

UNIVERSITI PUTRA MALAYSIA

AN IMPROVED FUZZY PARALLEL DISTRIBUTED -LIKE CONTROLLER FOR MULTI-INPUT MULTI-OUTPUT TWIN ROTOR SYSTEM

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FK 2009 72



An Improved Fuzzy Parallel Distributed –Like Controller for Multi-Input Multi-Output Twin Rotor System

By

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Thesis Submitted to the School of Graduate Studies, Univeristi Putra Malaysia, in Fulfillment of the Requirements of the Degree of Master of Science

May 2009



DEDICATION

To my Parents,

To my Brother and Sisters,

To my Grandmothers

Thair



Abstract of thesis presented to the Senate of University Putra Malaysia in fulfillment of the requirement of the degree of the Master of Science

An Improved Fuzzy Parallel Distributed –Like Controller for Multi-Input

Multi-Output Twin Rotor System

By

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December 2008

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Twin Rotor Multi Input Multi Output (MIMO) System (TRMS) is a laboratory set-up

design for which it has been used for control experiments, control theories

developments, and applications of the autonomous helicopter. Fuzzy Logic Control

(FLC) has been widely used with different control schemes to cope with control

objectives of TRMS. In this work, Self Tuning Fuzzy PD-like Controller (STFPDC)

is proposed to make the response of FLC more robust to the interactions and the non-

linearity of the process in terms of less rising time, settling time and overshoot.

Adaptive Neuro Fuzzy Inference System (ANFIS) based Fuzzy Subtractive

Clustering Method (FSCM) was used to remodel the proposed STFPDC to achieve

the control objectives on TRMS with less number of rules. MATLAB/SIMULINK

was involved to achieve the simulations in this work. The results showed the

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proposed controller could simplify the STFPDC to reduce the number of rules from 392 to 73, which is even less than the original FLC that has 196 rules.

The conclusion of this work is improving FLC response by using STFPDC and reducing the number of rules used to achieve this improvement by using ANFIS based on FSCM modeling. For future works, it is recommended to develop an optimization algorithm which achieves best selection for the range of influence which gives best response with less number of rules.



Abstrakt tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Master Sains

Kawalan Bagai Agihan Selari Kabur Yang Diperbaharui Bagi Sistem Putaran

Kembar Pelbagai Input Pelbagai Output

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dicadangkan boleh memudahkan

Sistem Rotor Kembar Pelbagai Input Pelbagai Output (MIMO) (TRMS) adalah satu reka cipta makmal yang sebelum ini digunakan untuk uji kaji kawalan, pembangunan teori kawalan dan aplikasi helikopter autonom. Kawalan logik kabur (FLC) telah banyak digunakan dengan skim kawalan yang berbeza bagi menampung tujuan kawalan TRMS. Dalam tugas ini, pengawal kabur putaran sendirian bagai PD (STFPDC) dicadangkan untuk menjadikan gerak balas lebih tegap daripada FLC biasa bagi interaksi dan proses yang taksekata dari segi pengurangan masa bangkit, masa tinggal dan keterlanjuran. Cara gugusan penolakan kabur yang berdasarkan sistem kesimpulan kabur penyesuaian saraf (ANFIS) telah digunakan untuk mengubah bentuk cadangan STFPDC untuk mencapai tujuan kawalan TRMS dengan bilangan peraturan yang kurang. MATLAB/SIMULINK telah dilibatkan untuk mencapai simulasi dalam tugas ini. Keputusan ini mununjukkan sistem kawalan yang

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STFPDC dengan mengurangkan bilangan

peraturan daripada 392 peraturan kepada 73 peraturan, malah kurang daripada FLC tulen yang mempunyai 196 peraturan.

Kesimpulan tugas ini ialah memperbaiki gerak balas FLC dengan menggunakan STFPDC dan mengurangkan bilangan peraturan yang digunakan untuk mencapai perbaikian dengan penggunaan pemodelan FSCM berdasarkan ANFIS. Bagi tugas yang akan datang, ia dipertimbangkan untuk memajukan satu algoritme optimasi di mana ia mencapai pilihan terbaik bagi banjaran pengaruh yang memberi gerak balas terbaik dengan bilangan peraturan yang kurang.



ACKNOWLEDGEMENTS

I would like to thank many people who assisted me to finish the research. My appreciation and thank to my supervisor, Assoc. Prof. Dr. Tang Sai Hong, for his help and support along with duration of doing this research. I would like to provide my warm appreciation to Assoc. Prof. Dr. Mohammed Hamiruce Marhaban who has provided me with guidance and knowledge support along with the duration of doing this research.

I take this opportunity to formally thank my fellow house mates, for their help and support throughout the whole project.

I would also especially like to thank my family who has always believed in me.

Thair Sh. Mahmoud

May 2009



I certify that an Examination Committee has met on 11 May 2009 to conduct the final examination of Thair Sh. Mahmoud on his thesis entitled "An Improved Fuzzy Parallel Distributed–Like Controller for Multi-Input Multi-Output Twin Rotor System" in accordance with the Universities and University Colleges Act 1971 and the Constitution of the Universiti Putra Malaysia [P.U.(A) 15 March 1998. The Committee recommends that the student be awarded the Master of Science.

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DECLARATION

I hereby declare that the thesis is based on my original work except for quotations

and citations, which have been duly acknowledged. I also declare that it has not been

previously or concurrently submitted for any other degree at Universiti Putra

Malaysia or other institutions.

Thair Sh. Mahmoud

Date: 11 May 2009

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LIST OF ABBRIVIATIONS AND NOTATIONS

AI Artificial Intelligence

ANFIS Adaptive Neuro Fuzzy Inference System

CGA Conjugate Gradient Algorithm

CPU Central Processing Unit

D/A Digital to Analog converter

DC Direct Current

DOF Degree Of Freedom

FIS Fuzzy Inference System

FSCM Fuzzy Subtractive Method

GUI Graphic User Interface

HLA Hybrid Learning Algorithm

ITSE Integral of Time Multiplied by Square Error Criterion

LQ Linear Quadratic

MIMO Multi Input Multi Output

MLP Multi Layer Preceptron

MRAN Minimal Resource Allocating network

NARX Non-linear Auto Regressive process with eXternal input

NN Neural Networks

PDC Parallel Distributed Compensator

PID Proportional Integral Derivative

PSO Particle Swarm Optimization

RBF Radial Basis Function

RGA Real valued Genetic Algorithms

RLS Recursive Least Square

STFPDC Self Tuning Fuzzy PD-like Controller



SISO Single Input Single Output

SNN Single Neural Net

TRMS Twin Rotor MIMO System

TSK Takagi-Sugeno-Kang

UAV Unmanned air Vehicle



CHAPTER 1

INTRODUCTION

1.1 Background

In recent years, Unmanned Aerial Vehicles (UAVs) have attracted significant research interests. UAVs can do piloted aerial vehicles jobs in risky places without risking pilot's life. UAVs are very useful in doing missions in hostile environments. Unmanned helicopters have practical interesting dynamic features among autonomous flying systems. The main difficulties in designing controllers for them come from nonlinearities and couplings.

Twin Rotor MIMO System (TRMS) is a laboratory set-up design for which it has been used for control experiments, control theories developments, and applications of the autonomous helicopter. Basically, it resembles in certain aspects the behavior of helicopter. It introduces some of the platforms and control architectures, and exemplifies a high order non-linear system with significant cross-couplings from control engineering point of view. A detailed approach to the control problems connected with the TRMS involves some theoretical knowledge of laws and physics (Feedback Corp. 1998).

Modeling and controlling of TRMS are considered challenging for control community. It has attracted many researchers in the last decade. The researchers have started working of



solving TRMS control problems with conventional PID controller. It seems to be inadequate for this complex problem, and resulting to a poor performance with the non-linearity and coupling effects. PID controller performance can be improved by adjusting gains, but this still has its limitation (Juang and Liu, 2006a). Artificial intelligence has also been used to improve control performance and reduce interactions effects.

Artificial Intelligence (AI) is necessary to achieve successful embedded control systems with good control performance. So, computation resources and complexity of the used AI algorithm need to be considered for less computation resources and more flexibility in developing the embedded software's that achieve the control objectives on the systems.

1.2 Problem Statement

TRMS control has been studied in the last few years as it represents a control and modeling challenge for researchers. Fuzzy Logic Control (FLC) has been widely used with different control schemes to cope with control objectives of TRMS. It has been shown that FLC was superior to classical controllers in terms of tracking and transient response of TRMS (Islam *et al.*, 2003; Juang and Liu, 2006a). FLC has been utilized in many different hybrid schemes, and implemented with the use of classical and/or intelligent control like Genetic Algorithms (GA). As mentioned by many researchers (Jang *et al.*, 2005; Adebrez *et al.*, 2006; Rahideh *et al.*, 2006; Jang *et al.*, 2006a, 2006b;



Jang *et al.*, 2008), fuzzy logic has been proposed in different schemes with the use of Genetic Algorithms (GA) and conventional PID controller.

From the literature, it looks that the limitation of FLC is related to the difficulty in predicting changes in the operating conditions of a system and then adjusting for them. Hence, it is desirable to develop a self-tuning fuzzy controller that can improve FLC performance based on its experience, and to adapt its response in relation to variations in TRMS dynamics (Zhang and Liu, 2006).

In this work Self Tuning Fuzzy PD-like Controller (STFPDC) is proposed to make the response more robust to the interactions and non-linearity of the process. For this work, it is obvious that STFPDC scheme is using eight fuzzy reasoning blocks; each with forty-nine rules at least for this system. The eight fuzzy reasoning blocks are needed to achieve control objectives for horizontal, vertical, and the two de-coupling parts; each with two fuzzy reasoning blocks. It is considered complex control scheme with high number of the used fuzzy rules.

1.3 Aims and Objectives of the Research

The aims of this study is to improve FLC trajectory tracking performance and reduce cross coupling of TRMS by adding self tuning fuzzy inference system to the original FLC. Then, ensure achieving same control objectives with simpler control strategy in terms of less number of rules.



To achieve these aims, three specific objectives need to be achieved:

- i) Proposing Self Tuning Fuzzy PD-like Controllers to improve FLC response in terms of solving control problems and interactions between each degree of freedom.
- ii) Developing an Adaptive Neural Fuzzy Inference System (ANFIS) based on Fuzzy Subtractive Clustering Method (FSCM) from the original proposed Self Tuning Fuzzy PD-like Controllers.
- iii) Controlling TRMS with new Adaptive Neuro Fuzzy Inference Systems based Self Tuning Fuzzy PD-Like Controllers.

1.4 Scope of the Work

In this work, ANFIS-STFPDC based Fuzzy Subtractive Clustering Method (FSCM) is proposed to control TRMS. This work will neither cover optimization of achieving best parameters selections of FLC nor any method achieve best selection of range of influence in FSCM. This project will try to solve trajectory and interaction problems between yaw and pitch angles of TRMS with better response than that of FLC using STFPDC. This work will achieve same performance of STFPDC with less number of rules. In this work, all the simulations and designs are made under MATLAB®/SIMULINK®.

