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Master of Technology in Knowledge Engineering

Unit 7:

Developing Intelligent Systems for Performing Business Analytics

Fuzzy Modelling

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Why Fuzzy Modelling

• In general, we design a fuzzy inference system based on the past known behaviour of a target system. The fuzzy system is then expected to be able to reproduce the behaviour of the target system.

» E.g.:

Target system: a human operator in charge of a chemical reaction process

Fuzzy system: a fuzzy controller

» E.g.:

Target system: a medical doctor

Fuzzy system: a fuzzy expert system for medical diagnosis

• The process called *Fuzzy modelling* is the standard method for constructing a fuzzy inference system.



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What is Fuzzy Modelling

- Features of fuzzy modelling
 - » Fuzzy modelling takes advantage of domain knowledge that might not be easily or directly employed in other modelling approaches.
 - » The rule structure of a fuzzy inference system makes it easy to incorporate human expertise about the target system directly into the modelling process.
 - » In fuzzy modelling, the use of numerical data also plays an important role.
- Conceptually, there are 2 stages of fuzzy modelling
 - **»** The identification of the surface structure
 - » The identification of deep structure



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Identification of Surface Structure — the first stage of fuzzy modelling

- Generally, it includes four tasks:
 - » Select relevant input and output variables
 - » Choose a specific type of fuzzy inference system
 - » Determine the number of linguistic terms associated with each input and output variables. The membership function of linguistic terms is determined in the second stage
 - » Design a collection of fuzzy if-then rules (fuzzy rule base) in a *symbolic* style:
 - i.e. IF x is A THEN y is B
- In this stage, we basically rely on
 - » Our own knowledge (common sense, simple physical laws, etc) of the target system
 - » Information provided by human experts who are familiar with the target system (which could be the human experts themselves)





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Identification of Deep Structure

— the second stage of fuzzy modelling

- Specifically, it may include the following tasks:
 - » Choosing an appropriate family of *parameterised* membership functions used in the rule base
 - **♦** Piecewise linear function
 - Triangular
 - Trapezoidal
 - **♦** non-linear, differentiable function:
 - Gaussian
 - Bell
 - » Interview human experts familiar with the target systems to determine the parameters of the membership functions used in the rule base
 - » Learning/Refining the parameters by certain optimisation techniques
 - **♦** Learning membership functions from training data: NN approaches
 - **♦** Optimisation: GA





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Brief Review of Fuzzy Inference System

- Basic architecture
 - » rule base
 - » data base
 - » reasoning mechanism
- Different fuzzy models
 - » Mamdani fuzzy model
 - **♦** first attempt to control system
 - » Sugeno fuzzy model (TSK model)
- Main steps
 - » Fuzzification
 - » Rule evaluation
 - » Defuzzification

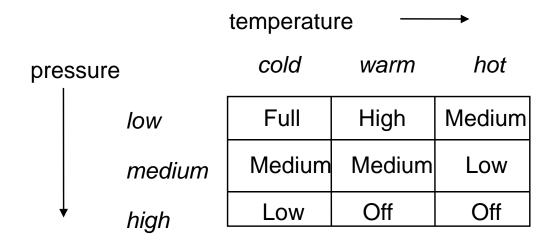




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Visualising Fuzzy Rules

- A fuzzy associative matrix (FAM) shows the maximum set of possible rules for a problem
- In the example there are nine possible rules to control a heater.



• Which box corresponds to the rule below?

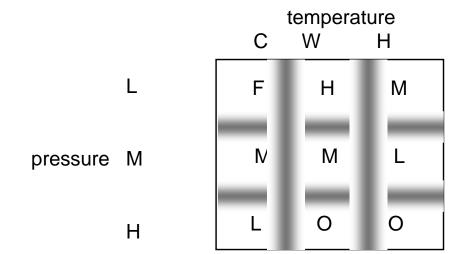
"If Temperature is WARM and Pressure is LOW Then set Heat to HIGH"





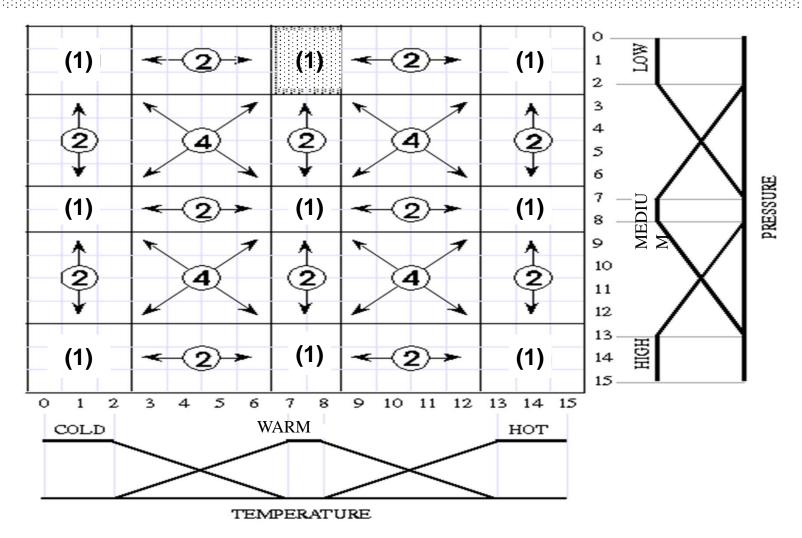
Visualising Fuzzy Rules (cont.)

• Both Temperature and Pressure are linguistic variables, the boundaries between their values (i.e. cold and warm) are not sharp





Visualising Fuzzy Rules (cont.)



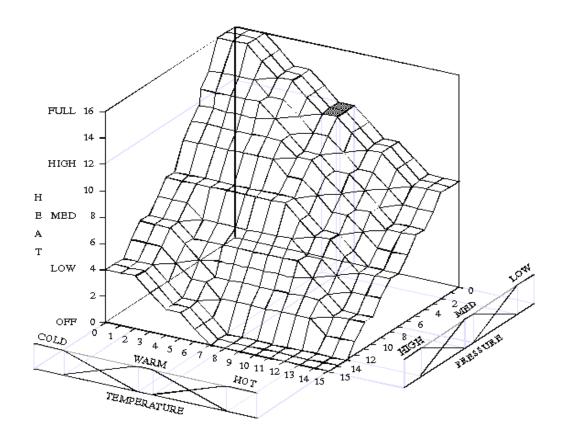
The numbers indicate how many rules can apply to that box.





Visualising Fuzzy Rules (cont.)

In the areas where more than one rule contributes to the output level, defuzzification is used to calculate a weighted average of the contributing rules. The result can be plotted as a control (or decision) surface.

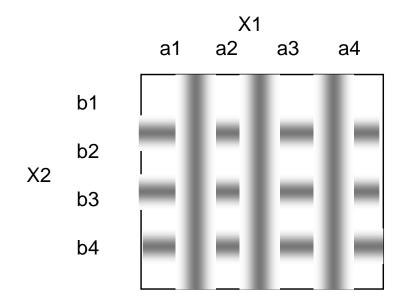




Input Space Partitioning — Grid partition

• Grid partition

- » Needs a small number of membership functions for each input.
- » Encounters problems when we have a large number of inputs.

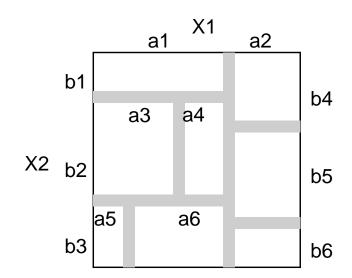




Input Space Partitioning — Tree partition

Tree partition

- » Each region is uniquely specified using a leaf in a decision tree
- » Relieves the problem of an exponential increase in the number of rules (in grid partition)
- » More membership functions for each input are used
- » Fuzzy membership functions do not usually bear clear linguistic meanings such as "small", "large", etc.
- Example: Two inputs, X1 and X2, one output Y
 - » X1 has six fuzzy values: a1, a2, a3, a4, a5, a6
 - **X2** has six fuzzy values: b1, b2, b3, b4, b5, b6



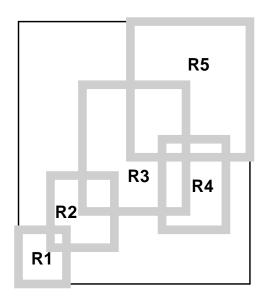




Input Space Partitioning — Scatter partition

Scatter partition

- » By covering a subset of the whole input space, can limit the number of rules to a reasonable amount
- » It is usually dictated by desired input-output data pairs





Exercise

Problem description:

An insurance company invented a new product (policy). The new product provides policy holder a choice of having an "investment" component linked with his/her normal insurance component. Comparing with the normal insurance, of course, the "investment" component takes a higher risk and hopefully obtains a better return.

It is generally believed that young people can bear relatively higher risk. Apart from age, family situation and income level of a customer may also affect his/her bearable level of risk.

So an insurance agent has to recommend a suitable amount of the "investment" based on the customer's age and bearable risk level.



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Exercise (cont.)

Problem description (cont.):

The insurance company has decided:

- » the amount of investment should be within the range of \$10,000 ~ \$1,000,000
- » to have an investment component, the customer's age should be between 20 to 50.
- The risk level is measured by the agent based on the bearable percentage (0% ~ 100%) of loss. e.g. a customer may provide his bearable loss as 50% of total investment money.

Some knowledge/experience from Mr. A (an insurance agent):

- » The older the person, the lower the investment amount.
- » The higher the bearable risk level, the higher the investment amount.





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Exercise (cont.)

- Your task:
 - » develop a fuzzy expert system to reproduce the work of Mr. A
- Possible steps of fuzzy modelling:
 - » identification of surface structure
 - input/output variables
 - **♦** type of fuzzy inference system
 - **♦** number of linguistic terms (fuzzy partition)
 - **♦** design rule base
 - » identification of deep structure
 - ♦ determine membership function for each linguistic term



