

Computational Intelligence

An Introduction

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July 28, 2023



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 - In classical AI, human experts would manually encode rules and knowledge into the AI system.
 - The focus was on symbolic reasoning and logic to process information and make decisions.
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 - Conventional AI lacked the ability to learn from data and adapt to new situations, as it primarily relied on explicit rules and pre-defined knowledge.
 - It had limitations in handling complex and real-world problems due to the rigid nature of rule-based systems.



- Modern AI: It is also known as Computational Intelligence. It shifted the focus from rule-based systems to data-driven approaches, where AI models learn patterns and relationships directly from data.
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 - The development of advanced algorithms and computational power enabled the rise of Machine Learning, Deep Learning, and other statistical techniques.
 - Modern AI can handle large amounts of data and identify complex patterns that may not be easily discernible by human experts.

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- Computational Intelligence has led to significant advancements in various applications, such as natural language processing, computer vision, speech recognition, and robotics, and has become an integral part of many industries and technologies in the present era.

Formal Computing

- The concept of computing refers to the process of using computers or computational systems to perform various tasks, solve problems, and process information.

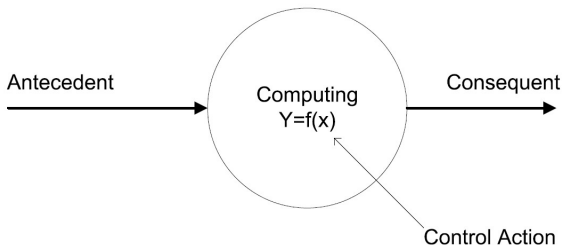


Figure: Basic of computing

- $y = f(x)$, f is a mapping function. f is also called a formal method or an algorithm to solve a problem.

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- Control action is formally defined (i.e. with mathematical model)

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- Searching and sorting techniques
- Solving graph problems (e.g. Shortest tour in Graph theory, Finding closest pair of points, etc.)



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- Network optimization

Soft Computing

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- The applications of Soft Computing have proved two main advantages. First, it made solving real world complex nonlinear problems, in which mathematical models are not available.
- Second, it introduced the human knowledge such as cognition, recognition, understanding, learning, and others into the fields of computing.

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- Algorithms are adaptive (i.e. it can adjust to the change of dynamic environment)
- Use some biological inspired methodologies such as genetics, evolution, Ant's behaviors, particles swarming, human nervous systems etc.

Methodologies of Soft Computing

Soft computing is a collection of computing techniques and methodologies that aim to deal with complex and uncertain problems by tolerating the notions of imprecision, vagueness, and partial truth.

- Fuzzy System: Imprecision and reasoning

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- Evolutionary Computation: Searching and optimization

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- The fuzzy set for representing the linguistic variable human age is partitioned into three fuzzy subsets, namely, young middle-age, old. Each fuzzy subset is characterized by an MF.
- A linguistic variable is a variable whose values are linguistic terms in a natural or artificial language. For example, the size of an object is a linguistic variable, whose value can be small, medium, and big.

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- The membership functions can take various shapes, such as triangular, trapezoidal, Gaussian, or sigmoidal, depending on the nature of data and application.



A fuzzy partition of human age

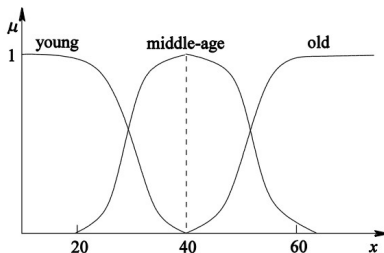
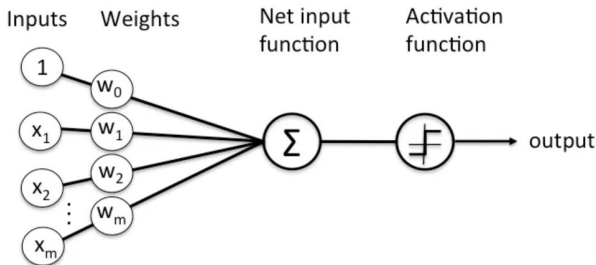


FIGURE 1: A fuzzy partition of human age. The fuzzy set for representing the linguistic variable *human age* is partitioned into three fuzzy subsets, namely, *young*, *middle-age*, *old*. Each fuzzy subset is characterized by an MF.

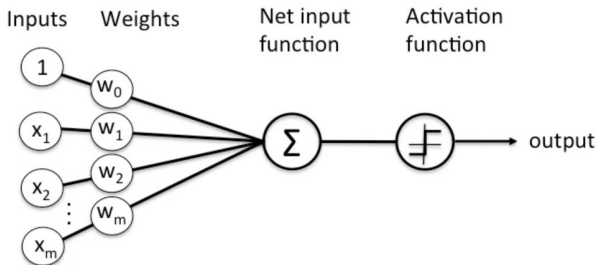
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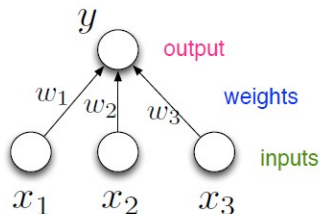


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- Neural networks are machine learning models inspired by the structure and functioning of the human brain.
- We can express a neural network as, it consists of interconnected artificial neurons, also known as nodes or units, organized into layers.
- The fundamental building block of a neural network is the neuron, which takes inputs, performs a computation, and produces an output.



Mathematical representation of Neural Network



$$y = g \left(b + \sum_i x_i w_i \right)$$

Diagram illustrating the mathematical representation of a neural network node. The equation is $y = g \left(b + \sum_i x_i w_i \right)$. Annotations include:

- y : output (pink arrow)
- g : nonlinearity (red arrow)
- b : bias (blue arrow)
- x_i : i'th input (green arrow)
- w_i : i'th weight (blue arrow)

Figure: Mathematical representation of Neural Network

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- A **generalization phase**: unseen samples (not used in the training) are tested with the neural networks model and a performance index is computed. This index characterizes the quality of the model.



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- Continue same process for number of epochs until error is minimized

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- But engineering problems' cost or loss functions are mostly complex and discontinuous in nature.
- However, evolutionary algorithms can handle these issues because it is derivative free and randomized optimization.



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- Evolutionary algorithms are efficient at exploring a wide search space, making them well-suited for finding global optima in complex and multi-modal optimization problems.

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- By combining/hybridizing two or more methodologies such as fuzzy logic, neural networks, and evolutionary algorithm methodologies together, Computational model provides powerful tools for tackling complex and dynamic challenges in diverse domains.
- Truly intelligent machines - Computers that can learn on their own and take decision as human being.