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Optimal MISO Repetitive Control System Design Using Mixed Time and Frequency Domain Criteria

This paper presents an optimal repetitive controller design approach when both frequency and time domain criteria are considered. These criteria include limited control force, convergence rate, and robustness. In particular, the design formulation is based on MISO (multiple-input-single-output) plants which aim at solving nodal point problems encountered in flexible dynamic systems. Nodal points represent physical constraints for feedback control, especially when their frequencies are within the control bandwidth. If multiple control paths with distinct nodal frequencies are available, the proposed method is able to design controllers to compensate the effect and reach certain design objectives. The ellipsoid algorithm is used to calculate controllers' parameters. Experiments were conducted to demonstrate the proposed method.

1 Introduction

Repetitive control algorithms using internal model principle (Francis and Wonham, 1975) have been studied by many researchers (Tomizuka et al., 1988; Hara et al., 1988). Applications such as noncircular cutting (Tsao and Tomizuka, 1988), disk drive tracking (Chew and Tomizuka, 1990), repeated motion of robot manipulators (Sadegh et al., 1990), material testing (Shaw and Srinivasan, 1993), and active harmonic noise cancellation (Hu, 1995a) have been reported in literature. When a plant has unstable zeros near the unit circle, it has been shown that conventional design approaches such as adding zero-phase filters (Hu et al., 1995b) may not give satisfactory result. Further, several design issues need to be addressed when dealing with distributed systems or when the fundamental period of the disturbance is high:

1 Robustness—especially around nodal points (those zeros close to the unit circle) where accurate identification is difficult due to a low S/N ratio.

2 Convergent speed—closed-loop poles inevitably approaching the unit circle due to a high fundamental period.

3 Limited control force—system may become unstable when the control force is saturated.

All these issues clearly indicate that an optimization scheme is necessary to select proper control parameters.

Nodal points in distributed parameter systems are actually transmission zeros of the plant. Attempting to cancel disturbances whose frequency contents are in the neighborhood of these zeros results in a very large control force. An obvious way to avoid this problem is to change transmission zeros' locations by properly placing actuators or sensors. Alternatively, for MIMO systems, it is possible to arrange actuators such that transmission zeros of all input paths are distinct. In this case, each controller should "give up" its control effort around its transmission zeros. These considerations represent various constraints when performing optimization.

This paper proposes an optimal design approach to an MISO (multiple-input-single-output) repetitive control problem. It is

motivated from an experimental observation of active noise control in ducts. When mounting control speakers on the side of a duct, there is a great chance that the generated sound wave of some particular frequencies will be self-destructed (transmission zeros). These frequencies are closely related to the dimensions and boundary conditions of the duct (Hu, 1995). From the explanation in preceding paragraphs, the control system's performance is limited when some of these frequencies fall into the bandwidth of design. Intuitively, multiple speakers can be arranged such that each speaker's transmission zeros are different to others. For example, as shown in Fig. 1, speaker B will have different transmission zeros from A when an additional segment of duct is added.

The organization of this paper is the following: Section 2 describes the problem formulation; Section 3 presents the solution procedures and Section 4 explains the experiments conducted to verify the design methods.

2 MISO Repetitive Control System Design

The block diagram of an *m*-input-1-output repetitive control system is depicted in Fig. 2. In Fig. 2, P_i is the dynamic of each control path, C_i is the compensator for each path and Δ_i represents multiplicative uncertainty, i = 1, 2, ..., m. d represents a periodic disturbance and *M* is the repetitive signal generator. It is assumed that both P_i 's and Δ_i 's are stable dynamics. Let $\overline{P_i} = P_i(1 + \Delta_i), i = 1, 2, ..., m$, and the multiplicative uncertainty Δ_i satisfy

$$\Delta_{i}(\omega) \Big| = \left| \frac{\bar{P}_{i} - P_{i}}{P_{i}} \right|_{z=e^{j\omega}} \leq \alpha(\omega) \; \forall \omega[0, \pi],$$

and i = 1, 2..., m

where $\alpha(\omega)$ is a conservative estimate of the uncertainty bound.

2.1 Stability and Robustness Analysis. When $\Delta = 0$, the internal stability of Fig. 2 requires that

$$\frac{-z^{-N}\mathbf{C}}{1-z^{-N}(1-\mathbf{PC})} \in \mathbf{RH}_{\infty}$$
(2.1)

A sufficient condition to guarantee stability by using the small gain theorem is (**P** is assumed stable)

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Fig. 1 (a) Same transmission zeros of A and B; (b) different transmission zeros

(C1)
$$z^{-N}\mathbf{C} \in \mathbf{RH}_{\infty}$$
 (2.2)

(C2)
$$||1 - \mathbf{PC}||_{\infty} < 1$$
 (2.3)

Note that from Eq. (2.2), C is allowed to be non-causal. When $\Delta \neq 0$, other than these two conditions, an additional condition should be added,

(C3)
$$\left| \frac{z^{-N} \mathbf{PC}}{1 - z^{-N} (1 - \mathbf{PC})} \right|_{z=e^{j\omega}} < \frac{1}{\sup_{i} |\Delta_i(\omega)|} \forall \omega$$
 (2.4)

At repetitive frequencies, i.e., $\omega = 2k\pi/N$, k = 0, 1, ..., N/2, we have,

$$\left|\frac{z^{-N}\mathbf{PC}}{1-z^{-N}(1-\mathbf{PC})}\right|_{z=e^{j2k\pi/N}}=1$$

From Eq. (2.4), the maximum allowable perturbation is also 1 which limits the robust performance. Therefore, a filter is added to sacrifice disturbance rejection ability or tracking accuracy in order to gain robustness, i.e., the repetitive signal generator is changed into

$$M = \frac{qz^{-N}}{1 - qz^{-N}}$$

where q is the selected filter (Tsao and Tomizuka, 1988). By adding this filter, the stability and robustness conditions (C1) to (C3) are modified as

$$(C4) \quad z^{-N}q\mathbf{C} \in \mathbf{RH}_{\infty} \tag{2.5}$$

(C5)
$$\|1 - \mathbf{PC}\|_{\infty} < 1$$
 (2.6)

(C6)
$$\left|\frac{z^{-N}q\mathbf{PC}}{1-z^{-N}q(1-\mathbf{PC})}\right|_{z=e^{j\omega}} < \frac{1}{\sup_{i} |\Delta_i(\omega)|} \quad \forall \omega \ (2.7)$$

A more relaxed stability condition of (C5) should be $||q(1 - \mathbf{PC})|_{\infty} < 1$ since $|q(\omega)| < 1$. However, implementing (C5) in design will result in more robustness since when $|q(\omega)|$ is small,

$$\left|\frac{z^{-N}q\mathbf{PC}}{1-z^{-N}q(1-\mathbf{PC})}\right|\approx |q\mathbf{PC}|<2|q|. \quad (2.8)$$

Thus, if q is selected as an FIR low-pass filter, more high-frequency model uncertainty can be tolerated. In fact, an ideal selection of q should be

$$q(\omega) = 1, \ \omega \in \Omega, \text{ and } |q(\omega)| = 0, \ \omega \notin \Omega.$$
 (2.9)

where Ω is the desired frequency range of disturbance rejection or tracking.

2.2 Performance Analysis. Two different performance requirements are investigated in this section. The first one is the speed of convergence. Assuming the z-transform of a periodic disturbance d is (denoted as \hat{d})

$$\hat{d} = \frac{F}{1 - z^{-N}}, \quad F = f_0 + f_1 z^{-1} + \ldots + f_{N-1} z^{-N+1}.$$
 (2.10)

The error of the repetitive control system, with the filter q added, is derived as (Fig. 2),

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$$\hat{e} = \frac{(1 - qz^{-N})}{[1 - z^{-N}q(1 - \mathbf{PC})]} \frac{F}{(1 - z^{-N})}$$
(2.11)

When $q(e^{j2k\pi/N}) \approx 1$ and N is large, it can be seen that the convergence of the error at repetitive frequencies depends on the closed-loop poles. To reduce the design complexity, the plant **P** is assumed to be stable and its poles are canceled directly. Further, if q and **PC** are FIR filters (by restricting the structure of **C** (see Eq. (3.4) later)), the closed-loop characteristic equation is

$$z^{N+N_p} + z^{N_p} q(z^{-1}) [1 - \mathbf{P}(z^{-1})\mathbf{C}(z^{-1})] = 0 \quad (2.12)$$

where N_p is the number of poles of $q(z^{-1})\mathbf{P}(z^{-1})\mathbf{C}(z^{-1})$ at origin. Notice that $q(z^{-1})\mathbf{P}(z^{-1})\mathbf{C}(z^{-1})$ may be noncausal. Let the characteristic roots be $re^{j\omega}$, 0 < r < 1, by maximum modulus theorem, we have

$$r^{N+N_{p}} = |z^{N_{p}}q(z^{-1})[1 - \mathbf{P}(z^{-1})\mathbf{C}(z^{-1})]|_{z=re^{j\omega}}$$

$$\leq \max_{0 \leq \omega \leq \pi} |z^{N_{p}}q(z^{-1})[1 - \mathbf{P}(z^{-1})\mathbf{C}(z^{-1})]|_{z=e^{j\omega}}$$

$$\leq ||1 - \mathbf{P}(z^{-1})\mathbf{C}(z^{-1})||_{\infty} \qquad (2.13)$$

since $z^{N_p}q(z^{-1})\mathbf{P}(z^{-1})\mathbf{C}(z^{-1})$ is analytic inside the unit circle and q is low pass (Eq. (2.9)). As a result, a proper constraint to limit the close-loop spectral radius is then,

$$\|1 - \mathbf{PC}\|_{\infty} < \delta$$

If the performance range (i.e., Ω in Eq. (2.9)) is considered, a more detailed constraint can be constructed as

(C7)
$$\sup_{\omega \in \Omega} |1 - \mathbf{PC}|_{z=e^{j\omega}} < \delta \text{ and}$$
$$\sup_{\omega \notin \Omega} |1 - \mathbf{PC}|_{z=e^{j\omega}} < 1 \qquad (2.14)$$

Second, to avoid control force saturation, it is necessary to constrain controllers' output levels according to their physical limits. An obvious choice is the A-norm minimization (or l_1 minimization) (Vidyasagar, 1986; Dahleh and Pearson, 1988), i.e., from Fig. 2,

$$\|u_i\|_{\infty} \le \|T_{u_i d}\|_{A} \|d\|_{\infty}, \quad i = 1, 2, \dots, m,$$

where $T_{u_i d} = \frac{-z^{-N}C_i}{1 - z^{-N}(1 - \mathbf{PC})}.$

this leads to the following condition (with the filter q added):

(C8)
$$||T_{u_id}||_A \le \rho_i, \quad T_{u_id} = \frac{-z^{-N}qC_i}{1-z^{-N}q(1-\mathbf{PC})}$$

for $i = 1, 2, ..., m.$ (2.15)

where $\rho_i \|d\|_{\infty}$ is less than the saturation limit of the actuator *i*.

3 Mixed Time and Frequency Domain Optimization

As explained by the design conditions (C4-C8), controllers can be calculated by optimizing various time and frequency domain criteria (Boyd et al., 1988; Helton and Sideris, 1988; Khargonekar and Rotea, 1991; Sznaier, 1994). For example, if



Fig. 2 Block diagram of an MISO repetitive control system ($P = [P_1 P_2 \dots P_m]$, $C = [C_1 C_2 \dots C_m]^T$, $\Delta = \text{diag} (\Delta_1, \Delta_2, \dots, \Delta_m)$ and $u = [u_1, u_2, \dots, u_m]^T$)

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we want to minimize the control forces under the constraint of a designated convergence speed, we could select:

$$\min_{z^{-N}\mathbf{C}\in\mathbf{RH}_{\infty}}\max_{i}\|T_{u_{i}d}\|_{A}$$
(3.1)

subject to
$$\|1 - \mathbf{PC}\|_{\infty} \le \delta, \quad \delta < 1$$
 (3.2)

Similar problems of this type are in the area of convex optimization. Two commonly-used algorithms to solve these problems are the decent and nondecent methods. Non-decent methods such as the ellipsoid algorithm (Boyd and Barratt, 1991) are particularly suitable for nondifferentiable norms like H_{∞} or l_1 . To implement the ellipsoid algorithm, it is necessary to calculate generalized gradients of cost and constraint functions at each iteration (Boyd and Barratt, 1991; Clarke, 1975; Polak, 1987). In this paper, we present the solutions in finite dimensional space (i.e., suboptimal solutions). To realize the numerical calculation, it is necessary to formulate the cost function and constraints in vector notations as explained in the following sections.

3.1 Vector Space Representations of the Control System. The plant is represented as the following,

$$\mathbf{P} = [P_1 \ P_2 \ \dots \ P_m], \ P_i = z^{-n_i} B_i(z^{-1}) G_i(z^{-1}),$$
$$i = 1, \dots, m \quad (3.3)$$

where n_i is the delay steps of P_i , G_i is minimum phase, and B_i contains all zeros outside the circle centered at origin with radius $\beta < 1$. The selection of β is to avoid stable pole-zero cancellations close to the unit circle since those zeros may not be accurately identified due to poor S/N ratios (nodal points as explained before). Let n be the maximum order of B_i 's, $i = 1 \sim m$, and

$$B_i(z^{-1}) = b_0^i + b_{-1}^i z^{-1} + \ldots + b_{-n}^i z^{-n}$$

where some of the coefficients may be zero if the B_i 's order is less than n.

The structure of C is selected as

$$\mathbf{C} = [C_1 \quad C_2 \quad \dots \quad C_m], \ C_i = z^{n_i} D_i(z, z^{-1}) G_i^{-1}(z^{-1}),$$

$$i = 1, \dots, m \quad (3.4)$$

where

$$D_{i}(z, z^{-1}) = d_{i}^{i} z^{l} + d_{l-1}^{i} z^{l-1} + \dots + d_{-l+1}^{i} z^{-l+1} + d_{-l}^{i} z^{-l} \quad (3.5)$$

By using Eq. (3.5), we shall perform optimization in finite dimensional space. Further, the filter q is assumed to be

$$q(z, z^{-1}) = q_r z^r + q_{r-1} z^{r-1} + \ldots + q_0 + \ldots + q_{-r} z^{-r}.$$
 (3.6)

The following formulae will be used in derivations later.

(a)
$$D_i(z, z^{-1}) = [z^i \ z^{l-1} \dots z^{-l}] \mathbf{d}_i,$$

where $\mathbf{d}_i = [d_i^i \ d_{l-1}^i \ \dots \ d_{-l}^i]^T$ (3.7)

(b)
$$z^{-(r+l)}q(z, z^{-1})D_i(z, z^{-1}) = [1 \ z^{-1} \dots z^{-2(l+r)}]\mathbf{Q}\mathbf{d}_i,$$

where
$$\mathbf{Q} = \begin{pmatrix} q_{r} & \bigcirc \\ | & \neg & q_{r} \\ q_{-r} & | \\ \bigcirc & q_{-r} \end{pmatrix} \in \mathbf{R}^{(2l+2r+1)\times(2l+1)}$$
 (3.8)

(c)
$$1 - \sum_{i=1}^{m} B_i D_i = [z^l \ z^{l-1} \ \dots \ z^{-(l+n)}](\mathbf{e} - \mathbf{Bd})$$
, where
 $\mathbf{B} = [\mathbf{B}_1 \ \mathbf{B}_2 \ \dots \ \mathbf{B}_m] \in \mathbf{R}^{(n+2l+1) \times m(2l+1)}$

$$\mathbf{B}_{i} = \begin{pmatrix} b_{0}^{i} & \bigcirc \\ \vdots & b_{0}^{i} \\ b_{-n}^{i} & \vdots \\ \bigcirc & b_{-n}^{i} \end{pmatrix} \in \mathbf{R}^{(n+2l+1)\times(2l+1)}$$
$$\mathbf{d} = \begin{pmatrix} \mathbf{d}_{1} \\ \mathbf{d}_{2} \\ \vdots \\ \mathbf{d}_{m} \end{pmatrix} \in \mathbf{R}^{m(2l+1)\times 1}, \quad \mathbf{d}_{i} = \begin{bmatrix} d_{i}^{i} \\ d_{i-1}^{i} \\ \vdots \\ d_{-l}^{i} \end{bmatrix},$$
$$i = 1, \dots, m, \quad \mathbf{e} = \begin{bmatrix} \bigcirc_{l\times 1} \\ 1 \\ \bigcirc_{(n+l)\times 1} \end{bmatrix}$$
(3.9)

(d)
$$q(1 - \sum_{i=1}^{n} B_i D_i)$$

= $[z^{l+r} \ z^{l+r-1} \ \dots \ z^{-(l+n+r)}] \overline{\mathbf{Q}} (\mathbf{e} - \mathbf{Bd}),$

where
$$\mathbf{\tilde{Q}} = \begin{pmatrix} q_{r} & 0 \\ q_{r} \\ q_{r} \\ q_{-r} & | \\ 0 & q_{-r} \end{pmatrix} \in \mathbf{R}^{(n+2l+2r+1)\times(n+2l+1)}$$
 (3.10)

3.2 Frequency Domain Design. The related functions in frequency domain are (C6) and (C7). Since the filter q can be shaped to gain robustness (Eq. (2.8) and (2.9)), we focus the discussion on (C7). Using the vector notations in Eqs. (3.7) – (3.10), we have

$$\left|1 - \sum_{i=1}^{m} B_i D_i\right|_{z=e^{j\omega}}^2 = \operatorname{Re} \left\{ (\mathbf{\Theta}^T \mathbf{e} - \mathbf{\Theta}^T \mathbf{Bd})^* (\mathbf{\Theta}^T \mathbf{e} - \mathbf{\Theta}^T \mathbf{Bd}) \right\}$$
$$= (\mathbf{e} - \mathbf{Bd})^T \operatorname{Re} \left\{ \mathbf{\Theta} (e^{-j\omega}) \mathbf{\Theta}^T (e^{j\omega}) \right\} (\mathbf{e} - \mathbf{Bd}) \quad (3.11)$$

where

$$\Theta(e^{j\omega}) = [e^{jl\omega}e^{j(l-1)\omega} \dots e^{-j(l+n)\omega}]^{T}$$

Re { $\Theta(e^{-j\omega})\Theta^T(e^{j\omega})$ }

$$= \operatorname{Re} \left\{ \begin{bmatrix} e^{-jl\omega} \\ e^{-j(l-1)\omega} \\ \vdots \\ e^{j(l+n)\omega} \end{bmatrix} [e^{jl\omega} e^{j(l-1)\omega} \dots e^{-j(l+n)\omega}] \right\}$$
$$= \begin{bmatrix} 1 & \cos \omega & \cos 2\omega & \cdots & \cos(2l+n)\omega \\ \cos \omega & 1 & \cos \omega & \vdots \\ \cos 2\omega & \cos \omega & 1 & \vdots \\ \vdots & \vdots & \vdots & \cos \omega \\ \cos (2l+n)\omega & \cos (2l+n-1)\omega & \vdots & 1 \end{bmatrix}$$

By knowing that

$$\|1 - \mathbf{PC}\|_{\infty}^{2} = \|1 - \sum_{i=1}^{m} B_{i} D_{i}\|_{\infty}^{2}$$
$$= \sup_{\omega} |1 - \sum_{i=1}^{m} B_{i} D_{i}|_{z=e^{j\omega}}^{2}, \qquad (3.12)$$

the generalized gradient g for $||1 - \mathbf{PC}||_{\infty}^2$ is derived by the maximum rule (Boyd and Barratt, 1991) as

$$g = -2\mathbf{B}^{T} \operatorname{Re} \left\{ \Theta(e^{-j\omega_{0}}) \Theta^{T}(e^{j\omega_{0}}) \right\} (\mathbf{e} - \mathbf{Bd}) \quad (3.13)$$

where

$$\omega_0 \in \{ \omega \in \mathbb{R} \mid \|1 - \mathbb{PC}\|_{\infty} = \|1 - \mathbb{PC}\|_{z=e^{j\omega}} \}$$

Equation (3.13) can be modified to include a weighting function. Other than the infinity norm, the H_2 norm is also a good

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index for minimizing the magnitude of (1 - PC) (under certain smooth conditions), i.e.,

$$||1 - \mathbf{PC}||_2^2 = (\mathbf{e} - \mathbf{Bd})^T (\mathbf{e} - \mathbf{Bd})$$
 (3.14)

The corresponding generalized gradient can be easily derived as

$$g(\mathbf{d}) = -2\mathbf{B}^{T}(\mathbf{e} - \mathbf{B}\mathbf{d}) \qquad (3.15)$$

There is a great flexibility of using the idea of \mathbf{H}_2 norm in selecting the performance range (i.e., Ω in Eq. (2.9)). For example, if we wish to improve convergence at $\omega \in [\omega_1, \omega_2]$ and gain robustness at $\omega \in [\omega_3, \omega_4]$, we could use the following formula

$$\min \{ \| \overline{W}_1 (1 - \mathbf{PC}) \|_2 + \| \overline{W}_2 \mathbf{PC} \|_2 \}$$
(3.16)

where

Ú

$$\overline{V}_{1}(\omega) = \begin{cases} 1 & \forall \omega \in [\omega_{1} \quad \omega_{2}] \\ 0 & \text{otherwise} \end{cases} \text{ and}$$
$$\overline{W}_{2}(\omega) = \begin{cases} 1 & \forall \omega \in [\omega_{3} \quad \omega_{4}] \\ 0 & \text{otherwise} \end{cases}$$

As a result, the representation of Eq. (3.16) in vector space are

$$\min_{D_i} \left\{ \int_{\omega_1}^{\omega_2} |1 - \sum_{i=1}^m B_i D_i|^2 d\omega + \int_{\omega_3}^{\omega_4} |\sum_{i=1}^m B_i D_i|^2 d\omega \right\} (3.17)$$

where

$$\int_{\omega_1}^{\omega_2} |1 - \sum_{i=1}^m B_i D_i|^2 d\omega + \int_{\omega_3}^{\omega_4} |\sum_{i=1}^m B_i D_i|^2 d\omega$$

= $(\mathbf{e} - \mathbf{Bd})^T \mathbf{W}_1 (\mathbf{e} - \mathbf{Bd}) + \mathbf{d}^T \mathbf{B}^T \mathbf{W}_2 \mathbf{Bd}$ (3.18) where

$$T_{i} = \frac{z^{-(r+l)}qD_{i}}{1 - z^{-N}q(1 - \sum_{i=1}^{m} D_{i}B_{i})}$$

The optimization is now focused on $||T_i||_A$ instead of $||T_{u_id}||_A$. Suppose $\{y_i(k)\}$ represent the impulse response of T_i and $\{x(k)\}$ an impulse input sequence, we have

$$y_{i}(k) = (\mathbf{e} - \mathbf{B}\mathbf{d})^{T} \mathbf{\bar{Q}}^{T} \begin{bmatrix} y_{i}(k - N + l + r) \\ y_{i}(k - N + l + r - 1) \\ \vdots \\ y_{i}(k - N + l + r - n) \end{bmatrix}$$

+ $\mathbf{d}_{i}^{T} \mathbf{Q}^{T} \begin{bmatrix} x(k) \\ x(k-1) \\ \vdots \\ x(k-2l-2r) \end{bmatrix}$, $k = 0, 1, 2,$ (3.21)

To minimize the computational load, l (length of the filter D_i) and r (length of the filter q) are restricted to satisfy (l + r) < N/3. Under this restriction, the first N + l + r + 1 terms of the impulse response, i.e., $\{y_i(k)\}_{k=0}^{N+l+r}$, are

$$[y_i(0) \ y_i(1) \ \dots \ y_i(2l+2r)]^T = \mathbf{Q}\mathbf{d}_i$$
 (3.22a)

$$[y_i(2l+2r+1) \quad y_i(2l+2r+2) \quad \dots \quad y_i(N-l-r-1)]^T$$

= $\bigcirc_{(N-3l-3r-1)\times 1}$ (3.22b)

$$[y_i(N-l-r) \quad y_i(N-l-r+1) \quad \dots \quad y_i(N+l+r)]^T$$
$$= \mathbf{Y}_i \dot{\mathbf{Q}} (\mathbf{e} - \mathbf{Bd}) \quad (3.22c)$$

$$\Omega(\omega) = \begin{bmatrix} \omega & \sin \omega & \frac{\sin 2\omega}{2} & \cdots & \frac{\sin (2l+n)\omega}{2l+n} \\ \sin \omega & \omega & \sin \omega & \vdots \\ \frac{\sin 2\omega}{2} & \sin \omega & \omega & \vdots \\ \vdots & \vdots & \vdots & \sin \omega \\ \frac{\sin (2l+n)\omega}{2l+n} & \frac{\sin (2l+n-1)\omega}{2l+n-1} & \vdots & \omega \end{bmatrix}$$

and

$$\mathbf{W}_1 = \mathbf{\Omega}(\omega_2) - \mathbf{\Omega}(\omega_1), \, \mathbf{W}_2 = \mathbf{\Omega}(\omega_4) - \mathbf{\Omega}(\omega_3)$$

The generalized gradient is therefore derived as,

$$g = -2\mathbf{B}^T \mathbf{W}_1(\mathbf{e} - \mathbf{B}\mathbf{d}) + 2\mathbf{B}^T \mathbf{W}_2 \mathbf{B}\mathbf{d} \qquad (3.19)$$

3.3 Time Domain Design. In minimizing the A-norm indicated in (C8) (Eq. (2.15)), we limit the l_1 norm of the impulse response from disturbance input to control forces. Since a closed-form representation of $||T_{u,d}||_A$ is impossible due to infinite terms in the impulse response, a simplified approach is used. Observe first that (see Eqs. (2.15), (3.5), and (3.6))

$$\|T_{u_i d}\|_{A} = \left\| \frac{z^{-(r+i)} q D_i G^{-1}}{1 - z^{-N} q (1 - \sum_{i=1}^{m} B_i D_i)} \right\|_{A} \le \|T_i\|_{A} \|G_i^{-1}\|_{A}$$
(3.20)

where

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$$\mathbf{Y} = \begin{pmatrix} y_i(0) & 0 & \dots & \dots & \dots & 0 \\ y_i(1) & y_i(0) & 0 & \dots & \dots & 0 \\ \vdots & \vdots & \ddots & & \vdots \\ y_i(2l+2r) & y_i(2l+r-1) & \dots & y_i(0) & 0 & \dots & 0 \end{pmatrix}$$

$$\in \mathbf{R}^{(2l+2r+1)\times(n+2l+2r+1)}$$

Now the target function $||T_i||_A$ is approximated as

$$|T_i(\mathbf{d})||_A \approx \sum_{k=0}^{N+l+r} |y_i(k, \mathbf{d})|$$
(3.23)

According to Sznaier (1994), the error of this approximation can be estimated. Assume that $T_i(z^{-1})$ is analytic for every $|z| \ge \delta$, $0 < \delta < 1$, we have

$$|y_{i}(k)| = \left|\frac{1}{2\pi j} \oint_{|z|=\delta} T_{i}(z^{-1}) z^{k-1} dz\right| \le ||T_{i}||_{\infty,\delta} \delta^{k} \quad (3.24)$$

where

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Duct's length: 0.5m Cross Section: 0.15m x 0.15m Cut-off Frequency: 1153 Hz

Fig. 3 Experimental setup

$$||T_i||_{\infty,\delta} = \sup_{\omega} |T_i(z^{-1})|_{z=\delta e^{j\omega}}$$

Therefore, the approximation error can be derived as

$$\sum_{n=N+l+r+1}^{\infty} |y_i(k)| \le \frac{\|T_i\|_{\infty,\delta} \delta^{(N+l+r+1)}}{1-\delta}$$
(3.25)

From Eq. (3.25), the error is small if N + l + r + 1 is large or δ is small. Note that δ is the spectral radius of the closed loop system. Using the sum rule (Boyd and Barratt, 1991), the generalized gradient for the approximated function (Eq. (3.23)) is,

$$g_i(\mathbf{d}) = \sum_{k=1}^{2l+2r+1} \operatorname{sign} (y_i(k-1))\mathbf{v}_k$$

+ sign $(y_i(k+N-l-r-1))$
 $\times \{-\mathbf{B}^T \overline{\mathbf{Q}}^T \mathbf{w}_k + \mathbf{X}_k \overline{\mathbf{Q}}(\mathbf{e} - \mathbf{Bd})\}$ (3.26)

where

$$\mathbf{v}_k$$
 is the *k*-th column vector of
$$\begin{bmatrix} \bigcirc_{(i-1)(2l+1)\times(2l+2r+1)} \\ \mathbf{Q}^T \\ \bigcirc_{(m-i)(2l+1)\times(2l+2r+1)} \end{bmatrix}$$

 \mathbf{w}_k is the *k*-th column vector of \mathbf{V}_i^T (see Eq. (3.22c)) and

 $\mathbf{X}_{k} = \begin{bmatrix} \mathbf{v}_{k} & \mathbf{v}_{k-1} & \dots & \mathbf{v}_{1} & \bigcirc_{m(2l+1)\times(n+2l+2r+1-k)} \end{bmatrix}$

Using the results derived in Sections 2 and 3, a design strategy was implemented. Experiments were conducted on a rectangular duct made of plastic glass. A DSP system based on TMS320c31 was installed to calculate the control signals (Fig. 3). The sampling rate was chosen as 4 kHz. The calibration value of the error microphone is 49 mV/Pascal. A noise speaker was installed to simulate a noise source. Two control speakers were



Fig. 4 Frequency response of plant $P_1(z^{-1})$ and $P_2(z^{-1})$

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Table 1Unstable zeros of both plants (stability margin =0.98; *: transmission zeros within the bandwidth)

	$B_1(z^{-1})$		$B_2(z^{-1})$	
Unstable zeros	1 2 3* 4 5	$\begin{array}{r} -3.4127 \\ 1.0302 \\ 0.7336 \pm 0.6598j \\ -0.4907 \pm 0.8598j \\ -0.1766 \pm 0.9751j \end{array}$	1 2 3* 4* 5	$\begin{array}{r} -2.0429\\ 1.0341\\ 0.9366\pm 0.3031j\\ 0.4463\pm 0.8816j\\ -0.3009\pm 0.9545j\end{array}$

mounted on the side of the duct. One of them included an additional duct segment to prevent identical nodal points for both control paths. It should be emphasized that the setup is not a 1-D problem because the length (0.5m) is too short to ignore evanescent modes (Morse and Ingard, 1968; Doak, 1973). Hence, the behavior of nodal points is a 3-D phenomenon in this case.

The plant P_1 (speaker 1) and P_2 (speaker 2) are identified using time-domain least square algorithms with frequency weighting (weighted band = $0 \sim 1$ KHz). Numerical calculations were carried out by the identification toolbox of MAT-LAB. Their frequency response are depicted in Fig. 4. The total number of parameters for P_1 and P_2 are 56 (4 delay steps; 30 poles; 26 zeros) and 52 (4 delay steps; 28 poles; 24 zeros), respectively. Other than 0 Hz, plant P_1 has a transmission zero around 470 Hz and P2 around 200 and 700 Hz. They are different because a small passage to the duct is added to speaker 2 (Fig. 3). The zero at 0 Hz are common to both plants. This is the zero caused by AC-coupling of the power and microphone amplifier and is impossible to remove. As shown later, special treatment has to be done to prevent DC drift in control signals. To avoid stable pole-zero cancellation near the unit circle, the stability margin is selected to be 0.98. As a result, the nonminimum phase parts of both plants are listed in Table 1 (see Eq. (3.3)).

Figure 5 shows the spectrum of the periodic noise tested. Its fundamental period is 245 samples (N = 245; fundamental frequency is 16.32 Hz under 4 kHz sampling). From Fig. 5, it can be seen that the noise signal is mostly within 500 Hz. Therefore, a zero-phase low-pass filter with cut-off frequency 666.7 Hz is selected for the filter q, i.e.,

$$q(z, z^{-1}) = 0.0019z^9 - 0.0029z^8 - 0.0067z^7 + 0.0088z^6$$

+ 0.0261z^5 - 0.0182z^4 - 0.0784z^3
+ 0.0266z^2 + 0.3080z + 0.4698 + 0.3080z^{-1}
+ 0.0266z^{-2} - 0.0784z^{-3} - 0.0182z^{-4}
+ 0.0261z^{-5} + 0.0088z^{-6} - 0.0067z^{-7}



Fig. 5 Spectrum of the harmonic noise (recorded from a four cylinder engine)



Fig. 6 Frequency responses of B/s (Eq. (3.3)) with added filters to give up control around nodal points

$$-0.0029z^{-8} + 0.0019z^{-9}$$
 (4.1)

Within this bandwidth, the transmission zero of plant P_1 is close to the 29th harmonic frequency of the noise (473.47 Hz) and P_2 to the twelfth (195.91 Hz) and 43rd (702.04 Hz) harmonic frequency. Like the dc component, controller for each plant has to give up cancellation at the corresponding frequencies.

Base on observations from plant identification, several design principles are listed as the following:

- Give up cancellation of noise around dc and nodal points;
- Enhance high-frequency robustness property;
- Improve convergence speed and noise rejection ability within the bandwidth;
- Keep control forces as small as possible.

To give up cancellation of noise around dc and nodal points, the corresponding internal models are added to the numerator of the signal generator M knowing that the controllers' internal states will not diverge if the error signal pass through the numerator first. In design stage, this means adding a series of filters to the plants. As a result, the frequency responses of B_i 's (Eq. (3.3)) are plotted in Fig. 6. Further, two frequency segments of the open loop gain ($|\mathbf{PC}|$ or $|1 - \sum_{i=1}^{m} B_i D_i|$) are selected to

satisfy:

- Unit magnitude and zero phase at frequency range 6.7 ~ 600 Hz;
- Magnitude as small as possible at frequency range 720 \sim 2000 Hz.

which represent disturbance rejection and robustness requirements, respectively.



Fig. 7 Open-loop gains for individual and overall control paths

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Fig. 8 Steady-state noise signal before and after cancellation

The optimization problem is then proposed as

$$\min \{ (\mathbf{e} - \mathbf{B}\mathbf{d})^T \mathbf{W}_1 (\mathbf{e} - \mathbf{B}\mathbf{d}) + \mathbf{d}^T \mathbf{B}^T \mathbf{W}_2 \mathbf{B}\mathbf{d} \}$$
(4.2)

Subject to

$$\max\left\{\sup_{i} \frac{\|y_{i}(\mathbf{d})\|_{1} \|G_{i}^{-1}\|_{A}}{c},\right.$$
$$\sup_{\omega} (\mathbf{e} - \mathbf{B}\mathbf{d})^{T} \operatorname{Re}\left\{\Theta(e^{-j\omega})\Theta^{T}(e^{j\omega})\right\}(\mathbf{e} - \mathbf{B}\mathbf{d})\right\} \leq 1$$

$$(4.3)$$

where $\mathbf{W}_1 = \Omega(0.3\pi) - \Omega(\pi/300)$, $\mathbf{W}_2 = \Omega(\pi) - \Omega(0.36\pi)$ and c is the maximum saturation level for the actuators. The length of D_1 and D_2 were chosen to be 37 (see Eqs. (3.4) and (3.7)) and the problem was solved by using the ellipsoid algorithm (Boyd and Barratt, 1991). Detailed iterative procedure is shown in the appendix but lengthy numerical results are not shown here (see Yu, 1995). The shaped loop gains are depicted in Fig. 7. From Fig. 7, the design objective was satisfied since each control path gave up control effort around corresponding nodal points and the overall loop gain was shaped to have good convergence property within the bandwidth.

Figure 8 shows the error versus noise signal measured at the feedback microphone while Fig. 9 shows their spectrum. Figure 9 indicates that the controller effectively cancel noise under the bandwidth limited by the filter q (Eq. (4.1)). Within 1 kHz, an average of 12.7 dB noise reduction was achieved. To see if each control path gave up its effort around nodal points, the spectrum of steady-state speaker input signals are plotted in

90 80 noise before cancellation 70 sound level (dB noise:after cancellatio 60 50 40 30 20 0 200 400 600 800 1000 Frequency (Hz)

Fig. 9 Spectrum of noise before and after cancellation



Fig. 10 Spectrum of the steady-state control signals

Fig. 10. Around 200 and 700 Hz, speaker 2 generated no control signals. Speaker 1 essentially gave up control at high frequency range including its nodal point (470 Hz).

5 Conclusion

In this paper, we present a new approach to design repetitive controllers. Comprehensive design goals involving robustness and performance are formulated as frequency-response or impulse-response constraints on some system functions. In order to solve this mixed frequency-domain and time-domain problem, the controller's structure is restricted to be finite dimensional. An interesting application of repetitive control is active harmonic noise cancellation (see Hu et al., 1995b). Because of the geometric constraint in wave propagation, one-loudspeaker ANC (Active Noise Control) system has some performance limits due to nodal points. Therefore, we designed a MISO repetitive ANC system and the proposed optimization techniques were applied. Experiments was conducted to demonstrate the effectiveness of this design approach.

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APPENDIX

Iterative Procedures to Solve Eqs. (4.2) and (4.3)

Step 1. Let k = 0 and make an initial guess of $\mathbf{d}_{(0)}$ and $A_{(0)}$ $\in \mathbf{R}^{n}, A_{(0)} \ge 0$, where n = 2(2l + 1).

Calculate the values of the constraint functions: Step 2.

$$\psi_2(\mathbf{d}_{(k)}) = \frac{1}{c} \max_i \|y_i\|_1 \|G_i^{-1}\|_A - 1 \quad (A.1)$$

$$\psi_1(\mathbf{d}_{(k)}) = \max_{\omega} (e - \mathbf{B}\mathbf{d}_{(k)})^T \mathbf{W}(\omega)$$

ŀ

$$(\mathbf{e} - \mathbf{Bd}_{(k)}) - 1$$
 (A.2)

$$\psi(\mathbf{d}_{(k)}) = \max \{ \psi_1(\mathbf{d}_{(k)}), \psi_2(\mathbf{d}_{(k)}) \}. \quad (A.3)$$

If $\psi_2(\mathbf{d}_{(k)})$ is the maximum of Eq. (A.3), the generalized gradient for $\psi(\mathbf{d}_{(k)})$ is

$$u(k) = \frac{\|G_i^{-1}\|_{\lambda}}{c} \sum_{s=1}^{2l+2r+1} \{ \operatorname{sign} (y_i(s-1)) \mathbf{v}_s + \operatorname{sign} (y_i(s+N-l-r-1)) \\ \times [-\mathbf{B}^T \mathbf{\bar{O}}^T \mathbf{w}_s + \mathbf{X}_s \mathbf{\bar{O}} (\mathbf{e} - \mathbf{B} \mathbf{d}_{(k)})] \}$$

where $||G_i^{-1}||_A$ and $y_i = \{y_i(t)\}_{t=0}^{N+l+r}$ achieve the maximum of Eq. (A.1), \mathbf{v}_s , \mathbf{w}_s , \mathbf{X}_s are defined in Eq. (3.26) and \mathbf{Q} is defined in Eq. (3.8).

Otherwise (i.e. $\psi_1(\mathbf{d}_{(k)})$ makes the maximum),

$$h_{(k)} = -2\mathbf{B}^T \mathbf{W}(\omega_0)(\mathbf{e} - \mathbf{B}\mathbf{d}_{(k)})$$

where ω_0 is the frequency achieving the maximum of Eq. (A.2).

Step 3. Check if $\psi(\mathbf{d}_{(k)})$ is less than zero. If yes, jump to Step 6, otherwise continue.

Step 4. Check if the following inequality is true.

$$\psi(d_{(k)}) > \sqrt{h_{(k)}^T A_{(k)} h_{(k)}}$$

If yes, stop iteration and re-initialize the guesses of $A_{(0)}$ and $\mathbf{d}_{(0)}$ or examine if the constraint functions are

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too stringent (it means that the ellipsoid established by $A_{(0)}$ and $\mathbf{d}_{(0)}$ does not satisfy the constraint).

Step 5. Calculate the following values:

$$\alpha = \psi(\mathbf{d}_{(k)})/\sqrt{h_{(k)}^{T}A_{(k)}h_{(k)}},$$

$$\tilde{h}_{(k)} = h_{(k)}/\sqrt{h_{(k)}^{T}A_{(k)}h_{(k)}},$$

$$\mathbf{d}_{(k+1)} = \mathbf{d}_{(k)} - \frac{1+n\alpha}{1+n}A_{(k)}\tilde{h}_{(k)},$$

$$A_{(k+1)} = \frac{n^{2}}{n^{2}-1}(1-\alpha^{2})(A_{(k)})$$

$$-\frac{2(1+n\alpha)}{(1+n)(1+\alpha)}A_{(k)}\tilde{h}_{(k)}\tilde{h}_{(k)}^{T}A_{(k)}),$$

and

$$k = k + 1$$
.

Go back to Step 2.

Step 6. Calculate the value of the cost function

$$\phi(\mathbf{d}_{(k)}) = \{ (\mathbf{e} - \mathbf{B}\mathbf{d})^T \mathbf{W}_1 (\mathbf{e} - \mathbf{B}\mathbf{d}) + \mathbf{d}^T \mathbf{B}^T \mathbf{W}_2 \mathbf{B}\mathbf{d} \},$$

and its generalized gradient

$$g_{(k)} = -2\mathbf{B}^T \mathbf{W}_1(\mathbf{e} - \mathbf{B}\mathbf{d}) + 2\mathbf{B}^T \mathbf{W}_2 \mathbf{B}\mathbf{d}.$$

Step 7. Calculate the following values:

$$\tilde{g} = g_{(k)} / \sqrt{g_{(k)}^T A_{(k)} g_{(k)}},$$

$$\mathbf{d}_{(k+1)} = \mathbf{d}_{(k)} - A_{(k)} \tilde{g} / (n+1),$$

and

$$A_{(k+1)} = \frac{n^2}{n^2 - 1} \left(A_{(k)} - \frac{2}{n+1} A_{(k)} \tilde{g} \tilde{g}^T A_{(k)} \right);$$

Step 8. Check if the following inequalities are satisfied,

$$\psi(\mathbf{d}_{(k)}) \leq \epsilon_1 \text{ and } \sqrt{g_{(k)}^T A_{(k)} g_{(k)}} \leq \epsilon_2$$

where ϵ_1 and ϵ_2 are pre-specified tolerance levels. If yes, stop iteration and \mathbf{d}_k is the answer. Otherwise let k = k + 1 and go back to *Step 2*.