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NEW RESULTS IN DISCRETE-TIME LOOP TRANSFER RECOVERY

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ABSTRACT
For discrete-time compensators incorporating prediction observers asymptotic loop transfer recovery is not feasible. Instead loop transfer recovery objectives must be satisfied via exact recovery techniques. In this note the model-based compensators which achieve exact recovery are parametrized in terms of the system zeros and the corresponding zero-directions. Full-order as well as minimal-order observers are treated. Further it is shown how exact recovery is also applicable to non-minimum phase plants. In this case the achievable performance is parameterized explicitly.

1 INTRODUCTION
In recent years the LQG/LTR feedback design methodology for robust model-based compensation has received much attention [see e.g. 1-6]. This procedure works for continuous-time systems - and it is always effective for minimum-phase plants. Unfortunately a similar procedure is not generally feasible in discrete-time. If filtering observers are used asymptotic recovery (the LTR step) is often possible [11]. However, the application of filtering observers require that the processing time of computing the control signal is negligible in comparison to the sampling interval. Very often such an assumption cannot be satisfied in practice, and prediction observers must be used. For compensators based on prediction observers, however, the asymptotic procedures will not be effective, since in general the difference between a full-state loop transfer (target design) and the full asymptotic loop transfer remains finite [4,11]. A detailed discussion of the mechanisms behind this fact is given in [4].

Loop transfer recovery is still possible, however, but different methods must be applied. In [4,11] such methods are discussed - and referred to as exact loop transfer recovery. In [4] the conditions for exact recovery for full-order observers were outlined, and some preliminary design considerations for minimum-phase continuous-time systems based on full-order observers were presented in [10]. In this note a more general treatment of exact recovery in discrete-time is provided. Exact recovery for minimum-phase as well as non-minimum phase plants based on full-order observers are discussed. Further results on exact recovery based on minimal-order observers are presented, and it is shown that in certain - common - cases very powerful designs procedures are possible. This is the first treatment of LTR for minimal-order observers in discrete-time. Earlier studies [16,17] were in continuous-time, but due to the same problems as for full-order observers the continuous-time methods cannot be generalized to discrete-time. Hence new methods based on exact recovery must be developed. Notice that the issue of recovery for non-minimum phase is particularly relevant in discrete-time since the sampling process often produces zeros outside the unit-circle [13]. An advantage of using the exact recovery concepts presented here is that the controllers are of finite gains, whereas the usual continuous-time LQG/LTR method often produces high-gain controllers.

The paper is organized as follows. In § 2-4 the full-order observer case is treated, and in § 5-7 minimal-order observer results are presented follow in § 8 by some examples.

2 EXACT LOOP TRANSFER RECOVERY
In the following square discrete-time minimum phase systems $S(A,B,C)$ are considered. It will be assumed that the model is minimal. The plant transfer matrix $G(z)$ and the model-based compensator $H(z)$ are given

$$G(z) = C(z)B,$$  \hspace{1cm} \text{dim } G(z) = m \times m

$$\Phi(z) = (zI - A)^{-1},$$  \hspace{1cm} \text{dim } \Phi(z) = n \times n

$$H(z) = K(zI - A + BK + FC)^{-1}F, \text{dim } H(z) = m \times m$$  \hspace{1cm} (2.1)

Here $K$ is the full-state feedback gain and $F$ is the full-order observer-gain. Let the number of transmission zeros be $p$. In order to formulate the loop-shape robustness constraints the uncertainties (disturbances, noise and modelling errors) are reflected to the plant input mode $[4,14]$. The target loop transfer is then the full-state loop transfer $K\Phi$ and the full loop transfer is $HG$ $[3,5]$. The difference between these two indicators is defined as the loop recovery error $E_z$:

$$E_{1}(z) = K\Phi(z)B - H(z)6(z)$$  \hspace{1cm} (2.2)

In order to have exact recovery it is required that $E_{1}(z) \equiv 0$ for all $z$. For square systems Goodman [4] has shown that

$$E_{1}(z) = M_{1}(z)(I + M_{1}(z))^{-1}(I + K\Phi(z)B)$$  \hspace{1cm} (2.3)

$$M_{1}(z) = K(zI - A + FC)^{-1}B$$

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It is, however, straightforward to derive the same results for non-square systems as well. Now let $H_1(z)$ be rewritten in the residual form:

$$H_1(z) = \frac{K v_1 T w_1}{z - \lambda_i} \sum_{i=1}^{n} (2-4)$$

where $v_i$ and $w_i$ are right and left eigenvectors associated with the eigenvalue $\lambda_i$ of $A - FC$. It is easy to show that

$$E_1(z) = 0 \quad \text{iff} \quad v_i = 0 \quad \text{iff} \quad (2-5a,b,c)$$

$$Kv_i = 0 \quad \text{or} \quad w_iB = 0, \quad i = 1, \ldots, n$$

If $A - FC$ is non-defective, the latter formulation of the exact recovery condition is suitable deriving the associated compensators.

3 SOLUTION OF THE EXACT LTR PROBLEM

From eigenvector assignment it is known that the left eigenvectors $w_i$ with the eigenvalue $\lambda_i$ of $A - FC$ are given by [9]:

$$\left[ \begin{array}{c} w_1^T \\ z_1^T \end{array} \right] \left[ \begin{array}{cc} \lambda_i I - A \\ -C \end{array} \right] = 0, \quad i = 1, \ldots, n$$

(3-1)

The condition $w_i B = 0$ from (2-5) imply that

$$\left[ \begin{array}{c} w_1^T \\ z_1^T \\ \vdots \\ w_n^T \\ z_n^T \end{array} \right] \left[ \begin{array}{cc} \lambda_i I - A \\ -C \end{array} \right] = 0$$

(2-2)

Maximally $p$ eigenvectors $w_i^T$ can satisfy this condition, if $\lambda_i$ is selected as a transmission zero of $S(A,B,C)$ [8]. Let these $p$ eigenvectors/ vectors be selected from (3-2), it is then straightforward to see that $F$ is parameterized by:

$$F = \left[ \begin{array}{c} w_1^T \\ \vdots \\ w_p^T \end{array} \right]^{-1} \left[ \begin{array}{c} z_1^T \\ \vdots \\ z_p^T \end{array} \right]$$

(3-3)

$$z_1^T = z_1^{10}, \quad w_1^T = w_1^{10}, \quad i = 1, \ldots, p$$

and $(\lambda_i, z_i, i = p+1, \ldots, n)$ are free design parameters, since $w_i^T$ is determined by $\lambda_i$ and $z_i$. The remaining $n-p$ conditions in (2-5c) must be satisfied by selecting $K$ suitably. Condition (2-5c) imply

$$K(v_1, \ldots, v_n) = [Q 0]$$

(3-4)

with $\dim Q = m \times p$ but otherwise arbitrary. Now

$$K = [Q 0]v^{-1} = Q \left[ \begin{array}{c} w_1^T \\ \vdots \\ w_p^T \end{array} \right] = Q \Gamma$$

(3-5)

with $\dim \Gamma = p \times n$. $\Gamma$ consists of the left eigenvectors $w_i^T$ constrained in (3-2), and is thus a matrix of fixed elements. Eq (3-3) and (3-5) are therefore simple parameterizations of the controller matrices which achieves exact recovery.

A few important consequences of exact LTR are discussed next:

- The parameterization of the state-feedback imply that $K$ must be selected as an output feedback controller, where $Q$ is the free parameter output feedback matrix. $\Gamma$ is the equivalent output matrix with $p$ independent columns. Since $p \times n = m \times m-p+1$ $(c > n$) eigenvalues can be assigned for such a problem [7]. Consequently all of the close-loop eigenvalues cannot be assigned freely, and no stability guarantees are available. However, in square discrete-time systems the $\text{rank}([\Gamma])$ is often maximal. This ensures that $G(z)$ has the maximum possible number of finite zeros. Which in turn will result in maximal freedom in selection of $K$.

- The selection of $F$ is only constrained by eq. (3-3) and stability can always be achieved.

- Good input sensitivity and stability for plant input modelling errors can only be achieved if $p < m$. If $\text{rank}(K) < m$ ($p < m$) the target open-loop transfer $KFB$ is rank defective and loop-shaping is not feasible.

- Dual results apply for the plant output breaking point.

- The structure of the controller $H(z)$ can be studied by looking at the system matrix for the controller $P_H$:

$$P_H = \left[ \begin{array}{c} I \\ A \end{array} \right] + BK + FC \quad F \quad K \quad 0$$

(3-6)

By using the transformation matrix $T = \text{diag}(V, I)$, eq. (3-6) can be transformed into:

$$P_H = \left[ \begin{array}{c} I \\ A \end{array} \right] + BK + FC \quad F \quad K \quad 0$$

(3-6)

where $V$, $Z_1$, and $Z_2$ has full rank. $A$ are the plant zeros and $\lambda_i$ are the remaining $m$ poles of $A - FC$ assigned in eq. (3-3). Notice that $A$ are the poles of $H(z)$ and $\lambda_i$ are output decoupling zeros of $H(z)$. Hence the resulting loop transfer $H_G$ will have $n$ poles.

- It has been assumed that $S(A,B,C)$ is minimal. The results could be extended to non-minimal systems as well - although this issue is not pursued here.

- Further the treatment is also possible for non-square systems. Since this is straightforward no details are given here.

- Notice that the exact recovery controllers outlined above are of finite gains, whereas the continuous-time LQG/LTR procedures usually produces a high-gain controller.

4 NON-MINIMUM-PHASE SYSTEMS

Sampling of a continuous-time system will often result in a non-minimum phase discrete-time
system [13]. If the LTR results from section 3 are used on a non-minimum phase system \( G(z) \), the resulting controller will be unstable. It is, however, still possible to achieve LTR for non-minimum phase systems. In order to facilitate exact recovery for non-minimum phase systems note that in selecting \( F \) only a subset \( j \) of the eigenvectors constrained by eq. (3-2) need to be chosen. In doing this, however, the dimension of \( Q \), the free parameters of \( K \), is reduced to \( m \times j \). Consequently such selection are only advisable for non-minimum phase systems. If only the plant's q minimum phase zeros are used in eq. (3-2), the equations for \( F \) and \( K \) become:

\[
F = \begin{bmatrix} \omega_1^T \\ \omega_2^T \end{bmatrix}^{-1} \begin{bmatrix} \omega_1 \\ \omega_2 \end{bmatrix}
\]

\[
K = \begin{bmatrix} T \\ w_1 \\ w_2 \end{bmatrix}^{-1} \begin{bmatrix} T \\ w_1 \\ w_2 \end{bmatrix}
\]

where \( \dim \omega = m \times q \).

Some of the consequences of exact LTR for non-minimum phase plants are:

* The following equation will be satisfied

\[
K + B = H(z)G(z)
\]

* The non-minimum zeros of \( G \) are not cancelled out on the right hand side. Hence \( H(z)G(z) \) and \( K + B \) are both non-minimum phase. This in turn limits the achievable performance [12].and zero loop-shapes for \( K + B \) are, of course, difficult to achieve. Notice how the achievable zero loop-shapes - under the exact recovery constraint - are parameterized explicitly in eq. (4-1) by the constraints of \( K \). This results is in agreement with the results in [18].

* The freedom in the selection of \( K \) will decrease by the number of non-minimum phase zeros in \( G(z) \).

* The consequences of exact LTR from section 3 are still valid.

### 5 MINIMAL ORDER OBSERVERS

In the following the discrete-time system \( S(A,B,C) \) will be partitioned as:

\[
S = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}
\]

\[
B = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}
\]

\[
C = \begin{bmatrix} I \\ 0 \end{bmatrix}
\]

There is no loss of generality in assuming that \( C = [I, 0] \) since any system can be transformed into this form. The system is assumed to be minimal-phase, with \( p \) zeros. The minimal order observer for (5-1) is [15]:

\[
z(k+1) = Dz(k) + Gu(k) - Ey(k)
\]

\[
x(k) = \begin{bmatrix} x_1(k) \\ x_2(k) \end{bmatrix} = \begin{bmatrix} V_1 T^{m-n} \\ V_2 I_{n-m} \end{bmatrix}
\]

where

\[
D = A_{22} - V_2 B \]

\[
G = B_2 - V_2 B \]

\[
E = A_{21} - V_2 A + V_2 V_2^T B_2 \]

and \( V_2 \) is the observer gain matrix.

The feedback law is:

\[
u(k) = -Kx(k) = -Kx_1(k) - Kx_2(k)
\]

It is assumed that \( (C,A) \) is observable, which implies that \( (A_{22}, A_{21}) \) is observable [15].

It is known that the separation principle applies for this feedback system. Hence stability is achieved by making the full-state and the minimal-order observer stable. The condition for LTR for the minimal-order observer based design is [16,17]:

\[
V_2^T (I + A_{12}^T B_2) V_2^{-1} A_{12} + (B_2 - V_2 B) = B_2 - V_2 B
\]

where

\[
\phi = (Iz - A_{22})^{-1}
\]

This condition is similar to the continuous-time version, but the design results from [16,17] can not be generalized, and new methods for utilizing (5-4) in discrete-time are derived in § 6. If (5-4) is satisfied then:

\[
K (\phi - V_2 A_{12}^{-1} B_2 - V_2 B) = 0
\]

is also satisfied [16,17]. Eq. (5-5) is a necessary and sufficient condition for LTR with minimal-order observers. If (5-4) is satisfied the full-state loop transfer \( K + B \) and the minimal-order observer based loop transfer are identical.

Let (5-5) be rewriten in the residual form:

\[
0 = \begin{bmatrix} I \\ \phi^T \end{bmatrix} \begin{bmatrix} I \\ \phi^T \end{bmatrix} w_1 w_2 (B_2 - V_2 B)
\]

where \( w_1 \) and \( w_2 \) are right and left eigenvectors associated with the eigenvalue \( \lambda_i \) of \( A_{22} - V_2 A_{12} \) and from eigenvector structure assignment [9] it is easily found that

\[
w_1^T = z_i \phi \]

\[
w_2^T = z_i^T
\]

\[
(5-7)
\]

\[
(5-8)
\]
It is easy to show that eq. (5-6) is satisfied if:
\[ K_w = 0 \text{ or } w_T (B_1 - V_2 B_1) = 0, \ i = 1, \ldots, n-m \]  
(5-9)

The condition implies 3 different design cases depending on the rank of \( B_1 \):

6 LTR SOLUTIONS FOR MINIMAL ORDER OBSERVERS

Case 1: \( r(B_1) = 0 \)

The recovery condition (5-9) now becomes:
\[ K_w = 0 \text{ or } w_T (B_1 - V_2 B_1) = 0, \ i = 1, \ldots, n-m \]  
(6-1)

The second condition in (6-1) together with (5-7) result in:
\[ z_{10} A_{12} 2^1 1^1 0^2 0^2 = 0, \ i = 1, \ldots, n-m \]  
(6-2)

This condition can be satisfied if \( \lambda_{0i} \) is selected as the transmission zeros of \( S(A,B,C) \), see [6]. Eq. (6-2) can be satisfied for maximally \( p \) eigenvalues \( \lambda_{0i} \) [6]. Let these eigenvalues be selected from (6-2), it is then straightforward to see that \( V_2 \) is parameterized by:
\[ V_2 = \begin{bmatrix} w_T \end{bmatrix} \begin{bmatrix} z_1 \end{bmatrix} \begin{bmatrix} z_{n-m} \end{bmatrix} = -w^T z \]  
(6-3)

\[ z_i = z_{10} A_{12} 2^1 1^1 0^2 0^2, \ i = 1, \ldots, p \]

and \( \lambda_{0i}, z_i, i = 1, \ldots, n-m \) are free design parameters. The first equation in (5-7) must be satisfied for the remaining \( n-m-p \) conditions by selecting \( K_2 \) as:
\[ K_2 = \begin{bmatrix} V_1, \ldots, V_{n-m} \end{bmatrix} = [ Q, 0 ] \]  
(6-4)

with \( \dim Q = m \times p \) but otherwise arbitrary.

Now
\[ K_2 = [ Q, 0 ] V^{-1} \]
\[ = \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} = Q \Gamma \]  
(6-5)

with \( \dim \Gamma = p \times (n-m) \). \( \Gamma \) consists of the left eigenvectors \( w_{10i} \) constrained in (6-2).

Case 2: \( r(B_1) = m \)

The condition \( r(B_1) = m \) indicate that the system \( S(A,B,C) \) has \( p \times (n-m) \) zeros. The recovery conditions can now be satisfied only by \( V_2 \) and \( K_2 \) is free to design.

The recovery condition is:
\[ w_T (B_1 - V_2 B_1) = 0, \ i = 1, \ldots, n-m \]  
(6-6)

This equation can be rewritten as (by using eqs. (5-7) and (5-8)):
\[ z_{10} A_{12} 2^1 1^1 0^2 0^2 = 0 \]  
(6-7)

The \( n-m \) equations can be satisfied by selecting \( \lambda_{0i} \) as the zeros of the system \( S(A,B,C) \), see [6].

The solution is:
\[ V_2 = B_2 B_1^{-1} \]  
(6-9)

Case 3: \( 0 < r(B_1) < m \)

The recovery condition (5-9) is:
\[ K_w = 0 \text{ or } w_T (B_1 - V_2 B_1) = 0, \ i = 1, \ldots, n-m \]  
(6-10)

The second recovery condition can again be rewritten as:
\[ z_{10} A_{12} 2^1 1^1 0^2 0^2 = 0 \]  
(6-11)

Maximally \( p \) eigenvalues \( w_{10i} \) satisfy this condition by selecting the eigenvalues \( \lambda_{0i} \) as the zeros of the system \( S(A,B,C) \), and \( z_{10i} \) as the corresponding zero directions (see [6]). The first equation in (6-10) must then satisfy the remaining \( n-m-p \) conditions by suitably selecting \( K_2 \).

The solution in this case is similar to case 1.
\[ V_2 = - \begin{bmatrix} w_T \end{bmatrix} \begin{bmatrix} z_1 \end{bmatrix} \begin{bmatrix} z_{n-m} \end{bmatrix} = -w^T z \]  
(6-12)

with \( z_i = z_{10} A_{12} 2^1 1^1 0^2 0^2, \ i = 1, \ldots, p \)

\[ K_2 = \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} \begin{bmatrix} \begin{bmatrix} w_T \end{bmatrix} \end{bmatrix} = Q \Gamma \]  
(6-13)

with \( \dim \Gamma = p \times (n-m) \), \( \dim Q = m \times p \) but otherwise arbitrary.

\( V \) can be rewritten into a form which emphasizes the fact that case 3 is inbetween case 1 and case 2. To see this, we assume that \( B_1 \) is transformed into:
\[ B_1 = \begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} \]
where \( A \) is a diagonal matrix.

Let \( A_{12} \), \( B_{12} \) and \( V_{2} \) be partitioned as:

\[
A_{12} = \begin{bmatrix}
A_{121} & \vdots \\
A_{122} & \vdots
\end{bmatrix} \quad (6-14)
\]

\[
B_{2} = \begin{bmatrix}
B_{21} & B_{22}
\end{bmatrix} \quad n - m
\]

\[
V_{2} = \begin{bmatrix}
V_{21} & V_{22}
\end{bmatrix} \quad n - m
\]

The second condition in (6-10) can now be written as:

\[
w_{2}^T (B_{21} B_{22} - (V_{21} V_{22}) \begin{bmatrix}
\Lambda & 0 \\
0 & 0
\end{bmatrix}) i = 0 \\
i = 1, \ldots, n-m \quad (6-15)
\]

By selecting \( V_{21} = B_{21} \Lambda^{-1} \) \( (6-16) \), \( (6-15) \) will be reduced to:

\[
w_{2}^T B_{22} = 0, \quad i = 1, \ldots, n-m \quad (6-17)
\]

where the left eigenvectors \( w_{2}^T \) in (6-17) are given by:

\[
w_{2}^T = Z_{i}^T A_{122} \Lambda_{122}^{-1} (\Lambda_{22}) \quad (6-18)
\]

where \( \Lambda_{22} = (I_{z} - \bar{\Lambda}_{22})^{-1} \)

and \( \bar{\Lambda}_{22} = A_{22} - 21 {\Lambda}_{22}^{-1} A_{122} \)

\[
T w_{i} B_{22} = 0 \quad i = 1, \ldots, n-m \quad (6-19)
\]

Now it is straightforward to see that \( V_{2} \) is parameterized by:

\[
V_{2} = \begin{bmatrix}
B_{21} A_{121}^{-1} & -Z_{1}^T Z_{n-m}
\end{bmatrix} \quad (6-20)
\]

with \( Z_{i}^T = Z_{10}^T + \bar{\Lambda}_{22} \Lambda_{122} (\Lambda_{10}) \), \( i = 1, \ldots, p \)

The resulting \( V_{2} \) is:

\[
V_{2} = \begin{bmatrix}
B_{21} A_{121}^{-1} & -Z_{1}^T Z_{n-m}
\end{bmatrix} \quad (6-21)
\]

where \( -Z_{n-m} \) is given in (6-20).

The remaining \( n-m-p \) conditions in (6-10) must be satisfied by selecting \( K_{2} \) as before:

\[
K_{2} = \begin{bmatrix}
W_{10} & \vdots \\
W_{p0}
\end{bmatrix} \quad \Gamma \quad (6-22)
\]

with \( \dim \Gamma = p \times (n-m) \).

A few important consequences of exact LTR for minimal-order observers are now discussed here:

* If the system does not have any zeros, exact recovery is still possible with the solution \( K = 0 \), i.e., no feedback from the state estimates. However, in square discrete-time systems, \( r(0) \) is often maximal, which ensures that \( r(0) \) is maximal and that \( C(z) \) has the maximum possible number of finite zeros, \( p = n-m \). In this special case exact LTR is possible only by selecting the observer gain \( V_{2} \). The feedback gain \( K \) is free to choose, and it is possible to use systematic design rules (e.g., LQG-design) for the \( K \) selection for stability and loop-shape requirements. This is a very useful result for LTR design in discrete-time systems because a full-state target design can be recovered, without affecting this original design, simply by choosing the minimal-order observer gain, whereas it is not possible with a full-order observer. Here the full-state design is constrained. Note that by using a minimal-order observer in compensators will require that the processing time of computing the control signal is negligible in comparison to the sampling interval. The processing time in this case will, however, be reduced compared with the processing time when a filtering observer is used and therefore the minimal-order observer is more attractive than the filtering observer.

* The result in case 3 (6-21) is the general result for exact recovery with minimal-order observer, since the solution constrains case 1 and 2 as special cases.

* Good input sensitivity and stability robustness for plant input modelling errors can always be achieved if the target loop \( K \# B \) has full rank. This is only guaranteed if \( p \geq m \) in case 1 and 3. In case 2 \( K \# B \) has generically full rank, and therefore good feedback properties can be achieved.

* Finally note that dual results for the plant output cannot be invoked, due to the missing duality of minimal-order observers.

7 NON-MINIMUM-PHASE SYSTEMS

The results for LTR with minimal-order observers of § 6 were based on a minimum-phase assumption. If this assumption is not valid some new results can be obtained. In the following the tree usual cases will be discussed independently, but a basic prerequisite will be the recovery conditions.

\[
K_{2} (I_{z} - A_{22} + V_{2} A_{12})^T (B_{2} - V_{2} B_{1}) = 0 \quad (7-1)
\]

\[
K_{2} Z_{n-m} (I_{z} - A_{22} + V_{2} A_{12})^T C = 0 \quad i = 1, \ldots, n-m \quad (7-2)
\]
where the symbols are defined in § 5.6.

Further let the number of plant zeros be p and the number of minimum-phase zeros be q.

**Case 1.** \( r(B) = 0 \)

In this case the recovery condition becomes:

\[
K_w = 0 \quad \text{or} \quad w_i \Lambda_i = 0, \quad i = 1, \ldots, n-m \quad (7-2)
\]

Due to the stability requirements only a subset \( q \) of the possible solutions to the condition \( w_i \Lambda_i = 0 \) can be selected, i.e. the \( q \) solutions:

\[
T \lambda_i \Lambda_i = 0 \quad (7-3)
\]

where \( \lambda_i \) are the zeros of \( S(A,B,C) \) - see [6].

The remaining \( n-q \) conditions constrains \( K \).

As in § 4 the solution becomes:

\[
V_2 = [T^{-1} z_1 \ldots T^{-1} z_{n-m}] [-1 1 \ldots -1] \quad (7-4)
\]

\[
T_{10} = z_{10} A_{10} \quad (7-5)
\]

The resulting \( n-m-q \) conditions for \( K \) are:

\[
K = \bar{O}^{-1} q \bar{T} \quad (7-6)
\]

\[
\bar{O} \text{ is a matrix of free parameters.}
\]

\[
\text{The remaining } n-m-q \text{ pairs } (\lambda_i, z_i) \text{ are free parameters.}
\]

**Case 2.** \( r(B) = m \)

Now the recovery condition becomes:

\[
K_w = 0 \quad \text{or} \quad w_i \Lambda_i = 0, i = 1, \ldots, n-m \quad (7-7)
\]

As before only \( q \) solutions to the conditions \( w_i \Lambda_i = 0 \) can be used, i.e.

\[
T \lambda_i \Lambda_i = 0 \quad (7-8)
\]

and \( \lambda_i \) is a zero of \( S(A,B,C) \) - see [6] for details.

The remaining \( n-m-q \) conditions must be satisfied by selecting \( K_2 \) appropriately. The expressions for \( V_2 \) and \( K_2 \) are similar to (7-4) with \( w_i \Lambda_i = 0 \).

**Case 3.** \( 0 < r(B) < m \)

In this case the recovery condition as \( (7-5) \). The \( q \) possible stable solutions to \( w_i \Lambda_i = 0 \) are given by eq. (7-6). The last \( n-m-q \) constraints \( K_2 \) - and the expressions for \( V_2 \) and \( K_2 \) are similar to eq. (7-4), with eq. (7-6) substituted for eq. (7-3).

A general comment for these results concerns the selection of \( K_2 \). In all three cases the matrix \( K \) is not free to assign, hence stability-design and loop-shape design are not as straightforward as one would desire. Otherwise the results from § 4 are also valid here. Notice again that the achievable loop-shapes - subject to the exact recovery constraint - are parameterized explicitly in terms of \( K \), i.e. the free parameters \( \bar{O} \) and the left eigenvectors \( w_i \Lambda_i \) and \( K_1 \).

**B EXAMPLES**

Consider the plant:

\[
G(s) = \frac{1+2s}{s(s^2+0.8s+1)}
\]

Let the sampling time be 0.25 sec. The discrete-time version \( G(z) \) then has zeros at:

\[
z_1 = 0.8825, \quad z_2 = -0.2502, \quad z_3 = 3.3968
\]

and \( G(z) \) is non-minimum phase. By applying the exact recovery procedure for full-order observers of § 3 the compensator becomes:

\[
H(z) = \frac{w_1(z-z_1) + w_2(z-z_2)}{(z-z_1)(z-z_2)}
\]

Where \( w_1 \) and \( w_2 \) are the 2 elements of \( \bar{O} \). The resulting loop transfer is then:

\[
K_0 = \frac{G(z)H(z)}{d(z)}
\]

Here \( d \) denotes the characteristic polynomial of \( A \).

As expected the non-minimum phase zero shows up in \( K_0 \). \( w_1 \) and \( w_2 \) are free design parameters which determines the shape of \( K_0 \) and stability of the closed-loop system. Notice how the performance for the non-minimum phase control-loop is characterized directly by \( w_1 \) and \( w_2 \).

As the second example consider the plant:

\[
A = \begin{bmatrix}
1.0044 & -5.2447E-3 & 1.4436E-2 \\
5.1372E-5 & 1.0001 & 2.3999E-5 -5.8045E-1 \\
-5.2161E-5 & 5.3818E-3 & 9.9980E-1 2.2215E-2
\end{bmatrix}
\]

\[
B = \begin{bmatrix}
3.5825E-3 & -8.6185E-2 \\
9.9749E-4 & 2.1474E-5 \\
-1.4399E-3 & 1.2011E-3 \\
-3.4725E-3 & -8.1575E-5
\end{bmatrix}
\]

This is an example from [4] transformed into form required for minimal-order observer design. In [4] it was attempted to design a discrete LQG/LTR regulator, but a finite recovery error was obtained for all frequencies. Here a minimal-order observer will be applied. The system is minimum-phase with zeros at \((-0.99802, -0.99468)\). The sampling-time is 0.01 sec.

A target feedback design is given by:

\[
K = \begin{bmatrix}
3.3072E+2 & 1.8503E+3 & 2.2942E+4 & -9.2927E+3 \\
-1.0656E+3 & -4.2362E+3 & -7.3194E+4 & 2.8251E+4
\end{bmatrix}
\]
A nominal observer is designed as \( V = -W^T \)
with eigenvalues at \((5.32E-3, -1.8E+4)\).
A recovery trajectory is defined from \( V \) to the
exact LTR value \( V_L = B_0 B_1^{-1} \) by moving the

eigenvalues \( \lambda \) and zero-directions \( z \) from
the
nominal to the LTR-values as functions of \( q \), so that
\[
\lambda_i(q=0) = \lambda_i^0, \quad \lambda_i(q=\infty) = \lambda_i^{LTR}
\]
and equally for \( z_i \). And \( V_L(q=\infty) = B_0 B_1^{-1} \).

The plot of the singular values of the full loop
transfer is shown in fig. 1 and 2 for different
values of \( q \). Clearly recovery is achieved. The
final value of \( V_L \) which achieves exact recovery is:
\[
V_L = B_0 B_1^{-1} = \begin{bmatrix}
-1.432E-2 & -1.3920 \\
-2.9920E-5 & -3.4812
\end{bmatrix}
\]

REFERENCES