

# **Implementation of Fuzzy Logic**

## **Selected Applications**

SPRA028  
January 1993



## **IMPORTANT NOTICE**

Texas Instruments Incorporated (TI) reserves the right to make changes to its products or to discontinue any semiconductor product or service without notice, and advises its customers to obtain the latest version of relevant information to verify, before placing orders, that the information being relied on is current.

TI warrants performance of its semiconductor products and related software to current specifications in accordance with TI's standard warranty. Testing and other quality control techniques are utilized to the extent TI deems necessary to support this warranty. Specific testing of all parameters of each device is not necessarily performed, except those mandated by government requirements.

Please be aware that TI products are not intended for use in life-support appliances, devices, or systems. Use of TI product in such applications requires the written approval of the appropriate TI officer. Certain applications using semiconductor devices may involve potential risks of personal injury, property damage, or loss of life. In order to minimize these risks, adequate design and operating safeguards should be provided by the customer to minimize inherent or procedural hazards. Inclusion of TI products in such applications is understood to be fully at the risk of the customer using TI devices or systems.

TI assumes no liability for applications assistance, customer product design, software performance, or infringement of patents or services described herein. Nor does TI warrant or represent that any license, either express or implied, is granted under any patent right, copyright, mask work right, or other intellectual property right of TI covering or relating to any combination, machine, or process in which such semiconductor products or services might be or are used.

## **Preface**

These three reports describe selected applications that the Texas Instruments Digital Signal Processing (DSP) Group, in conjunction with TI research groups, has developed in the emerging field of fuzzy logic.

The first paper gives an overview of the fuzzy logic theory and suggests methods of implementation that are illustrated in the next two papers.

The second paper describes the implementation of fuzzy logic as a serial algorithm on a TMS320 DSP system. The algorithm is based on parts of the theory described in the third paper.

The third paper presents a theory for a dedicated silicon-level fuzzy logic implementation.

## Contents

### What is Fuzzy Logic? An Overview of the Latest Control Methodology

*Timothy A. Adcock*

**Page**

Introduction .....	2
The Traditional Approach .....	2
Fuzzy Control .....	3
Fuzzification .....	3
Inference Rule Definition .....	4
Fuzzy Logic Rule Definition .....	4
Defuzzification .....	4
Maximum Defuzzification Method .....	5
Centroid Calculation Defuzzification Method .....	6
Summary .....	6

### Implementation of Fuzzy Logic Servo Motor Control on a Programmable TI TMS320C14 DSP

*Mathew George, Jr.*

**Page**

Abstract .....	10
Introduction .....	10
Servo Motor System (Power-14) .....	10
PID Implementation .....	11
Fuzzy Logic Theory for Servo Motor Control .....	11
Fuzzy Logic Implementation .....	12
Results .....	17
Conclusion .....	18
Appendix A: PID Code .....	19
Appendix B: Fuzzy Logic Code .....	23
References .....	34

### The Programmable Fuzzy Logic Array

*Philip Thrift*

**Page**

Introduction .....	38
Data Flow .....	38
Summary .....	41
Appendix A .....	41
References .....	42

# **What is Fuzzy Logic?**

## **An Overview of the Latest Control Methodology**

**Timothy A. Adcock**  
**Digital Signal Processing — Semiconductor Group**  
**Texas Instruments Incorporated**

## Introduction

The name “Fuzzy Logic” seems to imply an imprecise methodology that is useful only when accuracy is not necessary or important. That is what many people assume when they first hear about fuzzy logic—and understandably so. In a world increasingly manipulated by computers with their absolute “1” or “0” and “on” or “off” concepts, a term like fuzzy logic suggests inaccuracy or imprecision. Even Webster’s dictionary defines “fuzzy” as:

**fuzz·y (–ē) adj. 2. not clear, distinct, or precise; blurred**

This is not true of fuzzy logic. Fuzzy logic can address complex control problems, such as robotic arm movement, chemical or manufacturing process control, antiskid braking systems, or automobile transmission control with more precision and accuracy, in many cases, than traditional control techniques have.

Fuzzy logic was invented and named by Lotfi Zadeh, a professor at the University of California at Berkeley. Fuzzy logic is a methodology for expressing operational laws of a system in linguistic terms instead of mathematical equations. Many systems are too complex to model accurately, even with complex mathematical equations, but fuzzy logic’s linguistic terms provide a useful method for defining the operational characteristics of such a system. These linguistic terms are most often expressed in the form of logical implications, such as If – Then rules:

*If air\_temp is WARM, then set fan\_speed to MEDIUM.*

The terms WARM and MEDIUM are actually sets that define ranges of values known as membership functions. By choosing a range of values instead of a single discrete value to define the input variable “air\_temp”, you can control the output variable “fan\_speed” more precisely. Fuzzy logic controllers can often improve the performance of a control system by reducing the chance of wild functions in the output that may be caused by variations in the measured input variables.

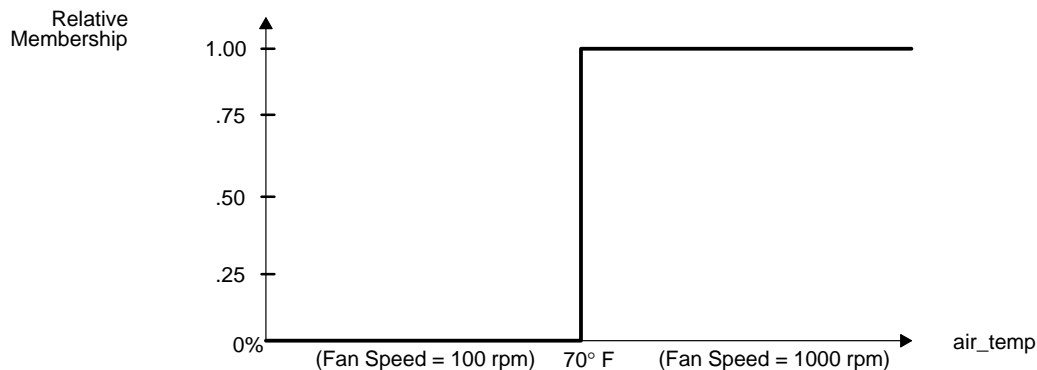
## The Traditional Approach

To illustrate the difference between fuzzy logic and the traditional approach, here is a control problem. First, consider how the traditional—often called “crisp”—controller would handle it:

*If air\_temp is  $\geq 70^\circ$  Fahrenheit, then set fan\_speed to “1000 rpm”.*  
*If air\_temp is  $< 70^\circ$  Fahrenheit, then set fan\_speed to “100 rpm”.*

A nonfuzzy, or “crisp,” controller relies on a discrete valued decision point. For this type of system, the input must reach an exact value before the control system reacts in a certain way. Even small variances in this input value may cause the output to react drastically differently. For instance, if the temperature is  $70^\circ$  or above, the first rule will set the fan\_speed to 1000 rpm. If the temperature is below  $70^\circ$ , the second rule will set the fan\_speed much lower, to 100 rpm. Figure 1 shows a diagram of this crisp valued controller.

**Figure 1. Crisp Controller**



What would happen if the temperature were  $69.5^\circ$ ? Or, more importantly, what would happen to the control system if the temperature were transitioning from below  $70^\circ$  to above  $70^\circ$ ? The temperature might even fluctuate back and forth slightly above or below  $70^\circ$  (e.g.,  $69.0^\circ$  to  $71.0^\circ$ ). This would cause the control system to alter the fan speed wildly for changes in the input variable air\_temp, although the temperature change may not be significant.

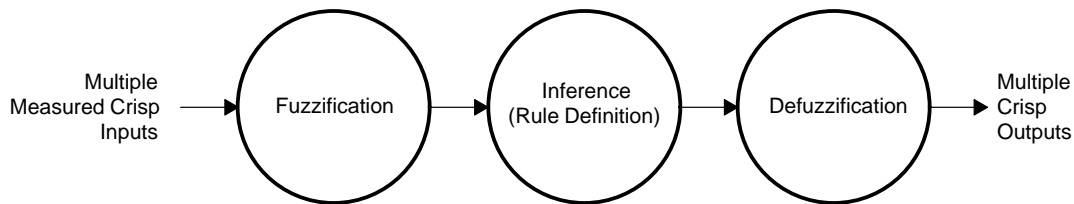
These transition points are difficult for “crisp” control systems to handle, but they are exactly where “fuzzy logic” excels.

## Fuzzy Control

Fuzzy logic is implemented in three phases (see Figure 2):

1. Fuzzification (crisp input to fuzzy set mapping).
2. Inference (fuzzy rule generation).
3. Defuzzification (fuzzy to crisp output transformation).

**Figure 2. Fuzzy Logic Phases**

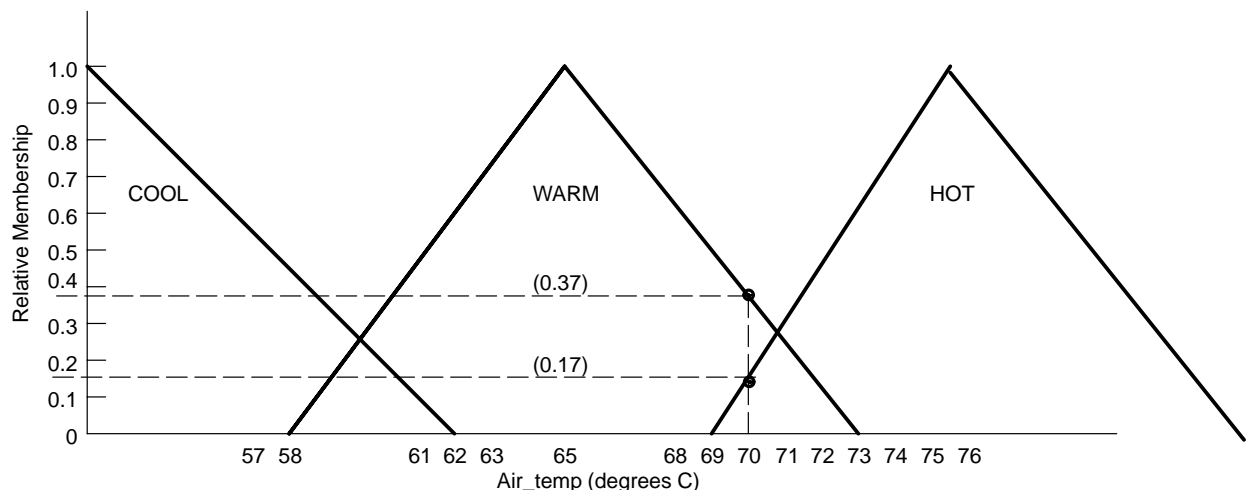


## Fuzzification

In the first fuzzy logic phase—fuzzification—actual measured input values are mapped into fuzzy membership functions. As an example, a climate-control system has been developed with fuzzy logic.

To create a climate control system, we first developed membership functions for the input variable “air\_temp”. These membership functions are defined by both a range of values and a degree of membership. In fuzzy logic, it is important to distinguish not only which membership functions a variable belongs to, but also the relative degree to which it is a member. This gives the variable a “weighted” membership in a membership function. A variable can have a weighted membership in several membership functions at the same time. The membership functions for “air\_temp” are shown in Figure 3.

**Figure 3. Air\_Temp Input Variable Membership Functions**



As shown in Figure 3, fuzzy membership functions span a range of values and can actually overlap. Three sets of membership values are defined above for the variable “air\_temp”. They are COOL, WARM, and HOT. The degree of membership is found by finding the intersection point of a distinct input value on the horizontal axis with the line defining one or more fuzzy membership functions. This intersection point is assigned a corresponding value on the vertical axis to define the relative membership in a set for an actual measured input value. Notice that when “air\_temp” is at a particular value, it may be contained in one or more fuzzy sets. For instance, at 70°, “air\_temp” is a member of the function HOT with a relative membership of 0.17. It is also a member of the function WARM with a relative membership of 0.37. Unlike a crisp system in which a value either is or is not a member of a function, a fuzzy logic system can take action based not only on membership in a fuzzy set, but also on the degree to which a variable is included in a membership function. In this case, because “air\_temp” at 70° is more WARM (0.37) than it is HOT (0.17), the controller will take that into account when defining what output action to take.

### **Inference Rule Definition**

Once membership functions have been defined for input and output variables, a control rule base can be developed to relate the output actions of the controller to the observed inputs. This phase is known as the inference, or rule definition portion, of fuzzy logic. Any number of rules can be created to define the actions of the fuzzy controller. Some examples are shown below.

### **Fuzzy Logic Rule Definition**

If air\_temp is COOL, then set fan\_speed to SLOW.  
If air\_temp is HOT, then set fan\_speed to FAST.  
If air\_temp is WARM, then set fan\_speed to MEDIUM.

These If-Then rules can relate multiple input and output variables. Because the rules are based on word descriptions instead of mathematical definitions, any relationship that can be described with linguistic terms can typically be defined by a fuzzy logic controller. This means that even nonlinear systems can be described and easily controlled with a fuzzy logic controller. In addition, since variables have weighted memberships—in particular membership functions—the rules that are composed of these variables are weighted as well. This means that different rules have different impacts on the controller, according to the measured input variable. For a multiple-input/multiple-output system with many defining rules, a wild fluctuation in any single input will be tempered by these rule weightings. Because of this, fuzzy logic systems are very robust and often allow many rules to be removed or altered without significantly impacting the controller.

### **Defuzzification**

After the fuzzy logic controller evaluates inputs and applies them to the rule base, it must generate a usable output to the system it is controlling. This may mean setting a voltage or current to a particular value to control the speed of a fan in the example above, or it may mean defining the optimal speed of a robotic arm as it nears its target. The fuzzy logic controller must convert its internal fuzzy output variables into crisp values that can actually be used by the controlled system. You can perform this portion of the fuzzy control algorithm, known as defuzzification, in several ways. Two of the most common methods are:

- maximum defuzzification method (page 5).
- centroid calculation defuzzification method (page 6).

Remember from fuzzification that in mapping input variables to membership functions, a particular measured value of the input variable determined the relative membership of that input variable in an input membership function. To determine the mapping of output variables to their corresponding output membership functions, the weighted input membership function and corresponding rule base determine the relative membership in the output function. Whatever relative membership was given to the input variable will also be given to the output variable, as assigned by its corresponding rule. For air\_temp = 70°, the output variables are assigned a value that corresponds to the input value shown in Table 1.



**Table 1. Fan\_Speed (Membership Function Relative Membership)**

Input Variable	Defining Rules	Output Variable
air_temp (WARM) = 0.37	If air_temp = WARM, then set fan_speed to MEDIUM	fan_speed (MED) = 0.37
air_temp (HOT) = 0.17	If air_temp = HOT, then set fan_speed to FAST	fan_speed (FAST) = 0.17

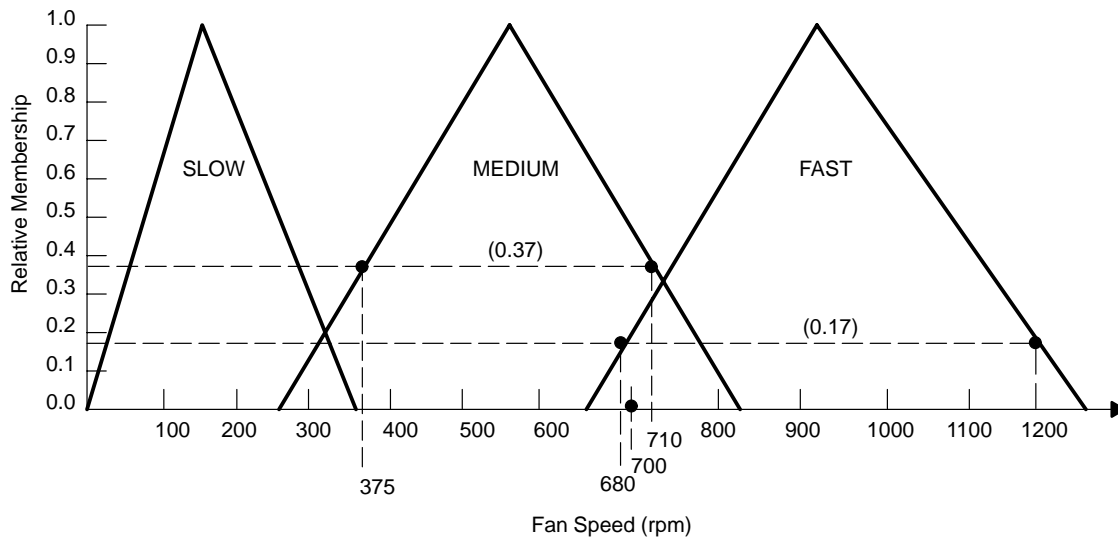
The output variable fan\_speed is given the same relative mapping as the input variable air\_temp that is defined by a particular rule.

Figure 4 illustrates the output variable membership functions. In this case, a distinct value on the horizontal axis is defined by the relative membership on the vertical axis. To create the actual crisp output value for the controller system output, membership functions are used with the output variable. In Table 1, the input value air\_temp = 70 resulted in two weightings for fan\_speed:

- Fan\_speed = 0.37 was assigned to the output membership function MEDIUM.
- Fan\_speed = 0.17 was assigned to the output membership function FAST.

As shown in Figure 4, the actual output value is determined by beginning at the weighting factor on the vertical axis and moving horizontally until an intersection point is reached on the lines defining its associated membership function. This intersection point is then transposed to the horizontal axis to determine the crisp output value.

**Figure 4. Fan\_Speed Output Variable Membership Functions**



### Maximum Defuzzification Method

One method of defuzzification is known as the maximum method. In this method, if more than one rule is active, the maximum relative membership is used to determine the output value. In the above example, making air\_temp = 70° created two possible values for fan\_speed:

$$\begin{aligned} fan\_speed &= 0.37 \\ fan\_speed &= 0.17 \end{aligned}$$

The maximum defuzzification method provides a single output by choosing the active rule with the greatest relative membership value in the output membership function. In the preceding example, the following rule is chosen because it has the highest membership value for fan\_speed.

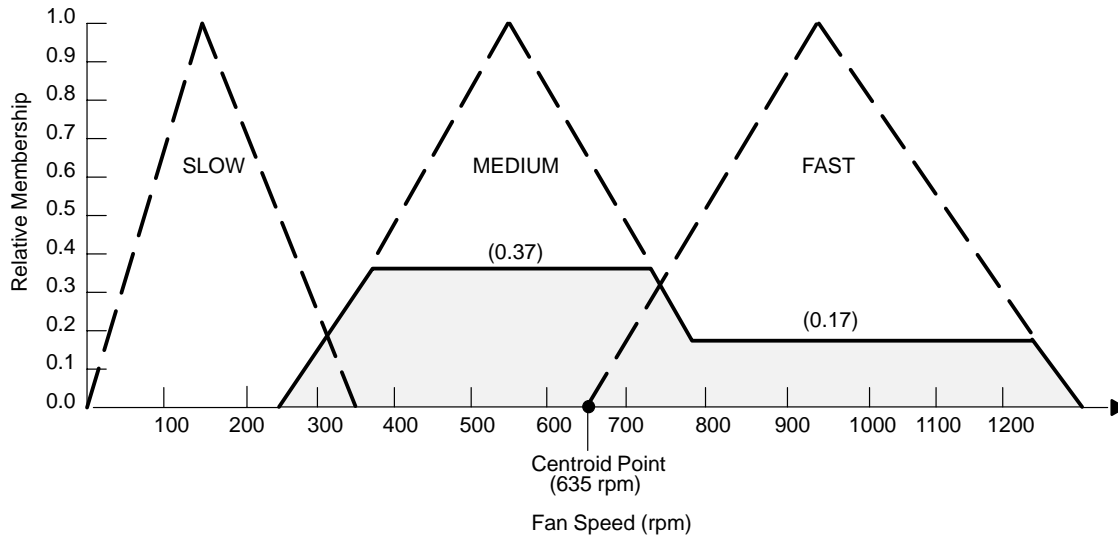
$$\text{If air\_temp} = \text{WARM, then set fan\_speed to MEDIUM. Fan\_speed (MED)} = 0.37$$

The value 0.37 on the vertical axis intersects the membership function MEDIUM at two points—one on the positive slope (at 375 rpm) and one on the negative slope (at 710 rpm) of the function. The two points represent two possible solutions that must be resolved.

## Centroid Calculation Defuzzification Method

Another method for calculating the output value is the centroid method. In this method, a weighted average of all the active rules determines an output by summing all of the applicable output variables over their relative membership values. Although this method is more computationally intensive, it creates a distinct output value based on the relative memberships of all of the active rules that apply (see Figure 5). This method eliminates the problem of multiple solutions observed with the maximum method. A processor architecture with a hardware multiply-accumulate feature like that of the TMS320 DSP family excels at this method.

**Figure 5. Fan\_Speed Output Centroid Calculation**



## Summary

By using fuzzy logic, you can simplify complex control problems that once required a high-powered microprocessor to execute in real time; you can now execute them on a low-cost Texas Instruments TMS320 DSP or TMS370 microprocessor. The following application note shows the benefits of controlling a simple DC motor with fuzzy logic using a TMS320C14 digital signal processor.

# **Implementation of Fuzzy Logic Servo Motor Control on a Programmable Texas Instruments TMS320C14 DSP**

**Mathew George, Jr.  
Digital Signal Processing — Semiconductor Group  
Texas Instruments Incorporated**

## Abstract

This paper describes the implementation of a fuzzy logic compensator on a Texas Instruments TMS320C14 DSP-based servo motor control development system. The system contains a real motor that is controlled by the programmable DSP. An on-chip debugger and servo motor program allowed both simple code modification and interactive control of the motor. A fuzzy logic algorithm was directly substituted for the original PID algorithm; this resulted in comparable motor response and algorithm performance. This implementation proves the feasibility of real-time fuzzy logic-based servo motor control on a real system.

## Introduction

Fuzzy logic is relatively new theory. Most of the readily available hands-on fuzzy logic system examples have been software simulations or bulky real systems. However, a TI commercial microprocessor (the TMS320C14) can serve as a simple, real-time, real-system platform for applying and investigating fuzzy logic. This facilitates both the understanding and implementation of fuzzy logic as a real-time programmable solution for the general engineering public.

Servo motor control is a viable and useful implementation. A programmable PID motor control board uses a Texas Instruments TMS320C14 chip and has an actual motor whose performance can be observed. Faster and newer parts are available, but the 'C14 is optimized for motor control with such features as on-board pulse-width modulation (PWM) generation capabilities. The board also has on-chip debugger code and interactive PID control code. The PID compensator code (written in TMS320 assembly language, which is upward-compatible with code executed by such newer fixed-point TI DSPs as the TMS320C25 or TMS320C50) uses position and velocity of the motor for inputs and motor input current as the output. The control code will allow the operator to interactively change such values as servo position and velocity and to monitor position error. Modifying the code for fuzzy logic required replacing the PID compensator section of the code with a fuzzy logic compensator.

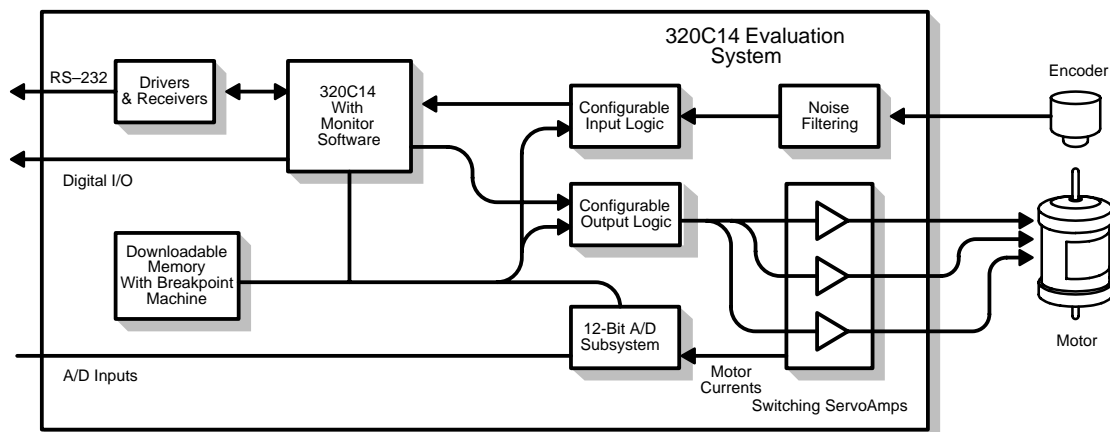
The membership function for the compensator defines the error between present motor position and the desired (command) position of the controller. Five linguistic variables characterize the function: negative medium, negative small, zero, positive small, and positive medium. The function is represented as overlapping isosceles right triangles for ease of fuzzification. Various rules for an inverted pendulum control system (ball and stick) are explained in [1]. These same eleven rules were used for servo motor control.

The algorithm's three sections are implemented as a series of software loops: fuzzification (i.e., input evaluation), fuzzy inference (rule contribution that uses a table look-up), and defuzzification (using center-of-gravity method). It is based on an algorithm developed for [4]. Each section uses various arrays that are modified in that section. The sizes of these loops are directly proportional to the number of inputs, outputs, and rules used in the system.

## Servo Motor System (Power-14)

The heart of the Power-14 board is a Texas Instruments TMS320P14 chip (one-time programmable TMS320C14). (See Figure 1).

Figure 1. Power 14 System



You communicate to the board through an RS-232 serial port connection by using standard terminal or terminal emulation software (such as Procomm). The 'P14 peripherals are optimized for control applications. An event manager can be operated in a PWM (pulse-width modulation) mode that is ideal for motor control. Monitor code is loaded into 'P14 external memory and run, which provides a command line debugger. The debugger has all standard debugger functions, such as memory dumps and modification, stepping through code, breakpoints, etc. The monitor can be used to load and run the servo motor program with PID compensator. The program is interactive and lets you control the motor from the keyboard. Position, velocity, and the PID values can be set. Data acquisition functions allow an ASCII text input stimulus table to be loaded onto the Power-14, an acquisition run to be executed, and the resulting ASCII output table to be written to a PC file for graphing. The rest of the board contains support circuitry: amplifiers for the motor and a serial port interface. An encoder on the motor is used as a position sensor for a compensator input. The velocity is found by executing a back-difference. Note that there is no separate velocity sensor.

## PID Implementation

The source code for this system implements the PID compensator in one file (See Appendix A). The algorithm is a direct implementation of the PID equation. The Proportional, Integral, and Differential variables are derived from the motor encoder sensor detecting position. On each cycle, the present position is taken from the encoder and stored in Position. The error is found by subtracting Position from DesiredPosition and storing it in ErrNow. Thus, ErrNow is the position input for the proportional section of the PID. A back-difference is then taken with ErrNow and ErrLast (the error in the previous cycle) to give ErrDiff. ErrDiff is an approximation of the velocity and is therefore the second input (Differential) of the PID. The Integral is found by adding the ErrNow value and storing it in Kintegrator. The following equation then holds the compensator output:

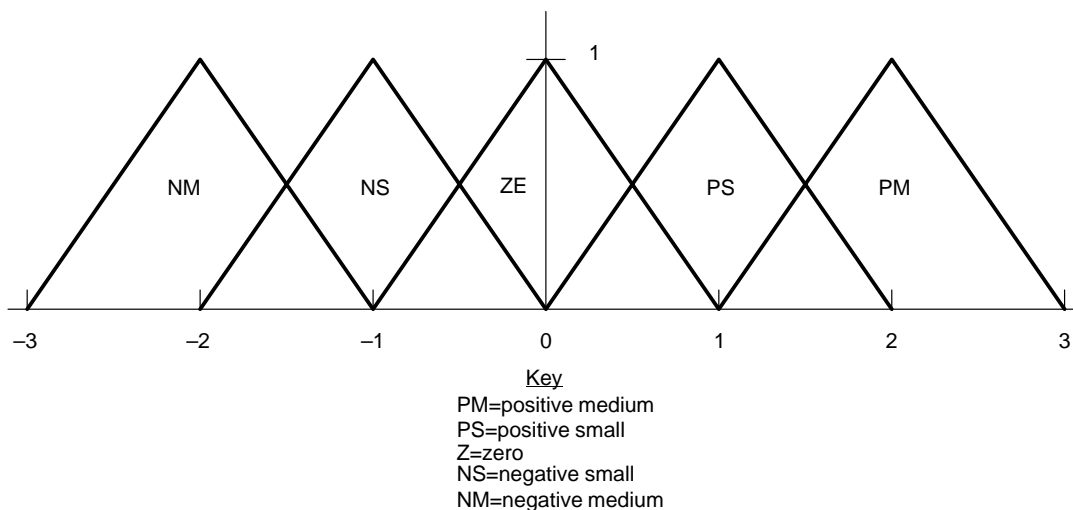
$$PWM = 9 * P * ErrNow + 9 * I * Kintegrator + 9 * D * ErrDiff \quad (1)$$

The value is scaled for the PWM mode and stored in NewServo. Thus, for the compensator output, the actual motor current input value is converted to a PWM value for implementation. The PWM output is then sent to power amplifiers that finally drive the motor. The PWM frequency can be controlled from the PID program.

## Fuzzy Logic Theory for Servo Motor Control

The membership function for this system is simple. Five fuzzy logic ranges (linguistic variables) were chosen to remain consistent with [1] and used for both inputs. Sets of values are represented as overlapping isosceles right triangles (Figure 2).

Figure 2. Membership Function



Two input variables, position (Theta) and velocity (dTheta) of the motor, are used in this fuzzy logic system. One output variable, motor current, is used, which will be proportional to the PWM output that is actually written. The rules for the compensator as mentioned were taken from [1]. Thus Theta, dTheta, and motor current operate according to the following "if a and b, then c" rules, as shown in Table 1.

**Table 1. List of Rules**

If Theta =	And dTheta =	Then Motor Current =
Z	Z	Z
PS	Z	NS
PM	2	NM
NS	Z	PS
NM	Z	PM
Z	NS	PS
Z	NM	PM
Z	PS	NS
2	PM	NM
PS	NS	Z
NS	PS	Z

These rules can then be indexed according to the scale shown in Figure 2. The mapping seen in Table 2 will be used in TMS320 programming.

**Table 2. Indexed List of Rules**

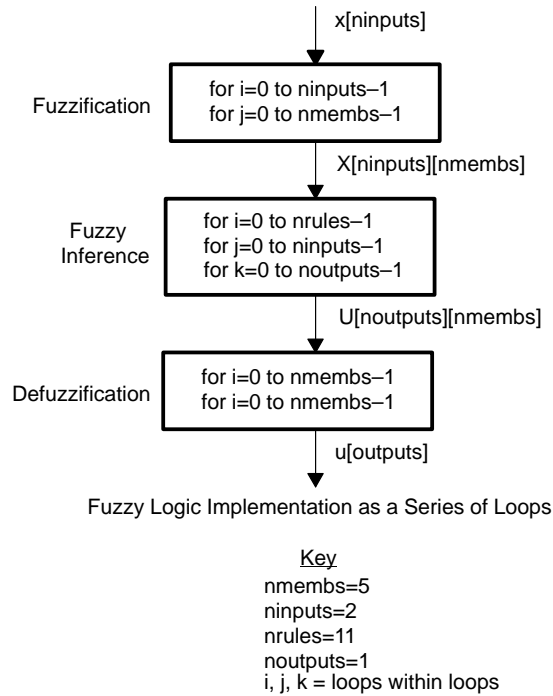
If Theta =	And dTheta =	Then Motor Current =
0	0	0
1	0	-1
2	0	-2
-1	0	1
-2	0	2
0	-1	1
0	-2	2
0	1	-1
0	1	-2
1	-1	0
-1	1	0

The defuzzification is done by the center-of-gravity method.

## Fuzzy Logic Implementation

Arrays are used in the fuzzy logic calculations and modified in the various loops that implement the compensator. Appendix B lists the TMS320C14 code. Figure 3 shows the three sections of the compensator: fuzzification, fuzzy inference, and defuzzification. Notation for the arrays follows C language standard, with subscripts from 0 to n-1. The 'C14 algorithm is based on an algorithm developed for [4]. The figure key summarizes the values of the system that will be used in the examples in this section.

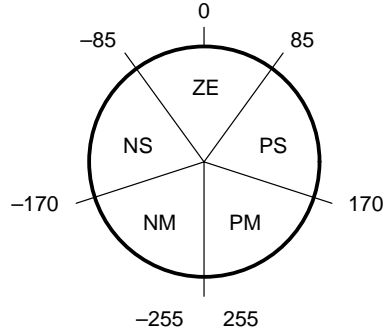
**Figure 3. List of Arrays**



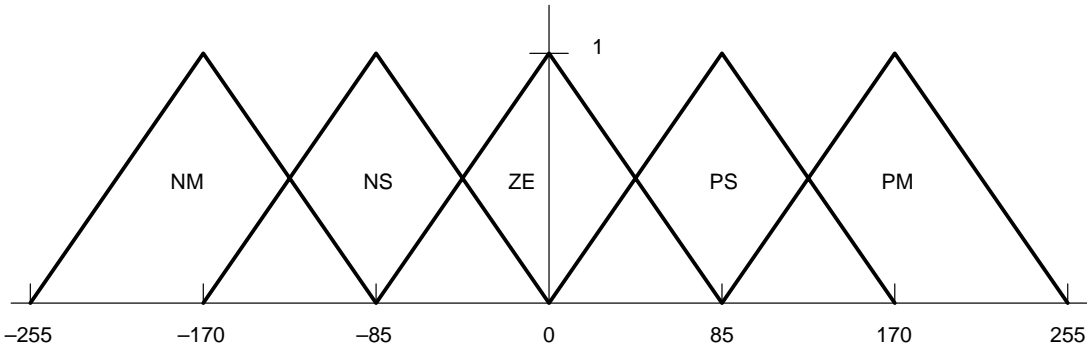
Note which arrays are modified in each section and their size boundaries. They are discussed later in more detail. The two inputs seen in  $x[ninputs]$  are position (found from the encoder) and velocity (found from an approximation of the derivative by taking the back-difference of the position). These are the Theta and dTheta variables described previously. As the compensator code begins, the position and velocity inputs (ErrNow and ErrDiff, respectively) are copied to the array  $X[ninputs]$ . The position is mapped so that one rotation of the motor ranges from  $-255$  to  $+255$  (See Figure 4 a). This relationship is mapped onto the x-axis of the membership function and thus fuzzifies the position of the motor (Figure 4 b). Note that this figure is not drawn to scale.

**Figure 4. Motor Shaft/Membership Function Mapping**

**a. Motor Range**

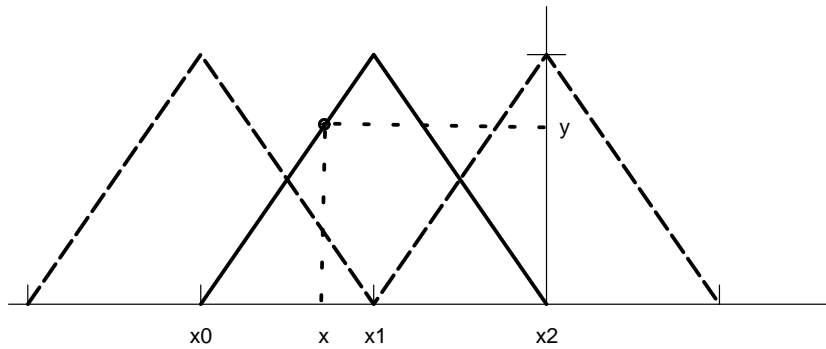


**b. X-Axis Map**



In the fuzzification loop, the degree of membership of each input relative to the input membership function is evaluated and written in the array  $X[ninputs][nmembs]$ . The value for a particular linguistic variable is the  $y$  value of the triangle for a particular value of  $x$  (See Figure 5).

**Figure 5. Analysis for One Linguistic Variable**



The  $y$  value is found by using the simple algebraic equation for a line ( $y=mx+b$ ). This equation may be geometrically reduced to one of the two following equations, depending on which side of the triangle the value of  $x$  lies:

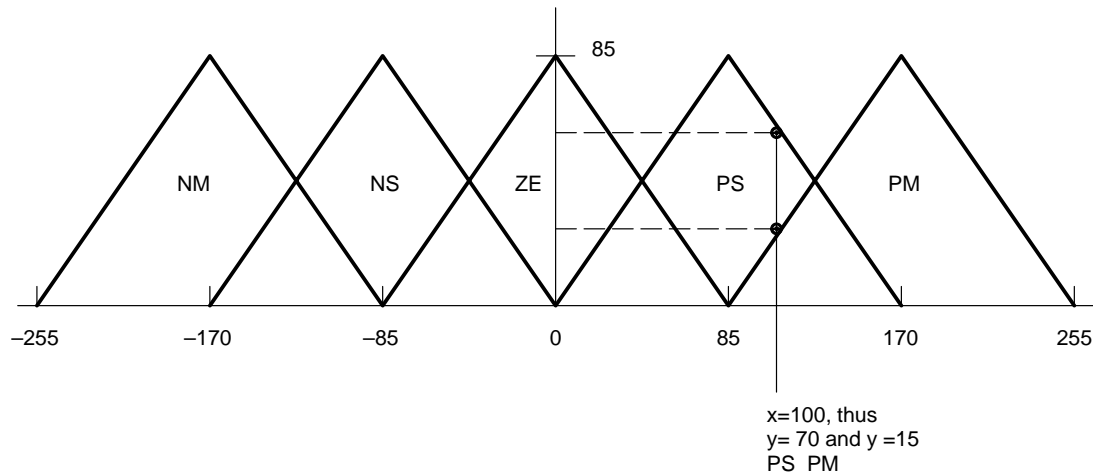
$$\begin{aligned} \text{if } \{x : x_0 < x < x_1\} \text{ then } y &= (x - x_0) / (x_1 - x_0) \text{ else} \\ \text{if } \{x : x_1 < x < x_2\} \text{ then } y &= (x_2 - x) / (x_2 - x_1) \end{aligned} \quad (2)$$



The division needed is costly on most microprocessors, usually requiring at least a number of cycles equivalent to the number of bits of the number being divided. But if the slope  $m$  is made equal to 1 by causing the elements of the membership functions to be isosceles right triangles, the equation can be reduced to  $y=x-x_0$ . This translates into a simple one-cycle subtraction. Note that forcing the elements of the membership functions to be isosceles right triangles also forces the peak of the membership function to no longer be 1. Rather the peak value  $=x_1-x_0=x_2-x_1=85$  (as seen in Example 1). This action also eliminates the need for using a  $Q$  format [6] to represent the fractional values from Equation 2 if the triangle were not isosceles.

Example 1 demonstrates this fuzzification calculation. If the position input value were 100 (i.e.,  $x[0]=100$ ), it would have nonzero degrees of membership in the PS and PM linguistic variables. You can also see this in Figure 5. To calculate the actual degree of membership value, the value for  $x$  is plugged into Equation 2 for both PS and PM boundaries. Thus, in PS the contribution is 70, while in PM it is 15. The rest of the linguistic variables are 0 because there is no contribution. (Of course, in software, all linguistic variables must be evaluated.) For this example,  $X[0][nmembs]=[0, 0, 0, 70, 15]$ .

### Example 1. Fuzzification Example



The next step involves fuzzy inference. In this loop, the maximum and minimum functions, as explained in [2], are implemented. In the actual code, the array indexing of the membership values is made nonnegative by adding a bias of three. Therefore, instead of NM to PM being indexed from  $-2$  to  $+2$ , as seen in Figure 4, they are indexed from 0 to 4. Table 3 shows how the 11 rules are indexed and reindexed in the array `RULE_TABLE[rule][input+output]`.

**Table 3. Original and Reindexed Table of Rules**

Original Index			Reindexed		
$x(0)$	$x(1)$	$u(0)$	$x(0)$	$x(1)$	$u(0)$
0	0	0	2	2	2
1	0	-1	3	2	1
2	0	-2	4	2	0
-1	0	1	1	2	3
-2	0	2	0	2	4
0	-1	1	2	1	3
0	-2	2	2	0	4
0	1	-1	2	3	1
0	1	-2	2	4	0
1	-1	0	3	1	2
-1	1	0	1	3	2

Some explanation is required for this decoding. The indexed rules match the explicit rules as described in Table 1. Also, the values given by accessing the RULE\_TABLE are limited to the values indexed by nmembs. This characteristic is heavily used in the index manipulation and allows nmemb to be interchanged with RULE\_TABLE[rule][input+output] in the appropriate parts of the algorithm.

To find the minimum value of X[ninputs][nmembs] decoded from the rule table inputs and stored in minZ (which is initialized to the maximum y value—in this case, 85), the equation is:

$$\min Z = \min (X[\text{input}] [\text{RULE\_TABLE} [\text{rule}] [\text{input}]]) \quad (3)$$

(Since noutputs=1, the loop is simplified, and minZ does not need to be the general case array minZ[nrules]). Note that only the first two columns (the input columns) of RULE\_TABLE are used in this part of the fuzzy inference section.

Then, for each rule, the max is taken of the output value U[nmembs], which is initialized to 0. The general case U[output][ nmembs] is simplified because only one output is decoded from the rule table outputs and minZ. This equation can be summarized as:

$$\begin{aligned} U[\text{nmembs}] &= U[\text{RULE\_TABLE}[\text{rule}] [\text{ninput} + \text{output}]] \\ &= \max (U[\text{RULE\_TABLE}[\text{rule}] [\text{ninput} + \text{output}]], \min Z) \end{aligned} \quad (4)$$

Note that in this section only the last column (the output column) is used. The following equation summarizes the minimum and maximum functions that are executed for each rule to result in the array U[nmembs] by plugging Equation 3 into Equation 4:

$$\begin{aligned} U[\text{nmembs}] &= \max \{ U[\text{RULE\_TABLE}[\text{rule}] [\text{ninput} + \text{output}]] , \\ &\quad \min (X[\text{input}] [\text{RULE\_TABLE}[\text{rule}] [\text{input}]] ) \} \end{aligned} \quad (5)$$

The following example illustrates the fuzzy inference section. One cycle of the loop for rule=0 is shown. For the input array:

$$\begin{aligned} X[\text{ninputs}] [\text{nmembs}] &= \begin{vmatrix} 0 & 0 & 70 & 15 & 0 \\ 0 & 60 & 25 & 0 & 0 \end{vmatrix} \\ &= \begin{vmatrix} 0 & 0 & 70 & 15 & 0 \\ 0 & 60 & 25 & 0 & 0 \end{vmatrix} \end{aligned} \quad (6)$$

and for rule=0, input=0, and input=1, plug into Equation 3:

$$\begin{aligned} \min Z &= \min \begin{vmatrix} X[0] [2] \\ X[1] [2] \end{vmatrix} = \min \begin{vmatrix} 70 \\ 25 \end{vmatrix} = 25 \end{aligned} \quad (7)$$

then for the max rule=0, input=0, and input=1, plug into Equation 4:

$$U[nmemb] = U[2] = \max \left| \begin{array}{c} U[2] \\ \min Z \end{array} \right| = \max \left| \begin{array}{c} 0 \\ 25 \end{array} \right| = 25 \quad (8)$$

This process continues for the list of 11 rules so that the array U[nmemb] contains the maximum values of the min—i.e., the contribution of each rule to the inference.

The final step involves defuzzifying the U[noutputs][nmemb] array. Since noutputs=1, U simplifies to U[nmemb]. The U[nmemb] array now has five values in it for its corresponding five positions. The center-of-gravity calculation is done with two loops. The first finds the numerator by using the multiplier to weight U[nmemb] by its position. The second loop finds the denominator by summing the position. The inevitable 16-bit divide loop then finds the output value u[output]. This u[output] represents the motor current mentioned in *Fuzzy Logic Theory for Servo Motor Control* (page 11). The divide operation is done on the 'C14 by a 16-cycle loop. This value is then scaled for the output and written to the NewServo memory location that sends it to the PWM generator.

As an example of defuzzification,

if U[nmemb] = [0 15 70 35 0],

then the output is:

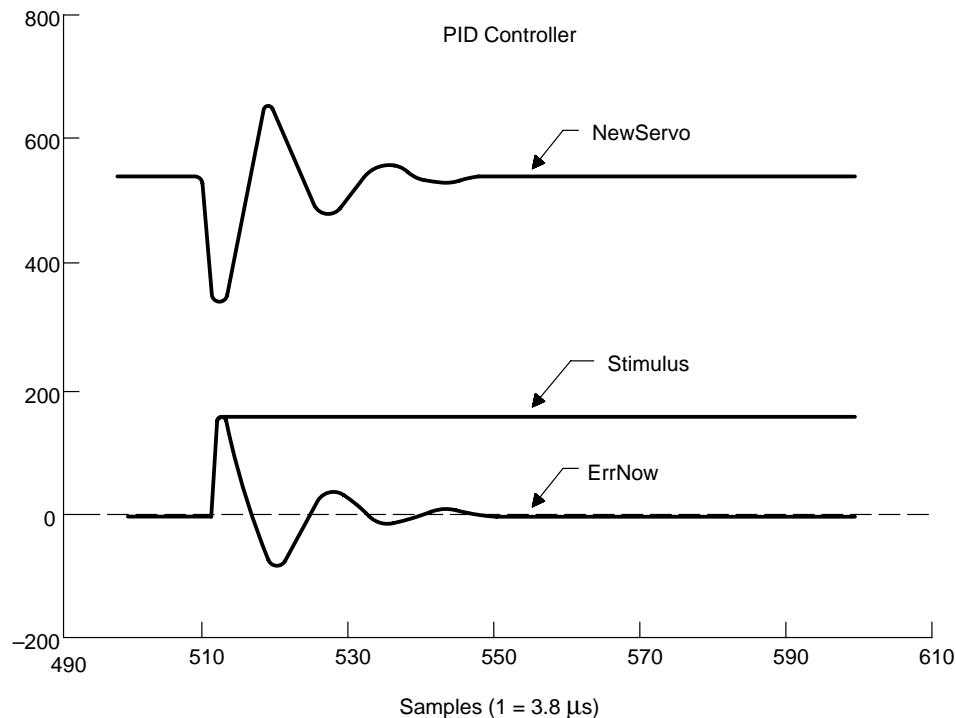
$$\begin{aligned} U[\text{output}] &= 0*(-170) + 15*(-85) + 70*(0) + 35*(85) + 0*(170) \\ &= 14.17 \end{aligned}$$

These loops thus evaluate the control input necessary for the servo motor.

## Results

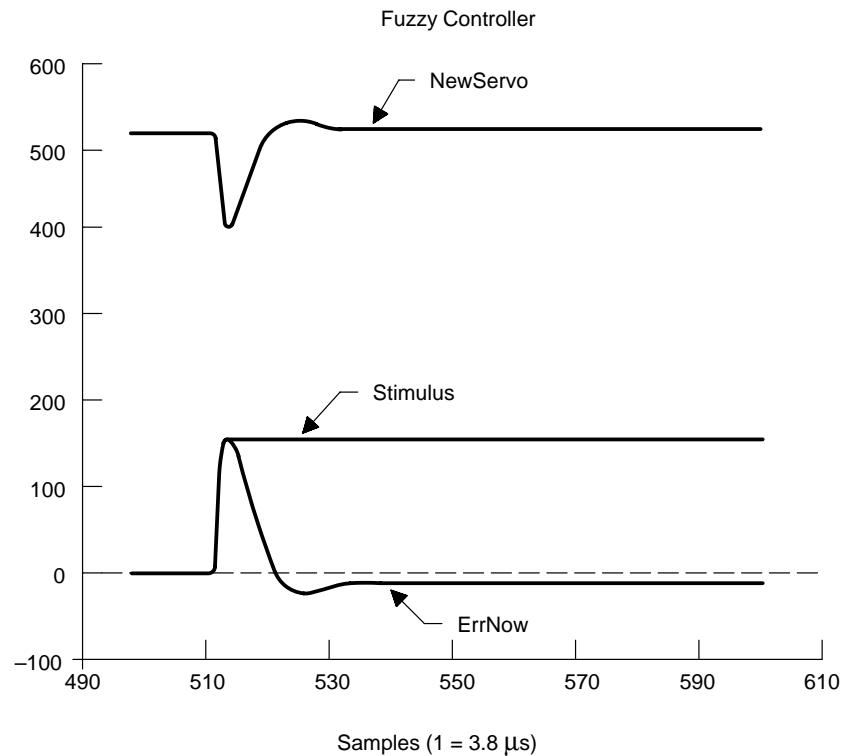
The fuzzy logic code with two inputs, eleven rules, one output, and five linguistic variables requires about 2000 instruction cycles to execute on the TMS320C14. This means a 400-μs period (2.5-kHz frequency) because one instruction cycle is 200 nanoseconds on the 'C14. The update period of the motor is 3.8 ms (263-Hz frequency), so the fuzzy code is obviously adequate (for comparison, the PID code execution required nine μs (111-kHz frequency), for the update period). Figure 6 shows the PID performance for a step with the classical control overshoot and settling times.

**Figure 6. PID Controller**



The fuzzy logic curve (Figure 7) is smoother than the PID controller curve but doesn't go to zero.

**Figure 7. Fuzzy Controller**



The update period is adequate for applications such as servo motor, robotics, motion control, and automotive control. Better performance may be needed for such applications as hard-disk drives. Later-generation digital signal processors, such as the TMS320C5x, operate at up to 25 nanoseconds with much more efficient instruction code.

## Conclusion

The final fuzzy logic system behaved favorably when compared to the conventional PID system. The system proved the feasibility of implementing a real-time fuzzy logic servo motor control. Proof of the fuzzy control was shown by positioning the motor spindle with an error outside the membership function, thus causing the control to desist. Further development could include a graphics display and the ability to vary the rules. The TMS320C14 board fits in a 12 x 8 x 6-inch suitcase conveniently and requires only an AC power supply and an RS-232 keyboard connection. This product easily demonstrates real-system fuzzy logic control on a microprocessor.

Fuzzy logic has great potential as a programmable solution for general engineering. For applications where performance is a priority, a hard-wired silicon solution (which may even configure as a microprocessor peripheral) based on [4] is being developed. The programmable solution may assist in the transition to this hard-wired option, depending on the software/hardware tradeoffs.

## APPENDIX A PID Code

```
.title "Servo compensator" $Revision: 3.4 $"
; *****
; $Header:: C:/src/c14/ps/vcs/comp.s_v 3.4 01 Oct 1991 17:24:18 "$
;
; $config$="/T8 /K! /L;* .ref.def /R:*-/B80"
; !config!="Mcomp.s"
;
; !NAME!
;         comp.s
;
; !PATHS!
;         modules
;
;         !0!
; DESCRIPTION
;         Servo compensator – uses a PID algorithm
;
; PRINCIPLE AUTHORS:
;         Dave Sewhuk
;
; CREATION DATE:
;         December 23, 1990 22:54:08
;
; COPYRIGHT NOTICE:
;         (C)Copyright 1990 Teknic Inc. All rights reserved.
;
; !end!
; *****
; *****
; HEADERS UTILIZED
;         .include "macrodef.inc"
;         .include "c14io.inc"
;
; *****
; *****
; !NAME!
;         comp.s
;
; !PATHS!
;         Exported Variables
; !0!
;
; .def  PwmChannel,PwmPeriod
; .def  NewServo,ErrDiff,Kintegrator
; .def  ErrNow,ErrLast,DesiredPosition
;
; .bss  PwmChannel,1           ; Current PWM channel
; .bss  PwmPeriod,1           ; Current period of PWM channels
; .bss  NewServo,1            ; New servo value
; .bss  ErrDiff,1             ; Error difference
; .bss  Kintegrator,1         ; Error integrator
; .bss  ErrNow,1              ; Current error function
; .bss  ErrLast,1             ; Last error function result
; .bss  DesiredPosition,1     ; Servo to this position
```

```

; !end!
; *****
; *****
; !NAME!
;         comp.s
;
; !PATHS!
;         Local Variables
; !0!

; !end!
; *****
; *****
;         .text
;         .def  servo_ISR
; !NAME!
servo_ISR:
;
; !PATHS!
;         functions\all
;         functions\interrupt
;         modules\utils.s
; !0!
; DESCRIPTION:
;
;         This is the servo compensator. The distribution algorithm is
;         the PID format. The compensator performs the following function
;          $PWM = 9 * P * error + SUM(9 * e_k) * I + 9 * (err \text{ diff}) * D$ 
;         with the PWM output clipped to timer hardware limits.
;
; RETURNS:
;         No changes to foreground environment.
;
; STACK LEVELS:
;         0
;
; EXTERNAL DATA REFERENCES
;         .ref  Position          ; Position from feedback ISR
;         .ref  ONE,CONSTFFFF    ; Useful constants
;         .ref  Pgain,Igain,Dgain ; Servo constants from user interface
;         .ref  ISR_TMP          ; Scratch
;
; *****
; !skip start!

SOVM
; Saturate math
LAC ErrNow          ; Save old error
SACL                ErrLast

```

```

        ZALH        DesiredPosition
; Calc new error
        SUBH        Position
        SACH        ErrNow
;
; Run PID algorithm on new sample
;
        LAC ErrNow          ; Update error difference
        SUB ErrLast
        SACL        ErrDiff

        ZALH        ErrDiff      ; Scale up value a bit/saturation
        ADDH        ErrDiff      ; *2
        ADDH        ErrDiff      ; *3
        ADDH        ErrDiff      ; *4
        ADDH        ErrDiff      ; *5
        ADDH        ErrDiff      ; *6
        ADDH        ErrDiff      ; *7
        ADDH        ErrDiff      ; *8
        ADDH        ErrDiff      ; *9
        SACH        ErrDiff

        ZALH        ErrNow      ; Scale up/saturation
        ADDH        ErrNow      ; *2
        ADDH        ErrNow      ; *3
        ADDH        ErrNow      ; *4
        ADDH        ErrNow      ; *5
        ADDH        ErrNow      ; *6
        ADDH        ErrNow      ; *7
        ADDH        ErrNow      ; *8
        ADDH        ErrNow      ; *9
        SACH        ISR_TMP
; Saved scaled up error

        ZALH        Kintegrator
; Update integrator
        ADDH        ErrNow
        SACH        Kintegrator

        LT Pgain          ; Do proportional part
        MPY ISR_TMP
        PAC

        LT Igain          ; Do integral part
        MPY Kintegrator

        LTA Dgain         ; Do differential part
        MPY ErrDiff
        APAC
        SACH        NewServo      ; Save result
;
; Check the answer so that it fits within the limits of the timer's
; period: [0..PwmPeriod*4)
;
        LAC PwmPeriod,1      ; Check positive: 2*Period-4
        SUB ONE,3
        SUB NewServo          ; - value

```

```

        BGZ pwmisr20
        LAC PwmPeriod,1                ; Oops, pin at biggest
        SUB ONE,3
        SACL          NewServo
        B    pwmisrSet
pwmisr20:                                ; Negative OK?
        LAC NewServo
        ADD PwmPeriod,1                ; Add largest offset
        BGEZ          pwmisrSet
        ZAC
        SUB PwmPeriod,1                ; Make largest negative
        SACL          NewServo
;
; Set Output PWM to new value
;
pwmisrSet:
        LAC PwmPeriod,1                ; Get center period value
        ADD NewServo                    ; Add newly computed value
        SACL          NewServo
        .if    ChipV1R1                ; PWM bug workaround
        .ref    CONST3
        SUB CONST3
        BLEZ          pwmisr10
; No fixes needed
        LAC NewServo                    ; Lower 2 bits set?
        AND CONST3
        SUB CONST3
        BNZ pwmisr10                    ; OK, No fixes needed
        LACK          4
        ADD NewServo
        SACL          NewServo
pwmisr10:
        .endif
        LACK          ActionBank        ; Set bank for timers
        SACL          ISR_TMP
        OUT ISR_TMP,BSR
        OUT NewServo,ACT0                ; Output channel 1 timer
        LAC PwmPeriod,2                ; Calculate complimentary output
        SUB ONE,2
        SUB NewServo
        SACL          NewServo
        OUT NewServo,ACT1                ; Set complimentary channel 2 timer
        RET                                ; All done.. back to the ISR already
                                           ; in progress.
; !skip end!
; !END!
;=====
; END OF FILE
;=====
        .end

```



## APPENDIX B Fuzzy Logic Code

```
.title "Fuzzy Servo compensator   Revision: 3.4 $"
; *****
; $Header:: C:/src/c14/ps/vcs/comp.s_v 3.4 01 Oct 1991 17:24:18 "$
;
; $config$="T8 /K! /L;* .ref.def /R:*- /B80"
; !config!="Mcomp.s"
;
; !NAME!
;   compfuz.s
;
; !PATHS!
;       modules
;
;                               !0!
; DESCRIPTION
;   Servo compensator – uses a fuzzy algorithm (modifying PID code)
;
; PRINCIPAL AUTHORS:
;       Dave Sewhuk
;       Joe George w/ Fuzzy based on C++ code by Phillip Thrift
; CREATION DATE:
;       December 23, 1990 22:54:08
;       Jan. 1992
; COPYRIGHT NOTICE:
;       (C) Copyright 1990 Teknic Inc. All rights reserved.
;       (C) Copyright 1992 Texas Instruments Incorporated. All rights reserved.
; !end!
; *****

; *****
; HEADERS UTILIZED

        .include "macrodef.inc"
        .include "c14io.inc"
;
; *****

; *****
; !NAME!
;       comp.s
;
; !PATHS!
;       Exported Variables
; !0!

        .def  PwmChannel,PwmPeriod
        .def  NewServo,ErrDiff,Kintegrator
        .def  ErrNow,ErrLast,DesiredPosition
        .def  ninput,noutput,triangle,nmemb
        .def  NM,NS,ZE,PS,PM,nloc,NMw,NSw,Zw,PSw,PMw
        .def  nrule,rule_table,input,output,rule,memb
        .def  loc,num,den,TEMP,TEMPH,x,X,U
```

```

        .bss PwmChannel,1          ; Current PWM channel
        .bss PwmPeriod,1          ; Current period of PWM channels
        .bss NewServo,1           ; New servo value
        .bss ErrDiff,1            ; Error difference
        .bss Kintegrator,1        ; Error integrator
        .bss ErrNow,1             ; Current error function
        .bss ErrLast,1            ; Last error function result
        .bss DesiredPosition,1    ; Servo to this position

; The following are fuzzy section, initialized and uninitialized

scalein    .set 256/3
scaleout    .set 2*100/3

;          .sect "pfuzzin"
          .asect "pfuzzin", 10h
          .label prule_table

rule_table:
        .word 2,2,2
        .word 3,2,1
        .word 4,2,0
        .word 1,2,3
        .word 0,2,4
        .word 2,1,3
        .word 2,0,4
        .word 2,3,1
        .word      2,4,0
        .word 3,1,2
        .word 1,3,2

ninput      .word 2
noutput     .word 1

triangle    .word 3
nmemb       .word 5
NMaddr      .word NM
NM          .word -3*scalein, -2*scalein, -1*scalein
NS          .word -2*scalein, -1*scalein, 0*scalein
ZE          .word -1*scalein, 0*scalein, 1*scalein
PS          .word 0*scalein, 1*scalein, 2*scalein
PM          .word 1*scalein, 2*scalein, 3*scalein

nloc        .word 5
NMwaddr     .word NMw
NMw         .word -2*scaleout
NSw         .word -1*scaleout
Zw          .word 0*scaleout
PSw         .word 1*scaleout
PMw         .word 2*scaleout

```

```

nrule          .word 11
rule_tableaddr .word rule_table
input          .word 0
output         .word 0
rule           .word 0
memb           .word 0
loc            .word 0
num            .word 0
den            .word 0
minZ           .word 0

TEMP           .word 0
TEMPH          .word 0
ONE2           .word 1

xaddr          .word x
Xaddr          .word X
Uaddr          .word U

rule_end:

rambeg         .usect "fuzzin", rule_end-rule_table

x              .usect "fuzzunin", 2
X              .usect "fuzzunin", 2*11
U              .usect "fuzzunin", 5

; !end!
; *****
; *****
; !NAME!
;          comp.s
;
; !PATHS!
;          Local Variables
; !0!

; !end!
; *****

```

```

;*****
;
;      .text
;      .def  servo_ISR
; !NAME!
servo_ISR:
;
; !PATHS!
;      functions\all
;      functions\interrupt
;      modules\utils.s
; !O!
; DESCRIPTION:
;
;      This is the servo compensator. The distribution algorithm is
;      modified from PID to fuzzy format.
;
; RETURNS:
;      No changes to foreground environment.
;
; STACK LEVELS:
;      0
;
; EXTERNAL DATA REFERENCES
;      .ref  Position                ; Position from feedback ISR
;      .ref  ONE,CONSTFFFF           ; Useful constants
;      .ref  ISR_TMP                 ; Scratch
;
;*****
; !skip start!
;
;      SOVM                        ; Saturate math
;
; First need to copy data over from program
;
;      LACK      prule_table
;      ADD ONE, 11
;      ADD ONE, 8
;
;      LDPK      0
;      LARK      AR0, 6Fh
;      LARK      AR1, rambeg        ; beginning of RAM block
RAMLOOP: MAR *, 1
;      TBLR      *+
;      LDPK      1
;      ADD ONE
;      LDPK      0
;      MAR *,0
;      BANZ      RAMLOOP
;
;                                     ; Get Theta and dTheta values
;
;      LDPK      1
;      LAC ErrNow        ; Save old error
;      SACL      ErrLast
;
;      ZALH      DesiredPosition    ; Calc new error
;      SUBH      Position
;      SACH      ErrNow

```

```

                LDPK            0
                MAR            *,0                ; store x input value (Theta)
LAR  AR0,xaddr
SACH  *+

                LDPK            1
LAC  ErrNow
SUB  ErrLast
SACL ErrDiff

                LDPK            0
                SACL            *                ; dTheta

; Calculate X
; indices for X[input][memberfn]
xfn:
    ZAC
    SACL    input                ; for (i=0;
xfnl:      ZAC
    SACL    memb                ; for (j=0;
    LAC     ninput                ; i<ninput
    SUB     input
    BLEZ    ilend
xfnml:     LAC  nmemb                ; j<nmemb
    SUB     memb
    BLEZ    mlend
    LAC     xaddr                ; load &x[input] AR0
    ADD     input
    SACL    TEMP
    LAR     AR0, TEMP
    LAC     NMaddr                ; load &memb into AR1
    LT      triangle
    MPY     memb
    APAC
    SACL    TEMP
    LAR     AR1, TEMP
    CALL    membership                ; Get membership value with AR's loaded
                                        ; and leave answer in ACC
    SACL    TEMPH                ; Save value
    LT      nmemb                ; Calculate index for X[input][memb]
    MPY     input
    PAC
    ADD     memb
    ADD     Xaddr
    SACL    TEMP
    MAR     *,1
    LAR     AR1, TEMP
    LAC     TEMPH                ; Store value in X[input][memb]
    SACL    *
    LAC     memb                ; memb++
    ADD     ONE2
    SACL    memb
    B       xfnml
mlend:     LAC  input                ; input++
    ADD     ONE2
    SACL    input

```

```

        B          xfnl                      ; exit memb loop

ilend:      NOP
; exit input loop

; Zero out U values
        LAR        AR0, nmemb
        LAR        AR1, Uaddr
        ZAC
Uzl: MAR    *,1
        SACL        *+
        MAR        *,0
        BANZ       Uzl

; calculate value of Z
zfn:
        ZAC
        SACL        rule                    ; for (i=0;
zfnrl:     ZAC                                ; for (j=0;
        SACL        input                    ; i<nrule
        LAC         nrule                    ; initialize Z[rule]
        SUB         rule                    ; kluge since PM = 1*scalein
        BLEZ        zrlend
;   LAC         ONE2
        LAC         PM
        SACL        minZ
zfnil:     LAC        ninput                 ; i<ninput
        SUB         input
        BLEZ        zilend
        LAC         ninput                 ; find &RULE_TABLE[rule][input]
        ADD         noutput
        SACL        TEMP
        LT          TEMP
        MPY         rule
        PAC
        ADD         input
        ADD         rule_tableaddr
        SACL        TEMP
        MAR         *,0                    ; find RULE_TABLE[rule][input]
        LAR         AR0, TEMP
        LAC         *
        SACL        TEMP                    ; TEMP = RULE_TABLE[rule][input]
        LT          nmemb                    ; find &X
        MPY         input
        PAC
        ADD         TEMP
        ADD         Xaddr
        SACL        TEMP                    ; &(X[input][RULE_TABLE[rule][input])
        LAR         AR0, TEMP                ; check for min
        MAR         *,0
        LAC         minZ
        SUB         *
        BLZ         notmin
        LAC         *
        SACL        minZ

```

```

notmin:      LAC input                ; input++
            ADD  ONE2
            SACL input
            B    zfnil

; check max for U
zilend: LAC  ninput                    ; get &U[RULE_TABLE[rule][input+output]
            ADD  noutput
            SACL TEMP
            LT   TEMP
            MPY  rule
            PAC
            ADD  ninput                ; total kluge for input+output
            ADD  rule_tableaddr
            SACL TEMP
            MAR  *, 0
            LAR  AR0, TEMP
            LAC  *
            ADD  Uaddr                 ; got RULE_TABLE[rule][input+output]
            SACL TEMP
            MAR  *,0
            LAR  AR0, TEMP
            LAC  minZ                  ; U[RULE_TABLE[rule][input+output]]
            SUB  *                      ; if U<minZ (looking for max)
            BLZ  notmax
            LAC  minZ                  ;
            SACL *                      ; store new U

notmax:      LAC rule                  ; rule++
            ADD  ONE2
            SACL rule
            B    zfnrl

zrlend: NOP

; Need to defuzzify output. u = num/den = sigma U*Loc/ sigma U
; (ignoring wt.)
defuzz:
            ZAC
            SACL loc                    ; for (i=0;
            SACL TEMP
            SACL TEMPH
            LAR  AR0, NMwaddr
            LAR  AR1, Uaddr

```

```

; Sigma loc*U
df11: LAC nmemb ; i < nloc;
      SUB loc
      BLEZ dlend1
      MAR *,0
      LT *+
      MAR *,1
      MPY *+
      ZALH TEMPH
      ADDS TEMP
      APAC
      SACH TEMPH
      SACL TEMP
      LAC loc ; i++
      ADD ONE2
      SACL loc
      B df11
dlend1: LAC TEMP
      SACL num ; Store
      NOP

; Sigma U
      ZAC ; for (i=0;
      SACL loc
      SACL TEMP
      LAR AR0, Uaddr
      MAR *,0
df12: LAC nloc ; i<nloc;
      SUB loc
      BLZ dlend2
      LAC TEMP
      MAR *,0
      ADD *+
      SACL TEMP
      LAC loc ; ++i
      ADD ONE2
      SACL loc
      B df12
dlend2: LAC TEMP
      BZ dzero
      SACL den
      NOP
      B cdiv
dzero: ZAC
      B endfuz

; divide num/den and scale for new servo

```



```

cdiv:      MAR *,0
          LT      TEMPH
          MPY     den
          PAC
          SACH    TEMP          ; Sign
          LAC     den
          ABS
          SACL     den
          ZALH    TEMPH
          ADDS    num
          ABS
          LARK    AR0,15
DIV:       SUBC   den
          BANZ    DIV

          SACL    TEMPH          ; Quotient
          LAC     TEMP          ; Sign
          BGEZ    done
          ZAC
          SUB     TEMPH
          SACL    TEMPH          ; New Servo

done:      ZAC
          LAC     TEMPH
          B       endfuz

membership:
          MAR     *,0          ; &x[input]
          LAC     *
          MAR     *,1          ; &memb
          SUB     *+          ; if x>x0
          BLEZ    zvalue
          MAR     *,0
          LAC     *
          MAR     *,1
          SUB     *+          ; if x>x1
          BLZ     div1
          MAR     *,0
          LAC     *
          MAR     *,1
          SUB     *+          ; if x>x2
          BLZ     div2
          B       zvalue
div1: MAR     *, AR0          ; x-x0
          LAC     *
          MAR     *, AR1
          MAR     *_-
          MAR     *_-
          SUB     *
          RET
div2: MAR     *, AR1          ; x2-x
          MAR     *_-
          LAC     *
          MAR     *, AR0
          SUB     *
          RET

```

```

zvalue: LACK 0          ; value = 0
      RET
endfuz: LDPK 1
      SACL    NewServo   ; Save result
      ZAC     ; negation kluge
      SUB     NewServo
      SACL    NewServo
; yoda: B      servo_ISR
;
; Check the answer so that it fits within the limits of the timer's
; period: [0..PwmPeriod*4)
;

```

```

; LACK    0FFh
; SACL    PwmPeriod
LAC      PwmPeriod,1          ; Check positive: 2*Period-4
SUB      ONE,3
SUB      NewServo            ; - value
BGZ      pwmisr20
LAC      PwmPeriod,1          ; Oops, pin at biggest
SUB      ONE,3
SACL     NewServo
B        pwmisrSet
pwmisr20:                               ; Negative OK?
LAC      NewServo
ADD      PwmPeriod,1          ; Add largest offset
BGEZ     pwmisrSet
ZAC
SUB      PwmPeriod,1          ; Make largest negative
SACL     NewServo
;
; Set Output PWM to new value
;
pwmisrSet:
LAC      PwmPeriod,1          ; Get center period value
ADD      NewServo            ; Add newly computed value
SACL     NewServo
.if      ChipV1R1             ; PWM bug workaround
.ref     CONST3
SUB      CONST3
BLEZ     pwmisr10             ; No fixes needed
LAC      NewServo             ; Lower 2 bits set?
AND      CONST3
SUB      CONST3
BNZ      pwmisr10             ; OK, No fixes needed
LACK     4
ADD      NewServo
SACL     NewServo
pwmisr10:
.endif
LACK     ActionBank           ; Set bank for timers
SACL     ISR_TMP
OUT      ISR_TMP,BSR
OUT      NewServo,ACT0        ; Output channel 1 timer
LAC      PwmPeriod,2          ; Calculate complimentary output
SUB      ONE,2
SUB      NewServo
SACL     NewServo
OUT      NewServo,ACT1        ; Set complimentary channel 2 timer
RET                                             ; All done.. back to the ISR already
                                             ; in progress.
; !skip end!
; !END!

;=====
; END OF FILE
;=====
.end

```

## References

- [1] B. Kosko. *Neural Networks and Fuzzy Systems: A Dynamical Systems Approach to Machine Intelligence*. Englewood Cliffs, New Jersey: Prentice-Hall, Inc., 1992.
- [2] Y.F. Li and C.C. Lau. "Development of Fuzzy Algorithms for Servo Systems," *IEEE Control Systems Magazine*. April, 1989, pp. 65-71.
- [3] K. Self. "Designing with Fuzzy Logic," *IEEE Spectrum*. November, 1990, pp. 42-44, 105.
- [4] P. Thrift. *The Programmable Fuzzy Logic Array*. Dallas, Texas: Texas Instruments Central Research Laboratories, 1992.
- [5] *Power-14/Power Source User's Manual*. Rochester, New York: Teknic, Inc., 1989.
- [6] *TMS320C1x User's Guide*. Dallas, Texas: Texas Instruments, 1991.

# **The Programmable Fuzzy Logic Array**

**Philip Thrift  
Central Research Laboratories  
Texas Instruments Incorporated**

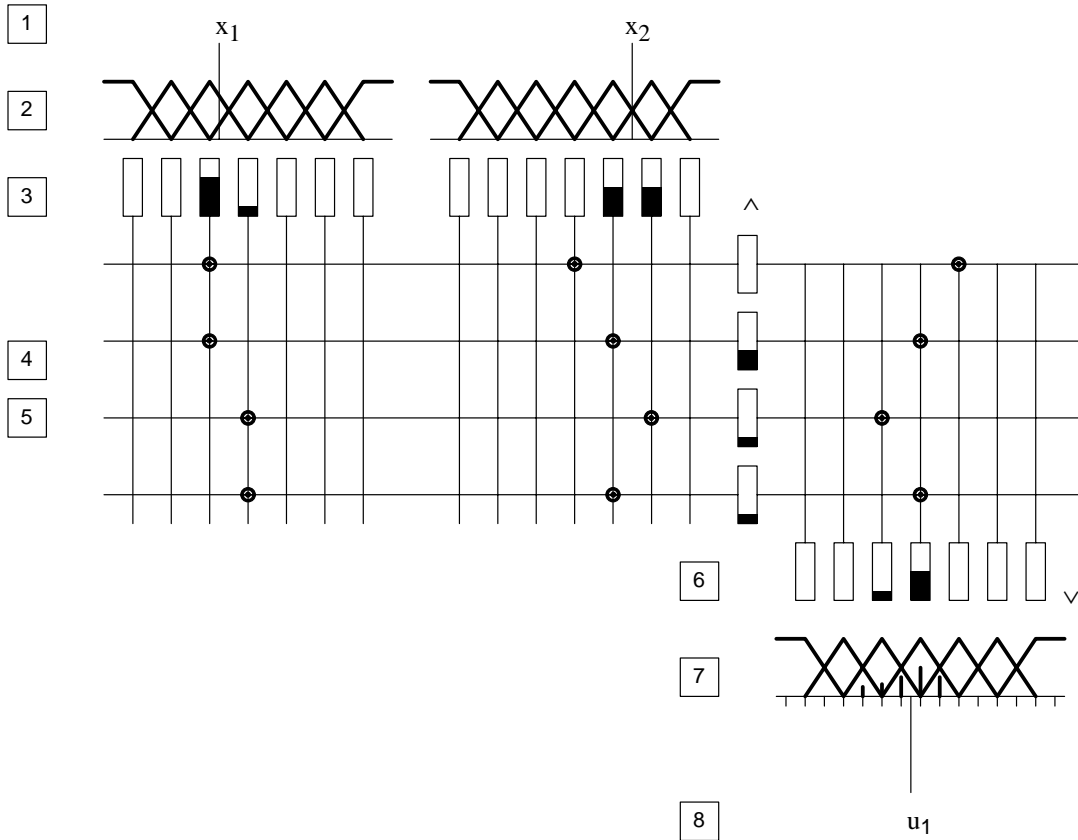
## Introduction

The Programmable Fuzzy Logic Array (PFLA) produces nonlinear multidimensional mappings by encoding fuzzy rule systems in a programmable array architecture. It can be used as a component of a graphical-user interface (GUI) for designing fuzzy controllers, as well as a software blueprint for mapping onto VLSI hardware. An advantage of the PFLA over other fuzzy representations, such as the FAM (Fuzzy Associative Memory) [2], is the PFLA's ability to easily visualize several inputs and outputs simultaneously.

## Data Flow

Figure 1 shows the PFLA data flow. Succeeding text describes PFLA in general mapping terms from  $P$  inputs to  $Q$  outputs, but only  $P = 2$  and  $Q = 1$  are shown in Figure 1. Each item number in the text corresponds to a step of Figure 1.

Figure 1. PFLA Data Flow



1. Each input  $x_i : i = 1, \dots, P$  to the PFLA is a numerical value in a range  $[a_i, b_i]$ .
2. Defined over each input range  $[a_i, b_i] : i = 1, \dots, P$  are fuzzy membership functions  $F_i^1, \dots, F_i^{n_i}$ , where  $n_i$  is the number of membership functions defined for input  $i$ . Each membership function varies between 0 and 1. The cases shown in Figure 1 are trapezoidal membership functions, which are defined in Appendix A. Other parametric families of membership functions can be substituted. Also, for each fuzzy membership function  $F_i^j$ , there is a corresponding label. A typical labeling scheme (labels are not shown in Figure 1) for  $n_i = 7$  is: negative large (NL), negative medium (NM), negative small (NS), zero (ZE), positive small (PS), positive medium (PM), positive large (PL). Although the fuzzy sets for each input in Figure 1 appear symmetrically spaced, these are not necessarily the optimal settings. If trapezoidal memberships functions are used, four numbers  $t_i^j = [t_{i,1}^j, t_{i,2}^j, t_{i,3}^j, t_{i,4}^j]$  must be specified for each membership function  $F_i^j$ .
3. Each input  $x_i$  is evaluated by each of its fuzzy membership functions to produce a value  $f_i^j$ :

$$f_i^j = F_i^j(x_i) : i = 1, \dots, P \quad j = 1, \dots, n_i \quad (9)$$

These values appear in the boxes as shown as a thermometer level.  $f_i^j$  also refers to the box that contains its value. In the example shown, only two boxes for each input have positive evaluation.

4. Fuzzy rules are encoded in a crossbar pattern. Corresponding to each of the input boxes  $f_i^j$  in 3, above, is a vertical wire dropping down. A horizontal wire crosses those vertical wires and also the vertical wires corresponding to output boxes in 6, below. A connection is indicated by a •. For each horizontal wire, there is, at most, one connection per input and output variable. Each horizontal connection pattern encodes a rule. For example, the connection pattern on the first horizontal line encodes the rule:

$$\text{If } x_1 \text{ is NS and } x_2 \text{ is ZE, then } u_1 \text{ is PS.} \quad (10)$$

Four rules are shown in Figure 1.  $R$  rules can be specified by a table of numbers:

$$r_1^1, \dots, r_1^P, s_1^1, \dots, s_1^Q \quad (11)$$

$$r_R^1, \dots, r_R^P, s_R^1, \dots, s_R^Q \quad (12)$$

where  $r_k^i$  is in  $\{1, \dots, n_i, \text{NULL}\}$ ,  $s_k^j$  is in  $\{1, \dots, m_j, \text{NULL}\}$ . Here,  $m_j$  is the number of fuzzy sets for output  $j$ . The  $k$ th horizontal wire specifies the rule

$$F_1^{r_k^1}, \dots, F_P^{r_k^P} \rightarrow G_1^{s_k^1}, \dots, G_Q^{s_k^Q} \quad (13)$$

The NULL indicates that there is no connection for this input/output variable. In Figure 1, the connections would be

3,4,5  
3,5,4  
4,6,3  
4,5,4

5. Intercepting the horizontal wires between the inputs and outputs are the  $\wedge$  (“wedge”) boxes. A conventional  $\wedge$  operator is the numerical minimum of the values (this is used in Figure 1), but other operators are possible (for example, product, or any of a set of so-called t-norms [1]).

At the  $k$ th  $\wedge$  box, this is produced:

$$h_k = \wedge (f_1^k, f_2^k, \dots, f_P^k) \quad (14)$$

If  $f_k^j$  is NULL, this argument is omitted.

6. The  $\vee$  (“vee”) boxes are computed according to the values of the connections above them. A conventional  $\vee$  operator is the numerical maximum of the values (see Figure 1). The values of the  $\vee$  boxes are

$$g_j^i = \vee [h_k : s_k^i = j] \quad i = 1, \dots, Q \quad j = 1, \dots, m_1 \quad (15)$$

Other operators [1] can be substituted for this operator (for example, probabilistic sum:  $x \oplus y = x + y - xy$ , etc.).

7. Defined over each output range  $[c_i, d_i] : i = 1, \dots, Q$  are fuzzy membership functions  $G_1^1, \dots, G_i^{m_i}$ , where  $m_i$  is the number of membership functions defined for output  $i$ . Also defined for each output  $i$  is a vector  $l_i$  of locations quantizing the range:  $l_i = [l_i^1, \dots, l_i^{q_i}]$ , where  $q_i$  is the number of locations specified for output  $i$ . In Figure 1, 17 locations are shown for output 1.

$$w_i^j = G_i^j(l_i^j), \quad i = 1, \dots, Q \quad j = 1, \dots, m_i \quad (16)$$

Here,  $G_i^j$  is applied by component to get the resulting vector. The  $w_i^j$  are precomputed and stored for each  $\vee$  box. Then, this is computed for output  $i$ :

$$v_i = \vee^*(w_i^1 \wedge^* g_i^1, \dots, w_i^{m_i} \wedge^* g_i^{m_i}) \quad (17)$$

Here, the “wedge” and “vee” operators are not necessarily the ones used in the boxes above. In Figure 1,  $v_i$  is shown as a sequence of vertical bars;  $\vee^*$  is the maximum operation, and  $\wedge^*$  is the product.

8. The final stage is to compute  $u_i$  from  $v_i$  for each output:

$$u_i = \frac{v_i \cdot 1_i}{v_i \cdot 1} \quad (18)$$

where  $1 = [1, \dots, 1]$  is a vector of 1s.

If, corresponding to each output fuzzy set  $G_i^j$ , there is a single distinct location  $l_i^j, w_i^j = G_i^j(l_i^j)$ , and  $\wedge^*$  is the product operation, then the above defuzzification procedure reduces to

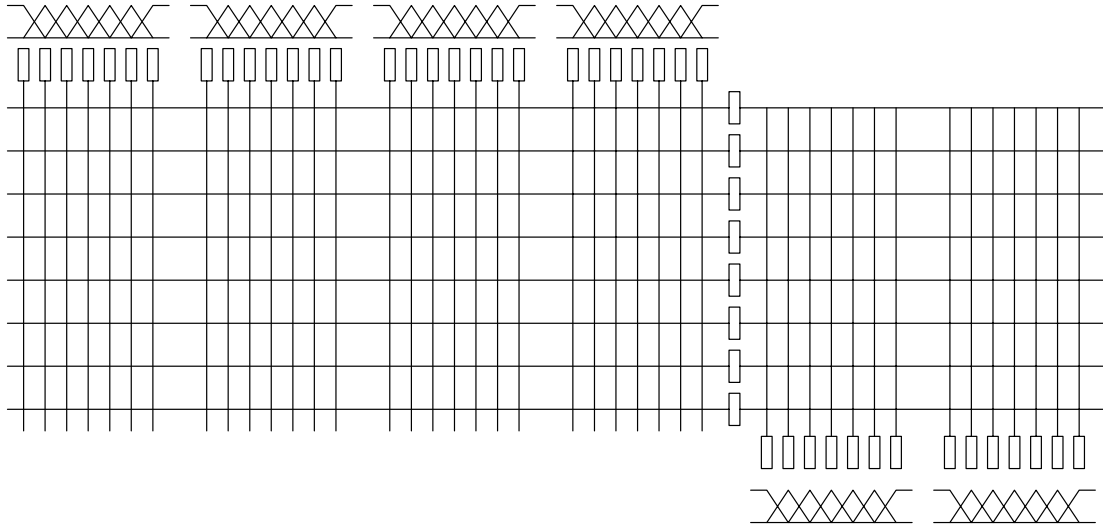
$$u_i = \frac{w_i^1 \cdot g_i^1 \cdot l_i^1 + \dots + w_i^{m_i} \cdot g_i^{m_i} \cdot l_i^{m_i}}{w_i^1 \cdot g_i^1 + \dots + w_i^{m_i} \cdot g_i^{m_i}} \quad (19)$$

This provides “weighted point mass defuzzification”. In this case, only a “weight”  $w_i^j$  and a location  $l_i^j$  must be stored for each  $\vee$  box.

Figure 2 shows the state of the system for particular inputs  $x_1, x_2$ . Thermometer levels indicate the values in both the fuzzification and the  $\wedge$  and  $\vee$  boxes (here, *min* and *max* are the operations performed). Darkened rule connections indicate values that propagated through the array (since *min* and *max* are used, on each output line there is, in general, one *max* winner and one *min* winner). Figure 2 also shows the layout for a four-input, two-output (4-2) system. In a GUI, you can use point-and-clicks to set the rule connections and membership function positioning.



**Figure 2. PFLA 4–2 Layout**



### Summary

This is the information that provides the setting for the PFLA:

#Inputs: P	#Input fuzzy sets: $n_1, \dots, n_P$
#Outputs: Q	#Output fuzzy sets: $m_1, \dots, m_Q$
Input fuzzy sets	$[a_i, b_i], t_i^1, \dots, t_i^{n_i}, i = 1, \dots, P$
#Rules: R	Operators: $\wedge \vee$
#Rule table:	$r_k^1, \dots, r_k^P, s_k^1, \dots, s_k^Q, k = 1, \dots, R$
Output defuzzifiers	$[c_j, d_j], l_j, w_j^1, \dots, w_j^{m_j}, j = 1, \dots, Q \wedge * \vee *$

### Appendix A

A trapezoidal membership function on an interval  $[a, b]$  is specified by four numbers  $t_1, t_2, t_3, t_4$ , satisfying

$$a \leq t_1 \leq t_2 \leq t_3 \leq t_4 \leq b$$

The trapezoidal function is defined by

$$TZ(a, t_1, t_2, t_3, t_4, b)(x) = \begin{cases} 0 & \text{for } x \text{ in } [a, t_1] \\ (x - t_1)/(t_2 - t_1) & \text{for } x \text{ in } (t_1, t_2) \\ 1 & \text{for } x \text{ in } [t_2, t_3] \\ (t_4 - x)/(t_4 - t_3) & \text{for } x \text{ in } (t_3, t_4) \\ 0 & \text{for } x \text{ in } [t_4, b] \end{cases}$$

Note that if  $t_1 = t_2$  or  $t_3 = t_4$ , this part of the definition is void.

## References

- [1] Pedrycz, W. *Fuzzy Control and Fuzzy Systems*. John Wiley & Sons Inc., 1989.
- [2] Kosko, B. *Neural Networks and Fuzzy Systems: A Dynamical Systems Approach to Machine Intelligence*. Prentice-Hall, Inc., 1992.