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A DIAGONALIZATION TECHNIQUE FOR THE  
COMPUTATION OF SENSITIVITY FUNCTIONS  
OF LINEAR TIME-INVARIANT SYSTEMS

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# A Diagonalization Technique for the Computation of Sensitivity Functions of Linear Time-Invariant Systems

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## ABSTRACT

This note presents a method for computing the sensitivity functions of parametrized linear time-invariant systems, for the case where the system matrix  $A$  is diagonalizable. The method is based on a formula, derived in the note, for the sensitivity of an exponential of a diagonalizable matrix.

## 1. Introduction

When designing a linear control system by semi-infinite optimization techniques, one is required to compute time and frequency domain responses as well as their sensitivities to the designable compensator parameters, see, e.g., [Pol.1]. While there is a considerable literature on the solution of state equations and the evaluation of frequency responses, see, e.g., [Lau.1, Lau.2], there are hardly any results available dealing with the efficient computation of response sensitivities.

This paper generalizes the unpublished results in [Bec.1] to obtain an efficient method for computing time domain response sensitivities for an important class of problems. Our method results from the following observations.

Consider a parametrized linear time-invariant system whose dynamics are given by:

$$\dot{x}(t,p) = A(p)x(t,p) + B(p)u(t), \quad (1.1a)$$

$$y(t,p) = C(p)x(t,p) + D(p)u(t), \quad (1.1b)$$

where  $x(t,p) \in \mathbb{R}^n$  is the state,  $p \in \mathbb{R}^N$  is the design parameter vector,  $A: \mathbb{R}^N \rightarrow \mathbb{R}^{n \times n}$ ,  $B: \mathbb{R}^N \rightarrow \mathbb{R}^{n \times k}$ ,  $C: \mathbb{R}^N \rightarrow \mathbb{R}^{m \times n}$  and  $D: \mathbb{R}^N \rightarrow \mathbb{R}^{m \times k}$  are continuously differentiable matrices, and the input  $u: \mathbb{R} \rightarrow \mathbb{R}^k$  is of the form  $u(t) = \sum_{i=1}^m t^{\alpha_i} e^{\lambda_i t} c_i$ , where the  $c_i \in \mathbb{R}^k$ , the  $\alpha_i$  are nonnegative integers and the  $\lambda_i \in \mathbb{C}$ .

Since the solution of (1.1a) is given by

$$x(t,p) = e^{A(p)t} x(0,p) + \int_0^t e^{A(p)(t-\tau)} B(p) u(\tau) d\tau, \quad (1.2a)$$

its partial derivatives are given by

$$\frac{\partial x(t,p)}{\partial p^i} = \frac{\partial e^{A(p)t}}{\partial p^i} x(0,p) + e^{A(p)t} \frac{\partial x(0,p)}{\partial p^i} + \int_0^t e^{A(p)(t-\tau)} \frac{\partial B(p)}{\partial p^i} u(\tau) d\tau + \int_0^t e^{A(p)(t-\tau)} B(p) \frac{\partial u(\tau)}{\partial p^i} d\tau, \quad (1.2b)$$

$$\int_0^t \frac{\partial}{\partial p^i} (e^{A(p)(t-\tau)}) B(p) u(\tau) d\tau + \int_0^t e^{A(p)(t-\tau)} \frac{\partial B(p)}{\partial p^i} u(\tau) d\tau,$$

for all  $i \in \underline{N} \triangleq \{1, 2, \dots, N\}$ . When the matrix  $A(p)$  is diagonalizable and the input  $u(t)$  is componentwise polynomial, (1.2a) represents a viable method for computing the state response. In this case, the efficient computation of partial derivatives  $\frac{\partial x(t, p)}{\partial p^i}$  requires an efficient method for computing  $\frac{\partial e^{A(p)t}}{\partial p^i}$ . In [Bec.1] we find such a technique, based on Lie bracket decompositions, for the case where  $A(p)$  has distinct eigenvalues. In this paper, the results in [Bec.1] are extended to include the case where the matrix  $A(p)$  has repeated eigenvalues, but is diagonalizable. We shall comment in the conclusion on possible ways of dealing with the nondiagonalizable case.

## 2. A Formula for the Matrix $\frac{\partial e^{A(p)t}}{\partial p^i}$

If we are to use (1.2b) to compute  $\frac{\partial x(t, p)}{\partial p^i}$ , we must evaluate  $\frac{\partial e^{A(p)t}}{\partial p^i}$ .

**Proposition 1:** Let  $A(p)$  be defined as in equation (1.1a), then

$$\frac{\partial e^{A(p)t}}{\partial p^i} = e^{A(p)t} \int_0^t e^{-A(p)\tau} \frac{\partial A(p)}{\partial p^i} e^{A(p)\tau} d\tau. \quad (2.1)$$

**Proof:** By assumption,  $A(\cdot)$  is continuously differentiable. Hence  $e^{A(p)t}$  is continuously differentiable in  $(p, t)$ , and it therefore follows that (see [Mar.1])

$$\begin{aligned} \frac{d}{dt} \left[ \frac{\partial e^{A(p)t}}{\partial p^i} \right] &= \frac{\partial}{\partial p^i} \left[ \frac{d e^{A(p)t}}{dt} \right] = \frac{\partial}{\partial p^i} \left[ A(p) \cdot e^{A(p)t} \right] \\ &= A(p) \cdot \frac{\partial e^{A(p)t}}{\partial p^i} + \frac{\partial A(p)}{\partial p^i} e^{A(p)t} \end{aligned} \quad (2.2)$$

Integrating the linear differential equation (2.2) from 0 to  $t$ , we obtain that

$$\frac{\partial e^{A(p)t}}{\partial p^i} = e^{A(p)t} \left[ \frac{\partial e^{A(p)t}}{\partial p^i} \right]_{t=0} + e^{A(p)t} \int_0^t e^{-A(p)\tau} \frac{\partial A(p)}{\partial p^i} e^{A(p)\tau} d\tau \quad (2.3)$$

Since  $\left[ \frac{\partial e^{A(p)t}}{\partial p^i} \right]_{t=0} = 0$ , (2.1) follows directly.

The evaluation of  $e^{A(p)t}$  in (2.1) is relatively easy and can be carried out by some of the methods described in [Mol.1], [Par.1], [Lau.1]. In general, the evaluation of the second term,  $\int_0^t e^{-A(p)\tau} \frac{\partial A(p)}{\partial p^i} e^{A(p)\tau} d\tau$ , in (2.1) is more problematic. However, the following two observations lead to the conclusion that there may be cases where this term can be computed without resorting to numerical integration.

(a) Let  $V \in \mathbb{R}^{n \times n}$  be such that  $A(p)V - V A(p) = 0$ . Then  $e^{A(p)t} V = V e^{A(p)t}$  and

$$\int_0^t e^{-A(p)\tau} V e^{A(p)\tau} d\tau = tV. \quad (2.4)$$

(b) For any  $U \in \mathbb{R}^{n \times n}$ ,

$$e^{-A(p)t} \{A(p)U - UA(p)\} e^{A(p)t} = - \frac{d}{dt} (e^{-A(p)t} U e^{A(p)t}). \quad (2.5)$$

**Proposition 2 :** Let  $A(p)$  be defined as in equation (1.1a). For any  $p \in \mathbb{R}^N$ , if there exist  $V \in \mathbb{R}^{n \times n}$  and  $U \in \mathbb{R}^{n \times n}$ , such that

$$\frac{\partial A(p)}{\partial p^i} = V + \{A(p)U - UA(p)\}, \quad (2.6a)$$

$$A(p)V - V A(p) = 0, \quad (2.6b)$$

then

$$\int_0^t e^{-A(p)\tau} \frac{\partial A(p)}{\partial p^i} e^{A(p)\tau} d\tau = tV + e^{-A(p)t} \{e^{A(p)t} U - U e^{A(p)t}\}. \quad (2.7)$$

**Proof :** Since

$$\int_0^t e^{-A(p)\tau} \frac{\partial A(p)}{\partial p^i} e^{A(p)\tau} d\tau = \int_0^t e^{-A(p)\tau} \{V + [A(p)U - UA(p)]\} e^{A(p)\tau} d\tau$$



$$= tV + e^{-A(p)t} \{ e^{A(p)t} U - U e^{A(p)t} \}. \quad (2.8)$$

In establishing the existence and uniqueness of solutions  $(U, V)$  for Equation (2.6), we shall make use of the following result which can be found in [Tay.1].

**Proposition 3 :** Let  $\mathcal{Q}$  be a linear operator mapping a finite dimensional linear space  $\mathcal{V}$  into itself and let  $R(\mathcal{Q})$  and  $N(\mathcal{Q})$  denote the range space and the null space of  $\mathcal{Q}$ . Given any scalar  $\lambda$ , if  $q$  is the smallest nonnegative integer such that  $N[(\mathcal{Q} - \lambda)^q] = N[(\mathcal{Q} - \lambda)^{q+1}]$ , then

$$\mathcal{V} = N[(\mathcal{Q} - \lambda)^q] \oplus R[(\mathcal{Q} - \lambda)^q]. \quad (2.9)$$

**Corollary 4 :** If  $\mathcal{Q}$  is diagonalizable, then  $\mathcal{V} = R(\mathcal{Q}) \oplus N(\mathcal{Q})$ .

**Proof :** If  $\mathcal{Q}$  is diagonalizable, then  $N[(\mathcal{Q} - \lambda)] = N[(\mathcal{Q} - \lambda)^2]$  for any scalar  $\lambda$ .

**Proposition 5 :** Let  $A \in \mathbb{R}^{n \times n}$  and let the Lie bracket type operator  $\mathcal{Q} : \mathbb{R}^{n \times n} \rightarrow \mathbb{R}^{n \times n}$  be defined by

$$\mathcal{Q}(X) = [A, X] \triangleq AX - XA. \quad (2.10)$$

Then  $\mathbb{R}^{n \times n} = R(\mathcal{Q}) \oplus N(\mathcal{Q})$ , if and only if  $A$  is diagonalizable.

**Proof :** " $\Leftarrow$ ": Suppose that  $A$  is diagonalizable. Let  $\lambda_1, \lambda_2, \dots, \lambda_n$  be the eigenvalues of  $A$ , let  $u_1, u_2, \dots, u_n$  be a set of linearly independent corresponding right eigenvectors of  $A$  and let  $v_1^T, v_2^T, \dots, v_n^T$  be a set linearly independent of corresponding left eigenvectors of  $A$ . Then (see [Gan.1])  $(\lambda_i - \lambda_j)$  for all  $i, j \in \underline{n}$  is an eigenvalue of  $\mathcal{Q}$ , and  $u_i v_j^T$  is an eigenvector of  $\mathcal{Q}$  corresponding to  $(\lambda_i - \lambda_j)$  for  $i, j \in \underline{n}$ . Since the set  $\{ u_i v_j^T \mid i, j \in \underline{n} \}$  contains  $n^2$  linearly independent eigenvectors,  $\mathcal{Q}$  is diagonalizable. By Corollary 4,  $\mathbb{R}^{n \times n} = R(\mathcal{Q}) \oplus N(\mathcal{Q})$ .

"==>": We give a proof by contraposition. Assume that  $A$  is not diagonalizable. Then, by the Jordan form theorem [Gan.1], there exists a nonsingular  $n \times n$  matrix  $U$  such that

$$U^{-1}AU = \begin{bmatrix} J_1 & 0 & \cdots & 0 \\ 0 & J_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & J_p \end{bmatrix} = J \quad (2.11)$$

where the  $J_i$  are  $n_i \times n_i$  (note that  $n_i \neq m_i$ ) Jordan blocks associated with eigenvalue  $\lambda_i$ , and  $n = \sum_{i=1}^p n_i$ . If the columns of  $U$  are denoted by  $u_i$ , so that  $U = [u_1, u_2, \dots, u_n]$ , and the rows of  $U^{-1}$  are denoted by  $v_i^T$ , so that  $V = [v_1, v_2, \dots, v_n]^T \triangleq U^{-1}$ , then  $u_1, u_2, \dots, u_n$  are generalized right eigenvectors of  $A$  and  $v_1^T, v_2^T, \dots, v_n^T$  are generalized left eigenvectors of  $A$ , see [Gan.1]. Also the dyads in  $\{u_i v_j^T \mid i, j \in \underline{n}\}$  form a basis for  $\mathbb{R}^{n \times n}$ . Because  $J$  is block diagonal, the equations  $AU = UJ$  and  $VA = JV$  decompose into equations of the form

$$A u_i = \lambda_i u_i + \nu_i u_{i-1}; \quad v_i^T A = \lambda_i v_i^T + \mu_i v_{i+1}^T, \quad (2.12)$$

where  $\nu_i, \mu_i$  can have only the values 0 or 1. Consider the  $n_1$  equations in (2.12) corresponding to the first Jordan block  $J_1$ :

$$A u_1 = \lambda_1 u_1; \quad v_1^T A = \lambda_1 v_1^T + v_2^T \quad (2.13)$$

$$A u_2 = \lambda_1 u_2 + u_1; \quad v_2^T A = \lambda_1 v_2^T$$

$$\dots \quad \dots$$

$$A u_{n_1-1} = \lambda_1 u_{n_1-1} + u_{n_1-2}; \quad v_{n_1-1}^T A = \lambda_1 v_{n_1-1}^T + v_{n_1}^T$$

$$A u_{n_1} = \lambda_1 u_{n_1} + u_{n_1-1}; \quad v_{n_1}^T A = \lambda_1 v_{n_1}^T$$

Hence  $\mathcal{Q}(u_1 v_{n_1}^T) = 0$  and  $\mathcal{Q}(u_1 v_{n_1-1}^T) = -u_1 v_{n_1}^T$ . This implies that  $u_1 v_{n_1}^T \in N(\mathcal{Q}) \cap R(\mathcal{Q})$  and therefore that  $\mathbb{R}^{n \times n} \neq N(\mathcal{Q}) \oplus R(\mathcal{Q})$ .

**Corollary 6 :** Let  $A \in \mathbb{R}^{n \times n}$  be diagonalizable, then for any  $M \in \mathbb{R}^{n \times n}$ , there exist  $V, U \in \mathbb{R}^{n \times n}$  such that

$$M = V + [A, U]; \quad [A, V] = 0. \quad (2.14)$$

Furthermore, the solution  $V$  of  $[A, V] = 0$  in (2.14) is unique.

**Proof :** By Proposition 5, for any  $M \in \mathbb{R}^{n \times n}$  there exists  $U \in \mathbb{R}^{n \times n}$  and  $V \in N(\mathcal{A})$  such that  $M = V + \mathcal{A}(U)$ . By the definition of direct sum,  $V$  and  $\mathcal{A}(U)$  are unique. ■

**Corollary 7 :** Let  $A(p)$  be defined as in equation (1.1). If  $A(p)$  is diagonalizable for a given  $p \in \mathbb{R}^N$ , then

$$\frac{\partial e^{A(p)t}}{\partial p^i} = t V e^{A(p)t} + [e^{A(p)t}, U] \quad (2.15)$$

for  $i \in \underline{N}$ , where  $V, U$  satisfy following equations:

$$\frac{\partial A(p)}{\partial p^i} = V + [A(p), U]; \quad [A(p), V] = 0. \quad (2.16)$$

**Proof :** The proof follows directly from the results of Proposition 2 and Corollary 6. ■

### 3. A Procedure for Computing the Matrix $\frac{\partial e^{A(p)t}}{\partial p^i}$ via Diagonalization.

We shall now state our procedure, based on (2.15) and Corollaries 6 and 7, for computing  $\frac{\partial e^{A(p)t}}{\partial p^i}$  when  $A(p)$  is an  $n \times n$ , continuously differentiable, diagonalizable matrix. To simplify the notation we denote  $\frac{\partial A(p)}{\partial p^i}$  by  $M(p)$ .

**Procedure :**

**Data:** A continuously differentiable, diagonalizable  $n \times n$  matrix  $A(p)$ .

**Step 1:** Diagonalize  $A(p)$  by computing a matrix of linearly independent eigenvectors  $T(p) \in \mathbb{C}^{n \times n}$  and setting  $T^{-1}(p)A(p)T(p) = \Lambda(p) = \text{diag}\{\lambda_1(p), \lambda_2(p), \dots, \lambda_n(p)\}$ .

**Step 2:** Compute  $\tilde{M}(p) = T^{-1}(p)M(p)T(p)$ , and construct the matrices  $\tilde{U}(p), \tilde{V}(p) \in \mathbb{C}^{n \times n}$  as follows:

$$\tilde{V}_{ij}(p) = \begin{cases} \tilde{M}_{ij}(p) & \text{if } \lambda_i(p) = \lambda_j(p) \\ 0 & \text{otherwise} \end{cases} \quad (3.1)$$

$$\tilde{U}_{ij}(p) = \begin{cases} \frac{\tilde{M}_{ij}(p)}{\lambda_i(p) - \lambda_j(p)} & \text{if } \lambda_i(p) \neq \lambda_j(p) \\ 0 & \text{otherwise} \end{cases} \quad (3.2)$$

for all  $i, j \in \underline{n}$ .

**Comment:** Note that  $[\Lambda, \tilde{V}] = 0$ , and

$$\{\tilde{V} + [\Lambda, \tilde{U}]\}_{ij} = \tilde{V}_{ij} + \lambda_i \tilde{U}_{ij} - \tilde{U}_{ij} \lambda_j = \begin{cases} \tilde{V}_{ij} & \text{if } \lambda_i = \lambda_j \\ \tilde{M}_{ij} & \text{if } \lambda_i \neq \lambda_j \end{cases} \quad (3.3)$$

Hence  $\tilde{M} = \tilde{V} + [\Lambda, \tilde{U}]$ .

**Step 3:** Compute  $V(p) = T(p)\tilde{V}(p)T(p)^{-1}$  and  $U(p) = T(p)\tilde{U}(p)T(p)^{-1}$ , which solve equations (2.16), and set

$$\frac{\partial e^{A(p)t}}{\partial p^i} = t V e^{A(p)t} + [e^{A(p)t}, U]. \quad (3.4)$$

If the sensitivity of the diagonalized system is required, set

$$T^{-1} \frac{\partial e^{A(p)t}}{\partial p^i} T = t \tilde{V} e^{\Lambda(p)t} + [e^{\Lambda(p)t}, \tilde{U}]. \quad (3.5)$$

#### 4. Conclusion

We have presented a procedure for the computation of the matrix  $\frac{\partial e^{A(p)t}}{\partial p^i}$  which appears in (1.2b) for the case where  $A(p)$  is diagonalizable. Since this

procedure involved the diagonalization of the matrix  $A(p)$ , it should be clear that the completion of the computation of  $\frac{\partial x(t,p)}{\partial p^i}$  is quite straightforward and does not require numerical integration for inputs whose components are sums of terms of the form  $t^\alpha e^{\lambda t}$ , with  $\alpha$  a nonnegative integer. Our procedure is particularly efficient when the matrix  $A(p)$  in (1.1a) is large and/or when evaluations of  $e^{A(p)t}$  and  $\frac{\partial e^{A(p)t}}{\partial p^i}$  must be carried out for many different values of  $t$ .

Since matrix diagonalization is unstable when  $A(p)$  is defective or near-defective, we follow [Mol.1] and use the condition number,  $cond(T) = \|T\| \|T^{-1}\|$ , of the matrix of eigenvectors as a testing function for the defectiveness of  $A(p)$ . When  $A(p)$  is nearly (exactly) defective, the  $cond(T)$  is large (infinite). Hence any errors in  $A(p)$ , and roundoff errors in the eigenvalue computation, may be magnified in final result of the decomposition  $A(p) = T(p)\Lambda(p)T^{-1}(p)$  by  $cond(T)$ . Consequently, when  $cond(T)$  is large, the computed  $e^{A(p)t}$  and  $\frac{\partial e^{A(p)t}}{\partial p^i}$  will most likely be inaccurate.

When  $cond(T)$  is large, we propose to abandon the formulae (1.1a,b) and solve instead the following equation,

$$\frac{d}{dt} \begin{bmatrix} x(t,p) \\ \frac{\partial x(t,p)}{\partial p^i} \end{bmatrix} = \begin{bmatrix} A(p) & 0 \\ \frac{\partial A(p)}{\partial p^i} & A(p) \end{bmatrix} \begin{bmatrix} x(t,p) \\ \frac{\partial x(t,p)}{\partial p^i} \end{bmatrix} + \begin{bmatrix} B(p) \\ \frac{\partial B(p)}{\partial p^i} \end{bmatrix} u(t). \quad (4.1)$$

Solution methods based on the *Schur* transformation [Par.1] or the *Padé* approximation (see [Mol.1, Lau.1]) appear to be particularly appropriate for this case.

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