

where

\[
(2.8) \quad F(t) = R(t)^{-1}B^*\Pi(t+1)T,
\]

\[
(2.9) \quad R(t) = R + B^*\Pi(t+1)B
\]

and \( \Pi(t) \) satisfies the Riccati difference equation

\[
(2.10) \quad \Pi(t) = T [\Pi(t+1) - \Pi(t+1)BR(t)^{-1}B^*\Pi(t+1)]T + Q, \quad t < t_f - 1,
\]
with the final condition

\[ \Pi(t_f) = G. \]

The optimal trajectory \( z_\ast \) is given by

\[ z_\ast(t+1) = S(t)z_\ast(t), \quad t > t_0 \]

where

\[ S(t) = T-BF(t). \]

2.2 Control on the Infinite Interval

Here, \( t_f = \infty \) and \( G = 0 \). To simplify notation, we will write \( J(t_0, z(t_0), u) \) instead of \( J(0; t_0, z(t_0), u) \).

**Definition 2.1.** A control sequence \( u \in L^2(0, \infty; U) \) is an admissible control for the initial condition \( z \) if \( J(0, z, u) < \infty \).

The discrete-time linear-quadratic regulator problem on the infinite interval is

\[ (P2) \quad \text{Choose an admissible control } u_\ast \text{ to minimize } J(0, z, u), \text{ if an admissible control exists for the initial condition } z. \]

That a unique optimal control \( u_\ast \) exists whenever at least one admissible control exists follows from the fact that the quadratic part of \( J(0, z, u) \) is coercive on a subspace of \( L^2(0, \infty; U) \). See the discussion following Definition 4.1 of [9].
Definition 2.2. A bounded linear operator $\Pi$ on $Z$ is a solution to the Riccati algebraic equation if

\[(2.14) \quad \Pi = T^* [\Pi - \Pi B (R + B^* \Pi B)^{-1} B^* \Pi] T + Q.\]

The following theorem summarizes results from Zabczyk [28].

Theorem 2.3. The following are equivalent:

(i) There exists an admissible control for each $z \in Z$;

(ii) for each $z \in Z$, $\sup_{t\leq t_f} \langle \Pi(t)z, z \rangle_Z < \infty$, where $\Pi(t)$ is the Riccati operator in (2.10) and $\Pi(t_f) = 0$ for fixed $t_f$;

(iii) as $t \to \infty$, $\Pi(t)$ converges strongly to a nonnegative self-adjoint solution to the Riccati algebraic equation;

(iv) there exists a nonnegative self-adjoint solution to the Riccati algebraic equation.

For uniqueness of the solution to the Riccati algebraic equation and characterization of the optimal control, Zabczyk treated two cases: when $Q$ is coercive, and when the spectral radius of $T$ is less than 1 (i.e., the open-loop system is uniformly exponentially stable). Since neither is the case in the example we discuss in Section 4.2 and other applications in which we are interested, we will need the following hypothesis and theorem.

Hypothesis 2.4. The operators $T$, $B$ and $Q$ are such that, if $z(0) \in Z$ and $u$ is an admissible control for $z(0)$, then

\[(2.15) \quad \lim_{t \to \infty} |z(t)|_Z = 0.\]
Theorem 2.5. When Hypothesis 2.4 holds, there exists at most one nonnegative self-adjoint solution to the Riccati algebraic equation. If such a solution exists, then there exists a unique solution to problem (P2) for each initial condition $z(0) \in Z$, the minimum value of the performance index is

\[ J_* = \min_{u \text{ admissible}} J(0,\infty,z(0),u) = \langle \Pi z(0), z(0) \rangle_Z, \]

the optimal control has the feedback form

\[ u_*(t) = -Fz_*(t), \quad t > 0, \]

where

\[ F = R^{-1}B \Pi T, \]

\[ \tilde{R} = R + B \Pi B \]

and the optimal trajectory $z_*(t)$ satisfies

\[ z_*(t+1) = S z_*(t), \quad t > 0, \]

with

\[ S = T - BF. \]

Proof. Let $\Pi$ be such a solution and note that, for any finite $t_f$, $\Pi$ is a constant solution to (2.10) and (2.11) with $G = \Pi$. Then the corresponding
F(t) and \( \tilde{R}(t) \) defined by (2.8) and (2.9) are the constant operators in (2.18) and (2.19). For \( z(0) \in \mathcal{Z} \), define \( \overline{z}(0) = z(0) \),

\[
(2.22) \quad \overline{z}(t+1) = (T-BF)\overline{z}(t), \quad t > 0,
\]

and

\[
(2.23) \quad \overline{u}(t) = -Fz(t), \quad t > 0.
\]

Now suppose that \( u \) is an admissible control for \( z(0) \) and that \( z(t) \) is the corresponding solution to (2.3). For \( t_f > 0 \), the preceding results about the solution to Problem (P1) with \( G = \Pi \) imply

\[
J(\Pi; 0, t_f, z(0), u) < J(0; 0, t_f, z(0), u) + \langle \Pi z(t_f), z(t_f) \rangle_Z
\]

\[
= J(0; 0, \infty, z(0), u) + \langle \Pi z(t_f), z(t_f) \rangle_Z.
\]

Also,

\[
J(\Pi; 0, t_f, z(0), \overline{u}) = \langle \Pi z(0), z(0) \rangle_Z
\]

\[
= J(0; 0, t_f, z(0), \overline{u}) + \langle \Pi z(t_f), z(t_f) \rangle_Z.
\]

Since \( z(t_f) \to 0 \) as \( t_f \to \infty \), (2.24) shows that \( \overline{u} \) is both admissible and optimal for Problem (P2). Since \( \overline{z}(t_f) \to 0 \) as \( t_f \to \infty \), (2.25) shows (2.16).

As we see now, (2.16) must hold for any nonnegative self-adjoint solution of the Riccati algebraic equation; therefore, such a solution is unique.

**Remark 2.6.** When Hypothesis 2.4 does not hold, the Riccati algebraic equation may have more than one nonnegative self-adjoint solution. In this case, the minimal such solution -- there will be one -- gives the solution to Problem
(P2) as in Theorem 2.5. Throughout this paper, we assume that Hypothesis 2.4 holds.

Lemma 2.7. Suppose that $Q > m$ for some positive constant $m$, and set

$$C_n = \sum_{t=0}^{n} (T^*)^t Q T^t, \text{ for } n = 1, 2, \ldots.$$ 

Then $|C_n z|_Z$ is bounded in $n$ for each $z \in Z$ if and only if $C_n$ converges in norm to the operator

$$(2.26) \quad C = \sum_{t=0}^{\infty} (T^*)^t Q T^t$$

and

$$(2.27) \quad |T^t| < (|C|/m)(1 - m/|C|)^t, \quad t = 1, 2, \ldots.$$ 

Proof. Since $C_n$ is an increasing sequence of bounded self-adjoint linear operators, $C_n$ converges strongly to some bounded self-adjoint $C$ if and only if $\langle C_n z, z \rangle_Z$ is bounded in $n$ for each $z$, if and only if $|C_n z|_Z$ is bounded in $n$ for each $z$. This is a standard result. The proof of the Lemma is then a standard exercise using the Lyapunov functional $\langle Cz(t), z(t) \rangle_Z$ for the homogeneous part of (2.3).

Corollary 2.8. If $Q > m > 0$ and the Riccati algebraic equation has a nonnegative self-adjoint solution $H$, then the spectral radius of the operator $S$ in (2.21) is less than 1, and

$$(2.28) \quad |S^t| < (|H|/m)(1 - m/|H|)^t, \quad t = 1, 2, \ldots.$$ 

Proof. This follows from Lemma 2.7 and
(2.29)  \[ <Hz,z>_Z = \sum_{t=0}^{\infty} (S^*)^t [Q + F^*RF]S^t. \]

For Q coercive, Zabczyk proved a stronger result than part (iii) of Theorem 2.3: if a nonnegative self-adjoint solution to the Riccati algebraic equation exists, then \( |\Pi(t) - \Pi| \to 0 \) geometrically fast as \( t \to \infty \) (Also, see [13]). We will need such a result, along with an explicit convergence rate, for the approximation theory in Section 3.2. Since Zabczyk's proof does not yield an explicit convergence rate, we give the following.

**Theorem 2.9.** Suppose that there exists a nonnegative self-adjoint solution \( \Pi \) to (2.14) and that

\[
(2.30) \quad |S^t| < Mr^t, \quad t = 1, 2, \ldots ,
\]

where \( M \) and \( r \) are positive constants with \( r < 1 \) and \( S \) is the optimal closed-loop operator in Theorem 2.5. If \( \Pi(\cdot) \) is the operator in (2.10) with \( t_f = 0 \) and

\[
(2.31) \quad \Pi(0) > \Pi,
\]

then

\[
(2.32) \quad <Hz,z>_Z < <\Pi(-t)z,z>_Z < <Hz,z>_Z + (Mr^t)^2|\Pi(0)|, \quad t = 1, 2, \ldots .
\]

**Proof.** For \( t_0 \) a negative integer, let \( u_0 \) be the optimal control sequence for the finite-time Problem \( (P1) \) on the interval \([t_0, 0]\) with initial condition
$z(t_0) \in Z$, with $z_0$ the corresponding optimal trajectory. Also, let $u_*$ be the optimal control sequence on the infinite interval for Problem (P2) with initial condition $z(t_0)$, with $z_*$ the corresponding optimal trajectory.

Since $\Pi$ is a constant solution to (2.10) for the final condition $G = \Pi$, we have

$$\langle \Pi z(t_0), z(t_0) \rangle_Z = J(\Pi; t_0, 0, z(t_0), u_*)$$

(2.33)

$$\langle J(0; t_0, 0, z(t_0), u_0) + \langle \Pi z_0(0), z_0(0) \rangle_Z$$

$$\langle J(0; t_0, 0, z(t_0), u_0) + \langle \Pi(0) z_0(0), z_0(0) \rangle_Z$$

$$= \langle \Pi(t_0) z(t_0), z(t_0) \rangle_Z.$$

On the other hand (note that $z_*(t_0) = S^{-t} z(t_0)$),

$$\langle \Pi(t_0) z(t_0), z(t_0) \rangle_Z$$

(2.34)

$$\langle J(0; 0, -t_0, z(t_0), u_*) + \langle \Pi(0) z_*(-t_0), z_*(-t_0) \rangle_Z$$

$$\langle J(0; 0, \infty, z(t_0), u_*) + \langle \Pi(0) z_*(t_0), z_*(t_0) \rangle_Z$$

$$\langle \langle \Pi z(t_0), z(t_0) \rangle_Z + |\Pi(0)| \langle S^{-t_0} |z(t_0)|_Z \rangle^2.$$
3. Approximation Theory

3.1 The finite time interval problem

In this section we develop a general approximation framework for the finite time interval problem (P1) and describe associated convergence results.

For each $N = 1, 2, \ldots$, let $Z_N \subset Z$ be a finite dimensional subspace of $Z$ and let $P_N : Z \rightarrow Z_N$ denote the orthogonal projection of $Z$ onto $Z_N$ with respect to the $\langle \cdot, \cdot \rangle_Z$ inner product. We require the following hypotheses.

**Hypothesis 3.1** There exist operators $T_N : Z_N + Z_N$, $B_N : U + Z_N$, $Q_N : Z_N + Z_N$ and $G_N : Z_N + Z_N$ which satisfy

\[
T_N P_N + T \quad \text{strongly,}
\]
\[
T_N^* P_N + T^* \quad \text{strongly,}
\]
\[
B_N + B \quad \text{strongly,}
\]
\[
Q_N P_N + Q \quad \text{strongly,}
\]
\[
G_N P_N + G \quad \text{strongly,}
\]

as $N \rightarrow \infty$ with $T_N$ and $B_N$ bounded and $Q_N$ and $G_N$ bounded, self-adjoint and nonnegative.

**Hypothesis 3.2** The spaces $Z_N$ are approximating subspaces in the sense that the projections $P_N$ satisfy $P_N + I$ strongly on $Z$ as $N \rightarrow \infty$.

We note that since $U$ has been assumed to be finite dimensional, Hypothesis 3.1 above necessarily implies that $B_N + B$ and $B_N^* P_N + B^*$ in the uniform norm topology on $L(U,Z)$ and $L(Z,U)$ respectively.
We define a sequence of approximating discrete-time linear quadratic regulator problems on the finite time interval \([t_0, t_f]\) as follows:

\((P_{1N})\)

Find \(u^N_\alpha \in \ell^2(t_0, t_{f-1}; U)\) which minimizes

\[
J_N(G; t_0, t_f, z(t_0), u) = \sum_{t=t_0}^{t_f-1} \langle Q_N z_N(t), z_N(t) \rangle_Z + \langle Ru(t), u(t) \rangle_U + \langle G_N z_N(t_f), z_N(t_f) \rangle_Z
\]

subject to

\[
z_N(t+1) = T_N z_N(t) + B_N u(t), \quad t > t_0
\]

\[
z_N(t_0) = P_N z(t_0).
\]

The results stated in Section 2.1 concerning the existence and uniqueness of solutions to Problem \((P_1)\) apply to the Problems \((P_{1N})\) as well. Indeed, there exists a unique solution \(u^N_\alpha \in \ell^2(t_0, t_{f-1}; U)\) to Problem \((P_{1N})\) which is given in feedback form by

\[
u^N_\alpha(t) = -F_N(t) z^N_\alpha(t), \quad t_0 < t < t_{f-1}
\]

where

\[
F_N(t) = R_N(t)^{-1} B_N^T P_N(t+1) T_N
\]
with

\[ R_N(t) = R + E_N^* \Pi_N(t+1)B_N \]

and the operators \( \{\Pi_N(t)\}_{t=t_0}^{t_f} \) on \( Z_N \) satisfying the Riccati difference equation

\[ \Pi_N(t) = T_N^*[\Pi_N(t+1) - \Pi_N(t+1)B_N^*R_N(t)^{-1}B_N^*\Pi_N(t+1)]T_N + Q_N \]

with terminal condition

\[ \Pi_N(t_f) = G_N. \]

The optimal trajectory \( z_N^* \) is given by

\[ z_N^*(t+1) = S_N(t)z_N^*(t), \quad t > t_0, \]

\[ z_N^*(t_0) = P_Nz(t_0) \]

where

\[ S_N(t) = T_N - B_NF_N(t), \quad t > t_0. \]

The operators \( \{\Pi_N(t)\}_{t=t_0}^{t_f} \) are bounded, self-adjoint and nonnegative. The minimum value of the performance index (3.1) is given by

\[ J_N^* = J_N(G_N;t_0,t_f,z(t_0),u_N^*) = \langle \Pi_N(t_0)z_N^*(t_0), z_N^*(t_0) \rangle_z . \]

The fundamental convergence result is given in the following theorem.
Theorem 3.3 Let $u^N_*$ and $u_*$ be the unique solutions to problems $(P_{1N})$ and $(P_1)$, respectively, with $z^N_*$ and $z_*$ the corresponding optimal trajectories generated by (3.8) and (2.12). Let $J_N$, $\Pi_N$ and $F_N$ and $J$, $\Pi$ and $F$ be given by (3.1), (3.6) and (3.4) and (2.2), (2.10) and (2.8). Then, if Hypotheses 3.1 and 3.2 hold, we have

(i) $\lim_{N\to\infty} |u^N_* - u_*|^2 = 0$,

(ii) $\lim_{N\to\infty} |z^N_* - z_*|^2 = 0$,

(iii) $\lim_{N\to\infty} |J^N_* - J_*| = 0$,

(iv) $\lim_{N\to\infty} |\Pi_N(t)P_Nz - \Pi(t)z| = 0, z \in Z, t_0 < t < t_f$ and

(v) $\lim_{N\to\infty} |F_n(t)P_N - F(t)| = 0, t_0 < t < t_f - 1$.

Proof

We first note that $\Pi_N(t)$ being nonnegative implies that $|\hat{R}_N(t)| > |R|$ and consequently that $|\hat{R}_N(t)^{-1}| < |R|^{-1}$. It follows therefore that for $u \in U$.

\begin{equation}
(3.11) \quad |(\hat{R}_N(t)^{-1} - R(t)^{-1})u|_U
\end{equation}

$$= |\hat{R}_N(t)^{-1}(R(t) - \hat{R}_N(t))R(t)^{-1}u|_U$$
\[ \left| R(t) - R_N(t) \right| \leq |R|^{-1} |(R(t) - R_N(t))R(t)^{-1}u|_U. \]

The above estimate together with (2.9), (2.11), (3.5), (3.7) and Hypothesis 3.1 imply that

\[(3.12) \quad R_N(t_f-1)^{-1} + \hat{R}(t_f-1)^{-1} \]

as \( N \to \infty \) strongly on \( U \). Since \( U \) is finite dimensional the convergence in (3.13) is in fact uniform. It then follows immediately from (2.8), (3.4) and Hypothesis 3.1 that

\[(3.13) \quad F_N(t_f - 1) P_N + F(t_f - 1), \]

uniformly as \( N \to \infty \), and from (2.10) and (3.6) that

\[(3.14) \quad H_N(t_f - 1) P_N + H(t_f - 1) \]

strongly on \( Z \) as \( N \to \infty \). A simple induction yields (iv) and (v) from which (i), (ii) and (iii) then follow trivially.

**Remark** It will, on occasion, be the case that in constructing a particular approximation scheme \( T_N P_N \to T \) strongly but \( T_N^* P_N \to T^* \) only weakly (see, for example, [3]). However, by using the fact that

\[(3.15) \quad (T_N^* \Pi_N(t+1))^* = \Pi_N(t+1)T_N \]
implies that $T_N^*(t+1) + T_N^*(t+1)$ weakly if $\Pi_N(t+1) + \Pi(t+1)$ weakly, we conclude that Theorem 3.3 continues to hold under these somewhat weaker hypotheses with the strong convergence in (iv) replaced by weak and the uniform convergence in (v) replaced by strong.

Under certain additional hypotheses it can be shown that the operators $\Pi(t), t_0 < t < t_f$ given by (2.10), (2.11) are trace class (see [15]) and that

\begin{equation}
\lim_{N \to \infty} \| \Pi_N(t) \Pi_N - \Pi(t) \|_1 = 0, t_0 < t < t_f,
\end{equation}

where $\| \cdot \|_1$ denotes the trace norm, the strongest of all common operator norms. We require the following lemmas.

**Lemma 3.4** If $\{a_i\}_{i=1}^\infty$ is an absolutely summable sequence of real numbers then there exist sequences $\{b_i\}_{i=1}^\infty$ and $\{c_i\}_{i=1}^\infty$ such that $\lim_{i \to \infty} b_i = 0$, $\{c_i\}_{i=1}^\infty$ is absolutely summable and $a_i = b_i c_i$.

**Proof**

Let

\begin{equation}
\alpha = \sum_{i=1}^\infty |a_i|
\end{equation}

and for $j = 0,1,2,\ldots$ define nonnegative integers $n_j$ as follows. Let $n_0 = 0$ and let $n_j$ denote the first index for which

\begin{equation}
\sum_{i=1}^{n_j} |a_i| > \alpha - \frac{1}{j^3},
\end{equation}

$j = 1,2,\ldots$. Set
(3.19) \( b_i = \frac{1}{j}, c_i = ja_i, \ i = n_{j-1}+1, \ldots, n_j, \ j = 1, 2, \ldots \)

Then \( b_i c_i = a_i, i = 1, 2, \ldots, \lim_{i \to \infty} b_i = 0 \) and

\[
(3.20) \sum_{i=1}^{\infty} |c_i| = \sum_{j=1}^{n} \sum_{k=n_{j-1}+1}^{n_j} |a_k| < \alpha + \sum_{j=1}^{\infty} \frac{1}{j} < \infty.
\]

**Lemma 3.5** If \( L \) is a self-adjoint trace class operator on a separable Hilbert space \( H \), then \( L \) can be written as \( L^1L^2 \) where \( L^1 \) is compact and \( L^2 \) is trace class.

**Proof**

Let \( \{\lambda_i\}_{i=1}^{\infty} \) denote the eigenvalues of \( L \) repeated according to multiplicity and let \( \{\phi_i\}_{i=1}^{\infty} \) denote the corresponding eigenvectors. Then \( \{\lambda_i\}_{i=1}^{\infty} \) is a sequence of real numbers, each of finite multiplicity, and

\[
(3.21) \sum_{i=1}^{\infty} |\lambda_i| = |L|_1 < \infty.
\]

Applying the previous lemma there exist sequences \( \{\mu_i\}_{i=1}^{\infty} \) and \( \{\nu_i\}_{i=1}^{\infty} \) with \( \lim_{i \to \infty} \mu_i = 0, \) \( \sum_{i=1}^{\infty} |\nu_i| < \infty \) and \( \lambda_i = \mu_i\nu_i. \) Defining \( L^1 \) and \( L^2 \) by

\[
(3.22) L^1\phi = \sum_{i=1}^{\infty} \mu_i \langle \phi, \phi_i \rangle \phi_i, \ \phi \in H
\]

and

\[
(3.23) L^2\phi = \sum_{i=1}^{\infty} \nu_i \langle \phi, \phi_i \rangle \phi_i, \ \phi \in H
\]
respectively, the lemma immediately follows.

**Lemma 3.6** Let \( \{S_N\}_{N=1}^{\infty} \) be a sequence of bounded linear operators on a separable Hilbert space \( H \) which converges strongly to a bounded linear operator \( S \). Let \( \{L_N\}_{N=1}^{\infty} \) be a sequence of trace class operators on \( H \) which converges in trace norm to an operator \( L \). If \( L \) can be written as \( L = L_1 L_2 \) with \( L_1 \) compact and \( L_2 \) trace class then the sequence \( \{S_N L_N\}_{N=1}^{\infty} \) converges in trace norm to \( SL \).

**Proof**

The result follows immediately from

\[
\|S_N L_N - SL\|_1 \leq \|S_N (L_N - L)\|_1 + \|(S_N - S)L_1\|_1 \|L_2\|_1
\]

\[
\leq \|S_N\| \|L_N - L\|_1 + \|(S_N - S)L_1\| \|L_2\|_1.
\]

**Theorem 3.7** If \( Q \) and \( G \) are trace class operators then the operators \( \{\Pi(t)\}_{t=t_0}^{t_f} \) given by (2.10) and (2.11) are trace class. Moreover, if Hypotheses 3.1 and 3.2 hold and \( Q_N P_N \rightarrow Q \) and \( G_N P_N \rightarrow G \) in trace norm as \( N \rightarrow \infty \) then we have

\[
\lim_{N \rightarrow \infty} \|\Pi_N(t) P_N - \Pi(t)\|_1 = 0, \quad t_0 < t < t_f.
\]

**Proof**

That the operators \( \Pi(t), t_0 < t < t_f \) are trace class is an immediate consequence of the hypotheses of the theorem, (2.10), (2.11) and the fact that
the trace class operators form a two sided ideal of \( L(Z) \), the space of bounded linear operators on \( Z \) (see [15]).

The trace norm convergence stated in (3.25) will follow once we have shown that

\[
\lim_{N \to \infty} \| \Pi_N (t+1) P_N - \Pi(t+1) \|_1 = 0
\]

implies

(i) \[
\lim_{N \to \infty} \| T_N^* \Pi_N (t+1) T_N P_N - T_N^* \Pi(t+1) T_N \|_1 = 0
\]

and

(ii) \[
\lim_{N \to \infty} \| T_N^* \Pi_N (t+1) \hat{B}_N \Pi_N (t+1) T_N P_N - T_N^* \Pi(t+1) \hat{B} \Pi(t+1) T_N \|_1 = 0.
\]

To argue (i) we first note that Hypothesis 3.1 and Lemmas 3.5 and 3.6 imply

\[
\lim_{N \to \infty} \| T_N^* \Pi_N (t+1) P_N - T_N^* \Pi(t+1) \|_1 = 0.
\]

Taking adjoints we obtain

\[
\lim_{N \to \infty} \| \Pi_N (t+1) T_N P_N - \Pi(t+1) \|_1 = 0.
\]

Another application of the previous two lemmas yields

\[
\lim_{N \to \infty} \| T_N^* \Pi_N (t+1) T_N P_N - T_N^* \Pi(t+1) T_N \|_1 < \lim_{N \to \infty} \| T_N^* \Pi_N (t+1) T_N P_N - \Pi(t+1) \|_1
\]
where $\Pi(t+1) = \Pi_1(t+1)\Pi_2(t+1)$ is the factorization of $\Pi(t+1)$ described in Lemma 3.6.

The verification of (ii) is analogous and the theorem is proven.

We note that if Hypotheses 3.1 and 3.2 hold and if the operators $Q$ and $G$ are trace class with $Q_N$ and $G_N$ defined by

\begin{align}
Q_N &= P_N Q \\
G_N &= P_N G,
\end{align}

then Lemmas 3.5 and 3.6 imply that the trace norm convergence hypotheses in Theorem 3.7 hold. The significance of this observation will become apparent when examples are discussed in Section 4. Indeed, it is frequently the case in practice that $Q$ and $G$ have finite rank (and consequently are trace class) and the operators $Q_N$ and $G_N$ are defined as in (3.30), and (3.31).

### 3.2 Approximation on the Infinite Interval

Problem $(P_{2N})$ is Problem $(P2)$ for the control system in (3.2) and the performance index

\[
J_N(0,\infty,z_N(0),u) = \sum_{t=0}^{\infty} [\langle Q_N z_N(t), z_N(t) \rangle_Z + \langle R u(t), u(t) \rangle_U].
\]
Hypothesis 3.8. For each $N$, there exists exactly one nonnegative self-adjoint solution to the Riccati algebraic equation

$$\Pi_N = T_N^*[\Pi_N - \Pi_N B_N (R + B_N^* B_N)^{-1} B_N^* \Pi_N] T_N + Q_N.$$  

By Theorem 2.3, this implies that

$$\lim_{t \to -\infty} |\Pi_N - \Pi_N(t)| = 0$$

for each $N$, since $\dim(Z_N) < \infty$.

As in Theorem 2.5, we write

$$F_N = \tilde{R}_N^{-1} B_N^* \Pi_N T_N,$$

$$\tilde{R}_N = R + B_N^* B_N,$$

and

$$S_N = T_N - B_N F_N^*.$$  

From here on, $\Pi$ will be the nonnegative self-adjoint solution to the infinite dimensional Riccati algebraic equation (2.14) — when it exists — $F$ will be the corresponding feedback operator in (2.18) and $S$ will be the corresponding closed-loop operator in (2.21).
**Theorem 3.9.** If $\Pi_N^P$ converges strongly to some bounded linear operator $\Pi$, then $\Pi$ is a nonnegative self-adjoint solution to (2.14), $F_N^P$ converges in norm to $F$ and $S_N^P$ converges strongly to $S$.

**Proof.** This follows from Hypotheses 3.1 and 3.2, (3.33) and (3.35) - (3.37), and the fact that the control space $U$ has fixed finite dimension.

**Theorem 3.10.** Suppose that there exist positive constants $M$ and $r$, independent of $N$, with $r < 1$, such that

\[(3.38) \quad \Pi_N^P < M, \quad N = 1, 2, \ldots\]

and

\[(3.39) \quad |S_N^t| < Mr^t, \quad t = 1, 2, \ldots, \quad N = 1, 2, \ldots\]

Then a nonnegative self-adjoint solution $\Pi$ to (2.14) exists, and as $N \to \infty$,

\[(3.40) \quad \Pi_N^P \to \Pi \text{ strongly.}\]

If there exists a positive $m$, independent of $N$, such that

\[(3.41) \quad \theta_N > m, \quad N = 1, 2, \ldots\]

then (3.38) implies the existence of an $r$ less that one and independent of $N$ for which (3.39) holds.
Proof. For each $N$, let $\Pi_N(\cdot)$ satisfy (3.6) with $t_f = 0$ and $\Pi_N(0) = MI$, where $I$ denotes the identity operator on $Z_N$. From (2.32),

$$|\Pi_N - \Pi_N(-t)| \to 0 \text{ as } t \to \infty,$$

uniformly in $N$. Now, for $z \in Z$, write

$$(3.43) \quad \langle (\Pi_N - \Pi_N(-t))z, z \rangle_Z = \langle (\Pi_N - \Pi_N(-t))z, z \rangle_Z + \langle (\Pi_N(-t) - \Pi_N(-t))z, z \rangle_Z$$

$$+ \langle (\Pi_N(-t) - \Pi_N(-t))z, z \rangle_Z.$$

For $\varepsilon > 0$ choose $t > 0$ such that $|\Pi_N - \Pi_N(-t)|_Z < \varepsilon$ and $|\Pi_N(-t) - \Pi_N(-t)|_Z < \varepsilon$. Then, for $N$ and $N'$ large enough,

$$|\Pi_N(-t) - \Pi_N(-t)|_Z < \varepsilon. \quad \text{This shows that } \Pi_N \quad \text{is a Cauchy sequence in } Z$$

for each $z$. Therefore, $\Pi_N$ converges strongly to a nonnegative self-adjoint solution to (2.14).

An important application of this theorem is when the approximating open-loop operators $T_N$ have an exponential decay rate independent of $N$, $Q$ is coercive and $Q_N = P_N Q|_{Z_N}$. In this case, the zero control gives an upper bound, independent of $N$, on $\Pi_N$. Such is the case in the example discussed in Section 4.1 and in applications to flexible structures with no rigid-body modes and coercive structural damping.

Theorem 3.11. Suppose that $\Pi_N P_N$ converges strongly to $\Pi$, $Q_N P_N$ converges in trace norm to $Q$ (hence $Q$ is trace class), and (3.39) holds for positive $M$ and $r$ independent of $N$ with $r$ less than one. Then $\Pi_N P_N$ converges in trace norm to $\Pi$. 

Proof. Since \( \|s_N Q_N s_N^*\| < |s_N|^{-1} \|Q_N\|^{-1} \), the series in (2.26) converges in trace norm, uniformly in \( N \). The current result follows then from Lemmas 3.5 and 3.6.

Note that \( Q_N P_N \) converges in trace norm to \( Q \) if \( Q \) is trace class and \( Q = P_N Q P_N^* \).

**Theorem 3.12** If \( |\Pi_N| \) is bounded in \( N \), then a nonnegative self-adjoint solution \( \Pi \) to (2.14) exists, \( \Pi_N P_N \) converges weakly to \( \Pi \), and \( F_N P_N \) and \( S_N P_N \) converge strongly to \( F \) and \( S \), respectively.

Proof. According to [11 Theorem 6], \( \Pi_N P_N \) converges weakly to some nonnegative self-adjoint bounded \( \Pi \). It follows from (3.33) and Hypotheses 3.1 and 3.2 that \( \Pi \) satisfies (2.14) and that \( F_N \) and \( S_N \) converge as indicated.

Note that Theorem 3.12 holds if \( S_N P_N \) converges strongly but \( S_N^* P_N \) converges only weakly.

### 3.3 Implementation of the Approximation Schemes

In constructing the approximating operators \( T_N, B_N, Q_N \) and \( G_N \) a standard Galerkin approach is often taken; that is, \( T_N = P_N T, B_N = P_N B, Q_N = P_N Q \) and \( G_N = P_N G \). We note however that explicit representations for the operators \( T \) and \( B \) are frequently not available. In particular, this can occur when the discrete-time system (2.3) arises from the sampling of an infinite dimensional continuous-time system of the form (2.4). In this case it is the operators \( A \) and \( B \) which are approximated by a sequence of finite dimensional
operators \( A_N \) and \( B_N \) on \( Z_N \), from which an approximation to the semigroup \( \{ T(s) : s > 0 \} \) is obtained as \( T_N(s) = \exp(A_N s), s > 0 \). The operators \( T_N \) and \( B_N \) are then \( T_N = T_N(\tau) \) and \( B_N = \int_0^\tau T_N(s)B_N ds \), respectively. The strong convergence \( T_N P_N + T \) and \( B_N + B \) is then usually argued using an appropriate formulation of the Trotter-Kato theorem, a well known semigroup approximation result (see [15] [23]).

The expressions given by (3.3) - (3.7) are operator equations and although they are finite dimensional, they are not appropriate for computations. To make use of our approximation framework, we must first determine equivalent matrix formulations. Toward this end we assume, without loss of generality, that \( U = R^m \) with the standard basis and inner product and let \( \{ \phi_N^j \}_{j=1}^{K_N} \) be a basis for \( Z_N \). Define the \( K_N \times K_N \) Gram matrix \( M_N \) by

\[
[M_N]_{ij} = \langle \phi_N^i, \phi_N^j \rangle_{Z}.
\]

For an operator \( A \) we denote its matrix representation with respect to the bases defined above by \([A]\). Similarly, for an element \( z \in Z \) or \( u \in U \), we let its vector representation be given by \([z]\) or \([u]\) respectively. Standard calculations yield

\[
[T_N^*] = M_N^{-1} [T_N] M_N
\]

and

\[
[B_N^*] = [B_N]^T M_N.
\]

Defining
(3.47) \[ \hat{\Pi}_N(t) = M_N[\Pi_N(t)], \]

(3.48) \[ \hat{Q}_N = M_N[Q_N], \]

and

(3.49) \[ \hat{G}_N = M_N[G_N] \]

we obtain

(3.50) \[ [u_N^*(t)] = -[F_N(t)][z_N^*(t)], \quad t_0 < t < t_f^{-1}, \]

(3.51) \[ [F_N(t)] = [R_N(t)]^{-1}[B_N]^T\hat{\Pi}_N(t+1)[T_N], \]

(3.52) \[ [R_N(t)] = [R] + [B_N]^T\hat{\Pi}_N(t+1)[B_N], \]

(3.53) \[ \hat{\Pi}_N(t) = [T_N]^T\hat{\Pi}_N(t+1) - \]

\[ \hat{\Pi}_N(t+1)[B_N][R_N(t)]^{-1}[B_N]^T\hat{\Pi}_N(t+1)[T_N] + \hat{Q}_N, \quad t_0 < t < t_f^{-1}, \]

(3.54) \[ \hat{\Pi}_N(t_f) = \hat{G}_N. \]

Note that since \( Q_N \) and \( G_N \) are self-adjoint and nonnegative so too are \( \hat{Q}_N \) and \( \hat{G}_N \). Equations (3.50) - (3.54) are therefore in the form of the standard ones obtained for the feedback law for a discrete-time linear-quadratic regulator problem in \( R^N \). Consequently they can be solved using conventional techniques. The minimum value of the performance index is given by
Analogously, for the infinite time horizon problem, (3.33), (3.35) and (3.36) yield

\begin{align}
(3.56) \quad u^N_N(t) &= -[F^N_N][z^N_N(t)], \quad t > t_0,
(3.57) \quad [F^N_N] &= [\tilde{R}^N_N]^{-1}[B^N_N]^T\Pi^N_N[T^N_N]
(3.58) \quad [\tilde{R}^N_N] &= [R] + [B^N_N]^T\Pi^N_N[B^N_N]
\end{align}

where \( \Pi^N_N \) is the solution to the matrix algebraic Riccati equation

\begin{equation}
(3.59) \quad \Pi^N_N = [T^N_N]^T(\Pi^N_N - \Pi^N_N[\tilde{R}^N_N]^{-1}[B^N_N]^T\Pi^N_N)[T^N_N] + \hat{Q}_N,
\end{equation}

with \( \hat{Q}_N \) given by (3.48). The minimum value of the performance index is given by

\begin{equation}
(3.60) \quad J^*_N = [z^N_N(t_0)]^T\Pi^N_N[z^N_N(t_0)].
\end{equation}
4. **Examples and Numerical Results**

In this section we describe the application of the general approximation framework developed above to a variety of examples. In addition to theoretical considerations, in each of the examples below, we discuss some numerical results for an infinite-time horizon problem of the form given in Problem (P2). All numerical studies were performed on an IBM Personal Computer. The machine we used was equipped with an Intel 8086 math co-processor chip and 640K bytes of random access memory (of which less than 384K was required).

Matrix exponentials were computed from eigenvalue-eigenvector decompositions obtained using the QR algorithm. The matrix Riccati equations (3.59) were solved using a Schur-vector decomposition of the Hamiltonian matrix (see [18][24]). It should be noted that if the eigenvalue pairs of the Hamiltonian matrix for a continuous-time linear-quadratic regulator problem are asymptotic to \( \pm \gamma(n) \) as \( n \to \infty \), then the eigenvalue pairs of the Hamiltonian matrix for the corresponding discrete-time problem will be asymptotic to \( e^{\pm \gamma(n)\tau} \) as \( n \to \infty \). Consequently, for all but very small \( \tau \), conditioning problems arise more quickly than in the continuous-time case when the approximating matrix algebraic Riccati equations are solved.

4.1. **The Heat Equation with Boundary Input**

In this example we consider the scalar parabolic system with boundary control given by

\[
\frac{\partial w(s,x)}{\partial s} = \frac{\partial}{\partial x} a(x) \frac{\partial}{\partial x} w(s,x), \quad s > 0, \ x \in (0,1)
\]
(4.2) \( w(s,0) = 0, \ w(s,1) = v(s) \)

(4.3) \( w(0,x) = \phi(x) \)

with \( a \in H^1(0,1), \ a(x) > 0, \ x \in [0,1], \ \phi \in H^0(0,1) = L_2(0,1) \) and \( v \in L_2(0,\infty) \).

To formulate the discrete-time state equation for this system we let \( \tau \) denote the sampling interval and consider only piecewise constant controls \( v \) given by

(4.4) \( v(s) = u(t) \quad s \in [t\tau, (t+1)\tau), \)

\( t = 0, 1, 2, \ldots \). We choose as our state space \( Z \) the Sobolev space \( H^0(0,1) \) with the usual inner product

(4.5) \( \langle \phi, \psi \rangle_Z = \langle \phi, \psi \rangle_0 = \int_0^1 \phi(\theta) \psi(\theta) d\theta. \)

The state \( z(t) \in Z \) is

(4.6) \( z(t) = \lim_{s \to t\tau} w(s,\cdot), \ t = 1, 2, \ldots \)

(4.7) \( z(0) = \phi. \)

For \( t \in \{0,1,2,\ldots\} \), we define \( y(s) \in Z \) by

(4.8) \( y(s) = w(s,\cdot) - \psi_0 u(t), \ s \in (t\tau, (t+1)\tau) \)
(4.9) \[ y(t\tau) = z(t) - \psi_0 u(t), \]

where \( \psi_0 \in Z \) is given by \( \psi_0(x) = x, x \in [0,1] \). A straightforward calculation reveals that \( y(s) = y(s,\cdot) \) satisfies

\[ (4.10) \quad y(s) = DaDy(s) + a'u(t), \quad s \in (t\tau, (t+1)\tau) \]

\[ (4.11) \quad y(s)|_0 = 0, \quad y(s)|_1 = 0, \quad s \in (t\tau, (t+1)\tau) \]

\[ (4.12) \quad y(t\tau) = z(t) - \psi_0 u(t), \]

where \( D \) denotes the differentiation operator on \( H^1(0,1) \).

Let \( A: \text{dom}(A) \subset Z \to Z \) be given by

\[
\text{Dom}(A) = H^2(0,1) \cap H^1_0(0,1) = \{ \phi \in H^0(0,1): \phi \in H^2(0,1), \phi(0) = \phi(1) = 0 \}
\]

(4.13) \( A\psi = DaD\psi \).

The operator \( A \) is densely defined and self-adjoint. It satisfies

\[ (4.14) \quad \langle Az, z \rangle_Z \leq -\omega |z|_Z^2, \quad z \in \text{Dom}(A) \]

for some \( \omega > 0 \) and has compact resolvent. Also, \( A \) is the infinitesimal generator of an analytic semigroup of contractions \( \{ T(s) : s > 0 \} \) on \( Z \) which, in light of (4.14), satisfies \( |T(s)| < e^{-\omega s}, s > 0 \). It follows therefore, that
\(y(s) = T(s-tT)y(tT) + \int_{tT}^{sT} T(s-\sigma)a'd\sigma u(t), \quad s \in [tT, (t+1)T).\)

The continuity of \(y\), (4.6), (4.8) and (4.9) imply

\[z(t) = y(tT) + \psi_0 u(t)\]

and

\[z(t+1) = y((t+1)T) + \psi_0 u(t),\]

and hence that

\[z(t+1) = y((t+1)T) + \psi_0 u(t)\]

\[= T(\tau)(z(t) - \psi_0 u(t)) + \int_{tT}^{(t+1)T} T((t+1)\tau-\sigma)a'd\sigma u(t) + \psi_0 u(t).\]

Defining the operators \(T \in L(Z)\) and \(B \in L(R^1, Z)\) by

\[Tz = T(\tau)z, \quad z \in Z\]

and

\[Bu = [(I - T(\tau))\psi_0 + \int_0^\tau T(\sigma)a'd\sigma]u, \quad u \in R^1,\]

we obtain

\[z(t+1) = Tz(t) + Bu(t), \quad t = 0, 1, 2, \ldots\]
\[ (4.22) \quad z(0) = \phi. \]

We take the performance index to be

\[ (4.23) \quad J(g_0;0,t_f,\phi,u) = \sum_{t=0}^{t_f-1} \left\{ q_0 |z(t)|^2_0 + ru(t)^2 \right\} + g_0 |z(t_f)|^2_0 \]

with \( q_0, g_0 > 0 \) and \( r > 0 \).

Applying the theory developed in Section 2.1, we have, for the finite time interval problem, that the optimal control is given by

\[ (4.24) \quad u_*(t) = -F(t)z_*(t), \quad t = 0, 1, 2, \ldots, t_f-1, \]

where for each \( t \), \( F(t) \) is the continuous linear functional on \( Z \) given by (2.8) - (2.11). It follows that \( F(t) \) has a representation \( f(t, \cdot) \in H^0(0,1) \) and that

\[ (4.25) \quad F(t)\psi = \int_0^1 f(t, \cdot)\psi(\theta) d\theta, \]

\( \psi \in H^0(0,1), \ t = 0, 1, 2, \ldots, t_f-1. \)

For the infinite time interval problem \( (t_f = \infty, g_0 = 0) \), it is immediately clear that Hypothesis 2.4 is satisfied. It is also clear that (4.14) implies that for each \( z(0) \in H^0(0,1) \), \( u(t) = 0, \ t = 0, 1, 2, \ldots \) is an admissible control, and hence that there exists a unique nonnegative self-adjoint solution of the Riccati algebraic equation (2.14). From (2.17) - (2.19) we obtain
(4.26) \( u_\ast(t) = -Fz_\ast(t), \quad t = 0, 1, 2, \ldots \)

where \( F \) is a continuous linear functional on \( Z \) and

(4.27) \( F\psi = \int_0^1 f(\theta)\psi(\theta)d\theta, \quad \psi \in \mathcal{H}_0^0(0, 1), \)

with \( f \in \mathcal{H}_0^0(0, 1) \).

We define an approximation scheme using a standard Ritz-Galerkin approach. We note that the operator \(-A\) is coercive and that it can be written as \(-A = L^*L\) where the operator \(L\) and its adjoint \(L^*\) are given by

(4.28) \( L\psi = a^{1/2}\psi, \quad \psi \in \mathcal{H}_0^1(0, 1), \)

and

(4.29) \( L^*\psi = -Da^{1/2}\psi, \quad \psi \in \mathcal{H}_1^1(0, 1), \)

respectively. We define the space \( V = \mathcal{H}_0^1(0, 1) \) together with the inner product

(4.30) \( \langle \phi, \psi \rangle_V = \langle L\phi, L\psi \rangle_Z, \quad \phi, \psi \in V. \)

We note that \( V \) is the energy space associated with the operator \(-A\), \( V = \text{Dom}((-A)^{1/2}) \), and it is the closure of \( \text{Dom}(A) \) with respect to the energy norm \( |.|_V \), which satisfies
\begin{equation}
\left| \phi \right|_V^2 = \langle L\phi, L\phi \rangle_Z = \langle -A\phi, \phi \rangle_Z, \quad \phi \in \text{Dom}(A).
\end{equation}

For each \( N = 2, 3, \ldots \) let \( \Delta_N \) denote the uniform partition of the interval [0,1] given by \( \{0, \frac{1}{N}, \frac{2}{N}, \ldots, \frac{N-1}{N}, 1\} \). Let \( \{e_{N,j}\}_{j=1}^{N-1} \) denote the usual linear B-splines on [0,1] corresponding to the partition \( \Delta_N \) and which satisfy \( e_{N,j}(0) = e_{N,j}(1) = 0, \) \( j = 1, 2, \ldots, N-1 \). The \( e_{N,j} \) are given by

\begin{equation}
(4.32) \quad e_{N,j}(\theta) = \begin{cases} 
N(\theta - \frac{(j-1)}{N}) & \theta \in [\frac{j-1}{N}, \frac{1}{N}] \\
N(\frac{(j+1)}{N} - \theta) & \theta \in [\frac{1}{N}, \frac{j+1}{N}] \\
0 & \text{elsewhere}
\end{cases}
\end{equation}

\( j = 1, 2, \ldots, N-1 \) and are depicted in the figure below.

Let \( Z_N = \text{span}\{e_{N,j}\}_{j=1}^{N-1} \). We note that \( Z_N \subset V \) and \( \text{dim} \ Z_N = N-1, \) \( N = 2, 3, \ldots \).

Define \( P_N : Z \rightarrow Z_N \) to be the orthogonal projection of \( Z \) onto \( Z_N \) with respect to the \( \langle \cdot, \cdot \rangle_Z \) inner product and \( P_N : V \rightarrow Z_N \) to be the orthogonal projection of \( V \) onto \( Z_N \) with respect to the \( \langle \cdot, \cdot \rangle_V \) inner product.

We define the operator \( \tilde{A}_N : Z_N \rightarrow Z_N \) as the inverse of the operator given
by

\[(4.33) \quad A_N^{-1} = P_N A^*_N |_{Z_N}.\]

The invertibility of $A$ of course follows from the coercivity of $-A$ (see (4.14)). On the other hand, a straightforward calculation yields

\[(4.34) \quad \langle A_N^{-1} z_N, z_N \rangle_V = |z_N|^2, \quad z_N \in Z_N.\]

Consequently the operator $A_N^{-1}$ is invertible and the operator $A_N$ is well defined. It is also self-adjoint. Indeed

\[(4.35) \quad \langle A_N z_N, y_N \rangle_Z = -\langle z_N, y_N \rangle_V, \quad z_N, y_N \in Z_N.\]

From (4.35) it also follows that

\[(4.36) \quad \langle A_N z_N, z_N \rangle_Z < -\omega |z_N|^2, \quad z_N \in Z_N.\]

It can be concluded therefore, that $A_N$ is the infinitesimal generator of a semigroup of bounded linear operators on $Z_N$, \{ $T_N(s) : s > 0$ \} with

$T_N(s) = \exp(A_N s)$, $s > 0$ and satisfying

\[(4.37) \quad |T_N(s)|_Z < e^{-\omega s}, \quad s > 0.\]

Elementary properties of spline functions (see [27]) imply $P_N \rightarrow I$ as $N \rightarrow \infty$, strongly on $V$. Furthermore, $A^{-1}$ compact and
(4.38) \[ |P_N A^{-1} z - A^{-1} z|_Z < \left| P_N A^{-1} z - \frac{A^{-1} z}{V} = |(P_N - I) A^{-1} z|_V \right. \]

imply that \( P_N A^{-1} \to A^{-1} \) as \( N \to \infty \) in the uniform operator topology on \( L(Z) \).

We have therefore, that

(4.39) \[ |A^{-1}_N P_N - A^{-1}|_Z \to 0 \]
as \( N \to \infty \).

From (4.37) and (4.39) we conclude (see [4], [15])

(4.40) \[ T_N(s)P_N z \to T(s)z \]

and

(4.41) \[ T^*_N(s)P_N z \to T^*(s)z \]
as \( N \to \infty \) for each \( z \in Z \), uniformly in \( s \) for \( s \) in compact subintervals.

With \( T_N = T_N(\tau) \), \( Q_N = q_0 P_N \), \( G_N = g_0 P_N \) and

(4.42) \[ B_N = (I - T_N)P_N \psi + \int_0^T T_N(\sigma)P_N a'd\sigma, \]

(4.40), (4.41) and elementary approximation properties of spline functions guarantee that Hypotheses 3.1 and 3.2 hold and hence that the convergence results for the finite time interval problem given in Theorem 3.3 apply.

For the infinite time interval Problem, (4.37) implies that Hypothesis 3.8 is satisfied. Moreover, if \( q_0 > 0 \) the conditions given in the statement of Theorem 3.10 are satisfied (see the Remark following the proof of Theorem 3.10) and consequently the convergence results for the infinite time horizon problem
given in Theorem 3.9 hold.

Define the $\mathbb{R}^{N-1}$ vector valued function $E_N$ on $[0,1]$ by

\begin{equation}
E_N(\theta)^T = (e_1^N(\theta), e_2^N(\theta), \ldots, e_{N-1}^N(\theta)),
\end{equation}

$\theta \in [0,1]$ and the $(N - 1) \times (N - 1)$ Gram matrix associated with the basis elements $\{e_j^N\}_{j=1}^{N-1}$

\begin{equation}
M_N = \langle E_N, E_N^T \rangle_z .
\end{equation}

A straightforward calculation yields

\begin{equation}
M_N = \frac{1}{N}
\begin{bmatrix}
\frac{2}{3} & \frac{1}{6} & 0 & \cdots & 0 \\
\frac{1}{6} & \frac{2}{3} & \frac{1}{6} & \cdots & 0 \\
0 & \frac{1}{6} & \frac{2}{3} & \frac{1}{6} & \cdots \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & \frac{1}{6} & \frac{2}{3} & \frac{1}{6}
\end{bmatrix}
\end{equation}

Let the $(N - 1) \times (N - 1)$ matrix $H_N$ be given by

\begin{equation}
H_N = \langle E_N, A_N E_N^T \rangle_z .
\end{equation}

Then

\begin{equation}
[H_N]_{ij} = \langle e_i^N, A_N e_j^N \rangle_z
\end{equation}

\begin{equation}
= -\langle e_i^N, e_j^N \rangle_z
\end{equation}
In the case $a(x) = a$, $x \in [0,1]$, a constant, we have

\[
\begin{bmatrix}
-2 & 1 & 0 \\
1 & -2 & 1 \\
0 & 1 & -2
\end{bmatrix}
\]

The matrix representation $[A_N]$ for the operator $A_N$ is given by

\[
[A_N] = M^{-1} H_N'
\]

while for the operators $T_N$, $Q_N$ and $G_N$ we have

\[
[T_N] = \exp([A_N] \tau),
\]

\[
[Q_N] = q_0 I_N
\]

and

\[
[G_N] = g_0 I_N
\]
where $I_N$ denotes the $(N - 1) \times (N - 1)$ identity matrix. If we define

\[ \psi_{ON}, a'_N \in \mathbb{R}^{N-1} \]

by

\begin{align*}
(4.53) & \quad \psi_{ON} = \langle E_N, \psi_0 \rangle_Z \\
(4.54) & \quad a'_N = \langle E_N, a' \rangle_Z, \\
\end{align*}

and

\[ j = 1, 2, \ldots, N-1 \]

we obtain

\begin{align*}
(4.55) & \quad [B_N] = (I_N - [T_N])M^{-1}_N \psi_{ON} + \int_0^T \exp([A_N]_N^{-1} a'_N) \text{d}\sigma \\
& \quad = (I_N - [T_N])M^{-1}_N \psi_{ON} + [A_N]^{-1}(M^{-1}_N [A_N]_N^{-1} a'_N). \\
\end{align*}

When the finite dimensional approximating gain matrices $[F_N(t)], t = 0, 1, 2, \ldots, t_f - 1$ for the finite time interval problem have been computed using

(3.51)-(3.54), an approximation $f_N(t, \cdot)$ to the feedback kernel $f(t, \cdot)$ can be obtained from

\begin{align*}
(4.56) & \quad f_N(t, \theta) = E_N(\theta)M^{-1}_N[F_N(t)]^T, \\
\end{align*}

\[ t = 0, 1, 2, \ldots, t_f - 1, \theta \in [0, 1]. \quad \text{We have} \]

\begin{align*}
(4.57) & \quad f_N(t, \cdot) + f(t, \cdot) \\
\end{align*}
in $H^0(0,1)$ as $N \to \infty$ for each $t = 0, 1, 2, \ldots, t_f - 1$.

Similarly, for the infinite time interval problem, an approximation $f_N$ to $f$ is given by

\begin{equation}
 f_N(\theta) = E_N(\theta)^T M_N^{-1} [F_N]^T,
\end{equation}

$\theta \in [0, 1]$ where the matrix $[F_N]$ is computed from (3.57) - (3.59). We have

\begin{equation}
 f_N \to f
\end{equation}

in $H^0(0,1)$ as $N \to \infty$.

We demonstrate the feasibility of our schemes on an infinite time interval problem of the form discussed above. Taking $q_0 = 1.0$, $r = 1.0$, $a(x) = a = 1.0$, $x \in [0,1]$ and $\tau = .01$ we obtained the approximating feedback kernals given in Table 4.2 and shown in Figure 4.3 below.
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Table 4.2
Figure 4.3
4.2. Hereditary or Time Delay Systems

In this example we consider linear hereditary control systems of the form

\begin{equation}
\dot{x}(s) = Lx_s + B_0 v(s) \quad s \geq 0
\end{equation}

where \( x(s) \in \mathbb{R}^n \), \( x_s \in L_2((-r,0);\mathbb{R}^n) \) for some \( r > 0 \), \( v \in L_2((0,s_1);\mathbb{R}^m) \) for all \( s_1 < \infty \) and \( B_0 \in L(\mathbb{R}^m,\mathbb{R}^n) \). The function \( x_s \) represents the history on the interval \([s-r,s]\); that is \( x_s(\theta) = x(s+\theta) \), \( \theta \in [-r,0] \). The operator \( L \) is assumed to be of the form

\begin{equation}
L \phi = \sum_{i=0}^{v} A_i \phi(-r_i) + \int_{-r}^{0} A(\theta) \phi(\theta) d\theta
\end{equation}

where \( A_i \in L(\mathbb{R}^n) \), \( i = 0,1,2,\ldots,v \), \( A(\cdot) \in L_2((-r,0); L(\mathbb{R}^n)) \) and \( 0 = r_0 < r_1 < \cdots < r_v = r \). The initial data is given by

\begin{equation}
x(0) = \eta, \quad x_0 = \phi,
\end{equation}

where \( \eta \in \mathbb{R}^n \) and \( \phi \in L_2((-r,0); \mathbb{R}^n) \).

Once again, we formulate the discrete-time control problem by letting \( \tau \) denote the sampling interval and considering piecewise constant control inputs of the form

\begin{equation}
v(s) = u(t) \quad s \in [t\tau, (t+1)\tau),
\end{equation}

t = 0,1,2,\ldots, \text{ where } u(t) \in \mathbb{R}^m \text{ for each } t. \text{ The state space is }

\begin{equation}
Z = \mathbb{R}^n \times L_2((-r,0);\mathbb{R}^n)
\end{equation}
with the inner product

\[(4.65) \quad \langle \eta, \phi \rangle, \quad \langle \xi, \psi \rangle_Z = \langle \eta, \xi \rangle_{\mathbb{R}^n} + \langle \phi, \psi \rangle_{L_2} \cdot \]

The state of the discrete-time control system is

\[(4.66) \quad z(t) = (x(t; \eta, \phi, v), x_{t+}(\eta, \phi, v)), \quad t = 0, 1, 2, \ldots, \]

where \(x(s; \eta, \phi, v)\) denotes the solution at time \(s\) to the system \((4.60) - (4.62)\) and \(x_s(\eta, \phi, v)\) its history on \([s-r, s]\). Then

\[(4.67) \quad z(t+1) = T(t)z(t) + \int_0^T T(\sigma) B\sigma u(t), \quad t = 0, 1, 2, \ldots, \]

\[(4.68) \quad z(0) = (\eta, \phi) \]

where \(\{T(s) : s > 0\}\) is the \(C_0\) semigroup of bounded linear operators on \(Z\) with infinitesimal generator \(A: \text{Dom}(A) \subset Z \rightarrow Z\) given by

\[
\text{Dom}(A) = \{(\xi, \psi) \in Z : \psi \in H^1((-r, 0); \mathbb{R}^n), \xi = \psi(0)\} \\
(4.69) \quad A(\psi(0), \psi) = (L\psi, D\psi),
\]

and the operator \(B: \mathbb{R}^m \rightarrow Z\) is defined by

\[(4.70) \quad Bu = (B_0 u, 0), \quad u \in \mathbb{R}^m. \]

Letting \(T = T(\tau)\) and \(B = \int_0^T T(\sigma) B\sigma d\sigma\), we obtain
\( z(t+1) = Tz(t) + Bu(t), \ t = 0,1,2,... \)

\( z(0) = (\eta, \phi). \)

Let \( Q_0, \ G_0 \in L(\mathbb{R}^n) \) be symmetric and nonnegative and let \( R \in L(\mathbb{R}^m) \) be symmetric and positive definite. We consider the discrete-time linear-quadratic regulator problem with state given by (4.71) and performance index

\[
J(G; 0, t_f, \eta, \phi, u) = \sum_{t=0}^{t_f-1} \langle Qz(t), z(t) \rangle_{\mathbb{R}^n} + \langle Ru(t), u(t) \rangle_{\mathbb{R}^m} + \langle Gz(t_f), z(t_f) \rangle_{\mathbb{R}^n}
\]

where the operators \( Q : \mathbb{R}^n \to \mathbb{R}^n \) and \( G : \mathbb{R}^n \to \mathbb{R}^n \) are given by

\[
Q(\xi, \psi) = (Q_0 \xi, 0)
\]

and

\[
G(\xi, \psi) = (G_0 \xi, 0)
\]

respectively. For the infinite time problem we of course have \( t_f = \infty \) and \( G_0 = 0 \).

For the finite time problem with the operators, \( T, B, Q \) and \( G \) as defined above, we apply the theory developed in Section 2.1 and conclude that the optimal control is given by

\[
u_*(t) = -F(t)z_*(t), \quad t = 0,1,2,..., t_f-1,
\]
where the operators \( F(t) \in L(Z, \mathbb{R}^n) \) are given by (2.8) - (2.11). The operator \( F(t) \) can be represented by a matrix of operators, \([F^0(t), F^1(t)]\) with \( F^0(t) \in L(\mathbb{R}^n, \mathbb{R}^m) \) and \( F^1(t) \in L(L_2((-r,0); \mathbb{R}^n); \mathbb{R}^m) \). It follows from (4.70) therefore that

\[
(4.76) \quad u_\ast(t) = -f^0(t)x_\ast(t\tau) - \int_{-r}^{0} f^1(t,\theta)(x_\ast)_\tau(\theta)d\theta, \quad t = 0, 1, 2, \ldots, t_f-1,
\]

where \( f^0(t) \) is an \( m \times n \) matrix and \( f^1(t,\cdot) \) is a square integrable \( m \times n \) matrix valued function on \((-r,0)\).

For the infinite time problem we assume that our original hereditary system and \( Q_0 \) are such that there exists an admissible control for each \( z(0) = (n,\phi) \in Z \) and that Hypothesis 2.4 is satisfied. Then Theorems 2.3 and 2.5 imply that there exists a unique nonnegative self-adjoint solution \( H \) to the Riccati algebraic equation (2.14) with the optimal control \( u_\ast \) given by

\[
(4.77) \quad u_\ast(t) = -Fz_\ast(t), \quad t = 0, 1, 2, \ldots,
\]

where \( F \in L(Z, \mathbb{R}^m) \) is given by (2.18) - (2.19). The feedback gain \( F \) can be represented by a matrix of operators \([F^0, F^1]\) with \( F^0 \in L(\mathbb{R}^n, \mathbb{R}^m) \) and \( F^1 \in L(L_2((-r,0); \mathbb{R}^n), \mathbb{R}^m) \). We have

\[
(4.78) \quad u_\ast(t) = -f^0x_\ast(t\tau) - \int_{-r}^{0} f^1(\theta)(x_\ast)_\tau(\theta)d\theta, \quad t = 0, 1, 2, \ldots,
\]

where \( f^0 \) is an \( m \times n \) matrix and \( f^1 \) is a square integrable \( m \times n \) matrix valued function on \((-r,0)\).

Numerical methods for the approximate solution of the continuous time linear quadratic regulator problem for hereditary systems in closed-loop form
have been studied extensively (see [7],[16],[19],[25],[26]). Most closely related to the approximation framework which we have developed here for the discrete-time problem are the treatments for the continuous-time problem in [2],[11],[14] and [17]. The first approximation scheme applied to the continuous-time linear-quadratic control problem for hereditary systems was the AVE scheme used in [1],[11],[25] and [26] which approximates the history by piecewise constant functions and derivatives with finite differences. In [1], Banks and Burns cast AVE in its modern semigroup approximation form, proved strong convergence of the approximating open-loop semigroups and used the scheme to compute the open-loop optimal control. Gibson [11] used the convergence results in [1] to prove strong convergence of the solutions to the approximating continuous-time Riccati equations and uniform norm convergence of the approximating feedback control laws. However, the rate of convergence as it was observed in numerical studies was relatively slow. Spline based schemes for continuous-time systems were developed for the finite time interval problem in [17] and for the infinite time interval problem in [2]. With regard to the minimization of the performance index these schemes represented an improvement over the AVE scheme. However, they appeared to yield only weak convergence of the solutions to the approximating Riccati equations and strong operator convergence of the approximating feedback gains.

Recently, a new spline-based state approximation for hereditary systems has been proposed in [14]. When applied to the continuous time control problem, this new method performs at the level of the earlier spline-based schemes and yields strong convergence of the approximating feedback gains. We have chosen this scheme to describe and implement here for the discrete-time problem.
To simplify the presentation, we consider systems of the form (4.60) having only a single discrete delay ($\nu = 1$) and no distributed delay term ($A(\theta) \equiv 0$). A detailed description of the scheme in complete generality can be found in [14].

For each $N = 1, 2, \ldots$ define

\begin{align*}
(4.79) \quad e_N^0 &= (I_n, 0) \quad \text{and} \quad e_N^j = (0, e_N^j), \quad j = 1, 2, \ldots, N+1,
\end{align*}

where $I_n$ denotes the $n \times n$ identity matrix and the $e_N^j$ are the $n \times n$ matrix valued piecewise linear elements on $[-r, 0)$ given by

\begin{align*}
(4.79) \quad e_N^j(\theta) = \begin{cases} 
\frac{N}{r} (\theta + \frac{r}{N}) I_n & \theta \in \left[ -\frac{r}{N}, 0 \right) \\
0 & \text{elsewhere}
\end{cases}
\end{align*}

\begin{align*}
(4.80) \quad e_N^j(\theta) = \begin{cases} 
\frac{N}{r} (\theta + j \frac{r}{N}) I_n & \theta \in \left[ -j \frac{r}{N}, -(j-1) \frac{r}{N} \right) \\
\frac{-N}{r} (\theta + (j-2) \frac{r}{N}) I_n & \theta \in \left[ -(j-1) \frac{r}{N}, -(j-2) \frac{r}{N} \right) \\
0 & \text{elsewhere}
\end{cases}
\end{align*}

\[ j = 2, 3, 4, \ldots, N \]
\[ e^{N+1}_N(\theta) = \begin{cases} \frac{-N}{r} (\theta + (N-1) \frac{\pi}{n}) \mathbf{1}_n, & \theta \in [-r, -(n-1) \frac{\pi}{n}] \\ 0, & \text{elsewhere} \end{cases} \]

Let

\[(4.81) \quad Z_N = \{ z \in Z : z = \sum_{j=0}^{N+1} \alpha_j e^j_N, \alpha_j \in \mathbb{R}^n \} \]

Note that \( \dim Z_N = n(N+2) \). We shall refer to the collection \( \{e^j_N\}_{j=0}^{N+1} \) as a "basis" for \( Z_N \) and a vector \( \alpha \in \mathbb{R}^n \) as being a "coordinate vector" for an element in \( Z_N \). Defining

\[(4.82) \quad E^T_N = (e^0_N, e^1_N, \ldots, e^{N+1}_N) \]

we have

\[(4.83) \quad Z_N = \{ z \in Z : z = E^T_N \alpha, \alpha \in \mathbb{R}^n \} \]

Let \( M_N \) denote the Gram matrix corresponding to the basis \( \{e^j_N\}_{j=0}^{N+1} \). A straightforward computation yields

\[(4.84) \quad M_N = \text{diag}(I_n, \frac{r}{n} m_N \otimes I_n) \]

where the \((N+1) \times (N+1)\) matrix \( m_N \) is given by
and \( \otimes \) denotes the Kronecker product.

Let \( P_N \) denote the orthogonal projection of \( Z \) on to \( Z_N \) with respect to the inner product (4.65). It follows that

\[
(4.86) \quad P_N(\xi, \psi) = (\xi, p_N \psi)
\]

where \( p_N \) is the orthogonal projection of \( L^2((-r,0); \mathbb{R}^n) \) on to \( \text{span}\{e^j_N\}_{j=1}^{N+1} \) with respect to the usual \( L^2 \) inner product. We have

\[
(4.87) \quad P_N(\xi, \psi) = E_N^T \alpha_N(\xi, \psi)
\]

where \( \alpha_N(\xi, \psi) \in \times \mathbb{R}^n \) is given by

\[
(4.88) \quad \alpha_N(\xi, \psi) = M_N^{-1} \text{col} (\xi, \psi_N^1, \psi_N^2, \ldots, \psi_N^{N+1})
\]

with

\[
(4.89) \quad \psi_N^j = \int_{-r}^{0} e^j_N(\theta) \psi(\theta) d\theta.
\]
We shall set $T_N = \exp(A_N \tau)$ where $A_N : Z_N \to Z_N$ is an appropriately defined finite dimensional approximation to the operator $A$ given by (4.69). Noting that $Z_N \not\subset \text{Dom}(A)$, we motivate the definition of $A_N$ by first formally extending the operator $A$ to an operator defined on $Z_N$.

For $z_N = (\xi_N, \psi_N) \in Z_N$ define

\begin{equation}
A_N^Nz_N = (A_0^N + A_1^N(-r), D^+N \psi_N + \delta_N(\xi_N - \lim_{\theta \to 0^+} N(\theta)))
\end{equation}

where $\delta$ is the Dirac delta impulse concentrated at zero and $D^+\psi$ denotes the right hand derivative of $\psi$. For each $N = 1, 2, \ldots$ let $A_N : Z_N \to Z_N$ be given by

\begin{equation}
A_N^NZ_N = (A_0^N + A_1^N(-r), p_N D^+N \psi_N + \delta_N(\xi_N - \lim_{\theta \to 0^+} N(\theta)))
\end{equation}

where

\begin{equation}
\delta_N = E_T^N Y_N
\end{equation}

with

\begin{equation}
Y_N = M_N^{-1} \col 0, \lim_{\theta \to 0^-} \psi_N^1(\theta), \ldots, \lim_{\theta \to 0^+} \psi_N^{N+1}(\theta).
\end{equation}

To compute $[A_N]$, the matrix representation for the operator $A_N$, we let $[z_N]$ denote the coordinate vector representation for an element $z_N \in Z_N$. Then from

\begin{equation}
[A_N^Nz_N] = [A_N][z_N]
\end{equation}
and

\[(4.95) \quad [A_N^{\alpha N}] = M^{-1}_N \alpha_N(A_N^{\alpha N}) = M^{-1}_N \langle E_N^\alpha A_N^\alpha N^{\alpha N} \rangle_Z = M^{-1}_N \langle E_N^\alpha A_N^\alpha E_N^\alpha T \rangle_Z [z_N] \]

we obtain

\[(4.96) \quad [A_N] = M^{-1}_N H_N \]

where

\[(4.97) \quad H_N = \langle E_N^\alpha A_N^\alpha E_N^\alpha T \rangle_Z \ . \]

Using the definitions of \(A_N\) and \(E_N\), a straightforward calculation (see [14]) yields

\[(4.98) \quad H_N = \begin{bmatrix}
A_0 & 0 & 0 & A_1 \\
\text{In} & 0 & h_N \otimes \text{In}
\end{bmatrix} \]

where the \((N+1) \times (N+1)\) matrix \(h_N\) is given by
The matrix representation for the operator $T_N = \exp(A_N \tau)$ can then be computed from

$$[T_N] = \exp([A_N] \tau).$$

We define the operators $B_N : R^m \rightarrow Z_N$, $Q_N : Z_N \rightarrow Z_N$ and $G_N : Z_N \rightarrow Z_N$ by

$$\bar{B}_N = P_N B$$

$$Q_N = P_N Q$$

and

$$G_N = P_N G$$

from which we obtain
(4.104) \[ B_N = \text{col} (B_0, 0, \ldots, 0) \in \mathbb{R}^{n(N+2) \times m} \]

(4.105) \[
\begin{bmatrix}
Q_0 & 0 & \cdots & 0 \\
0 & Q_0 & & 0 \\
\vdots & \ddots & \ddots & \vdots \\
0 & \cdots & 0 & Q_0
\end{bmatrix}
\in \mathbb{R}^{n(N+2) \times n(N+2)}
\]

and

(4.106) \[ C_N = \begin{bmatrix}
C_0 & 0 & \cdots & 0 \\
0 & C_0 & & 0 \\
\vdots & \ddots & \ddots & \vdots \\
0 & \cdots & 0 & C_0
\end{bmatrix}
\in \mathbb{R}^{n(N+2) \times n(N+2)} \]

Finally, defining

(4.107) \[ B_N = \int_0^T \exp (A_N s) E_N ds \]

we have

(4.108) \[ [B_N] = \int_0^T \exp ([A_N] s) [E_N] ds. \]

Once the matrix representations for the approximating feedback gains have been computed, \([F_N(t)], t = 0, 1, 2, \ldots, t_{f-1}\) from (3.51) - (3.54) for the finite time interval problem and \([F_N]\) from (3.57) - (3.59) (assuming, for the moment, that solutions to (3.34) exist) for the infinite time interval problem,
approximations for $f^0$, $f^1(t, \cdot)$, $t = 0, 1, 2, \ldots, t_f - 1$ and $f^0, f^1(\cdot)$ can be computed from

\begin{equation}
(4.109) \quad ((f^0_N(t))^T, (f^1_N(t, \cdot))^T) = E_N^{-1}N^T (F_N(t))^T, \quad t = 0, 1, 2, \ldots, t_f - 1,
\end{equation}

and

\begin{equation}
(4.110) \quad ((f^0_N)^T, (f^0_N(\cdot))^T) = E_N^{-1}N^T (F_N)^T
\end{equation}

respectively.

For the approximation scheme defined above, it is shown in [14] that $P_N \rightarrow I$ strongly on $Z$. Using a Trotter-Kato like result it is also shown that

\begin{equation}
(4.111) \quad \exp (A_N s) P_N \rightarrow T(s)
\end{equation}

and

\begin{equation}
(4.112) \quad \exp(A_N^* s) P_N \rightarrow T^*(s)
\end{equation}

strongly on $Z$ and uniformly in $s$ for $s$ in compact intervals. Hypothesis 3.1 is a simple consequence of these results. The present scheme, therefore, satisfies all of the hypotheses of Theorem 3.3 and we may conclude that the convergence results for the finite time interval problem given in the statement of the theorem hold. In particular, we have

\begin{equation}
(4.113) \quad f^0_N(t) \rightarrow f^0(t)
\end{equation}
in $\mathbb{R}^{m \times n}$ and

\begin{equation}
(4.114) \quad f_{N}^{1}(t, \cdot) + f^{1}(t, \cdot)
\end{equation}

in $L_{2}((-r,0); \mathbb{R}^{m \times n})$ for each $t = 0, 1, 2, \ldots t_{f}-1$.

With the operators $Q$ and $G$ given by (4.73) and (4.74) and the operators $Q_{N}$ and $G_{N}$ defined as in (4.102) and (4.103) it is clear that the hypotheses given in the statement of Theorem 3.7 are satisfied. We have therefore that for the present example the operators $\{\Pi(t)\}_{t=0}^{t_{f}}$ are trace class and

\begin{equation}
(4.115) \quad \lim_{N \to \infty} \|\Pi_{N}(t)P_{N} - \Pi(t)\|_{1} = 0, \quad t = 0, 1, 2, \ldots, t_{f}.
\end{equation}

For the infinite time problem and the approximation scheme discussed here, the situation with regard to convergence is much the same as it is for the continuous time problem (see [14]). We are unable to demonstrate the existence of an $M$ and an $r < 1$ for which (3.38) and (3.39) hold. In fact, our numerical studies point to the conclusion that condition (3.39) is violated by the present scheme. We observe the existence of a sequence of closed-loop eigenvalues of the approximating discrete-time control problems $(P_{2N})$ which tend toward the unit circle as $N \to \infty$.

On the other hand, upon solving the approximating problems it is also apparent that $|\Pi_{N}|$ remains bounded in $N$. Consequently we may apply Theorem 3.12 to conclude that a solution $\Pi$ to (2.14) does in fact exist, $\Pi_{N}P_{N} \to \Pi$ weakly and $F_{N}P_{N} \to F$ strongly as $N \to \infty$. We have therefore that

\begin{equation}
(4.116) \quad f_{N}^{0} \to f^{0}
\end{equation}
in $\mathbb{R}^{m \times n}$ and

(4.117) \[ f_N^1 + f^1 \]

weakly in $L_2((-r,0);\mathbb{R}^{m \times n})$ as $N \to \infty$.

We applied the scheme to the infinite-time problem with state

(4.118) \[ \ddot{y}(s) + y(s-1) = v(s). \]

Transforming (4.118) to an equivalent first order system we obtain a system of the form given in (4.60), (4.61) with $n = 2$, $r = 1$, $m = 1$,

\[
A_0 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad A_1 = \begin{bmatrix} 0 & 0 \\ -1 & 0 \end{bmatrix}, \quad B_0 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad A = 0 \text{ and } x(s) = \begin{bmatrix} y(s) \\ \dot{y}(s) \end{bmatrix}.
\]

Taking $Q_0 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ the performance index takes the form

(4.119) \[ J(0;0,\infty,y(0),\dot{y}(0),y_0,\dot{y}_0,u) = \sum_{t=0}^{\infty} y(t\tau)^2 + \dot{y}(t\tau)^2 + Ru(t)^2 \]

where $\tau$ is the length of the sampling interval. The optimal feedback control is given by

(4.120) \[ u_*(t) = -[f^0_1]_1 y(t\tau) -[f^0_2]_2 \dot{y}(t\tau) \]

\[ - \int_{-1}^{0} ([f^1(\theta)]_1 y(t\tau+\theta) + [f^1(\theta)]_2 \dot{y}(t\tau+\theta)) \, d\theta \]

where $[f^0_i]$ and $[f^1(\theta)]_i$, $i = 1, 2$ are the $i$th components of the $1 \times 2$ matrices $f^0$ and $f^1(\theta)$. 
Since by taking the initial conditions

\[(4.121)\quad y(0) = 0, \quad \dot{y}(0) = 0, \quad y_0(0) = 0, \quad -1 < \theta < 0\]

we have \(y(s) = 0, \quad s > 0\) regardless of how \(\dot{y}_0(\theta), \quad -1 < \theta < 0\) is chosen, it follows that \([f^1(\theta)]_2 = 0, \quad -1 < \theta < 0\). Indeed, the optimal control corresponding to the initial conditions \((4.121)\) with \(\dot{y}_0\) arbitrary is \(u(t) = 0, \quad t = 0,1,2,...\). Furthermore, the nature of the approximation scheme is such that we must have \([f^1_N(\theta)]_2 = 0, \quad -1 < \theta < 0, \quad N > 1\).

Setting \(\tau = .01\) we obtained the results given in Tables 4.4 and 4.5 and Figure 4.8 below when \(R = .05\). With \(R = 1.0\), the results given in Tables 4.6 and 4.7 and Figure 4.9 were obtained. As the cost of control increases the effect that the optimal control for the infinite dimensional problem has on higher modes decreases. Consequently, the finite dimensional approximations are more effective and convergence is more rapid.
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Table 4.4

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Table 4.5
### Table 4.6

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### Table 4.7

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Figure 4.8

Figure 4.9
4.3 Control of a Flexible Structure

We consider an Euler-Bernoulli beam cantilevered to a rigid hub which is free to rotate about its fixed center, point O. Also, a point mass \( m_1 \) is attached to the other end of the beam. The control is a torque \( u \) applied to the hub, and all motion is in the plane. See Figure 4.10 and Table 4.11.
r = hub radius  
\ell = beam length  
I_0 = hub moment of inertia about axis perpendicular to page through 0  
m_b = beam mass per unit length  
m_1 = tip mass  
EI = product of elastic modulus and second moment of cross section for beam  
fundamental frequency of undamped structure

Table 4.11

The angle $\theta$ represents the rotation of the hub (the rigid-body mode), $w(t,n)$ is the elastic deflection of the beam from the rigid-body position, and $w_1(t)$ is the displacement of $m_1$ from the rigid-body position. For technical reasons, we do not yet impose the condition $w_1(t) = w(t,\ell)$.

The control problem is to stabilize rigid-body motions and linear (small) transverse elastic vibrations about the state $\theta = 0$ and $w = 0$. Our linear model assumes not only that the elastic deflection of the beam is linear but also that the axial inertial force produced by the rigid-body angular velocity has negligible effect on the bending stiffness of the beam. The rigid-body angle need not be small.

For this example, it is a straightforward exercise to derive the coupled ordinary and partial differential equations of motion in $\theta$, $w$ and $w_1$. However, rather than writing these equations explicitly, it is easier and more useful
for our purposes to derive an abstract second order evolution equation for the structure. To do this, we define the generalized displacement vector

\[(4.122) \quad x = (\theta, w, w_1) \in H = \mathbb{R}^1 \times L^2(0,\xi) \times \mathbb{R}^1.\]

The kinetic energy in the system is then

\[(4.123) \quad \text{Kinetic Energy} = \frac{1}{2} \langle M_0 \dot{x}, \dot{x} \rangle_H \]

where \(M_0\) is the unique bounded self-adjoint linear mass operator \(M_0\) on \(H\) such that

\[(4.124) \quad \langle M_0 x, x \rangle = I_0 \theta + m_b \langle w + \psi_0 \theta, w + \psi_0 \theta \rangle_{L^2} + m_1 \langle w_1 + \psi_0(\xi) \theta, w_1 + \psi_0(\xi) \theta \rangle_{L^2},\]

where \(\psi_0 \in L^2(0,\xi)\) is given by \(\psi_0(\eta) = r + \eta\). It is easy to show that \(M_0\) is also coercive. The elastic strain energy is

\[(4.125) \quad \text{Strain Energy} = \frac{1}{2} a(x,x) \]

with

\[(4.126) \quad a(x,\hat{x}) = EI \langle D^2 w, D^2 \hat{w} \rangle_{L^2}.\]

We make \(a(\cdot,\cdot)\) into an inner product by setting

\[(4.127) \quad \langle x, \hat{x} \rangle_V = a(x,\hat{x}) + \theta \hat{\theta}.\]
and define the strain-energy space

\[(4.128) \quad V = \{ x = (\theta, \phi, \phi(\xi)) : \phi \in H^2(0,\xi), \phi(0) = D\phi(0) = 0 \} \]

The last term in (4.127) is necessary for the \(V\)-inner product because there is no strain energy associated with the rotation of the hub.

We define the stiffness operator \(A_0\) by

\[(4.129) \quad \text{Dom}(A_0) = \{ x = (\theta, \phi, \phi(\xi)) \in V : \phi \in H^4(0,\xi), D^2\phi(\xi) = 0 \}\]

and

\[(4.130) \quad A_0 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & EI & D^4 \\ 0 & -EI D^3 & 0 \end{bmatrix} \]

This operator is self-adjoint with compact resolvent and all positive eigenvalues except the one zero eigenvalue corresponding to the rigid-body mode. Note that \(V\) is the domain of the square root of \(A_0\).

With these mass and stiffness operators, we can write the equations of motion as

\[(4.131) \quad M_0\ddot{x}(s) + c_0 A_0 \dot{x}(s) + A_0 x(s) = B_0 u(s), \quad s > 0,\]

where \(c_0\) is a positive constant and the term \(c_0 A_0 \dot{x}\) represents viscoelastic damping in the beam. The input operator is
(4.132) \[ B_0 = (1,0,0). \]

Letting \( Z = V \times H \) with inner product \( \langle (v,h),(v',h') \rangle_Z = \langle v,v \rangle_V + \langle M_0 h,h \rangle_H \),
the first order form of this system is given by

(4.133) \[ \dot{z}(s) = A z(s) + B u(s), \quad s > 0, \]

where \( z = (x,x) \in Z \) and \( A \) is the unique extension of the operator

(4.134) \[ A = \begin{bmatrix} 0 & I \\ -M_0^{-1} A_0 & -M_0^{-1} A_0 \end{bmatrix}, \quad \text{Dom}(A) = \text{Dom}(A_0) \times \text{Dom}(A_0), \]

that generates a \( C_0 \)-semigroup on the space \( Z \). Of course, \( B \) is

(4.135) \[ B = \begin{bmatrix} 0 \\ M_0^{-1} B_0 \end{bmatrix}. \]

See [10] and [12]. The hub-beam-tip mass structure here is discussed in more
detail in [12], along with the continuous-time problem.

The discrete-time control system for sampling interval \( \tau \) is

(4.136) \[ z(t+1) = Tz(t) + Bu(t), \quad t = 0,1,2,\ldots, \]

where

(4.137) \[ T = T(\tau), \quad B = \int_0^\tau T(s)B \, ds \]

and \( \{T(s): s > 0\} \) is the semigroup generated by the \( A \) in (4.133).
As in the previous examples, we will solve a discrete-time optimal control problem on the infinite interval. In the performance index, we take the state weighting operator $Q$ to be the identity on $Z$. This means that $\langle Qz,z \rangle_Z$ is twice the total energy in the structure plus the square of the rigid-body rotation. Since there is one input, the control weighting $R$ is a scalar. The optimal control has the feedback form

\begin{equation}
(4.138) \quad u^*_x(t) = -\langle f, x(t) \rangle_V - \langle M_0 g, x(t) \rangle_H
\end{equation}

where $x(t)$ has the form (4.122) and

\begin{equation}
(4.139) \quad f = (f^1, f^2, f^3) \in V
\end{equation}

\begin{equation}
(4.140) \quad g = (g^1, g^2, g^3) \in H.
\end{equation}

Our approximation of the structure is based on a finite element approximation of the beam which uses Hermite cubic splines as basis functions ([27]). We define the sequence of spaces $V_N = \text{span} \{ e^j_N \}_{j=1}^{J_N}$ with

\begin{equation}
(4.141) \quad e^1_N = (1,0,0),
\end{equation}

\begin{equation}
(4.142) \quad e^j_N = (0, \phi^j_N, \phi^j_N(\&)), \quad j = 2,3,\ldots, J_N,
\end{equation}

where the $\phi^j_N$'s are the cubic splines. Each $V_N$ is a subspace of $V$, and our approximation scheme is a Ritz-Galerkin approximation obtained by projecting (4.131) onto $V_N$. See [12] for details. Writing
(4.143) \[ x_N(s) = \sum_{j=1}^{J_N} [x_N(s)]_j e_N^j, \]

we have

(4.144) \[ M_N [\ddot{x}_N(s)] + c_0 K_N [\dot{x}_N(s)] + K_N [x_N(s)] = B_{ON} u(s) \]

to solve for the vector \([x_N(s)]_j\) of time-dependent coefficients \([x_N(s)]_j\). The mass matrix \(M_N\) and the stiffness matrix \(K_N\) are given by

(4.145) \[ [M_N]_{ij} = \langle M_0 e_N^i, e_N^j \rangle_H, \quad [K_N]_{ij} = \langle e_N^i, e_N^j \rangle_V \]

and the input matrix is

(4.146) \[ B_{ON} = [1 0 0 ... 0]^T. \]

With \(z_N = (x_N, \dot{x}_N) \in V_N \times V_N\), (4.144) is the matrix representation of an evolution equation

(4.147) \[ \dot{z}_N(s) = A_N z_N(s) + B_N u(s) \]

where \(A_N\) and \(B_N\) approximate \(A\) and \(B\). It is shown in [12] that, as \(N\) increases, the semigroup \(\{T_N(s) : s > 0\}\) generated by \(A_N\) converges strongly to the semigroup \(\{T(s) : s > 0\}\) and that the adjoint semigroup \(\{T^*_N(s) : s > 0\}\) converges strongly as well. Since \(B_N\) is the Z-projection of \(B\) onto \(V_N \times V_N\), it converges strongly to \(B\).

For the approximating discrete-time control systems, we replace \(z(t), T,\)
B, T(·) and B in (4.136) and (4.137) with \( z_N, T_N, B_N, T_N(\cdot) \) and \( B_N \), respectively. For each \( N \), the solution to the infinite-time optimal control problem is based on the \( N^{th} \) Riccati operator equation (3.33). As in the previous examples, we solve the Riccati matrix equation (3.59) for \( \hat{\Pi}_N \), which is related to \([\Pi_N]\) (the matrix representation of the operator \( \Pi_N \)) as in Section 3.3, except here we have

\[
\hat{\Pi}_N = W_N [\Pi_N],
\]

where

\[
W_N = \begin{bmatrix}
\tilde{K}_N & 0 \\
0 & M_N
\end{bmatrix}
\]

and \( \tilde{K}_N \) is the stiffness matrix with 1 added to the first element. Since \( Q = I \) in the infinite dimensional problem, \( Q_N \) is the identity on \( V_N \times V_N \) and it follows from (3.48) that the matrix \( \hat{Q}_N \) for (3.59) is \( W_N \).

The optimal feedback control for the \( N^{th} \) problem is given by (3.56) with the matrices in (3.57) and (3.58), and it has the equivalent representation

\[
u^N_{\alpha}(t) = - \langle f_N, x_N(t) \rangle_V - \langle M_0 g_N, x_N(t) \rangle_H
\]

where

\[
f_N = (f_N^1, f_N^2, f_N^3) \in V,
\]

\[
g_N = (g_N^1, g_N^2, g_N^3) \in H
\]
as in (4.139) and (4.140). From (3.56), (4.143) and (4.150), it follows that

\[
\begin{bmatrix}
    f_N \\
    g_N
\end{bmatrix}
= 
\begin{bmatrix}
    E_N^T & 0 \\
    0 & E_N^T
\end{bmatrix} W^{-1} \begin{bmatrix} F_N \end{bmatrix}^T,
\]

where \( E_N^T = (e_1^N, e_2^N, \ldots, e_J^N) \).

For the sampling interval \( \tau = 0.01 \), the damping coefficient \( c_0 = 0.001 \) and the control weighting \( R = 1 \), Tables 4.12-4.15 give the values of the corresponding scalar and functional gains, \( f_N^i, g_N^i, i = 1, 2, 3 \) for various values of \( N \). The values of the functional gains \( D^2 f_N^2 \) and \( g_N^2 \) along the length of the beam also are plotted in Figures 4.16 and 4.17. We tabulated and plotted \( D^2 f_N^2 \) because this is what appears in the \( V \) inner product in (4.150) and also to show the \( H^2 \) convergence. We note that the form of the \( V \) inner product given by (4.127) is such that \( f_N^3 \) does not appear in the feedback law.
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Table 4.12

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Table 4.15
5. Concluding Remarks

We have presented an approximation theory for numerical solution of the discrete-time optimal linear regulator problem in Hilbert space, on both finite and infinite time intervals. The motivation for this theory comes from optimal control problems for systems involving diffusion equations, hereditary differential equations and distributed models of flexible structures. We have demonstrated the application of the theory to examples from all three areas.

The solution to the infinite dimensional optimal control problem is based on an infinite dimensional Riccati operator equation — a difference equation in the finite-time problem and an algebraic equation in the infinite-time problem. We have shown that the solution to the infinite dimensional problem can be approximated by the solutions to a sequence of finite dimensional problems each of which involves a finite dimensional Riccati matrix equation to be solved numerically. The finite dimensional problems are just the corresponding optimal control problems for finite element approximations to the infinite dimensional control system. For the infinite-time problem, the finite dimensional Riccati equations usually are solved via eigenspace decomposition of the Hamiltonian matrix.

In both continuous and discrete-time optimal regulator problems for distributed systems, the numerical solution often involves solution of large Riccati matrix equations. As we observed at the beginning of Section 4, the asymptotic relationship between the eigenvalues of a continuous-time Hamiltonian system and the eigenvalues of the corresponding discrete-time Hamiltonian system is exponential. This means that the approximating finite dimensional discrete-time Riccati equations for a given distributed system invariably are not as well conditioned as the corresponding continuous-time Riccati equations. Nonetheless, as our examples should illustrate, the
numerical solution of such problems is well within the reach of current computing. To emphasize this, we obtained all of the numerical results in this paper on an IBM Personal Computer (not an XT or AT) with 640K of random access memory and an Intel 8087 math coprocessor chip. The largest Riccati matrix equation that we solved here was a $30 \times 30$ steady state equation for the hub-beam-tip mass example. This solution takes 15 to 20 minutes on the PC. We have solved much larger Riccati equations easily on larger mainframe computers.
REFERENCES


An abstract approximation framework is developed for the finite and infinite time horizon discrete-time linear-quadratic regulator problem for systems whose state dynamics are described by a linear semigroup of operators on an infinite dimensional Hilbert space. The schemes included in the framework yield finite dimensional approximations to the linear state feedback gains which determine the optimal control law. Convergence arguments are given. Examples involving hereditary and parabolic systems and the vibration of a flexible beam are considered. Spline-based finite element schemes for these classes of problems, together with numerical results, are presented and discussed.