Adaptive Neuro Fuzzy Observer Estimator for Servo Motor

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Abstract: The servo engine is really an assembly of 4 things: a gear reduction product, a typical DC motor, a position sensing unit along with a management circuit. The DC engine is actually associated with a gear mechanism that delivers comment to a position sensor that is mainly a potentiometer. The point of this research is to design a NFSC which could be applied to a DC servo motor based explanatory antenna locating system. To accomplish this, initial, a Fuzzy Logic Controller (FLC) was designed utilizing master information. Next, NFSC was designed dependent on the FLC algorithms by utilizing Neural Network (NN) figuring out how to tune the Fuzzy Logic (FL) rule base through hybrid training method.

Keywords: ANFIS, Neural Network, Servo Motor

I. INTRODUCTION

A servo engine is an electric unit which can drive or perhaps spin an item with great accuracy. In case you would like to spin as well as object at several exact angles or maybe distance, then you make use of servo motor. It's simply made up of straightforward motor which run by servo mechanism. Servo suggests a mistake sensing feedback management which is used to fix the overall performance of a product. Additionally, it takes a typically advanced controller, frequently a separate module created especially for using with servomotors. Servo motors are actually DC motors that permit for accurate control of angular place. They're in fact DC motors whose pace is gradually lowered by the gears. The servo motors ordinarily have a revolution cutoff from 90° to 180°. A number of servo motors have revolution cutoff of 360° or over. But servo motors don't rotate continuously. The rotation of theirs is restricted in between the fixed angles.

The servo engine is actually a get together of 4 things: an ordinary DC engine, a rigging decrease item, a position detecting unit alongside an administration circuit. The DC engine is really connected with an apparatus system that gives input to a position sensor that is principally a potentiometer. By the gearbox, the yield of the engine is really exhibited through a servo spline to the servo arm. For normal servo motors, the rigging is for the most part included clear plastic while, for expanded vitality servos, the apparatus is really comprised of metallic.

A servo comprises of a Motor (AC or DC), a potentiometer, gear gathering, and a controlling circuit. As a matter of first importance, we use gear get together so as to decrease RPM and furthermore to raise torque of the engine. State from the outset position of servo engine shaft, the job of the potentiometer handle is quite that there's no power signal delivered at the yield port of the potentiometer.

Today an electric signal is actually given to another input terminal of the error detector amplifier. However distinction between these 2 signals, one comes from potentiometer and another will come from some other resource, will be prepared in feedback mechanism plus paper is going to be offered in term of errors signal. This particular mistake signal functions when the input for engine as well as motor starts rotating. At the present time engine shaft is related with a potentiometer and as engine swivels so the potentiometer and this will produce a sign. Thus as the potentiometer's precise position changes, its yield criticism signal adjustments. Following some time the job of potentiometer arrives at a spot that the yield of potentiometer is very same as outside sign advertised.

Around this problem, there is going to be no output signal from the amplifier to the engine input as there's no distinction between outside applied signal as well as the signal produced for potentiometer, what about the circumstance motor stops rotating.

As the name prescribes, a servomotor is a servomechanism. Even more especially, it is a servomechanism that utilizes position contribution to control its development and last position. The data to its control is some sign, either fundamental or moved, addressing the position facilitated for the yield shaft. The engine is made out of some kind of encoder to give position and speed input. At all bothersome case, only the position is evaluated. The intentional circumstance of the yield is stood apart from the charge position, the outer information to the controller. In case the output position differs from that required, a bungle sign is made which at that point makes the Motor pivot in either bearing, true to form to pass on the output shaft to the fitting position. As the positions approach, the error signal reduces to zero and the Motor stops. The uncommonly clearest servomotors use position-simply recognizing by methods for a potentiometer and shoot control of their Motor. The Motor constantly pivots at max throttle .This kind of servomotor isn't commonly used as a piece of mechanical development control, but instead it outlines the reason of the direct and unobtrusive servos used for radio-controlled models. Increasingly present day servomotors measure both the position and the speed of the pole. They may in like manner control the pace of their Motor, instead of ceaselessly running at max throttle.
Both of these upgrades, generally in blend in with a PID control calculation, license the servomotor to be passed on to its coordinated position even more quickly and even more unequivocally, with less overshooting.

II. METHODOLOGY
The one of a kind methodology of development of the fuzzy control rules dependent on the master understanding and in the wake of studying the different reactions got from the classical PI controller and fuzzy-PI controller has been exhibited. The methodology of this paper to design NFSC Controller for D.C. servomotor-based antenna pointing system has been effectively accomplished. This was checked through reenacted output reactions of the system to step input signal which fulfilled the design criteria. Likewise, the trial results got from the prototype demonstrated that the control system created and its algorithms worked and could be sent in driving the azimuth and height DC servomotors in order to direct an elegiac dish antenna and keep it generally inside the ideal viewable pathway with a specific satellite. The FLC without anyone else's input expanded the settling time to 2.4 seconds yet made some asces mnemonic memories of 2.0 seconds and diminished the overshoot to zero (0) %. The Takagi-Sugeno FLC recorded a superior asces memory time of 1.6 seconds and settling time of 2.1 seconds contrasted with the Mamdani FLC model. The seat stamping PID controller enlisted an asces time of 0.7 seconds, settling time of 4.3 seconds and an overshoot of 21 %. The ANFIS approach gave the best performance an asces time of 0.2 seconds, settling time of 0.8 seconds and an overshoot of 8.0 % hence meeting every one of the details of the tracking system. In this manner, the objectives were accomplished in totality for the design, software reproductions just as tests directed.

Neuro-Fuzzy System Controller Design
The ANFIS controller produces change in the reference drive voltage based on position error E and derivative in the position error (speed error) DE as will be defined in following equation. In this examination, first request Sugeno-type fuzzy inference will be utilized for ANFIS and the typical fuzzy principle takes the structure in equation (3.44). On the off chance that E is A, and DE is B, at that point
\[ z = f(E, DE) \]

Where A, and B are fuzzy sets in the antecedent and \( z = F(E, DE) \) is a hard function in the subsequent. The hugeness of each layer and operation of the 2-input-1-output ANFIS structure considered:

**Layer 1:** This layer (the fuzzification layer) empowers the passage of crude information or fresh inputs from the objective system into ANFIS. It is made out of various figuring nodes whose enactment functions are fuzzy logic membership functions, taken as triangular in this postulation. Each adaptive node creates the membership grades called fuzzy spaces for the input vectors A; \( I = 1, \ldots, n \) and B; \( i = 1, \ldots, n \), and where n is the quantity of membership functions of the inputs (E and DE) picked as n = 7. The degree to which the inputs exist in the fuzzy space is given a worth standardized among and the output is characterized by:

\[ 0^1_{A1} = \mu_{A1}(E), \quad 0^1_{B1} = \mu_{B1}(DE), i = 1, \ldots, n \] (2)

**Layer 2:** Is the standard layer where every node is fixed. When the areas of inputs in the fuzzy spaces are distinguished, the result of the degrees to which the inputs fulfill the membership functions is found. This item is known as the -ring strength of a standard whose output is given by (3). At the end of the day, it chooses the minimum (min) estimation of the inputs. In this layer, the absolute number of Takagi-Sugeno rules is 49.

\[ 0^2_i = W_i = \min (\mu_{A1}(E), \mu_{B1}(DE)), \quad (3) \]

**Layer 3:** In layer 3, the standardization layer, the ratio of each standard's terminating quality is determined regarding the total of the terminating qualities of the considerable number of rules. Each part in this layer is fixed. The \( i^{th} \) node output is the \( i^{th} \) input initiation level separated by the total of all the actuation levels of different inputs (4),

\[ 0^3_i = \overline{W_i} = \frac{W_i}{\sum_{j=1}^{n}W_j} \] (4)

**Layer 4:** In layer 4, the defuzzification layer, the output of every node is the weighted ensuing worth. Adaptive node i in this layer figure the commitment of rule towards the general output, with the accompanying node function (5),

\[ 0^4_i = \overline{W_i}Z_i = \overline{W_i}(p_iE + q_iDE + r_i) \] (5)

**Layer 5:** Layer 5 is the summation layer and its output, which is the total of the considerable number of outputs of layer 4, gives the general output for the particular inputs inside the fuzzy space. The single fixed node in this layer processes the general output as the aggregate of each standard's commitment (6),

\[ 0^5_i = \sum_{i=1}^{2} \overline{W_i}Z_i = \frac{W_1Z_1 + W_2Z_2}{W_1 + W_2} \] (6)

Before the ANFIS system can be utilized for expectation, the parameters of the rules are controlled by first creating an underlying FIS where first arbitrary values are doled out to the parameters. Next, an enhancement scheme is applied to decide the best values of the parameters that would supply rules to optimistically model the objective system. Subsequent to training, the rules remain with the goal that when new input information is displayed to the model, the rules give a relating sensible output. The streamlining strategy utilized is a hybrid learning algorithm that limits the error between the ANFIS model and the genuine system utilizing training information from the objective system to produce signals that proliferate in reverse and advances and update the parameters. The parameters to be prepared are \( A_i \) and Bi and of the reason parameters and pi, qi and ri, of the subsequent parameters.
III. RESULTS

ANFIS framework, has been tested by both simulation as well as experiments. These had been carried out in MATLAB SIMULINK, Proteus Professional Prototyping Software along with micro controller locations. The NFSC along with the PID controller was individually put on to both an manufacturing DC servomotor satellite tracking plant as well as an experimental prototype as the FLC was tested on the manufacturing type just. In total instances, step input signal was selected as the reference signal because it immediately represented changes of the position of the motors (load) and also provided simple mapping with the potentiometers. The results obtained by using FLC and NFSC had been in contrast to those of the PID for each situation. This particular comparison was made in order to assess overall performance of the NFSC and FLC (in presence of saturation non linearity) and then to verify the precision of the design.

In the improvement of the NFSC, two training systems for example back spread algorithm (which looks like ANN) and the hybrid enhancement procedure (which is normal for ANFIS) were accessible for deployment additionally, there are a few kinds of information enrollment functions that could be utilized and furthermore whether the output participation functions took linear or steady structures. Thusly, it was important to check the appropriateness of every advancement system in limiting the error between the ideal position and the real position of the DC servomotor preceding settling on one of them.

Table 1. ANFIS Training and MFs Selection

| S/No. | Input Membership Functions (MFs) | Output Membership Functions (MFs) | Training Error Based on Optimization Method |
|-------|----------------------------------|-----------------------------------|---------------------------------------------|
|       | Number | Type        | Number | Type | Back Propagation | Hybrid |
| 1     | 3,3    | Triangular | 25     | Constant | 0.135 | 0.165 |
| 2     | 5,5    | Triangular | 25     | Linear  | 0.020 | 0.250 |
| 3     | 5,5    | Trapezoidal| 25     | Constant| 0.220 | 0.187 |
| 4     | 5,5    | Trapezoidal| 25     | Linear  | 0.250 | 0.032 |
| 5     | 5,5    | Gaussian   | 25     | Constant| 0.187 | 0.140 |
| 6     | 5,5    | Gaussian   | 25     | Linear  | 0.250 | 0.002 |
| 7     | 7,7    | Triangular | 49     | Constant| 0.185 | 0.025 |
| 8     | 7,7    | Triangular | 49     | Linear  | 0.275 | 0.00015 |
| 9     | 7,7    | Trapezoidal| 49     | Constant| 0.127 | 0.120 |
| 10    | 7,7    | Trapezoidal| 49     | Linear  | 0.250 | 0.00038 |
| 11    | 7,7    | Gaussian   | 49     | Constant| 0.450 | 0.100 |
| 12    | 7,7    | Gaussian   | 49     | Linear  | 0.275 | 0.0006 |

Table 1. It was noticed that the ANFIS model deploying 7×7 input triangular MFs, 49 linear output MFs and prepared by the hybrid algorithm enrolled the most lowest mean squared error of just 1.5×10⁻⁴ and hence offered the best powerful and stabilized performance.

Surface plots and comparing contour plots indicating relationship between the two inputs parameters for example position error (input 2) and change in position error (input 1) and the ANFIS controller output (input signal to motor drive) before training and in the wake of training are given in Figure 3 & 4.
IV. CONCLUSION

The ANFI versions of centrifugal chiller grow are designed by utilizing exactly the same input output measurements information used to create the autoregressive moving average versions, as well as the proportional essential derivative balance program structures based on these designs are actually smart hybrid mixture of an adaptive neuro fuzzy buildings and proportional integral derivative controllers. As the arrangement transducers generally utilized in industry don't innately survey a quick speed, signal preparing is normally expected to improve the precision of speed estimation at each examining moment. This specific assessed signal is then used as the speed criticism for the speed circle impact in servo engine drives.

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