

A PROPOSAL FOR MULTI-VALUED LOGIC APPROACH
TO FUZZY CONTROLLERS IMPLEMENTATION

V. VARSHAVSKY,

V. MARAKHOVSKY, I. LEVIN, H. SAITO

March 12, 2007



Computer Organization Laboratory
The University of Aizu
Tsuruga, Ikki-machi, Aizu-Wakamatsu City
Fukushima 965-8580, Japan

| | |
|--|--------------------------------------|
| <p>Title: A Proposal for Multi-valued Logic Approach to Fuzzy Controllers Implementation</p> | |
| <p>Authors: V. Varshavsky, V. Marakhovsky, I. Levin, H. Saito</p> | |
| <p>Key Words and Phrases: Fuzzy Logic, Fuzzy Controller, Fuzzy rules, Fuzzy inference, Fuzzification and Defuzzification, Multi-Valued Logic, Multi-threshold Element, Functional Completeness, Summing Amplifier.</p> | |
| <p>Abstract: This paper offers a new technique for designing fuzzy controllers as analog hardware devices on the base of CMOS implementation of multi-valued logical functions. This approach is based on using a summing amplifier with saturation as a building block that can be considered as a multi-threshold logical element. The functional completeness in an arbitrary-valued logic of a summing amplifier with saturation is proven. This fact gives a theoretical background for an analog implementation of fuzzy devices. In contrast with the traditional software approach to fuzzy controller implementations based on explicit fuzzification, fuzzy inference, and defuzzification procedures, hardware implementations of fuzzy controllers as analog devices have advantages of higher speed, lower power consumption, smaller die area and more. Universal and proper design methods for such type of hardware are suggested. The paper illustrates design examples for real industrial fuzzy controllers and provides SPICE simulation results of their functioning.</p> | |
| <p>Report Date: 3/12/2007</p> | <p>Written Language: English</p> |
| <p>Any Other Identifying Information of this Report: Sent to the journal <i>Multi-Valued Logic and Soft Computing</i></p> | |
| <p>Distribution Statement: First Issue: 12 copies</p> | |
| <p>Supplementary Notes:</p> | |

Computer Organization Laboratory
The University of Aizu
Aizu-Wakamatsu
Fukushima 965-8580
Japan

A Proposal for Multi-valued Logic Approach to Fuzzy Controllers Implementation

V. VARSHAVSKY,

V. MARAKHOVSKY¹, I. LEVIN², H. SAITO³
^{1,3}The University of Aizu, JAPAN

² Bar Ilan University, ISRAEL

marak@u-aizu.ac.jp, levinil@eng.biu.ac.il, hiroshis@u-aizu.ac.jp

Abstract: - This paper offers a new technique for designing fuzzy controllers as analog hardware devices on the base of CMOS implementation of multi-valued logical functions. This approach is based on using a summing amplifier with saturation as a building block that can be considered as a multi-threshold logical element. The functional completeness in an arbitrary-valued logic of a summing amplifier with saturation is proven. This fact gives a theoretical background for an analog implementation of fuzzy devices. In contrast with the traditional software approach to fuzzy controller implementations based on explicit fuzzification, fuzzy inference, and defuzzification procedures, hardware implementations of fuzzy controllers as analog devices have advantages of higher speed, lower power consumption, smaller die area and more. Universal and proper design methods for such type of hardware are suggested. The paper illustrates design examples for real industrial fuzzy controllers and provides SPICE simulation results of their functioning.

Key-Words: - Fuzzy Logic, Fuzzy Controller, Fuzzy rules, Fuzzy inference, Fuzzification and Defuzzification, Multi-Valued Logic, Multi-threshold Element, Functional Completeness, Summing Amplifier.

1 Introduction

Fuzzy logic control is a methodology bridging *artificial intelligence* and traditional *control theory*. This methodology is usually applied in the only cases when accuracy is not of high necessity or importance. On the other hand, as it is stated in [1], “Fuzzy Logic can address complex control problems, such as robotic arm movement, chemical or manufacturing process control, antiskids braking systems, or automobile transmission control with more precision and accuracy, in many cases, than traditional control techniques have... . Fuzzy Logic is a methodology for expressing operational laws of a system in linguistic terms instead of mathematical equations.”

Wide spread of the fuzzy control and high effectiveness of its applications in a great extend is determined by formalization opportunities of necessary behavior of a controller as a “fuzzy” (flexible) representation. This representation usually is formulated in the form of logical (fuzzy) rules under linguistic variables of a type “If A then B”.

The Fuzzy Logic methodology [2, 3] comprises three phases:

1. The *Fuzzification* is a transformation of analog (continuous) input variables to linguistic ones, e.g., transformation of temperature into the terms *cool*, *warm*, *hot* or transformation of speed into the terms *negative big (NB)*, *negative small (NS)*, *zero (Z)*”, *positive small (PS)*, *positive big (PB)*. Such transformation is realized by introduction of so-called *membership functions*, which define both a range of value and a degree of membership. For linguistic variables it is important not only which membership function a variable belongs to, but also a relative degree (weight) to which it is a member. A variable can have a weighted membership in several membership functions at the same time.

2. The *Fuzzy inference* maps input linguistic variables onto output linguistic variables on the base a system of fuzzy rules “IF... THEN ...”-type. For instance: “IF the temperature is *worm* THEN the speed is *Positive Small (PS)*” or “IF the speed is *Negative Big (NB)* THEN force is *ZERO*”, etc. Since input linguistic variables are weighted, the output linguistic variables can be obtained weighted as well. Traditional fuzzy logic approach comprises Mamdani-type and Sugeno-type inference methods. The Mamdani-type method is more intuitive and assumes the output variables as a fuzzy set. Fuzzy rules in it contain a *fuzzy precondition* part (after IF) and a *fuzzy consequence* part (after THEN). The Sugeno-type method expects the output variables to be singletons or dealing with consequents that are equations. So it is better suited for mathematical analysis, nonlinear system modeling and interpolation.

3. In the *defuzzification* phase, the weighted values of output linguistic variables obtained as a result of fuzzy inference have to be transformed to analogue (continuous) variables. This procedure is also based on membership functions. Two major methods are used for defuzzification:

- the *maximum* defuzzification method, wherein an output value is determined by the linguistic variable with the maximum weight;
- the *centroid* calculation defuzzification method, wherein an output value is determined by the weighted influence of all the active output membership functions.

As a rule, or at least in a great part of applications, a fuzzy controller is a transformer of input analog signals into an analog output signal. A linguistic variable is a *subjective*

characteristic of an input analog variable, values of which are transformed on the base of given membership functions into a set of weighted values corresponding linguistic variables. This procedure is called a fuzzification and it contains as its composite part the analog-digital transformation.

A set of combinations of weighted linguistic variables corresponds to each value combination of input analog variables. On the base of a system of fuzzy inference rules it is possible to receive the set of weighted output linguistic variables. Using these variables and their membership functions, with help of one of well known defuzzification methods it is possible to form values of the analog output variable. The defuzzification procedure also includes digital-analog transformation.

At present the most wide-spread way of fuzzy logic control implementation is using the programmable fuzzy controllers, which are available on the market together with the means of computer aided programming (e.g. Motorola's 8-bit 68HC11 and 16-bit 68HC12 microcontrollers or specialized fuzzy processors of Siemens 80C517/80C535 families). However, in spite of the implementation evidence and fuzzy controllers' accessibility this approach to controller implementation possesses some disadvantages, e.g. such as high cost and low throughput (that is especially important when fuzzy control in the control contour is used) etc.

This work shows that for a sufficient wide set of applications fuzzy controllers can be implemented as rather simple CMOS devices, which can be used in embedded systems or as an IP core. What is the basic idea of the proposal?

A fuzzy controller is a deterministic device, for which one and only one value of the output analog variable corresponds to each value combination of the input analog variables. It means that the fuzzy controller should realize an analog function $Y = f(x_1, x_2, \dots, x_n)$ ¹.

There are two important questions:

1. How to transit from a standard specification of a fuzzy logic function to the specification of corresponding analog function?
2. How to transit from an analog function specification and/or from a standard specification of a fuzzy logic function to corresponding CMOS implementation?

First of all, let us address to membership functions. In most cases [2 – 4], membership functions have a triangle or trapeze form (see Fig.1).

¹ It should be noticed that in suppressing majority of publications on fuzzy controllers, this function is given as a response surface and practically without exception this surface has a piecewise linear form.

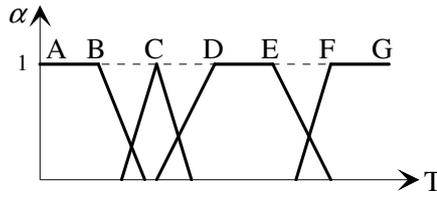


Figure 1. Types of membership functions.

In Fig.1 linguistic points (variables) A and B are *cold*, C is *fresh*, D and E are *warm*, F and G are *hot*. These points determine the connection of the linguistic variables with values of the analog variable T (T is *temperature*). Relatively to these points and similar points for other analog input variables we can compose a table of fuzzy rules connecting combinations of input linguistic variables with output linguistic variables.

On the base of membership functions we can put into accordance to the input and output linguistic variables a set of integer numbers splitting by appropriate way all diapason of changing of corresponding analog variables. Then the table of fuzzy rules will to determine by obvious way the function of multi-valued logic, values of which define the digit representation of the output linguistic variable on chosen value combinations of multi-valued input variables.

In other words, according to our concept, for a broad class of fuzzy controller specifications it is possible to construct corresponding tables connecting input and output membership functions. Frequently the membership functions evenly divide the ranges of output variables' variations. If it is not so, the membership functions can be brought to even scale by increasing the number of gradations or, as it will be shown later, by introducing a certain equalization procedure for logical levels. Therefore, specification tables represent nothing but tables determining a specific multi-valued logical function. And what is more, for a number of implementations it is possible to neglect weighting and determining input linguistic variables and simply to use continuous-valued variables.

The above idea was in the focus of our research. We dealt with searching for simple basic multi-valued functions, which, from the one hand, would present a complete functional basis in the multi-valued logic, and from the other hand, could be efficiently implemented by CMOS technology.

2 Possibility of Fuzzy Controller CMOS Implementation

2.1 Summing Amplifier as a Multi-Valued Logical Element

Summing amplifier's behavior, accurate to the members of the infinitesimal order that is determined by the amplifier's gain factor in disconnected condition (Fig.2), is described as follows:

$$V_{out} = \begin{cases} V_{dd} & \text{if } \sum_{j=1}^n \frac{R_0}{R_j} (V_j - \frac{V_{dd}}{2}) \leq -\frac{V_{dd}}{2} \\ \frac{V_{dd}}{2} - \sum_{j=1}^n \frac{R_0}{R_j} (V_j - \frac{V_{dd}}{2}) & \text{if } \frac{V_{dd}}{2} > \sum_{j=1}^n \frac{R_0}{R_j} (V_j - \frac{V_{dd}}{2}) > -\frac{V_{dd}}{2} \\ 0 & \text{if } \frac{V_{dd}}{2} \leq \sum_{j=1}^n \frac{R_0}{R_j} (V_j - \frac{V_{dd}}{2}) \end{cases} \quad (1)$$

where V_{dd} – the supply voltage, V_j – the voltage on j^{th} input, R_j – the resistance of j^{th} input, R_0 – the feedback resistance, and $V_{dd}/2$ – the midpoint of the supply voltage. Dependence of V_{out} on $\sum_{j=1}^n \frac{R_0}{R_j} \cdot (V_j - \frac{V_{dd}}{2})$ is shown in Fig.3 (a).

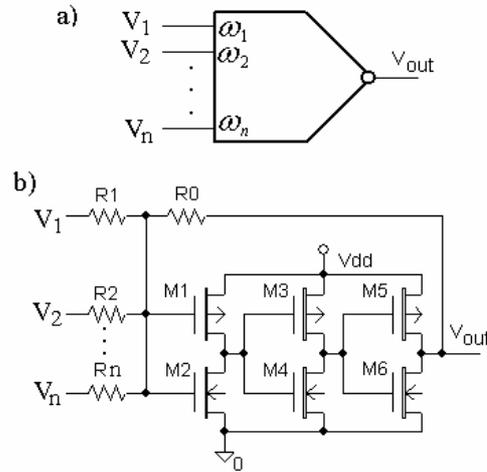


Figure 2. Summing amplifier: general designation (a); CMOS implementation using symmetrical invertors (b).

Let us split the source voltage V_{dd} on $m = 2k + 1$ voltage levels. Then replacing the input voltages $V_j - \frac{V_{dd}}{2}$ by m -valued logical variables $x_j = \frac{2 \cdot V_j - V_{dd}}{V_{dd}} \cdot k$ and the output voltage V_{out} by m -valued variable y and designating $R_0/R_j = \omega_j$ the system (1) can be represented as (2). Graphical view of (2) is shown in Fig.3 (b).

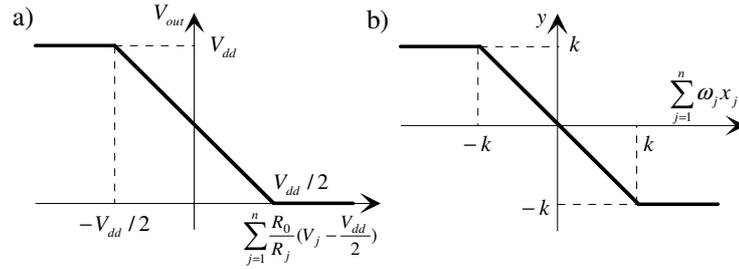


Figure 3. Summing amplifier's behavior: within voltage coordinates (a); within multi-valued variables coordinates (b).

$$y(X) = S\left(\sum_{j=1}^n \omega_j \cdot x_j\right) = \begin{cases} +k & \text{if } \sum_{j=1}^n \omega_j \cdot x_j \leq -k \\ -\sum_{j=1}^n \omega_j \cdot x_j & \text{if } k > \sum_{j=1}^n \omega_j \cdot x_j > -k \\ -k & \text{if } \sum_{j=1}^n \omega_j \cdot x_j \geq +k \end{cases} \quad (2)$$

Later on, we will call the functional element, whose behavior is determined by the system (2), a multi-valued threshold element. In the simplest case when $\omega_j = 1$, $j = 1, 2, 3$, we will call it a majority element and designate as $maj(x_1, x_2, x_3)$.

2.2 Functional Completeness of the Threshold Element

The basic operation (or a set of basic operations) is called functionally completed in arbitrary-valued logic, if any function of this logic can be represented as superposition of the basic operations.

There are some known functionally complete sets of functions. It is clear, that for proving the functional completeness of a certain new function it is sufficient to show that every function of the known functionally complete set can be represented as a superposition of the considered function. One of functionally complete functions in m -valued logic is the Webb's function [8]:

$$w(x, y) = [\max(x, y) + 1]_{\text{mod } m} \quad (3)$$

Therefore, for proving functional completeness of threshold operation in multi-valued logic it is sufficient to show how the Webb's function can be represented through this operation [5 – 7].

First, let us represent the function $\max(x_1, x_2)$ by threshold functions. To do this let us consider the function $f_a(x)$, such as

$$f_a(x) = \max(x, a) = \begin{cases} a & \text{if } a \geq x \\ x & \text{if } x > a \end{cases} \quad (4)$$

The diagram of this function is shown in Fig.4 (a).

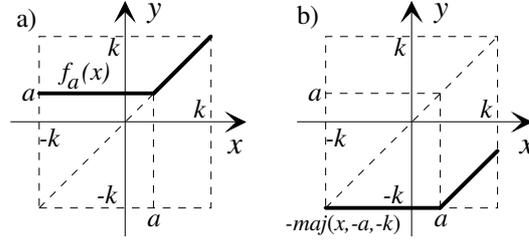


Figure 4. Diagrams of $f_a(x)$ (a) and $-maj(x, -a, -k)$ (b) functions.

The $-maj(x, -a, -k)$ function diagram is shown in Fig.4 (b). Actually, as far as $x < a$ $x - a - k < -k$ and $-maj(x, -a, -k) = -k$. Note that for all values of x ,

$$f_a(x) = -maj(x, -a, -k) + a + k$$

as it follows from Fig.4, hence

$$f_a(x) = -maj[-maj(x, -a, -k), a, k]. \quad (5)$$

Taking into consideration $-maj(a, b, c) = maj(-a, -b, -c)$, it follows from (5) that

$$\max(x_1, x_2) = maj(maj(x_1, -x_2, -k), -x_2, -k). \quad (6)$$

Now let us consider the representation of the function $y = (x+1)_{\text{mod } m}$, $x \geq 0$, $0 \leq y \leq m-1$ through threshold functions. First of all we designate $m = 2k + 1$ and change the beginning of coordinates so that the function will have a view $y = (x+k+1)_{\text{mod}(2k+1)} - k$, $x \geq -k$, $-k \leq y \leq +k$. To implement this function on threshold elements let us turn to the sequence of pictures in Fig.5.

It is easy to see that

$$(x+k+1)_{\text{mod}(2k+1)} - k = \varphi_1(x) + 2\varphi_3(x)$$

and obviously, this function can also be implemented on threshold elements as

$$y = maj(maj(x, 1, 0), k \cdot maj(maj(x, 1-k, -k), -k, 0), k \cdot maj(maj(x, 1-k, -k), -k, 0)).$$

Hence, the functional completeness of the summing amplifier in arbitrary-valued logic is shown. The proof procedure of functional completeness naturally does not give information about methods of effective synthesis. Some methods of circuit design in the proposed base will be developed later.

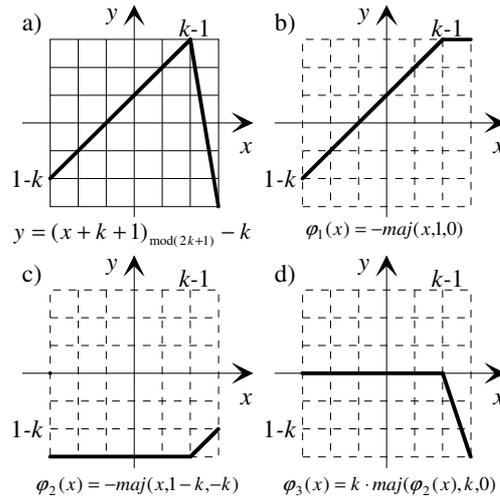


Figure 5. Implementation of the function $y = (x + k + 1)_{\text{mod}(2k+1)} - k$.

2.3 Fuzzy Devices as Multi-Valued and Analog Circuits

Conventional implementation of fuzzy devices usually has the structure shown in Fig.6. Analog variables $X = \{x_1, x_2, \dots, x_n\}$ enter the fuzzy device input. Fuzzifier converts a set of analog variables x_j into sets of weighted linguistic (digital) variables $A = \{a_1, a_2, \dots, a_n\}$.

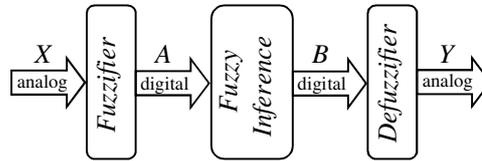


Figure 6. Conventional structure of a fuzzy device implementation.

Fuzzy Inference block generates on the base of the fuzzy rules a set of weighted linguistic variables values $B = \{b_1, b_2, \dots, b_k\}$.

Defuzzifier converts sets of weighted linguistic (digital) variables $B = \{b_1, b_2, \dots, b_k\}$ into a set of output analog variables $Y = \{y_1, y_2, \dots, y_k\}$.

As a rule, fuzzifier and defuzzifier include AD and DA (analog-digital and digital-analog) converters and are implemented on both levels (hardware and software). Fuzzy inference is usually implemented on the level of microprocessor software.

It is easy to see that each set of values of output analog variables unambiguously corresponds to some set of input analog variable values; hence a fuzzy device could be specified as a functional analog of a signal converter

$$Y(X) = \{y_1(X), y_2(X), \dots, y_k(X)\}$$

and its output Y determines a system of n -dimensional surfaces. In cases of sufficient simple membership functions (in known publications such functions are in majority), for fuzzy controller implementations as analog devices it is sufficient to provide a piece-wise linear approximation between a couples of points calculated as adjacent values of a multi-valued logic function.

Let $m = 2k + 1$ linguistic variables a_j ($a_j \in A$) correspond to values of analog variable x_j ($x_j \in X$). Then basing on a system of fuzzy rules, we can specify a system of m -valued logic functions, as follows:

$$B(A) = \{b_1(A), b_2(A), \dots, b_k(A)\}. \quad (7)$$

Note that most publications describing fuzzy controllers contain tables specifying fuzzy controllers' behavior as (7) and a plenty of publications contain piece-wise linear approximations of the corresponding surfaces.

The apparent conclusion can be made from the things mentioned above: if a fuzzy controller is represented as (7), it can be implemented as superposition of multi-valued threshold elements. In this case, owing to linear behavior of the threshold element in the zone between the saturation levels ((2) and Fig.3 (b)) natural linear approximation appears between the discrete points of specification.

In the last subsection of this section some illustrations will be given to show that for a number of real applications the suggested approach can provides simple and efficient circuits of controllers.

2.4 Fuzzy Controller Implementations as Circuits from Threshold Elements

2.4.1 Example 1

Let us consider the example, which is taken from [9, pp. 81 – 86]: “Design of a Rule-Based Fuzzy Controller for the Pitch Axis of an Unmanned Research Vehicle”.

The fuzzy control rules for the considered device depend on the error value $e = ref - output$ and changing of error $ce = \frac{old\ e - new\ e}{sampling\ period}$. Fuzzifier gives seven linguistic variables for each of input analog variables (NB – negative big; NM – negative middle; NS – negative small; ZO – zero; PS – positive small; PM – positive middle; PB – positive big).

The output has the same seven gradations. Corresponding 49 fuzzy rules are represented in Table 1.

Table of Fuzzy Rules **Table 1**

| | | e | | | | | | |
|----|----|----|----|----|----|----|----|----|
| | | NB | NM | NS | ZO | PS | PM | PB |
| ce | NB | ZO | PS | PM | PB | PB | PB | PB |
| | NM | NS | ZO | PS | PM | PB | PB | PB |
| | NS | NM | NS | ZO | PS | PM | PB | PB |
| | ZO | NB | NM | NS | ZO | PS | PM | PB |
| | PS | NB | NB | NM | NS | ZO | PS | PM |
| | PM | NB | NB | NB | NM | NS | ZO | PS |
| | PB | NB | NB | NB | NB | NM | NS | ZO |

Let us split evenly the source voltage (e.g. 3.5V) onto seven logical levels corresponding to linguistic levels and enumerate them with integer numbers from -3 to +3. Then Table 2 will represent Table 1 as the function of seven-valued logic.

It is seen from Table 2 that the function is symmetric with respect to “North-West – South-East” diagonal and its values can be calculated as $e - ce$. This dependency is shown in Fig.7.

The Seven-Valued Function **Table 2**

| | | e | | | | | | |
|----|----|----|----|----|----|----|----|---|
| | | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
| ce | -3 | 0 | 1 | 2 | 3 | 3 | 3 | 3 |
| | -2 | -1 | 0 | 1 | 2 | 3 | 3 | 3 |
| | -1 | -2 | -1 | 0 | 1 | 2 | 3 | 3 |
| | 0 | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
| | 1 | -3 | -3 | -2 | -1 | 0 | 1 | 2 |
| | 2 | -3 | -3 | -3 | -2 | -1 | 0 | 1 |
| | 3 | -3 | -3 | -3 | -3 | -2 | -1 | 0 |

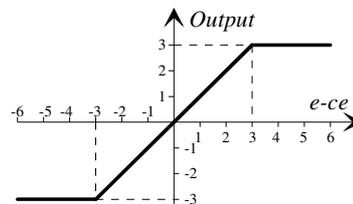


Figure 7. Graphical representation of the function specified by Table 2.

It apparently follows from comparison of Fig.3 (b) and Fig.7 that in order to reproduce the function specified by Table 2 it is sufficient to have one two-input summing amplifier and one one-input amplifier that we will call inverter.

Note that inversion of logic variables lying within $-k \div +k$ interval is the operation of diametric negation $\bar{x} = -x$; the operation $\overline{V_{out}} = V_{dd} - V_{in}$ corresponds to it in the term of the summing amplifier's output voltage. Thus CMOS circuit containing 12 transistors and 5 resistors, which implements our function, is shown in Fig.8.

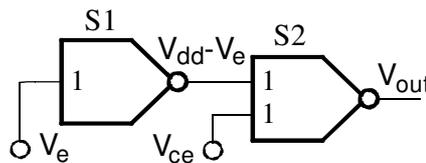


Figure 8. Implementation of the fuzzy controller specified by Table 2.

2.4.2 Example 2

This example is taken from [9, pp. 168 – 172]: “Manipulator for Man-Robot Cooperation (Control Method of Manipulator/Vehicle System with Fuzzy Inference)”.

In the considered example the experimental manipulator has two force/torque sensors. One of them is the operational force sensor F_h ; the other is “the environmental force sensor” ω . Each of input and output variables of the manipulator controller is represented with three linguistic variables – S (small), M (middle) and B (big). The controller has five fuzzy rules, as it follows:

- If $\omega = S$ then Output=B;
- If $\omega = B$ then Output=S;
- If $\omega = M$ and $F_h = S$ then Output=S;
- If $\omega = M$ and $F_h = M$ then Output=M;
- If $\omega = M$ and $F_h = B$ then Output=B;

The controller *Output* is three-valued logic function specified in Table 3.

It can be simply proved by trivial substitution that $Output = maj(2\omega, -F_h, 0)$ and CMOS implementation coincides with the circuit shown in Fig.8, if make substitutions $V_e = V_{F_h}$, $V_{ce} = V_{\omega}$ and change the weight of the input V_{ω} to 2.

The ternary function **Table 3**

| | | | | |
|----------|----|-------|----|----|
| | | F_h | | |
| | | -1 | 0 | +1 |
| ω | -1 | +1 | +1 | +1 |
| | 0 | -1 | 0 | +1 |
| | +1 | -1 | -1 | -1 |

2.4.3 Example 3. Fuzzy Controller for Washing Machine

This example is taken from Apronix Incorporated (<http://www.apronix.com/fuzzynet>).

A. Controller specification

Input variables:

Dirtiness of clothes *degree*: Large (L), Medium (M), and Small (S);

Type of dirtiness *degree*: Greasy (G), Medium (M), and Not Greasy (NG).

Output variable is washing time (minutes): Very Long (VL), Long (L), Medium (M), Short (S), and Very Short (VS).

Fuzzy rules are represented in Table 4.

Matrix of linguistic variables **Table 4**

| | | | | |
|----------------------|-----------|-----------------------------|----------|-----------|
| <i>Wash. time</i> | | <i>Dirtiness of clothes</i> | | |
| | | <i>S</i> | <i>M</i> | <i>L</i> |
| <i>Type of dirt.</i> | <i>NG</i> | <i>VS</i> | <i>S</i> | <i>M</i> |
| | <i>M</i> | <i>M</i> | <i>M</i> | <i>L</i> |
| | <i>G</i> | <i>L</i> | <i>L</i> | <i>VL</i> |

According to our approach the Table of linguistic variables (Table 4) can be transformed into the table of multi-valued logic variables (Table 5).

Matrix of multi-valued variables **Table 5**

| | | | | |
|-----------------------------|----|---------------------------------|----|----|
| <i>Wash. time</i> | | <i>Dirtiness of clothes (Y)</i> | | |
| | | -2 | 0 | +2 |
| <i>Type of dirt.</i> (X) | -2 | -2 | -1 | 0 |
| | 0 | 0 | 0 | +1 |
| | +2 | +1 | +1 | +2 |

In this table the output variable *Wash. time* has 5 logical levels but input variables *X* and *Y* have only three. Because of the change range of output and input variables should be the same in the Table 5 logical levels of input variables *X* and *Y* are -2, 0, +2.

B. Functional decomposition

Let us represent the washing time matrix as a sum of two matrixes:

$$\begin{array}{c} \text{wash time} \\ \left| \begin{array}{ccc} -2 & -1 & 0 \\ 0 & 0 & +1 \\ +1 & +1 & +2 \end{array} \right| = \left| \begin{array}{ccc} -1 & -1 & 0 \\ 0 & 0 & +1 \\ +1 & +1 & +2 \end{array} \right| + \left| \begin{array}{ccc} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right| \end{array} \quad (8)$$

or $(\text{wash. time}) = S(-\varphi_1 - \varphi_2)$ were S is the function of summing amplifier with saturation.

Let us take into consideration a function of one variable

$$\varphi_3(Y) = S(0.5 \cdot Y - 2) = |+2 + 2 + 1|. \quad (9)$$

In (9) Y corresponds to the *dirtiness of clothes* and varies from -2 to $+2$ as follows

$$Y = |-2 \quad 0 \quad +2|.$$

Now the following intermediate sum is introduced:

$$\varphi_3(Y) - 0.5 \cdot X - 2 = \left| \begin{array}{ccc} +1, +1, & 0 \\ 0, & 0, & -1 \\ -1, -1, & -2 \end{array} \right|. \quad (10)$$

Here X corresponds to the *type of dirtiness* and varies also from -2 to 2 as follows

$$X = \left| \begin{array}{c} -2 \\ 0 \\ +2 \end{array} \right|.$$

From (8) and (10) it is easy to see that (10) is $-\varphi_1$ and

$$(\text{wash. time}) = S(\varphi_3(Y) - 0.5 \cdot X - 2 - \varphi_2).$$

Now let us introduce the function:

$$\varphi_4(X, Y) = S(X + Y + 4) = \left| \begin{array}{ccc} 0, -2, -2 \\ -2, -2, -2 \\ -2, -2, -2 \end{array} \right| \quad (11)$$

and form the second intermediate sum:

$$0.5 \cdot \varphi_4(X, Y) + 1 = \left| \begin{array}{ccc} +1, 0, 0 \\ 0, 0, 0 \\ 0, 0, 0 \end{array} \right| = -\varphi_2. \quad (12)$$

Finally

$$\begin{aligned} (\text{wash time}) &= S[\varphi_3(Y) - 0.5 \cdot X - 1 + 0.5 \cdot \varphi_4(X, Y)] = \\ &S_4[S_1(0.5Y - 2) + 0.5 \cdot S_2(X + Y + 4) + 0.5 \cdot S_3(X) - 1]. \end{aligned} \quad (13)$$

In Fig.9 the CMOS implementation of the expression (13) is presented. The circuit is implemented as the superposition of four multi-valued threshold elements.

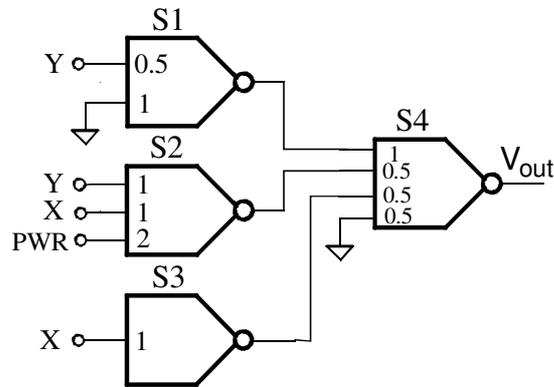


Figure 9. CMOS implementation of fuzzy controller for washing machine.

The result of the SPICE simulation of the circuit in Fig.9 is shown in Fig.10 in the form of response surface.

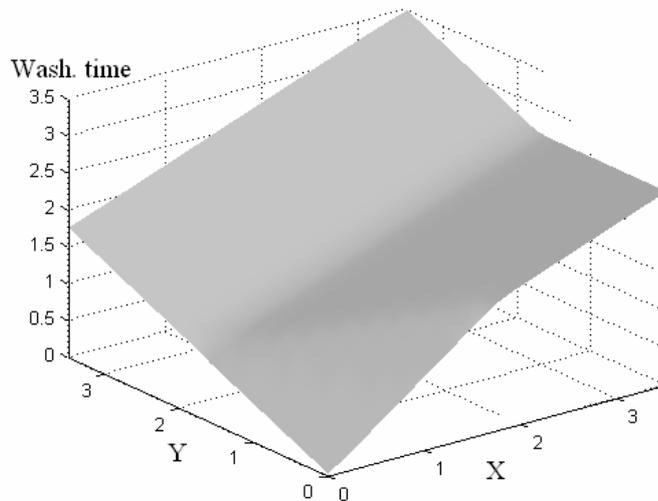


Figure 10. Results of SPICE simulation for controller in Fig.9.

In Fig.10 all variables are represented in voltages. The correspondence of logical values to voltages is shown in Table 6. It is easy to see that the controller output signal represented by the surface in Fig.10 has linear approximation between adjacent logical levels.

Correspondence voltages to logical levels **Table 6**

| | | | | |
|----|--------|-------|--------|------|
| -2 | -1 | 0 | 1 | 2 |
| 0V | 0.875V | 1.75V | 2.625V | 3.5V |

3 Universal Method of Fuzzy Inference Rules Implementation

It was shown in 2.2 and [5, 6] that a summing amplifier with saturation is a functionally complete element in any multi-valued logic (of an arbitrary value). Thus it may serve a basis for hardware implementation of fuzzy devices.

The study subject is design techniques for analog CMOS circuits that implement fuzzy controller multi-valued functions.

Without departing from the general character of the study, let us suppose that the logic has odd value ($m = 2k + 1$ and $-k \leq x_j \leq +k$). Let's also assume that $X = \{x_1, x_2, \dots, x_n\}$ is a set of input multi-valued variables and $y = F(X)$ is the output variable. Then for a function of multi-valued logic we may build an analog of the Shannon's decomposition in the binary logic:

$$y_i = F_i(X) = \bigcup_{\alpha=-k}^{+k} [\text{if } x_j = \alpha \text{ then } y = F(x_j = \alpha, X \setminus x_j)]. \quad (14)$$

Equation (14) can be further expanded and thus enabling to build a fuzzy circuit by using the variables exclusion method. To this purpose, we need a sub-circuit implementing the function:

$$\mathbf{if } Z = A \mathbf{ then } y = F(Z = A, X \setminus Z) \quad (15)$$

where $Z \subset X$ and A is a value combination of the variable set Z .²

Having a basic element (sub-circuit) realizing (15), we can implement a fuzzy device directly according to the system of fuzzy rules. However, note that equations (14) and (15) represent multi-valued functions in a piece-wise constant manner. An example of a 7-valued function is given in Fig.11 (a). Taking into account fuzzification and defuzzification procedures in fuzzy logic, corresponding multi-valued logic function should have at least piece-wise linear approximation between adjacent logical levels. Fig.11 (b) gives an example of such a representation of the function with evenly distributed logical values of the input and the output in the range of corresponding voltages.

² It is possible to add **else** in (15) that can be defined by circuit requirements.

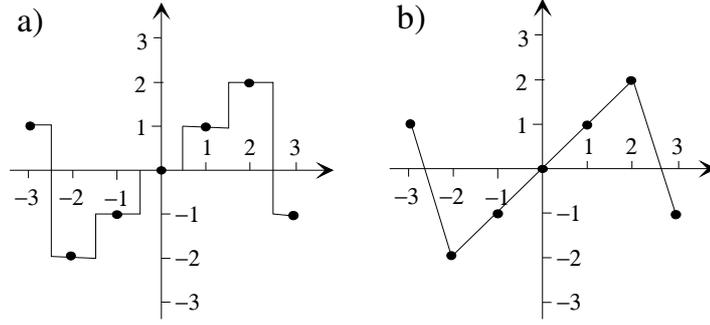


Figure 11. Example of a seven-valued function of one variable: piece-wise constant representation (a); peace-wise linear representation (b).

3.1 Masking summing amplifier input

Let us rewrite the definition (2) of the inverting summing amplifier with saturation in the following form:

$$S(A \cdot X; \beta) = \begin{cases} +k & \text{if } \sum_{i=1}^n \alpha_i \cdot x_i + \beta \leq -k \\ -\sum_{i=1}^n \alpha_i \cdot x_i - \beta & \text{if } +k > \sum_{i=1}^n \alpha_i \cdot x_i + \beta > -k \\ -k & \text{if } +k \leq \sum_{i=1}^n \alpha_i \cdot x_i + \beta \end{cases} \quad (16)$$

where $A = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ is a set of weight coefficients, $X = \{x_1, x_2, \dots, x_n\}$ is a set of analog or multi-valued variables, β is a constant symbolizing a threshold, $\pm k$ is a saturation value (in the case of m -valued logic, $m = 2k + 1$).

Let us introduce a masking function $M_\alpha(x)$ such that

$$M_\alpha(x) = \begin{cases} +k & \text{if } x \leq \alpha - 1 \\ k(\alpha - x) & \text{if } \alpha - 1 < x < \alpha + 1 \\ -k & \text{if } \alpha + 1 \leq x \end{cases} \quad (17)$$

where α ($-k \leq \alpha \leq k$) is a fixed value of the variable x . It can be easily seen that when $x = \alpha$, $M_\alpha(\alpha) = 0$. Fig.12 illustrates an example of the function $M_{-1}(x)$ for $m = 7$.

Taking into account that the source voltage V_{dd} has the logical value equal to $+k$ and the ground potential V_{gnd} has the logical value equal to $-k$, the mask-function can be easily implemented on the base of summing amplifier as

$$M_\alpha(x) = \begin{cases} S(kx; -\alpha \cdot V_{dd}) & \text{if } \alpha < 0 \\ S(kx) & \text{if } \alpha = 0 \\ S(kx; \alpha \cdot V_{gnd}) & \text{if } \alpha > 0 \end{cases} \quad (18)$$

where $x (V_{gnd} \leq x \leq V_{dd})$ is measured in voltages.

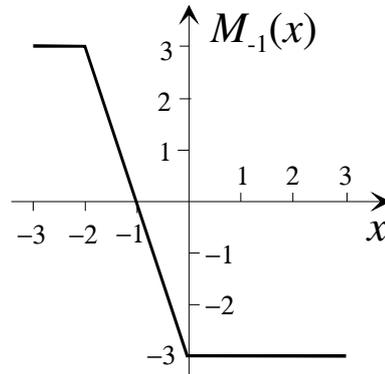


Figure 12. $M_{-1}(x)$ diagram for $m = 7$.

Using the mask-function $M_{\alpha}(x)$ it is possible to implement the rule

$$\text{if } x = \alpha \text{ then } y = F(x = \alpha, Y) \text{ else } y = 0, \quad (19)$$

which extracts the value of the function $F(x = \alpha, Y)$ in the point $x = \alpha$, as the circuit from summing amplifiers shown in Fig.13.

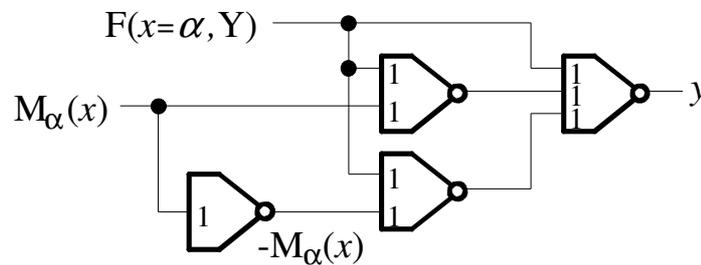


Figure 13. Implementation of the rule (19).

This implementation can be written in analytical form as

$$y = S\{S[M_{\alpha}(x); F(x = \alpha, Y)]; S[S(M_{\alpha}(x)); F(x = \alpha, Y)]; F(x = \alpha, Y)\} . \quad (20)$$

For example, in the case when $\alpha = -1$, $F(x = -1, Y) = 2$, and $m = 7$, the behavior of the circuit in Fig.13 can be represented as it is shown in Fig.14.

Analyzing the implementation of the rule (19) it is possible to see that in it the condition $x = \alpha$ is realized as the condition $M_{\alpha}(x) = 0$.

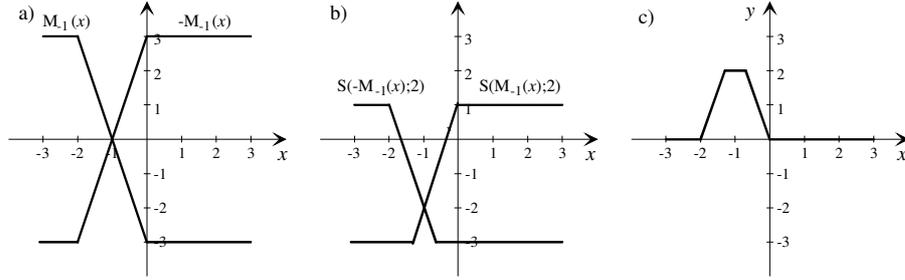


Figure 14. Example of implementing the rule (19).

3.2 Mask-functions of other types

To decrease the number of variables, which an m -valued logical function depends on, by one using the analog of Shannon's decomposition (14) we need to implement m rules of the type (19) and to find m components $F(x = \alpha, Y)$, $-k \leq \alpha \leq +k$ of the decomposition. Sometimes the number of rules can be reduced, if the function $F(x, Y)$ doesn't change on some interval of changing logical values of the variable x . A single rule can correspond to such interval of the variable x and the conditional part of this rule can have one of three views: $\alpha \leq x \leq \beta$, $x \leq \beta$, $\alpha \leq x$ where $-k \leq \alpha < \beta \leq +k$. For the condition $\alpha \leq x \leq \beta$ let us construct the following mask-function:

$$M_{\alpha, \beta}(x) = \begin{cases} +k & \text{if } x \geq \beta + 1 \\ -M_{\alpha-1}(x) - M_{\beta+1}(x) & \text{if } \alpha - 1 < x < \beta + 1 \\ -k & \text{if } x \leq \alpha - 1 \end{cases} \quad (21)$$

It is easy to see (Fig.15) that on the interval $\alpha \leq x \leq \beta$ this function takes the value 0.

In the case when $\alpha = -k$ or $\beta = k$, this mask-function will have one of the forms:

$$M_{-k, \beta}(x) = \begin{cases} +k & \text{if } x \geq \beta + 1 \\ +k - M_{\beta+1}(x) & \text{if } x < \beta + 1 \end{cases} \quad \text{or} \quad (22)$$

$$M_{\alpha, k}(x) = \begin{cases} -M_{\alpha-1}(x) - k & \text{if } \alpha - 1 < x \\ -k & \text{if } x \leq \alpha - 1 \end{cases} \quad (23)$$

and represent conditions $x \leq \beta$ or $\alpha \leq x$ respectively.

Mask-functions (21), (22), and (23) can be implemented on the base of summing amplifiers as

$$M_{\alpha, \beta}(x) = S[M_{\alpha-1}(x); M_{\beta+1}(x)], \quad (24)$$

$$M_{-k, \beta}(x) = S[V_{gnd}; M_{\beta+1}(x)], \quad (25)$$

$$M_{\alpha, k}(x) = S[M_{\alpha-1}(x); V_{dd}]. \quad (26)$$

Let us look how the masking can be performed for a wider scope of the variable changes, such as:

$$\text{if } \alpha \leq x \leq \beta \text{ then } y = F(\alpha \leq x \leq \beta, Y) = \Phi(Y) \text{ else } y = 0. \quad (27)$$

Using the mask-function $M_{\alpha,\beta}(x)$ it is possible to transform the rule (21) into the following form:

$$\text{if } M_{\alpha,\beta}(x) = 0 \text{ then } y = F(\alpha \leq x \leq \beta, Y) = \Phi(Y) \text{ else } y = 0. \quad (28)$$

The rule (28) can be implemented with the circuit shown in Fig.13, if to change in it the inputs $M_\alpha(x)$ and $F(x = \alpha, Y)$ with the inputs $M_{\alpha,\beta}(x)$ and $\Phi(Y)$ respectively.

Analytically this implementation is represented as

$$y = S\{S[M_{\lambda,\delta}(x); \Phi(Y)]; S[S(M_{\lambda,\delta}(x)); \Phi(Y)]; \Phi(Y)\}. \quad (29)$$

The sequence of pictures in Fig.15 illustrates the implementation of the rule

$$\text{if } -2 \leq x \leq +1 \text{ then } y = -2 \text{ else } y = 0$$

for the case of ($m = 9$)-valued logic.

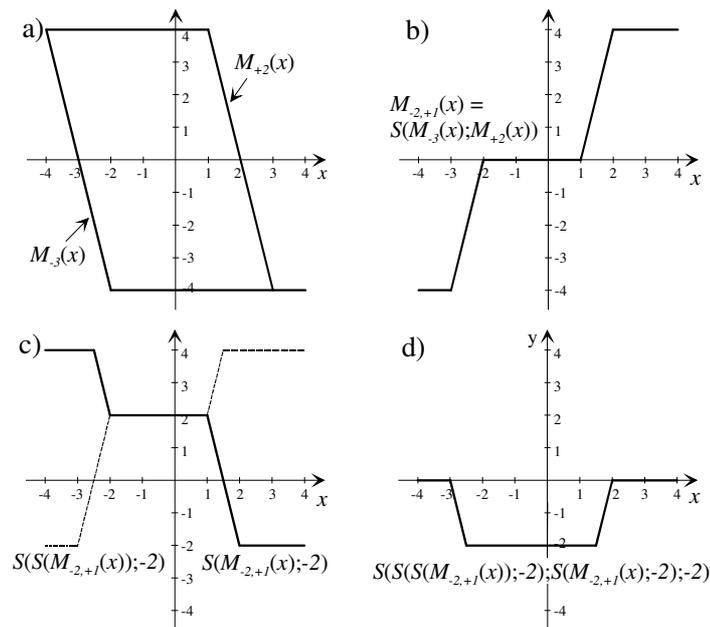


Figure 15. Example of mask-functions application.

3.3 An example of interval masking application

For further explaining the matter discussed in 3.2, let us recall an example from [4, pp. 123 – 128] “A Fuzzy Logic Force Controller for a Stepper Motor Robot”.

The fuzzy controller implements the function of two linguistic variables: *position error* and *force error*, which will be marked x and y respectively. The variables x and y each

have 7 linguistic values: NL, NM, NS, ZE, PS, PM, PL, and their membership functions are shown in Fig.16.

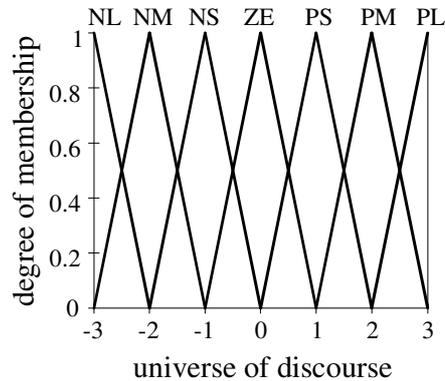


Figure 16. Fuzzy sets for *force error* and *position error* inputs.

The Inference Engine Rule Matrix for the output linguistic variable from the cited work looks as it is shown in Table 7.

Inference Engine Rule Matrix

Table 7

| | | position error (x) | | | | | | |
|---------------------|----|------------------------|----|----|----|----|----|----|
| | | NL | NM | NS | ZE | PS | PM | PL |
| force error (y) | NL | NM | NL | NL | NL | NL | NL | NM |
| | NM | NS | NM | NM | NM | NM | NM | NS |
| | NS | ZE | NS | NS | NS | NS | NS | ZE |
| | ZE | ZE | ZE | ZE | ZE | ZE | ZE | ZE |
| | PS | ZE | PS | PS | PS | PS | PS | ZE |
| | PM | PS | PM | PM | PM | PM | PM | PS |
| | PL | PM | PL | PL | PL | PL | PL | PM |

Let us transform the Table 7 into the Table 8 taking into account that we are going to produce fuzzy inference calculating values of the corresponding multi-valued logic function.

The Inference Engine Rule Matrix

as the multi-valued logic function

Table 8

| | | position error (x) | | | | | | |
|---------------------|----|------------------------|----|----|----|----|----|----|
| | | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
| force error (y) | -3 | -2 | -3 | -3 | -3 | -3 | -3 | -2 |
| | -2 | -1 | -2 | -2 | -2 | -2 | -2 | -1 |
| | -1 | 0 | -1 | -1 | -1 | -1 | -1 | 0 |
| | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | 1 | 0 | 1 | 1 | 1 | 1 | 1 | 0 |
| | 2 | 1 | 2 | 2 | 2 | 2 | 2 | 1 |
| | 3 | 2 | 3 | 3 | 3 | 3 | 3 | 2 |

Table 8 comprises only two different columns defining two functions depending on the variable *force error* (Table 9).

Two different functions of the *force error* **Table 9**

| | force error (y) | | | | | | |
|----------|-----------------|----|----|---|---|---|---|
| | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
| $F_1(y)$ | -2 | -1 | 0 | 0 | 0 | 1 | 2 |
| $F_2(y)$ | -3 | -2 | -1 | 0 | 1 | 2 | 3 |

Fig.17 illustrates graphs of these functions.

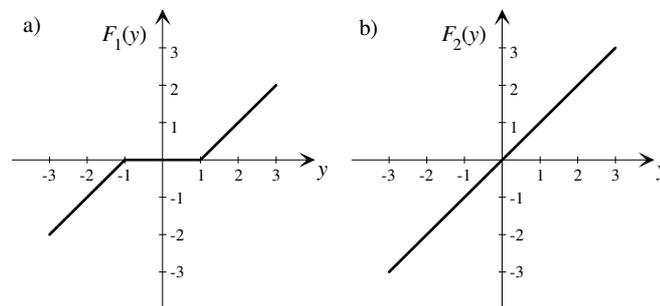


Figure 17. Components of the function defined by Table 7 and decomposed relative to variable x .

It is easy to see that the function $F_1(y)$ looks like mask-function $M_{-1,1}(y)$ but has different slopes of the lines. By analogy with (17), (18), (21), (24), Fig.15 (a), and Fig.15 (b), it is possible to construct the function $F_1(y)$ in accordance with graphics in Fig.18.

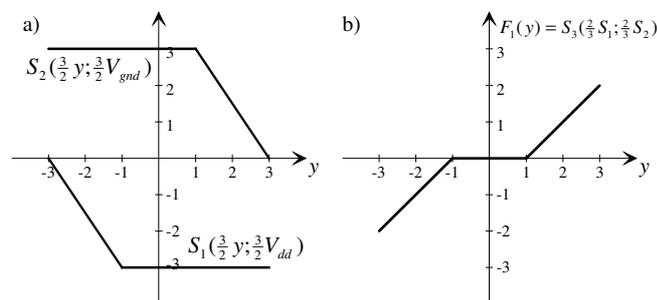


Figure 18. Constructing of the function $F_1(y)$.

As a result, the functions $F_1(y)$ and $F_2(y)$ can be implemented as

$$F_1(y) = S_3\left[\frac{2}{3} S_2\left(\frac{3}{2} y; \frac{3}{2} V_{gnd}\right); \frac{2}{3} S_1\left(\frac{3}{2} y; \frac{3}{2} V_{dd}\right)\right]; \quad F_2(y) = y. \quad (30)$$

It can be seen from the Table 7 and Table 8 that the behavior of the controller's output in the decomposition by variable x has the form:

$$\begin{aligned} &\text{if } NM \leq x \leq PM \text{ then } Output = F_2(y) \text{ else } Output = F_1(y) \text{ or} \\ &\text{if } -2 \leq x \leq +2 \text{ then } Output = F_2(y) \text{ else } Output = F_1(y). \end{aligned} \quad (31)$$

It is possible to split the rule (31) into two rules and represent them as:

$$\text{if } M_{-2,+2}(x) = 0 \text{ then } Output = F_2(y) \text{ else } Output = 0, \quad (32)$$

$$\text{if } M_{-2,+2}(x) \neq 0 \text{ then } Output = 0 \text{ else } Output = F_1(y). \quad (33)$$

The rule (32) can be implemented in accordance with (29) and (30) and (24) as

$$\begin{cases} M_{-2,+2}(x) = S_4(M_{-3}(x); M_{+3}(x)), \\ -M_{-2,+2}(x) = S_5(M_{-2,+2}(x)), \\ \Phi_1 = S_{11}\{S_6[M_{-2,+2}(x); F_2(y)]; S_7[-M_{-2,+2}(x); F_2(y)]; F_2(y)\}. \end{cases}$$

It is easy to check that the rule (33) can be implemented in accordance with the structural scheme shown in Fig.13, in which the output amplifier has the weight equal 2 of the input $F_1(y)$:

$$\Phi_2 = -S_{10}\{S_9[M_{-2,+2}(x); F_1(y)]; S_8[-M_{-2,+2}(x); F_1(y)]; 2F_1(y)\}.$$

Finally the output of the controller can be calculated as

$$Output = \Phi_1 + \Phi_2.$$

For producing this summation it is possible to use summing amplifier S_{11}

$$Output = S_{11}(-\Phi_1; -\Phi_2).$$

Fig.19 illustrates the structural scheme of the controller implementation with elements containing designations of input weights.

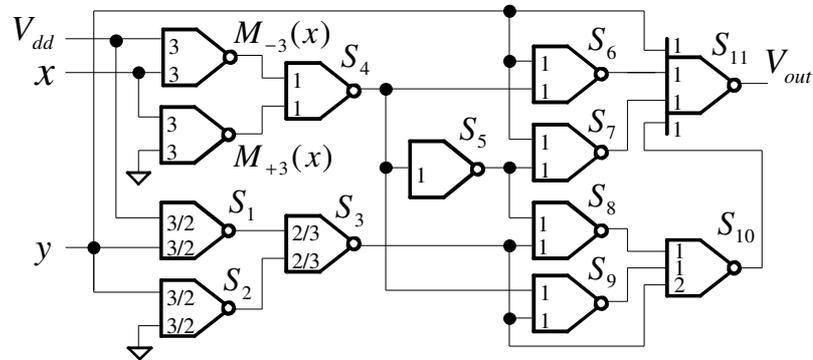


Figure 19. Structural diagram of the controller.

The controller circuit has been constructed from three-stage push-pull CMOS operational amplifiers with 1-MegOhm resistors in the feedback (Fig.2 (b)). It's functioning

has been checked with SPICE simulation (MSIM 8). MOSIS BSIM3v3.1, level 7 models of $0.4\mu\text{m}$ transistors have been used. In this paper, all other SPICE simulation experiments with designed circuits of controllers will be executed under the same conditions.

In the experiments with the controller presented in Fig.19, source voltage was 3.5V, input variable x changed linearly from 0V to 3.5V, input variable y changed discretely in accordance with its logical values and kept constant value within one cycle of x changing.

For the controller constructed from 3-stage elements results of SPICE simulation are shown in Fig.20. It is possible to see that the functioning of the controller is correct (logical values of the circuit output depend on the logical values of the input variables in accordance with the Table 2).

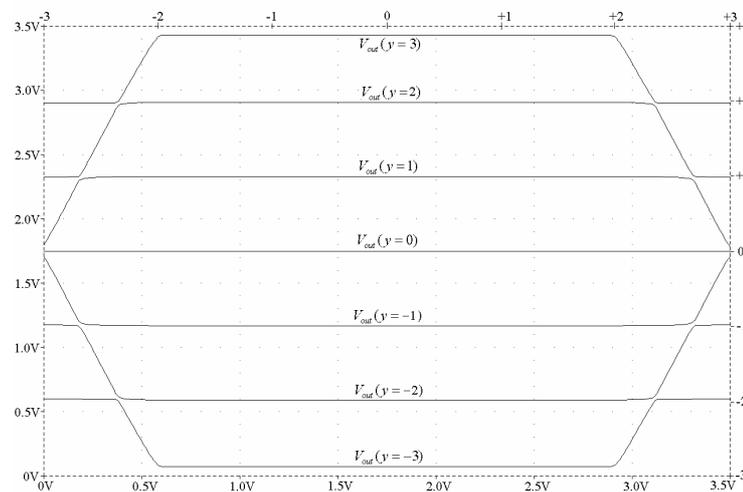


Figure 20. SPICE simulation results for the controller constructed from 3-stage summing amplifiers.

4 Particular Methods of Implementing Fuzzy Inference

The universal method of implementing multi-valued logic functions suggested in the previous section can be always used but often can give inappropriate results due to its universality. For this reason some particular design methods for fuzzy inference part of controllers were developed. These methods utilize specific properties of certain multi-valued logical function descriptions corresponding to sets of fuzzy inference rules.

According to the approach described above an initial set of fuzzy rules is represented in the form of a matrix or matrices defining multi-valued logical functions. Complex matrices

can not be directly implemented. They must be decomposed into component matrices with relatively simple configuration of elements allocation, for which rather simple implementations can be find. The topologies of valuable elements inside of such component matrices can be specified as symmetrical, diagonal, matrixes with linear configurations of elements, with elements located along rows and columns, matrices containing single valuable element and others.

The best way to introduce particular design methods is to show possible matrix decomposition into a set of implementable matrices on the base of real design example.

Let us take the description of the rather complex fuzzy controller from the patent [10] of Toyota Motors Corporation. The controller calculates a regeneration time decision coefficient R on the base on a differential pressure coefficient K_p and total fuel consumption Q_f . The set of fuzzy rules in terms of linguistic variables is represented in Table 10. Transformation of the input and output signals are performed in accordance with

Fuzzy rules for regeneration time decision coefficient R **Table 10**

| | | K_p | | | | | | |
|-------|----|-------|----|----|----|----|----|----|
| | | NB | NM | NS | ZO | PS | PM | PB |
| Q_f | NB | NB | NB | NM | NS | ZO | PB | PB |
| | NM | NB | NM | NS | NS | ZO | PB | PB |
| | NS | NM | NM | NS | ZO | ZO | PS | PB |
| | ZO | NM | NS | ZO | ZO | PS | PM | PB |
| | PS | NS | ZO | ZO | PS | PS | PM | PB |
| | PM | ZO | ZO | PS | PM | PM | PM | PB |
| | PB | PS | PS | PM | PM | PM | PB | PB |

Analysis of the membership functions [10] of linguistic variables representing input and output analog signals shows that the linguistic variables having maximum weight are evenly distributed within the change ranges of corresponding analog signals. It means that without loosing the accuracy of representation, these linguistic variables can be replaced with logical values as it is shown in Table 11. In this table, signals Q_f and K_p are changed with 7-valued logic variables x and y respectively.

The 7-valued logical function **Table 11**

| $x \setminus y$ | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
|-----------------|----|----|----|----|---|---|---|
| -3 | -3 | -3 | -1 | -1 | 0 | 3 | 3 |
| -2 | -3 | -2 | -1 | -1 | 0 | 3 | 3 |
| -1 | -2 | -2 | -1 | 0 | 0 | 1 | 3 |
| 0 | -2 | -1 | 0 | 0 | 1 | 2 | 3 |
| 1 | -1 | 0 | 0 | 1 | 1 | 2 | 3 |
| 2 | 0 | 0 | 1 | 2 | 2 | 2 | 3 |
| 3 | 1 | 1 | 2 | 2 | 2 | 3 | 3 |

4.1 Extracting a Symmetrical Component Matrix

Let the Table 11 of the controller is represented as initial matrix M , which, in its turn, can be represented as sum of two component matrixes (M_1 and M_2). Matrix M_1 corresponds to a symmetrical component and M_2 corresponds to nonsymmetrical residual component.

$$\begin{array}{c}
 M \\
 \left| \begin{array}{l} -3-3-2-1\pm 0+3+3 \\ -3-2-1-1\pm 0+3+3 \\ -2-2-1\pm 0\pm 0+1+3 \\ -2-1\pm 0\pm 0+1+2+3 \\ -1\pm 0\pm 0+1+1+2+3 \\ \pm 0\pm 0+1+2+2+2+3 \\ +1+1+2+2+2+3+3 \end{array} \right|
 \end{array}
 =
 \begin{array}{c}
 M_1 \\
 \left| \begin{array}{l} -3-3-2-2-1\pm 0\pm 0 \\ -3-2-2-1\pm 0\pm 0+1 \\ -2-2-1\pm 0\pm 0+1+1 \\ -2-1\pm 0\pm 0+1+1+2 \\ -1\pm 0\pm 0+1+1+2+2 \\ \pm 0\pm 0+1+1+2+2+3 \\ \pm 0+1+1+2+2+3+3 \end{array} \right|
 \end{array}
 +
 \begin{array}{c}
 M_2 \\
 \left| \begin{array}{l} \leq 0 \leq 0 \pm 0 + 1 + 1 + 3 + 3 \\ \leq 0 \pm 0 + 1 \pm 0 \pm 0 + 3 \geq 2 \\ \pm 0 \pm 0 \pm 0 \pm 0 \pm 0 \geq 2 \\ \pm 0 \pm 0 \pm 0 \pm 0 + 1 \geq 1 \\ \pm 0 \pm 0 \pm 0 \pm 0 \pm 0 \geq 1 \\ \pm 0 \pm 0 \pm 0 + 1 \pm 0 \pm 0 \geq 0 \\ + 1 \pm 0 + 1 \pm 0 \pm 0 \geq 0 \geq 0 \end{array} \right|
 \end{array}$$

Matrix M_1 is symmetrical relative to the side diagonal. Its components can be represented as a function $f_1(z)$ of one variable $z = (x + y)/2$. After performing the linear approximation between adjacent logical levels this function will have the form shown in Fig.21.

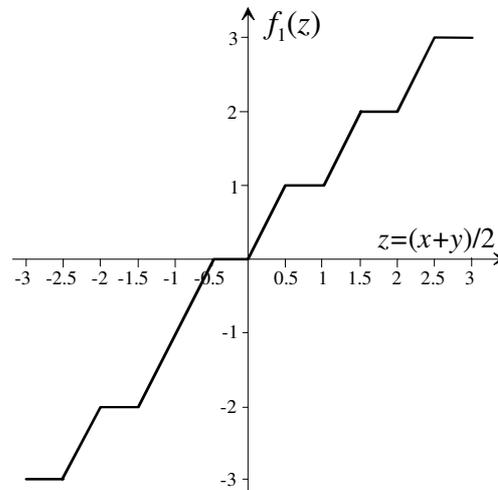


Figure 21. The function $f_1(z)$.

To implement the function $f_1(z)$ let us represent it as a sum of 5 subfunctions $\alpha_j(z)$ shown in Fig.22.

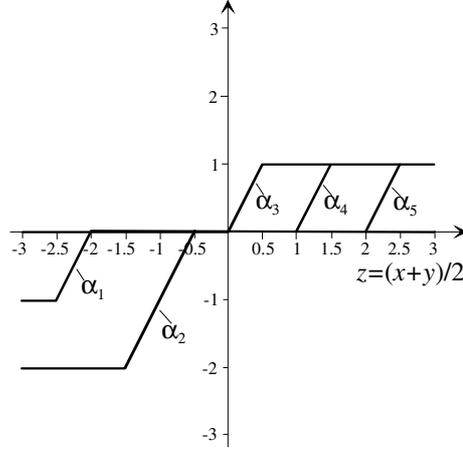


Figure 22. Five components of the function $f_1(z)$.

It is easy to see, that

$$f_1(z) = \sum_{j=1}^5 \alpha_j(z) \quad (34)$$

and let us consider formation $\alpha_j(z)$ using summing amplifiers on the example of $\alpha_1(z)$.

For this let us address to Fig.23.

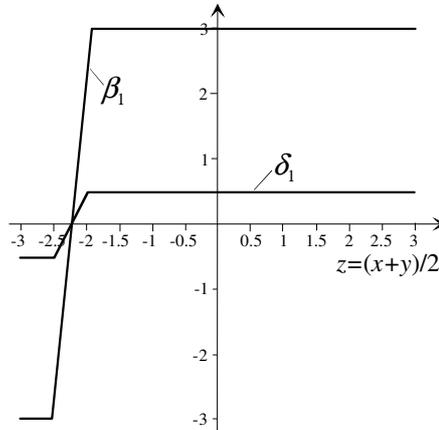


Figure 23. Representation of the function $\alpha_1(z)$.

The function $\beta_1(z)$ in Fig.23 can be implemented as $\beta_1(z) = -S(k_1 \cdot z + a_1)$. For $z = -2.25$ $\beta_1 = 0$ then $a_1 = 2.25 \cdot k_1$. Taken into account that $k_1 = 6/0.5 = 12$, we receive $a_1 = 27$ and $\beta_1(x, y) = -S(6 \cdot x + 6 \cdot y + 27)$, $\delta_1(x, y) = \frac{1}{6} \cdot \beta_1(x, y)$. Finally

$$\alpha_1(x, y) = \delta_1(x, y) - 0.5 = \frac{1}{6} \cdot \beta_1(x, y) - 0.5 \quad (35)$$

In the same way it is possible to find

$$\begin{aligned}
\alpha_2(x, y) &= \frac{1}{3}\beta_2(x, y) - 1 = -\frac{1}{3}S(3 \cdot x + 3 \cdot y + 6) - 1 \\
\alpha_3(x, y) &= \frac{1}{6}\beta_3(x, y) + 0.5 = -\frac{1}{6}S(6 \cdot x + 6 \cdot y - 3) + 0.5 \\
\alpha_4(x, y) &= \frac{1}{6}\beta_4(x, y) + 0.5 = -\frac{1}{6}S(6 \cdot x + 6 \cdot y - 15) + 0.5 \\
\alpha_5(x, y) &= \frac{1}{6}\beta_5(x, y) + 0.5 = -\frac{1}{6}S(6 \cdot x + 6 \cdot y - 27) + 0.5
\end{aligned} \tag{36}$$

The value of $f_1(z)$ can be calculated in accordance with (34) on one summing amplifier. Finally taking into account mutual compensation of constants, we have

$$f_1(x, y) = S\{-\frac{1}{6} \cdot [\beta_1(x, y) + 2 \cdot \beta_2(x, y) + \beta_3(x, y) + \beta_4(x, y) + \beta_5(x, y)]\}. \tag{37}$$

Thus, the implementation of the function $f_1(x, y)$, which represents the matrix M_1 , consists of six summing amplifiers.

4.2 Extracting a Matrix with Elements Separated by a line

This method is applicable for realization of matrices composed from two types of elements which can be separated with a line.

After extracting the symmetrical component the residual matrix is M_2 . This matrix has some elements of types $\leq a$ and $\geq b$. This means that instead of values a and b of the elements it is possible to substitute any logical value less then a and more then b respectively. Let us split the matrix M_2 in two matrices (M_3 and M_4) and try to implement the matrix M_3 . The matrix M_4 is a new residual matrix.

$$\begin{array}{c}
M_2 \\
\left| \begin{array}{cccccc}
\leq 0 \leq 0 & 0+1+1 & +3+3 \\
\leq 0 & 0+1 & 0 & 0 & +3 \geq 2 \\
0 & 0 & 0 & 0 & 0 & 0 \geq 2 \\
0 & 0 & 0 & 0 & 0 & +1 \geq 1 \\
0 & 0 & 0 & 0 & 0 & 0 \geq 1 \\
0 & 0 & 0+1 & 0 & 0 \geq 0 \\
+1 & 0+1 & 0 & 0 \geq 0 \geq 0
\end{array} \right| =
\begin{array}{c}
M_3 \\
\left| \begin{array}{cccccc}
0 & 0 & 0 & 0 & 0+3+3 \\
0 & 0 & 0 & 0 & 0+3+3 \\
0 & 0 & 0 & 0 & 0 & 0+3 \\
0 & 0 & 0 & 0 & 0 & 0+3 \\
0 & 0 & 0 & 0 & 0 & 0+3 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0
\end{array} \right| +
\begin{array}{c}
M_4 \\
\left| \begin{array}{cccccc}
\leq 0 \leq 0 & 0+1+1 \geq 0 \geq 0 \\
\leq 0 & 0+1 & 0 & 0 \geq 0 \geq 0 \\
0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
0 & 0 & 0 & 0 & 0 & +1 \geq 0 \\
0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
0 & 0 & 0+1 & 0 & 0 \geq 0 \\
+1 & 0+1 & 0 & 0 \geq 0 \geq 0
\end{array} \right|
\end{array}$$

Let us address to Fig.24. It is easy to see that elements of the matrix M_3 with two different values can be separated with help of two parallel lines: $x - 3y + 8 = 0$ and $x - 3y + 8 = 1$. Introduce a new variable

$$w = x - 3y + 8. \tag{38}$$

Value of the variable w in the point with coordinates (x, y) is proportional to the distance of this point from the line. In all points lying on and up of the line $w \leq 0$, and in all points laying on and down of the dashed line ($x - 3y + 7 = 0$) $w \geq 1$.

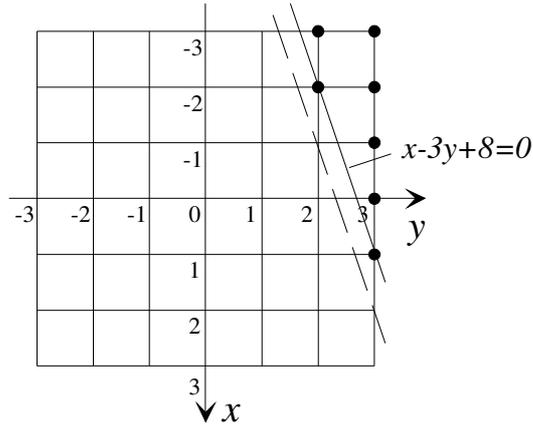


Figure 24. Separating valuable elements of the matrix M_3 .

It is easy to see, that the matrix M_3 representing the function $f_2(x, y)$ can be implemented as

$$f_2(x, y) = S\{S[3(-x + 3y - 8)] - 3\}. \quad (39)$$

In this implementation all valuable matrix elements are equal to “3”.

4.3 Extracting a Matrix with Rectangular Configuration of Valuable Elements

Let us introduce a *Pyramid Function* that is the function, which corresponds to a matrix with a single valuable element and represents rules of the type

$$\text{if } (x = a) \ \& \ (y = b) \ \text{then } f(x, y) = k \ \text{else } f(x, y) = 0. \quad (40)$$

This function is shown in Fig.25.

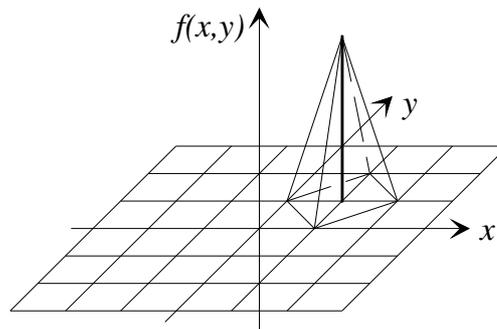


Figure 25. A pyramid function.

The Pyramid Function has some fixed value c at the point (a, b) so that at the rest of the space bordered by points neighboring to (a, b) this function is zero and transition from c to zero is linear. Neighborhood is defined by coordinate increments $\Delta_x = \pm 1$ and $\Delta_y = \pm 1$.

Let's turn to Fig.26 to construct the pyramid function.

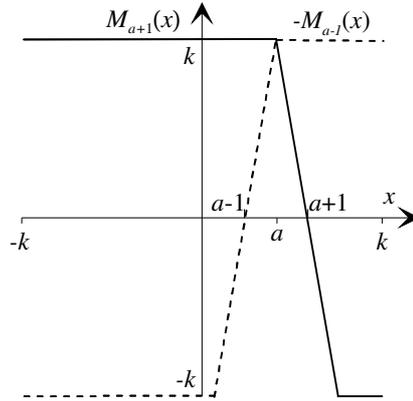


Figure 26. Component functions of the pyramid in Fig.26.

The figure shows two component mask-functions $M_{a+1}(x)$ and $-M_{a-1}(x)$ those are implemented for $(2 \cdot k + 1)$ -valued logic $(-k \leq x \leq +k)$ as:

$$\begin{aligned} M_{a+1}(x) &= S[k \cdot (x - a - 1)], \\ -M_{a-1}(x) &= S[k \cdot (-x + a - 1)]. \end{aligned} \quad (41)$$

Similar component functions are constructed for the variable y :

$$\begin{aligned} M_{a+1}(y) &= S[k \cdot (y - b - 1)], \\ -M_{a-1}(y) &= S[k \cdot (-y + b - 1)]. \end{aligned} \quad (42)$$

Simple substitution yields that the function

$$\gamma(x) = S[M_{a+1}(x) - M_{a-1}(x) - 2k] \quad (43)$$

has the form shown in Fig.27; it equals to “0” for $x = a$ and equals to “+k” for all other integer argument values.

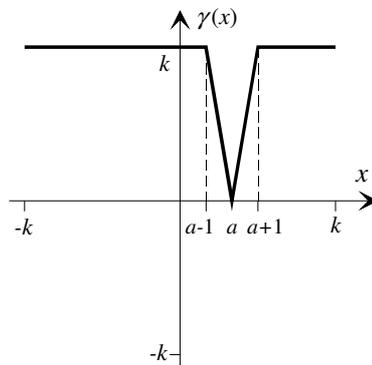


Figure 27. Graph of the function $\gamma(x)$.

In a similar manner, we can construct the following function of 2 variables:

$$\gamma(x, y) = S[M_{a+1}(x) - M_{a-1}(x) + M_{b+1}(y) - M_{b-1}(y) - 4k]. \quad (44)$$

Function $\gamma(x, y)$ equals to “0” for $(x = a) \& (y = b)$ and equals to “+k” for all other points.

Now it is easy to construct the pyramid of Fig.26 with height “k”:

$$f(x, y) = S[\gamma(x, y) - k]. \quad (45)$$

Pyramid of an arbitrary height is obtained by simple input gain factor scaling of the next amplifier. The pyramid sign can be elementarily changed at the stage of component mask-functions constructions.

We anticipate some complications in the case when it is needed to receive good "sewing" pyramids with already implemented functions. The pyramid function $f(x, y)$ (45) of Fig.25 type has projections on flats x and y shown in Fig.28.

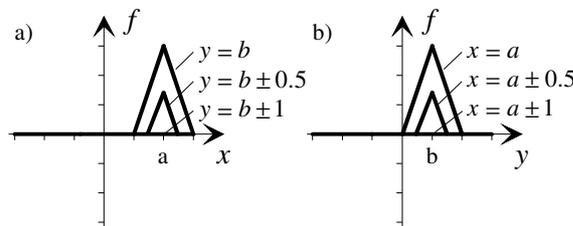


Figure 28. Graph of the pyramid function by coordinate x (a) and by coordinate y (b).

When “sewing” a pyramid function with other functions to get monotonous piece-linear approximation between adjacent logical values the view of the pyramid function by one of its coordinate can be changed to one of variants shown in Fig.29.

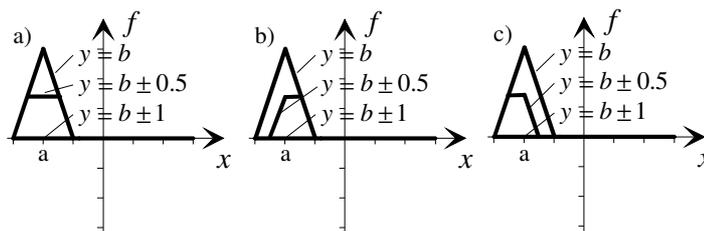


Figure 29. Possible graphs of a pyramid function by one of coordinates: center trapeze (a), right trapeze (b), and left trapeze (c).

To construct a pyramid function with projections onto flats x and y of the Fig.29 (a) type (center trapeze), it is sufficient to substitute in the functions (41) – (45) instead of original variables x, y new variables $z = (x + y)/2$, $w = (x - y)/2$ and instead of points a, b

points $c = (a + b)/2$, $d = (a - b)/2$ respectively. It means transition to the pyramid function shown in Fig.30, which is

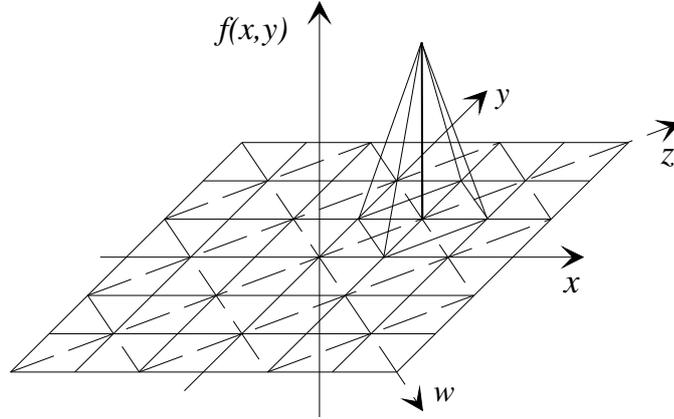


Figure 30. A pyramid in coordinates $z = (x + y)/2$ and $w = (x - y)/2$.

implemented by analogy with (41) – (45) as

$$f(x, y) = S\{S[\frac{1}{2}S[k(x + y - a - b - 1)] + \frac{1}{2}S[k(-x - y + a + b - 1)] + \frac{1}{2}S[k(x - y - a + b - 1)] + \frac{1}{2}S[k(-x + y + a - b - 1) - 2k] - k\}. \quad (46)$$

For implementing a function, which has the projection on one of its argument flats of the right trapeze or left trapeze type (Fig.30 (b) or Fig.30 (c)), let us introduce two intermediate functions $\varphi[x, \xi(y)]$ and $\psi[x, \xi(y)]$ shown in Fig.31.

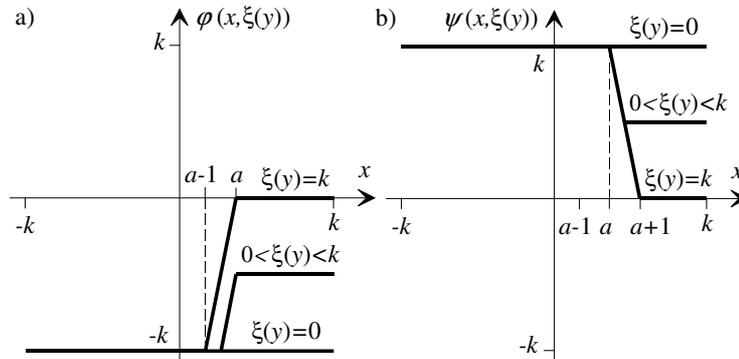


Figure 31. Intermediate functions $\varphi[x, \xi(y)]$ (a) and $\psi[x, \xi(y)]$ (b).

It is easy to see that the function $f(x, y) = \varphi[x, \xi(y)] + \psi[x, \xi(y)]$ has the right-hand trapeze projection on the flat x (Fig.29 (b)). If the function $\xi(y)$ has a triangle form, $y = b$, and $\xi(b) = k$, the function $f(x, y)$ is a pyramid function. The way of constructing the function $f(x, y)$ with left-hand trapeze projection on the flat x (Fig.29 (c)) now is obvious.

It is not difficult to check that the functions $\varphi(x, \xi(y))$ and $\psi(x, \xi(y))$ can be implemented as

$$\varphi[x, \xi(y)] = S[M_{a-1}(x) + 2k - \varphi(y)], \quad (47)$$

$$\psi[x, \xi(y)] = S[-M_{a+1}(x) + 2k - \varphi(y)] + 2k - \varphi(y). \quad (48)$$

The pyramid function approach is not limited to rules with point condition (40) and may be extended to rules with interval condition of the type

$$\text{if } (a_1 \leq x \leq a_2 \text{ and } b_1 \leq y \leq b_2) \text{ then } f(x, y) = k \text{ else } f(x, y) = 0. \quad (49)$$

Interval conditions can be implemented by simple changing constants for mask-functions. It is easy to see that the rule (49) represents matrices with rectangular configurations of valuable elements those can be implemented as truncated square pyramids.

Note that the function similar to (44) may be constructed for an arbitrary variables number. Pyramid implementation for two variables requires 6 amplifiers. Each additional variable introduction requires two additional amplifiers.

4.4 Extracting a Matrix with Valuable Elements Laying on a Diagonal

Let us split the Matrix M_4 on two matrixes (M_5 and M_6) and try to implement the matrix M_5 . Matrix M_6 is the next residual matrix.

$$\begin{array}{c} M_4 \\ \left| \begin{array}{cccccc} \leq 0 \leq 0 & 0+1+1 \geq 0 \geq 0 \\ \leq 0 & 0+1 & 0 & 0 \geq 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & +1 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ 0 & 0 & 0+1 & 0 & 0 \geq 0 \\ +1 & 0+1 & 0 & 0 \geq 0 \geq 0 \end{array} \right| = \begin{array}{c} M_5 \\ \left| \begin{array}{cccccc} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0+1 \\ 0 & 0 & 0 & 0 & 0+1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0+1 & 0 & 0 & 0 \\ 0 & 0+1 & 0 & 0 & 0 & 0 \end{array} \right| + \begin{array}{c} M_6 \\ \left| \begin{array}{cccccc} \leq 0 \leq 0 & 0+1+1 \geq 0 \geq 0 \\ \leq 0 & 0+1 & 0 & 0 \geq 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\ +1 & 0 & 0 & 0 & 0 \geq 0 \geq 0 \end{array} \right| \end{array}$$

In its turn, the matrix M_5 can be composed from two matrices

$$\begin{array}{c} M_5 \\ \left| \begin{array}{cccccc} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0+1 \\ 0 & 0 & 0 & 0 & 0+1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0+1 & 0 & 0 & 0 \\ 0 & 0+1 & 0 & 0 & 0 & 0 \end{array} \right| = \frac{1}{3} \begin{array}{c} M_{51} \\ \left| \begin{array}{cccccc} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0+3 \\ 0 & 0 & 0 & 0 & 0+3 & 0 \\ 0 & 0 & 0 & 0+3 & 0 & 0 \\ 0 & 0 & 0+3 & 0 & 0 & 0 \\ 0 & 0+3 & 0 & 0 & 0 & 0 \end{array} \right| + \frac{1}{3} \begin{array}{c} M_{52} \\ \left| \begin{array}{cccccc} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right| \end{array}$$

and represented as the sum of corresponding functions $f_3(x, y) = \frac{1}{3}[f_{31}(x, y) + f_{32}(x, y)]$.

In the matrix M_{51} , elements with the value “+3” lay on the line $x + y - 2 = 0$. This matrix can be described by the function of one variable $f_{31}[z = (x + y)/2]$, which is defined by the rule

$$\text{if } z = 1 \text{ then } f_{31}(z) = 3 \text{ else } f_{31}(z) = 0.$$

The function $f_{31}(z)$ can be constructed as it is shown in Fig.32.

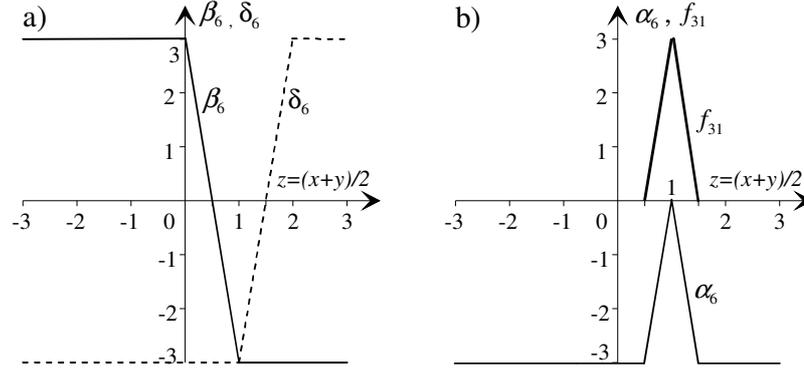


Figure 32. Functional representations of the matrixes M_5 .

It is easy to see from this figure that

$$f_{31}(z) = \alpha_6(z) + 3, \quad \alpha_6(z) = S[\beta_6(z) + \delta_6(z) + 6],$$

$$\beta_6(z) = S(6z - 3), \quad \delta_6(z) = S(-6z + 9).$$

For good “sewing” the function f_{32} with f_{31} and the function f_3 with f_1 the function f_{32} has to be implemented as pyramid function of variables $z = (x + y)/2$ and $w = (x - y)/2$ as it is shown in Fig.30. Substitution variables z and w in formula (46) instead of $(x + y)/2$ and $(x - y)/2$ respectively, $a = +1$, $b = +1$, $k = 3$, and changing the sign of the function gives

$$f_{32}(z, w) = S\{S[\frac{1}{2}\beta_6(z) + \frac{1}{2}\delta_6(z) + \frac{1}{2}\beta_7(w) + \frac{1}{2}\delta_7(w) + 6] + 3\}$$

where $\beta_6(z)$, $\delta_6(z)$ are already implemented and

$$\beta_7(w) = S(6w + 3), \quad \delta_7(w) = S(-6w + 3).$$

Finally, the function implementation

$$f_3(x, y) = \frac{1}{3}S\{\beta_6(x, y) + \delta_6(x, y) + S[\frac{1}{2}\beta_6(x, y) + \frac{1}{2}\delta_6(x, y) + \frac{1}{2}\beta_7(x, y) + \frac{1}{2}\delta_7(x, y) + 6] + 9\} + 1, \quad (50)$$

where

$$\beta_6(x, y) = S(3x + 3y - 3), \quad \delta_6(x, y) = S(-3x - 3y + 9),$$

$$\beta_7(x, y) = S(3x - 3y + 3), \quad \delta_7(-3x + 3y + 3)$$

corresponds to the matrix M_5 .

4.5 Implementation of the Matrix M_6

Let us split the matrix M_6 on two matrices (M_7 and M_8) and try to implement the matrix M_7 . The new residual matrix is M_8 , all elements of which are defined.

$$\begin{array}{c}
 M_6 \\
 \left| \begin{array}{cccccc}
 \leq 0 \leq 0 & 0+1+1 \geq 0 \geq 0 \\
 \leq 0 & 0+1 & 0 & 0 \geq 0 \geq 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \geq 0 \\
 +1 & 0 & 0 & 0 & 0 \geq 0 \geq 0
 \end{array} \right| = \begin{array}{c}
 M_7 \\
 \left| \begin{array}{cccccc}
 0 & 0 & 0+1+1+1+1 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right| + \begin{array}{c}
 M_8 \\
 \left| \begin{array}{cccccc}
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0+1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 \\
 +1 & 0 & 0 & 0 & 0 & 0
 \end{array} \right|
 \end{array}$$

The matrix M_7 has one rectangular component and can be represented as the function

$f_4(x, y)$ that is defined by the rule

$$\text{if } (x = -3) \ \& \ (y \geq 0) \ \text{then } f_4(x, y) = 1 \ \text{else } f_4(x, y) = 0.$$

By analogy with constructing formulas (41) – (45) let us compose two auxiliary functions $\gamma_1(x)$ for the condition $x = -3$ and $\gamma_1(y)$ for the condition $y \geq 0$. These functions are represented in Fig.33 and can be constructed as

$$\gamma_1(x) = S[-M_{-3}(x)], \quad \gamma_1(y) = S[M_{-1}(y) + 3].$$

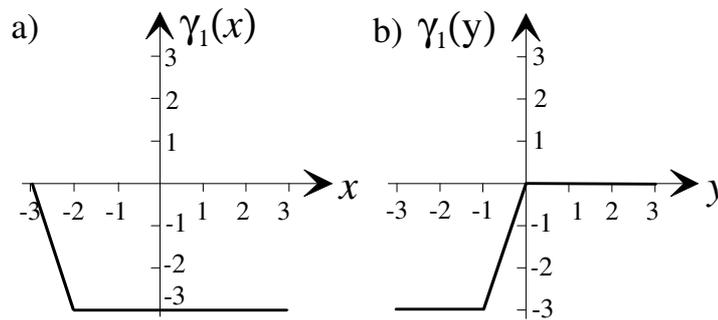


Figure 33. Two auxiliary functions: $\gamma_1(x)$ for the condition $x = -3$ (a) and $\gamma_1(y)$ for the condition $y \geq 0$ (b).

Therefore

$$\gamma_1(x, y) = S[-M_{-3}(x) + M_{-1}(y) + 3].$$

Taking into account that for $k = 3$

$$-M_{-3}(x) = S(-3x - 9), \quad M_{-1}(y) = S(3y + 3)$$

it is not difficult to find

$$f_4(x, y) = \frac{1}{3} S[S(-3x - 9) + S(3y + 3) + 3] + 1. \quad (51)$$

The residual matrix M_8 has only two nonzero elements with coordinates $(x = +3, y = -3)$ and $(x = -2, y = -1)$. Let us designate the functions, with help of which the controller function can be corrected in these points, as f_5 and f_6 , respectively. Depending on coordinates of valuable element and conditions of good coupling the element with already implemented fragments different implementation methods can be used.

The function $f_5(x, y)$ is equal to 0 everywhere except the point $(x = +3, y = -3)$, at which $f_5(x, y) = 1$. For monotonic piece-wise linear coupling this function with the function $f_1(x, y)$ its projections on the flats x and y must have the form of Fig.29 (a) or (b). The function $f_5(x, y)$ can be implemented as pyramid function in accordance with (46) but the following approach, which is shown in Fig.34, gives better implementation.

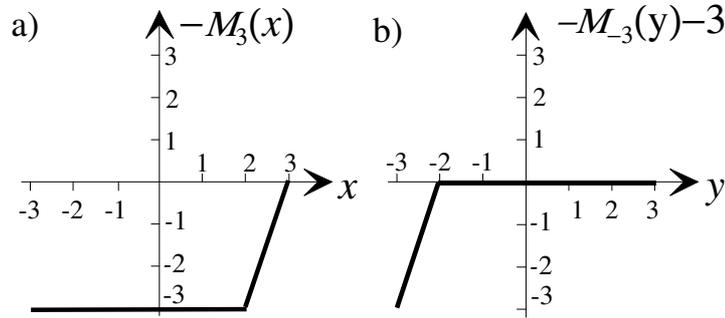


Figure 34. Auxiliary functions: $-M_3(x)$ (a) and $M_{-3}(y) - 3$ (b).

Now it is not difficult to construct the function

$$f_5(x, y) = \frac{1}{3} \{S[-M_{-3}(y) - 3 - M_3(x)] - M_3(x)\}$$

and finally the function f_5 is implemented as

$$f_5(x, y) = \frac{1}{3} S[S(-3y - 9) + S(-3x + 9) - 3] + \frac{1}{3} S(-3x + 9). \quad (52)$$

As experiments showed, the monotonic piece-wise linear approximation between the logical level in the point $(-2, -1)$ and logical levels in the adjacent points of the functions $f_1(x, y)$ and $f_4(x, y)$ will be obtained, if the pyramid function f_6 in the point $(-2, -1)$ is implemented in accordance with the formula (46).

$$f_6(x, y) = \frac{1}{3} S \left\{ S \left[\frac{1}{2} S(3x + 3y + 6) + \frac{1}{2} S(-3x - 3y - 12) + \frac{1}{2} S(3x - 3y) + \frac{1}{2} S(-3x + 3y - 6) - 6 \right] - 3 \right\}.$$

After some transformations providing the possibility to save one summing amplifier this function looks as following

$$f_6(x, y) = \frac{1}{3} S \left[\frac{1}{2} S(-3x - 3y - 6) + \frac{1}{2} S(3x + 3y + 12) + \frac{1}{2} S(-3x + 3y) + \frac{1}{2} S(3x - 3y + 6) + 6 \right] + 1. \quad (53)$$

4.6 Implementation of the Controller

Correctness of the designed controller and its functioning has been checked with SPICE simulation (MSIM 8). In simulation experiments MOSIS BSIM3v3.1 level 7 models of 0.4µm transistors have been used.

As building blocks for the controller circuit two types of summing amplifiers were used (ordinary and powerful). They are built on the base of three-stage push-pull CMOS operational amplifier with 1.5-MegOhm resistor in the feedback. Examples of such summing amplifiers are shown in Fig.35.

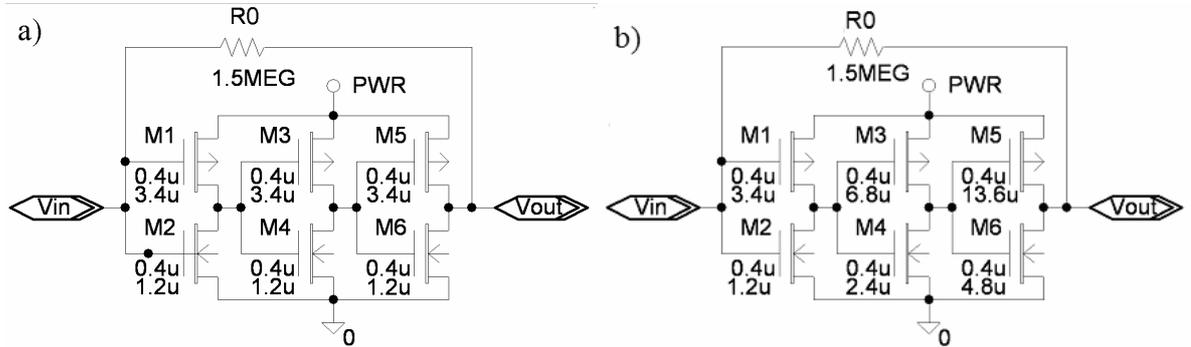


Figure 35. Operational amplifier with feedback: ordinary (a), powerful (b).

Transistors in this figure are marked with two numbers, which designate transistor dimensions (width and length).

The controller schematic for experiments is represented in Fig.36. Two powerful elements PHS1 and PHS2 of the controller produce signals $-y$ and $-x$ respectively. Other elements are ordinary.

Analytical description of the controller schematic can be derived on the base of the function implementations (35) – (37), (39), (50) – (53) and has the following form

$$f_1(x, y) = S_9 \left[\frac{1}{6} S_3(6x + 6y + 27) + \frac{1}{3} S_4(3x + 3y + 6) + \frac{1}{6} S_5(6x + 6y - 3) + \frac{1}{6} S_6(6x + 6y - 15) + \frac{1}{6} S_7(6x + 6y - 27) \right],$$

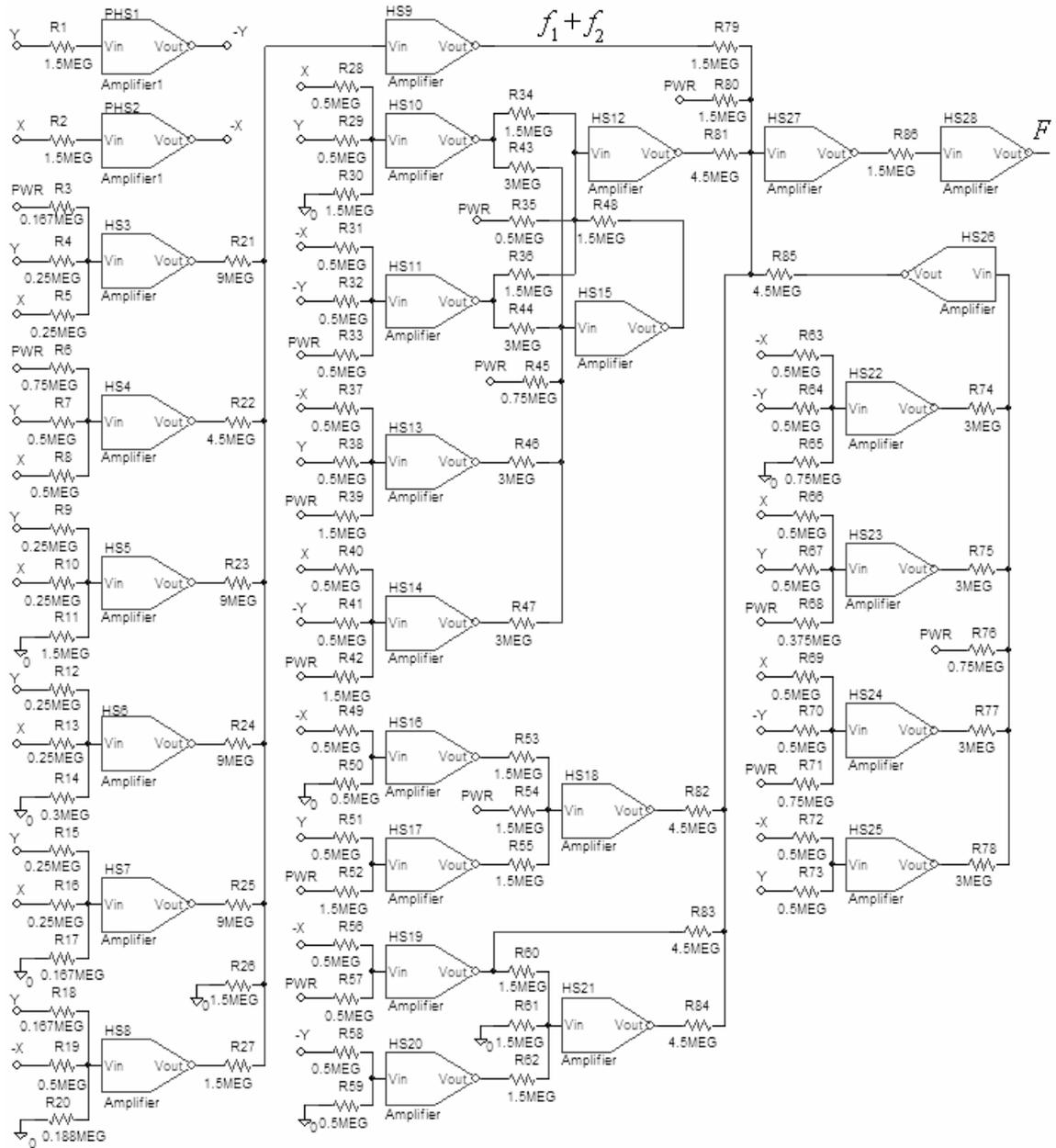


Figure 36. The controller schematic for experiments.

$$f_2(x, y) = S_9(S_8(-3x + 9y - 24) - 3);$$

$$f_3(x, y) = \frac{1}{3} S_{12} \{ S_{10}(3x + 3y - 3) + S_{11}(-3x - 3y + 9) + S_{15}[\frac{1}{2} S_{10}(3x + 3y - 3) + \frac{1}{2} S_{11}(-3x - 3y + 9) + \frac{1}{2} S_{14}(3x - 3y + 3) + \frac{1}{2} S_{13}(-3x + 3y + 3) + 6] + 9 \} + 1;$$

$$f_4(x, y) = \frac{1}{3} S_{18} [S_{16}(-3x - 9) + S_{17}(3y + 3) + 3] + 1;$$

$$f_5(x, y) = \frac{1}{3} S_{21} [S_{19}(-3x + 9) + S_{20}(-3y - 9) - 3] + \frac{1}{3} S_{19}(-3x - 9);$$

$$f_6(x, y) = \frac{1}{3} S_{26} [\frac{1}{2} S_{23}(-3x - 3y - 6) + \frac{1}{2} S_{22}(3x + 3y + 12) + \frac{1}{2} S_{25}(-3x + 3y) + \frac{1}{2} S_{24}(3x - 3y + 6) + 6] + 1;$$

$$F(x, y) = \sum_{j=1}^6 f_j(x, y) = S_{28} [S_{27} (S_9 + \frac{1}{3}S_{12} + \frac{1}{3}S_{18} + \frac{1}{3}S_{19} + \frac{1}{3}S_{21} + \frac{1}{3}S_{26} + 3)].$$

Enumeration of summing amplifiers in this description corresponds to enumeration of elements in the controller circuit. The controller contains 28 amplifiers and 86 resistors. Resistor values have been calculated as $R_j = R_0 / w_j$ where w_j is logical weight of the j -th element input signal.

In experiments, source voltage was 3.5V. Input variable x changed linearly from 0V to 3.5V, input variable y changed discretely and kept constant value within one cycle of x changing. The voltage range was evenly divided onto seven logical levels so that the logical levels “-3” and “+3” corresponded to voltages V_{gnd} and V_{dd} respectively.

Results of SPICE simulation of the controller schematic are represented in Fig.37.

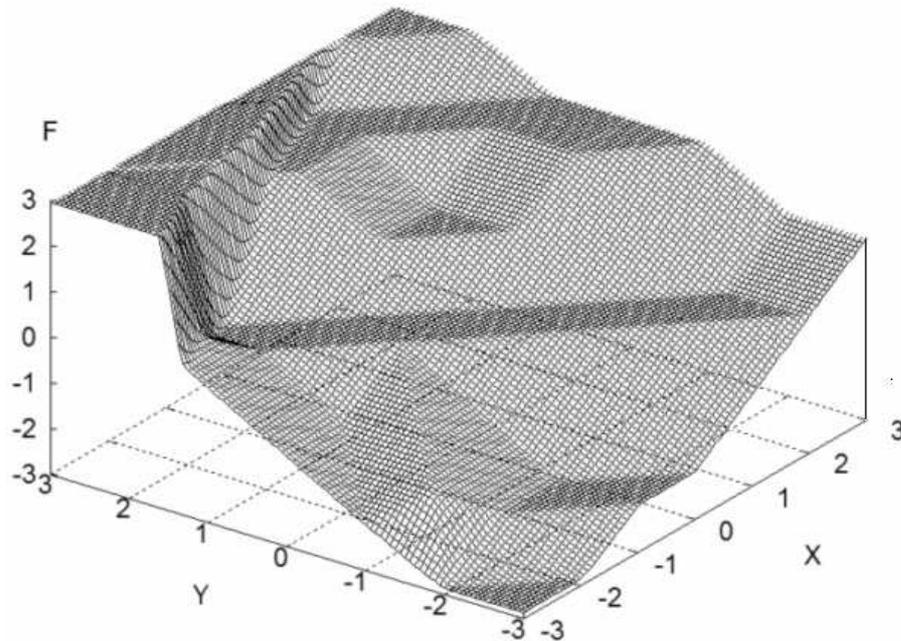


Figure 37. The response surface of the controller

This figure has been constructed by using GNUplot and illustrates the response surface in the coordinates X, Y . Analyzing the surface it is possible to conclude that the functioning of the controller is correct because of logical values of the circuit output depend on the logical values of the input variables in accordance with the Table 11. Moreover, the controller output signal has monotonic piece-wise linear approximation between adjacent logical levels. Thus the designed controller can be used as an analog device, which has analog inputs and produces an analog output signal.

5 Transformation of Analog Signals into Multi-valued Logic Form

In previous sections functions of multi-valued logic were specified by tables of fuzzy rules over linguistic variables. It was implicitly assumed that values of analog signals, which corresponded to linguistic variables, were evenly distributed in the range of voltages representing analog signals. Otherwise by artificial means the number of linguistic variables can be increased that leads to growing the implementation complexity.

In this section a procedure of transforming input analog variables into multiple-valued logic variables with evenly distributed logical levels is suggested. This procedure in some sense is analogous to the procedure of fuzzification in fuzzy controllers. Because of using a fuzzy control description for implementation of controllers as multi-valued logical functions the same term “fuzzification” for the suggested procedure of logical levels equalization for input variables will be applied.

The same procedure is supposed to be used when output multi-valued variables with evenly distributed logical levels demand backward transformation to an analog form with not evenly distributed voltages corresponding to logical levels. In this case the term “defuzzification procedure” will be applied.

5.1 Fuzzification Procedures

Let us examine more attentively the fuzzification procedure for the case of linear membership functions or membership functions, which sufficiently simply can be represented as piecewise-linear, and propose sufficiently simple universal method. Here the standard determination of a membership function is used. The membership function determines the weight of the corresponding linguistic variable b for each value of an analog variable X :

$$w_b = F(b, X); 0 \leq w_b \leq 1.$$

The simplest example of membership functions is given in Fig.38.

“The membership function is a graphical representation of the magnitude of participation of each input. It associates a weighting with each of the inputs that are processed, define functional overlap between inputs, and ultimately determines an output response. The rules use the input membership values as weighting factors to determine their influence on the fuzzy output sets of the final output conclusion. Once the functions are inferred, scaled, and combined, they are defuzzified into a crisp output which drives the

system. There are different memberships functions associated with each input and output response. Some features to note are:

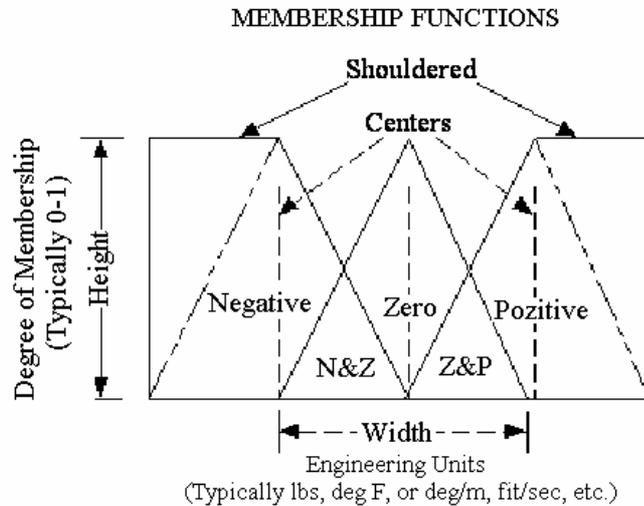


Figure 38. The simplest type of membership functions.

SHAPE - triangular is common, but bell, trapezoidal, haversine and, exponential have been used (More complex functions are possible but require greater computing overhead to implement.);

HEIGHT or magnitude (usually normalized to 1);

WIDTH (of the base of function);

SHOULDERING (locks height at maximum if an outer function. Shouldered functions evaluate as 1.0 past their center);

CENTER points (center of the member function shape);

OVERLAP (N&Z, Z&P, typically about 50% of width but can be less)".³

Fig.38 illustrates the features of the triangular membership function which is used in the following example.

The procedure of fuzzification and constructing corresponding diagram is examined on an example of the Container Crane fuzzy Controller, membership functions for which are given in Fig.39.⁴

It is assumed, without disrupting the generality of reasoning, that with changing the angle within the limits ($-90^\circ \div +90^\circ$) and the distance in the limits ($-10 \div +30$) yards the corresponding analog voltages vary within the range ($0 \div 3.5$) V. The source voltage of the controller circuit is also 3.5V.

³ Citation is taken from "Fuzzy Logic – an Introduction", part 4, by Steven D. Kaehler, http://www.seattlerobotics.org/encoder/mar98/fuz/fl_part4.html

⁴ http://www.fuzzytech.com/e/e_a_pfd.html

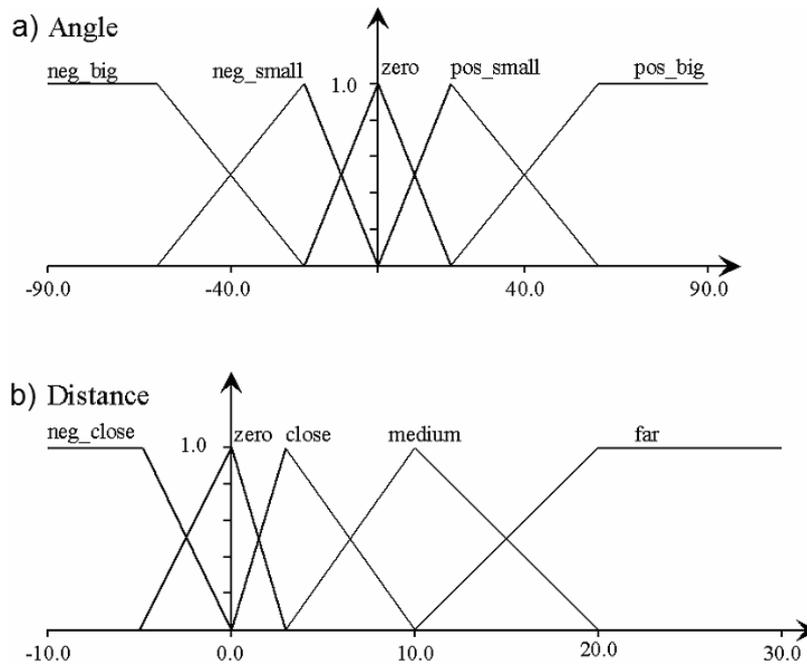


Figure 39. Membership functions for the Container Crane Fuzzy Controller: of the variable “angle” (a) and of the variable “distance” (b).

Table 12 determines the function of fuzzification for the piecewise-linear membership functions of the variable *angle* shown in Fig.39 (a). Linearity of these membership functions gives the possibility to connect the points of logical values by straight lines. The corresponding fuzzification (equalization) function is given in Fig.40.

Angle membership functions

Table 12

| linguistic variable | neg_big | neg_small | zero | pos_small | pos_big |
|---------------------|--------------------------------|---------------|-------------|--------------|------------------------------|
| angle | $-90^{\circ} \div -60^{\circ}$ | -20° | 0° | 20° | $60^{\circ} \div 90^{\circ}$ |
| voltage | $(0 \div 0.583)V$ | $1.361V$ | $1.75V$ | $2.139V$ | $(2.917 \div 3.5)V$ |
| logic values | -2 | -1 | 0 | +1 | +2 |
| voltage | $0V$ | $0.875V$ | $1.75V$ | $2.625V$ | $3.5V$ |

In this figure the variations of voltages from the average (equilibrium) point of summing amplifier are plotted along the axes.

For implementation of the function $V_{out} = F_1(V_{in})$ shown in Fig.40 three auxiliary functions should be introduced. These functions are represented in Fig.41. Their sum with saturation on the levels $\pm 1.75V$ determines the fuzzificated input function for the controller fuzzy inference part, which, as it has been already proved in previous sections, can be implemented as a multi-valued logic function.

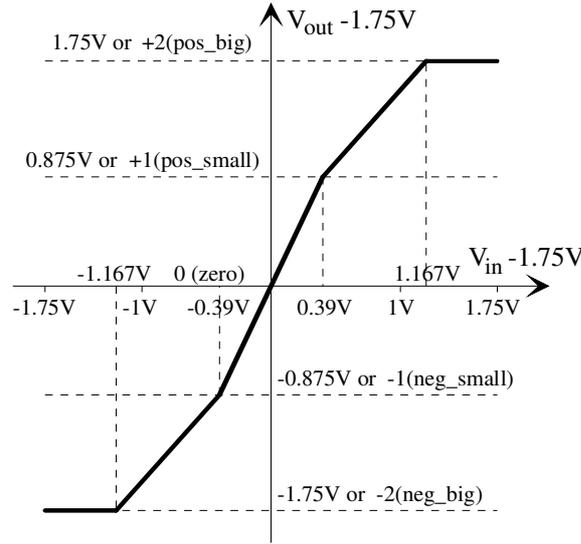


Figure 40. Piecewise-linear function for fuzzification of the variable “Angle”.

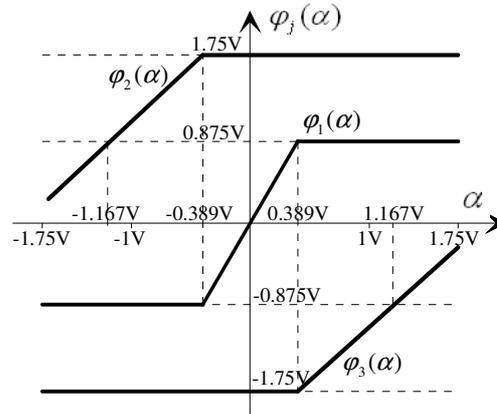


Figure 41. Component functions for the function represented in Fig.40.

In Fig.41 the angle α and values of functions $\varphi_j(\alpha)$ are represented in positive and negative voltages. These component functions and the fuzzifier output function $F_1(\alpha)$ can be implemented by the following way:

$$\begin{aligned}
 \varphi_1(\alpha) &= -0.5S(4.5\alpha); \\
 \varphi_2(\alpha) &= -S(1.125\alpha + 2.19); \\
 \varphi_3(\alpha) &= -S(1.125\alpha - 2.19); \\
 F_1(\alpha) &= S(-\varphi_1(\alpha) - \varphi_2(\alpha) - \varphi_3(\alpha)) = \\
 &= S\left(\frac{1}{2}S(4.5\alpha) + S(1.125\alpha + 2.19) + S(1.125\alpha - 2.19)\right).
 \end{aligned} \tag{54}$$

Now let us show how to construct and implement the fuzzification function for the input variable *distance*. As can be inferred from Fig.39 (b), the membership functions are characterized first by asymmetry of the measured distance ((-10 ÷ +30) yards) and second

by the explicit asymmetry of the linguistic variable positions along the distance axis. It assumed that the complete range of the measured distance corresponds to the complete range of the supply voltages ($0V \div 3.5V$) or ($-1.75V \div +1.75V$) in deviations from the equilibrium point of amplifier. For this case the fuzzification function is determined by Table 13. In this table the linguistic variable *close* corresponds to value “log. 0” and the linguistic variable *zero* corresponds to the value “log.-1”. The balance point of the amplifier input voltage corresponds to linguistic variable *medium*.

Distance membership functions

Table 13

| linguistic variables | neg_close | zero | close | medium | far |
|----------------------|-----------------|----------|-----------|----------|-----------------|
| distance | ≤ -5 yards | 0 yards | 3 yards | 10 yards | ≥ 20 yards |
| voltage | $\leq 0.4375V$ | $0.875V$ | $1,1375V$ | $1.75V$ | $\geq 2.625V$ |
| logic values | -2 | -1 | 0 | +1 | +2 |
| voltage | $0V$ | $0.875V$ | $1.75V$ | $2.625V$ | $3.5V$ |

Corresponding function $V_{out}(V_{in})$ is given in Fig.42. For implementation of this function $V_{out} = F_2(V_{in})$ it is necessary to realize four auxiliary functions, whose sum with saturation on the levels $\pm 1.75V$ will give the desired result. The auxiliary functions are given in Fig.43. Their values and value of the variable d (*distance*) are represented in negative and positive voltages.

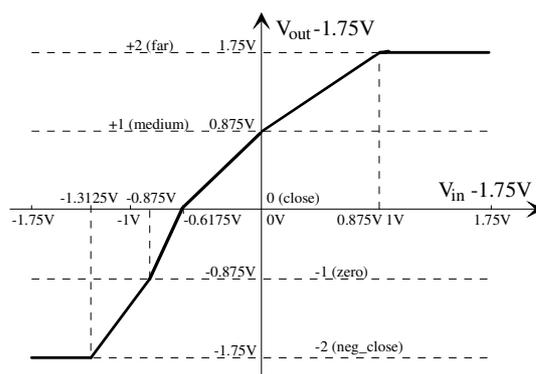


Figure 42. Piecewise-linear fuzzification function for the variable *distance*.

The ways of forming the component functions given in Fig.42 and the function $F_2(d)$ are shown below:

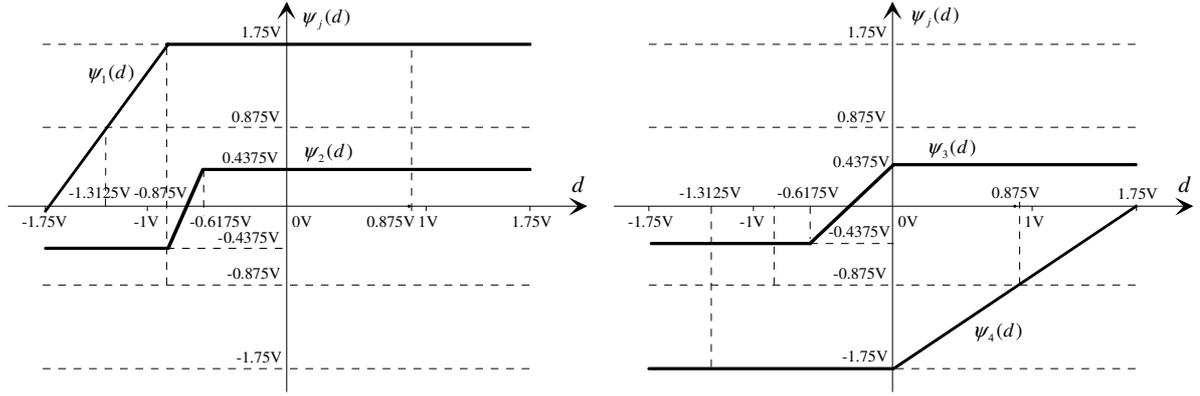


Figure 43. Component functions for the function represented in Fig.42.

$$\begin{aligned}
 \psi_1(d) &= -S(2 \cdot d + 3.5); \\
 \psi_2(d) &= -0.25S(13.6 \cdot d + 10.14); \\
 \psi_3(d) &= -0.25S(5.67 \cdot d + 1.75); \\
 \psi_4(d) &= -S(d - 1.75); \\
 F_2(d) &= S(-\psi_1(d) - \psi_2(d) - \psi_3(d) - \psi_4(d)) = \\
 &= S[S(2 \cdot d + 3.5) + 0.25S(13.6 \cdot d + 10.14) + 0.25S(5.67 \cdot d + 1.75) + S(d - 1.75)].
 \end{aligned} \tag{55}$$

5.2 Fuzzifier Implementations

For the completion of the fuzzifier design it only remains to determine the values of the input resistances of summing amplifiers and to conduct SPICE simulation for checking correctness of the implementations (54) and (55). These implementations are represented graphically in Fig.44 (a) and Fig. 44 (b) respectively. Their schematics, which have been used for SPICE simulations, are shown in Fig.45 (a) and Fig.45 (b).

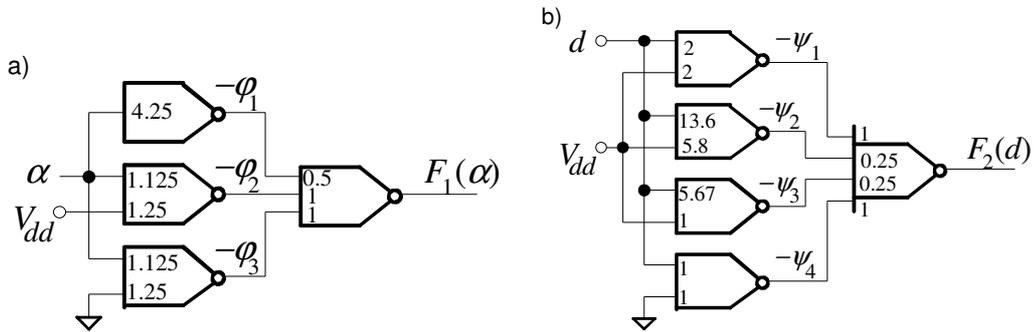


Figure 44. Fuzzifiers of the variable *angle* (a) and for variable *distance* (b).

Summing amplifiers used in the schematics are constructed on the base of three-stage push-pull CMOS operational amplifier in accordance with Fig.35 (a).

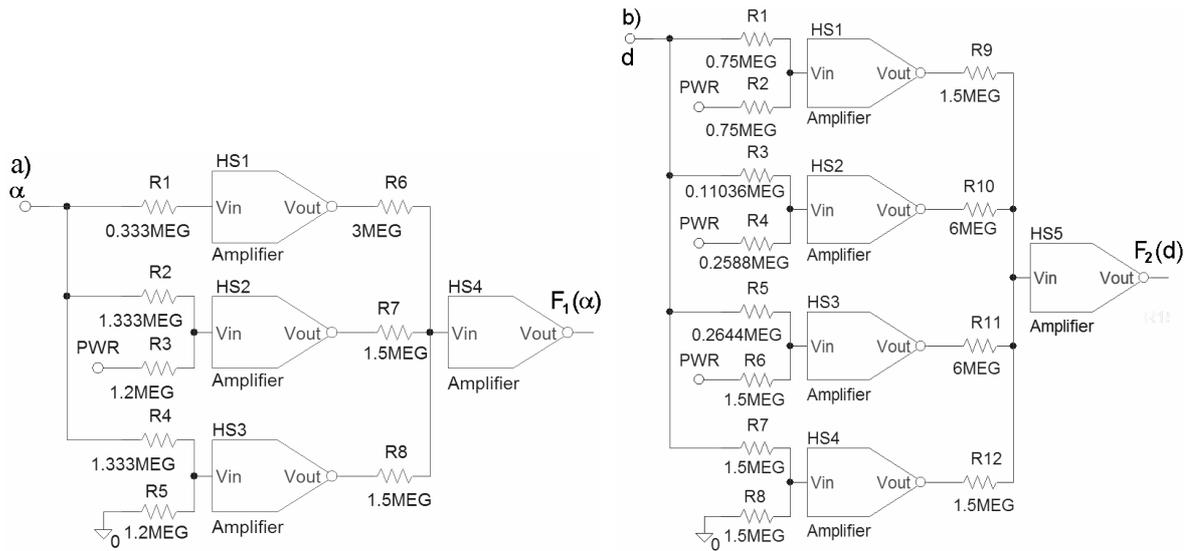


Figure 45. Fuzzifier schematics the *angle* (a) and for the *distance* (b).

Results of SPICE simulation of the fuzzifiers for variables “angle” and “distance” are shown in Fig.46 and Fig.47 respectively.

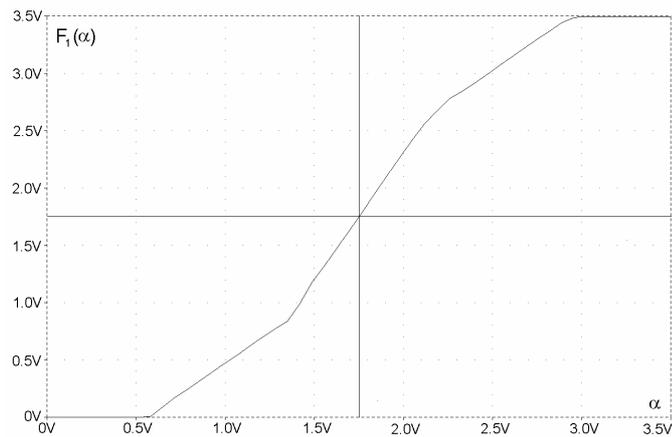


Figure 46. Output of the fuzzifier shown in Fig.45 (a) and derived by SPICE simulation.

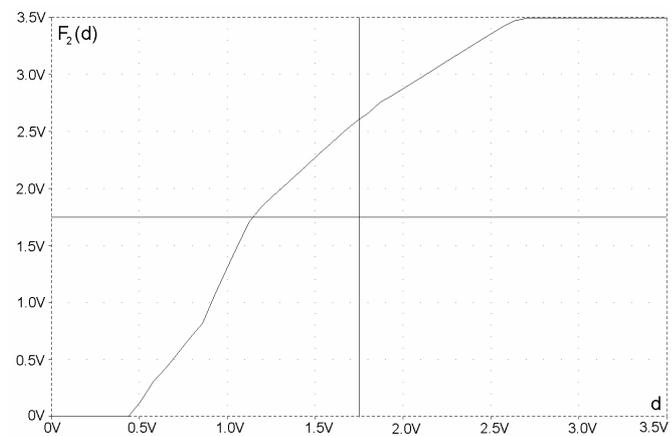


Figure 47. Output of the fuzzifier shown in Fig.45 (b) and derived by SPICE simulation.

It is easy to see that the simulation plots are exactly the same as it is required for fuzzification of the input variables *angle* (Fig.40) and *distance* (Fig.42). This proves the correctness of the fuzzifier implementations.

It should be noted that in the case of software implementation of the fuzzification and defuzzification functions, their component functions may be chosen not only piecewise linear but providing any reasonable approximations.

6 Conclusions

It was shown that all parts of fuzzy controllers can be effectively implemented on the base of summing amplifiers with saturation in accordance with the proposed methodology. This methodology is oriented to hardware implementation of fuzzy controllers as analog devices. In comparison with software implementation of the controllers, their hardware implementation have advantages of better response time and reliability, low power consumption, smaller die area etc. It should be noticed that the methodology also admits software implementation of the controllers by means of simulation using the summation operation with restrictions.

In all examples of controllers presented in the paper, the push-pull summing amplifier containing three CMOS invertors is used. The summing amplifier however can be of another type, e.g., the differential type or any other type of an operational amplifier.

Some may object that summing amplifiers in all examples of controllers designed with help of suggested methodology contain resistors of large values and it is very difficult to implement these resistors in CMOS VLSI technology. Indeed it is correct. In our case p-well resistors (1-10K Ohms/sq.) or pinch resistors (5-20K Ohms/sq.) can be used. These resistors are compatible with CMOS technology but occupy very large die area, possess bad accuracy, and have big temperature and voltage coefficients. By these reasons the possibility of creating a dynamical model of the summing amplifier with saturation using capacitors instead of resistors has been considered. This consideration gave positive results and will be published soon.

The proposed methodology has been applied for designing several devices specified as fuzzy controllers, showed high efficiency and gave very economical implementations. Techniques of synthesizing fuzzy devices in the offered base should get further developing and problems of implementability under the conditions of real production should be resolved in the nearest future.

References:

- [1] *Implementation of Fuzzy Logic*, Texas Instruments, SPRA028, January 1993.
- [2] *An Introduction to Fuzzy Logic Applications in Intelligent Systems*, by Ronald R. Yager, Lotfi A. Zadeh (Editor), Kluwer International Series in Engineering and Computer Science, 165, Jan. 1992, 356 p.
- [3] *Fuzzy Sets, Fuzzy Logic, and Fuzzy Systems*, by George J. Klir (Editor), Bo Yuan (Editor), Selected Papers by Lotfi A. Zadeh (Advances in Fuzzy Systems - Applications and Theory), World Scientific Pub Co.; Vol. 6, June 1996, 826 p.
- [4] *Fuzzy Logic Technology and Applications*, by Robert J. Marks II (Editor), IEEE Technology Update Series, Selected Conference Papers, 1994, 575 p.
- [5] V. Varshavsky, V. Marakhovsky, I. Levin, and N. Kravchenko, *Fuzzy Device*, Japanese Patent Application No. 2003-190073, filed to Japan's Patent Office, July 2nd, 2003.
- [6] V. Varshavsky, V. Marakhovsky, I. Levin, and N. Kravchenko, Summing Amplifier as a Multi-Valued Logical Element for Fuzzy Control, *WSEAS Transactions on Circuit and Systems*, Issue 3, Vol. 2, July 2003, pp. 625 – 631.
- [7] V. Varshavsky, I. Levin, V. Marakhovsky, A. Ruderman, and N. Kravchenko, CMOS Fuzzy Decision Diagram Implementation, *WSEAS Transactions on Systems*, Issue 2, Vol. 3, April 2004, pp. 615 – 620.
- [8] Post E., Introduction to a general theory of elementary propositions, *Amer. J. Math.*, 43, 1921, pp. 163-185.
- [9] *Fuzzy Control Systems*, Abraham Kandel (Edited by), Lotfi A. Zadeh (Foreword by), CRC Press LLC, Sept. 1993, pp. 81 – 86, pp. 168 – 172.
- [10] K. Kimura, T. Kawawa, *Exhaust Emission Control Device for Internal Combustion Engine*, Toyota Motor Corp., Patent of Japan No. 05-312020, Date of publication: 22.11.1993.
- [11] Mori Takakazu, Hamada Eiichi, Nishiyama Kunio, *Controller for Vehicle Seat*, Toyota Motor Corp., Patent of Japan No. 05-085235, Date of publication: 06.04.1993.
- [12] *Design of Analog Fuzzy Logic Controllers in CMOS Technologies*, by C. Dualibe, M. Verleysen, P.G.A. Jespers, Kluwer Academic Publishers, 2003.
- [13] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N. "Functionally Complete Element for Fuzzy Control Hardware Implementation", *the 2004 47th IEEE Midwest Symposium on Circuits and Systems*, Hiroshima, Japan, July 25-28, Vol. 3, 2004, pp. 263 – 266.

- [14] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N. "Fuzzy Controller CMOS Implementation", *WSEAS Transactions on Circuits and Systems*, Issue 9, Vol. 3, November 2004, pp.1762 – 1769.
- [15] Varshavsky, V.; Levin, I.; Marakhovsky, V.; Ruderman, A.; Kravchenko, N., "Fuzzy Decision Diagram Realization by Analog CMOS Summing Amplifiers", *Proceedings of the 11th IEEE International Conference on Electronics, Circuits and Systems (ICECS 2004)*, Tel Aviv, December 2004, pp. 286-289.
- [16] Varshavsky, V.; Marakhovsky, V.; Levin, I., "CMOS Fuzzification Circuits for Linear Membership Functions", *WSEAS Transactions on Systems*, Issue 4, Vol. 4, April 2005, pp.238 – 243.
- [17] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N., "Multi-Valued Logic Device", New Japanese Patent Application No. 2004-087880, filed to Japan's Patent Office, March 24th, 2004.
- [18] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N., "Functional Device", New Japanese Patent Application No. PCT/JP2004/9442, filed to Japan's Patent Office, July 2, 2004.
- [19] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N., "Median Circuit and Fault-Tolerant Circuit", New Japanese Patent Application No.2004-235239, filed to Japan's Patent Office, August 12, 2004.
- [20] Varshavsky, V.; Marakhovsky, V.; Levin, I.; and Kravchenko, N., "Fuzzification Circuit, Defuzzification Circuit and Fuzzy Functional Circuit", New Japanese Patent Application No. 2005-51291, filed to Japan's Patent Office, Feb. 25th, 2005.
- [21] Varshavsky, V.; Marakhovsky, V.; Levin, I., "CMOS Fuzzification Circuits for Linear Membership Functions", *WSEAS Transactions on Systems*, Issue 4, Vol. 4, April 2005, pp.238 – 243.