

MOTIVATION

- The increased penetration of new technologies has heightened the variability of the grid, impacting the efficiency of damping controller (DC) devices. New adaptable approaches are necessary.
- Current adaptable WADCs face some challenges, such as dependency on up-to-date models for model-based approaches and security constraints for real-time fine-tuning in the case of data-driven approaches.
- How can inverter-based resources (IBR) be optimally used as DC only when necessary?

COORDINATION OF DAMPING CONTROLLERS

The coordination seeks the on/off switching combination of all DCs that best improves grid damping for the current operating condition and disturbance. This is achieved by identifying the coordination that minimizes the total action, a metric based on the kinetic energy released by synchronous generators.

Optimal controller coordination

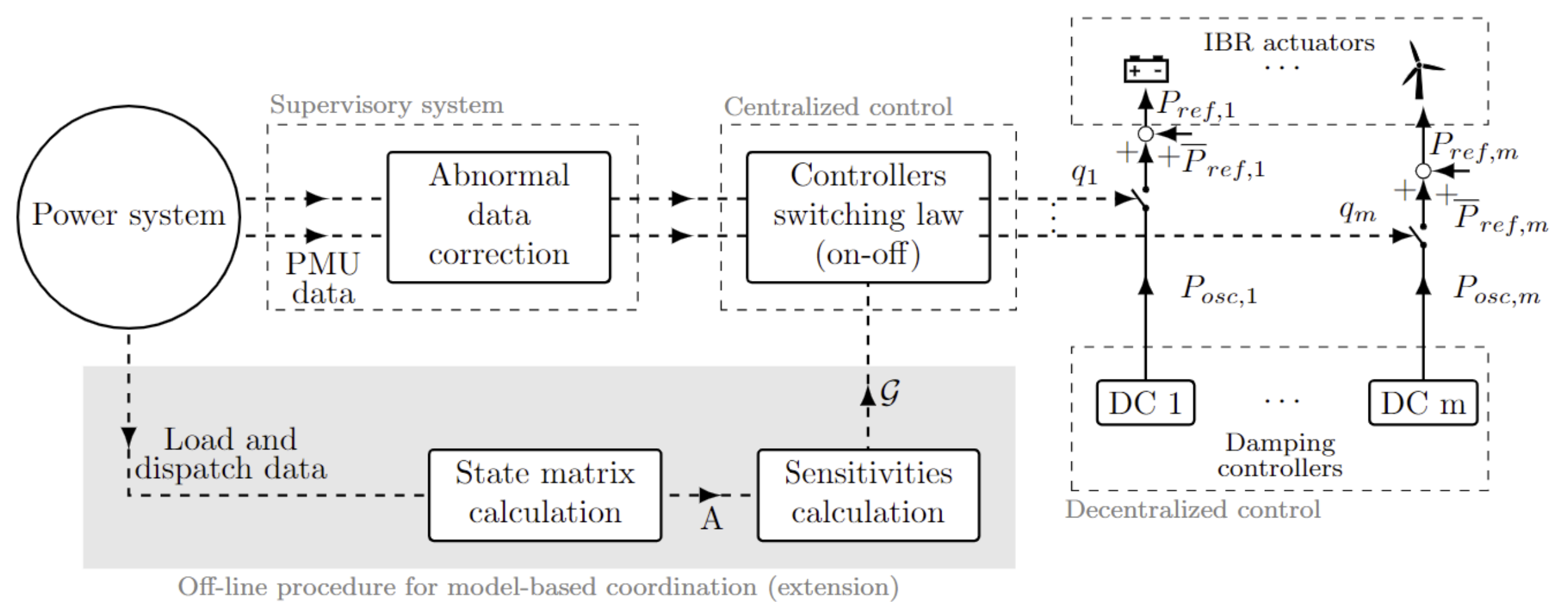
$$\gamma^* = \arg \min_{\gamma_k \forall k \in \{1, \dots, n_c\}} \hat{S}_\infty(y_0, \gamma_k)$$

Total Action

$$S_\infty = \lim_{\tau \rightarrow \infty} S(\tau) = \lim_{\tau \rightarrow \infty} \int_{t_0}^{t_0 + \tau} E(t) dt$$

Oscillation Energy

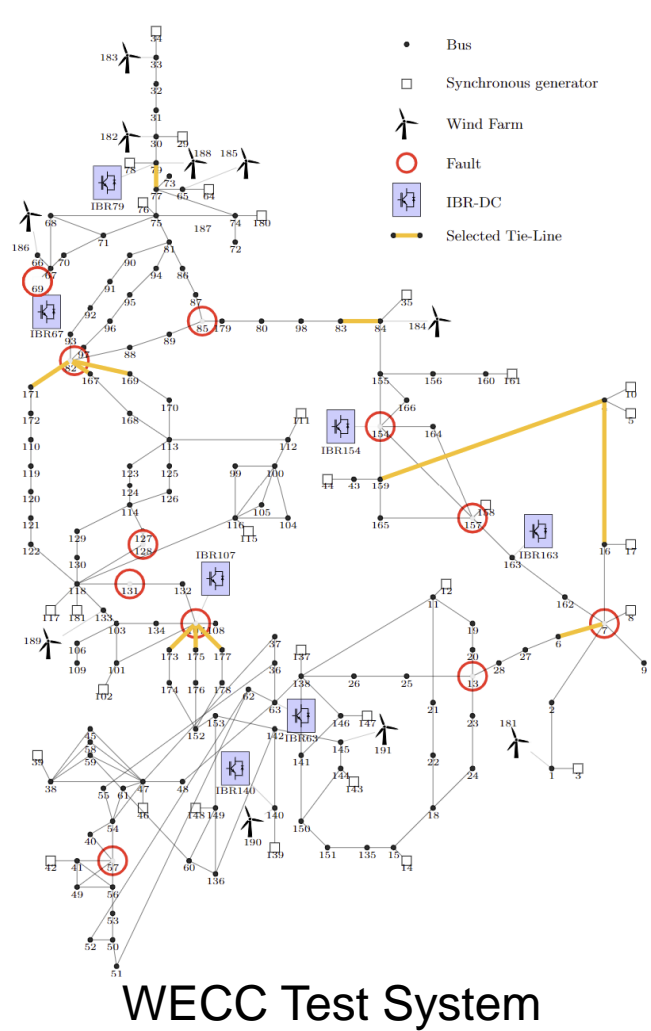
$$E(t) = \sum_{j=1}^p H_j \omega_s \Delta \omega_j^2(t)$$



Damping controllers' coordination framework

DATA-INFORMED COORDINATION

To achieve coordination, a deep neural network is trained as a total action function approximation algorithm. This network learns the nonlinear relationship between inputs, DC combinations, and total action.



Data collection and training process

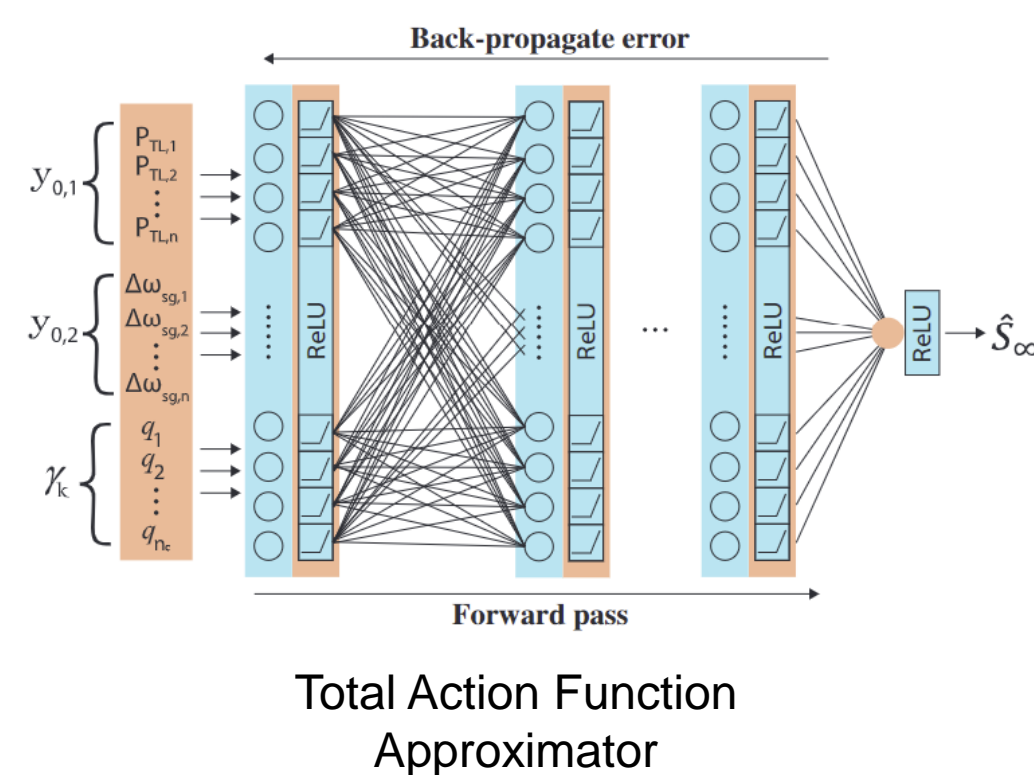
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Algorithm 1 Data collection
1: Define  $\mu_i \forall i \in \{1, \dots, n_{\mu}\}$  ▷ Vector of op. conditions
2: Define  $v_j \forall j \in \{1, \dots, n_d\}$  ▷ Vector of disturbances
3: Define  $\gamma_k \forall k \in \{1, \dots, 2^{n_c}\}$  ▷ Switching combination
4: for each  $\mu_i$  do
5:   Run power flow
6:    $y_{0,i} = [y_0]$ 
7:   for each  $v_j$  do
8:     Run time-domain simulation
9:     if  $t = t_0$  then
10:       $y_{0,j} = [\Delta \omega_{s,j}]$ 
11:      if  $t = t_1$  then
12:        Activate coordination  $\gamma_k$ 
13:      if  $t = t_2$  then
14:        Calculate total action  $S_{\infty, j, k}$ 
15:        Store in  $\Phi$  the subset  $\phi_{j,k} = \{S_{\infty, j, k}, y_{0,i}, y_{0,j}, \gamma_k\}$ 

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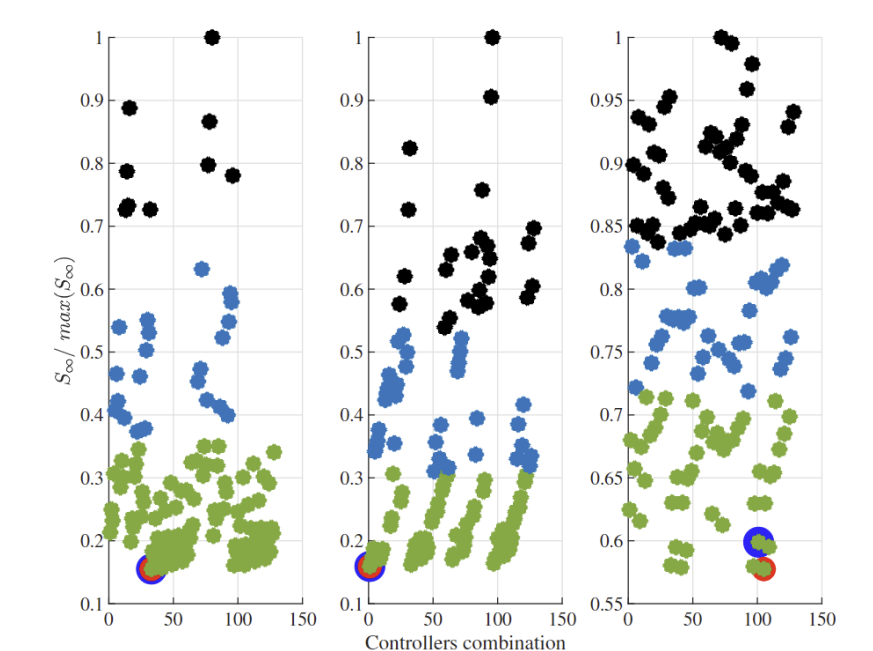
DigSILENT | PowerFactory

python TensorFlow



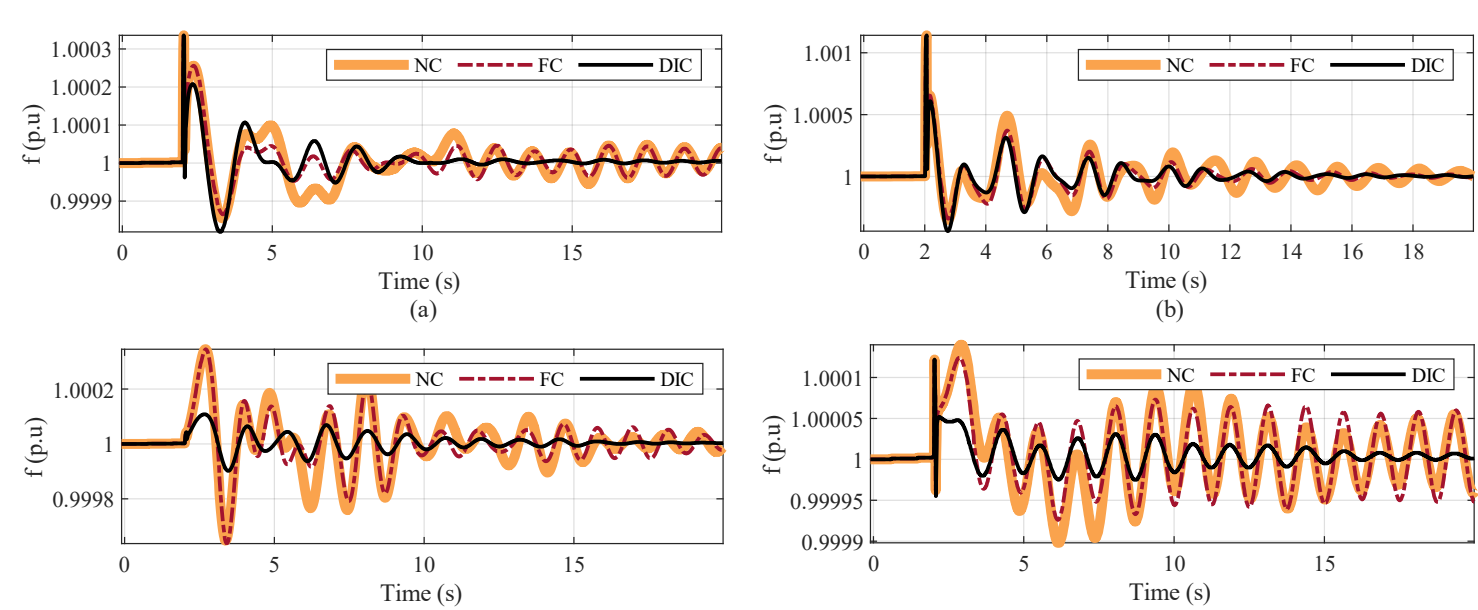
Total Action Function Approximator

$$\gamma^* = \arg \min_{\gamma_k \forall k \in \{1, \dots, n_c\}} \hat{S}_\infty(y_0, \gamma_k)$$

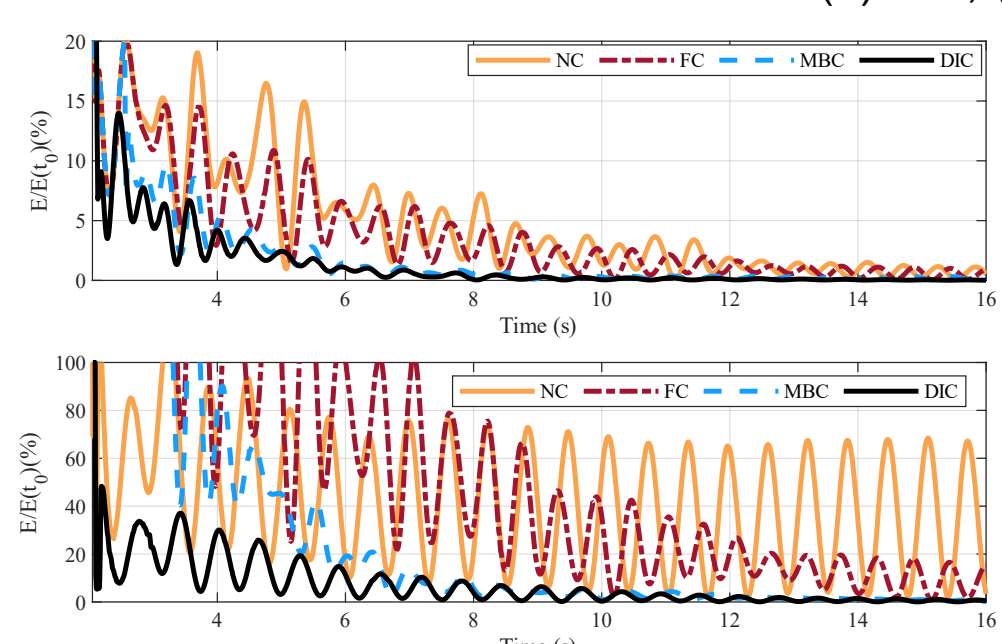


Controllers' combination search space for three different cases

EVALUATION



Frequency comparison for different control scenarios. Frequency at bus: (a) 117, (b) 14, (c) 161, and (d) 69.



Comparison between MBC and DIC for short-circuit at bus: (a) 157, and (b) 69.

TABLE II: TA reduction per control scheme

Coordination	Short-circuit at bus 157		Short-circuit at bus 69	
	Total Action	Reduction (%)	Total Action	Reduction (%)
No DC	1.241	-	0.840	-
FC	0.998	19.60	0.998	-18.79
DIC	0.328	73.53	0.295	64.87
MBC	0.500	59.72	0.473	43.65

REMARKS

- The results show that adaptability can be achieved by using a DNN as a switching law to find the combination that minimizes TA.
- It is not necessary to consider all possible disturbances and operating conditions; a carefully selected subset has been proven to be sufficient for generalization.
- The computational time of the coordination is 78 ms per 1,000 tests.
- Coordination allows for the efficient use of IBR as DC.

REFERENCES

- Zelaya-Arrazabal, F., Pulgar-Painemal, H., Liu, J., Li, F., & Silva-Saravia, H. (2023). Coordination of Damping Controllers: A Data-Informed Approach for Adaptability. arXiv preprint arXiv:2312.07739.
- Zelaya-Arrazabal, F., Liu, J., Zhao, J., Pulgar-Painemal, H., Li, F., & Silva-Saravia, H. (2022, October). Data-driven adaptive dynamic coordination of damping controllers. In 2022 North American Power Symposium (NAPS) (pp. 1-6). IEEE.

