

Sugeno Fuzzy Models

- The main difference between *Mamdani* and *Sugeno* is that the Sugeno output membership functions are either **linear or constant**.
- Mamdani-style inference requires finding the centroid of a two-dimensional shape by integrating across a continuously varying function. In general, this process is not computationally efficient.
- Michio Sugeno suggested to use a single spike, a *singleton*, as the membership function of the rule consequent. A **singleton**, or more precisely a **fuzzy singleton**, is a fuzzy set with a membership function that is unity at a single particular point on the universe of discourse and zero everywhere else.
- The Sugeno Fuzzy model (also known as the TSK fuzzy model) was proposed by Takagi, Sugeno, and Kang. A typical fuzzy rule in a Sugeno fuzzy model has the form:

if x is A and y is B then
$$z = f(x, y)$$

- Where A and B are fuzzy sets in the antecedent, while z= f(x, y) is a crisp function in the consequent.
- Usually f(x, y) is a polynomial of the input variables x and y. but it can be any function.
- When f(x, y) is a first-order polynomial, the resulting fuzzy inference system is called a *first-order Sugeno fuzzy model*, for example z = ax + by + c.
- When f is a constant, we then have a zero-order Sugeno fuzzy model,
 (z = ax + by + c where (a = b = 0)) which can be viewed either as a special case of the Mamdani Fuzzy inference system in which each rule's consequent is specified by a fuzzy singleton.



Each rule weights its output level, z_i, by the firing strength of the rule, w_i. For example, for an AND rule with Input 1 = x and Input 2 = y, the firing strength is

$w_i = AndMethod(F_1(x), F_2(y))$

Where **F1(x)**, **F2(y)** are the membership functions for Inputs 1 and 2.

• The final output of the system is the weighted average of all rule outputs, computed as

Final Output =
$$\frac{\sum_{i=1}^{N} w_i z_i}{\sum_{i=1}^{N} w_i}$$

Where **N** is the number of rules

• A Sugeno rule operates as shown in the following diagram.

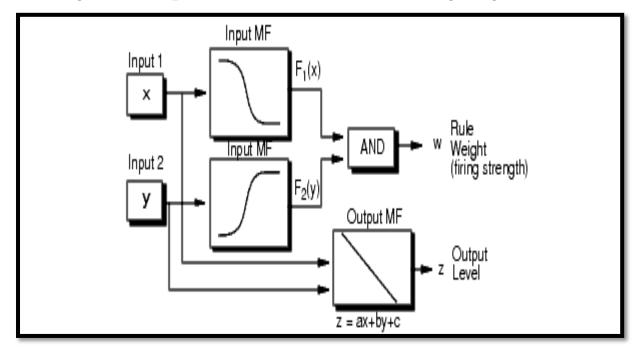
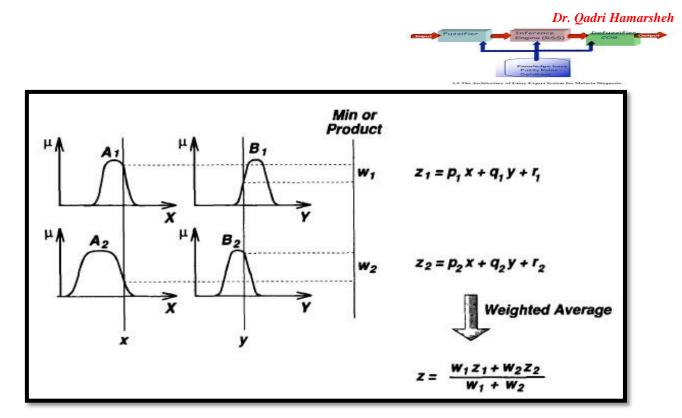


 Figure shows the fuzzy reasoning procedure for a first-order Sugeno fuzzy model. Since each rule has a *crisp output*, the overall output is obtained via *weighted average*, thus avoiding the time-consuming process of defuzzification required in a Mamdani model.

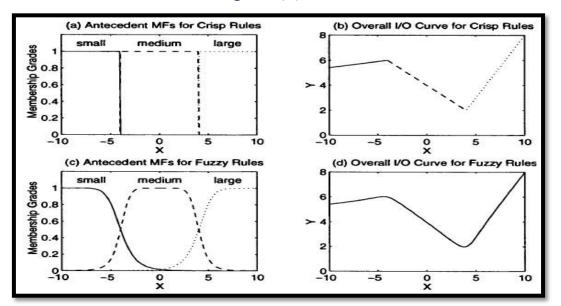


1st order Sugeno Fuzzy Model

• Example 1: Fuzzy and nonfuzzy rule set – a comparison

$$\begin{cases} If X is small then Y = 0.1X + 6.4 \\ If X is medium then Y = -0.5X + 4 \\ If X is large then Y = X - 2 \end{cases}$$

- If "*small*", "*medium*", and "*large*" are **nonfuzzy** sets with membership functions shown Figure (a), the overall input-output curve is **piecewise** linear, as shown in Figure (b).
- If "*small*", "*medium*", and "*large*" are **nonfuzzy** sets with membership functions shown Figure (c), the overall input-output curve becomes a smoother one, as shown in Figure (d).



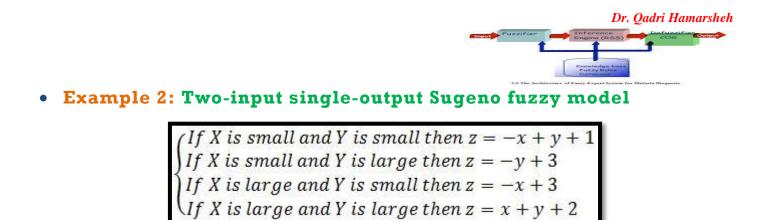
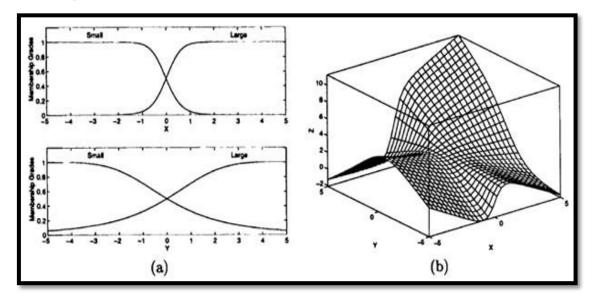
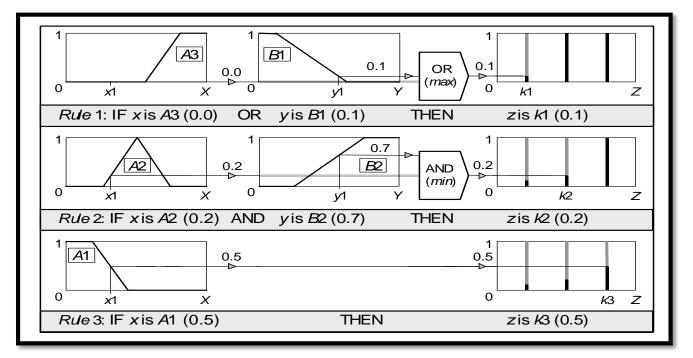


Figure (a) plots the membership functions of input X and Y, and Figure (b) is the resulting input-output surface. The surface is complex; the surface is composed of four planes, each of which is specified by output equation of a fuzzy rule.



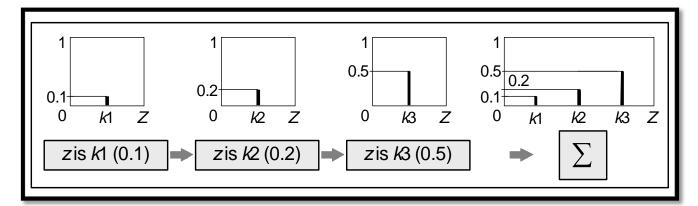
Example 3:

Sugeno-style rule evaluation





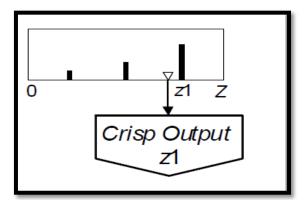
Sugeno-style aggregation of the rule outputs



Weighted average (WA):

$$WA = \frac{\mu(k1) \times k1 + \mu(k2) \times k2 + \mu(k3) \times k3}{\mu(k1) + \mu(k2) + \mu(k3)} = \frac{0.1 \times 20 + 0.2 \times 50 + 0.5 \times 80}{0.1 + 0.2 + 0.5} = 65$$

Sugeno-style defuzzification



How to make a decision on which method to apply – Mamdani or Sugeno?

- Mamdani method is widely accepted for capturing expert knowledge. It allows us to describe the expertise in more intuitive, more human-like manner. However, Mamdani-type fuzzy inference entails a substantial computational burden.
- On the other hand, Sugeno method is computationally effective and works well with optimization and adaptive techniques, which makes it very attractive in control problems, particularly for dynamic nonlinear systems.