

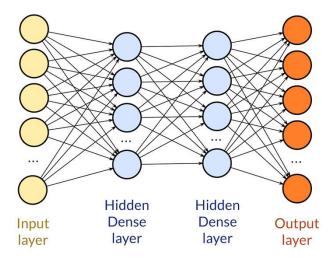
# Deep Learning

#### **Course Instructor:**

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### **Deep Learning**

- Machine Learning is based on Statistical structure.
- Deep Learning is a subfield of Artificial Intelligence and Machine Learning that is inspired by the structure of a human brain.
- Deep Learning algorithms attempt to draw similar conclusions as humans would by continually analyzing data with a given logical structure called Neural Network.



# Why Deep Learning is getting so famous?

- Deep Learning is a part of a broader family of machine learning methods based on artificial neural networks with representation learning.
- Representation Learning a.k.a feature learning is a set of technique that allows a system to automatically discover the representations needed for feature detection or classification from raw data.
- *In DL Feature Engineering is not necessary.*
- **Applicability:** Computer Vision, Speech Recognition, Image Processing, Bioinformatics, Drug Design, etc.
- Performance

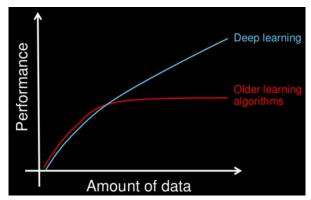






# Vs Machine Learning

- Data Dependency
- Hardware Dependency
- Training Time
- Feature Selection



Eg: In Resume based prediction: Features such as 10<sup>th</sup> percentage, 12<sup>th</sup> percentage, No. of achievements, No. of courses, etc.

Interpretability

# Why Now?

- Public Available Datasets:
  - Image
  - Video
  - Text
  - Audio
- Hardware Cost Reducing
- Availability of different Frameworks:
  - TensorFlow Google
  - PyTorch Facebook
  - Keras
- Deep Learning Architectures

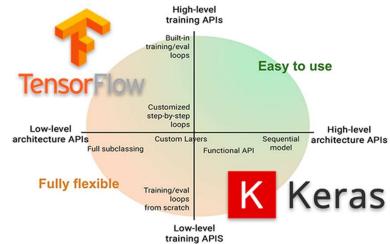












# State-of-theart Architectures

Image Classification: ResNET

Text Classification: BERT

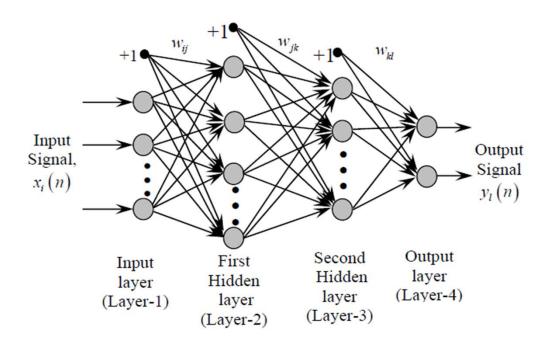
Image Segmentation: WNet

Image Translation: Pix2Pix

Object Detection: YoLo

Speech Generation: WaveNET

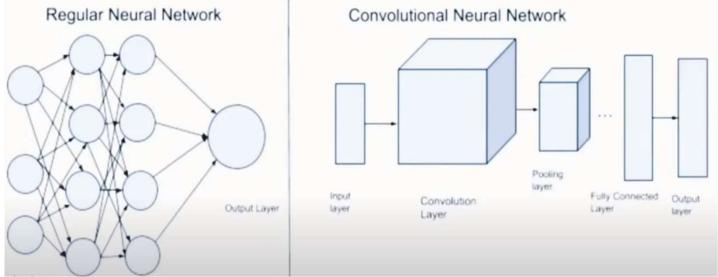
#### 1. Multilayer Perceptron



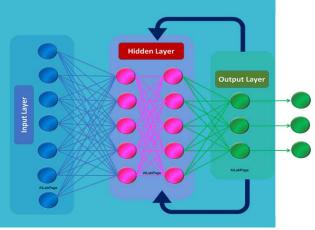
-> Works best in supervised problems

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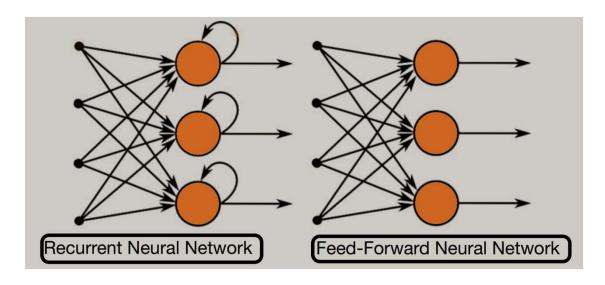
- 1. Multilayer Perceptron
- 2. Convolutional Neural Network



-> Works best with image and video processing



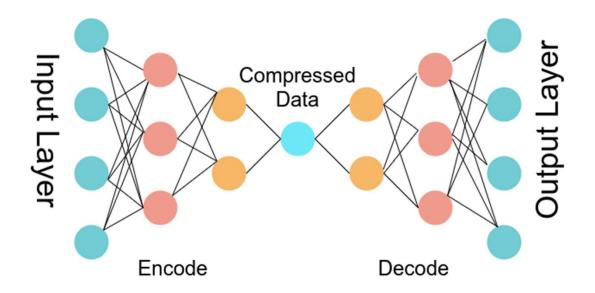
- 1. Multilayer Perceptron
- 2. Convolutional Neural Network
- 3. Recurrent Neural Network: Eg: LSTM



-> Works best with NLP applications

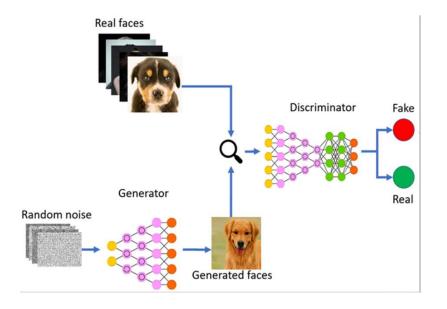


- 1. Multilayer Perceptron
- 2. Convolutional Neural Network
- 3. Recurrent Neural Network
- 4. Autoencoders

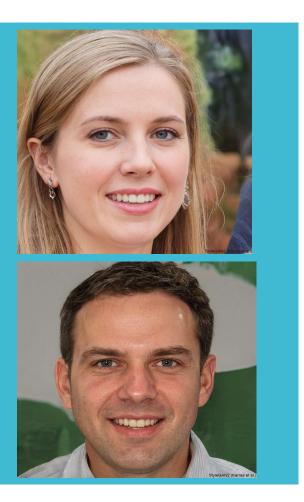


-> Works best with image compressions or regeneration of inputs

- 1. Multilayer Perceptron
- 2. Convolutional Neural Network
- 3. Recurrent Neural Network
- 4. Autoencoders
- 5. GAN (Generative Adversarial Network): Can generate hand written texts, images, stories, music, videos, etc.



-> Works best generating texts and images



thispersondoesnotexist.com (1024×1024)







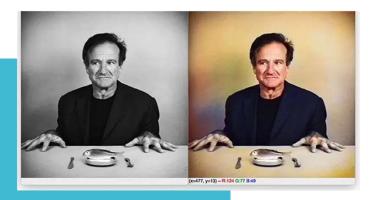




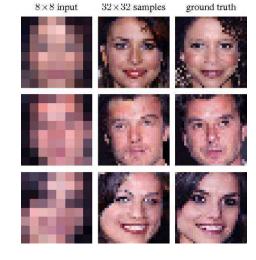






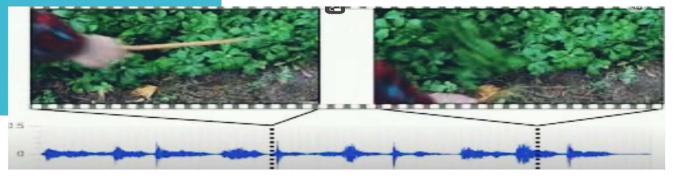






# Applications Eg:







# History of Deep Learning

- 1943: Warren McCulloch and Walter Pitts propose the first artificial neuron model, laying the foundation for neural network research. 1957: Frank Rosenblatt introduces the perceptron, a type of neural network capable of learning through a single layer.
- 1969-1970: Marvin Minsky and Seymour Papert publish "Perceptrons," highlighting limitations and challenges of single-layer perceptrons, dampening interest in neural networks.
- 1986: Geoffrey Hinton, David Rumelhart, and Ronald Williams publish the "Parallel Distributed Processing" book, reviving interest in neural networks and introducing backpropagation for training multi-layer perceptrons.
- 1990s: Neural networks gain popularity in various applications, but face limitations in training deep architectures due to the vanishing gradient problem.
- 2006: Geoffrey Hinton, along with Simon Osindero and Yee-Whye Teh, introduces the concept of unsupervised pretraining, a breakthrough for training deep neural networks.
- 2012: The ImageNet competition is won by a deep convolutional neural network (CNN) developed by Alex Krizhevsky, marking the beginning of the deep learning revolution in computer vision.
- 2014: Google acquires DeepMind, an AI company founded by Demis Hassabis, Shane Legg, and Mustafa Suleyman, known for its work in deep learning and reinforcement learning.
- 2015: Generative adversarial networks (GANs) are introduced by Ian Goodfellow and his colleagues, enabling the generation of realistic synthetic data.
- 2016: AlphaGo, developed by DeepMind, defeats world champion Go player Lee Sedol, showcasing the power of deep learning in mastering complex games.
- 2017: Transfer learning gains prominence, allowing pre-trained models to be adapted to new tasks with limited data.
- 2018: OpenAI introduces GPT (Generative Pre-trained Transformer), a language model based on transformers, capable of understanding and generating human-like text.
- 2019: BERT (Bidirectional Encoder Representations from Transformers) by Google demonstrates the effectiveness of pretrained language models in various natural language processing tasks.
- 2020s: Continued advancements in deep learning, including large language models, reinforcement learning breakthroughs, and applications in diverse fields such as healthcare, finance, and autonomous systems.