CRII: CPS: Architecture and Distributed Computation in the Networked Control Paradigm: An Autonomous Grid Example PENNSTATE.

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(1)

Abstract

This project is focused on the fundamental research in establishing a foundational framework towards the development of an autonomous Cyber-Physical System (CPS) through distributed computation in a Networked Control Systems (NCS) paradigm. Specific attention is focused on an application where the computational, and communication challenges are unique due to the sheer dimensionality of the physical system. An example of such CPS is the smart power grid, which includes large-scale deployment of distributed and networked Phasor Measurement Units (PMUs) and wind energy resources. A systematic approach is proposed for wide-area oscillation damping control, which can handle data packet dropout in the communication channels in such smart power grids. The major challenges identified in the controller design are: a) computational burden, b) communication network delays and data drops. To handle the challenges of computational burden, frequency domain abstraction is explored to reduce the dimension of the model. To that end it is demonstrated that a widely-used technique Balanced Truncation (BT) fails to produce stable reduced order model (ROM) when modern grid with wind generation is considered while a relatively new approach - Iterative Rational Krylov Algorithm (IRKA) is able to give an ROM retaining the essential properties of the system. To handle the delays and data drops the dynamic property of the ROM is exploited in an Observer-driven Reduced Copy (ORC) approach. The key contribution comes from the analytical derivation of the impact of coupling between the cyber and the physical layer on ORC performance. Nonlinear time-domain simulations following large disturbances (e.g., faults, line outages, etc.) demonstrate that the ORC gives significantly better performance compared to conventional feedback under higher data drop situations.





Analytical Derivation of ORC Performance Affected by Cyber-physical Coupling

□ The error between the reduced order linearized system state trajectory and that of reduced copy

 $\xi(t) = e^{A_i(t-t_k)}(x_{ik}^0 - \bar{x}_k) + e^{A_i(t-t_k)}(I - \Psi_k)(\bar{x}_k - x_{nk}) + \int e^{A_i(t-\tau)}(\tilde{A} - \tilde{B}K)e^{(A_n - B_nK)\tau}(\Psi_k(\bar{x}_k - x_{nk}) + x_{nk})d\tau$

- 1. The open-loop system and the closed-loop nominal system are stable.
- 2. The observer tracks the actual states perfectly: $(x_{ik}^0 \bar{x}_k) = 0$.
- □ The expectation of the maximum norm of error can be derived as:

 $\mathbf{T}[\|c(i)\| = \left\| - V(1-i) + V \right\| \tilde{i} = \tilde{D}V \| + V \| \tilde{i} = \tilde{D}V \|$

□ Balanced truncation (BT): Based on Singular value decomposition and uses controllability and	$\mathbf{E}[\ \xi(t)\ _{max}] \propto K_1(1-\rho) + K_2 \ A - BK\ + K_3 \ A - BK\ \rho $ (2)
observability gramians for ordering Hankel singular values and form ROM.	Important Notations:
□ Iterative Rational Krylov Algorithm (IRKA): Based on Moment matching approach which uses sigma	1 . \tilde{A} and \tilde{B} : Deviation in A and B matrix from nominal operating condition; 2 . ρ : Data receiving rate in the
points to iteratively correct the ROM.	communication channel; 3. K: Controller gain; 4 . K_1 , K_2 , and K_3 : Proportionality constants.
□ These techniques were tested in two scenarios: Conventional power grid with SGs only (PS-SG) and	Observations from Equation (2):
modern power grid with inverter-interfaced WFs (PS-DFIG).	1 st –Term: Impact of cyber-only term indicates the error norm increases with increase in data dropout.
Communication Dropout-resilient Control: Principle of Proposed ORC	2 nd –Term: Impact of physical-only term indicates increase in error norm under off-nominal condition. 3 rd –Term: The cyber-physical coupling term indicates a non-trivial impact on the error norm.
An observer at each sensor location uses the reduced-order linearized model of the power system	Conclusion [2-5]
to estimate the states. \Box	Both BT and IRKA can produce acceptable ROM for conventional power grids but for modern
For each feedback signal the corresponding control input $(u(t))$ at the actuator and sensor locations are calculated using two reduced order models of the power system, see Fig. 1.	power grid BT results in an unstable ROM whereas IRKA produces stable ROM capturing poorly- damped modes.
When data packet drops out the states of each copy are allowed to evolve naturally otherwise	□ The uncertainty in the cyber layer due to data packet drop and the off-nominal operation of the
switched to reset its states.	physical layer affect the ORC performance in a coupled manner, where the coupling is non-trivial.

Simulation Results



0.035

0.03

0.025

0.02

Table: Comparison of relative H_2 error norm in power system with two model reduction approaches (r = 32)

Type of	Model reduction approach	
PS	BT	IRKA
PS-SG	1.7737×10^{-4}	4.1345×10^{-5}
PS-DFIG	∞	5.9×10^{-3}

R=25%

40-41



-80 -60 -40 -20 0 -0.5 -0.4 -0.3 σ -0.2Fig.3. a): Eigenvalue plot for full system (n = 133) and reduced-order systems (r = 32) in conventional power grid with SGs. b): Zoomed inter-area modes.

-8 + 4

R=75%

60-61 Outage

 \square

40-41

0.035

0.03

0.025

0.02

0.015

18-42

 $\|\xi(t)\|_{j}$

 $-50 \sigma 0$ 50 -0.4 -0.2 0 - 0.2 0.4 0.6Fig. 4. a): Eigenvalue plot for full system (n= 148) and reduced-order system (r=32) in modern power grid with inverter-interfaced WF. b): Zoomed inter-area modes.

18-42

60-61

Outage

0.035

0.03

0.025

0.02

Communication dropout-resilient control: ORC Approach

100 Monte Carlo Runs Performed for each Case

R=50%

40-41



Fig. 5. Dynamic response following a pulse disturbance at the input of the conventional and modern power grid when represented with their ROMs (r= 32) with different model reduction techniques.





Fig.6. Dynamic performance after a three-phase fault near bus 18/60/40 following by the outage of one of the tie-lines between buses 18-42/60-61/40-41, respectively, see Fig. 2.

Fig. 7. Boxplots of maximum error bound of state trajectories in the inter-sample interval with non-ideal communication channel calculated from linear timedomain simulation. 100 Monte Carlo runs were conducted for each outage and data rates, totals 900 runs.

60-61 Outage

Fig.8. Bound on error norm of state trajectories in the inter-sample interval with ideal communication channel: (a) measure of bound on error norm from equation (2), (b) calculated bound on error from linear simulation

Publications

18-42

A. Yogarathinam, J. Kaur, and N.R. Chaudhuri, "Modeling Adequacy for Studying Power Oscillation Damping in Grids with Wind Farms and Networked Control Systems (NCS)", in proceedings of IEEE Power and Energy Society General Meeting (PESGM), 2016. 2. J. Kaur, and N.R. Chaudhuri, "Challenges of Model Reduction in Modern Power Grid with Wind Generation", in proceedings of IEEE of 48th North American Power Symposium (NAPS), 2016.

. Yogarathinam, and N.R. Chaudhuri, "Data Packet-drop Resilient Wide-Area Damping Control Using DFIG-based Wind Farms", in proceedings of IEEE of 48th North American Power Symposium (NAPS), 2016. 3. A.

4. A. Yogarathinam, and N.R. Chaudhuri, "An Approach for Wide-Area Damping Control using Multiple DFIG-based Wind Farms to Deal with Communication Dropouts", in proceedings of IEEE of 7th Innovative Smart Grid Technologies (ISGT) Conference, 2016. 5. A. Yogarathinam, and N.R. Chaudhuri, "An Approach for Wide-Area Damping Control using Multiple DFIG-based Wind Farms Under Stochastic Data Packet Dropouts." Under review in IEEE Transactions on Smart Grid.

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