Practical Fuzzy-Logic Design

The Fuzzy-Logic Advantage

FEATURE ARTICLE

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O ver the past few years, systems designers have heard some rather controversial discussion about a new design technology called fuzzy logic. Some disciples of fuzzy logic have pitched it as the magic bullet for every type of engineering problem, and they've made wild and unsupported claims.

Such behavior raises skepticism among control-engineering specialists, causing some to completely condemn fuzzy logic as a marketing fad of no technical value.

The discussion in Asia and Europe about fuzzy logic has been much less religious and has kept closer to its real benefits. Fuzzy logic is considered one of the many tools necessary to build systems solutions in a fast, cost-effective, and transparent fashion [1]. It's not more or less than that.

Previous *INK* articles have generated explicit questions in how fuzzy logic achieves superior performance

In this article, Constantin shows us a situation where fuzzy logic solves a problem that traditional logic cannot. His intention is not to replace conventional techniques, however, just extend them.

Photo 1—When transporting concrete modules for bridges and tunnels by a crane, sway of the load must be controlled. A fuzzy-logic controller uses the experience of a human crane operator. This crane has two heads, each capable of transporting 64 tons, and it employs an antisway controller based on fuzzy logic.
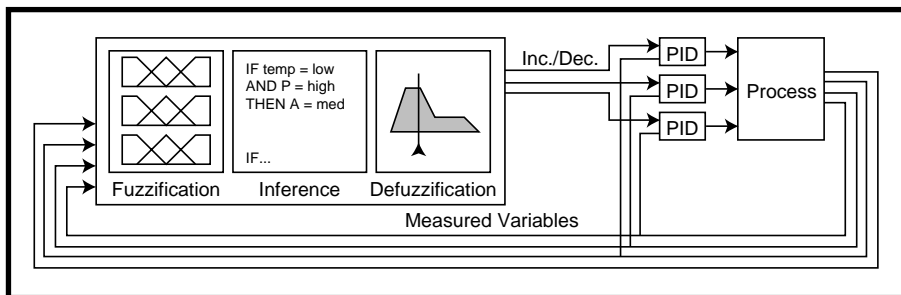
Figure 1—*In many applications, keeping a single figure of the process constant is relatively easy and can be facilitated by conventional PID controllers. However, to optimize the operation point of the process, a supervisory controller modifies the set points of the PID controllers.*

compared to other engineering design techniques. My goal is to show you exactly that.

To illustrate how fuzzy logic solves real-world problems and to compare it head-to-head with conventional design techniques, I'll use the case study of the antisway crane controller shown in Photo 1. I selected this real-world application for a number of reasons.

Unlike other technical real-world applications, the works of a crane controller can be understood quickly.

Antisway control of a crane is a simple problem, but conventional engineering techniques have a hard time arriving at a practical solution. In contrast, fuzzy logic rapidly delivers a good solution which has been successfully implemented on many crane controllers.

In this article, I'll assume you're familiar with the concept of fuzzy logic. If not, refer to this issue's "An Introduction to Fuzzy Logic" by Chris Sakkas, to the last fuzzy-logic issue (*INK* 56), or to *Fuzzy Logic and Neuro-Fuzzy Applications Explained* [2].

SINGLE-VARIABLE CONTROL

Before I start with the case study, let me clarify some control-engineering basics. Closed-loop control mostly involves keeping a single figure constant. It can be an electrical current, the mass flow of a liquid, or the temperature of a room.

To keep room temperature constant, a sensor delivers the value of the actual room temperature. The difference between this temperature and the desired room temperature is the temperature error.

The controller's objective is to get the temperature error to zero. To do

this, it may turn a heater or air conditioner on or off. Such controllers are referred to as on/off-type controllers.

If the command variable of the process is continuous, such as with a gradual heat control, a so-called proportional-type (P) controller is used. This implements a control strategy that computes its output—the command variable of the process—by multiplying the error by a constant factor.

Many real-world processes involve a time lag. When controlling room temperature and the heater turns on, it takes a while before the heat reaches the sensor.

Because the controller gets the measured room-temperature variable with a time delay, it can overreact and apply more heat than what's required to get the temperature error to zero. This overreaction can cause unwanted oscillation of the room temperature.

To compensate for this effect, proportional-differential (PD) controllers add the time derivative of the error multiplied by a constant to the error.

Proportional-integral-differential (PID) controllers also consider the integral of the error and add it to the output signal. This method ensures that the error eventually reaches zero because even a small offset in the error value is compensated for as its integral raises to a high value over time.

Because conventional controllers only use the error or signals derived from the error as input and the command variable of the process as output, they are called single-loop controllers. These controllers are usually simple to program and easy to tune.

Replacing such controller types by a fuzzy-logic controller only delivers a better performance in cases where the conventional controller doesn't cope well with the nonlinearities of the process under control [2, p. 82].

MULTIVARIABLE CONTROL

A much higher potential for fuzzy logic lies in the fact that it also facilitates the design of multivariable control systems. In a multivariable control loop, the control strategy considers not only one input variable (e.g., error) but different types of measured variables.

In the case of the room thermostat, a multivariable control strategy con-
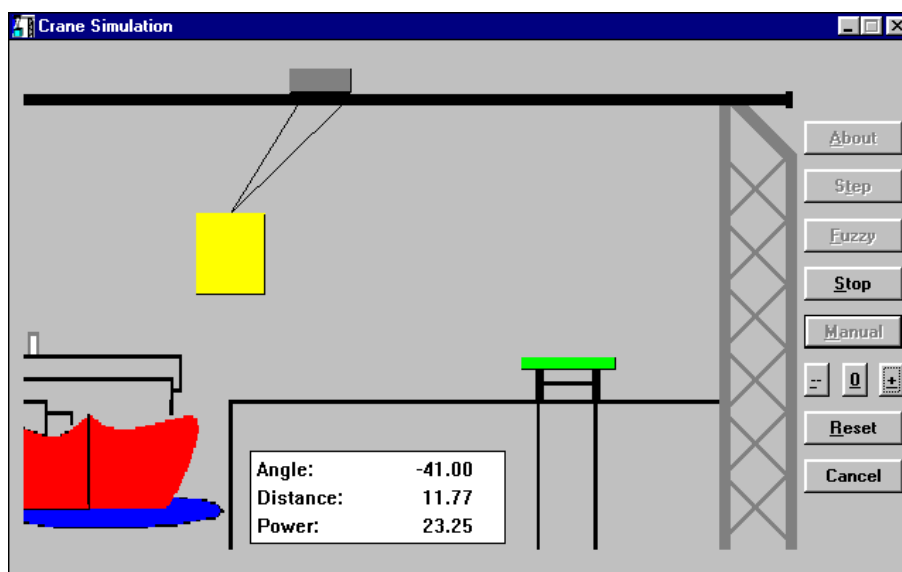

Photo 2—*The software simulation of container-crane operation visualizes the operation of the fuzzy-logic antisway controller. The fuzzy-logic control strategy is to get the container from the ship to the target as quickly as possible. When the target is reached, the sway must be compensated before releasing the container.*

Photo 3—*Each row corresponds to a fuzzy rule. The columns Angle and Distance underneath IF form the conditions of the rules. Underneath THEN, the column Power forms the conclusion of the rules. The DoS (Degree of Support) column can contain an individual weight of each rule.*

siders temperature error and room air humidity to set the command variables of the heater and air conditioner.

To deal with multiple input variables, a controller must assume a mathematical model of the controlled process. For the multivariable room-thermostat control strategy, such a model can be the humidity and temperature relationship which describes comfortable conditions.

In most real-world applications, the mathematical relation between the different variables cannot be described so easily. So, multivariable control is not commonly used.

However, application areas exist where multivariable-control strategies must be used, such as continuous process control as found in the food and chemical industries. Keeping single figures of the processes constant is easy in most cases. Controlling the plant's optimal operation point involves multiple variables.

Because of the difficulties involved with deriving mathematical models, automated control is rare and human operators often adjust the set points of the individual PID controllers.

This type of multivariable control is also referred to as supervisory control because the multivariable control strategy analyzes and supervises the set points of underlying control loops.

In contrast to conventional design techniques, fuzzy logic enables the design of such multivariable control strategies directly from human-opera-

tor experience or experimental results (see Figure 1).

Using such existing knowledge and circumventing the effort of rigorous mathematical modeling, the fuzzy-logic design approach delivers efficient solutions faster.

The question therefore becomes, "How can operator experience be put into a fuzzy-logic system?" The following case study of a container-crane antisway control illustrates this.

CONTAINER-CRANE CONTROL

Container cranes load and unload containers to and from ships in most

harbors (see Photo 2). They pick up single containers with flexible cables mounted at the crane head.

The crane head moves on a horizontal track. When a container is picked up and the crane head starts to move, the container begins to sway. While sway does not affect the transport, a swaying container can't be released.

There are two trivial solutions to this problem. One is to position the crane head exactly over the target position and wait until the sway dampens to an acceptable level. On a nonwindy day, this eventually happens, but it takes far too much time. A container ship has to be loaded and unloaded in minimum time.

The alternative is to move the container so slowly that no sway occurs. Again, this technique works on a nonwindy day and takes too much time.

Another solution is to build container cranes where additional cables fix the position of the container during operation. This alternative is very expensive.

CONTROL-MODEL ALTERNATIVES

Many engineers have attempted to automate this control task via conventional PID, model-based, and fuzzy-logic control strategies.

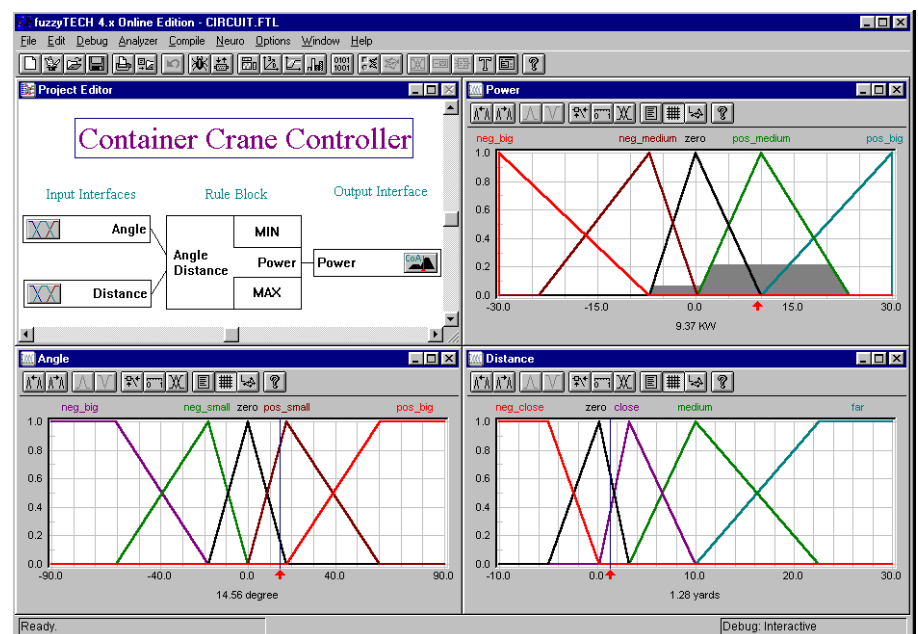Conventional PID control was unsuccessful because the control task is

Photo 4—*In this design of the fuzzy-logic antisway controller in* fuzzyTECH, *the upper left window shows the structure of the system with two inputs, one output, and one rule block. The upper right window visualizes the defuzzification of* Power, *and the lower windows show the fuzzification of* Angle *and* Distance.
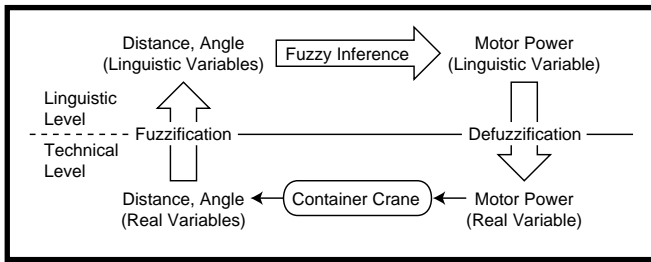
Figure 2—*The complete control loop of the fuzzy-logic crane controller has three steps. The variables* Distance *and* Angle *measured at the crane are translated into linguistic variables (fuzzification) and used to evaluate the condition of the fuzzy rules (fuzzy inference). A linguistic value for* Power *is translated back into a numerical value (defuzzification).*

inherently nonlinear. For example, sway minimization is important only when the container is close to the target.

Others have tried to derive a mathematical model of the crane to use in a model-based controller. They came up with a fifth-degree differential equation that describes the mechanical behavior. In theory, a controller design based on this model works. In reality, it doesn't.

One reason for failure is that the weight of the container is unknown. Also, the crane-motor behavior isn't as linear as assumed in the model. Its gear box involves slack, its head only moves with friction, and its cables involve elasticity. As well, disturbances such as wind gusts aren't included in the model.

LINGUISTIC CONTROL STRATEGIES

On the other hand, a human operator can control a crane without differential equations. (Chances are, if the operator knew how to use differential equations, he wouldn't operate cranes.)

An operator doesn't even use the cable-length sensors that a model-based solution requires. Once the container is picked up, the operator starts the crane with medium motor power to see how the container sways.

Depending on the reaction, the motor power is adjusted to get the container a little behind the crane head. In this position, maximum speed is reached with minimum sway.

In approaching the target position, the operator reduces motor power or applies negative power. The container gets a little ahead of the crane head until the container almost reaches target position.

Motor power is then increased so the crane head is over target position and sway is zero. No differential equations are required, and disturbances and nonlinearities are compensated by the operator's observation of the container's position.

The operator's control strategy can be described by several rules:
- start with medium power
- if you're still far away from the target, adjust the motor power so the container gets a little behind the crane head
- if you're closer to the target, reduce speed so the container gets a little ahead of the crane head
- when the container is very close to target position, power up the motor
- when the container is over the target and the sway is zero, stop the motor

The advantage of fuzzy logic is that it can use the same sort of rules as the human operator.

To automate control of this crane, sensors are used for the head position (Distance) and the angle of the container sway (Angle). Using these inputs to describe the current condition of the crane, the rules can be translated into the if/then statements in the

fuzzy-logic table in Photo 3.

The first condition describes the value of Distance, and the second the value of Angle. The conditions are combined by AND since both have to be valid for the respective situation.

Once you have rules describing the desired behavior of a system, the question becomes how to implement these rules. Consider using a programming language to code the if/then rules. Unfortunately, the conditions of the rules use words that need to be defined, and exact definitions don't exist.

However, fuzzy logic provides non-exact definitions for the words. Thus, you use linguistic rules for control-system design directly.

FUZZY-LOGIC CRANE CONTROLLER

Figure 2 shows the complete structure of a fuzzy-logic controller. All sensor signals have to be translated into linguistic variables. A measured distance of 12 yd. has to be translated to the linguistic value "still medium, just slightly far." This step is called *fuzzification* because it uses fuzzy sets for translating real variables into linguistic variables.

Once all input variable values are translated into linguistic variable values, the so-called fuzzy-inference step evaluates the if/then fuzzy rules that define system behavior. This step yields a linguistic value for the linguistic variable. So, the linguistic result for Power could be "a little less than medium."
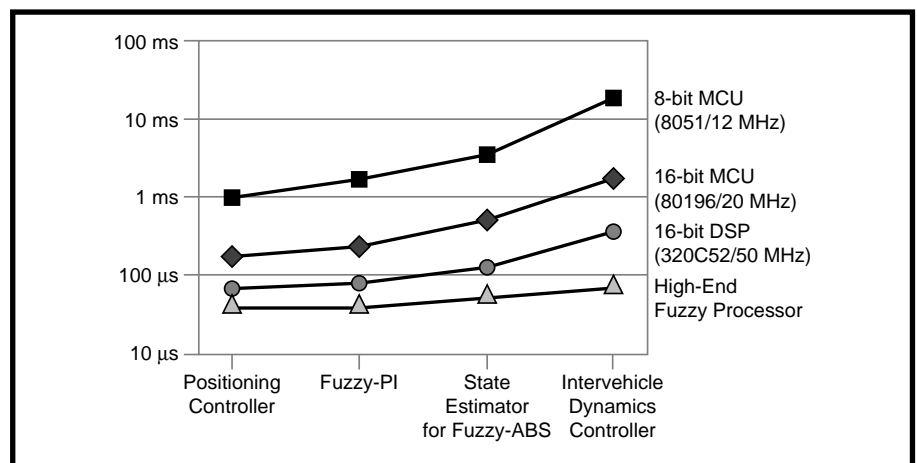

Figure 3—*Using a set of standardized benchmark systems, the performance of different hardware platforms can be compared. On a 8051 MCU, a small fuzzy-logic system can be computed in about a millisecond. Faster microcontrollers and DSP can compute even very large fuzzy-logic systems in fractions of a millisecond.*

Defuzzification translates this linguistic result into a real value that represents the power setting of the motor in kilowatts.

DESIGN AND IMPLEMENTATION

The development of fuzzy-logic systems involves specific design steps:

- design the inference structure—specify how the output variables connect to the input variables by the rule blocks.
- define the linguistic variables—the variables form the vocabulary used by the fuzzy-logic rules expressing the control strategy.
- create an initial fuzzy-logic rule base using all available knowledge on how the system should perform.
- debug, test, and verify the system offline for completeness and non-ambiguity—use a software simulation or sample data of the process if it exists in this step.
- debug online—connect the fuzzy-logic system to the process under control and analyze its performance in operation. Because fuzzy logic lets you modify the system in a straightforward way from the performance you observe, this step can expedite system design rapidly.

Most systems today are developed using software tools like the one in Photo 4. Such development tools not only support all listed design steps, but they also can generate the entire system for different hardware platforms. For microcontrollers (MCUs), the tools generate assembly code. For PLCs, they generate function blocks. For PC/ workstations, they generate C code.

Many years ago, the computation of the fuzzy-logic algorithm was so inefficient on a standard MCU that dedicated fuzzy-logic processors were developed. Today, code efficiency has improved dramatically.

Figure 3 shows different MCUs, a DSP, and today's fastest fuzzy-logic processor computational performance for four benchmark systems. The time shown on the vertical logarithmic scale is the total time required to compute the complete fuzzy-logic system.

On a standard 12-MHz 8051 MCU, small fuzzy-logic systems compute in just one millisecond, and 16-bit MCUs can compute large fuzzy-logic systems in the same amount of time. This speed enables the integration of a fuzzy-logic system with most embedded system designs. However, some embedded system applications require even faster computation (e.g., antilock brakes in cars, ignition control, or hard-drive positioning).

To expedite the fuzzy-logic algorithm, some semiconductor manufacturers include specific fuzzy-logic instructions with their new-generation MCUs. For example, Motorola's 68HC12 MCU features a complete fuzzy-logic function set in assembly language.

LESSONS LEARNED

Fuzzy logic enables the use of experience and experimental results to deliver more efficient solutions. It does not replace or compete with conventional control techniques.

Rather, fuzzy logic extends the way automated control techniques are used in practical applications by adding supervisory control capabilities. The container-crane example demonstrates that fuzzy logic delivers a transparent, simple solution for a problem that's much harder to solve using conventional engineering techniques. [▲]

Constantin von Altrock began research on fuzzy logic with Hewlett-Packard in 1984. In 1989, he founded and still manages the Fuzzy Technologies Division of Inform Software, a market leader in fuzzy-logic development tools and turn-key applications. You may reach Constantin at cva@ inform-ac.com

SOFTWARE

You may download the complete animated software simulation of the container crane for Windows from Circuit Cellar's BBS together with a simulation-only version of *fuzzy*TECH and the FTL source code of the crane controller. You may download the source code of the benchmark suite from *fuzzy*TECH's Web site (http://www.inform-ac.com).

REFERENCES

[1] C. von Altrock, "Fuzzy Logic Applications in Europe," in J. Yen, R. Langari, and L.A. Zadeh (eds.) *Industrial Applications of Fuzzy Logic and Intelligent Systems*, IEEE Press, Chicago, 275–310, 1995.
[2] C. von Altrock, *Fuzzy Logic and NeuroFuzzy Applications Explained*, Prentice Hall, Englewood Cliffs, NJ, 1995.

SOURCES

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