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# Recurrent modified Elman neural network control of PM synchronous generator system using wind turbine emulator of PM synchronous servo motor drive

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#### ABSTRACT

The recurrent modified Elman neural network (NN) controlled a permanent magnet (PM) synchronous generator system, which is driven by wind turbine emulator of a PM synchronous motor servo drive, is developed to regulate output voltage of rectifier (or AC to DC power converter) and inverter (or DC to AC power converter) in this study. First, the wind turbine emulator of a closed loop PM synchronous motor servo drive is designed to produce the maximum power for the PM synchronous generator system. Then, the rotor speed of the PM synchronous generator, the output DC bus voltage and current of the rectifier are detected simultaneously to yield maximum power output of the rectifier through DC bus power control. Because the PM synchronous generator system is a nonlinear and time varying dynamic system, the online training recurrent modified Elman NN control system with the variable learning rate is derived based on Lyapunov stability theorem, so that the stability of the system can be guaranteed. Finally, some experimental results are verified to show the effectiveness of the proposed recurrent modified Elman NN control system.

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#### 1. Introduction

Since the petroleum gradually exhausting and environmental protection progressively rising, the usage of the clean energy sources has become very importance in the electric power applications. Clean energy sources such as wind, photovoltaic, and fuel cells can be interfaced to a multilevel converter system for a high power application [1–3].

Wind turbine acted as sources of energy has progressively increased in the whole earth. A few control methods and conversion technologies in wind energy application are developed in the wind energy conversion systems. The PM synchronous generator system has been used for wind power generating system due to many advantages such as simpler structure, better reliability, lower maintenance and higher efficiency [4–8]. Therefore, the PM synchronous generator generation system stands for a significant trend in progress of wind power applications [4–8]. The output power behavior of wind turbine is nonlinear. The provided power of vertical-axis turbines is very sensitive to the load variation due to different structure effect [4–8]. Thus, the control of operating point is indispensable for maximum output power. The controllable rectifier is used to convert varied AC voltage generated by PM synchronous generator into DC bus voltage. Then, the controllable inverter is used to convert DC bus voltage into AC at a fixed frequency in order to provide for the stand alone or grid applications of electrical utilizations. The major purposes utilized wind turbines are to extract maximum power of turbine and deliver appropriate energy to stand alone power or grid power. According to these purposes, the better structure of the power conversion in wind turbines is the AC to DC to AC power converter [9,10].

Over the past decade, many different control approaches were used to generator and wind turbine emulator for energy generation [11–18]. In [11,12], a fuzzy logic control is adopted to control the power of the wind electrical conversion system transmitted to the grid and generator speed. The advantage in using fuzzy logic controller against standard proportional-integral (Pl) controller, is pointed out in better response to frequently changes in wind speed. In [13], the rule-based fuzzy-logic based maximum power point tracking (MPPT) strategy is proposed for PM synchronous generator variable speed wind turbine generation systems. The fuzzy-logic-based output power smoothing method of a wind energy conversion system (WECS) with a PM synchronous generator using the inertia of wind turbine was proposed in [14]. In [15], a sliding mode control (SMC) strategy associated to the field-ori-





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### Nomenclature

- d-axis and q-axis stator voltages of the PM synchronous  $v_{d1}, v_{q1}$ generator
- *d*-axis and *q*-axis stator currents of the PM synchronous  $i_{d1}, i_{a1}$ generator
- *d*-axis and *q*-axis inductances of the PM synchronous  $L_{d1}, L_{q1}$ generator
- phase winding resistance of the PM synchronous gener- $R_{s1}$ ator
- $\omega_{r1}, n_{r1}$ rotor angular velocity of the PM synchronous generator in rad/s, in rpm
- permanent magnet mutual flux linkage of the PM syn- $\lambda_{pm}$ chronous generator
- Р number of poles of the PM synchronous generator
- tip ratio of the wind turbine ß
- turbine rotor radius of the wind turbine  $R_1$
- output torque of the wind turbine  $T_1$ electromagnetic torque of the PM synchronous genera-Te tor
- moment of inertia of the PM synchronous generator  $J_1$
- $B_1$ viscous friction coefficient of the PM synchronous gen-
- erator
- density of air  $\rho_1$
- Kt torque constant of the PM synchronous generator  $A_1$ exposed area of the wind turbine
- $D_p(\beta)$ coefficient of power performance
- $P_1$ output mechanical power of the wind turbine wind speed  $v_1$
- $\theta_{r1}$ rotor position of the PM synchronous generator
- electric angular angle of the inverter  $\theta_{i1}$
- $i_{dr}^{*}, i_{qr}^{*}, i_{dr}^{*}, i_{ar}^{*}, i_{br}^{*}, i_{cr}^{*}$ *d*-axis and *q*-axis control currents of the rectifier
- - desired phase currents of the PM synchronous generator in phases ar, br and cr
- *iar*, *ibr*, *icr* measured phase currents of the PM synchronous generator in phases ar, br and cr
- Tar, Tbr, Tcr sinusoidal pulse-width-modulation control signals of the rectifier in phases ar, br and cr
- $V_d, V_d^*$ measured magnitude of the DC bus voltage of the rectifier, desired magnitude of the DC bus voltage of the rectifier
- *d*-axis and *q*-axis control currents of the inverter
- $i_{ai}^{*}, i_{qi}^{*}$  d-axis and q-axis control currents of the inverter  $i_{ai}^{*}, i_{bi}^{*}, i_{ci}^{*}$  desired phase currents of the inverter in phases ai, biand ci
- $i_{ai}$ ,  $i_{bi}$ ,  $i_{ci}$  measured phase currents of the inverter in phases ai, biand ci
- T<sub>ai</sub>, T<sub>bi</sub>, T<sub>ci</sub> sinusoidal pulse-width-modulation control signals of the inverter in phases *ai*, *bi* and *ci*
- $V_{rms}$ ,  $V_{rms}^*$  measured root-mean-square magnitude of the AC 60 Hz line voltage of the inverter, desired root-meansquare magnitude of the AC 60 Hz line voltage of the inverter
- $c_{i,m}^1, c_{i,m}^2, c_{j,m}^3, c_{k,m}^4$  inputs of nodes i in the input layer, the hidden layer, the context layer and the output layer of the mth recurrent modified Elman NN
- $g_{i,m}^1, g_{j,m}^2, g_{k,m}^3, g_{0,m}^4$  activation functions in the input layer, the hidden layer, the context layer and the output layer of the mth recurrent modified Elman NN
- $node_{i,m}^1$ ,  $node_{j,m}^2$ ,  $node_{k,m}^3$ ,  $node_{o,m}^4$  node functions in the input layer, the hidden layer, the context layer and the output layer of the *m*th recurrent modified Elman NN
- $d_{i,m}^1, d_{j,m,}^2, \bar{d}_{k,m}^3, d_{o,m}^4$  output of nodes in the input layer, the hidden layer, the context layer and the output layer of the mth recurrent modified Elman NN
- Ν
- $l_1$

- total number of node in context laver n
- self-connecting feedback gain of context layer in the  $\alpha_m$ mth recurrent modified Elman NN
- connective weights between input wavelet layer (i  $\mu_{ii.m}^2$ layer) and hidden layer (*j* layer) in the *m*th recurrent modified Elman NN
- connective weights between context wavelet layer (k  $\mu_{ki.m}^3$ layer) and hidden layer (*j* layer) in the *m*th recurrent modified Elman NN
- $\mu_{io.m}^4$ connective weights between hidden layer (*i* layer) and output layer (o layer) in the mth recurrent modified Elman NN
- $\mu^1_{oim}$ recurrent weights between output layer (o layer) and input layer (*i* layer) in the *m*th recurrent modified Elman NN
- $d_{01}^4 = U_{R1}, \ d_{02}^4 = U_{R2}$  output of the output layer in the first *m*th recurrent modified Elman NN, output of the output layer in the second recurrent modified Elman NN
- $\psi_m$ collections vector of the adjustable parameters in *m*th recurrent modified Elman NN
- inputs vector of the output layer in the *m*th recurrent  $\chi_m$ modified Elman NN
- learning rates in the *m*th recurrent modified Elman NN,  $\gamma_m, \gamma_m^*$ optimal learning rates in the *m*th recurrent modified Elman NN
- $V_{c,m}$ , cost function in the *m*th recurrent modified Elman NN
- change of connective weights between context layer (k  $\Delta \mu_{ki.m}^4$ layer) and hidden layer (*j* layer) in the *m*th recurrent modified Elman NN
- error term in the mth recurrent modified Elman NN  $\rho_{j,m}$
- discrete-type Lyapunov function in the *m*th recurrent  $L_{2,m}$ modified Elman NN
- $P_{j,m} \equiv \partial d_{j,m}^2 / \partial u_{kj,m}^3$  partial differentiation with respective to connective weights between the context layer and the hidden layer in the mth recurrent modified Elman NN
- $Q_{j,m} \equiv \partial d_{j,m}^2 / \partial u_{ij,m}^2$  partial differentiation with respective to connective weights between the input layer and the hidden layer in the mth recurrent modified Elman NN
- $R_{j,m} \equiv \partial d_{j,m}^2 / u_{o,m}^1$  partial differentiation with respective to recurrent weights between the output layer and the input layer in the mth recurrent modified Elman NN  $u^*_{rm,m}$ ideal neural controller
- minimum approximation error  $\mathcal{E}_m$
- $F_m(s)$ error function of the control effort for the rectifier and the inverter
- $\delta_m$ approximation error bound
- error tracking index  $Z_m$
- $L_{1,m}$ Lyapunov function
- uncertain bound  $\delta_m$
- $\hat{\delta}_m$ estimation value of uncertain bound
- estimation error of uncertain bound  $\delta_m$
- $k_{p1}, k_{p2}$ the constants of the output voltage for rectifier and inverter
- Vabr the line-line output voltage of the PM synchronous generator
- the regulated values of the output voltage for rectifier ζ1, ζ2 and inverter
- control effort of control system for rectifier and inverter  $\Delta \zeta_m$
- recurrent modified Elman NN control system for recti $u_{rmc,m}$ fier and inverter
- recurrent modified Elman NN controller for rectifier and  $u_{rm,m}$ inverter
- compensated controller for rectifier and the inverter  $u_{c,m}$
- the Euclidean norm in  $\Re^n$  $\|\cdot\|$

number of iterations total number of node in hidden layer Download English Version:

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