

Introduction to Soft Computing
Prof. Debasis Samanta
Department of Computer Science & Engineering
Indian Institute of Technology, Kharagpur

Lecture - 13
Fuzzy logic controller (Contd.)

So, there are 2 broad approaches, so far the fuzzy logic controller design is concerned. One approach is Mamdani approach and another is Takagi Sugeno approach. 2 approaches are different, the way they treated the fuzzy logic controller design. We have discussed the Mamdani approach and we see that the Mamdani approach follow a rule base and then fuzzy linguistic state and then Fuzzification of the input and then produce the fuzzy output and then crisp value. The method more or less same, they are in the Takagi Sugeno approach, but the though the way they treat the fuzzy inference engine is different.

Now, fuzzy inference engine rather more what is called the interpretable so far the Mamdani approach is concerned. However, in case of Takagi Sugeno approach, it is less interpretable. So, interpretation means, anybody can see from the design how it works, but Takagi Sugeno approach is bit difficult because it follows certain mathematical treatment so far the inference engine is concerned.

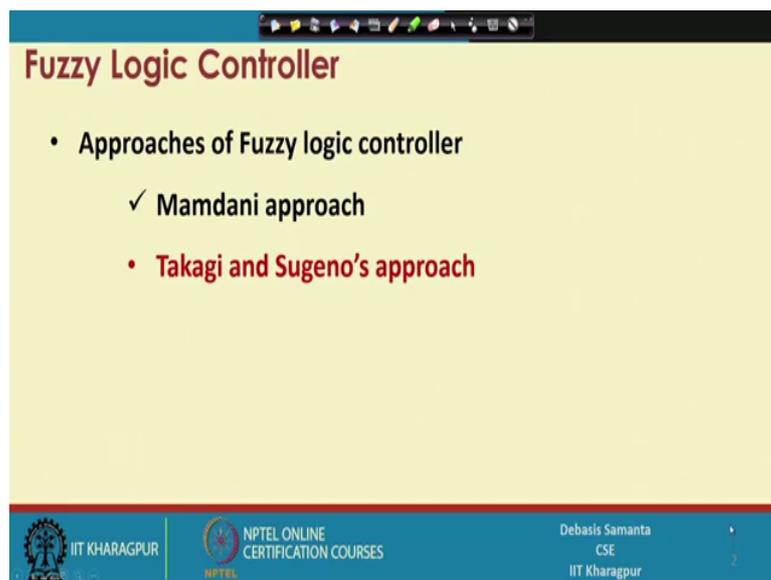
Now, so far the output quality is concerned, the Mamdani approach is less accurate, whereas Takagi Sugeno is give more accurate output and output calculation if we consider according the to the 2 different approaches, Mamdani approach follows the standard Fuzzification method. So, Fuzzification method if we follow COG, it is computationally expensive, whereas the Takagi Sugeno approach follow the simple numerical calculation and it is faster. So, we can broadly can say that Mamdani approach is easy to interpret, but less accurate and computationally bit expensive. On the other hand, Takagi Sugeno approach is difficult to interpret, accurate; more accurate than the Mamdani approach and then calculation is fast.

So, if you want to design a fast and accurate fuzzy logic controller, then we can follow Takagi Sugeno approach, but there is one issue, Takagi Sugeno approach as it needs some mathematical treatment. So, whatever the rules are there they needs to be stored in the form of a, some mathematical representation and that is a big challenge for the designer. So, if the designer is not so much experienced, then they can follow certain difficulty in this direction,

whereas Mamdani approach is very easy to frame the rules and then rule base and then inference engine.

So, these are the 2 difference are there. Obviously, there is a trade off. Now, let us see how the Takagi Sugeno approach so far the fuzzy logic controller is concerned and then how it works.

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Fuzzy Logic Controller

- Approaches of Fuzzy logic controller
 - ✓ Mamdani approach
 - **Takagi and Sugeno's approach**

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Now, we will consider one example, so that you can understand Takagi Sugeno approach and mainly in terms of input example case studies rather we will explain the step, whatever the methods and technologies are there, I will discuss in time.

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Takagi and Sugeno's approach

- In this approach, a rule is composed of **fuzzy antecedent** and **functional consequent** parts.
- Thus, any *i*-th rule, in this approach is represented by If (x_1 is A_1^i) and (x_2 is A_2^i) ... and (x_n is A_n^i) ✓
- Then, $y^i = a_0^i + a_1^i x_1 + a_2^i x_2 + \dots + a_n^i x_n$ ✓
- where, $a_0, a_1, a_2, \dots, a_n$ are the co-efficients.

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Anyway, let us proceed here the Takagi Sugeno approach. Now, according to this approach, every rule is represented as it is in the same as Mamdani approach using the, if then clause.

Now, for a system if there are n input, then a rule in the Takagi Sugeno approach take this form. It is the rule that is with the n input x_1, x_2 and x_n . So, if n input at any instant having its own value and if these are the input, which basically related to the fuzzy state A_1 , these are input fuzzy state A_2 and these are n th input, the fuzzy state A_n , then the rule will take this form.

Now, this is the rule and if this is the rule, then the output of the rule also can be expressed mathematically and then for this input the output is shown here. So, it is basically input x_1, x_2, x_n and a_0, a_1, a_2 and dot dot a_n are the coefficients. Now, how these coefficients can be obtained, I will discuss it. Anyway, this is the coefficient will be supplied by the fuzzy engineer based on the different rule.

So, for the i -th rule, so these are the coefficient will be totally different and this is a difficult job time to time for engineer to decide right value of this coefficient, because the output of a rule depends on the right choice of the this coefficient value. Anyway, so if fuzzy, we can depends on the wisdom of the fuzzy engineer and let the fuzzy engineer suggest these are the coefficients for the i -th rule and therefore, for any input which satisfies this, the rule we can calculate the output. So, these are the basic thing and this is very difficult job for the engineer

to decide this one, once you decide it then rest of the things is very straightforward and simple.

Now, let us see how the rest of the part can be for the 2 rules, for the rules that we have discussed which takes this form like.

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Takagi and Sugeno's approach

- The weight of i -th rule can be determined for a set of inputs $x_1, x_2, x_3, \dots, x_n$ as follows.
- $w^i = \mu_{A_1}^i(x_1) \times \mu_{A_2}^i(x_2) \times \dots \times \mu_{A_n}^i(x_n)$
- where A_1, A_2, \dots, A_n indicates membership function distributions of the linguistic hedges used to represent the input variables and μ denotes membership function value.

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Now, according to this approach, it first calculates the weight of a rule. For example, for the i -th rule that we have discussed with the n th input, the weight; the i -th weight can be calculated using this formula. This formula is like this, x_1 , which belongs to the fuzzy state A_1 taking its membership value, so μ_{A_1} and similarly for the other x_2 which is in the fuzzy state A_2 , taking this membership value as μ_{A_2} and so on.

For the n -th input taking the fuzzy state A_n and taking its membership value. So, it basically takes the product of the different membership values for all inputs, belongs to the different fuzzy state. So, taking these are the values, we can calculate the weight of the i -th rule. So, for whatever the rules are relevant for each rule we will be able to calculate weight.

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Takagi and Sugeno's approach

- $y^i = a_0^i + a_1^i x_1 + a_2^i x_2 + \dots + a_n^i x_n$ ✓
- $w^i = \mu_{A_1}^i(x_1) \times \mu_{A_2}^i(x_2) \times \dots \times \mu_{A_n}^i(x_n)$ ✓
- The combined action then can be obtained as

$$y = \frac{\sum_i^k w^i y^i}{\sum_i^k w^i}$$

where k denotes the total number of rules

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Now, after the weight is calculated, the next step is to calculate the output value these call the y^i . As I told you, so y^i is associated with each rule and this is y^i value for the i -th rule and this is a weight for the i -th rule, that we have calculated in the last step and taking these values for all the rules.

Suppose there are k rules and then taking the sum of product of all the w^i into y^i and divided by the sum of all weights, it will give you the final output and that is the output and you can say that this output y , is basically the crisp output and this is the one difference from the Mamdani approach to the Takagi Sugeno approach, where we need to calculate the fuzzy output first and from the fuzzy output we have to calculate the crisp output.

But here, directly from the fuzzy input as the rule we can calculate the weights for each rule and then the output value of this one and then using this formula the final output, which is in the crisp value can be calculated. So, this way this is first, because straight away we can avoid the Fuzzification module there and then we can directly come into the rule.

Now I can, let us elaborate the Takagi Sugeno approach with an example, so that you can understand. So, this is the method actually by which the Takagi Sugeno approach calculates the output and this is basically the mechanism, where the fuzzy inference engine work. Now, whatever the rule base and then Fuzzification and everything that is the same as Mamdani approach, now let us consider, as an illustration, one example. (Refer Slide Time: 08:48)

Illustration:

Consider two inputs I_1 and I_2 . These two inputs have the following linguistic states :

I_1 : L (low), M (Medium), H (High)

I_2 : NR (Near), FR (Far), VF (Very Far)

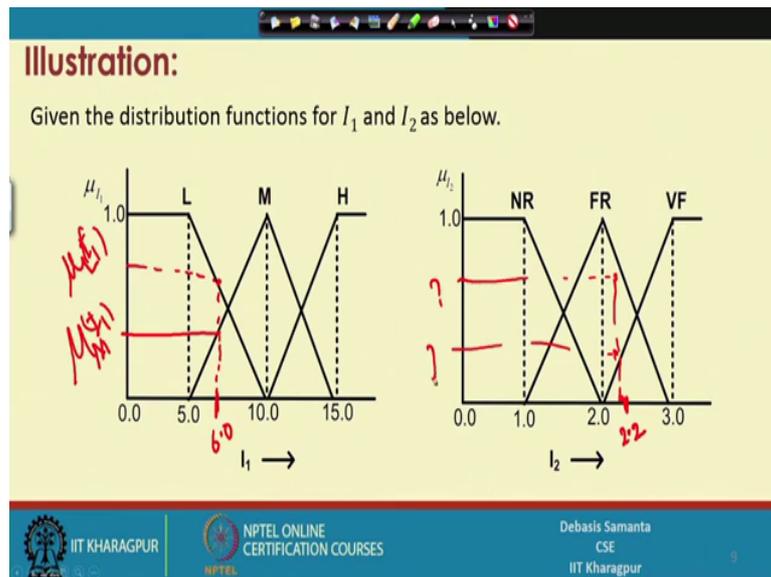
Note:
The rule base of such a system is decided by a maximum of $3 \times 3 = 9$ feasible rules.

So, let us consider is a 2 input system. 2 inputs are denoted as I_1 and I_2 , they let us consider there are abstract input and also consider that the 2 inputs I_1 and I_2 has the fuzzy linguistic, namely low, medium and high. That means, I_1 can be considered as a fuzzy state low, fuzzy state M or fuzzy state high with the different values of the membership or the different values of I_1 .

Similarly, the other input I_2 has the 3 fuzzy linguistic again. They are called near denoted NR and far denoted FR and very far denoted as VF. So, there are 3, linguistics for the first input and then 3 linguistics state or fuzzy state for the second input. Now again if we follow the rule base matrix for this and you can understand the rule base matrix has, how many rules? Total 9 rules. So, if this is the L M and R, sorry it is a, so rule base system it is like this, so L, M and H. So, the 3 input for I_1 , these are the I_1 and another 3 input for I_2 . So, this is the I_2 and has NR, FR and then VF.

Now, here in this rule it is basically, so there are 9 rules. So, rule 1, the y_1 output, y_2 and y_3 , similarly y_4 , y_5 , y_6 , y_7 , y_8 and y_9 . Now, in the Mamdani approach, we have observed that all the rules values here in terms of fuzzy linguistic, where as in the Takagi Sugeno approach all the rule values in terms of mathematical representation by which the y_1 , y_2 , y_3 all this things can be calculated. So, if there is a rule if I_1 is L and I_2 is NR, then the output y_1 , will be calculated using some mathematical notation. Now, for this again example, see what is the mathematical notation that we can consider out of these 9 rules, it is expressed here.

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Now, before going to this, again let us consider the fuzzy linguistic state for 3 fuzzy states. So, far the, I_1 input is concerned, which are denoted as L M and H is shown here. So, it is the fuzzy state for the L M and H, the fuzzy membership function which is like this, for L it is like this and for H it is like this. So, these are the fuzzy state defined for the, I_1 and the input of I_1 , is in the range 0 to 15 as it is mentioned here.

Now, similarly for the other input I_2 , the range of values that is in 0.02, 3.0 and all of and the fuzzy state that we have considered here, it is shown here. For NR it is look like this, for FR it is like this the membership function and for VF the membership function is like this. So, this is the fuzzy state for the 2 inputs I_1 and I_2 , μ_{I_1} is the membership values for the different fuzzy state belongs to the input I_1 and μ_{I_2} is a membership values for the different fuzzy state belongs to the different input as I_2 .

Now,. So, this is the Fuzzification module that is here or it is basically the fuzzy design is there. That means, all the input should be represented in the form of a fuzzy linguistic and we have discussed the 3 different fuzzy linguistic related to this illustration. Now having these the fuzzy linguistics there, now we have to calculate for a particular instant, that means, for a particular values of I_1 and for a particular values of I_2 at any moment, how the fuzzy input can crisp input can be converted to the fuzzy input.

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Illustration:

The output of any i -th rule can be expressed by the following:

$$y^i = f(I_1, I_2) = a_j I_1 + b_k I_2$$

where, $j, k = 1, 2, 3$.

Suppose:

$a_1^i = 1, a_2^i = 2, a_3^i = 3$ if $I_1 = L, M$ and H , respectively.

$b_1^i = 1, b_2^i = 2, b_3^i = 3$ if $I_2 = NR, FR$, and VF , respectively.

We have to calculate the output of FLC for $I_1 = 6.0$ and $I_2 = 2.2$

It is basically the same method that we have discussed there and we can follow it. Now here, in this slide say suppose each fuzzy rule that is denoted here in this expression. As I told you, again the same thing I can write it. So, it is basically y_1, y_2, y_3 right, y_4, y_5, y_6, y_7, y_8 and y_9 , 9 rules are there. For each rule, that can be discussed in terms of the input value I_1 and I_2 and suppose we assume these are the, what is called the mathematical representation of output values.

Now, if we see, so this is the input I_1 and I_2 and then these are the coefficient. Now, all this coefficient needs to be decided by the designer. So, it is basically they are, basically how many coefficients are there L, M and H . If we consider 1 coefficient a_1 and a_2, a_3 for this, similarly here NR then FR and VF , then we can consider b_1, b_2, b_3 so there.

Now, so we have to decide the different values for this rule, if a_1 is 1, b_1 is this and this one. So, there are 9 different values to be considered here, for j equals to 1, 2, 3 and some k equals to say, 9 different values to be considered there. Now for simplicity, suppose we consider the values a_i like this one, a_1 for any rule I_1 is 1, a_2 for any rule is 2, a_3 for any rule I_1 is 3.

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Illustration:

The output of any i - th rule can be expressed by the following.

$$y^i = f(I_1, I_2) = a_j^i I_1 + b_k^i I_2;$$

where, $j, k = 1, 2, 3$.

Suppose:

$a_1^i = 1, a_2^i = 2, a_3^i = 3$ if $I_1 = L, M$ and H , respectively.

$b_1^i = 1, b_2^i = 2, b_3^i = 3$ if $I_2 = NR, FR$, and VF , respectively.

We have to calculate the output of FLC for $I_1 = 6.0$ and $I_2 = 2.2$

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So, you have a simplified assumption and suppose it is decided like this one. So, alternatively I can say like, y^1 this is equals to $a_1^1 I_1 + b_1^1 I_2$ this is the 1 rule like. similarly y^2 there is another rule also can be considered likewise. So, all this expression, that means, for the output can be represent like this. Now, having this is the representation of a particular rule and corresponding to that particular rule basically the output value, compress the rule base for the fuzzy system according to the Takagi Sugeno approach.

Now, let us consider a particular instant for a certain values of I_1 and I_2 . We consider, this is the input at any moment, that I_1 is 6.0 and I_2 is 2.2. Now given this is the input, we have to calculate the fuzzy output according to the Takagi Sugeno approach. Let us see, how it can be calculated? Now, again for this I_1 equals to 6.0 and I_2 , 2.2, we have to calculate the μ values of the mem elements. That belongs to the 2 different fuzzy state. Here, so I_1 is 6.0, so this basically this is a 6.0 and for the I_2 we have considered 2.2, so it is the 2.2 right. So, if this is even, then so this basically 6.0 belongs to that 2 fuzzy state M and as well as another fuzzy state L. So, I_1 is 6.0 belong to the fuzzy state M, has the $\mu_M I_1$ value this one.

So, it is basically $\mu_M I_1$ value this one. Similarly, the 6.0 the input I_1 belongs to the fuzzy state L, having this value. So, it is basically $\mu_L I_1$ is the value. Likewise, 2.2 has the 2 fuzzy state namely, VF and then another is FR 2 1. So, this value and this value are to be calculated. So, this μ values can be calculated from this fuzzy description of the state are

the same way, as it is followed there in the Mamdani approach, using the principle of similarity of triangles, we can calculate it. So, this mu values will be obtained. Once the, this mu values are known to us, then we can calculate the output value easily.

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Solution:

a) The input $I_1 = 6.0$ can be called either L or M.
Similarly, the input $I_2 = 2.2$ can be declared either FR or VF.

b) Using the principal of similarity of triangle, we have the following.

- $\mu_L(I_1) = 0.8$
- $\mu_M(I_1) = 0.2$
- $\mu_{FR}(I_2) = 0.8$
- $\mu_{VF}(I_2) = 0.2$

Now here, for an example we can calculate the mu value this is for the 2 input, this one as this one, mu 1 mu L I 1, mu M this one and this one.

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Solution:

c) For the input set, following four rules can be fired out of all 9 rules.

- R1: I_1 is L and I_2 is FR
- R2: I_1 is L and I_2 is VF
- R3: I_1 is M and I_2 is FR
- R4: I_1 is M and I_2 is VF

So, these are the 4 different values that can be calculated from the fuzzy linguistics state description and using the similarity of principle that we have already discussed about in Mamdani approach.

Now, so these values are known to us now. Now, we will use these values and then the output values for each can be calculated and so this is the method that can be calculated here. Now, here you can see, again just like Mamdani approach there are 20 rules, here 9 rules. Now, out of the 9 rules we have to select those rules, which are basically fireable. Fireable means, those rules are relevant in the context of current input. Now, as you see the, for the I_1 equals to 6.0 and I_2 equals to 2.2 and when I_1 is 6.0 it belongs to the fuzzy state L and M. Similarly, when I_2 is 2.2, it belongs to the fuzzy state FR and VF.

So, putting all these things together, there are therefore 4 rules, the 4 rules are discussed here. So, if I_1 is L and I_2 is FR, then it basically gives y_1 , 1 rule. Similarly r_2 gives you y_2 and this one gives you y_3 and this gives you y_4 . So, the 4 rules and related to the 4 rules, the 4 different output can be obtained. So, the task, the object, the next task is basically for each rule we have to calculate w_i . These are w_i that is the weighted value of the rule, for this rule also 2, for this rule 3 and for this rule 4.

So, we have to calculate these are the values, so far the output is concerned and these are the values, so far the rules strength is concerned or weight is concerned. Once, we have these values, then we can calculate w_i and y_i , where i equals to 1 to 4, is a numerator and then the summation of w_i , i equals to 1 to 4. So, this will give you the final output according to the Takagi Sugeno approach. Now, let us see in the current context, how the results can be obtained for the 2 things are there. We have considered the different values of a_1 and b_1 in the context of this example.

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Solution:

d) Now, the weights for each of the above rules can be determined as follows.

- $R1: w^1 = \mu_L \times \mu_{FR} = 0.8 \times 0.8 = 0.6$ ✓
- $R2: w^2 = \mu_V \times \mu_{VF} = 0.8 \times 0.2 = 0.16$ ✓
- $R3: w^3 = \mu_M \times \mu_{FR} = 0.2 \times 0.8 = 0.16$ ✓
- $R4: w^4 = \mu_V \times \mu_{VF} = 0.2 \times 0.2 = 0.6$ ✓

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Now here, so this basically shows the computation of the o h value, as I told you it is a simple product of the membership value is there. So, for the rule 1 the value that can be obtained is 0.6, for the rule the value that can be obtained 0.16 and like this one. So, these values can be obtained based on the different values of the membership values in the context of current input. Current input here or here so far for the I 1 is concerned and these are the I 2 is concerned. So, w 1 can be calculated. Now, once the w 1 is known, then we will see how the y 1 for each rule, R 1 and R 2 can be calculated.

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Solution:

e) The functional consequent values for each rules can be calculated as below.

- $y^1 = I_1 + 2I_2 = 6.0 + 2 \times 2.2 = 10.4$
- $y^2 = I_1 + 3I_2 = 6.0 + 3 \times 2.2 = 12.6$
- $y^3 = 2I_1 + 2I_2 = 2 \times 6.0 + 2 \times 2.2 = 16.4$
- $y^4 = 2I_1 + 3I_2 = 2 \times 6.0 + 3 \times 2.2 = 18.6$

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So, here is the method by which the y_1 that means output value can be calculated. I told you that, we will consider the different coefficients. For the rule 1, we take a_1 is 1 and b_1 is 2 and for others, a_2 is 3; a_2 is 2 and b_2 is 3. So, these are the values of rule can be calculated like so for the different values. So, here basically a_1 is 1 and b_1 is 2 and in this case b_2 is 3 and a_1 is 1.

So, these are the, basically the rule mathematical representation of the different rules are there and then we can obtain the values of the outputs using this. This is basically the mathematical representation of the rule 1 and this is a mathematical representation of the rule 2 and mathematical rule 3 and rule 4. So, I have given this as an example, but actually the, these are the tricks by which this rule can be decided or defined in the system and once this rule is defined according the system, it is a straight forward to calculate in terms of the different value. For example here, I_1 is 6.0 and 2.2 and taking the v values 2 1. So, this rule I_1 can be calculated this 10.4.

So, this is the, I_1 one; y_1 the value is this one and for the, I_2 the value will be there and I_3 the value will be there and y_4 the value will be there. So, 4 rules related to I_1 is 6.0 and I_2 is 2.2 can be calculated, their corresponding output values. We have calculated the weights of all these rules also. So, these values then can be used finally to calculate the final output.

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Solution:

f) Therefore, the output y of the controller can be determined as follows.

$$y = \frac{w^1 y^1 + w^2 y^2 + w^3 y^3 + w^4 y^4}{w^1 + w^2 + w^3 + w^4}$$
$$y = 12.04$$

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So, the final output is the formula according the Takagi Sugeno approach. That says that, it is basically product of the weight, weighted output values. So, it is basically, these are the sum of the product of weighted output values, divided by the sum of the weights and if we follow the calculation as we have obtained in the last calculations, then finally the y value that can be obtained as this one.

So, that straight forward way from the rule base we can select the relevant rules and from the relevant rules we can calculate the weights of each rules and then output values of each rules and using this expression according the Takagi Sugeno approach we will be able to calculate the output value and once the output value is known then will we can take a decision and you can note that this is the output that we have obtained is crisps. As I told you, no need to do any, follow any defuzzification method in this case state way.

Here, basically state way actually, this is because state way for each rule we have taken the crisp value. All these y_1 , y_2 are all these are the crisp value for each rule. That is why, state way we can obtain the crisp value as the final output. So, this is the method that is followed there in Takagi Sugeno approach and so this is the way that the 2 controllers, the Mamdani approach and the Takagi Sugeno approach can be designed and 2 approaches have their own advantage as well as the disadvantage and as I told you, so Mamdani approach is easy to interpret, easy to interpret means rules are interpretable. On the other hand, the Takagi Sugeno approach the rules are difficult to interpret, so less interpretability. Difficult to interpret because, it is expressed in terms of some mathematical formula, which needs some coefficients to be design decided.

Now, deciding the coefficient it basically is a task for the fuzzy engineer or fuzzy designer. If the fuzzy engineer can decide the coefficient correctly or accurately, then it will give the accurate result. So, accuracy of the fuzzy logic controller according to the Takagi Sugeno approach solely depends on the performance or experience or prudence of the fuzzy designer. That is the only thing; otherwise, fuzzy logic controller according to the Takagi Sugeno approach is fast and more accurate, compared to the Mamdani approach.

So, it is up to the fuzzy designer, which approach he wants to follow, if it is a critical application, where accuracy and then the speed of the controller is very important, then they should follow Takagi Sugeno approach. Otherwise, both the approach are equally applicable to design any fuzzy logic controller. So, this is the 2 controller we have discussed and then

using this the concept, we can follow any fuzzy system can be designed from the fuzzy logic. So, this is end of the fuzzy logic discussion and we will study about some case studies that will be given as a some special problem for your practice.

Thank you.