# Adaptive Neuro Fuzzy Inference System (ANFIS)

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## Outline

- Fuzzy logic
- Fuzzy membership
- Fuzzy inference system
- Neural fuzzy inference system
  - Architecture
  - Learning model
  - Implementation

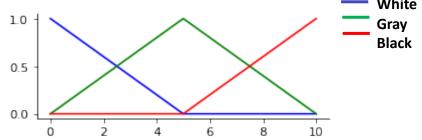
Activities in this session will be: Building a FIS and ANFIS

## Fuzzy Logic

- Fuzzy Logic teaches computers to make human-like decisions in an uncertain world.
- "The closer one looks at a real-problem, the fuzzier becomes the solution" Zadeh, 1973

 Can deal with the decision-making problems involving uncertainty and vagueness of real world applications.



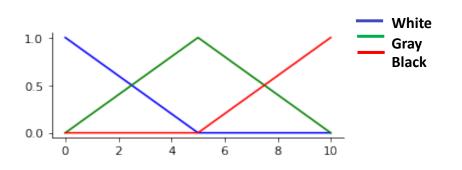


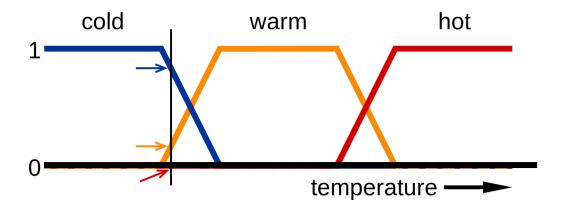
- Fuzzy Logic is a methodology predicated on the idea that the "truthiness" of something can be expressed over a continuum.
- This is to say that something isn't true or false but instead partially true or partially false.

• A **fuzzy variable** has a **crisp value** which takes on some number over a pre-defined domain (in fuzzy logic terms, called a **universe**). The crisp value is how we think of the variable using normal mathematics.

• A fuzzy variable also has several **terms** that are used to describe the variable. These terms are usually adjectives like "poor," "mediocre," and "good."

• Each term has a **membership function** that defines how a crisp value maps to the term on a scale of 0 to 1. In essence, it describes "how good" something is.





• A fuzzy system consists of a set of fuzzy **IF-THEN rules** that describe the input-output mapping relationship of the networks.

 These rules are simply mappings that describe how one or more fuzzy variables relates to another. These are expressed in terms of an IF-THEN statement; the IF part is called the antecedent and the THEN part is the consequent.

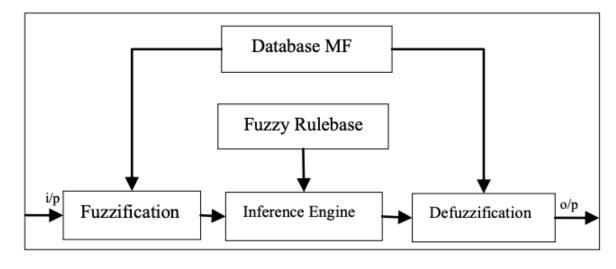
## Fuzzy Inference

• Fuzzy inference is the process of formulating input/output mappings using fuzzy logic.

 Knowledge is encoded as using a set of explicit linguistic rules, which can be easily understood by people without technical expertise.

• Fuzzy systems implement nonlinear systems using linguistic variables when adequate knowledge about the system is available.

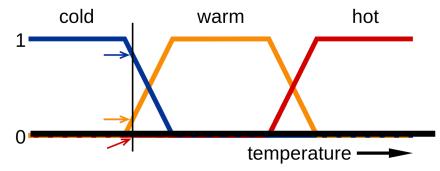
## Framework of Fuzzy Logic System



- Fuzzification, which translates crisp (real-valued) inputs into fuzzy values;
- Rule base reasoning, an inference engine that applies a fuzzy reasoning mechanism to obtain a fuzzy output using rules;
- Defuzzification, translates this latter output into a crisp value, as shown in the figure.

## **Fuzzification**

• The purpose of fuzzification is to map system input values from 0 to 1 via defined input membership functions.



- Input: A crisp value restricted to to the universe.
- Output: a fuzzy degree of membership in the qualifying linguistic set

$$\mu_A$$
: membership function  $\mu$ :  $\chi$ : universe set  $\mu$ :  $\chi$ :  $\chi$ 

# Rule-based Reasoning

 In rule-based reasoning, the fuzzy input values membership values are mapped to classify fuzzy output through a table containing if-then rules.

Rules are expressed as a logic implication p → q where p is called the
 antecedent of the rule and q is called the consequence of the rule.

## Defuzzification

• Defuzzification is a process which produces single system output (crisp) values by using a defuzzification formula and fuzzy output membership outputs.

## Tipping Example

#### Antecednets (Inputs)

- Service
  - Universe (ie, crisp value range): How good was the service of the wait staff, on a scale of 0 to 10?
  - Fuzzy set (ie, fuzzy value range): poor, acceptable, amazing.

#### Food quality

- Universe: How tasty was the food, on a scale of 0 to 10?
- Fuzzy set: bad, decent, great

#### Consequents (Outputs)

- Tip
  - Universe: How much should we tip, on a scale of 0% to 25%
  - Fuzzy set: low, medium, high

#### Rules

- IF the service was poor and the food quality was poor THEN the tip will be low.
- IF the service was average, THEN the tip will be medium.
- IF the service was good and the food quality was good, THEN the tip will be high.

Source: https://pythonhosted.org/scikit-fuzzy/auto\_examples/plot\_tipping\_problem\_newapi.html

## Usage



#### Rules

IF the service was poor and the food quality was poor THEN the tip will be low.

IF the service was average, THEN the tip will be medium.

IF the service was good and the food quality was good, THEN the tip will be high.

# Adaptive Neuro-Fuzzy Inference Systems (ANFIS)

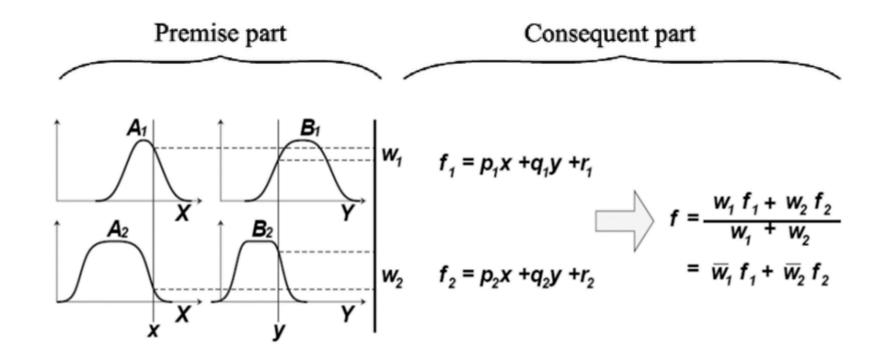
# Adaptive Neuro-Fuzzy Inference Systems (ANFIS)

• ANFIS is a kind of **artificial neural network** that is based on Takagi—Sugeno **fuzzy inference system**.

Network nodes in different layers have different structures.

## **ANFIS**

- Consider a first-order Sugeno fuzzy model, with two inputs, x and y, and one output, z.
- Rule set
  - Rule 1: If x is  $A_1$  and y is  $B_1$ , then  $f_1 = p_1x + q_1y + r_1$
  - Rule 2: If x is  $A_2$  and y is  $B_2$ , then  $f_2 = p_2x + q_2y + r_2$

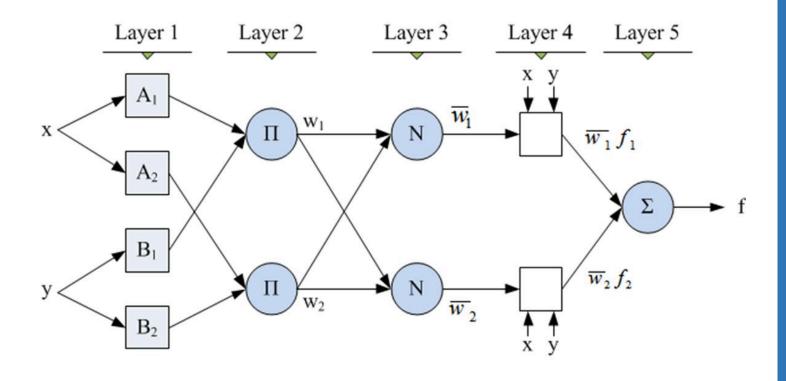


## **ANFIS Architecture**

- 5-Layer ANFIS architecture:
  - Fuzzy layer
  - Product layer
  - Normalized layer
  - De-fuzzy layer
  - Total output layer

• Square: Fixed node

• Circle: Adaptive node



Layer 1 Layer 2 Layer 3 Layer 4 Layer 5  $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_2$   $\overline{w}_2$ 

- Layer 1 Fuzzy Layer
- Layer 1: every node is an adaptive node with node function:

$$O_{1,i} = \mu_i (x_i)$$

• Parameters in this layer are called premise parameters.

Layer 1 Layer 2 Layer 3 Layer 4 Layer 5  $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_1$   $\overline{w}_2$   $\overline{w}_2$   $f_2$ 

- Layer 2: Product Layer
- Layer 2: every node is fixed whose output (representing firing strength) is the product of the inputs:

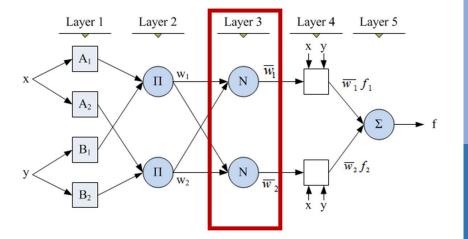
$$O_{2,i} = w_i = \pi_j \mu_i$$

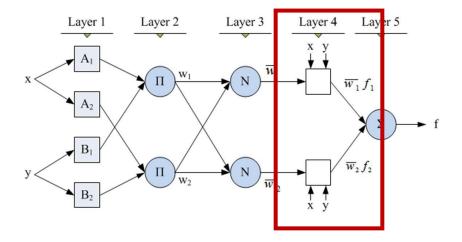
Any operator that performs fuzzy AND.

- Layer 3: Normalized Layer
- Layer 3: every node is **fixed** (normalization):

$$O_{3,i} = \overline{w_i} = \frac{w_i}{\sum_j w_j}$$

- Indicated normalization to the firing strength from previous layer.
- Output is called normalized firing strengths.





- Layer 4: De-fuzzy Layer
- Layer 4: every node is adaptive (consequent parameters):

$$O_{4,i} = O_{3,i}f_i = \overline{w_i}(p_0 + p_1x_1 + \dots + p_nx_n)$$

- Layer 5: Total Output Layer
- Layer 5: single fixed node, sums up inputs:

Layer 1 Layer 2 Layer 3 Layer 4 Layer 5 
$$\overline{W}_1$$
  $\overline{W}_1$   $\overline{W}_1$   $\overline{W}_2$   $\overline{W}_2$ 

$$O_{5,i} = \sum_{i} \overline{w_i} f_i = \frac{\sum_{i} w_i f_i}{\sum_{i} w_i}$$

Adaptive network is functionally equivalent to a Sugeno fuzzy model!

## Learning Algorithm of ANFIS

the membership function parameters that best allow the associated fuzzy inference system to track the given input/output data.

- The parameters associated with the membership function change through the learning process.
- Goal: tune all the modifiable parameters, to formulate the ANFIS output match the training data.

	Forward Pass	Backward Pass
<b>Premise Parameters</b>	Fixed	<b>Gradient descent</b>
<b>Consequent Parameters</b>	Least square	Fixed
Signals	Node output	Error signal



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