

Decentralized Iterative Learning Control for Heterogeneous System with Arbitrary Interconnections

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Summary

- Subunits share their interconnection signals with directly interconnected neighbors.
- Iterative learning controllers improve their control signal in a decentralized manner.
- Then the controllers make the outputs of the subunits to track the desired outputs on a fixed time in an iteration domain.

Background

- A significant amount of research efforts has been focused on control of spatially interconnected systems, due to its emerging applications in diverse areas.
- Fast steering mirrors and deformable mirrors have recently been dealt as key applications of spatially interconnected systems.
- Control strategies such as H_∞ control and iterative learning control(ILC) which can be synthesized in the interconnected systems have advantages in terms of practicality, cost-effectiveness, high-performance.

Iterative Learning Control(ILC)

- Iterative learning control is a method to improve the performance of a system that operates repetitively on a finite time interval. For example, suppose that there is a ILC K which makes a given plant asymptotically stable. Then it can be expressed by

$$\lim_{i \rightarrow \infty} e_i(t) \rightarrow 0 \quad \forall t = [0, T]. \quad (1)$$

- Such improvement can be achieved by taking error signals from previous iterations as its feedback signal.

$$u_{i+1}(t) = u_i(t) + Qe_i(t) \quad (2)$$

- A huge amount of researches has been concentrated on the ILC of the spatially interconnected systems with various applications.

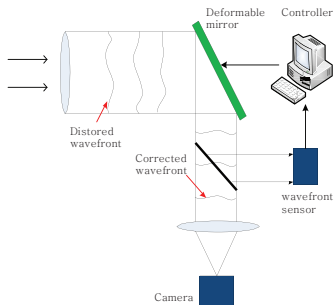


Figure: Adaptive optics system¹

- Adaptive optics system is a control system to compensate wavefront errors of images.
- Such compensation can be achieved by using lots of actuators to deform the mirror.

¹ Hyo-Sung Ahn, Taekyung Lee, and Young-Soo Kim, "Iterative learning control for spatially interconnected systems," 2011.

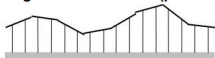
The selection of the deformable mirror type

Segmented facesheet (piston):



Liquid Crystals, DLP's
very high actuator density

Segmented facesheet (piston/tip/tilt):



very large (M1's) or MEMS devices
high actuator density

Continuous facesheet:



various types (electrostatic, reluctance,
piezo, ...)
medium sized diameters (cm range)
'no' photon loss, low spatial aliasing

Figure: Three types of deformable mirrors

In this research, a deformable mirror is considered as the continuous facesheet type.

Using 2D plate theory, the deformable mirror can be expressed as

$$\sigma d \frac{\partial w^2}{\partial^2 t} + D \left(\frac{\partial^4 w}{\partial x^4} + 2 \frac{\partial^4 w}{\partial^2 x \partial^2 y} + \frac{\partial^4 w}{\partial^4 y} \right) = u(t, x, y) \quad (3)$$

where w denotes the deformation of the plate at (t, x, y) , σ denotes the mass density, d denotes the thickness of the plate, $D = \left(\frac{Ed^3}{12(1-\nu^2)} \right)$ denotes the bending/flexural rigidity (E is the elastic modulus, and ν denotes the poisson's ratio), and u denotes the distributed prepressure generated from actuators.

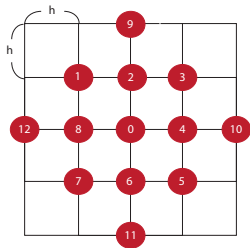


Figure: An example of interconnection structure

Modified equation of motion

In the previous figure, 2D Taylor expansions of each node with respect to the node 0 yeild

$$\begin{aligned}
& \sum_{k=0}^{12} a_k w(x_0 + \bar{x}_k, y_0 + \bar{y}_k) \\
&= w(x_0, y_0) \sum_{k=0}^{12} a_k + \left(\frac{\partial w(x_0, y_0)}{\partial y} \sum_{k=0}^{12} a_k \bar{y}_k + \frac{\partial w(x_0, y_0)}{\partial x} \sum_{k=0}^{12} a_k \bar{x}_k \right) \\
&+ \frac{1}{2} \left(\frac{\partial^2 w(x_0, y_0)}{\partial y^2} \sum_{k=0}^{12} a_k \bar{y}_k^2 + 2 \frac{\partial^2 w(x_0, y_0)}{\partial x \partial y} \sum_{k=0}^{12} a_k \bar{x}_k \bar{y}_k + \frac{\partial^2 w(x_0, y_0)}{\partial x^2} \sum_{k=0}^{12} a_k \bar{x}_k^2 \right) \\
&+ \frac{1}{6} \left(\frac{\partial^3 w(x_0, y_0)}{\partial y^3} \sum_{k=0}^{12} a_k \bar{y}_k^3 + 3 \frac{\partial^3 w(x_0, y_0)}{\partial x \partial y^2} \sum_{k=0}^{12} a_k \bar{x}_k \bar{y}_k^2 + 3 \frac{\partial^3 w(x_0, y_0)}{\partial x^2 \partial y} \sum_{k=0}^{12} a_k \bar{x}_k^2 \bar{y}_k + \frac{\partial^3 w(x_0, y_0)}{\partial x^3} \sum_{k=0}^{12} a_k \bar{x}_k^3 \right) \\
&+ \frac{1}{24} \left(\frac{\partial^4 w(x_0, y_0)}{\partial y^4} \sum_{k=0}^{12} a_k \bar{y}_k^4 + 4 \frac{\partial^4 w(x_0, y_0)}{\partial x \partial y^3} \sum_{k=0}^{12} a_k \bar{x}_k \bar{y}_k^3 + 6 \frac{\partial^4 w(x_0, y_0)}{\partial x^2 \partial y^2} \sum_{k=0}^{12} a_k \bar{x}_k^2 \bar{y}_k^2 \right. \\
&\quad \left. + 4 \frac{\partial^4 w(x_0, y_0)}{\partial x^3 \partial y} \sum_{k=0}^{12} a_k \bar{x}_k^3 \bar{y}_k + \frac{\partial^4 w(x_0, y_0)}{\partial x^4} \sum_{k=0}^{12} a_k \bar{x}_k^4 \right) + O(h^5).
\end{aligned} \tag{4}$$

Modified equation of motion

From the equation (4), we can see that the partial derivative term can be linearized as

$$\sum_{k=0}^{12} a_k w(x_0 + \bar{x}_k, y_0 + \bar{y}_k) \approx \left(\frac{\partial^4 w}{\partial x^4} + 2 \frac{\partial^4 w}{\partial^2 x \partial^2 y} + \frac{\partial^4 w}{\partial^4 y} \right). \quad (5)$$

Then we can obtain the spatially discretized equation of motion as

$$\sigma d \frac{dw(t, x_i, y_j)^2}{d^2 t} + D \underbrace{\left(\sum_{k=0}^{12} a_k w(t, x_i + \bar{x}_k, y_j + \bar{y}_k) \right)}_{\text{interconnection term}} = u(t, x_i, y_j). \quad (6)$$

Modified equation of motion

The parameters can be identified using the Vandermonde matrix.

$$\begin{bmatrix} 1 & \dots & 1 \\ \bar{x}_0 & \dots & \bar{x}_{12} \\ \bar{y}_0 & \dots & \bar{y}_{12} \\ \bar{x}_0^2 & \dots & \bar{x}_{12}^2 \\ \bar{y}_0^2 & \dots & \bar{y}_{12}^2 \\ \bar{x}_0\bar{y}_0 & \dots & \bar{x}_{12}\bar{y}_{12} \\ \bar{x}_0^3 & \dots & \bar{x}_{12}^3 \\ \bar{y}_0^3 & \dots & \bar{y}_{12}^3 \\ \bar{x}_0^2\bar{y}_0 & \dots & \bar{x}_{12}^2\bar{y}_{12} \\ \bar{x}_0\bar{y}_0^2 & \dots & \bar{x}_{12}\bar{y}_{12}^2 \\ \bar{x}_0^4 & \dots & \bar{x}_{12}^4 \\ \bar{y}_0^4 & \dots & \bar{y}_{12}^4 \\ \bar{x}_0^2\bar{y}_0^2 & \dots & \bar{x}_{12}^2\bar{y}_{12}^2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \\ a_7 \\ a_8 \\ a_9 \\ a_{10} \\ a_{11} \\ a_{12} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 24 \\ 24 \\ 8 \end{bmatrix} \rightarrow \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \\ a_7 \\ a_8 \\ a_9 \\ a_{10} \\ a_{11} \\ a_{12} \end{bmatrix} = \frac{1}{h^4} \begin{bmatrix} 20 \\ 2 \\ -8 \\ 2 \\ -8 \\ 2 \\ -8 \\ 2 \\ -8 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad (7)$$

Modified equation of motion(State-space form)

By replacing the spatial domain (x, y) with (s_1, s_2) and introducing the spatial operator S_1 and S_2 which have the features

$$\begin{aligned} S_1 w(t, s_1, s_2) &= w(t, s_1 + 1, s_2), S_1^{-1} w(t, s_1, s_2) = w(t, s_1 - 1, s_2) \\ S_2 w(t, s_1, s_2) &= w(t, s_1, s_2 + 1), S_2^{-1} w(t, s_1, s_2) = w(t, s_1, s_2 - 1), \end{aligned} \quad (8)$$

the equation described by (6) can be transformed as

$$\begin{aligned} \sigma d\ddot{w}(t, s_1, s_2) + \frac{D}{h^4} \left(20 + 2 \left(S_1^{-1} S_2 + S_1 S_2 + S_1 S_2^{-1} + S_1^{-1} S_2^{-1} \right) \right. \\ \left. - 8 \left(S_2 + S_1 + S_2^{-1} + S_1^{-1} \right) + \left(S_2 S_2 + S_1 S_1 + S_2^{-1} S_2^{-1} + S_1^{-1} S_1^{-1} \right) \right) w = u. \end{aligned} \quad (9)$$

Modified equation of motion(State-space form)

By defining the state variables

$$\mathbf{x} = \begin{bmatrix} \mathbf{x}_1(t, s_1, s_2) \\ \mathbf{x}_2(t, s_1, s_2) \end{bmatrix} = \begin{bmatrix} \mathbf{w}(t, s_1, s_2) \\ \dot{\mathbf{w}}(t, s_1, s_2) \end{bmatrix}, \quad (10)$$

the equation of motion can be represented as the spatially interconnected form² as

$$\dot{\mathbf{x}} = \mathbf{A}_{TT}\mathbf{x}(t, s_1, s_2) + \mathbf{A}_{TS}\mathbf{v}(t, s_1, s_2) + \mathbf{B}_T\mathbf{u}(t, s_1, s_2) \quad (11)$$

$$\mathbf{v} = \mathbf{A}_{ST}\mathbf{x}(t, s_1, s_2) + \mathbf{A}_{SS}\mathbf{v}(t, s_1, s_2) + \mathbf{B}_S\mathbf{u}(t, s_1, s_2) \quad (12)$$

$$\mathbf{y} = \mathbf{C}_T\mathbf{x}(t, s_1, s_2) + \mathbf{C}_S\mathbf{v}(t, s_1, s_2). \quad (13)$$

²R. D'Andrea and G. E. Dullerud, "Distributed control of spatially interconnected systems," *IEEE Transactions on Automatic Control*, vol. 48, no. 9, pp. 1478–1495, Sep.2003.

where

$$A_{TT} = \begin{bmatrix} 0 & 1 \\ 20\alpha & 0 \end{bmatrix} \text{ where } \alpha = -\frac{D}{\sigma d h^4} \quad (14)$$

$$A_{TS} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 2 & -8 & 2 & -8 & 2 & -8 & 2 & -8 & 1 & 1 & 1 & 1 \end{bmatrix} \quad (15)$$

$$A_{ST} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}^T \quad (16)$$

$$A_{SS} = [0] \quad (17)$$

$$B_T = \begin{bmatrix} 0 \\ \frac{1}{\sigma d} \end{bmatrix} \quad (18)$$

$$B_S = [0] \quad (19)$$

$$C_T = 1 \quad (20)$$

$$C_S = 0. \quad (21)$$

Problem formulation

- Subunit model:

$$x_i^j(k+1) = A_{TT}^j x_i^j(k) + A_{TS}^j v_i^j(k) + B_{Tu}^j u_i^j(k) \quad (22)$$

$$w_i^j(k) = A_{ST}^j x_i^j(k) + A_{SS}^j v_i^j(k) + B_{Su}^j u_i^j(k) \quad (23)$$

$$y_i^j(k) = C_T^j x_i^j(k) + C_S^j v_i^j(k) + D_u^j u_i^j(k) \quad (24)$$

$$x_i^j(0) = x_0^j, \text{ for } k = 0, 1, \dots, K. \quad (25)$$

where $j \in \mathcal{V}$, i is the iteration number, $x_i^j(k) \in \mathbb{R}^{m_j}$ is the state, $v_i^j(k) \in \mathbb{R}^{n_j}$ is the interconnected input, $u_i^j(k) \in \mathbb{R}^{r_j}$ is the control input, $w_i^j(k) \in \mathbb{R}^{n_j}$ is the interconnected output, and $y_i^j(k) \in \mathbb{R}^{r_j}$ is the output at the subunit j .

- Interconnection topology: $\mathcal{G} = (\mathcal{V}, \mathcal{E})$.

Modified state-space model

The variables in (22) – (25) can be rewritten as the super-vector containing information at the whole time instants

$$\mathbf{X}_i^j = \begin{bmatrix} x_i^j(1) & x_i^j(2) & \cdots & x_i^j(K) \end{bmatrix}^T \quad (26)$$

$$\mathcal{V}_i^j = \begin{bmatrix} v_i^j(0) & v_i^j(1) & \cdots & v_i^j(K) \end{bmatrix}^T \quad (27)$$

$$\mathbf{W}_i^j = \begin{bmatrix} w_i^j(0) & w_i^j(1) & \cdots & w_i^j(K) \end{bmatrix}^T \quad (28)$$

$$\mathbf{U}_i^j = \begin{bmatrix} u_i^j(0) & u_i^j(1) & \cdots & u_i^j(K-1) \end{bmatrix}^T \quad (29)$$

$$\mathbf{Y}_i^j = \begin{bmatrix} y_i^j(0) & y_i^j(1) & \cdots & y_i^j(K-1) \end{bmatrix}^T \quad (30)$$

Modified state-space model

In addition, introduce so-called stacking vectors containing information at both temporal and spatial position

$$\tilde{\mathbf{X}}_i = [\mathbf{X}_i^1 \quad \mathbf{X}_i^2 \quad \cdots \quad \mathbf{X}_i^L]^T \quad (31)$$

$$\tilde{\mathbf{V}}_i = [\mathcal{V}_i^1 \quad \mathcal{V}_i^2 \quad \cdots \quad \mathcal{V}_i^L]^T \quad (32)$$

$$\tilde{\mathbf{W}}_i = [\mathbf{W}_i^1 \quad \mathbf{W}_i^2 \quad \cdots \quad \mathbf{W}_i^L]^T \quad (33)$$

$$\tilde{\mathbf{U}}_i = [\mathbf{U}_i^1 \quad \mathbf{U}_i^2 \quad \cdots \quad \mathbf{U}_i^L]^T \quad (34)$$

$$\tilde{\mathbf{Y}}_i = [\mathbf{Y}_i^1 \quad \mathbf{Y}_i^2 \quad \cdots \quad \mathbf{Y}_i^L]^T \quad (35)$$

Modified state-space model

Then the overall relation among the subsystems can be expressed by

$$\tilde{\mathbf{X}}_i = \tilde{\mathbf{A}}_{TT}\tilde{\mathbf{X}}_0 + \tilde{\mathbf{A}}_{TS}\tilde{\mathbf{V}}_i + \tilde{\mathbf{B}}_{Tu}\tilde{\mathbf{U}}_i \quad (36)$$

$$\tilde{\mathbf{W}}_i = \tilde{\mathbf{A}}_{ST}\tilde{\mathbf{X}}_0 + \tilde{\mathbf{A}}_{SS}\tilde{\mathbf{V}}_i + \tilde{\mathbf{B}}_{Su}\tilde{\mathbf{U}}_i \quad (37)$$

$$\tilde{\mathbf{Y}}_i = \tilde{\mathbf{C}}_T\tilde{\mathbf{X}}_0 + \tilde{\mathbf{C}}_S\tilde{\mathbf{V}}_i + \tilde{\mathbf{D}}_u\tilde{\mathbf{U}}_i \quad (38)$$

Modified state-space model

where

$$\tilde{\mathbf{A}}_{TT} = \text{diag}[\hat{\mathbf{A}}_{TT}^l] \quad (39)$$

$$\tilde{\mathbf{A}}_{TS} = \text{diag}[\hat{\mathbf{A}}_{TS}^l] \quad (40)$$

$$\tilde{\mathbf{B}}_{Tu} = \text{diag}[\hat{\mathbf{B}}_{Tu}^l] \quad (41)$$

$$\tilde{\mathbf{A}}_{ST} = \text{diag}[\hat{\mathbf{A}}_{TS}^l] \quad (42)$$

$$\tilde{\mathbf{A}}_{SS} = \text{diag}[\hat{\mathbf{A}}_{SS}^l] \quad (43)$$

$$\tilde{\mathbf{B}}_{Su} = \text{diag}[\hat{\mathbf{B}}_{Su}^l] \quad (44)$$

$$\tilde{\mathbf{C}}_T = \text{diag}[\hat{\mathbf{C}}_T^l] \quad (45)$$

$$\tilde{\mathbf{C}}_S = \text{diag}[\hat{\mathbf{C}}_S^l] \quad (46)$$

$$\tilde{\mathbf{D}}_u = \text{diag}[\hat{\mathbf{D}}_{TT}^l] \quad (47)$$

for $l = 1, 2, \dots, L$.

Control strategy

Theorem

Consider the equations represented by (36), (37), and (38), which describe the whole system comprising subsystems with identical initial conditions. For a finite time interval $k = 0, 1, 2, \dots, K$, the desired trajectory is given by $y_d^j(k)$ which can be transformed into $\tilde{\mathbf{Y}}_d$ in a stacking vector form with the decentralized ILC law described by

$$\tilde{\mathbf{U}}_{i+1} = \tilde{\mathbf{U}}_i + \mathbf{T}_e(\tilde{\mathbf{Y}}_d - \tilde{\mathbf{Y}}_i) \quad (48)$$

where

$$\mathbf{T}_e = \begin{bmatrix} z_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & z_q \end{bmatrix}. \quad (49)$$

Theorem

(continue)

For such a system, the error $\mathcal{E}^ = \tilde{\mathbf{Y}}_d - \tilde{\mathbf{Y}}_i$ converges to zero asymptotically if there exists a diagonal matrix \mathbf{T}_e such that the inequality is satisfied as*

$$\rho(\mathbf{I} - \mathbf{T}_e \mathbf{T}_S) < 1. \quad (50)$$

where $\rho(\cdot)$ denotes the spectral radius.

Simulation

Suppose that there are three subunits($j = 1, 2, 3$) for a finite time instants described by $k = [1, 2, \dots, 50](sec)$. The state-space models of the three subunits are numerically given as

$$\left[\begin{array}{c|c|c} A_{TT}^j & A_{TS}^j & B_{Tu}^j \\ \hline A_{ST}^j & A_{SS}^j & 0 \\ \hline C_T^j & 0 & 0 \end{array} \right] \quad (51)$$

$$x_i(0) = 0 \quad (52)$$

where

$$A_{TT}^j = -j \quad (53)$$

$$A_{TS}^j = \begin{bmatrix} j & j & j \end{bmatrix} \quad (54)$$

$$B_{Tu}^j = 1 \quad (55)$$

$$A_{ST}^j = \begin{bmatrix} j & j & j \end{bmatrix}^T \quad (56)$$

$$A_{SS}^j = \text{diag}\{0.1 \times j\} \quad (57)$$

$$C_T^j = 0.5 \text{ for } j=1,2,3. \quad (58)$$

Simulation

The desired trajectory vector is given for the time instant $k = 1, 2, 3, \dots, 50(sec)$,

$$\begin{bmatrix} y_d^1(k) \\ y_d^2(k) \\ y_d^3(k) \end{bmatrix} = \begin{bmatrix} \cos(0.1k) \\ k(k+1) \\ \tan(k)^2 \end{bmatrix} \quad (59)$$

Simulation

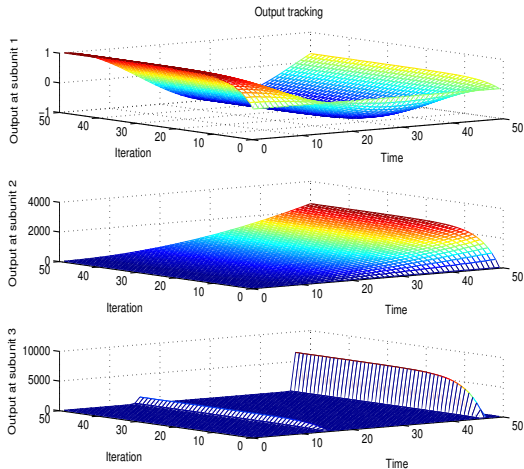


Figure: Output tracking performance

Simulation

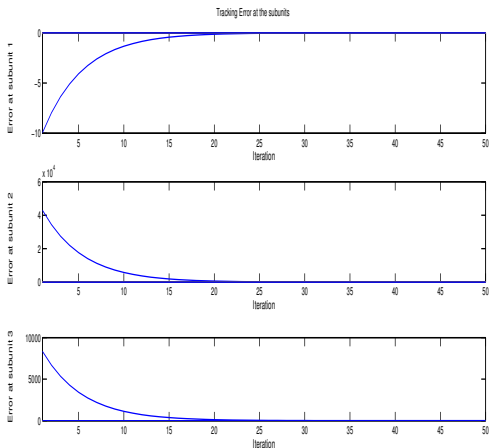


Figure: Error convergence

Conclusion

- We proposed a decentralized iterative learning controller for a spatially heterogeneous system with arbitrary interconnections.
- As shown from the numerical simulation results, the tracking error converges to zero as the iterative number increases.
- As a future work, it is desirable to get more realistic model to describe the whole adaptive optics system.

Thank you for your attention.
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