THE RELATIONSHIP BETWEEN *THE* ZAMES REPRESENTATION AND LQG COMPENSATORS

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ABSTRACT

Recently Zames [1] has introduced the so-called model
reference transformation which can be used to conveniently parametrize the class of linear time-invariant multivariable compensators that lead to stable feedback multivariable compensators that lead to stable recuper.
Control systems. A very popular design methodology for The results of Zames [1] state that the closed-loop
designing stable and robust multivariable control systems is that based on the Linear-Quadratic-Gaussian (LQG) design methodology with Loop Transfer Recovery (LTR). Thus, it is natural to inquire upon the relationship where the mom transfer matrix $Q(s)$ is defined by the between the Zames representation and the ICC (ITP some verse the mom transfer matrix $Q(s)$ is defined by the between the Zames representation and the LQG/LTR com-

compensator C(s) and the open-loop plant P(s) as

compensator C(s) and the open-loop plant P(s) as pensators. This paper summarizes this relationship. $\begin{array}{cc}\text{compensa:} \ \text{follows:}\end{array}$

1. INTRODUCTION

The Zames representation [1] has attracted a significant attention in the literature; see also [2]. A recent paper by Gustafson and Desoer [3] presents a Tt follows that the compensator $C(s)$ is given by design methodology which has its roots upon the Zames representation. On the other hand, the most popular design methodology for linear time-invariant multivariable systems remains the LQG-based design method 14],[7), with loop-transfer-recovery (LTR) 15],[6],[8]. The LQG/LTR method generates nominally stable compen-
The LQG/LTR method generates nominally stable compen-
 sators with superior robustness properties [6] in the
absence of non-minimum phase zeroes. The design of S absence of non-minimum phase zeroes. The design of LQG/LTR feedback control systems can now be routinely carried out by loop-shaping methods in the frequency and the input sensitivity matrix \underline{S}_i (s) domain, [5], [6], [8].

Only time will tell whether design methodologies based upon the Zames representation will approach the popularity of LQG/LTR methods. In the meantime it is The recent interest [1], [2], [3] in this representation
of interest to relate these methodologies. Thinges on the fact that the sensitivity matrices S (s)

In Section 2 we summarize the Zames Representation.
In Section 3 we start with the class of the so-called. readily apparent in the literature are stated without using Eq. (5). The resultant closed loop trans
proof. Next, we specify the Zames representation for trix, given by Eq. (3), is then calculated by proof. Next, we specify the Zames representation for , the class of LQG/LTR designs.

Figure 1 illustrates a block diagram of a multi- 3. THE CLASS OF LQG COMPENSATORS variable linear time-invariant feedback control system. The mxm transfer matrix $P(s)$ represents the open-loop In this section we examine the class of Linearplant, and the mxm transfer matrix $C(s)$ represents the α Quadratic-Gaussian (LQG) compensators with Loop-Trans-
dynamic compensator. Elementary algebra leads to the α fer-Becovery (LTB) [4] [5] [6]. It is guite bel

$$
\underline{y}(s) \stackrel{\Delta}{=} \underline{G}_{CL}(s) \underline{r}(s) \tag{1}
$$

$$
\underline{G}_{CL}(s) = [\underline{I} + \underline{P}(s) \underline{C}(s)]^{-1} \underline{P}(s) \underline{C}(s)
$$

= $\underline{P}(s) \underline{C}(s) [\underline{I} + \underline{P}(s) \underline{C}(s)]^{-1}$ (2),

transfer matrix G_{CL} (s) can be expressed in the form

$$
G_{\text{cr}}(s) = P(s)Q(s) \tag{3}
$$

$$
Q(s) = C(s) \left[\underline{1} + \underline{P}(s) \underline{C}(s) \right]^{-1}
$$

=
$$
\left[\underline{1} + C(s) P(s) \right]^{-1} C(s)
$$
 (4)

$$
\underline{C}(s) = \underline{Q}(s) \left[\underline{I} - \underline{P}(s) \underline{Q}(s) \right]^{-1}
$$

=
$$
\left[\underline{I} - \underline{Q}(s) \underline{P}(s) \right]^{-1} \underline{Q}(s)
$$
 (5)

$$
L_{\alpha}(s) = [L + P(s)C(s)]^{-1} = L - P(s)Q(s)
$$
 (6)

$$
E_{\frac{1}{2}}(s) = [\underline{I} + \underline{C}(s) \underline{P}(s)]^{-1} = \underline{I} - \underline{Q}(s) \underline{P}(s)
$$
 (7)

hinges on the fact that the sensitivity matrices S_o (s) and S_i (s), given by Eqs. (6) and (7), are linearly related to the matrix $Q(s)$. Thus, by shaping $Q(s)$, for any given $\underline{P}(s)$, one can obtain "good" sensitivity ma-Model Based Compensators (MBC) and specialize them to any given P(s), one can obtain "good" sensitivity ma-
the class of LOG/LTR compensators: many results not trices, and then evaluate the dynamic compensator C(s) the class of LQG/LTR compensators; many results not trices, and then evaluate the dynamic compensator C(s)
readily apparent in the literature are stated without using Eq. (5). The resultant closed loop transfer ma-

$$
\underline{G}_{CL}(s) = \underline{I} - \underline{S}_0(s) \tag{8}
$$

2. THE ZAMES PARAMETRIZATION IDEALLY, one would like to have S (jw)=0 for all w. In this section we summarize the Q-parametrization However, this is not possible, and approximations always introduced by Zames [1]; see also [2].

fer-Recovery (LTR) $[4]$, $[5]$, $[6]$. It is quite helpful following closed-loop transfer matrix, G_{CL} (s), from the the closer of Model-Based-Compensators
reference input \underline{r} (s) to the output \underline{r} (s), i.e. (MBC); the MBC nonmeclature is non-standard.

 (1) 3.1 The Open-Loop Plant

where time domain description of the open-loop plant

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$$
\underline{y}(t) = \underline{C} \underline{x}(t)
$$

with $\underline{x}(t) \in R^n$, $\underline{u}(t) \in R^m$, $\underline{v}(t) \in R^m$. Define the nxm matrix ¢(s) by where the matrix K is the solution of the algebraic

$$
\underline{\Phi}(s) \stackrel{\Delta}{=} (s\underline{I} - \underline{A})^{-1}
$$
 (10)

Evidently, the open-loop plant transfer matrix $\underline{P}(s)$ is $-\frac{0}{p}$

$$
P(s) = C(sI - A)^{-1}B = C \Phi(s)B
$$
 (11)

3.2 Model Based Compensators (MBC)

In reference to the feedback system in Fig. 1 we and that the LQG compensator transfer matrix $C_{1,\infty}(s)$ define a model based compensator (MBC) with the trans- see Eq. (12)- approaches fer matrix

$$
\underline{C}(s) = \underline{C}_{\text{HBC}}(s) = \underline{C}[s\underline{I} - \underline{A} + \underline{B} \underline{G} + \underline{H} \underline{C}]^{-1} \underline{H}
$$
\n
$$
= \underline{C}[\underline{\Phi}^{-1}(s) + \underline{B} \underline{G} + \underline{H} \underline{C}]^{-1} \underline{H}
$$
\n(12) Thus, the LOG/L

tionship of the MBC to the LQG compensator is apparent; there are no right-half plane pole-zero cance
if we got $r(s) = A/(s)-0$ in Fig. 1, then $y(s)$ would re-
The resulting closed-loop transfer matrix if we set $r(s) = d$ (s)=0 in Fig. 1, then $v(s)$ would represent the innovations vector, and $\hat{x}(s)$ the state es- \overline{G} timate. Also, one can think of the MBC in Fig. 2 as been obtained by cascading a non-minimal stable Luenberger observer with a suitable control gain, G , that may be obtained from a pole-placement algorithm

tem shown in Figure 3.
Note that all parameters in C_{MBC}(s) in Eq. (12) and the corresponding Zames parametrization Q(s)=Q^{LQG}(s) equa
Fig. 2 are fixed, except for the mxn filter gain matrix the now be readily calculated H , and the nxm control gain matrix G . If $[A, B]$ is stabilizable and $[A, C]$ is detectable, then it is wellknown (from non-minimal observer theory and pole-placement theory) that there exist gain matrices \underline{H} and \underline{G} As $\beta \rightarrow \infty$, we deduce from Eqs. (18) and (20) that such that the closed loop system of Fig. I is stable using the MBC of Eq. (12).
Next we present the closed-loop transfer function

NBC
 $\frac{LQ}{C}$ (s), γ (s) = $\frac{C}{C_L}$ (s) $\frac{LQ}{C_L}$ (s) associated with the feed-

http://ext.control system of Fig. 1 using the MBC of Fig. 2. Fig. 3 followed by the plant inverse P^{-1} (s). back control system of Fig. 1 using the MBC of Fig. 2. Tedious algebra and the use of the matrix inversion :Tedious algebra and the use of the matrix inversion In the LQG/LTR design procedure the filter gain ma-

$$
C_{\text{L}}^{\text{RBC}}(s) = \underline{P}(s) \left[\underline{I} + \underline{G} \ \underline{\Phi}(s) \underline{B} \right]^{-1} \underline{G} \ \underline{\Phi}(s) \underline{H} \left[\underline{I} + \underline{C} \ \underline{\Phi}(s) \underline{H} \right]^{-1}
$$
\n
$$
= \underline{P}(s) \left[\underline{I} + \underline{G} \ \underline{\Phi}(s) \underline{B} \right]^{-1} \underline{G} \ \underline{\Phi}(s) \underline{H} \left[\underline{I} + \underline{C} \ \underline{\Phi}(s) \underline{H} \right]^{-1} \tag{13}
$$

that the $Q(s)$ matrix in the Zames representation $[1]$ in the frequency domain. Indeed, at is possible to select H so that is as follows for MBC derived control systems

$$
Q(s) = Q^{MBC}(s) = \left[\underline{1} + \underline{G} \Phi(s) \underline{B}\right]^{-1} \underline{G} \Phi(s) \underline{H} \left[\underline{1} + \underline{C} \Phi(s) \underline{H}\right]^{-1} \qquad (14) \qquad \underline{C} \Phi(j\omega) \underline{H} \left[\underline{1} + \underline{C} \Phi(j\omega) \underline{H}\right]^{-1} \approx \underline{I} \tag{23}
$$

It is the selection of the constant filter gain matrix H and of the constant control gain matrix G that would $Q^{LQG}(j\omega) \approx p^{-1}(j\omega)$

3.3 LQG/LTR Compensators

The structure of an LQG compensator is identical to that of an MBC (see Figure 2). What distinguishes an LQG compensator from an MBC is the specific way that by compensator from an rice is the spectric way that
one calculated the Zames Q matrix for the class
noncloted algobraic Digital constitutions for the IOC one of LQC/LTR designs. In the opinion of the author, the propriate algebraic Riccati equations for the LQG op-
timization problem [4] [7] These will not be reviewed Zames decomposition offers no particular insights into timization problem [4], [7]. These will not be reviewed, Zames decomposition offers no particular insights in
since as embitized IOC concernation does not provide any. The LQG/LTR problem, except that it confirms (in a since an arbitrary LQG compensator does not provide any the LQG/LIR problem, except that it confirms (in a
more incight than that contained in Eq. (14) cound-about way) that in the absence of non-minimum more insight than that contained in Eq. (14).

P(s) is as follows: with LTR [5], [8]. Suppose that the open-loop plant has $\dot{x}(t) = \underline{A} x(t) + \underline{B} u(t)$
 $\dot{x}(t) = \underline{A} x(t) + \underline{B} u(t)$ and the control rain matrice, suppose that the control qain matrice. zeros. Furthermore, suppose that the control gain matrix G in Fig. 2 and in Eq. (12) is computed as follows:

$$
\underline{G} = \frac{1}{\rho} \underline{B}^{-1} \underline{K} \; ; \qquad \rho > 0 \tag{15}
$$

Riccati equation

$$
\underline{0} = -\underline{K} \underline{A} - \underline{A} \underline{K} - \beta \underline{C} \underline{C} + \frac{1}{\rho} \underline{K} \underline{B} \underline{B} \underline{K}
$$
 (16)

In particular in the LTR procedure we let $\beta \rightarrow \infty$. Then, the results in Refs. [5], [8] indicate that as $\beta \nrightarrow \infty$, the control gain matrix G has the property that

$$
\frac{1}{\beta} \quad \underline{G} \rightarrow \underline{W} \subseteq ; \quad \underline{W} = \text{orthonormal matrix} \tag{17}
$$

$$
C(s) = C(s) + C \Phi(s) B^{-1}C \Phi(s) H
$$
\n
$$
C(s) = C_{LQG}(s) + C \Phi(s) B^{-1}C \Phi(s) H
$$
\n
$$
= E^{-1}(s) C \Phi(s) H
$$
\n(18)

Thus, the LQG/LTR compensator G_{c} (s) performs an ap-A block diagram of $C_{\text{top}}(s)$ is shown in Fig. 2. The proximate inversion of the plant $P(s)$; the assumption A block diagram of C_{MBC} (s) is shown in Fig. 2. The proximate inversion of the plant⁻² P(s); the assumption block diagram is drawn in such a way so that the rela-
block diagram is drawn in such a way so that the r

LQG(s) tends to
CL
CL
CL
LCG(s)
$$
\rightarrow
$$
 C $\underline{\phi}(s)$ H[I \rightarrow C $\underline{\phi}(s)$ H]⁻¹ (19)

which is the closed-loop transfer of the feedback sys-

$$
LQG(s) = C_{LQG}(s) \left(\underline{r} + \underline{P}(s) C_{LQG}(s) \right)^{-1}
$$
 (20)

$$
\underline{L}^{LQG}(s) \to \underline{p}^{-1}(s) \underline{C} \underline{\Phi}(s) \underline{H} [\underline{I} + \underline{C} \underline{\Phi}(s) \underline{H}]^{-1}
$$
 (21)

trix H is calculated by suitable solution of a Kalman Filter design problem so that the loop transfer matrix $C \Phi(s)H$ in Fig. 3 and the closed-loop transfer function

$$
H^{\frac{1}{2}} \t(13) \tC \phi(s) H[I+C \phi(s)H]^{\frac{1}{2}} \t(22)
$$

By comparing Eqs. (3) and (13) we can readily conclude meet the posed performance and robustness specifications
that the O(s) matrix in the Zames representation [1] in the frequency domain. Indeed, at low frequencies,

$$
\underline{C} \underline{\Phi}(\mathbf{j}\omega) \underline{H}[\underline{I} + \underline{C} \underline{\Phi}(\mathbf{j}\omega) \underline{H}]^{-1} \approx \underline{I}
$$
 (23)

resulting in

 $\mu_{\rm{max}}=1$. If $\mu_{\rm{max}}=1$ is the this-dependent function of the this-dependent function $\mu_{\rm{max}}=1$

$$
O^{\text{LQG}}(\text{i}\omega) \approx P^{-1}(\text{i}\omega) \tag{24}
$$

shape $Q(s)$.
and output sensitivity which is almost zero -see Eq.(6)-

$$
S_{0}(\mathbf{j}\omega) \approx \underline{0} \tag{25}
$$

4. CONCLUDING REMARKS

A greater insight can be obtained in Eq. (14).
A greater insight can be obtained for LQG designs . eedback loops. However, this has been known an_i-way.

5. REFERENCES

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on integratio Central Vol. 2012 (s) 2 April (s) (s) (s) (s) (s) (s) + (s) on Automatic Control, Vol. AC-26, No. 2, April $\frac{r(s)}{r(s)}$ $\frac{q(s)}{s}$ $\frac{q(s)}{s}$ P(s) 1981, pp. 301-320.
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Figure 3: Block diagram of the limiting feedback control system obtained by the LQG/LTR procedure. The constant matrix H is the **Kalman Filter gain matrix.**

