



# AI Searching Techniques

Computational Intelligence

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# Last-Time

- The Local-Global Principle for Similarity Measures
- Graph Representations and Graph Similarities
- Taxonomic Similarities

# Learning Objectives

- Introduce the students to the theory of computational intelligence.
- Learn about the different computational intelligence paradigms
- Understand how to apply CI paradigms to solve different problems

# Outlines

Computational Intelligence

Computational Intelligence Paradigms

# Computational Intelligence: Definition

## What is computational Intelligence?

- Is a **branch of AI**, that focus on building machine intelligence by using a set of **nature-inspired algorithms** and approaches that address complex real-world problems to which mathematical or **traditional modeling** is **inefficient**.
- Computational Intelligence achieve intelligence by modeling biological and natural intelligence.
- The study of adaptive mechanisms to enable or facilitate intelligent behavior in complex and changing environments.

# Computational Intelligence: Main Paradigms

1. Artificial Neural Networks (ANN)
2. Evolutionary Computation (EC)
3. Swarm Intelligence (SI)
4. Artificial Immune Systems (AIS)
5. Fuzzy Systems (FS)

# Computational Intelligence: Main Paradigms

- Each of the CI paradigms has its origins in biological systems:
  - ANNs model biological neural systems
  - EC models natural evolution (including genetic and behavioral evolution)
  - SI models the social behavior of organisms living in swarms or colonies
  - AIS models the human immune system
  - FS attempts to resemble human reasoning in decision making

# Computational Intelligence: Definition

- Computational Intelligence and **Soft Computing**
  - The term coined by **Lotfi Zadeh**. Soft computing techniques resemble **biological processes** more closely than traditional techniques.
  - It is a different grouping of CI paradigms. It include **support vector machine** and **bayesian networks**.
  - There is no a general agreement on what is computational intelligence



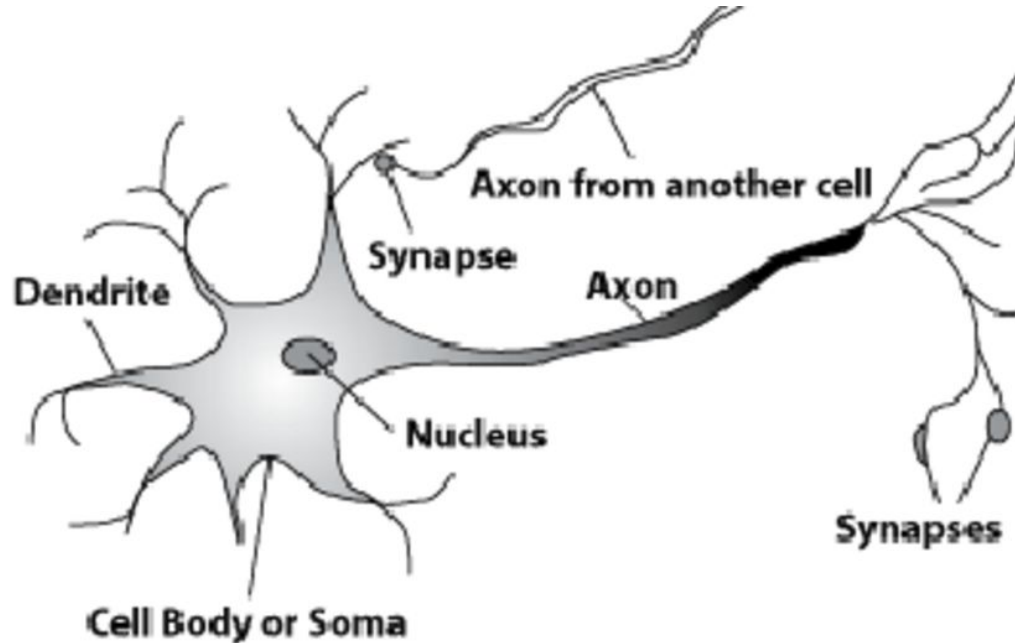
# Artificial Neural Networks (ANN)

The ability of the human brain to perform complex tasks such as **pattern recognition**, **perception** and **motor control**, **learn**, **memorize**, generalize and **reuse knowledge** prompted research in algorithmic modeling of **biological neural systems**.

It is estimated that there is in the order of **10-500 billion neurons** in the human cortex, with 60 trillion synapses. The neurons are arranged in approximately **1000 main modules**, each having about **500 neural networks**.

# Artificial Neural Networks (ANN)

The basic building blocks of biological neural systems are nerve cells, referred to as neurons.



# Artificial Neural Networks (ANN)

Neurons are **massively interconnected**, where an interconnection is between the axon of one neuron and a dendrite of another neuron. This connection is referred to as a synapse. This connection is referred to as a synapse.

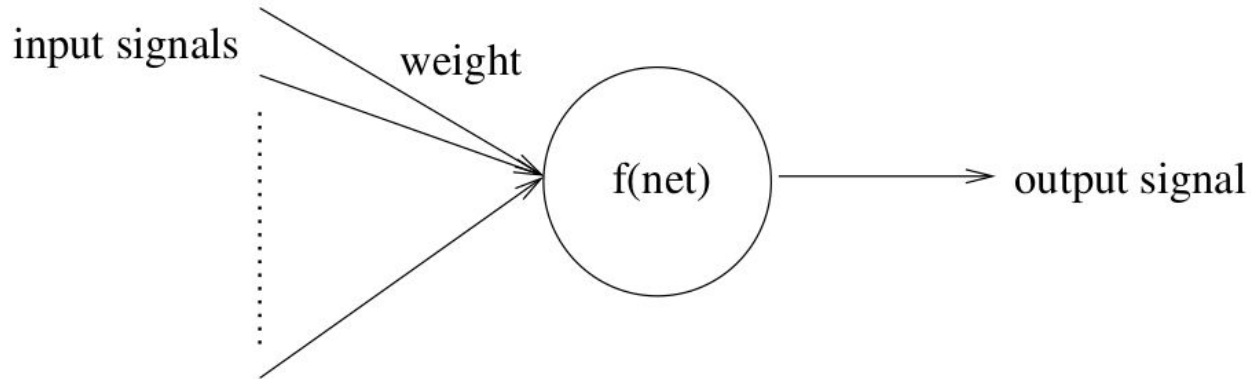
**Signals propagate** from the dendrites, through the cell body to the axon; from where the signals are propagated to all connected dendrites.

A **signal** is transmitted to the axon of a neuron only when the cell **"fires"**. A neuron can either **inhibit** or **excite** a signal.

# Artificial Neural Networks (ANN)

An artificial neuron (AN) is a model of a biological neuron (BN).

Each AN receives signals from the environment, or other ANs, gathers these signals, and when fired, transmits a signal to all connected ANs.



# Artificial Neural Networks (ANN)

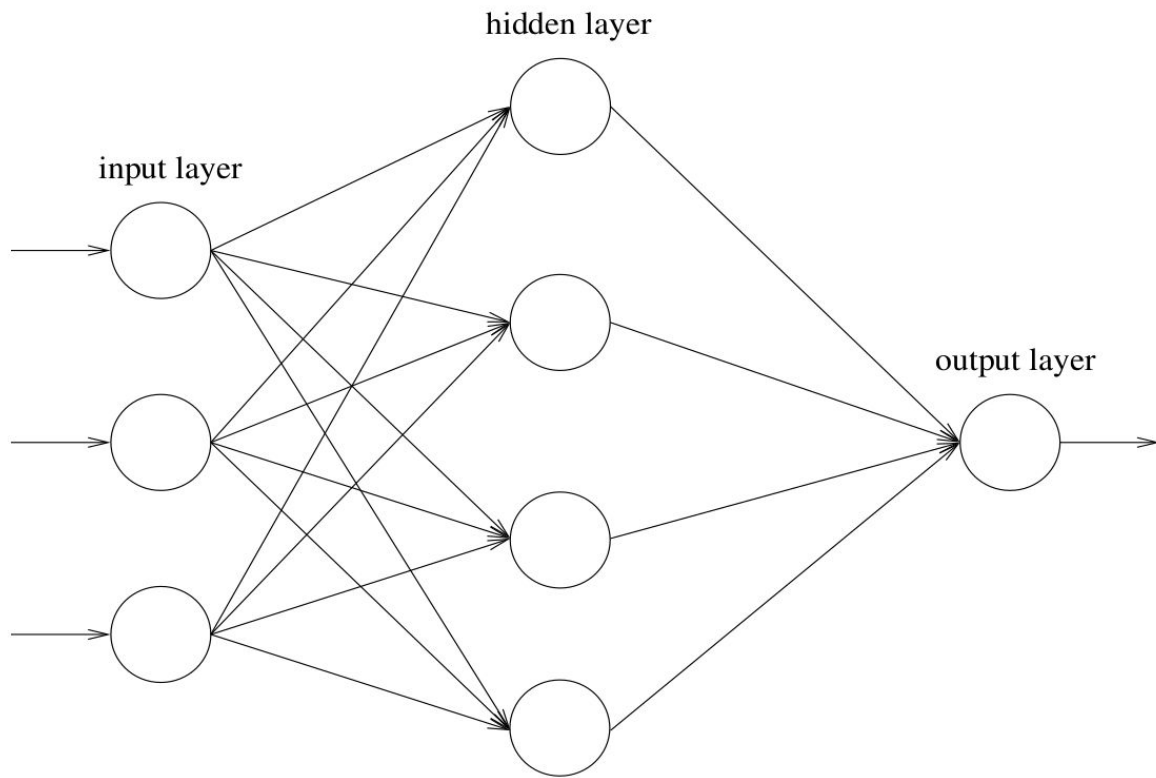
- **Input signals** are inhibited or excited through **negative** and **positive numerical weights** associated with each connection to the AN.
- The firing of an AN and the strength of the exiting signal are controlled via a function, referred to as the **activation function**.
- The AN collects all incoming signals, and computes **a net input signal** as a function of the respective weights.
- The net input signal serves as input to the **activation function** which calculates the **output signal** of the AN.

# Artificial Neural Networks Structure

## What is an ANN?

- An artificial neural network (NN) is a layered network of ANs.
- An NN may consist of an input layer, hidden layers and an output layer.
- ANs in one layer are connected, fully or partially, to the ANs in the next layer.
- Feedback connections to previous layers are also possible.

# Artificial Neural Networks (ANN)



# Artificial Neural Networks (ANN)

Several different NN types have been developed:

- [Single-layer NNs](#), such as the Hopfield network
- [Multilayer feedforward NNs](#) (backpropagation NNs)
- [Temporal NNs](#), such as the Elman and Jordan simple recurrent networks as well as time-delay neural networks.
- [Self-Organizing NNs](#), such as the Kohonen self-organizing feature maps and the learning vector quantizer



# Artificial Neural Networks (ANN)

Current [successes](#) in neural modeling are for small artificial NNs aimed at [solving a specific task](#).

NN types have been used for a wide range of applications, including diagnosis of diseases, [speech recognition](#), [data mining](#), composing music, image processing, forecasting, robot control, credit approval, classification, pattern recognition.

# Evolutionary Computation: Definition

It mimics processes from natural evolution. The key concept in EC is **survival of the fittest**.

In natural evolution, survival is achieved through **reproduction**.

**Offspring**, reproduced from two parents, contain genetic material of both parents – hopefully the best characteristics of each parent.

Those individuals that inherit bad characteristics are weak and lose the battle to survive.

# Evolutionary Computation

- Uses a population of individuals, where an individual is referred to as a **chromosome**.
- A chromosome defines the characteristics of individuals in the population.
- Each characteristic is referred to as a **gene**.
- The value of a gene is referred to as an allele.
- For each generation, individuals compete to reproduce offspring.

# Evolutionary Computation

- Offspring are generated by combining parts of the parents, a process referred to as **crossover**.
- Individual in the population can also undergo **mutation** which alters some of the genes of the chromosome.
- The survival strength of an individual is measured using a **fitness function** which reflects the objectives and constraints of the problem to be solved.
- After each generation, individuals may undergo **culling**, or individuals may **survive** to the next generation.

# Evolutionary Computation

Different classes of evolutionary algorithms (EA) have been developed:

- **Genetic algorithms** which model genetic evolution.
- **Genetic programming** which is based on genetic algorithms, but individuals are programs (represented as trees).
- **Cultural evolution** which models the evolution of culture of a population and how the culture influences the genetic and phenotypic evolution of individuals.
- **Coevolution** where initially “dumb” individuals evolve through cooperation, or in competition with one another, acquiring the necessary characteristics to survive.

# Evolutionary Computation

Evolutionary computation has been used successfully in real-world applications:

- Data mining
- Combinatorial optimization
- Fault Diagnosis
- Classification
- Clustering
- Scheduling

# Swarm Intelligence

- Artificial Swarm Intelligence → Many minds are better than one
- Social creatures (birds, bees, fish, wolves, ants, spiders) can work together as a unified, decentralized, self-organized, and dynamic system to accomplish specific goals.
- Originated from the study of colonies, or swarms of social organisms.
- Studies of the social behavior of organisms (individuals) in swarms prompted the design of very efficient optimization and clustering algorithms.
- Collaborative behaviors of swarms and colonies are achieved via subtle connections that establish real-time feedback-loops among members

# Swarm Intelligence

## Examples of Artificial Swarm Intelligence

- Particle Swarm Optimization.
- Ant Colony Optimization.
- Bees Algorithm.
- Wolf Pack Algorithm.
- Social Spider Swarm Optimization.
- Fish School Search.



# Artificial Immune Systems

Human Immune System has an amazing **pattern matching** ability, used to distinguish between foreign cells entering the body (referred to as **non-self**, or antigen) and the cells belonging to the body (referred to as **self**).

Different computational models are inspired from HIS:

- Positive and Negative Selection
- Clonal Selection Theory
- Danger Theory

# Artificial Immune Systems

## Applications of Artificial Immune System

- Pattern Recognition
- Anomaly Detection
- Classification
- Computational Optimization

The challenge when building AIS is how we can model the task as an immune process.

# Fuzzy Systems

- Introduced by Lotfi Zadeh in 1965 ( fuzzy set theory)
- An element belongs to a set to a certain degree of certainty.
- Model the human reasoning process, where the decisions are not simply binary decisions (true or false)
- Fuzzy sets and logic allow the modeling of common sense and can handle uncertainty
- Fuzzy sets and fuzzy logic allow what is referred to as approximate reasoning.
- The challenge is how to design and select fuzzy function

# Fuzzy Systems

Fuzzy systems are usually implemented a set of IF-THEN rules

- If temperature is very cold turn off fan
- If temperature is cold turn down fan
- If temperature is normal maintain fan speed
- If temperature is hot speed up fan

Applications of Fuzzy Systems include:

- Control systems, Automotive, Defense , Cyber Security,

# Questions