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1	Neurofuzzy Implementation in Smart Toolpost to Improve
2	Performance
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6	

7 Abstract

Machining is a complex process that requires a high degree of precision with tight geometrical 8 tolerance and surface finish. Those are confronted by the existence of vibration in the turning 9 machine tool. Overcoming a micro level vibration of a cutting tool using smart materials can 10 save old machines and enhance flexibility in designing new generations of machine tools. Using 11 smart materials to resolve such problems represent one of the challenges in this area. In this 12 work the transient solution for tool tip displacement, the pulse width modulation (PWM) 13 technique is implemented for smart material activation to compensate for the radial disturbing 14 cutting forces. A Neurofuzzy algorithm is developed to control the actuator voltage level to 15 improve dynamic performance. The deployment of the finite element method in this work as a 16 dynamic model is to investigate the ability of the in intelligent techniques in improving cutting 17 tool accuracies. The influence of minimum number of PWM cycles with each disturbing force 18 cycle is investigated in controlling the tool error growth. Toolpost structural force excitation 19 due to the PWM cycles was not given adequate attention in previous publications. A 20 methodology is developed to utilize toolpost static force-displacement diagram to obtain 21 required activation voltage to shrink error under different dynamic operating conditions using 22 neurofuzzy. 23

24

25 Index terms— Tool vibration, Smart Material, Vibration suppression, Cutting tool, Neurofuzzy

²⁶ 1 INTRODUCTION

mproving quality of surface finish and geometrical accuracies during machining using active material was under 27 intensive investigation (Park, et al. 2007). Raw material conversion to new product requires material removal 28 processes using machine tool. Demand for higher productivity in automated manufacturing brought attention 29 for controlling machine tool dynamics for better machining accuracy. Both economic and ecological factors 30 encouraged old conventional machines to continue in service by overcoming tool vibration problems. Various 31 factors might affect the machining process, some of them are non-measurable and others might change in real-32 time. However, the wider use and the availability of cost effective microcontrollers encouraged the implementation 33 34 of intelligent control schemes to overcome such time dependent machining problems ??Krzysztof, et al. 2011). 35 The tiny unfavorable relative motion between the cutting tool and the working piece that associated with high 36 excitation forces encouraged the use of smart material actuators to counteract such motion errors (Radecki, et 37 al. 2010). The rigid fixture is a good choice for minimizing displacements of cutting tools from its nominal position during machining. Unfortunately, such option is not available in all applications. The reconfigurable 38 manufacturing era prefers fixtures consumes less space with minimum weight (Gopalakrishnan, et al. 2002; Moon 39 and Kota 2002). 40

When the control system, and real time microprocessor implementation were examined no details were given for the design and selection of actuator, tool holder, and tool bit stiffness, and, actuator switching. Also in

5 B) BUILDING THE FINITE ELEMENT EQUATIONS

43 the case of future geometrical changes, the validity of using lumped masses in system modeling is questionable.

⁴⁴ Information is required regarding the nature and type of signals controlling smart material and how might affect ⁴⁵ toolpost dynamic response. Recently a tool adaptor is used with built-in active vibration damping device to

46 dynamically stabilize the turning process (Harms, et al. 2004). The vibration compensation system is based

47 on a multilayer piezoactuator in collocation with a piezoelectric force sensor. An analogue controller based on

48 integral force feedback method is used for active damping. Latest dynamic modeling of smart toolpost (Rashid,

49 2005) is based on continuous elastic structural toolpost overcoming previous limitations of lumped mass modeling

50 without further steps toward the development of a generalized scheme for tool error attenuation. Such models

⁵¹ are then implemented for designing an adaptive controller using fuzzy controller (Rashid, 2006).

This work is implementing the finite element method (FEM) to model flexible smart tool post incorporating PZT actuator, tool holder, supporting fixture, and tool bit, and, discusses the effect of structural properties on the critical frequencies as compared to lumped mass modeling. Also investigate the effectiveness of the developed Neurofuzzy algorithm in controlling error attenuation under different excitation cutting force frequencies. The

tool radial motion that causes dimensional variation in the work-piece is emphasized. The smart toolpost static

57 forcedisplacement analysis under different voltage input is integrated with the development the Neurofuzzy 58 scheme1 VII (A)

to predict the activation voltage in dynamic error attenuation. A special attention is given for the model to be a robust for large variations in design parameters. Such a finite element model offers a methodology for

⁶¹ micro-vibration attenuation in smart toolpost using smart materials and intelligent schemes like Neurofuzzy.

62 **2** II.

63 **3** THE TOOLPOST FEM MODEL

⁶⁴ In this work Lead Zirconate Titanate (PZT), is employed as a smart material actuator. This is encouraged by a

65 well-developed theoretical analysis of this material. Also it is the most common used piezoelectric materials. The

Toolpost model incorporates actuator, tool carrier (holder), supporting diaphragm, and tool bit (spring buffer between the tool carrier and the net actuating force at tool tip) as shown in Figure ??.

⁶⁸ 4 a) Model Governing Equations

The model consists of a conventional stacked PZT actuator contains polarized ferroelectric ceramic in the direction of actuation, adhesive, supporting structure, and electrically wired electrodes as shown in Figure ??.

The finite element modeling of the PZT actuator and toolpost is developed (Piefort 2001) by using the general constitutive equations of linear piezoelectricity and equations of mechanical and electrical balance.

The momentum balance equation is And the electric balance equation is Knowing, Where represents the stress vector, , the strain vector, , the electric field, , the electric displacement, , the elastic coefficients at constant ,] [S ?

, the dielectric coefficients at constant , and , the piezoelectric coupling coefficients. As well is the mechanical
displacement vector and is the acceleration. In addition is the electric potential (voltage). The boundary
conditions are shown in Figure ??, as represented by the fixed end conditions for both, actuator left side
and, diaphragm outer edge. The model description is completed by specifying the applied voltage at actuator
electrodes' using the PWM technique weighted by a factor worked out from the developed fuzzy algorithm and

⁸¹ based on the inputs from the calculated toolpost dynamic response.

⁸² 5 b) Building The Finite Element Equations

The finite element model of this work is built by using the piecewise application of classical variational methods 83 on smaller and simpler sub-domains connected to each other by a finite number of nodes. A 8-node isoparametric 84 solid element is used for domain discretization. The unknowns are the displacements vector and the electric 85 potential values at node i. The formulation of the dynamic equations of a piezoelectric continuum is discussed 86 thoroughly in the literature (Allik and Hughes 1970;Lerch 1990). Taking into account the constitutive equations 87 (1) and by introducing the Lagrangian and virtual work formulation into the Hamilton's principle we that satisfies 88 the arbitrary deviation of the displacements and the) 2 (] .{ } { T u ? = ? ? ?) 3 (0 } .{ = ? D ? ?? = ? = 89 } { }, .{ } { E u S S } {T } {S } {E } {D] [E c } {E } {S] [e } {u 2 2 / } { } } { t u u ? ? = ? ? ? i u i ? } 90 { i u 2 Global (A) 2011 December }]{ [}]{ [} {) 1 (} {] [}]{ [} { E S e D E e S c T S T E ? + = ? = 91 92 electrical potentials

and their compatibilities with the compatibilities with the associated boundary conditions,

, and, are the mechanical mass, stiffness, and, damping matrices, respectively. is the piezoelectric coupling matrix and, is} { i ? } { }]{ [}]{ [}]{ [}]{ [i i u i uu i uu i uu f k u k u c u m = + + + ? ? ? ? } } } } { }]{ [] i i i T u q k u k = + ? ?? ?] [uu m] [uu k] [uu c] [?? k] [? u k] [?? k

97 the dielectric stiffness matrix. and } { i q are the nodal mechanical force and electric charge vectors, 98 respectively. and, are the nodal displacement and } { i f potential vectors, respectively. For the sake of brevity, 99 the scheme by which the elemental contributions are assembled to form the global system matrices are discussed

in ??Zienkiewicz and, Taylor 2000a, b).

101 III.

¹⁰² 6 NATURAL FREQUENCY COMPARISON BETWEEN ¹⁰³ LUMPED AND FEM MODELING

The lumped mass modeling of the PZT actuator and tool carrier generate a simple closed form solution that is of interest to designers and modelers. However using such model in different applications requires more alertness. Most applications require a precise displacement sensing and accurate prediction for natural frequency to ensure the effective control for smart material actuator. a) Comparative Results for PZT Actuator Mode shapes and the resonant frequencies for undamped system are obtained by using Eigenvalue analysis. Free vibration implies that0 } { = i f and 0 } { = i q }

in equation (4). Modal analysis is based on the orthogonality principle of natural modes and expansion theorem (Zienkiewicz and Taylor 2000 b).

The fundamental angular undamped natural frequency of a beam of fixed -free end condition is represented by:n ? = (5)

Usually the actuator is composed of several PZT layers, electrodes, adhesive, and supporting structure as shown in Figure ??. To compare between simple calculations and the FEM solutions the actuator effective stiffness assumed to be the stiffness sum in the Plotted results in Figure ?? series for all individual layers neglecting all piezoelectric effects.) 6 () (i i i L E A KA ? =

Actuator effective lumped mass is estimated to be a 20 or 30% of the summed layers masses as indicated in Figure ??.) 7 () () 3 . 0 2 . 0 () (i i i eff A L A or M ? ? = Then,) 8 () (eff A A Lumped M K = ?

The FEM solutions for the actuator first natural frequency under short circuit (zero piezo effect) and represent 30% of the combined actuator and tool carrier masses.

 122 $\,$ While the effective mass near to the diaphragm endeff CE M) (

is 30% of the combined tool carrier and diaphragm masses. Higher mass percentages resulted in a higher deviation from the FEM solutions. Solutions for the two-degree of freedom system incorporating piezoelectric coupling effects are investigated by Abboud, et al. 1998), However, the significant deviation between the FEM and the lumped mass solution and its range of applicability has not investigated. KA is calculated as in Eq. (??) which is shown in Fig. ?? as a coupling stiffness joining Diaphragm stiffness (KD) is based on plate theory with fixed central hole at both inner and outer edges, (Roark and Young 1975). Then the stiffness matrix of this 2-degree of freedom system shown in Figure ?? is:) 9 (] [??????????? +?? + = KD KC KC KC KA K

Modal shapes and frequencies that resulted from the FEM model compare to the lumped mass model in Figure 131 ?? The second critical frequency ratio on the semi-log plot of Figure ?? shows independency of such ratio on tool 132 carrier stiffness when selecting higher ratio of (KD/KA) which is not recommended for actuator design. At low 133 (KD/KA) the dependency of frequency ratio on (KC/KA) is much more distinguished. In general the frequency 134 ratio is tending to change remarkably when (KC/KA) goes beyond ten. Such disagreement between lumped 135 mass models and FEM solution requires more awareness in vibration controller design using smart materials. 136 IV.

TOOLPOST FORCE GENERATION VERSUS DISPLACE MENT

Effective tool error attenuation depends on PZT actuator capabilities for resisting tool axial force within the 139 required limited range of motion. To obtain such data a force versus displacement curve is developed for the 140 investigated toolpost in Figure ??. Figure ?? shows the force-displacement characteristics for different values of 141 tool carrier to actuator stiffness ratio (KC/KA). The plotted curves in Figure ?? are emphasizing the importance 142 of increasing (KC/KA). Similar plots are generated to obtain the effects of increasing (KT/KA) and (KT/KA). 143 Resulted figures indicate the worth of reducing the structural support stiffness (diaphragm) in the direction of 144 PZT activation to improve actuation force for error attenuation. Guessing first actuator design can be conducted 145 according to information offered by forcedisplacement calculations. A special treatment for dynamic effects during 146 machining is discussed in the next sections. 147

The smart material data and, the investigated toolpost dimensions that applied to both static and dynamic calculations are given in Table 1. A general theory for a piezoelectric actuator subjected to mechanical excitations and feedback voltages is discussed in (Tzou 1991).

¹⁵¹ 8 NEUROFUZZY ALGORITHM FOR VOLTAGE ACTIVA ¹⁵² TION

Obtained results from Figures 5-7 prove significant deviation of lumped mass modeling from the finite element solution of the continuous elastic structure especially in the range of low (KD/KA) and high (KC/KA) where the PZT actuation is maximum as pointed out in Figure ??. Therefore the finite element method is the only reliable and available tool of solution in assessing switching methodology and system damping in the smart toolpost toward effective error reduction. Transient solution for tool displacement is achieved by solving Eq. (4) in the time domain for the system shown in Figure ??. The smart toolpost configuration and its associated data are given in Table 1. The voltage activation to the smart material is either triggered by a piezo stack with force sensing layer or by using a suitable type of displacement sensor. In both methods sensing the location should reflect cutting tool position error correctly. Switching circuits design is not discussed in this work; however the required voltage intensity level and the resulted motion are emphasized in this work.

Setting up the switching voltage by a series of PWM cycles should be judged by a sensed cutting force value from the peak force spectrum at the peak force frequency (), then, the initial peak voltage is estimated from Figure ??. A complete period of the peak force cycle (

) is divided into number of duty cycles (or NPWM). Then, for any of these divisions, the time duration of
the PWM high DC-voltage is calculated based on the obtained voltage factor from the neurofuzzy algorithm
that will be discussed next. A time delay in voltage activation can be incorporated as a function of the peak
force frequency period. Two switching's are associated with each PWM cycle segment, therefore switching rate
is . Effects of switching voltage input, forcing frequency , and, damping level upon toolpost time response are

171 parameters to be discussed in a smart toolpost transient solution.

¹⁷² 9 b) Controller Configuration

It is difficult to acquire a controller that ensures continuous error tracking under stabilized condition for smart 173 toolpost under continuous exposal to an erratic real time force inputs. The use of intelligent controller is generated 174 by the random nature of system excitations which largely depends on unpredictable parameters such as structural 175 properties, friction, and other variable dynamic forces. A neural network can model the response of such system 176 by means of a nonlinear regression in the discrete time domain. The result is a network, with adjustable weights, 177 that might approximate the system dynamics. Though it is a problem since the knowledge is stored in an opaque 178 fashion and the learning results in a large set of parameter values which almost impossible to be interpreted in 179 words. Conversely using a fuzzy rule based controller that consists of readable if-then statements which is almost 180 a natural language, cannot learn new rules alone. The neurofuzzy controller might be preferred over the others 181 for such application since it combines the two and it has a learning architecture (Lin, J. and Chao, W.S., 2009). 182 To construct a neurofuzzy controller with ANFIS (Adaptive Neuro Fuzzy Inference System), we need a set of 183 input-output data. In this work, two input signals are considered. The first input is normalized error defined by 184 representing the negative error by the tool tip displacement away from the work piece axis. Then is the negative 185 of the normalized tool tip error with respect to the maximum static displacement of the peak radial cutting 186 force? The universe of discourse of the input variable is defined to be within the range A are fuzzy sets whose 187 membership functions are denoted by the same symbols as the fuzzy values. Given an inputs (e 1, e 2, ..., e 188 k) the final output of the fuzzy model is inferred by taking the weighted average of the i f is: i n 1 i i i n 1 i w f 189 w y ? ? = = = (12) where 0 i w ? 190

and i f is calculated for the input by consequent equation of the i-th rule, and the weight i w implies the overall truth value of premise of the i-th rule for input calculated asp e i p A k 1 p. i w ? = =(13)d

¹⁹⁴ 10) The Neuro-fuzzy control algorithm

To facilitate the learning (or adaptation) of the Takagi-Sugeno fuzzy model, it is convenient to implement the 195 fuzzy model into a framework of adaptive network that can compute gradient vectors systematically. The resultant 196 network architecture called ANFIS (Adaptive Neuro-Fuzzy Inference System). ANFIS is described by a similar 197 Takagi-Sugeno model with a single difference that in this case the inputs, e 1 (? -Error) and, e 2 (?? -Rate of 198 Error) are range values. The fuzzy set for ? being 1 A = { L="Low", M/L="Medium to low", M="Medium", 199 M/H="Medium to High" and H="High"}, and fuzzy set for being 2 A = { L="Low", M/L="Medium to 200 low", M="Medium", M/H="Medium to High" and H="High"}. Fig. ??1 illustrate graphically the neurofuzzy 201 reasoning mechanism to derive an output y from a given inputs ? and ? ? . Output i f is one of the voltage 202 factor VF for i-th rule where the size of the rule base is 25. The dynamic simulation is conducted with several 203 types and sizes of membership functions for the fuzzy sets 1 A and, 2 A. The triangular membership functions 204 and a size of five for each of the two fuzzy sets were found the simplest and best suited for this case. 205

The square elements in Fig. ??1 represent the adaptive nodes depending on the parameter set of the adaptive network. The circles represent fixed nodes, which are independent of the parameter set. The first layer is composed of adaptive nodes representing the triangular membership functions (Jang, J.-S. R.; Sun, C.-T. & Mizutani, E. 1996) associated with each linguistic value. The second layer implements the fuzzy rules. Each node in this layer calculates the firing strength of a rule by means of multiplication between the membership degrees of the two inputs. The third layer consists of adaptive nodes which include the output membership.

The other two layers consist of fixed nodes that implement the weighted average procedure to obtain the voltage factor VF as shown in Fig. **??1**. As the size of the rule base of the Sugeno fuzzy inference system (SFIS) is 25, we will have to identify 75 consequent **??1**996). This can be obtained from the neural network (NN) using training set { ? , ? ? , VF } which are collected from the dynamic simulation results by using the Sugeno fuzzy inference system. A back-propagation learning algorithm is used to identify these parameters in two steps. In the forward pass, the input membership functions are fixed and consequent parameters associated with the output are calculated by applying the least square estimation method. Using these parameters, the NN generates an estimate of the output voltage factor VF. The difference between this estimate and the motor's value from the training set is then backpropagated in a second pass when the premise parameters associated with the input membership functions are calculatedVII (A)

222 2011 December e) Solution Method for Dynamic Equations

The system of equations for such a nonlinear problem is best solved by classical Newmark algorithms (Abboud, 223 et al. 1998). Time step-by-step integration is used in solving Eq. (4). Basically the final results are obtained 224 by attaining the solution at present time step from known solution at the previous time step. This approach 225 takes into consideration the higher order time approximations. Also it assumes a constant acceleration over a 226 small time interval (time-step). By considering the Taylor series quadratic expansion for the function and its 227 derivatives, (Abboud, et al. 1998) then, (14) From the dynamic equation Eq. (??) we have (15) Equations (228 ??4) and (??5) allows three unknowns and, to be determined and for brevity the detail of solving these equations 229 and the values of and, are given in (Abboud, et al. 1998). The step-by-step integration scheme assumes a known 230 structural damping. The damping matrix is assumed to be a linear combination of stiffness and mass matrices 231 (Rayleigh damping) (Bathe 1982): (16) Where and are constants to be determined from two assumed modal 232 damping ratios () (1% and 5%) for first and second natural frequencies respectively which are related to modal 233 234 damping by the available established relations.

235 VI.

11 RESULTS OF FUZZY CONTROLLED RESPONSE FOR INTEGRATED TOOLPOST

Requirements to reduce tool holder size and weight encourage developing new tactics of using smart actuators to attain high precision by compensating unfavorable motion errors. Estimation of cutting tool radial force might involve several variables. In general the static force relation) can be used as a first guess in error attenuation:r r r K t f V d K F) (??? = a general constant (17)??, r K

and, ? are to be calibrated for each tool-workpiece, tool-work material combinations, process types, tool-wear 242 condition, workpiece hardness, tool geometry, and speed. For the presented results, the applied voltage to the 243 actuator is estimated first from both Eq. (???) and Figure ??. The subsequent applied voltage values are then 244 obtained from the neurofuzzy output voltage factor of Figure ??1 based on the resulted error and the rate of 245 error values. Actuator data for the obtained final results are given in Table 1. Using few numbers of PWM cycles 246 per force period can cause unfavorable switching dynamic excitation by the actuator to tool post. The results 247 in Figure ??2are the outcome of five PWM cycles, twenty PWM cycles per force period produce more favorable 248 results but more than twenty have little effect. Cutting force fluctuations have a component that is proportional 249 to the undeformed chip thickness and a component due to the rate of penetration called a plowing effect. For 250 comparison, the results of increasing ? in equation (16) by ten folds of the selected datum of 1% damping ratio 251 for first mode and 5% for the second mode indicated a significant reduction in tool tip normalized error. In 252 addition to the effectiveness of the Neurofuzzy algorithm both damping and NPWM (253

262 **12 2011**

263 **13** December

The negative normalized error in Figures ??2-13 indicates outward tool tip retraction away from workpiece axis. Tool bit to actuator stiffness ratio (KT/KA) has an importance in terms of force availability for tool tip error elimination and accurate displacement sensing as shown in Figure ??. Stiffness ratios greater than ten produce identical displacements between the tool tip and tool carrier main body. Taking into consideration the geometrical factors, significant deviation starts when stiffness ratio (KT/KA) drops below one. The importance of such parameter depends on the design configuration of the tool post and the acceptable range for the tool error. of tool tip errors. Results are compared for two frequencies as shown in Figure ??3.

271 VII.

272 14 CONCLUSIONS

The application of neurofuzzy techniques to control a smart tool post has been presented. An adaptive learning algorithm for the neurofuzzy controller has been developed. The advantages of the proposed schemes are that an accurate model to describe the dynamics of the tool post is no longer needed, and the choice of learning

parameters for the controller is not critical. With the proposed method, the controller can be easily designed 276 and expanded. Designing against cutting tool error in turning machines using smart material reduce industrial 277 waste, save money and, improve design flexibility for future tool generations. In this work critical frequencies of 278 two modeling schemes are compared for a smart tool post under open and short circuit conditions. The results 279 indicate significant differences between lumped mass modeling and FEM solution at low diaphragm to actuator 280 stiffness ratio (KD/KA) and high tool carrier to actuator stiffness (KC/KA). This range of stiffness ratio has 281 been investigated to ensure better error attenuation in smart material actuation for such applications. The work 282 283 outcome can identify the stiffness range of lumped mass modeling that has more realistic representation of the dynamic response control. Also, suggest the use of high tool bit to tool carrier stiffness for better actuation 284 capability and smaller tool tip error. Generated results suggest a reasonable number of at least twenty PWM 285 segments should be used in representing the force cycle to reduce switching dynamic transient effects. The 286 developed methodology in generating voltage activation factor to modify the static voltage-force-displacement 287 value proved absolute effectiveness in error attenuation. The developed neurofuzzy algorithm to predict the 288 voltage activation factor is based on both normalized error and rate of change in error and proved an ultimate 289 success independent of forcing frequency. The neurofuzzy algorithm for voltage activation has contributed in 290 reducing the too post error.



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Figure 1:

¹December

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Figure 2:



Figure 2: PZT Stacked Actuator

Figure 3:



 $\mathbf{5}$

and open circuit (Op)



Figure 4: Tool Carrier and Actuator with Diaphragm

Figure 5:



Figure 6:



Figure 7: 10)December



Fig. 7: Ratio of FEM third critical to second lumped mass critical frequency versus (KC/KA) at different (KD/KA) for open (Op) and short (Sh) circuits

Figure 8:



Figure 8: Smart toolpost displacement versus the applied force for different tool carrier to actuator stiffness ratio.

Figure 9:

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Item	Value	Units
Cylindrical PZT-8 Stack		
PZT Thickness	0.09e-03 m	
Electrode Thickness	0.03e-03 m	
Structural support	0.03e-03 m	
Adhesive Thickness	10.0e-06 m	
Number of layers	500	
Effective Radius	5.0e-3	m
Steel Cylindrical Tool Carrier (holder)		
Radius	10.0e-3	m
Length	$55.0\mathrm{e}{-3}~\mathrm{m}$	
Steel Tool Bit Effective Length		
Assumed Effective Length	20.0e-3	m
Steel Diaphragm		
Thickness	0.5e-3	m
Outside Radius	20.0e-3	m
V.		

Figure 10: Table 1 :

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13