Motor Speed Maximum Control in the Resonance Ratio Controller for Two-Mass System using Self-Organizing Fuzzy Controller

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Abstract: According to the importance of controlling the maximum speed and overshoot in electric motors, and also according to the fact that at the start of a two-mass system with a resonance ratio controller, we will see severe fluctuations in the speed. The control the motor maximum speed in speed desired value is aim in this paper. To take action to achieve this goal, we will use the design incorporating a self-organizing fuzzy control system with a coefficient of resonant two-mass system. Self-organizing fuzzy controller is a combination of fuzzy control and adaptive, which along with the combination of self-organizing fuzzy control methods, can create instructional with the ability to modify the rules in the knowledge base and incorporated into it, two-mass system with a resonance ratio control system, the maximum engine speed can be adjusted to any desired value.

Keywords: two-mass system; self-organizing fuzzy control; resonance ratio control.

1. INTRODUCTION

With industry development and control technique advanced, high performance and high efficiency realize in dynamic control scale [1,2]. A mechanical system can model to one multi-mass system in industry motor drive [3,4]. In many states, two-mass system purpose for first resonance mode [5,6]. In general movement control applications, system consist of a motor, a gear wheel, transfer implements and also a load [7,8]. One of the problems in two-mass systems control is convenient and automatic control of the maximum motor speed. The two-mass resonance system is shown in Fig. 1 [9,10]. On the other hand, some important factors in controlling motor optimal, control the amount of overshoot, reduce the ripple amplitude, motor oscillations and reaching a steady state in the shortest time possible [11].

Fig1. Two-mass resonance system

Many papers have been published in the field of two-mass system [12,13]. An application of a fuzzy controller strengthened with a gray estimator to a nonlinear two-mass system control is presented in [14], which fuzzy controller is designed to regulate the speed of the system. A nonlinear approach for variable-speed wind turbine control using a two-mass model and a wind speed estimator is presented in [15], which the model of the two-mass system is motivated by the need to deal with flexible modes induced by the low-speed shaft stiffness. In order to have sufficient damping, in [16] three different controller base on integral–proportional-derivative controller are designed for the speed control of a two-mass system using a normalized model and polynomial method. A quasi-time-optimal, nonlinear state-feedback control for the torsional torque of a two-mass system is proposed in [17], which controller design is based on results of the general framework of optimal control and leverages the assumption of a bang-bang structure for the actuation. The stability analysis of small-signal models of two-mass and
three-mass wind turbine interconnected to thermal power system is presented in [18], which the use of multi-mass model is motivated by the need to deal with flexible modes induced due to intermittent varying of wind speed and low speed shaft stiffness. A two degree of freedom mass-spring-damper model on a moving lubricated belt is considered in [19], which the friction model includes the boundary, mixed and hydrodynamic regimes of lubrication contact regimes.

In this paper, the design of such a control system will expand under the control of the maximum speed of the motor as one of the two objects two-mass system, and that this functionality to the system, that the user only has to provide the maximum speed of the system in terms of percent of optimum speed to the system, to achieve its goal of controlling the electric motor.

2. TWO-MASS SYSTEM MODEL

Fig. 2 shows two-mass resonance system block diagram. As was seen in the figure, shaft coupling (TS) consist of sum of two signals which one of them is proportional with difference motor speed (oM) and load speed (oL) and other one is proportional with difference motor torque angle (θM) and load torque angle (θL). According to system damping negligibility, shaft torque supposes motor angle difference and load proportional. Choosing three state variables, oL, oM and TS with two input variables, motor torque (TM) and load disturbance torque (TL), system state equations are [20]:

\[
\frac{d}{dt} \omega_M = -\frac{B_M}{J_M} \omega_M - \frac{1}{J_M} T_S + \frac{1}{J_M} T_M
\]

\[
\frac{d}{dt} T_S = (K_S - \frac{B_M B_S}{J_M}) \omega_M - (K_S - \frac{B_L B_S}{J_L}) \omega_L - B_S (-\frac{1}{J_M} + \frac{1}{J_L}) T_S + \frac{B_S}{J_M} T_M + \frac{B_L}{J_L} T_L
\]

\[
\frac{d}{dt} \omega_L = -\frac{B_L}{J_L} \omega_L + \frac{1}{J_L} T_S - \frac{1}{J_L} T_L
\]

\[
\frac{d}{dt} \theta_M = \omega_M
\]

\[
\frac{d}{dt} \theta_L = \omega_L
\]

where J_M and J_L are the motor inertia and the load inertia, B_M and B_L are the motor viscous damping coefficient and the load viscous damping coefficient, and K_S is the shaft stiffness. The damping ratio \( \eta \) and resonant frequency \( \omega_R \) are given by [21]:

\[
\omega_R = \sqrt{\frac{K_S}{J_L} (1 + K_J)}
\]

\[
\eta = \frac{B_S}{2 \sqrt{K_S J_L}} (1 + K_J)
\]

where \( K_J = J_L/J_M \) is inertia ratio. The anti-resonant frequency is:

\[
\omega_A = \sqrt{\frac{K_S}{J_L}}
\]

If \( K_J \) and \( B_S \) are small, mechanical vibration occurs easily. Increasing the inertia ratio \( K_J = J_L/J_M \) and shaft stiffness will decrease the mechanical vibration of the system. Apparently, with the increase of \( K_S \) and \( J_L \), damping ratio goes down. Good response characteristics depend on both damping ratio and undamping nature frequency. One important factor has been noticed, in that the resonance characteristics of a two-mass system can be described by its resonance ratio \( (K_R) \) which is the quotient of the anti-resonance frequency and resonance frequency of the system:
3. TWO-MASS RESONANCE RATIO CONTROL

Resonance ratio control is one of the effective control methods for two-mass resonance system determined where ratio between motor resonance frequency and robotic arm base on measured reaction feedback [22,23]. In two-mass resonance systems couple and motor speed are measurable only and rest of variables must estimate by observers [24,25]. Fig. 3 show the block diagram of the two-mass resonance system controller. It consists two-system, disturbance observer and resonance ratio controller. Fig. 4 show the block diagram of transfer function controller of two-mass resonance controller. In the resonant ratio controller, disturbance observer is much faster than the resonance frequency and anti-resonance, so used the factor 1-K in feedback estimated disorder.

4. SELF-ORGANIZING FUZZY CONTROLLER

The overall performance of the self-organizing fuzzy controller based on the Fig. 5, so that in the first phase of the phase of the corresponding entry [26,27]. And then modify the existing rules in the
knowledge base will help adaptive control method and base on the law selected has been modified, the calculated desired output and after passing through a unit De-fuzzification, is non-fuzzy and is applied to the system. First, with the help of table I, which is experimentally obtained, 49 of the initial values to use self-organizing fuzzy control system is available in the knowledge base, accordingly, when entering the desired speed and speed measurement error into the system, so of fuzzy logic inference based on respective one of 49 of the existing knowledge base, is selected. Non-fuzzy coefficient of resonant controller and then to apply two-mass system [28, 29].

Then, with the help of self-organizing adaptive control methods, fuzzy decision table values are modified. Self-organizing part, as follows to change the rules in the knowledge base acts:

\[ E = \frac{1}{2} \| y(t) - y_{sp} \|^2 \]  

(10)

\[ c_i(k+1) = c_i(k) - \alpha \frac{\partial E}{\partial c_i} \]  

(11)

where \( C_i(k) \) is the membership function of law L in the course of training k, \( C_i(k+1) \) is the membership function of law L in the course of training, \( \alpha \) is learning rate and \( \partial E/\partial c_i \) is derivative function when compared with \( C_i \).

\[ \frac{\partial E}{\partial c_i} = \sum_{l=1}^{n} \left( \mu_l(e) \times \mu_l(ce) \times (y(t) - y_{sp}) \right) \]  

(12)

where \( \mu_l(e) \) is join the error of the law L, \( \mu_l(ce) \) is join the error changes the law L, \( Y_{sp} \) is value period is required and \( y(t) \) is the value of the measured period.

| Table 1. KNOWLEDGE BASE AND RULES AVAILABLE AT THE TIME NT IN TRAINING K |
|-----------------|-----|-----|-----|-----|-----|-----|-----|-----|
| e               | de  | NB  | NM  | NS  | ZE  | PS  | PM  | PB  |
| center          | -0.6| -0.4| -0.2| 0   | 0.2 | 0.2 | 0.4 | 0.6 |
| NB              | -0.6| -0.6| -0.6| -0.6| -0.6| -0.4| -0.2| 0   |
| NM              | -0.4| -0.6| -0.6| -0.6| -0.4| -0.2| 0   | 0.2 |
| NS              | -0.2| -0.6| -0.6| -0.4| -0.2| 0   | 0.2 | 0.4 |
| ZE              | 0   | -0.6| -0.4| -0.2| 0   | 0.2 | 0.4 | 0.6 |
| PS              | 0.2 | -0.4| -0.2| 0   | 0.2 | 0.2 | 0.4 | 0.6 |
| PM              | 0.4 | -0.2| 0   | 0.2 | 0.4 | 0.6 | 0.6 | 0.6 |
| PB              | 0.6 | 0   | 0.2 | 0.4 | 0.6 | 0.6 | 0.6 | 0.6 |
5. SIMULATION RESULTS

In this section the simulation results of the system is shown. Items listed in the table II, is considered as a common factor in all the simulations. In this part the initial conditions include the follows: \( J_M = 0.3 \), \( J_L = 1.5 \), \( T_L = 0.1 \), \( K = 6 \). Curves for maximum desired value by 4%, 7% and 9% to curve the system in the non-use of self-organizing fuzzy controller are show in Fig. 6. Fig. 7 to better illustrate the distinction between curves under 4%, 7%, 9% is shown. As seen, in the case of a self-organizing system is no fuzzy controller, fluctuations in engine speed is observed. We also see the maximum speed with 1073 the amount of non-use of the controller. But after using fuzzy self-organizing systems by setting the maximum speed on the optimal value, the excess amount of base speed can be expressed in terms of percentage (4% = 1040 and 1090=1070=7% and 9%), does not exceed the maximum speed in the cycle of the set and also the volatility has decreased to a considerable extent and the cycle is under control. Even when set to the maximum rate of 9%, we will see that. Although in the case of self-organizing fuzzy controller, we have a maximum speed, but the advantage of this controller to reduce engine speed fluctuations, will show itself well. Figs. 8 and 9 was shown the effect of the load torque to the goal control system and we can see that reached to maintain maximum speed at the desired level and also decreased levels of volatility through the use of the controller. Due to closer look at system performance in terms of torque load in different conditions control system fuzzy self-organizing, in Figs. 10 and 11, have shown that the torque load with a value of 50 N.m since the 25s, the system as it is expected to continue its activities and the addition of load torque will have little impact on the maintenance of the maximum speed at the desired level and the volatility also does not have much impact and to the use of this system greatly reduced the volatility.

**TABLE2. COMMON FACTORS IN SIMULATION**

| Factor | Value  |
|--------|--------|
| \( \omega \) | 1000 rpm |
| \( g_1 \) | 4.7 |
| \( K \) | 6 |
| \( K_p \) | 2.8 |
| \( K_i \) | 4.8 |

![Fig2. Curves for maximum desired value by 4%, 7%, 9% to curve the system in the non-use of self-organizing fuzzy controller](image)

**Fig2. Curves for maximum desired value by 4%, 7%, 9% to curve the system in the non-use of self-organizing fuzzy controller**

![Fig3. Zoom of the figure 6](image)

**Fig3. Zoom of the figure 6**
Motor Speed Maximum Control in the Resonance Ratio Controller for Two-Mass System using Self-Organizing Fuzzy Controller

**Fig4.** The curve compared to the maximum amount of 5% and 7% toward curve the system in the non-use of self-organizing fuzzy controller with torque load 50 Nm

**Fig5.** The curve compared to the maximum amount of 5% and 7% toward curve the system in the non-use of self-organizing fuzzy controller with torque load 50 Nm

**Fig6.** The curve compared to the maximum amount of 4% and 9% toward curve the system in the non-use of self-organizing fuzzy controller with torque load 50 Nm with a step function at the time of 25 seconds

**Fig7.** The curve compared to the maximum amount of 5% and 7% toward curve the system in the non-use of self-organizing fuzzy controller with torque load 50 Nm with a step function at the time of 25 seconds
6. CONCLUSION

As has been shown in experimental results simulated, with fuzzy self-organizing systems can be controlled the maximum value motor speed in the controller coefficient of resonant two-mass system. By providing desired maximum value as input system, observed that in the whole process of the maximum motor speed does not exceed the amount set and also minimizes the oscillations in the system and reduced the extent of it. Because of this feature, in order to control the motor speed of the system, it can be used in various processes that maximize the speed and motor oscillations and stable operation of the system is important.

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Motor Speed Maximum Control in the Resonance Ratio Controller for Two-Mass System using Self-Organizing Fuzzy Controller

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