Modeling and Fuzzy Command of a Wind Generator

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To cite this article:
Nejib Hamrouni, Amel Ghobber, Moncef Jraidi, Ahmed Dhouib. Modeling and Fuzzy Command of a Wind Generator. Science Journal of Circuits, Systems and Signal Processing. Vol. 6, No. 4, 2017, pp. 35-43. doi: 10.11648/j.cssp.20170604.11

Received: November 14, 2017; Accepted: November 30, 2017; Published: January 2, 2018

Abstract: A problem of mechanical modeling and robustly stabilization of a wind generator is considered. To overcome the non-linearity of the system, the model of the wind generator is approximated by a Takagi-Sugeno fuzzy model. To stabilize the obtained fuzzy model, two command approaches were developed. They are the fuzzy controller using the parallel distributed compensation (PDC) and the $H_\infty$ controller based-fuzzy observer. Numerical optimization problems using linear matrix inequality (LMI) and convex techniques are used to analyze the stability of the wind generator. Finally, simulation examples illustrating the control performance and dynamic behavior of the wind generator under various command approaches are presented.

Keywords: Modeling, TS-Fuzzy Control, $H_\infty$ Command, LMI Approach, Stability

1. Introduction

Wind energies were well developed and exploited and became competitive thanks to the evolution of the power electronics. Nevertheless, the optimal exploitation of the renewable energy sources encountered many problems which are related mainly to the uncertain variation of the wind speed. To resolve these problems and to improve the penetration rate and the stability level of the connected wind system, several controls are developed focusing this subject. These commands illustrate performance limits such as instability and implementation complexity.

Recently many studies have been devoted to the stability of the non-linear system. Fuzzy control constitutes a preferment command which has attracted an increasing attention because it can offer an effective solution to the control of complex, uncertain and undefined system [1-3]. The well-known Takagi-Sugeno (TS) fuzzy method has become a convenient approach for dealing with a complex system. This approach provides an effective representation of the system with the aid of fuzzy sets, fuzzy rules and a local linear model. The Takagi Sugeno fuzzy model [4] constitutes a productive way to describe and control the dynamics of nonlinear systems [5]. The TS fuzzy dynamics model is a system described by the fuzzy ‘if-then’ rules which offer local linear representation. The advantages of using this approach for design control are that the closed loop stability analysis using Lyapunov method becomes easier to apply and the controller synthesis can be reduced to convex problem [6]. Once the nonlinear models are transferred to a fuzzy model, control design can be carried out. Some of them are fuzzy controller using parallel distribution compensation (PDC) and the $H_\infty$ controller based-fuzzy observer. The PDC approach employs multiple linear controllers corresponding to the locally linear plants [7]. Many applications illustrate the effectiveness of the TS models and parallel distribution compensation [4, 8]. Fuzzy observer-based $H_\infty$ controller has been developed in [9, 10]. Stability and aptitude to reject exterior disturbances provided by this command approach were approved in [11].

In this work, the TS-fuzzy approach is used to approximate the nonlinear wind generator model. Two approaches of robust and powerful commands to stabilize the mechanical part of the wind generator are presented. In the first part, a fuzzy controller design uses the concept of PDC [7, 12] is studied. In the second part, a fuzzy observer-based $H_\infty$ controller is developed in order to improve the performance of the system and to minimize the disturbance effect of the wind speed. Sufficient stability conditions are expressed in terms of LMI which can be solved very efficiently using convex optimization techniques [6]. Finally, simulation examples are given to illustrate both stability and robustness analysis of the proposed control systems.
2. Description and Modeling of the Wind Generator

The drive train of the wind turbine generator system consists of the following elements: a blade-pitching mechanism, a hub with blades, a rotor shaft, a gearbox and generator. The common way to model the drive train is to treat the system as a number of discrete masses connected together by springs defined by damping and stiffness coefficients (Figure 1).

In order to simplify the control of the wind generator, we introduce some simplifications on the six mass models. The turbine inertia can be calculated from the combined weight of the blades and the hub. Therefore, the mutual damping and elasticity between the hub and the three blades is ignored ($K_{p1}=K_{p2}=K_{p3}\approx0$ and $f_{p1}=f_{p2}=f_{p3}\approx0$). The torque losses of the blades ($\alpha_{p1}, \alpha_{p2}$ and $\alpha_{p3}$) are ignored because turbine speed is very weak. Moreover, it is assumed that the three blades have uniform weight distribution ($F_{p1}=F_{p2}=F_{p3}$) and the turbine torque is assumed to be equal to the sum of the torque acting on the various blades. Thus, the turbine can be looked as a large disk with small thickness and the wind system (three blades, hub, gearbox and generator) can be modeled by three masses coupled through a gearbox as indicated by Figure 2.

![Figure 1. Transmission model of 6 masses connected together.](image1)

![Figure 2. Three-mass model.](image2)
In addition, compared with the mechanical characteristics of the blades and the generator, the gearbox inertia and mutual damping with adjacent masses moment of inertia are ignored. Therefore, the complex system can be simplified to be two-mass model through reforming the dynamic system with the equivalent stiffness and damping coefficients. As a consequence, the two-mass model contains two parts which represent the wind turbine and the generator. They are connected by a flexible shaft. This model will be used to investigate the control of the wind turbine.

\[
\begin{align*}
J_b, \theta_1, \Omega_b, C_1 \\
F_p \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow 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The TS system is defined as follows:

\[ y = \Omega_y \quad \text{and} \quad C = [0 \quad 0 \quad 1 \quad 0] \quad \text{; } z = \begin{bmatrix} V \\ \beta \end{bmatrix} \]

### 2.1. TS Fuzzy Model

The objective of this approach is to approximate the nonlinear system by linearized sub-systems. It has been used as an alternative to classical models to capture dynamic performances under different operating conditions and in different functioning zones.

The description of the nonlinear system in terms of “If-Then” rules combined with a mathematical description of non-linear systems is called a Takagi-Sugeno fuzzy model. The concept of sector nonlinearity provided means for exact approximation of nonlinear system by linearized sub-systems. It has been used to transform the scalar functions and the bornitude of the continuous variables \( V \) and \( \beta \). These variables are limited as given by the following equations.

\[
\begin{align*}
\forall \; (t) \in [0, 2...M] & \quad y(t) = C_i x(t) \\
\text{Where:} & \\
A_i & = \sum_{i=1}^{M} \mu_i(z(t)) (A_i x(t) + B_i u(t)) \\
B_i & = \sum_{i=1}^{M} \mu_i(z(t)) B_i \\
C_i & = \sum_{i=1}^{M} \mu_i(z(t)) C_i \\
\end{align*}
\]

The output signal is obtained by the same technique:

\[
y(t) = \sum_{i=1}^{M} \mu_i(x(t)) C_i x(t)
\]

The term \( \mu_i(z(t)) \) determines the activation terms of the associated local models. According to the functioning zone of the system, these terms indicate the contribution of the local model. They allow a progressive passage of a local model to another and they depend on the state vector. They can be in triangular or Gaussian forms. They satisfy the conditions given in [17].

\[
\begin{align*}
\sum_{i=1}^{M} \mu_i(x(t)) & = 1 \\
0 & < \mu_i(x(t)) \leq 1
\end{align*}
\]

### 2.2. TS Fuzzy Description of the Wind Generator

We consider the TS fuzzy models based on the nonlinear sectors [18, 19] to represent the nonlinear model of the wind system composed of a turbine, a driving shaft and an induction machine. This approach based on the transformation of the scalar functions and the bornitude of the continuous variables \( V \) and \( \beta \). These variables are limited as given by the following equations.

\[
V_{\min} \leq V \leq V_{\max} \quad \text{and} \quad \beta_{\min} \leq \beta \leq \beta_{\max}
\]

According to the model given by the equation (10), the system has two non-linearities depending on \( V \) and \( \beta \). To linearize this model, we will use the presentation of TS previously presented. For those non-linearities, the base comprises four rules “If-Then”. Thus, the nonlinear wind system is represented by the following fuzzy model:

\[
\begin{align*}
\text{If } \beta \text{ is } F_1 \text{ and } V \text{ is } F_2 & \text{ then} \\
\dot{x}(t) & = A_1 x(t) + B_1 u(t) + B_{21} w \\
y & = C_1 x(t)
\end{align*}
\]

\[
\begin{align*}
\text{If } \beta \text{ is } F_1 \text{ and } V \text{ is } F_2 & \text{ then} \\
\dot{x}(t) & = A_2 x(t) + B_1 u(t) + B_{22} w \\
y & = C_2 x(t)
\end{align*}
\]

\[
\begin{align*}
\text{If } \beta \text{ is } F_1 \text{ and } V \text{ is } F_2 & \text{ then} \\
\dot{x}(t) & = A_3 x(t) + B_1 u(t) + B_{23} w \\
y & = C_3 x(t)
\end{align*}
\]
If $\beta$ is $F_i^2$ and $V$ is $F_2^2$ then

$$\dot{\mathbf{x}}(t) = A_4 \mathbf{x}(t) + B_4 \mathbf{u}(t) + B_{24} \mathbf{w}$$

$$y = C_4 \mathbf{x}(t)$$

Then the inferred system is given by:

$$\begin{aligned}
\dot{\mathbf{x}}(t) &= \sum_{i=1}^{M} \mu_i(z(t)) \left[ (A_i \mathbf{x}(t) + B_i \mathbf{u}(t) + B_{2i} \mathbf{w}) \right] \\
y &= \sum_{i=1}^{M} \mu_i(z(t)) C_i \mathbf{x}(t)
\end{aligned}$$

(15)

Where $\mu_i(z(t))$ are activation terms of the subsystems. They are given by:

$$\begin{aligned}
\mu_1(z) &= F_1^2(\beta) F_2^2(V) \\
\mu_2(z) &= F_1^2(\beta) F_2^2(V) \\
\mu_3(z) &= F_1^2(\beta) F_2^2(V) \\
\mu_4(z) &= F_1^2(\beta) F_2^2(V)
\end{aligned}$$

With $F_i^2$ are the degrees of membership function of the activation terms. They are given by:

$$\begin{aligned}
F_1^2(\beta) &= \beta - \beta_1 & F_2^2(\beta) &= \beta - \beta_2 \\
F_1^2(\beta) &= \frac{1}{V - V_1} & F_2^2(\beta) &= \frac{1}{V_2 - V}
\end{aligned}$$

(17)

With $V_{1}=V_{\text{min}}, V_{2}=V_{\text{max}}, \beta_1=\beta_{\text{min}}$ and $\beta_2=\beta_{\text{max}}$.

For $i \in \{1,2, \ldots, M\}$, the matrices $A_1, B_1, B_{2i}$ and $C$ are given by:

$$A_1 = A_2 = \begin{bmatrix}
0 & 1 & -1 & 0 \\
0 & 0 & 0 & 0 \\
\frac{K_h}{j\omega} & \frac{D_2}{j\omega} & 0 & 0 \\
0 & 0 & 0 & -1
\end{bmatrix} \ ; \ A_3 = A_4 = \begin{bmatrix}
0 & 1 & -1 & 0 \\
0 & 0 & 0 & 0 \\
\frac{K_h}{j\omega} & \frac{D_2}{j\omega} & 0 & 0 \\
0 & 0 & 0 & -1
\end{bmatrix}
$$

$$B_{21} = B_{22} = \begin{bmatrix}
0 \\
0 \\
0 & \frac{C_{aer,\beta_1}}{j\omega} \\
0 & 0 & \frac{C_{aer,\beta_2}}{j\omega}
\end{bmatrix} \ ; \ B_1 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
\frac{D_2}{j\omega} & \frac{D_2}{j\omega} & 0 & 0 \\
0 & 0 & 0 & 0
\end{bmatrix} \ ; \ B_{22} = B_{24} = \begin{bmatrix}
0 & 0 \\
0 & 0 \\
0 & \frac{C_{aer,\beta_1}}{j\omega} \\
0 & 0 & \frac{C_{aer,\beta_2}}{j\omega}
\end{bmatrix} \ ; \ C = \begin{bmatrix}
0 & 0 & 1 & 0
\end{bmatrix}
$$

With:

$$C_{aer,\beta_1} = C_{aer,\beta}(\beta = \beta_1), \quad C_{aer,\beta_2} = C_{aer,\beta}(\beta = \beta_2), \quad C_{aer,\beta_1} = C_{aer,\beta}(V = V_1), \quad C_{aer,\beta_2} = C_{aer,\beta}(V = V_2)$$

$$u(t) = -\frac{\sum_{j=1}^{M} w_j(x(t)) K_j x(t)}{\sum_{j=1}^{M} w_j} = -\sum_{j=1}^{M} \mu_j(z(t)) K_j x(t)$$

(18)

Where $K_j$ are the feedback gain matrices for the subsystems.

3. Stabilisation of the Wind Generator

3.1. By Fuzzy Controller: Parallel Distribution Compensation

The concept of this command, named parallel distributed compensation (PDC), is used to stabilize the fuzzy system. It employs linear controllers corresponding to the locally linear plant model. The general diagram of the PDC is given by Figure 4.

The general structure of the PDC is given by [20]: If $z_1(t)$ is $M_1$ and $z_2(t)$ is $M_2$, ... , and $z_p$ is $M_p$ then $u(t) = -K \cdot x(t)$, $j=1,2,\ldots,M$. Consequently the overall PDC fuzzy controller is represented by:

$$\dot{\mathbf{x}}(t) = \sum_{i=1}^{M} \sum_{j=1}^{M} \mu_i(z(t)) \mu_j(z(t)) (A_i - B_i K_j) x(t) + B_{2j} \mathbf{w}$$

(19)

The fuzzy controller design consists in determining the local feedback gain $K_j$ (j=1, 2, 3, 4) for the corresponding parts of TS models so that the zero equilibrium of the closed loop fuzzy systems was globally stable. We apply the quadratic stability theorem of the global system, yields the following results [9]: the equilibrium of the closed loop system (19) is globally stable, if there is a common positive definite matrix $P$ which satisfies the following conditions:
\[
\begin{align*}
-XA_i^T - A_iX + B_iM_i + (B_iM_i)^T & > 0 \\
-XA_i^T - A_iX - XA_i^T - A_iX + M_i^T B_i^T + B_iM_i + M_i^T B_i^T + B_iM_i & \geq 0
\end{align*}
\] (20)

The objective of the command is in finding \( K_1, K_2, K_3 \) and \( K_4 \) and \( P > 0 \) that satisfy the conditions presented by equation 20. An approach to design a stable fuzzy controller for the wind generator is to transform the condition in equation 20 into convex problem [1, 21-23]. The solution of the stable PDC controller design problem via linear matrix inequalities (LMI) for the system leads to:

\[ K_j = M_jX^{-1} \text{ with } P = X^{-1} \] (21)

If the solution of the LMI (eq. 20) is found, it means that local state feedback gains \( K_j (j=1, 2, 3 \) and 4) provide quadratic stability of the closed loop TS fuzzy systems. Hence, using matlab toolbox, the controller parameters are found to be:

\[
K_1 = \begin{bmatrix}
-4225 \\
-1 \times 10^8
\end{bmatrix}
\]

\[
K_2 = \begin{bmatrix}
-4225 \\
-1 \times 10^8
\end{bmatrix}
\]

\[
K_3 = \begin{bmatrix}
-4225.5 \\
-1.08 \times 10^8
\end{bmatrix}
\]

\[
K_4 = \begin{bmatrix}
-4225.45 \\
-1.065 \times 10^8
\end{bmatrix}
\]

\[\ldots\]

\[\ldots\]

Table 1. Parameters of the wind system.

| Symbol | Quantity | Value |
|--------|----------|-------|
| \( K_e \) | Elasticity of driving shaft | 1.566 \times 10^6 Nm\(^{-1}\) |
| \( D_r \) | Damping factor shaft-nacelle | 3029.5 Nms rad\(^{-1}\) |
| \( D_g \) | Damping factor IG-gearbox | 15.993 Nms rad\(^{-1}\) |
| \( J_g \) | Inertia of IG | 5.9 kg m\(^2\) |
| \( J_p \) | Inertia of blades | 83000 kg m\(^2\) |
| \( \rho \) | Air density | 1.225 kg m\(^{-3}\) |
| \( R \) | Blade radius | 30.3 m |
| \( \tau_B \) | Time constant of the blade movement | 100 ms |
| \( P \) | Rated power of the IG | 1 MW |
| \( p \) | Pole pair number of the IG | 3 |
| \( \lambda_{opt} \) | Optimal specific speed | 7 |
| \( C_{p_{max}} \) | Coefficient of maximal power | 0.48 |
| \( \gamma \) | Desired level disturbance attenuation | 0.6 |
| \( C_{aer,\beta_1} \) | Aerodynamic torque corresponding to the pitch angle \( \beta_1 \) | 723980 Nm |
| \( C_{aer,\beta_2} \) | Aerodynamic torque corresponding to the pitch angle \( \beta_2 \) | 376070 Nm |
| \( C_{aer,V_1} \) | Aerodynamic torque corresponding to the wind speed \( V_1 \) | 106440 Nm |
| \( C_{aer,V_2} \) | Aerodynamic torque corresponding to the wind speed \( V_2 \) | 85370 Nm |

The choice of the wind speed profile is arbitrary. For that, we consider a sinusoidal form of a wind speed which varies between 17 and 32 km/h. The main objective is to test and analyze the performance of the PDC controller.

We notice that the form of the wind speed generates disturbances on the machine speed according to Figure 5a. The generator speed is not stable and oscillates (Figure 5b). These oscillations can cause vibrations and harmful effects.
on the driving shaft. The PDC controller is insufficient and does not allow a stability of the overall system. To overcome this problem, we will use another preferment command which is able to eliminate the oscillatory effect on the operating of the overall system.

3.2. By Fuzzy Observer-Based $H_\infty$ Controller

The approach is to develop a robust command which permits to obtain robustness stability in analytical way.

According to the TS system represented by equation (12), some state variables are un-measurable. Hence, it was obligatory to synthesize an observer which make possible to estimate those variables $(\theta, \beta)$. The design of fuzzy observer obliges that the local models of the system should be locally observables, and is obtained by interpolation of several Luenberger observer. The observer is introduced as follows:

If $\dot{z}_1(t) = F_{ii}$ and $\ldots \dot{z}_p(t) = F_{ip}$, then for $i=1 \ldots 4$, the observer states are governed by [24]:

$$
\dot{\hat{x}}(t) = \begin{cases}
A_i \hat{x}(t) + B_i u(t) + B_{2i} w(t) - L_i (y(t) - \hat{y}(t)) \\
\hat{y}(t) = C_i \hat{x}(t)
\end{cases}
$$

(22)

Where $L_i$ is the observer gain for the $i^{th}$ observer rule. The overall fuzzy observer is represented as follows:

$$
\dot{\hat{x}}(t) = \sum_{i=1}^{4} \mu_i(\hat{x}(t)) \left( A_i \hat{x}(t) + B_i u(t) + B_{2i} w(t) - L_i (y(t) - \hat{y}(t)) \right)
$$

$$
\hat{y}(t) = \sum_{i=1}^{4} \mu_i(\hat{x}(t)) C_i \hat{x}(t)
$$

(23)

Where $\hat{x}(t) \in \mathbb{R}^n$ and $\hat{y}(t) \in \mathbb{R}^h$ are respectively the estimated state and output vector.

The stabilization of the generator speed is essentially a disturbance rejection problem. Thus, a robust control using $H_\infty$ technique is well adopted to resolve this kind of control problem. It is assumed that the fuzzy systems are locally controllable. Hence, a fuzzy controller with the following rules can be used.

$$
\mu_i(\hat{x}(t)) = \begin{cases}
1 & \text{if } \hat{x}(t) \in \mathcal{R}_i \\
0 & \text{otherwise}
\end{cases}
$$

(24)

With $K_j$ are the controller gains.

Let use a new variable which present the estimation error. It is given by:

$$
e(t) = \hat{x}(t) - x(t)
$$

(25)

The closed loop fuzzy model of the wind system integrating the TS model (15), the fuzzy observer (23) and the controller (24) became:

$$
\begin{bmatrix}
\hat{x}(t) \\
x(t)
\end{bmatrix} = \begin{bmatrix}
\sum_{i=1}^{4} \mu_i(\hat{x}(t)) \left( \tilde{A}_{ij} \hat{x}(t) + \tilde{B}_i \hat{w}(t) \right) \\
0
\end{bmatrix}
$$

(26)

With:

$$
\tilde{A}_{ij} = \begin{bmatrix}
A_i + B_1 K_j & B_1 K_j \\
0 & A_i + L_i \tilde{C}
\end{bmatrix}; \tilde{B}_i = \begin{bmatrix}
B_{2i} \\
0
\end{bmatrix}; \tilde{C} = \begin{bmatrix}
C \\
0
\end{bmatrix}
$$

The general diagram of the fuzzy observer-based $H_\infty$ controller is given by Figure 6.

For the formulation of the main result, we recall the following theorem and definition.

**Definition:** given a scalar $\gamma > 0$, system (26) is stable with $H_\infty$ disturbance attenuation $\gamma$ if it is exponentially stable with:

$$
\sigma_1 \sigma_2 < 0
$$

(27)

$\sigma_1 \sigma_2$ is the desired level disturbance attenuation [6].

**Theorem:** the system (26) is stable with $H_\infty$ disturbance attenuation $\gamma$, if there are symmetrical matrices $\succ 0$, $\succ 0$, $\succ 0$, $\succ 0$, $\succ 0$, $\succ 0$ which satisfy the following linear matrix inequalities for $i,j \in \{1,2,3,4\}$ [6]:

$$
X(A_i + A_j)^T + (A_i + A_j)X + B_i (M_i + M_j) + (M_i + M_j)^T B_i^T + \gamma^2 B_{2i} B_{2i}^T < P_i
$$

(28)

$$
A_i^T Y + YA_i + C_i^T J_i^T + J_i C < Q_i, \quad i = 1,2, \ldots, m
$$

(29)

$$
\begin{bmatrix}
P_1 & \ldots & 0 \\
\vdots & \ddots & \vdots \\
P_m & \ldots & 0
\end{bmatrix}
\begin{bmatrix}
X C^T \\
X C^T \\
C X
\end{bmatrix} < 0
$$

(30)

$$
\begin{bmatrix}
P_1 & \ldots & 0 \\
\vdots & \ddots & \vdots \\
P_m & \ldots & 0
\end{bmatrix}
\begin{bmatrix}
X C^T \\
X C^T \\
C X
\end{bmatrix} < 0
$$

(31)
With the controller and observer gains are given by:

\[
\begin{align*}
K_i &= M_i \chi^T \\
L_i &= Y_i^T
\end{align*}
\]  

(32)

By applying the above theorem and the LMI optimization toolbox in matlab, to solve the equations 29-32, with four fuzzy rules, we obtain set feasible solutions. Hence, the controller parameters are:

\[
K_1 = 10^6 \begin{bmatrix} 0.0273 & 0.0000 & -0.0001 & -0.0000 \\ 2.1542 & 0.0042 & -0.0042 & 0.0000 \end{bmatrix}, \quad K_2 = 10^6 \begin{bmatrix} 0.0275 & 0.0000 & -0.0001 & 0.0000 \\ 2.1543 & 0.0042 & -0.0042 & 0.0000 \end{bmatrix}, \\
K_3 = 10^6 \begin{bmatrix} 0.1096 & 0.0002 & -0.0002 & 0.0000 \\ 2.1504 & 0.0042 & -0.0042 & 0.0000 \end{bmatrix}, \quad K_4 = 10^6 \begin{bmatrix} 0.1022 & 0.0002 & -0.0002 & 0 \\ 2.15043 & 0.0042 & -0.0042 & 0 \end{bmatrix}
\]

The observer gains are given by:

\[
L_1 = \begin{bmatrix} -0.6766 & -0.6546 \times 10^3 \\ -1.367 & 0.5244 \end{bmatrix}, \quad L_2 = \begin{bmatrix} -0.5749 & -0.5488 \times 10^3 \\ -1.1588 & 0.3478 \end{bmatrix}, \quad L_3 = \begin{bmatrix} -0.5442 & -0.52 \times 10^3 \\ -1.1062 & 0.475 \end{bmatrix}, \quad L_4 = \begin{bmatrix} -0.6043 & -0.5727 \times 10^3 \\ -1.2213 & 0.3947 \end{bmatrix}
\]

Figure 7 represents the variation curve of the generator speed. A stabilization of the generator speed around 30rd/s is ensured. The proposed control minimizes the generator speed ripple and maintains stability of the system during wind disturbances. Thus, this result demonstrates the performances of the command which is insensitive to the disturbance coming from the variation of the wind speed. The command seems a powerful technique for the disturbance rejection. In addition, the generator speed follows its reference with an error equal to 8%.

Curve 8 represents the variation of the state variables of the TS fuzzy system. Figure 8a represents the estimated and measure angle torsion of the driving shaft. Figure 8b represents the variation of the pitch angle of the blades. The measured and estimated values are confused justifying consequently the performances of fuzzy observer-based \(H_\infty\) controller.

4. Conclusion

In this paper, an approach of modeling and control of a nonlinear system based-wind generator is discussed. In the first part, a TS fuzzy model is used to approximate the dynamics of wind generator composed of a turbine, a driving shaft and an induction machine. In the second part, two command approaches are developed around the complex system in order to stabilize the mechanical model and to minimize the generator speed ripple. Numerical optimization problems using linear matrix inequality and convex techniques are used to design the controller and the observer parameters.

Stability results for closed model based fuzzy controller using the concept of PDC have been examined assuming that all of the system states are measurable. The generator speed
curve shows that the fuzzy controller can’t stabilize the response of system when the wind speed varies. The proposed command scheme using fuzzy observer-based $H_{\infty}$ controller illustrates some good performances. It minimizes the generator speed ripple, maintains stability of the system during wind disturbances and permits the attenuation of the external disturbances. The obtained results illustrate the effectiveness of this approach to stabilize the nonlinear system. Moreover, they showed the ability of the command to reject the disturbances and make possible to obtain a stable wind generator without oscillations. Therefore, we consider that the fuzzy observer-based $H_{\infty}$ controller is appropriate for the nonlinear system control with external disturbances.

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