

CS-C1000 – Introduction to Artificial Intelligence

Deep Learning

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March 19, 2021

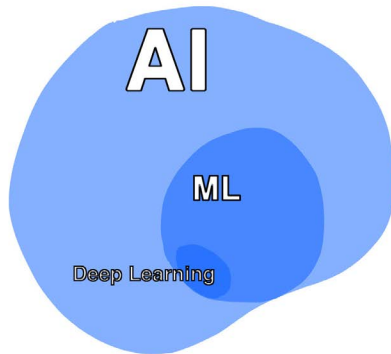
 @arnosolin

 arno.solin.fi

Outline

- ▶ What is Deep learning?
- ▶ Neurons and networks
- ▶ Types of deep learning models
- ▶ Applicability and examples

AI → ML → Deep Learning

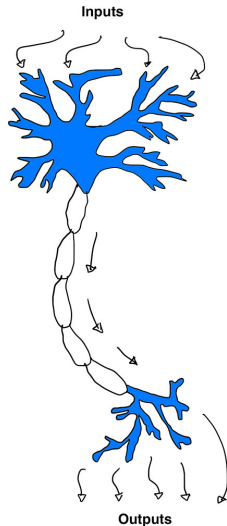


Deep learning

- ▶ Deep learning is a class of **machine learning** methods.
- ▶ Uses multiple layers of nonlinear processing units for **feature extraction** and **transformation**.
- ▶ Learns in **supervised** (e.g., classification) and/or **unsupervised** (e.g., pattern analysis) manners.
- ▶ Ideally, tries to capture **multiple levels of representations** that correspond to different levels of abstraction (forms a hierarchy of concepts).

Neural networks

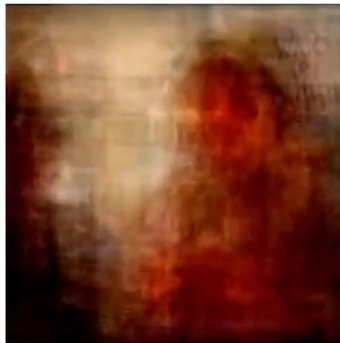
- ▶ Artificial neural networks are inspired by actual networks of neurons.
- ▶ The human brain has billions of **neurons**.
- ▶ Each neuron can have tens of thousands of connections depending on its type.



Presented clip



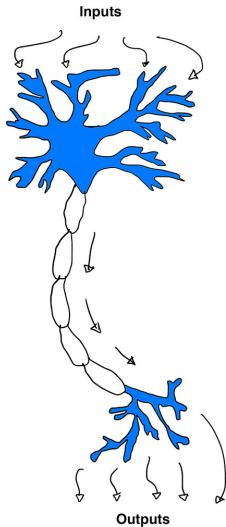
Clip reconstructed
from brain activity



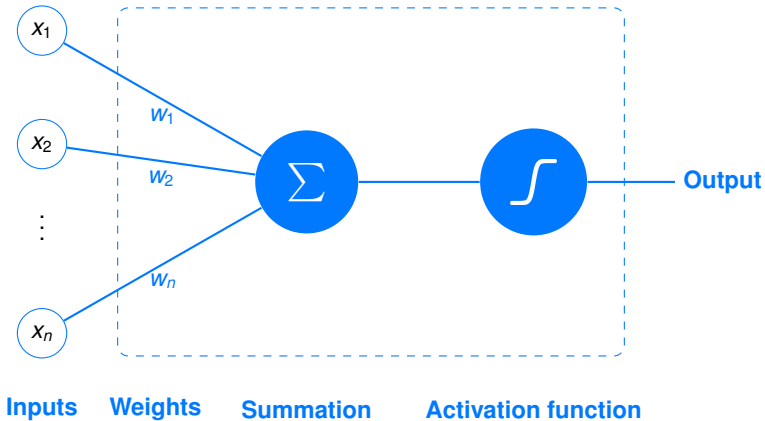
Reconstructing visual experiences from brain activity:
<https://www.youtube.com/watch?v=nsjDnYxJ0bo>

Neural networks

- ▶ Single neurons are **simple** (nothing 'intelligent' about them).
- ▶ When there are enough of them, they can power-up the human brain.
- ▶ The key is how they are connected and interact.

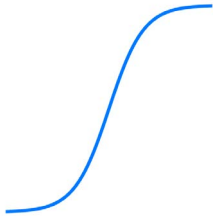


Artificial neuron



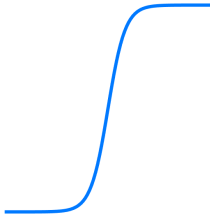
Activation functions

Logistic



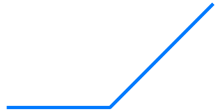
$$g(x) = \frac{1}{1 + e^{-x}}$$

Hyperbolic



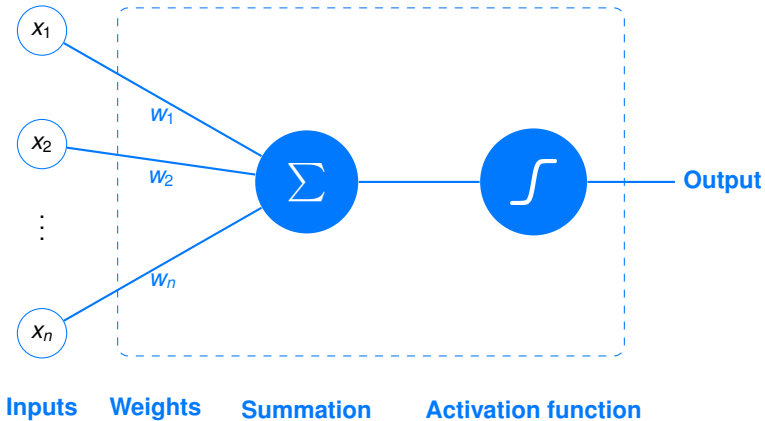
$$g(x) = \tanh(x)$$

Rectifier linear unit
(ReLU)



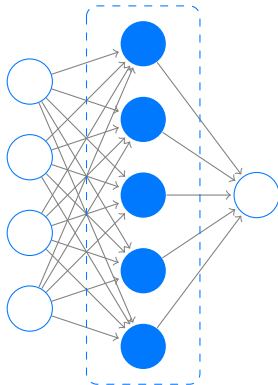
$$g(x) = \max(x, 0)$$

Artificial neuron



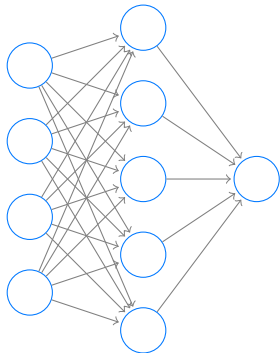
Artificial Neural Networks (ANNs)

- ▶ Collection of connected units or nodes called artificial neurons.
- ▶ Composing layers consisting many (simple) units.
- ▶ Typically you might need many units for capturing interesting phenomena.



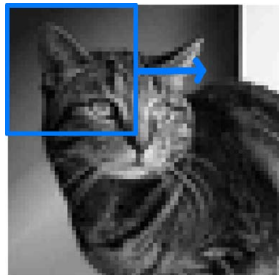
Deep neural networks

- ▶ In deep neural networks, there are **several layers**.
- ▶ Each successive layer uses the output from the previous layer as input.
- ▶ Makes the model very **flexible** (see previous lecture).

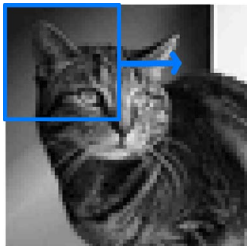


Convolutional Neural Networks (CNNs)

- ▶ Most commonly applied to analyzing visual imagery.
- ▶ Convolutions sweep the image in order to extract features.
- ▶ Good for capturing **translation invariant** characteristics (no matter where the feature in question appears in the image).



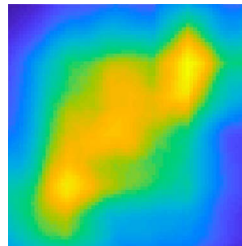
Convolution?



Input image

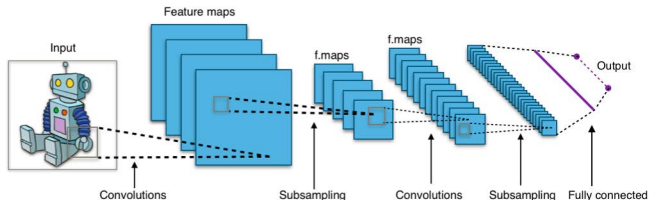


Convolution kernel
(what to extract)



Output
'cat-likeness'

How to train these monsters?



- ▶ Real-world models are often constructions of many layers.
- ▶ Recall that there are **weights** associated with the units.
- ▶ These weights are parameters that need to be ‘learned’ (through optimization).

Image: Typical cnn, Wikimedia Commons.

What problems can there be?

Great flexibility comes with some costs:

- ▶ Typically the number of weights/parameters is **huge**.
- ▶ Can lead to problems with **overfitting** (see previous lecture).
- ▶ Models are typically big (as in our previous exercise session!) and can require a lot of **computational resources** (thus GPUs are used).
- ▶ **Interpretability** typically low (black box).

Applicability

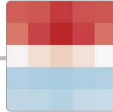
- ▶ Well-suited especially for applications where the **input is complicated** (e.g., images or audio).
- ▶ But on the other hand, the (sensible) latent space is quite **concentrated**.
- ▶ **For example:** Of all possible combinations of pixel colors, only a very small (and concentrated) subset actually make sense to us.



A random image.

Interpretability

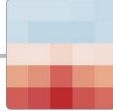
Windows (4b:237)
excite the car detector
at the top and inhibit
at the bottom.



Car Body (4b:491)
excites the car
detector, especially at
the bottom.



Wheels (4b:373) excite
the car detector at the
bottom and inhibit at
the top.



● positive (excitation)
● negative (inhibition)



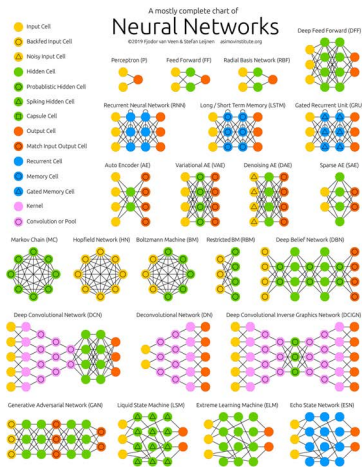
A car detector (4c:447)
is assembled from
earlier units.

Example of a car detector neuron

By OpenAI (<https://distill.pub/2020/circuits/zoom-in>)

A Zoo of Neural Network Models

- ▶ Deep learning / Neural networks do not form a single model family
- ▶ There are numerous different model types that go under this category, and there is no idea going through all of them here
- ▶ We only cover the basics of some of the most known ones

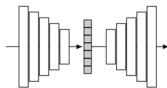


See more <https://www.asimovinstitute.org/neural-network-zoo/>

Autoencoders



Input



low-
dimensional
encoding



Output

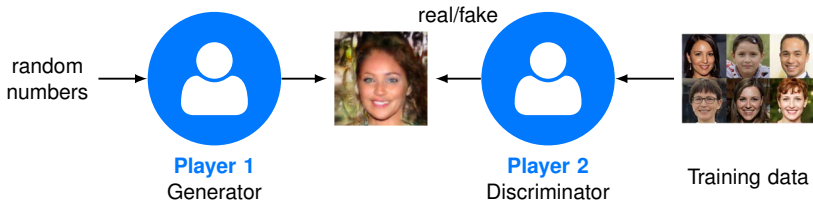
- ▶ A type of artificial neural network used to learn **efficient data codings** in an **unsupervised** manner.
- ▶ Learns a representation (encoding/**encoder**) for a set of data, typically for dimensionality reduction.
- ▶ Also a **decoder**/reconstructor is learnt, where it tries to reconstruct the original content.

Recurrent Neural Networks (RNNs)

- ▶ The connections between the nodes form a directed graph along a temporal sequence.
- ▶ Used for time-evolving phenomena (time-series).
- ▶ Applicable to tasks such as unsegmented, connected **handwriting recognition** or **speech recognition**.

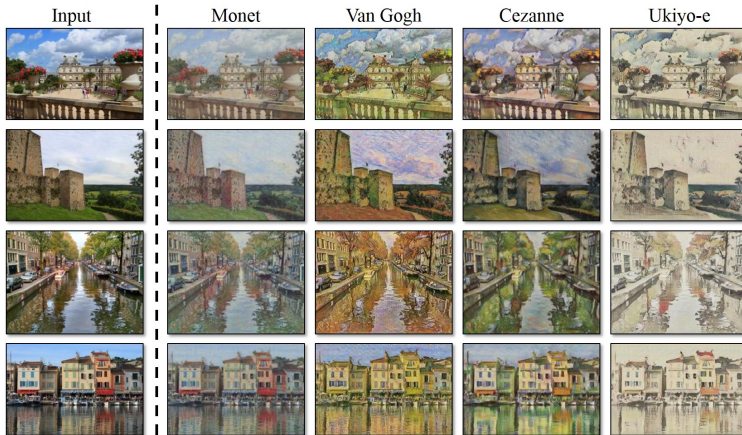
When
is
the
break ?

Generative Adversarial Networks (GANs)



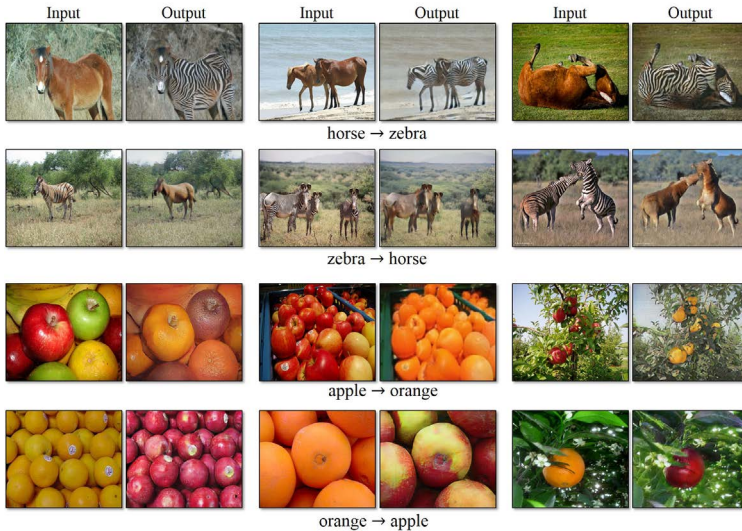
- ▶ GANs are like two-player games.
- ▶ The generative network generates candidates while the discriminative network evaluates them (real vs. fake).
- ▶ Both parts learn and try to get better.

GAN examples



Zhu *et al.* Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks. ICCV 2017.

GAN examples



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GAN examples

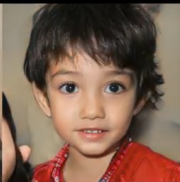


Zhu *et al.* Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks. ICCV 2017.

Source A: gender, age, hair length, glasses, pose



Source B:
everything
else



Result of combining A and B

A Style-Based Generator Architecture for Generative Adversarial Networks:

<https://www.youtube.com/watch?v=kSLJriaOumA>

Transformers

- ▶ A Transformer is a deep learning model that utilizes the mechanism of **attention**
- ▶ Mimics **cognitive attention**: enhances the important parts of the input data and fades out the rest
- ▶ Primarily in the field of **natural language processing** (NLP), in machine translation, text generation, etc.

Transformers Example

TEXT PROMPT

an illustration of a baby daikon radish in a tutu walking a dog

AI-GENERATED IMAGES



- ▶ DALL·E a 12-billion parameter version of GPT-3 trained to generate images from text descriptions, using a dataset of text–image pairs.

See <https://openai.com/blog/dall-e/>

Transformers Example

TEXT PROMPT

an armchair in the shape of an avocado [...]

AI-GENERATED IMAGES



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See <https://openai.com/blog/dall-e/>

Transformers Example

TEXT PROMPT

a store front that has the word 'openai' written on it [...]

AI-GENERATED IMAGES



- ▶ DALL-E a 12-billion parameter version of GPT-3 trained to generate images from text descriptions, using a dataset of text–image pairs.

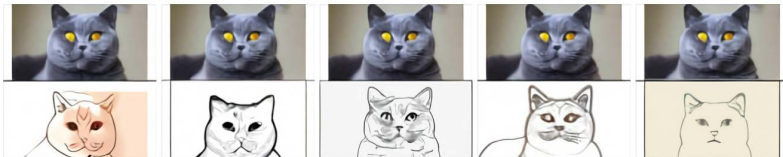
See <https://openai.com/blog/dall-e/>

Transformers Example

TEXT AND IMAGE PROMPT

the exact same cat on the top as a sketch on the bottom

AI-GENERATED IMAGES



- ▶ DALL·E a 12-billion parameter version of GPT-3 trained to generate images from text descriptions, using a dataset of text–image pairs.

See <https://openai.com/blog/dall-e/>

Recap

- ▶ Deep learning is a class of machine learning methods.
- ▶ Multiple layers of nonlinear processing units.
- ▶ Each successive layer uses the output from the previous layer as input.
- ▶ Capable of great flexibility in feature extraction and transformation.
- ▶ Produces state-of-the-art results image, audio, and text modelling.

What next?

- ▶ There is a lineup of [past visiting lectures in MyCourses](#). The intention is that you watch at least one of those.
- ▶ The second [Computer Exercise](#) is next Tuesday.
- ▶ The next [lecture](#) is next Friday. We will have [Janne Pulkkinen](#) from KELA visiting live during the first hour.

AI