From boolean algebra to convolutional neural networks

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Apr 2019



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CTU, 2019

What is intelligence

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- •



What is intelligence

- plan and reason,
- •
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- •
- ·
- •



What is intelligence

- plan and reason,
- solve complex problems,
- •
- •
- •



What is intelligence

- plan and reason,
- solve complex problems,
- make decisions under uncertainty,
- •
- •



What is intelligence

Intelligence is a provable ability to

- plan and reason,
- solve complex problems,
- make decisions under uncertainty,
- undestand the world,

•



What is intelligence

- plan and reason,
- solve complex problems,
- make decisions under uncertainty,
- undestand the world,
- and to learn and adapt.



Measurable change of behaviour acquired through

• instruction - supervised learning,

- Simplifications:
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 - •



Measurable change of behaviour acquired through

- instruction supervised learning,
- experience unsupervised learning,

•

Simplifications:



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Measurable change of behaviour acquired through

- instruction supervised learning,
- experience unsupervised learning,
- study semi-supervised learning.

Simplifications:

•

- •
- •



Measurable change of behaviour acquired through

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Simplifications:

• behaviour: obtain a set of observable features X and perform a decision Y.

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Measurable change of behaviour acquired through

- instruction supervised learning,
- experience unsupervised learning,
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Simplifications:

- behaviour: obtain a set of observable features X and perform a decision Y.
- given the knowledge of probability distributions p(F|S), p(S)
- and the reward function R(Y,S).





- Y: propose a retirement contract
- •
- •





- Y: propose a retirement contract
- X: features of the person
- •





- Y: propose a retirement contract
- X: features of the person
- x₁: age < 30
- •





- Y: propose a retirement contract
- X: features of the person
- x₁: age < 30
- x₂: smoker





Early retirement contract:

- Y: propose a retirement contract
- X: features of the person
- x₁: age < 30
- x₂: smoker

How many neurons are needed for such a decision?



Bio-inspired computational model:

McCulloh, Pritts: A Logical Calculus of Ideas Immanent in Nervous Activity. (1943)

- 1. The activity of the neuron is an all-or-none process.
- 2. A certain fixed number of synapses must be excited within the period of latent addition in order to excite a neuron at any time, and this number is independent of previous activity and position on the neuron.
- 3. The only significant delay within the nervous sytem is synaptic delay.
- 4. The activity of any inhibitory synapse absolutely prevents excitation of the neuron at that time.
- 5. The structure of the net does not change with time.



Towards learning of neurons

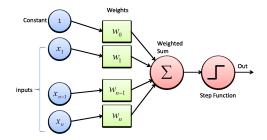
Ronald Hebb: The Organization of Behavior. (1949)

When an axon of cell A is near enough to excite a cell B and repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A's efficiency, as one of the cells firing B, is increased.



Perceptron

Rosenblatt: Principles of Neurodynamics. American Journal of Psychology (1969)





Perceptron training

```
while err do

err = false

foreach x in positive samples:

if (w^T x < 0) w = w + x

err = true

foreach x in negative samples:

if (w^T x > 0) w = w - x

err = true
```



Perceptron limitations

Papert, Minski: Perceptrons: An Introduction to Computational Geometry. (1967): Perceptron cannot do XOR, NXOR the research based on perceptron is doomed computational infeasibility to train large networks

- Little research done until 1980 (?).

- Solutions ?

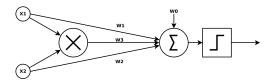


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Perceptron and XOR problem

Perceptron cannot do XOR, NXOR

- projection into a higher-dimensional space
- additional feature $x_3 = x_1 x_2$

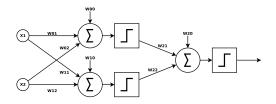




Perceptron and XOR problem

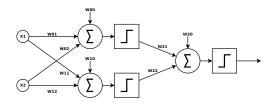
Perceptron cannot do XOR, NXOR

- multiple layers





How to train a multi-layered NN



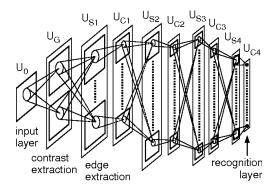
Backpropagation

- same idea as perceptron training, but by layers (last to first)



(Neo)cognitron

- Fukushima 1975: Cognitron: layered NN (unsupervised learning)
- Fukushima 1980: Neocognitron: layered NN (unsupervised learning) for text recognition

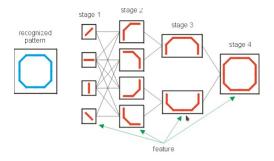




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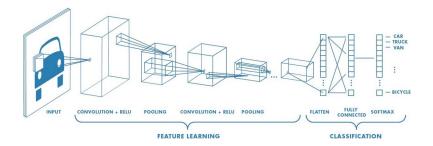


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Foundations of Artificial Intelligence

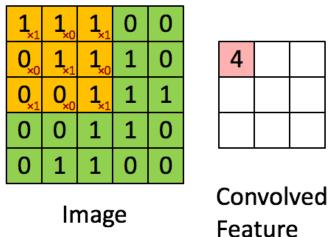
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Convolutional neural networks





Convolutional neural networks







The problem of good data

Data bias:

- google images for "person"
- why is that?

Dataset size:

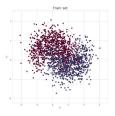
- number of annotated data needed to teach a NN vs number of NN weights?

Overfitting vs. accurracy:

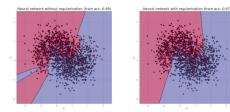
- How well will my model work on non-trained data?



What is overfitting and how to prevent it?









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Lecture on Bio-inspired Machine Learning

What does convetional programmer do?

- Programming complex systems is overwhelmingly boring.
- ML gets us rid of programming, it works out of the box!

So what does AI specialist do instead of programming?



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So what does AI specialist do instead of programming?

- Dataset annotation
- Parameter tuning
- Trial-and-error testing





Geoffrey Hinton materials:

http:

//www.cs.toronto.edu/~hinton/coursera_lectures.html

Introduction to deep learning: https://www.youtube.com/playlist?list= PLZHQObOWTQDNU6R1_67000Dx_ZCJB-3pi



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Next lecture?

Chronorobotics vs. AI in robotics

Vote here https://doodle.com/poll/bubn6dfa772t7h8x

