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Contributions to the control configuration selection

With the ever increasing complexity of the process plants and manufacturing processes, the objectives of process control strategies cannot be attained unless a suitable control configuration is selected. To select an appropriate control configuration, it is important to determine which variables should be measured and how the process should be actuated. Therefore, the first step is to determine the optimal locations for the sensors and actuators. This makes providing the accurate and reliable process measurements and suitable actuations possible for the control purposes. For the multivariable processes, this step is followed by choosing the appropriate input and output pairs for the design of SISO (or block) controllers. This is due to the popularity of the distributed and decentralized control in industrial control systems. The reason for this popularity is that the centralized control of large-scale complex systems are expensive and difficult, due to the computational complexity, the problems related to reliability and the limitations in communications. On the other hand, decentralized controllers are easy to understand for operators, easy to implement and to re-tune [1]-[2].

In this paper both key issues in control configuration selection are addressed. These two key issues have been studied extensively for deterministic systems. For the placement of the sensors and the actuators, several techniques have been proposed over the last few decades. These techniques take into account different performance criteria [3]-[9]. One of the most reliable criterions for determining sensor and actuator locations is the improvement of state controllability and observability of the process [3]. In these methods, the problem of determining the sensor locations is viewed as the problem of maximizing the output energy generated by a given state. The problem for the actuator locations is viewed as the problem of minimizing the input energy required to reach a given state. In [4]-[6], several gramian-based methods from this category for optimal placement of the sensors and the actuators have been proposed. These methods have been improved and have been extended to unstable systems in [9] and further to nonlinear systems in [8]-[7].

The second key issue of control configuration selection which is input-output pairing has also been studied extensively for multivariable deterministic systems. The results in this context are based on different interaction measures. Interaction measures make it possible to study input-output interactions and to partition a process into subsystems in order to reduce the coupling, to facilitate the control and to achieve a satisfactory performance. There are two broad categories of interaction measures in the literature. The first category is the relative gain array (RGA) and its related indices [10]-[16] and the second category is the family of the gramian-based interaction measures [17]-[26].

In recent years, gramians become popular in the process of control configuration selection. The gramians are matrices with the embedded controllability and observability information. The

controllability and observability gramians were first introduced in [27] and [28] and more recently in [29]. It is well-known that the controllability gramian shows the level of controllability. Similarly, the observability gramian contains information of the level of observability for a system. Gramians have been also been extended, improved and have been used in different applications such as model reduction [27]-[39].

References:

1. R. Scattolini, Architectures for distributed and hierarchical model predictive control - a review, *Journal of Process Control*, Vol. 19, No. 5, pp. 723-731, 2009.
2. A. Khaki-Sedigh, and B. Moaveni, Control Configuration Selection for Multivariable Plants, *Lecture Notes in Control and Information Sciences*, Springer, 2009.
3. M. Van de Wal, and B. De Jager, A review of methods for input/output selection, *Automatica*, Vol. 37, No. 4, pp.487-510, 2001.
4. D. Georges, The use of observability and controllability Gramians or functions for optimal sensor and actuator location infinite-dimensional systems, *Proceedings of IEEE Conference on Decision and Control*, Vol. 4 , pp. 3319-3324,1995.
5. B. Marx, D. Koenig and D. Georges, Optimal sensor/actuator location for descriptor systems using Lyapunov-Like equations, *Proceedings of IEEE Conference on Decision and Control*, pp. 4541-4542, 2002.
6. B. Marx, D. Koenig, D. Georges, Optimal sensor and actuator location for descriptor systems using generalized gramian and balanced realization, *Proceedings of American Control Conference*, pp. 2729-2734, 2004.
7. A.K. Singh and J. Hahn, Sensor Location for Stable Nonlinear Dynamic Systems: Multiple Sensor Case, *Industrial & Engineering Chemistry Research*, Vol. 45, No. 10, pp. 3615-3623, 2006.
8. A.K. Singh and J. Hahn, Determining Optimal Sensor Locations for State and Parameter Estimation for Stable Nonlinear Systems, *Industrial Engineering & Chemistry Research*, Vol. 44, No. 15, pp. 5645-5659 , 2005.
9. H. R. Shaker and M. Tahavori, Optimal sensor and actuator location for unstable systems, *Journal of Vibration and Control*, Vol. 19, No. 12, pp. 1915-1920, 2012.
10. E. H. Bristol, On a new measure of interaction for multivariable process control, *IEEE Transactions on Automatic Control*, Vol. 11, pp. 133-134, 1966.
11. S. Skogestad and M. Morari, Implications of large RGA elements on control performance, *Industrial Engineering Chemistry Research*, Vol. 26, pp. 2323-2330, 1987.
12. M. F. Witcher, and T. J. McAvoy, Interacting control systems: Steady-state and dynamic measurement of interaction, *ISA Transactions*, Vol. 16, No. 3, pp. 35-41, 1977.
13. A. Niederlinski, A heuristic approach to the design of linear multivariable interacting control systems, *Automatica*, Vol. 7, pp. 691-701, 1971.
14. E. H. Bristol, Recent results on interaction in multivariable process control. 71st AIChE Conference, 1978.
15. E. Gagnon, A. Desbiens, and A. Pomerleau, Selection of pairing and constrained robust decentralized PI controllers, *Proceedings of American Control Conference*, pp. 4343-4347, 1999.
16. Mao-Jun He, Wen-Jian Cai, Wei Ni, Li-Hua Xie, RNGA based control system configuration for multivariable processes, *Journal of Process Control*, Vol. 19, pp. 1036-1042, 2009.
17. A. Conley and M. E. Salgado, Gramian based interaction measure, *The Proceedings of 39th IEEE Conference on Decision and Control*, pp. 5020-5022, 2000.
18. M. E. Salgado and A. Conley, MIMO interaction measure and controller structure selection, *International Journal of Control*, Vol. 77, pp. 367-383, 2004.
19. H. R. Shaker, F. Shaker, Control Configuration Selection for Linear Stochastic Systems, *Journal of Process Control*, Vol. 24, pp. 146-151, 2014.
20. B. Halvarsson, Comparison of some gramian based interaction measures, *Proceedings of IEEE Multi-conference on Systems and Control*, pp.138-143, 2008.
21. M. E. Salgado and A. Conley, MIMO interaction measure and controller structure selection, *International Journal of Control*, Vol. 77, pp. 367-383, 2004.
22. H. R. Shaker and J. Stoustrup, Control configuration selection for multivariable descriptor systems, *Proceedings of the 2012 American Control Conference*, pp. 6294-9299, 2012.
23. H. R. Shaker and J. Stoustrup, An interaction measure for control configuration selection for multivariable bilinear systems, *Nonlinear Dynamics*, Vol. 72, No.1-2, pp 165-174, 2013.
24. H. R. Shaker and M. Tahavori, Frequency-interval control configuration selection for multivariable bilinear systems, *Journal of Process Control*, Vol. 23, No. 6, pp 894-904, 2013.
25. H. R. Shaker and M. Komareji, Control Configuration Selection for Multivariable Nonlinear Systems, *Industrial & Engineering Chemistry Research*, Vol. 51, No. 25, pp 8583-8587 , 2012.
26. B. Moaveni, A. Khaki Sedigh, Input-output pairing analysis for uncertain multivariable processes, *Journal of Process Control*, Vol. 18, No. 6, pp. 527-532, 2008.
27. E. G. Lee and L. Marcus, *Foundations of Optimal Control Theory*, New York, Wiley, 1967.

28. R W. Brockett, Finite Dimensional Linear System, New York. Wiley, (1970).
29. B. C. Moore, Principal component analysis in linear systems: controllability, observability, and model reduction, IEEE Transactions on Automatic Control, pp. 17-32, AC-26, 1981.
30. W. Gawronski, and J.-N. Juang , Model Reduction in Limited Time and Frequency Intervals , International Journal of System Sciences , Vol. 21, No. 2, pp. 349-376, 1990.
31. M. Tahavori and H. R. Shaker, Model reduction via time-interval balanced stochastic truncation for linear time invariant systems. , International Journal of Systems Science, Vol. 44, No. 3, pp. 493-501, 2013.
32. H. R. Shaker, and R. Wisniewski , Model reduction of switched systems based on switching generalized gramians, International Journal of Innovative Computing, Information and Control, Vol. 8, No. 7(B), 2012.
33. H. R. Shaker, and R. Wisniewski, Switched systems reduction framework based on convex combination of generalized gramians, Journal of Control Science and Engineering, 2009.
34. H. R. Shaker and R. Wisniewski, Generalized gramian framework for model/controller order reduction of switched systems, International Journal of Systems Science, Vol.42, No.8, pp.1277-1291, 2011.
35. H. R. Shaker, and M. Tahavori, Time-Interval Model Reduction of Bilinear Systems, International Journal of Control, 2013.
36. M. Tahavori and H. R. Shaker, Relative error model reduction via time-weighted balanced stochastic singular perturbation , Journal of Systems Science , Journal of Vibration and Control, Vol. 18, No.13, pp. 2006-2016, 2012.
37. M. Tahavori and H. R. Shaker, Time-weighted balanced stochastic model reduction, 50th IEEE Conference on Decision and Control and European Control Conference (CDC-ECC), pp. 7777-7781, 2011.
38. H. R. Shaker, and M. Tahavori, Frequency-Interval Model Reduction of Bilinear Systems, IEEE Transactions on Automatic Control, will appear in July issue, 2014.
39. H. R. Shaker , M. Tahavori, Control Reconfigurability of Bilinear Hydraulic Drive Systems, Proceedings of IEEE Conference on Fluid Power and Mechatronics, Beijing, China, 2011.