



## Technical communique

Optimal ripple-free deadbeat control using an integral of time squared error (ITSE) index<sup>☆</sup>Francisco J. Vargas<sup>\*</sup>, Mario E. Salgado, Eduardo I. Silva

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## ABSTRACT

This paper describes an optimal ripple-free deadbeat control strategy for single-input–single-output (SISO) linear sampled data plants. The cost function to be minimized is a linear combination of a time-weighted cumulative term that penalizes the tracking error, that is, an integral of time squared error (ITSE) cost term, and a cumulative term which penalizes the control signal deviations from its steady-state value. The optimization problem turns out to be convex, and closed-form solutions are obtained. An example is included to illustrate our results.

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## 1. Introduction

In optimal control, the choice of the performance index is a key issue, since it defines the quality measure used to assess controller performance. Commonly used time-domain indices include the ISE (integral<sup>1</sup> squared error) index, which is related to the 2-norm, the IAE (integral of the absolute error) index, related to the 1-norm, and the ITSE (integral of time squared error) index (Duarte-Mermoud & Prieto, 2004; Ogata, 1997).

The ITSE index has been used in connection with the tuning of proportional–integral–derivative (PID) controllers (Ogata, 1997). A deeper treatment of this index can be found in Carrasco and Salgado (2009), where a set of analytical tools for the optimal design of controllers for stable discrete-time single-input–single-output (SISO) plants is presented. Also, Barbargires and Karybakas (1994) presents optimal deadbeat controller design for discrete-time SISO plants based on a time-weighted performance index. However, ripple-free behaviour is not considered in that work.

This article deals with the design of controllers that, for step references, achieve ripple-free deadbeat (RFDB) control for sampled-data plants and, at the same time, minimize the time-weighted cost function

$$J = \lambda J_e + (1 - \lambda) J_u, \quad (1)$$

where  $\lambda \in [0; 1]$  is a weighting parameter, and

$$J_e = \sum_{k=0}^{\infty} k e(k)^2, \quad J_u = \sum_{k=0}^{\infty} (u(k) - u_{ss})^2. \quad (2)$$

In (2),  $e(k)$  denotes the tracking error,  $u(k)$  the control input, and  $u_{ss}$  the steady-state value of  $u(k)$ . Due to the time weighting of  $e(k)$ , we have that, for  $J_e$  to become small, the tracking error must converge to zero rapidly. This property suggests that the unconstrained optimization of  $J_e$  will have a negative impact on the control effort. The presence of  $J_u$  allows one to deal with that problem by choosing an appropriate value for  $\lambda$ .

The main contribution of this article is finding the optimal RFDB controller that minimizes the index  $J$  in (1). A numerical example is provided to illustrate our results.

## 2. Ripple-free deadbeat control

Deadbeat is a well-known technique for the design of discrete-time control systems, whose purpose is to perfectly track a reference in the minimum number of sampling periods (see, e.g., Emami-Naein & Franklin, 1982; Goodwin, Graebe, & Salgado, 2001; Salgado, Oyarzún, & Silva, 2007). The number of sampling

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<sup>1</sup> In this paper we work in discrete-time. Therefore integration must be interpreted as a cumulative sum.

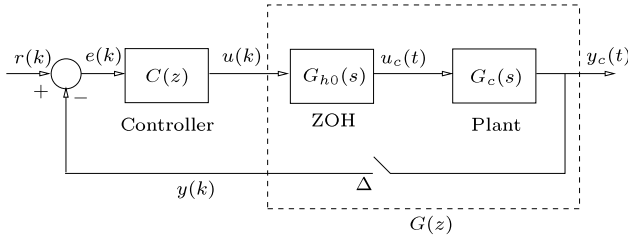


Fig. 1. Standard one-degree-of-freedom sampled-data control loop.

times required for the tracking error to converge to zero is known as the *deadbeat horizon*. In the context of sampled-data systems, deadbeat control should avoid intersample ripple in the continuous-time output of the plant. This requirement has motivated the development of techniques for designing RFDB controllers (see, e.g., Casavola, Mosca, & Zecca, 1999; Nobuyama, 1993; Paz, 1999). The optimal design of RFDB controllers has been already addressed using  $l_\infty$ ,  $l_1$ , and  $\mathcal{H}_2$  norms (see, e.g., Casavola et al., 1999; Salgado & Oyarzún, 2007; Salgado et al., 2007).

Fig. 1 shows the sampled-data control loop considered in this paper. The goal is to track a step reference  $r(k) = v\mu(k)$ , achieving  $y_c(t) = v, \forall t \geq \eta\Delta$ , and a control signal such that  $u(k) = u_{ss}$  (a constant),  $\forall k \geq \eta$ . Here,  $\eta \in \mathbb{N}_0$  is the deadbeat horizon. For simplicity,  $v = 1$  is assumed. The continuous-time model of the plant is assumed to have the form  $G_c(s) = \frac{B_c(s)}{A_c(s)}e^{-s\tau}$ , with  $\tau \geq 0$ , where  $B_c(s)$  and  $A_c(s)$  are coprime polynomials in  $s$ . The sampled data representation of  $G_c(s)$ , using a zero-order hold, is given by  $G(z) = \frac{B(z)}{A(z)}$ , where  $B(z)$  and  $A(z)$  are coprime polynomials in  $z$  of degrees  $n$  and  $m$ , respectively, with  $n \geq m$ . Without loss of generality, we assume that  $B(1) = 1$ . On the other hand,  $A(z)$  is factorized as  $A(z) = A_+(z)A_-(z)$ , where the polynomial  $A_+(z)$ , of degree  $n_+$ , contains all roots of  $A(z)$  in  $\{z \in \mathbb{C} : |z| \geq 1\}$ , while  $A_-(z)$ , of degree  $n_-$ , contains all roots in  $\{z \in \mathbb{C} : |z| < 1\}$ .

To achieve an RFDB response, it is necessary that the synthesized controller does not cancel the minimum-phase zeros of  $G(z)$  (Goodwin et al., 2001; Salgado et al., 2007). The solution to the problem thus requires that the complementary sensitivity  $T(z)$  and the control sensitivity  $S_u(z)$  are finite impulse response (FIR) transfer functions of order  $\eta$  (Karybakas & Barbargires, 1994; Sirisena, 1985). This ensures that both  $e(k)$  and  $u(k)$  settle to constant values in the specified deadbeat horizon. Also, by requiring that  $T(1) = 1$ , we guarantee zero steady-state error. On the other hand, the minimum deadbeat horizon to get RFDB control is given by  $\eta_{\min} = n + n_+$  (Nobuyama, 1993), and, hence, any arbitrary horizon can be written as  $\eta = \eta_{\min} + \ell$ , with  $\ell \in \mathbb{N}_0$ . When  $\ell = 0$ , there exists only one controller, say  $C_0(z)$ , providing an RFDB response. That controller satisfies

$$A(z)L_0(z) + B(z)P_0(z) = z^{\eta_{\min}}A_-(z), \quad (3)$$

where  $C_0(z) = P_0(z)/L_0(z)$  is biproper with  $L_0(z)$  of order  $n$  such that  $L_0(1) = 0$ . Note from (3) that  $P_0(z) = \tilde{P}_0(z)A_-(z)$ , with  $\tilde{P}_0(1) = 1$ . It is known that every stabilizing RFDB controller  $C(z)$  can be written as Salgado et al. (2007)

$$C(z) = \frac{P_0(z) + X(z)A(z)}{L_0(z) - X(z)B(z)}, \quad (4)$$

where  $X(z)$  is an FIR transfer function of order  $\ell$  such that  $X(1) = 0$ . Thus,  $X(z) = (z-1)\tilde{D}(z)z^{-\ell}$ , where  $\tilde{D}(z)$  is a polynomial of order  $\ell - 1$  or less. The optimal design of an RFDB controller amounts to finding the polynomial  $\tilde{D}(z)$  that minimizes  $J$ .

### 3. Optimal designs

This section presents a technique to design optimal controllers using the index defined in (1). By using Parseval's theorem Goodwin et al. (2001), we can write  $J_e$  and  $J_u$  in (2) as follows:

$$J_e = -\frac{1}{2\pi j} \oint \frac{dE(z)}{dz} E(z^{-1}) dz, \quad (5)$$

$$J_u = \frac{1}{2\pi j} \oint \frac{1}{z} F(z)F(z^{-1}) dz, \quad (6)$$

where  $E(z) = S(z)R(z)$  is the  $\mathcal{Z}$  transform of the tracking error, and  $F(z) = S_u(z)R(z) - u_{ss}\frac{z}{z-1}$  is the  $\mathcal{Z}$  transform of  $u(k) - u_{ss}$ . Furthermore, given that  $r(k) = \mu(k)$ , we can proceed as in Salgado et al. (2007) to write  $E(z) = M_e(z) - N_e(z)\tilde{D}(z)$  and  $F(z) = M_u(z) - N_u(z)\tilde{D}(z)$ , where

$$M_e(z) = \frac{z^{\eta_{\min}} - B(z)\tilde{P}_0(z)}{z^{\eta_{\min}-1}(z-1)} = \sum_{i=0}^{\eta_{\min}-1} m_{e_i}z^{-i}, \quad (7)$$

$$N_e(z) = \frac{B(z)A_+(z)}{z^{\eta-1}} = \sum_{i=0}^{\eta-1} n_{e_i}z^{-i}, \quad (8)$$

$$M_u(z) = \frac{A(z)\tilde{P}_0(z) - A(1)z^{\eta_{\min}}}{z^{\eta_{\min}-1}(z-1)} = \sum_{i=0}^{\eta_{\min}-1} m_{u_i}z^{-i}, \quad (9)$$

$$N_u(z) = -\frac{A(z)A_+(z)}{z^{\eta-1}} = \sum_{i=0}^{\eta-1} n_{u_i}z^{-i}, \quad (10)$$

and  $m_{e_i}$ ,  $n_{e_i}$ ,  $m_{u_i}$ , and  $n_{u_i}$  are real coefficients (note that  $\tilde{P}_0(1) = 1$ ,  $B(1) = 1$ , and  $m \leq n$  implies that  $M_e(z)$ ,  $N_e(z)$ ,  $M_u(z)$ , and  $N_u(z)$  are FIR transfer functions). To further simplify the problem formulation, we define

$$\boldsymbol{\zeta}(z) \triangleq [z^0 \ \dots \ z^{\ell-1}]^T, \quad \tilde{\mathbf{d}} \triangleq [\tilde{d}_0 \ \dots \ \tilde{d}_{\ell-1}]^T. \quad (11)$$

Given that  $\tilde{D}(z) = \boldsymbol{\zeta}(z)^T \tilde{\mathbf{d}}$ , the problem of finding the optimal polynomial  $\tilde{D}(z)$  reduces to finding the optimal vector of coefficients  $\tilde{\mathbf{d}}$ . The costs  $J_e$  and  $J_u$  can be written as functions of  $\tilde{\mathbf{d}}$ , as

$$J_e(\tilde{\mathbf{d}}) = T_e - \tilde{\mathbf{d}}^T \cdot \mathbf{K}_e + \tilde{\mathbf{d}}^T \cdot \mathbf{L}_e \cdot \tilde{\mathbf{d}}, \quad (12)$$

$$J_u(\tilde{\mathbf{d}}) = T_u - \tilde{\mathbf{d}}^T \cdot \mathbf{K}_u + \tilde{\mathbf{d}}^T \cdot \mathbf{L}_u \cdot \tilde{\mathbf{d}}, \quad (13)$$

where  $T_e, T_u \in \mathbb{R}$ ,  $\mathbf{K}_e, \mathbf{K}_u \in \mathbb{R}^{\ell \times 1}$ , and  $\mathbf{L}_e, \mathbf{L}_u \in \mathbb{R}^{\ell \times \ell}$  are given by

$$T_e = -\frac{1}{2\pi j} \oint \frac{dM_e(z)}{dz} M_e(z^{-1}) dz, \quad (14)$$

$$T_u = \frac{1}{2\pi j} \oint \frac{1}{z} M_u(z)M_u(z^{-1}) dz, \quad (15)$$

$$\mathbf{K}_e = -\frac{1}{2\pi j} \oint \left( M_e(z^{-1}) \frac{d[N_e(z)\boldsymbol{\zeta}(z)]}{dz} + \frac{dM_e(z)}{dz} N_e(z^{-1})\boldsymbol{\zeta}(z^{-1}) \right) dz, \quad (16)$$

$$\mathbf{K}_u = \frac{1}{2\pi j} \oint \frac{1}{z} (M_u(z)N_u(z^{-1})\boldsymbol{\zeta}(z^{-1}) + M_u(z^{-1})N_u(z)\boldsymbol{\zeta}(z)) dz, \quad (17)$$

$$\mathbf{L}_e = -\frac{1}{2\pi j} \oint \frac{d[\boldsymbol{\zeta}(z)N_e(z)]}{dz} N_e(z^{-1})\boldsymbol{\zeta}(z^{-1})^T dz, \quad (18)$$

$$\mathbf{L}_u = \frac{1}{2\pi j} \oint \frac{1}{z} \boldsymbol{\zeta}(z)N_u(z)N_u(z^{-1})\boldsymbol{\zeta}(z^{-1})^T dz. \quad (19)$$

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