

# Passive Dancer Structure for Optical Gyroscope Winding Process and Tension Control Applying Fuzzy Learning Algorithm

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**Abstract**— Winding process of optical fiber gyroscopes greatly affects its quality. Due to its delicate characteristics, (order of 100 micrometers) tension forces applied to fiber during winding process should be kept within very low limits. Any change beyond the tension forces limit of the fiber optic could cause deformation or destruction of the fiber. This paper proposes tension sensor and control structure to achieve high precision of fiber tension during winding process. The proposed structure relaxes control objective and hence, allows the use of slow dynamics high torque motors (required for driving large inertia supply spool), yet achieve high transient performance. In addition, Fuzzy Model Reference Learning controller is used to compensate for time changing parameters of the system during the fiber winding process. In this paper, the proposed tension sensing and control mechanism is introduced; its mathematical model is developed. A Fuzzy Model Reference Learning control algorithm is implemented in the tension control loop. The results of the proposed control algorithm is compared to conventional PID controller at different stages of winding process. The results validate the proposed mechanism and prove the ability of the control algorithm to adapt to the changing parameters of the process during the course of winding process.

**Keywords**— *Modeling; Tension control; Optical Gyroscope; Fiber coil winding; Fuzzy Learning Control; Model Reference*

I.

## INTRODUCTION

Optical gyroscopes are among instruments that is used to measure the rotational motion [1, 2]. They have fast reaction time due to their structure that have no moving part. In addition, they have wide dynamic range, no "g" sensitivity drift and high accuracy and reliability. Optical gyroscopes operate based on Sagnac effect principle [3]. The most common used optical gyroscopes are Ring laser gyroscopes (RLGs) and fiber-optic gyroscopes (FOGs) [4]. While RLGs require expensive mirrors and other mechanical components, FOGs represent a cost effective gyroscope.

However, to achieve high accuracy, it is required to develop optical coils with lengthy optical fiber (one or

two kilometers) in special pattern windings. FOGs deploy Sagnac effect by injecting two in phase laser beams from the two ends of fiber coil. Hence, the two laser beams will travel inside optical coil in a direction against each other, one of them in the direction of the coil rotational motion and hence its speed increases while the opposite one will travel with slower speed. The emerging beams will have phase shift relatively proportional to the rotational speed of the coil. To increase its accuracy, for detecting very slow angular motion, optical gyros require lengthy fiber winding coil. Length of the fiber winding is proportional to the produced coil weight which negatively affects the FOGs applications. Hence, a tiny optical fiber (order of 100 micrometer diameter) is used for FOGs coils to keep their weight in a reasonable range. With complex winding process and fragile fiber, a possible deformation of the fiber cross section could occur.

Fiber tension control is an essential process in the optical fiber winding system. The winding tension directly influences the quality of wound coil products that makes the winding tension control a key feature in produced optical gyroscopes. Variable tension during winding process could result in deformation of the fiber cross section in different parts, hence refractive index of the fiber will change resulting in low quality optical gyroscope coils. To avoid fiber deformation during winding process, the fiber tension should be maintained within very low limits defined by fiber characteristics. Therefore, commonly used tension control systems in industrial winding applications cannot satisfy the special characteristic of the optical fiber used for gyroscopes. The speed of the supply spool has significant influence on the fiber tension. The control of such speed to achieve the desired tension is challenging control problem due to slow dynamics of supply spool system (high inertia of the fiber windings and spool).

One of the proposed control techniques for such fiber winding process was to have multiple control loops with single control objective (fiber tension). The main control loop is achieved by controlling the speed of the supply spool while the auxiliary control loop is added by what is called "dancer" mechanism [5]. While the driving system of the supply spool has high inertia and able to rotate the supply spool, the dancer mechanism is driven by low inertia motor. Hence, the dancer mechanism has very fast transient response. Accordingly, the tension control performance enhances.

However, the presence of two control loops with one control objective complicate the controller design and reduce the system stability. The challenge is evident when a pressure transducer is used to measure the tension forces. At this case, small variation of the speed between the two driving motors (the supply spool and the product spool motors) would result in abrupt variation on fiber tension. With the presence of large time constant of the supply spool driving motor, the required fast response cannot be achieved [6]. Moreover, there is still inertia for the dancer motor.

In addition to the complicated control loops of fiber winding process, the moment of inertia for both the supply and product mechanism are changing during winding process. This is because the optical fiber transfers from the supply spool to the product spool. Keeping in mind the length of the optical fiber used for optical gyroscopes (2 km length), the changing of inertia for both supply and product mechanism is significant and should be taken into consideration.

Fuzzy Logic controllers (FLC) [7] have a great advantage over conventional control techniques for the following reasons: 1) They can present nonlinear mapping surface by membership functions and inference mechanism inherited in their structure; 2) They are independent on mathematical modeling and require only behavioral performance to set out linguistic variables and rule basis; and 3) They are better to present system uncertainties.

FLC with adaptive capabilities was developed by Procyk and Mamdani [8]. The parameters of the fuzzy controller are subjected to adaptation during system operation to meet a predefined performance under plant parameter variation. Fuzzy model reference learning control (FMRLC) [9, 10] uses reference models to describe the required performance to the adaptation mechanism. The term learning reflect the memorizing capability of the fuzzy controller where enhancement in the system performance can be experienced by frequent exposure of the system to the same dynamic range of state variables. The algorithm was successfully implemented to control the speed of induction motor drives [11, 12].

This paper introduces a novel tension transducer structure that relaxes the control objective yet provide fast transient performance even in presence of large inertia driving motors for both supply and product spools. The proposed transducer replaces the dancer mechanism, introduced in [5], such that the control objective successfully achieved by one control loop. In addition, Fuzzy Model Reference Learning Control (FMRLC) is used to adapt for the variation of moment of inertia of both supply and product driving mechanism.

The paper is organized as follows: The second section introduces the structure of the proposed optical fiber winding tension control. The third section presents the kinematic model, then, section three provides the dynamic model of the driving system, sensor model, and tension control system of fiber tension winding system. Section four discusses the

FMRLC algorithm. Section five introduces the simulation results. Finally, conclusion and future work are presented in section six.

## II. STRUCTURE OF TENSION CONTROL AND FIBER WINDING SYSTEM

The proposed structure composed of three mechanisms, (1) Product spool, (2) Supply spool and (3) tension transducer as shown in Figure 1). Both of supply and product spools are driven by permanent magnet brushed DC motor (PMBDCM) which is commonly used in low power low voltage applications. The three mechanisms are supported at an aluminum platform. The tension transducer is mounted at the center of the tension mechanism between the two spools. The tension transducer is a rotational potentiometer with its base mounted on the aluminum platform and an aluminum arm is mounted to the rotating potentiometer shaft. Two free rotating pulleys are mounted in the far edges of the arm. The shaft of the potentiometer is linked to the platform by calibrated helical spring.

The optical fiber is wound around two free rotating pulleys at the edges of transducer arm. The helical spring applies forces to the stretched fiber between the two pulleys. The applied forces are proportional to the angle of the potentiometer between horizontal line and rotating arm.

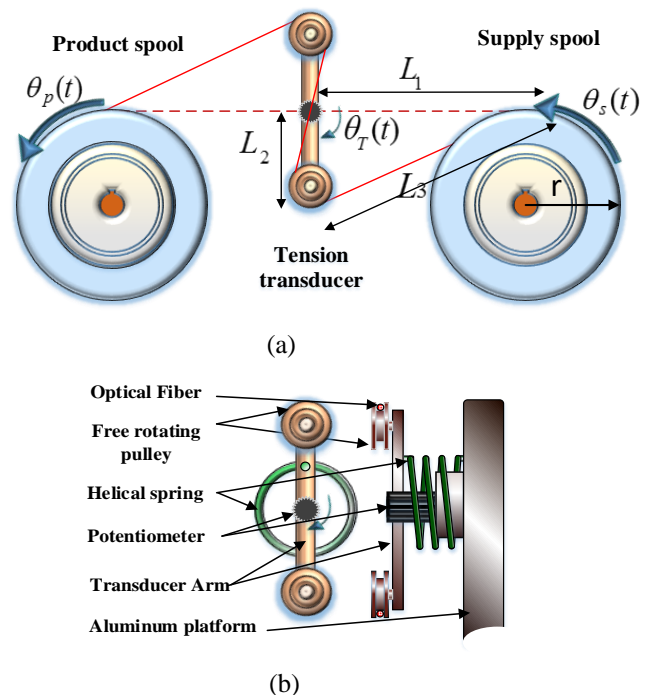


Figure 1 Schematic diagram of (a) tension sensing and control system and (b) tension transducer Structure

## III. KINEMATIC MODELING:

This section discusses the relation between relative angular speed of the supply and product spools and the angular deviation of the transducer arm from its horizontal position ( $\theta_T(t)$ ). The transducer angle ( $\theta_T(t)$ ) is linearly proportional to the tension produced by the helical spring that links system platform to transducer arm. The cross relation between system

geometrical quantities can be driven from Figure 1-a. At any instance  $t$ ,  $\theta_p(t)$  and  $\theta_s(t)$  are the product and supply spools angular position respectively,  $L_1$  is the stretched half-length of optical fiber,  $L_2$  is the half length of transducer arm,  $L_3$  is stretched fiber between supply spool and the free rotating pulley at the edge of transducer and  $r$  is the radius of the product and supply spools. The transducer deviation angle is proportional to fiber length between the supply spool and the product spool. To avoid complex modeling, the following assumptions are made:

- 1- The elongation due to tension forces aligned to optical fiber longitudinal axis is neglected. The relative elongation of the fiber due to applied tension forces is given by [13]:

$$\frac{\Delta L}{L} = \frac{F}{E_G A} \quad (1)$$

where  $E_G$  is the proportionality constant between the perturbation force per area and the relative deformation (Young modulus). The relative elongation is  $0.00075 \mu\epsilon$  for 50 g tension forces which is very small and can be neglected.

- 2- The curvature of the optical fiber around supply and product spools and free rotating pulleys is neglected. This is valid if  $L_1 \gg r$  and the radius of the free rotating pulleys is much smaller than the half length of the transducer arm ( $L_2$ ).

The tension applied to the optical fiber is proportional to the integration of relative rotational speed ( $\Delta\dot{\theta}$ ) between product and supply spools. The angle of tension transducer  $\theta_T$  is proportional to the angular shift between supply spool and product spool  $\Delta\theta$ . The proportional constant between them is dependent on the dimensions of fiber winding mechanism. The length  $l$  of tangled optical fiber between the two pulleys (supply and product) is given by:

$$l = \Delta X + 2L_1 = 2(L_3 + L_2) \quad (2)$$

where  $\Delta X$  is the linear displacement shift between supply spool and product spool. This displacement is proportional to the angular shift between supply motor and product motor  $\Delta\theta$  as follow:

$$\Delta X = \Delta\theta \times r + \frac{\Delta r}{2} \quad (3)$$

where  $\frac{\Delta r}{2}$  is the average thickness of the fiber around the spool. and  $\Delta\theta$  can be achieved by integrating the speed difference between the product and supply spools as follow:

$$\Delta\theta = \int_0^t (\dot{\theta}_2 - \dot{\theta}_1) dt \quad (4)$$

From (2) and (3) we get

$$2(L_3 + L_2 - L_1) = \Delta\theta \times r + \frac{\Delta r}{2} \quad (5)$$

Hence,  $L_3$  can be calculated using known dimensions of system  $L_1$ ,  $L_2$ ,  $\Delta r$  and  $r$

$$L_3 = \frac{1}{2} \left( \Delta\theta \times r + \frac{\Delta r}{2} - (L_2 - L_1) \right) \quad (6)$$

Finally, the transducer angle  $\theta_T$  can be calculated as follow:

$$\theta_T = \cos^{-1} \frac{L_3^2 - (L_1 + L_2)^2}{-2L_1 L_2} \quad (7)$$

#### IV. WINDING SYSTEM DYNAMIC MODEL

The model of optical fiber coil winding was developed for stress and control analysis in [4, 14]. In this paper, a modified structure is proposed. The aim of this section is to provide a modified model of proposed winding system for control analysis.

The total damping (friction) of the mechanical system at product driving side  $D_{Tp}$  is given by:

$$D_{Tp} = D_m + D_{ps} \quad (8)$$

where  $D_m$ ,  $D_{ps}$  and  $D_{ss}$  are damping of motor, product spool and supply spool respectively.

Similarly, the total damping (friction) at supply side driving  $D_{Ts}$  is given by:

$$D_{Ts} = D_m + D_{ss} \quad (9)$$

Total moment of inertia of the rotor at product driving side  $J_{Tp}$  is given by:

$$J_{Tp} = J_m + J_{ps} + J_{pf}(t) \quad (10)$$

where  $J_m$ ,  $J_{ps}$ ,  $J_{ss}$ ,  $J_{sf}(t)$  and  $J_{pf}(t)$  are moment of inertia of the motor, product spool, supply spool, supply fiber and the product fiber respectively. The moment of inertia of the wound optical fiber decreases with time as the optical fiber leave supply spool and wound around product spool. Similarly, the total inertia at supply driving side  $J_{Ts}$  is given by:

$$J_{Ts} = J_m + J_{ss} + J_{sf}(t) \quad (11)$$

Figure (2) shows a simplified schematic of optical fiber arrangement around supply and product spool. The number of the layers and rounds is dependent on length of the fiber ( $L$ ), its diameter ( $d$ ), width of the spool ( $D$ ) and its radius ( $r$ ). If  $\rho$  is the fiber density then, the mass per unit length is given by:

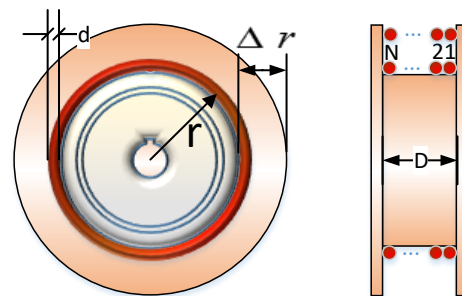


Figure 2 Simplified schematic of optical fiber arrangement around supply and product spool

$$\lambda_M = \rho \times \frac{\pi d^2}{4} \quad (12)$$

The number of rounds per layer is given by:

$$N = \frac{D}{d} \quad (13)$$

And the number of layers is given by:

$$N_L \cong \frac{L}{2\pi r N} \quad (14)$$

Hence, the thickness of the fiber around the spool per unit length is given by:

$$\sigma = \frac{N_L \times d}{L} \cong \frac{1}{2\pi r N} \times d \quad (15)$$

Accordingly, the total thickness of the fiber around the spool is given by:

$$\Delta r = \sigma \times L \quad (16)$$

And the instantaneous length of the fiber around the spool is given by:

$$\ell(t) \cong (r + \Delta r/2) \int_0^t \omega(t) \cdot dt \quad (17)$$

Hence, the instantaneous length of the fiber around the supply spool is given by:

$$\ell_s(t) \cong L - (r + \Delta r/2) \int_0^t \omega_s(t) \cdot dt \quad (18)$$

with initial length  $\ell_s(0) = L$  And the instantaneous length of the fiber around the product spool is given by:

$$\ell_p(t) \cong (r + \Delta r/2) \int_0^t \omega_p(t) \cdot dt \quad (19)$$

with initial length  $\ell_p(0) = 0$

Based on equations (12), (15) and (18) the instantaneous moment of inertia for the fiber winding around the Supply spool is given by:

$$J_{sf}(t) = \lambda_M \times (\ell_s(t)) \times (r + [\sigma \times (\ell_s(t))]/2)^2 \quad (20)$$

Similarly, the instantaneous moment of inertia for the fiber winding around the product spool is given by:

$$J_{pf}(t) = \lambda_M \times (\ell_p(t)) \times (r + \sigma \times (\ell_p(t)/2)^2 \quad (21)$$

The optical fiber is extended from the supply spool to the product spool through free rotating pulleys supported at far ends of transducer arm. As the tension increases, the potentiometer rotates squeezing the spring. When the tension decreases, the potentiometer rotates stretching the spring and relaxing fiber tension. The angle of the potentiometer is proportional to the fiber tension. According to the above analysis, the establishment of the optical fiber tension model is done. The transfer function of dc motor with product spool [6, 15], is given by:

$$\frac{\dot{\theta}_p(S)}{V_{ap}} = \frac{K_m}{[(R_a + L_a S)(J_{Tp})S + (D_{Tp})] + K_m K_b} \quad (22)$$

Where  $V_{ap}$  is the product armature voltage and  $\dot{\theta}_p(S)$  is the angular velocity of the product shaft in rad/s.  $K_b$  is the back electromotive force constant,  $K_m$  is the motor torque constant,  $R_a$  is the armature resistance, and  $L_a$  is the armature inductance.

Figure (3) shows the mathematical model of product and supply spool driving motors.

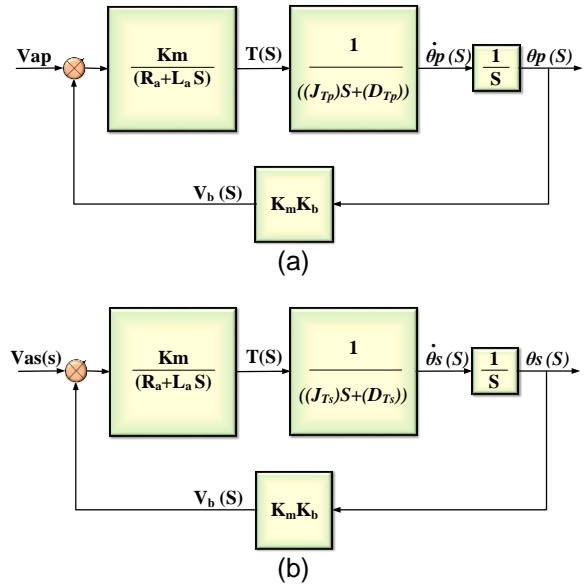


Figure 3 Mathematical model of (a) product dc motor with spool and (b) supply dc motor with spool

The input to the product motor is  $V_{ap}$ . The measured variables are the angular velocity of the product shaft ( $\dot{\theta}_p(S)$ ) and the shaft angle ( $\theta_p(S)$ ) in rad [16]. The transfer function of dc motor with supply spool, is given by:

$$\frac{\dot{\theta}_s(S)}{V_{as}(S)} = \frac{K_m}{[(R_a + L_a S)(J_{Ts})S + (D_{Ts})] + K_m K_b} \quad (23)$$

where  $V_{as}(S)$  is the supply armature voltage and  $\dot{\theta}_s(S)$  is the angular velocity of the supply shaft in rad/s. The mathematical model of supply dc motor with spool is shown in Figure (3) (b).

The input to the supply motor is  $V_{as}(S)$ . The measured variables are  $\dot{\theta}_s(S)$  and the shaft angle  $\theta_s(S)$  in rad. Figure (4) shows the physical structure of the tension transducer and its schematic presentation. The tension transducer consists of damper and spring [8] that produces tension forces proportional to angular rotation [17].

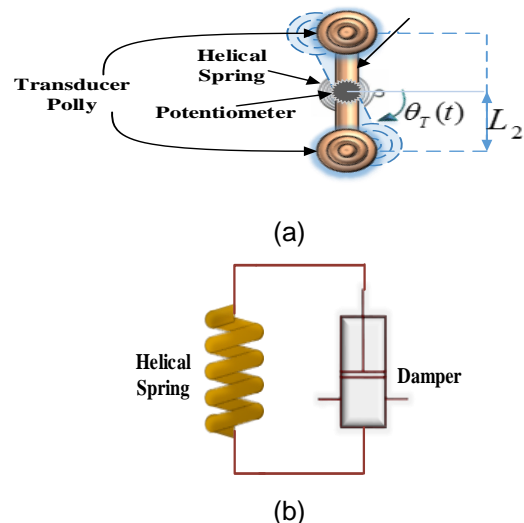


Figure 4 Tension transducer (a) physical structure and (b) its schematic presentation

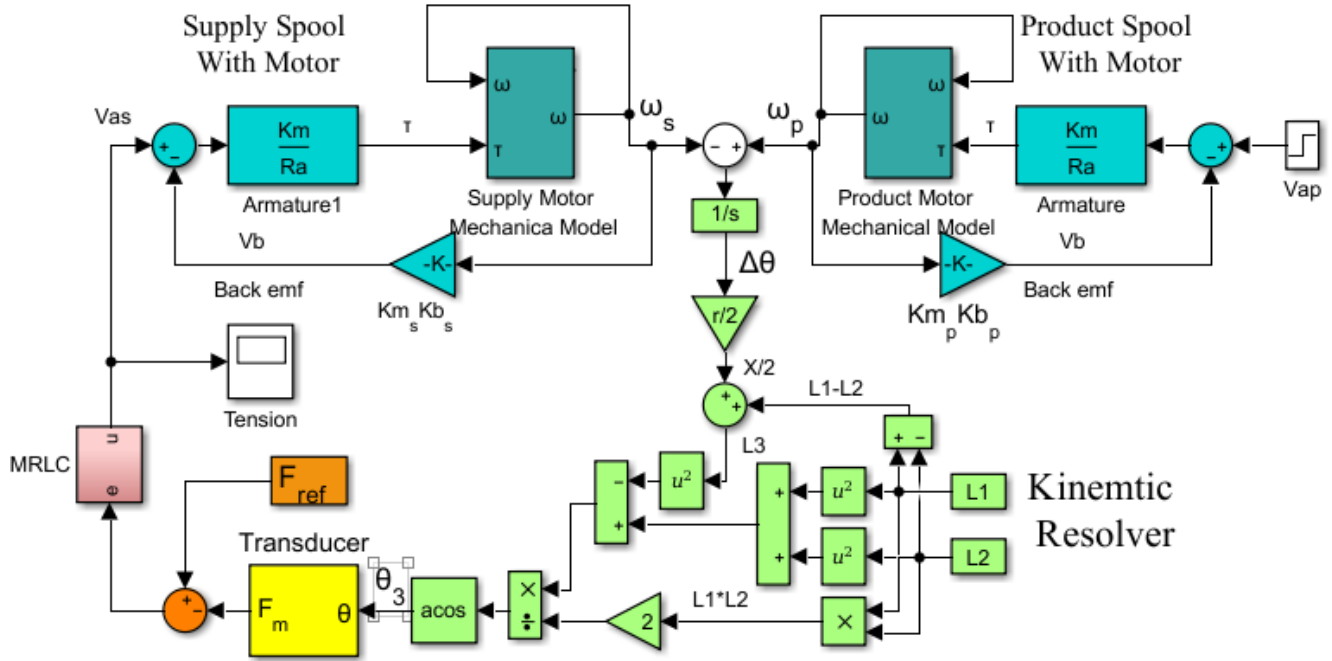


Figure 5 Simulink model of the fiber tension winding system

The transfer function of tension transducer is given by:

$$f(S) = -\frac{J_{sen}S^2 + D_{sen}S + K_{sen}}{2L_2} \theta_T(S) + f_0 \quad (24)$$

where  $f(S)$  is the tension forces applied to the fiber,  $J_{sen}$  is the moment of inertia of the tension transducer,  $D_{sen}$  is the damping (friction) of the tension transducer,  $K_{sen}$  is a constant factor of transducer spring and  $f_0$  is the initial tension experienced by the fiber due to initial torque applied by the initially twisted spring. The negative sign at the first term of equation (24) indicate that the measured angle is against the torque direction and hence the fiber tension direction (the tension decreases as the measured angle increases).

The winder testing machine simulation model is shown in Figure (5). The control input to the system is voltage source  $V_{as}(S)$ . The monitored output is  $f(S)$  to be regulated using the proposed FMRLC algorithm.

The rotor and the pulley are assumed to be rigid. The armature inductance of supply and product motor are neglected. The friction of tension transducer is twice the friction of the motor.

Figure 6 Simulink model of the fiber tension winding system

#### V. FMRLC CONTROL ALGORITHM

Figure (6) shows the proposed control structure. The algorithm consists of reference model, inverse model, adaptation mechanism, and fuzzy controller. Reference model presents desired achievable performance required to follow by the system. Reference model parameters should be chosen reasonably to avoid instability of learning mechanism.

The reference model is a disturbance rejection 2nd order system as shown in Figure (7) Its forward

transfer function is of type one to insure disturbance rejection with zero steady state error and is given by:

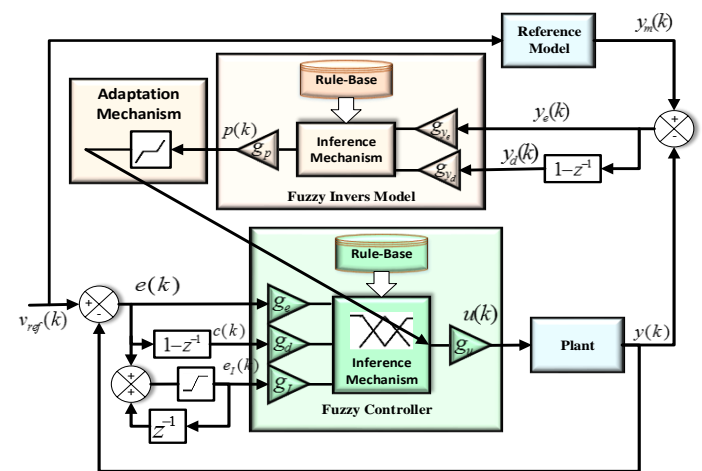


Figure 7 FMRLC algorithm used for the control loop of SVC voltage regulator

$$\frac{Y_d(s)}{Y_m(s)} = \frac{\omega_n^2}{s(s + 2/\tau)} \quad (25)$$

where  $\omega_n$  is calculated by the desired overshoot and  $\tau$  as follow:

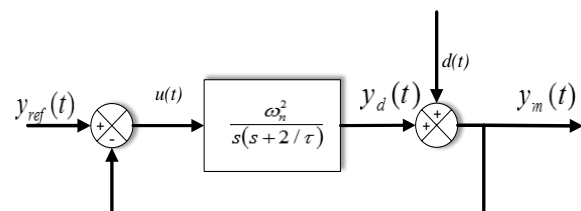


Figure 7 Reference model structure

$$\omega_n = \frac{1}{\tau} \sqrt{\frac{\pi^2 + \ln^2(\%OS/100)}{-\ln(\%OS/100)}} \quad (26)$$

The inverse model is a first order (PD) fuzzy system with rule-base set to describe inverse dynamics of winding process. It generates adequate amount of control action that forces the system to follow the reference model. Based on knowledge of system dynamics, fuzzy inverse dynamics was encoded in rule base of the inverse model as shown in table (1). The inputs to the inverse model are given by:

$$y_e(k) = y_m(k) - y(k) \quad (27)$$

$$y_d(k) = \left( \frac{y_e(k) - y_e(k-1)}{T} \right) \quad (28)$$

Table 1 Rule-base for the fuzzy Inverse model

$\begin{matrix} y_e \\ y_d \end{matrix}$	NB	NE	ZE	PO	PB
NB	NB	NB	NB	NE	ZE
NE	NB	NB	NE	ZE	PO
ZE	NB	NE	ZE	PO	PB
PO	NE	ZE	PO	PB	PB
PB	ZE	PO	PB	PB	PB

NB: Negative big  
 NE: Negative  
 ZE: Zero  
 PO: Positive  
 PB: Positive big

where  $y$ ,  $y_m$ ,  $y_e$  and  $y_d$  are the output of the process, the output of reference model, the error between reference model output and process output, and the derivative of that error respectively. The third part of control algorithm is learning and adaptation mechanism that is responsible for encoding the output of inverse model into the rules at the rule-base of fuzzy controller. The fuzzy controller rules describe the nonlinear control surface that compensates and linearizes the overall system to match the reference model. The adaptation mechanism is also responsible for adapting the control surface to compensate for time varying parameters of winding process. Triangular membership functions are used for fuzzy controller as

shown in figure (8) where its centers  $b_i$  subjected to adaptation as follow:

$$b_i(k) = \begin{cases} b_{\min} & b_i(k-1) + \Delta b_i(k) \leq b_{\min} \\ b_i(k-1) + \Delta b_i(k) & b_{\min} < b_i(k-1) + \Delta b_i(k) < b_{\max} \\ b_{\max} & b_i(k-1) + \Delta b_i(k) \geq b_{\max} \end{cases} \quad (28)$$

$$\Delta b_i(k) = \eta g_p p(k)$$

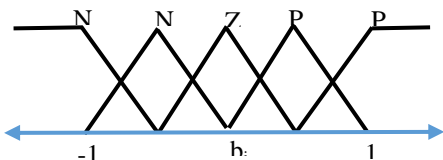


Figure 8 Membership functions of the Fuzzy controller

where,  $\eta$  is the learning factor,  $g_p$  is the adaptation gain,  $p(k)$  is the output of the inverse model and  $b_{\min}$  and  $b_{\max}$  are the minimum and maximum control action values respectively. Both  $\eta$  and  $g_p$  can be combined as a one gain ( $g_p$ ).

The addition of limit values to  $b_i(k)$  ( $b_{\min}$  and  $b_{\max}$ ) is used to avoid generating control actions exceeding process input limits.

Equation (28) shows that FMRLC algorithm provides both adaptation and learning capabilities. This is due to the fact that adaptation process is independent of inputs of the main fuzzy controller. In addition, adaptation process is seeking a certain performance defined by reference model regardless of process changing parameters. To avoid learning instability, dead band was added to adaptation mechanism according to following equation:

$$b_i(k) = \begin{cases} b_i(k-1) + g_p p(k) & -P_{DB} \geq g_p p(k) \geq P_{DB} \\ b_i(k-1) & \text{otherwise} \end{cases} \quad (29)$$

where  $P_{DB}$  is the dead band limit of the learning process. Note that  $\eta g_p$  is replaced by  $g_p$  in equation (29). The main fuzzy controller is a PID fuzzy controller with inputs described by following equations:

$$e(k) = r(k) - y(k) \quad (30)$$

$$c(k) = \frac{e(k) - e(k-1)}{T} \quad (31)$$

$$e_i(k) = T \left( e(k) + \sum_{i=1}^{k-1} e(i) \right) \quad (32)$$

Where  $e$ ,  $c$ , and  $e_i$  are the error between reference input  $r$  and the measured output  $y$ , its derivative and its integral respectively in the current sample  $k$ . (While PD fuzzy controllers are commonly used in control loop of nonlinear systems, it was found that the system with the PD fuzzy controller would possess a large steady state error. A PI fuzzy controller would experience undesired transient performance. Although the PID fuzzy controller increases the dimension of the rule-base matrix (and hence complicate the learning process), it has better performance. The dynamic range that is covered by the membership functions is set to the interval [-1, 1]. The adjustment of the signals dynamic range to dynamic range covered by the membership functions is carried out by the input-output scaling factors  $g_x$  where  $x$  represents the label of the signal at which the gain is placed on.

In order to avoid accumulated large values during transient period from integral part, a limiter is added

after digital integrator described by equation (32). Such large values (over the dynamic range of the proper signal) could cause slowdown of the controller performance. The limiter maximum and minimum values are set in order of magnitude of output signal dynamic range.

The activation level of rule primes is given by:

$$\mu_A^i = \min(\mu_{A_j}^1, \mu_{A_k}^2) \quad (33)$$

where  $i$  is the index of the rule,  $A_j$  is linguistic value for the input (center of the  $i^{\text{th}}$  input membership function i.e.  $A_0 =$  negative big,  $A_1 =$  negative ... etc), and finally,  $\mu_{A_j}^1$  is membership value of 1st input  $A_j$  linguistic value. Center Of Area (COA) is used for the defuzzification process and is given by:

$$y^{crisp} = \frac{\sum_{i=1}^R b_i \sup_y \{ \mu_{\hat{B}^i}(y) \}}{\sum_{i=1}^R \sup_y \{ \mu_{\hat{B}^i}(y) \}} \quad (34)$$

where  $y^{crisp}$  is the crisp outputs,  $R$  is the numbers of rules at rule base, ( $\sup(x)$  denotes supremum value of  $\mu(x)$  which can be assumed as the upper bound of the chopped output membership function) and  $\mu_{\hat{B}^i}$  is the implied fuzzy sets for the  $i^{\text{th}}$  rules of fuzzy controller.

The implied fuzzy set  $\mu_{\hat{B}^i}$  is given by:

$$\mu_{\hat{B}^i}(y) = \mu_A^i * \mu_{B^i}(y) \quad (35)$$

where  $\mu_{B^i}(y)$  represents the membership function of the output (a triangle membership function was used). Assuming the use of minimum and maximum function for inference mechanism then:

$$\sup_y \{ \mu_{\hat{B}^i}(y) \} = \mu_A^i \quad (36)$$

and hence equation (34) can be rewritten as:

$$y^{crisp} = \frac{\sum_{i=1}^R b_i \mu_A^i}{\sum_{i=1}^R \mu_A^i} \quad (37)$$

## VI. SIMULATION RESULT

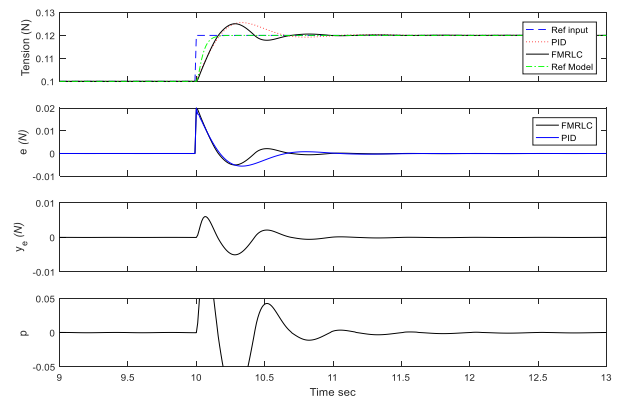
The simulation of tension control model applying both FMRLC and PID controller is conducted for different winding stages. The tension was set to 0.1 Newton with addition of periodic 0.02 Newton. The period of the change was set to 10 sec. A full simulation run was performed for 80000 sec representing the time used to fully wind the fiber from the supply to the product spool. At the early stage, the inertia of supply driving motor with the spool was maximum while the inertia of product driving motor with the spool was minimum (the whole fiber is on the supply spool). The tension control loop was applied to

the supply driving system and significantly affected by its inertia. Table 2 shows the percentage change of the inertia for both supply and product referred to its initial inertia. The negative sign represent the decrement of the inertia and vice versa.

**Table 2 Percentage change of inertia for both supply and product driving motors and spools referred to its initial inertia at three stages of winding process.**

	Stage I	Stage II	Stage III
Time (sec)	9 sec	16009 sec	64009 sec
$J_{sf}(t)$	-0.01%	-9%	-39.18%
$J_{pf}(t)$	0.01%	19.2%	76.8%

Figures (9-11) show the simulation results at different control stages indicated in table (2). For each figure, the top curves represent the reference input (blue dashed line), the response applying FMRLC algorithm (black solid line), the reference model output (Green dashed line), and the response applying PID control algorithm (red dotted line). The 2<sup>nd</sup> curves from the top represent the error between the reference input and the measured tension applying FMRLC (black solid curve) and PID (blue solid curve). The 3<sup>rd</sup> curve from the top represents the error between the output of the reference model and the measured tension for FMRLC algorithm only. Finally, the bottom curve represents the output of the inverse model that indicates the learning activity. It can be noted that both FMRLC algorithm and PID have similar performance at stage I (figure (9)). However, the system performance applying FMRLC improved dramatically compared to PID controller at stage II and III. The FMRLC performance matches closely the reference model as can be noted from  $y_e$  curve in figures (10) and (11). A slight improvement of the transient response can be noted for PID control algorithm due to the decreased inertia of the supply spool. Hence, PID controller has inconsistent performance due to changing inertia that can't be compensated. On the other hand, FMRLC performance is kept constant (following the reference model performance) even when inertia of supply driving system decreased by about 40% of its initial value. This is due to the learning capabilities of the algorithm.



**Figure 9 Simulation results of fiber winding process applying FMRLC and PID controller at early stage (9 to 13 secs)**

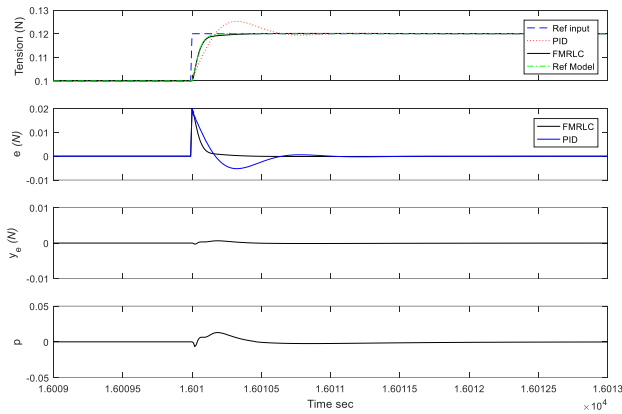


Figure 10 Simulation results of fiber winding process applying FMRLC and PID controller at early stage (16000 to 16013 secs)

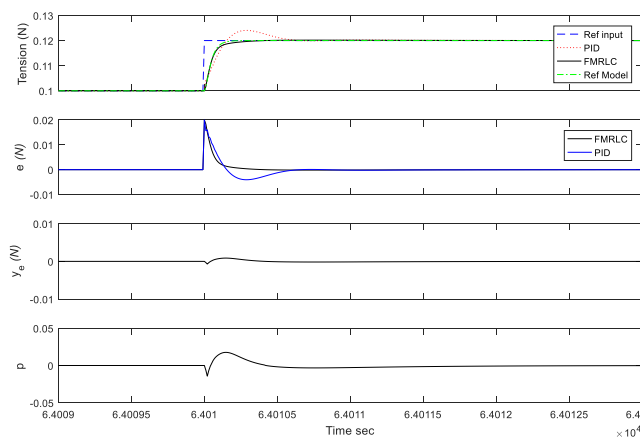


Figure 11 Simulation results of fiber winding process applying FMRLC and PID controller at early stage (64000 to 64013 secs)

## VII. CONCLUSIONS:

This paper presented novel structure and control algorithm of tension control for optical gyroscope fiber winding process. The structure is based on passive dancer that reduces the complexity of the control objectives. A study of the system with its time varying parameters has been conducted. Both, kinematic, and dynamic models have been developed. FMRLC algorithm has been applied to tension control loop of the supply driving system. For verification for FMRLC, the performance of conventional PID controller is compared to the performance of FMRLC. The results show that FMRLC is able to follow the performance described by the reference model at all control stages even with inertia changing of the supply spool.

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Nomenclatures:	
$\theta_p(t)$ ... Product angular position	$D_{sen}$ ... The damping (friction) of the tension transducer
$\theta_s(t)$ ... Supply angular position	$K_{sen}$ ... A constant factor of transducer spring
$L_1$ ... The stretched half length of the optical fiber	$V_b(t)$ ... The back electromotive force (back emf)
$L_2$ ... The half-length of the transducer arm	$K_b$ ... The back electromotive force constant
$L_3$ ... The stretched fiber length between supply spool and the free rotating pulley at the edge of transducer.	$T_m(S)$ ... The motor torque
$r$ ... The radius of the product and supply spools	$K_m$ ... A motor constant of proportionality
$\frac{\Delta L}{l}$ ... The relative elongation of the fiber due to applied tension forces	$R_a$ ... The armature resistance
$E_G$ ... The proportionality constant between the perturbation force per area and the relative deformation	$L_a$ ... The armature inductance
$\Delta\theta$ ... The angular shift between supply motor and product motor	$V_{ap}(S)$ ... The product armature voltage
$l$ ... The length of tangled optical fiber between the two pulleys (supply and product)	$V_{as}(S)$ ... The supply armature voltage
$\Delta X$ ... The linear displacement shift between supply spool and product spool	$\hat{\theta}_s(S)$ ... The angular velocity of the product shaft in rad/s
$\frac{\Delta r}{2}$ ... The average thickness of the fiber around the spool	$\hat{\theta}_p(S)$ ... The angular velocity of the product shaft
$L$ ... Length of the fiber	$K_p$ ... Proportional gain
$d$ ... Fiber diameter	$K_i$ ... Integral gain
$D$ ... Width of the spool	$K_c$ ... Derivative gain
$\rho$ ... The fiber density	$y$ ... Output of the process
$\lambda_M$ ... The mass per unit length	$y_m$ ... Output of reference model
$N$ ... The number of fiber rounds per layer	$y_e$ ... Error between reference model output and process output
$N_L$ ... The number of fiber layers	$y_d$ ... Derivative of error between reference model output and process output
$\sigma$ ... The thickness of the fiber around the spool per unit length	
$\Delta r$ ... The total thickness of the fiber around the spool	
$\ell(t)$ ... The instantaneous length of the fiber around the spool	
$\ell_s(t)$ ... The instantaneous length of the fiber around the supply spool	
$\ell_p(t)$ ... The instantaneous length of the fiber around the product spool	
$D_{Ts}$ ... The total damping (friction) of the mechanical system at supply driving	
$D_{Tp}$ ... The total damping (friction) of the mechanical system at product driving	
$D_m$ ... Motor damping	
$D_{ps}$ ... The damping of product spool	
$D_{ss}$ ... The damping of supply spool	
$J_{Ts}$ ... Total moment of inertia at supply driving side	
$J_{Tp}$ ... Total moment of inertia at product driving side	
$J_m$ ... A moment of inertia of the motor	
$J_{ps}$ ... A moment of inertia of product spool	
$J_{ss}$ ... A moment of inertia of the supply spool	
$J_{sf}(t)$ ... The instantaneous moment of inertia of the supply fiber	
$J_{pf}(t)$ ... The instantaneous moment of inertia of the product fiber	
$f(S)$ ... Tension force of the optical fiber	
$J_{sen}$ ... The moment of inertia of the tension transducer	