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Nomenclature

Roman symbols

Wind turbine swept surface \boldsymbol{A}

CDC bus capacity

 I_t One-mass wind turbine aggregated inertia

 K_{C_p} Constant tip speed ratio speed control law coefficient

 L_l Inductance of the grid connection impedance

PGenerator pole pairs

 Q_s Generator stator reactive power

RWind turbine radius

 V_{DC} DC bus voltage

 V_{bus} Infinite bus voltage

Aerodynamic power coefficient

System frequency

 f^r Rated frequency

 r_s Resistance of a single phase of the stator windings

Resistance of the grid connection impedance r_l

Rated collection grid voltage

Rated wind turbine output voltage

Wind speed

Greek symbols

 Γ_m Generator torque

 Γ_t Wind turbine torque

β Pitch angle

Flux linkage per rotating speed unit λ_m

Gearbox multiplication ratio

Nominal generator speed ω_{mn}

Generator electrical angle speed ω_r

Wind turbine speed ω_t

Air density ρ

Time constant of the pitch angle controller

 θ_m Generator shaft angular position

Generator rotor electric angle θ_r

Superscripts and Subscripts

Reference value

abcVector of abc components

Vector of qd components qd

Variable related to the generator shaft m

rVariable related to the generator rotor

Variable related to the generator stator s

Variable related to the turbine t

Variable related to the grid connection point z

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A. WIND FARM MODEL

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In this section, we provide the full model of the wind turbine used in the simulations. The wind turbine model consists of several sub-blocks. We summarise each block in terms of inputs, outputs and detailed equations. The Park and inverse Park transformation of a rotation angle θ are indicated by $T(\theta)$ and $T^{-1}(\theta)$, respectively.

1) Wind turbine dynamics:

Block inputs: θ_m^* , Γ_m and v_w . Block outputs: θ_m and ω_m .

$$\beta(s) = \frac{1}{\tau s + 1} \theta_m^*(s),$$

$$c_p(\Lambda, \beta) = c_1 (c_2 \frac{1}{\Lambda} - c_3 \beta - c_4 \beta^{c_5} - c_6) e^{-c_7 \frac{1}{\Lambda}},$$

$$\Gamma_t = \frac{1}{2} c_p \rho A v_w^3 \frac{1}{\omega_t},$$

$$\omega_t = \frac{1}{I_t} (\Gamma_t + \nu \Gamma_m),$$

$$\dot{\theta}_m = \omega_m = \nu \omega_t,$$

$$(1)$$

where $[c_1\cdots c_9]$ are wind turbine characteristic parameters and Λ is defined as $\frac{1}{\Lambda}=\frac{1}{\lambda+c_8\beta}-\frac{c_9}{1+\beta^3}$ with $\lambda=\frac{\omega_t R}{v_w}$.

2) Wind turbine speed controller:

Block input: ω_m .

Block outputs: Γ_m^* and θ_m^* .

$$K_{C_p} = \frac{1}{2} \rho A R^3 \frac{c_1 (c_2 + c_6 c_7)^3 e^{-\frac{c_2 + c_6 c_7}{c_2}}}{c_2^2 c_7^4}$$

$$\Gamma_m^* = \frac{1}{\nu^3} K_{C_p} \omega_m^2,$$

$$\theta_m^*(s) = \frac{K_p s + K_i}{s} (\omega_m(s) - \omega_{mn}).$$
(2)

3) Generator dynamics:

Block inputs: v_s^{qd} and ω_m . Block outputs: i_s^{qd} , i_s^{abc} and Γ_m .

$$v_s^{qd} = \begin{bmatrix} r_s & \omega_r L_d \\ -\omega_r L_q & r_s \end{bmatrix} i_s^{qd} + \begin{bmatrix} L_d & 0 \\ 0 & L_q \end{bmatrix} \frac{d}{dt} i_s^{qd} + \lambda_m \omega_r \begin{bmatrix} 1 \\ 0 \end{bmatrix},$$

$$\Gamma_m = \frac{3}{2} P(\lambda_m i_{sq} + (L_d - L_q) i_{sq} i_{sd}),$$

$$i_s^{abc} = T^{-1}(\theta_r) i_s^{qd},$$
(3)

4) Generator vector controller:

Block inputs: Γ_m^* , i_s^{abc} , ω_m and θ_m . Block outputs: v_s^{qd} and v_s^{abc} .

$$\begin{split} i_{s}^{qd} &= T(\theta_{r})i_{s}^{qbc} \\ i_{sq}^{*} &= \frac{2}{3P} \frac{\Gamma_{m}^{*}}{\lambda_{m}}, \\ i_{sd}^{*} &= \frac{2}{3P} \frac{Q_{s}^{*}}{\omega_{m}\lambda_{m}}, \\ \hat{v}_{sq} &= \frac{K_{pq}s + K_{iq}}{s} (i_{sq}^{*} - i_{sq}), \\ \hat{v}_{sd} &= \frac{K_{pd}s + K_{id}}{s} (i_{sd}^{*} - i_{sd}), \\ v_{s}^{qd} &= \hat{v}_{s}^{qd} + \begin{bmatrix} 0 & \omega_{r}L_{d} \\ -\omega_{r}L_{q} & 0 \end{bmatrix} i_{s}^{qd} + \lambda_{m}\omega_{r} \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \\ v_{s}^{abc} &= T^{-1}(\theta_{r})v_{s}^{qd}. \end{split}$$

where $\omega_r = P\omega_m$, $\theta_r = P\theta_m$, $v_s^{qd} = [v_{sq}, v_{sd}]^{\mathrm{T}}$, $\hat{v}_s^{qd} = [\hat{v}_{sq}, \hat{v}_{sd}]^{\mathrm{T}}$ and $i_s^{qd} = [i_{sq}, i_{sd}]^{\mathrm{T}}$.

5) DC bus dynamics:

Block inputs: i_l^{abc} , v_l^{abc} , i_s^{abc} and v_s^{abc} .

Block outputs: V_{DC} and i_{DCm} .

$$\frac{d}{dt}V_{DC} = \frac{1}{C}(i_{DCm} - i_{DCl}),$$

$$\{v_l^{abc}\}^{\mathrm{T}}i_l^{abc} = V_{DC}i_{DCl},$$

$$\{v_s^{abc}\}^{\mathrm{T}}i_s^{abc} = V_{DC}i_{DCm}.$$

6) Grid side system dynamics:

Block inputs: v_z^{abc} and v_l^{abc} .

Block output: $i_{l}^{\tilde{a}bc}$

$$v_l^{abc} = r_l i_l^{abc} + L_l \frac{d}{dt} i_i^{abc} + v_z^{abc}.$$

7) Grid side controller:

Block inputs: V_{DC} and i_{DCm} .

Block output: i_{DCL}^*

$$i_{DCl}^* = \frac{K_{pg}s + K_{ig}}{s}(V_{DC}^* - V_{DC}) + i_{DCm}$$

(7)

8) Grid current controller

Block inputs: $V_{D\!C}$, i_{DCl}^* , v_z^{abc} and i_l^{abc} .

Block output: v_l^{abc} .

$$\begin{split} i_{lq}^* &= \frac{2}{3} \frac{V_{DC}}{v_{zq}} i_{DCl}^*, \\ i_{ld}^* &= 0, \\ \hat{v}_{lq} &= \frac{K_{pc} s + K_{ic}}{s} (i_{lq}^* - i_{lq}), \\ \hat{v}_{ld} &= \frac{K_{pc} s + K_{ic}}{s} (i_{ld}^* - i_{ld}), \\ v_{lq} &= v_{zq} - 2\pi f i_{ld} L_l - \hat{v}_{lq}, \\ v_{ld} &= 2\pi f i_{lq} L_l - \hat{v}_{ld}, \\ \hat{\omega} &= \frac{s + 0.129}{s} v_{zd}, \\ i_{l}^{qd} &= T(2\pi f t - \hat{\omega} t) i_{l}^{abc}, \\ v_{l}^{qd} &= T(2\pi f t - \hat{\omega} t) v_{l}^{abc}. \end{split}$$

where $i_l^{qd} = [i_{lq}, i_{ld}]^{\mathrm{T}}$, $v_l^{qd} = [v_{lq}, v_{ld}]^{\mathrm{T}}$ and $v_z^{qd} = [v_{zq}, v_{zd}]^{\mathrm{T}}$.

B. WIND FARM PARAMETERS

In this section, we summarise the parameters of the wind farm used in the simulation.

- 1) Wind turbine: $c_1 = 1$, $c_2 = 39.52$, $c_6 = 2.04$, $c_7 = 14.47$, $c_3 = c_4 = c_5 = c_8 = c_9 = 0, R = 40 \text{m}, A = 5,026.5 \text{m}^2,$ $\rho = 1.225 {\rm kg/m^3}, \ \nu = 90, \ I_t = 4 {\rm kg \cdot km^2}, \ \tau = 0.1 {\rm s} \ {\rm and}$
- 2) Wind turbine speed controller: $\omega_{mn} = 1,602 \mathrm{min}^{-1}, K_v =$ $0.1^{\circ} \cdot s/\mathrm{rad}$ and $K_i = 0.02^{\circ}/\mathrm{rad}$.
- 3) Generator: 2 pairs of poles, $r_s = 15 \text{m}\Omega$, $\lambda_m = 2.35 \text{V} \cdot \text{s/rad}$, $L_q = 0.12732 \mathrm{mH}$ and $L_d = 0.12764 \mathrm{mH}$.
- 4) Generator vector controller: $Q_s^*=10{
 m VAr},~K_{pq}=0.0637{
 m V/A},~K_{iq}=7.5{
 m V/(A\cdot s)},~K_{pd}=0.0638{
 m V/A}$ and $K_{id} = 7.5 \mathrm{V/(A \cdot s)}.$
- 5) DC bus: $C=10 \mathrm{mF}$ and $V_{DC}^*=2.6 \mathrm{kV}$. 6) Grid side system: $r_l=20 \mathrm{m}\Omega,~L_l=1 \mathrm{mH},~U_w^r=0.97 \mathrm{kV},$ $U_g^r=66 {
 m kV}$ and $f^r=50 {
 m Hz}$. 7) Grid side controller: $K_{pg}=0.6032 {
 m A/V}$ and $K_{ig}=$
- $14.2122A/(V \cdot s)$.
- 8) Grid current controller: $K_{pc} = 0.2803 \text{V/A}$ and $K_{ic} =$ $10V/(A \cdot s)$.

C. IEEE 14-BUS SYSTEM

In this section, we introduce the IEEE 14-bus model. This standard power system consists of 14 buses, 5 generators, 11 loads and 20 transmission lines. The diagram of the IEEE 14-bus model is presented in Fig. 1 [1].

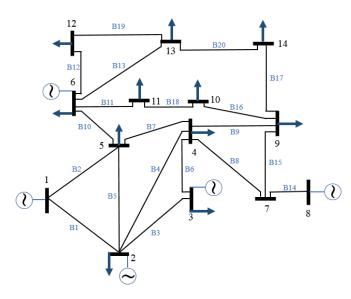


Fig. 1. The diagram of IEEE 14-bus system, indicating generators, loads positions and transmission lines.

D. IEEE 14-BUS SYSTEM PARAMETERS

In this section, we summarise the bus and line parameters of the IEEE 14-bus system in Table I and Table II, respectively.

TABLE I BUS PARAMETERS

Bus	Generation		Load	
No.	Real(MW)	Reactive(MVAR)	Real(MW)	Reactive(MVAR)
1	100.0	-16.9	0.0	0.0
2	100.0	42.4	21.7	12.7
3	100.0	23.4	94.2	19.0
4	0.0	0.0	47.8	3.9
5	0.0	0.0	7.6	1.6
6	100.0	12.2	11.2	7.5
7	0.0	0.0	0.0	0.0
8	100.0	17.4	0.0	0.0
9	0.0	0.0	29.5	16.6
10	0.0	0.0	9.0	5.8
11	0.0	0.0	3.5	1.8
12	0.0	0.0	6.1	1.6
13	0.0	0.0	13.5	5.8
14	0.0	0.0	14.9	5.0

TABLE II Line parameters

Bus	Between	Line impedance		Half line
No.	buses	Resistance(pu)	Reactance(pu)	Susceptance(pu)
1	1-2	0.01938	0.05917	0.02640
2	2-3	0.04699	0.19797	0.02190
3	2-4	0.05811	0.17632	0.01870
4	1-5	0.05403	0.22304	0.02460
5	2-5	0.05695	0.17388	0.01700
6	3-4	0.06701	0.17103	0.01730
7	4-5	0.01335	0.04211	0.0064
8	5-6	0.0	0.25202	0.0088
9	4-7	0.0	0.20912	0.0085
10	7-8	0.0	0.17615	0.0090
11	4-9	0.0	0.55618	0.0082
12	7-9	0.0	0.11001	0.0084
13	9-10	0.03181	0.08450	0.0079
14	6-11	0.09498	0.19890	0.0087
15	6-12	0.12291	0.25581	0.0092
16	6-13	0.06615	0.13027	0.0086
17	9-14	0.12711	0.27038	0.0089
18	10-11	0.8205	0.19207	0.0077
19	12-13	0.22092	0.19988	0.0098
20	13-14	0.17093	0.34802	0.0091

E. ROBUSTNESS ANALYSIS OF ALGORITHMS

To analyse the robustness of Algorithm 1 and Algorithm 2, we add Gaussian noise to the measurements with the signal-noise-ratio (SNR) of 60dB and 40dB, respectively. We estimate the quantity $C\Pi$ by Algorithm 1 and the quantity ΥB by Algorithm 2 using the data containing the noise. Fig. 2 to Fig. 5 concludes the performances of Algorithm 1 and Algorithm 2 under different levels of noise by showing the regions between the upper and lower envelopes of all obtained magnitude plots of the ROMs obtained for the 20 realisations of the noise. In particular, the red regions refer to the estimation with sufficient measurements ($\widetilde{\nu}=1000$ or $\widetilde{q}=1000$), the blue regions refer to the estimation with limited measurements ($\widetilde{\nu}=30$ or $\widetilde{q}=1$) and the brown dashed lines represent the ideal estimation (no noise). These figures confirm that the proposed data-driven method is robust with respect to the measurement noise.

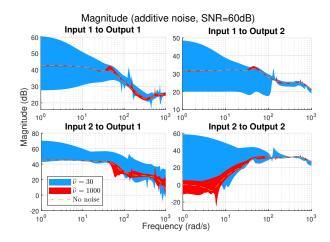


Fig. 2. Magnitude plots of the reduced-order models under 20 realizations of additive Gaussian noise. The red regions, blue regions and brown dashed lines refer to the estimation with 1000 measurements ($\tilde{\nu}=1000$), estimation with 30 measurements ($\tilde{\nu}=30$) and the ideal estimation (no noise), respectively.

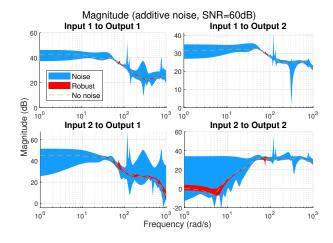


Fig. 3. Magnitude plots of the reduced-order models under 20 realizations of additive Gaussian noise. The red regions, blue regions and brown dashed lines refer to the robust algorithm (Algorithm 2), non-robust method (Theorem 2) and the ideal estimation (no noise), respectively.

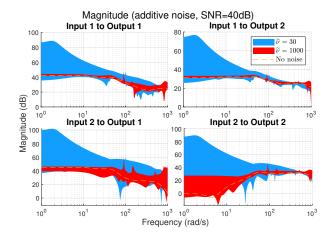


Fig. 4. Magnitude plots of the reduced-order models under 20 realizations of additive Gaussian noise. The red regions, blue regions and brown dashed lines refer to the estimation with 1000 measurements ($\widetilde{\nu}=1000$), estimation with 30 measurements ($\widetilde{\nu}=30$) and the ideal estimation (no noise), respectively.

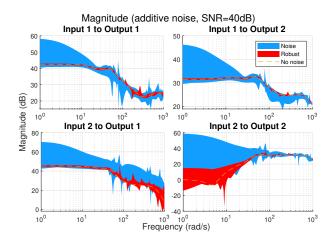


Fig. 5. Magnitude plots of the reduced-order models under 20 realizations of additive Gaussian noise. The red regions, blue regions and brown dashed lines refer to the robust algorithm (Algorithm 2), non-robust method (Theorem 2) and the ideal estimation (no noise), respectively.

F. CODE AVAILABILITY

We develop a demo to illustrate the proposed data-driven model order reduction (MOR) method. Provided with a simple benchmark model of the multiple-input-multiple-output (MIMO) systems, the corresponding reduced-order model (ROM) and bode plots are produced by this illustrative demo. For more benchmark models, see [2]. The demo is available at https://github.com/zilong-gong/WindFarmMOR.

REFERENCES

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