



INTELLIGENT ROBUST CONTROL OF PRECISION  
POSITIONING SYSTEMS USING ADAPTIVE  
NEURO FUZZY INFERENCE SYSTEM

BY

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## ABSTRACT

Recently, there has been an increasing interest in the application of robust control theory for Precision Positioning Systems (PPS). This is mainly driven by the need to provide guaranteed stability in spite of uncertainties and disturbances associated with these systems. However, robust control techniques require a dynamic model of the plant under study and bounds on modelling uncertainty to develop control laws with guaranteed stability. Although identification techniques for modelling dynamic systems and estimating model parameters are well established, very few procedures exist for estimating uncertainty bounds. A conservative bound is usually chosen to ensure robust stability for a reasonable range of variations about the nominal model. Nevertheless, high performance requirement of PPS will be severely affected. In this research an intelligent uncertainty function is developed to improve the performance of  $H_\infty$  robustly controlled high precision positioning system in terms of reduced conservatism. The proposed approach can be systematically applied. First, the nominal model of the positioning system is identified; output performance and control signal requirements are then determined by proper selection of performance and control weighting functions. Adaptive Neuro Fuzzy Inference System (ANFIS) is used to produce the uncertainty bounds of model uncertainty that results from unmodeled dynamics and parameter variations. The synthesis of the  $H_\infty$  controller will incorporate these weighting functions. Then to further improve the controlled system performance, an unconstrained optimization procedure is developed to obtain the best possible performance weighting function. Moreover, an intelligent disturbance weighting function is developed to eliminate the effect of crosstalk between the axes.  $\nu$ -gap metric is utilized to validate the identified uncertainty set for robust controller design.  $\mu$ -analysis is used to evaluate the robustness of the system. The computational time and number of iterations of the proposed intelligent estimation method are decreased to  $< 0.1$  of that required by a neural network method with less or equal  $\nu$ -gap metric value. Simulation and experimental results using different servo motion plants reveal the advantages of combining intelligent uncertainty identification and robust control. Improved performance has been achieved for rotational motion, single axis and two-axis servo systems. Settling time  $< 0.8$  seconds, rise time  $< 0.5$  and steady state error within sensor resolution are achieved for the rotational motion system. In the case of the X-Y positioning systems, tracking errors are reduced to less than 100% of that obtained using a well tuned conventional PID controller and less than 10% of that obtained using a nominal  $H_\infty$  robust controller.  $\nu$ -gap metric value of  $< 1.0$  and larger stability region can be readily obtained for both cases. Robust stability and performance are also guaranteed. The generality of the problem formulation enables the application for more complicated systems.

## خلاصة البحث

لقد أزداد مؤخرا الأهتمام بتطبيق نظرية السيطرة المتينة على أنظمة الحركة ذات الدقة العالية. ويعود السبب في ذلك للحاجة الماسة لتوفير سيطرة مؤكدة على هذه المنظومات حتى في حال وجود عدم وثوقية أو اضطرابات مرافقة. و حيث ان تقنيات السيطرة المتينة تتطلب توفير نموذج ديناميكي محدد للنظام قيد الدراسة بالإضافة الى تعريف حدود عدم الوثوقية في النموذج. و على الرغم من توفر تقنيات متطورة لتمثيل الأنظمة الديناميكية و تعريفها و تخمين عواملها لا نجد الا طرق قليلة لتعريف حدود عدم الوثوقية. و لقد جرت العادة على اختار حدود عدم وثوقية محافظة للحصول على السيطرة المتينة مما يؤثر سلبا على نوعية المتطلبات العالية لأداء منظومة الحركة الدقيقة . لذلك تم في هذا البحث تطوير دالة كشف عدم وثوقية ذكية لتحسين السيطرة و الأداء المتينين لمنظومة سيطرة عالية الدقة باستخدام  $H_{\infty}$  مع تقليل المحافظة الزائدة. بالأمكان تطبيق الطريقة المقترحة بشكل منظم حيث يتم في البداية تعريف منظومة الحركة المعنية, ولتوفير متطلبات اشارة السيطرة و الاداء الخارجي يتم اختيار دوال اهمية مناسبة لهذا الغرض, و من ثم يتم تعليم منظومة ذاتية التكيف ذات نظام عشوائي- عصبي ANFIS لموائمة تأثير حدود عدم الوثوقية الناتجة عن العوامل الديناميكية الغير معرفة او المتغيرة بشكل دقيق. ويتم بعد ذلك تركيب مسيطر متين  $H_{\infty}$  بالاستفادة من دوال الأهمية المستحصلة سابقا. و لزيادة تحسين أداء منظومة السيطرة تستخدم خطوات تحسين أمثل غير مقيدة للحصول على افضل دالة اهمية للأداء. بالإضافة الى تعريف دالة أهمية ذكية أخرى للتخلص من تأثيرات الاضطراب الناتجة من التعشيق بين المحاور الحركية المتعددة. و لقد استخدمت الفجوة المترية- $\nu$  للتصديق على دقة مجموعة عدم الوثوقية المعرفة. كما تم استخدام طريقة  $\mu$  لتحليل مدى متانة المنظومة الناتجة. كشفت نتائج تجارب المحاكاة و التطبيق العملي للمسيطر الذكي المتين على أكثر من جهاز حركة دقيقة عن فوائد الجمع بين التعريف الذكي لحدود عدم الوثوقية و السيطرة المتينة. حيث تم الحصول على تحسن باداء منظومتي حركة دقيقة احادية و ثنائية المحاور. كذلك تم تحقيق قيم صغيرة للفجوة المترية- $\nu$  مع مساحة سيطرة كبيرة, حيث تم الحصول على سيطرة و اداء متينين. بالاضافة الى تحسن عالي بأداء المسار و الخطط لمنظومة ثنائية المحاور مما يدل على مدى فاعلية الطريقة المستخدمة. ان الطريقة المقترحة ذات خاصية عمومية شاملة و عملية مما يتيح امكانية تطبيقها لمنظومات أكثر تعقيدا.

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## DECLARATION

I hereby declare that this thesis is the result of my own investigations, except where otherwise stated. I also declare that it has not been previously or concurrently submitted as a whole for any other degrees at IIUM or other institutions.

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## LIST OF ABBREVIATIONS

AFC	Adaptive Fuzzy Control
AFIS	Adaptive Fuzzy Inference Machine
ANFIS	Adaptive Neuro Fuzzy Inference System
ANN	Artificial Neural Network
AMB	Active Magnetic Bearings
AR	Auto Regressive
ARC	Adaptive Robust Control
ARMA	Auto Regressive Moving Average
ARMAX	Auto Regressive Moving Average with External Inputs
ARX	Auto Regressive with External Input
A/D	Analogue to Digital
CCC	Cross-Coupled Controller
CIN	Confidence Interval Network
CMM	Coordinate Measuring Machines
CNC	Computer Numerical Control
CNF	Composite Nonlinear Feedback
CP	Continuous- Path
DAQ	Data Acquisition System
DC	Direct Current
DFLS	Dynamic Fuzzy Logic System
DNLRX	Dynamic Nonlinear Regression with Direct Application of EXcitation

DOB	Disturbance Observer- Based
DDOB	Digital Disturbance Observer- Based
DOF	Degree Of Freedom
DRNN	Dynamic Recurrent Neural Network
DSP	Digital Signal Processor
D/A	Digital to Analogue
EDM	Electric Discharge Machining
emf	Electro-Motive Force
FNN	Fuzzy Neural Network
FIS	Fuzzy Inference Systems
FPE	Final Prediction Error
GIMC	Generalized Internal Model Control
GMS	Generalized Maxwell Slip
HPPS	High Precision Positioning System
IFOC	Indirect Flux Oriented Control
ILC	Iterative Learning Control
IMC	Internal Model Control
IPF	Intelligent Pre-shaping Filter
LBLDCM	Linear Brushless Direct Current Servo Motors
LFT	Linear Fractional Transformation
LMI	Linear Matrix Inequality
LPMSM	Linear Permanent Magnet Synchronous Motor
LUSM	Linear UltraSonic Motor
MEM	Model Error Modelling
MIMO	Multi-Input-Multi-Output