
Neural Network Zoo

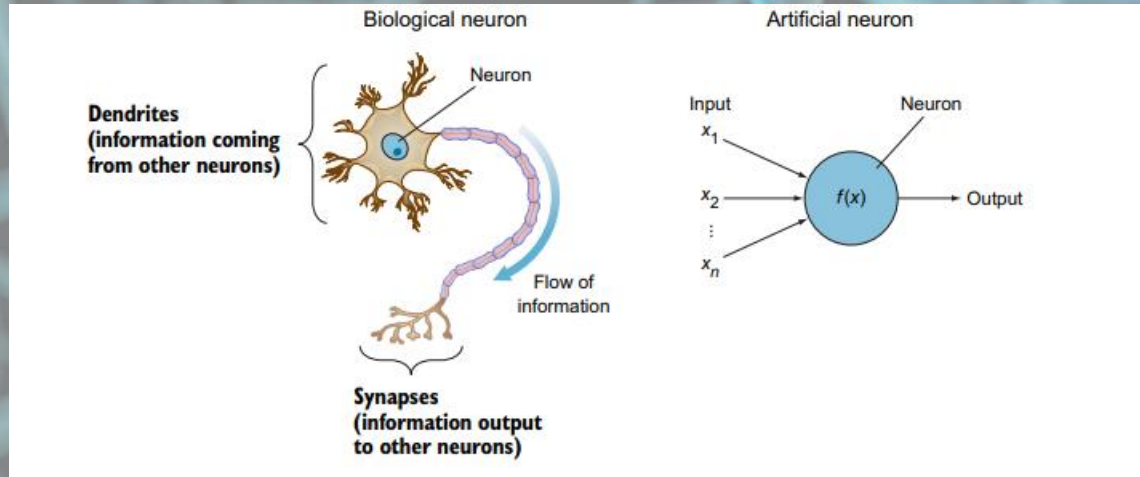
SAIMA SANO

ITAI 2376

ANNA DEVARAKONDA

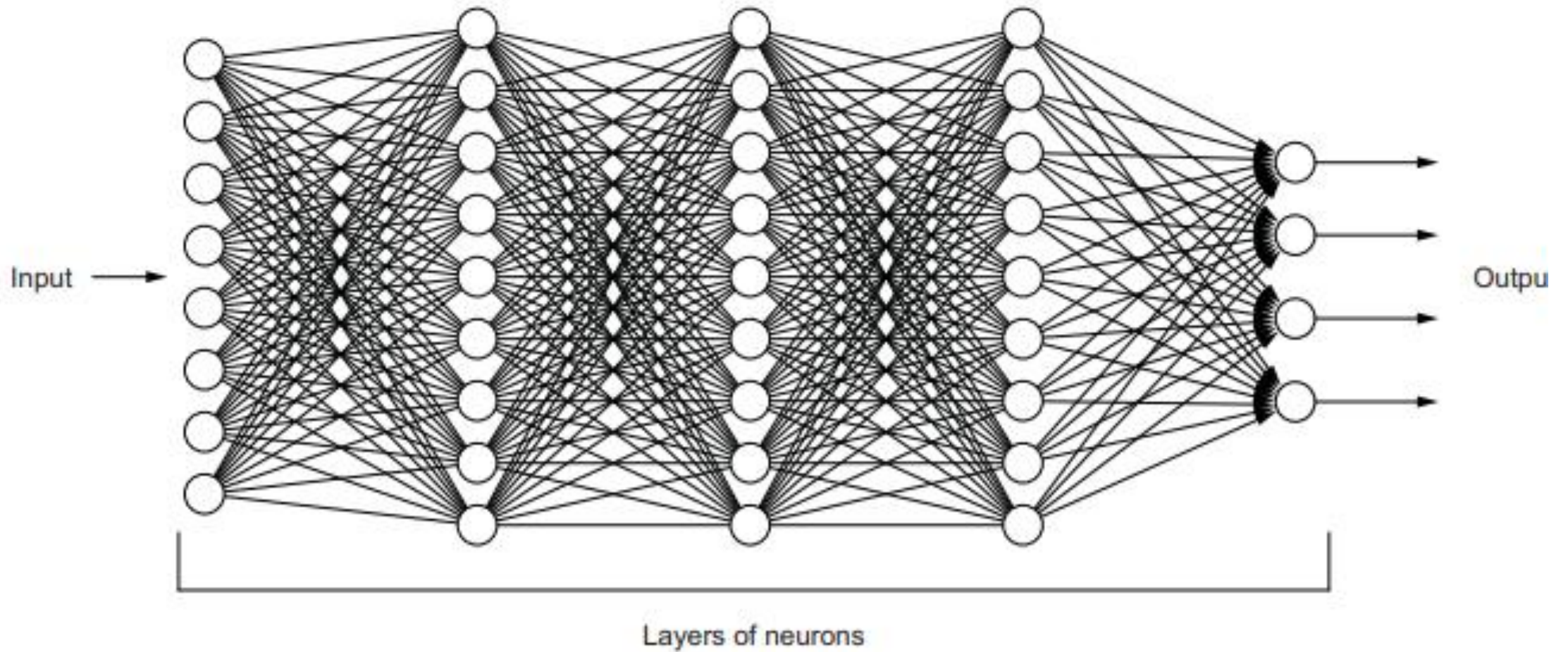
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What is Neural Network?

A **neural network** is made up of layers of interconnected nodes (or "neurons") that process data by assigning weights and applying activation functions. These networks learn to recognize patterns and make decisions through training on large datasets. It consists of three layers. **Input Layer** which receives the initial data. **Hidden Layers** are intermediate layers that perform complex computations and feature extraction. **Output Layer** is the last layer that produces the final result (e.g., classification or prediction). Each connection between nodes has a weight, and each node applies an activation function (like ReLU, sigmoid, or tanh) to determine whether and how much signal to pass on.



What is Deep Learning?

Deep learning is a type of machine learning that uses multi-layered neural networks to automatically learn representations and features from raw data with minimal human intervention.

Pattern Spotter – The Eagle

Convolutional Neural Network

An eagle's extraordinary visual acuity allows it to pinpoint distant prey with precision, Convolutional Neural Network (CNN) processes vast visual data to classify images and detect specific objects within them. Both excel at identifying targets within complex visual fields, albeit one through biological evolution and the other through advanced artificial intelligence





The Silent Reconstructor – The Owl

AutoEncoder

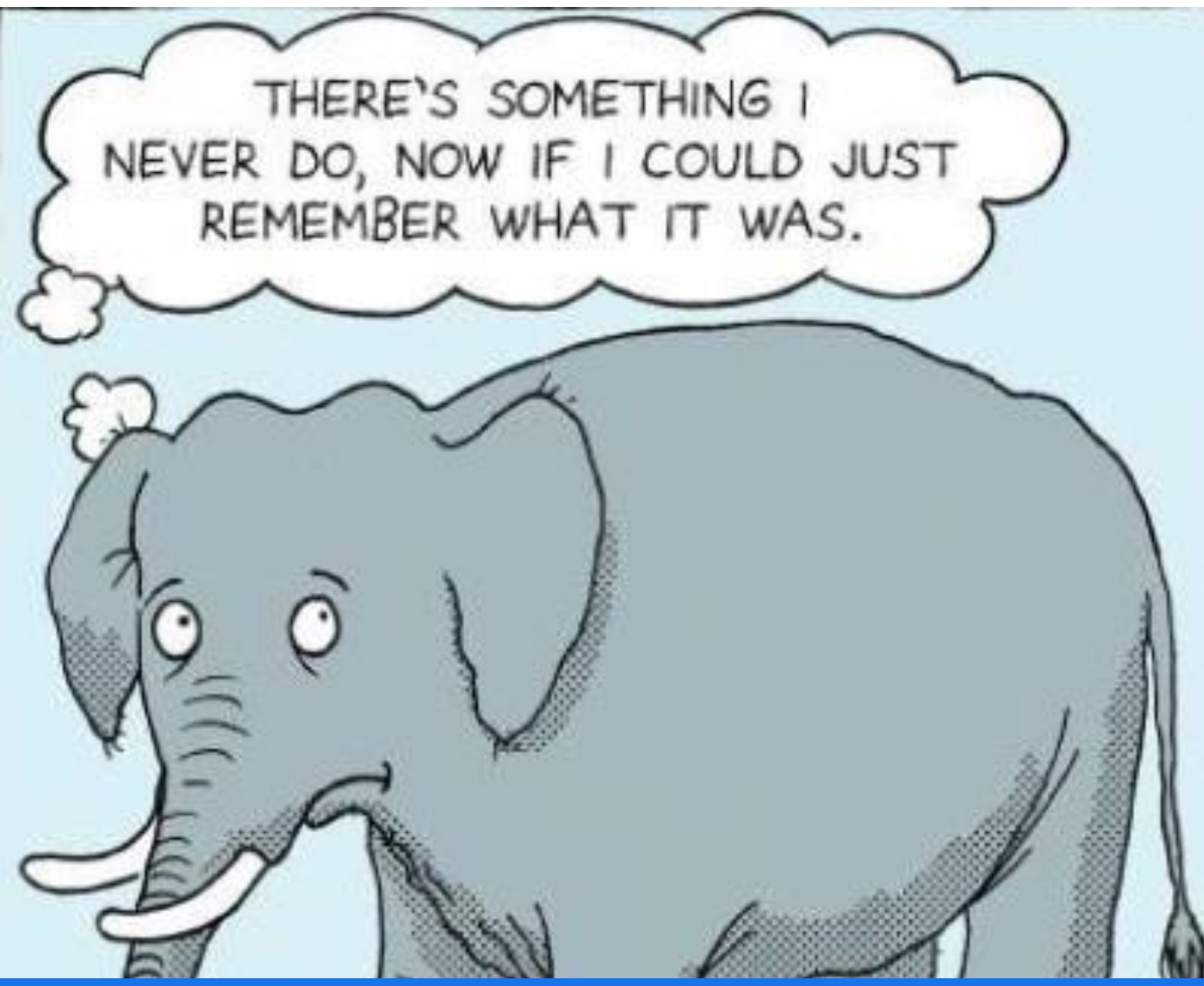
An autoencoder, by compressing and reconstructing data while learning to ignore noise, mirrors an owl's ability to filter out ambient sounds to pinpoint the precise location of its prey. Both systems excel at extracting critical information from complex inputs, one through neural network architecture and the other through highly evolved sensory perception.



The Sequence Mover – The Snake

Recurrent Neural Network

A snake moves sequentially, segment by segment, with its current position influencing the next, much like a Recurrent Neural Network (RNN) processes data in a sequence. An RNN's internal "memory" allows information from previous steps to influence the processing of current inputs, enabling it to understand context and dependencies in sequential data such as speech or text.



The Memory Guardian _ The Elephant

Long Short-Term Memory

An LSTM, or Long Short-Term Memory network, excels at remembering important information over long sequences and selectively forgetting irrelevant details, akin to an elephant's remarkable long-term memory and ability to recall specific, vital details across its lifespan. Both are designed to manage and retain crucial information over extended periods, whether for understanding complex patterns in data or navigating a vast environment.



The Multi-Tasker – The Octopus

Transformer

An octopus, with its decentralized nervous system and eight highly independent arms capable of performing distinct tasks simultaneously while maintaining overall coordination, resembles the parallel processing and distributed attention mechanisms of a Transformer neural network. Both are adept at integrating diverse, independently processed information to form a comprehensive understanding or achieve a complex goal.



The Steady Learner – The Tortoise

Feedforward Neural Network

A feedforward neural network processes information in one direction, from input to output, without cycles or memory of past inputs, much like a tortoise steadily moving forward along a predetermined path. Both systematically progress through their defined layers or movements without looking back or adapting based on prior steps.



The Distance Tracker – The Bee

Radial Basis Function

A Radial Basis Function (RBF) Network responds most strongly to inputs similar to those it has "learned" (its centers), much like a bee's specialized foraging behavior where it focuses on and extracts resources most efficiently from flowers within a specific, localized area around its hive. Both systems exhibit a localized response, where their activity peaks in proximity to certain optimal points or patterns.



Similarities and Differences Between Different Network

The "Neural Network Zoo" exercise is an excellent opportunity for reflection and deeper understanding. As I have gotten to know the unique "animals" in our zoo – the CNN Eagle, RNN Snake, LSTM Elephant, and so on. I have seen how each network, while keeping the general neuron-and-layer framework, is best suited for specific tasks. The CNN Eagle, with its convolution layers and pooling, is also adept at detecting spatial patterns and is thus a good fit for viewing images and videos, much like an eagle well-developed visual acuity for stalking. The RNN Snake, with its circular nodes, is made for sequential data and is thus suitable for natural language processing or time series analysis, much like a snake's memory and ability to track complicated routes. The LSTM Elephant, an improvement on RNN, removes short-term memory issues so that it can more easily learn longer-term dependencies in sequences, just like an elephant's long-term memory for complex social hierarchies. By contrasting these "animals," we more clearly see that although all neural networks are capable of learning from data, their differing structures (local receptive fields in CNNs, recurrent connections in RNNs/LSTMs) pretrain them to be extremely well-suited to learn different types of data and sets of problems and depict the diverse and compelling applications of deep learning in numerous disciplines.



Work Cited

- Goodfellow, I., Bengio, Y., & Courville, A. (2016). [*Deep Learning*](#). MIT Press.