VS 265 - Neural Computation

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Class meets TTH 3:30-5 online

Challenge problem assignments (60% of grade)

Final Project (40% of grade)

Readings:
   Handouts
   Hertz, Krogh & Palmer, *Introduction to the Theory of Neural Computation*
   Dayan & Abbott, *Theoretical Neuroscience*
   MacKay, *Information Theory, Inference and Learning Algorithms*
   Sterling & Laughlin, *Principles of Neural Design.*

All reading materials and assignments on website at
   http://redwood.berkeley.edu/courses/vs265

Piazza discussion forum
Readings for this week
(available on the website)

Today:

- Sterling & Laughlin, Chapter 1

Next week:

- Sterling & Laughlin, Chapters 2-4
- Mead, *Analog VLSI and Neural Systems, Chapter 1: Introduction and Chapter 4: Neurons*
What have brain scans and single-unit recording taught us about the computations underlying perception and cognition?
Why hasn’t machine intelligence scaled with Moore’s law?
After **50 years** of concerted research efforts...

- there is little understanding of how neurons interact to process sensory information or to control actions.

- machines are still incapable of solving simple perceptual or motor control tasks.

We are missing something fundamental on both fronts: *we are ignorant of the underlying principles governing perception and action.*
How did we get here?
Among the most challenging scientific questions of our time are the corresponding analytic and synthetic problems: How does the brain function? Can we design a machine which will simulate a brain?

-- *Automata Studies*, 1956
THE SUMMER VISION PROJECT

Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".
Subgoal for July

Analysis of scenes consisting of non-overlapping objects from the following set:

balls

bricks with faces of the same or different colors or textures

cylinders.

Each face will be of uniform and distinct color and/or texture.

Background will be homogeneous.

Extensions for August

The first priority will be to handle objects of the same sort but with complex surfaces and backgrounds, e.g. cigarette pack with writing and bands of different color, or a cylindrical battery.

Then extend class of objects to objects like tools, cups, etc.
Machines will be capable, within twenty years, of doing any work that a man can do.
— Herbert Simon, 1965

Within a generation...the problem of creating ‘artificial intelligence’ will be substantially solved.
— Marvin Minsky, 1967

I confidently expect that within a matter of 10 or 15 years, something will emerge from the laboratory which is not too far from the robot of science fiction fame.
— Claude Shannon, 1961
The Lighthill debate (1973)

http://www.aiai.ed.ac.uk/events/lighthill1973/

Sir James Lighthill vs. PROF. RICHARD GREGORY
Experimental Psychologist

PROF. DONALD MICHIE
Edinburgh University

PROF. JOHN McCARTHY
Stanford University
The Lighthill debate (1973)

http://www.aiai.ed.ac.uk/events/lighthill1973/

Sir James Lighthill
Our first foray into Artificial Intelligence was a program that did a credible job of solving problems in college calculus. Armed with that success, we tackled high school algebra; we found, to our surprise, that it was much harder. Attempts at grade school arithmetic, involving the concept of numbers, etc., provide problems of current research interest. An exploration of the child’s world of blocks proved insurmountable, except under the most rigidly constrained circumstances. It finally dawned on us that the overwhelming majority of what we call intelligence is developed by the end of the first year of life.

--Minksy, 1977
Even ‘simple’ nervous systems can exhibit profound visual intelligence.

Jumping spider

Box jellyfish

Sand wasp
“The theory reported here clearly demonstrates the feasibility and fruitfulness of a quantitative statistical approach to the organization of cognitive systems. By the study of systems such as the perceptron, it is hoped that those fundamental laws of organization which are common to all information handling systems, machines and men included, may eventually be understood.”  -- Frank Rosenblatt

Perceptron model
(Rosenblatt, ca. 1960)

\[ u = w_0 + \sum_{i=1}^{n} w_i x_i \]

\[ y = \sigma(u) \]
Neocognitron
(Fukushima 1980)
An AI Breaks the Writing Barrier
A new system called GPT-3 is shocking experts with its ability to use and understand language as well as human beings do

Word has been making its way out from the technology community: The world changed this summer with the rollout of an artificial intelligence system known as GPT-3. Its ability to interact in English and generate coherent writing have been startling hardened experts, who speak of “GPT-3 shock.”

I copied and pasted the first paragraph of George Washington’s 1796 Farewell Address:

“The period for a new election of a citizen to administer the executive government of the United States being not far distant, and the time actually arrived when your thoughts must be employed in designating the person who is to be clothed with that important trust, it appears to me proper, especially as it may conduce to a more distinct expression of the public voice, that I should now apprise you of the resolution I have formed, to decline being considered among the number of those out of whom a choice is to be made.”

GPT-3 gave me its translation: “I am not going to run for president.” Take a bow, HAL 9000.
The approach of David Marr (ca. 1980)

<table>
<thead>
<tr>
<th>Computational theory</th>
<th>Representation and algorithm</th>
<th>Hardware implementation</th>
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<td>What is the goal of the computation, why is it appropriate, and what is the logic of the strategy by which it can be carried out?</td>
<td>How can this computational theory be implemented? In particular, what is the representation for the input and output, and what is the algorithm for the transformation?</td>
<td>How can the representation and algorithm be realized physically?</td>
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*Figure 1–4.* The three levels at which any machine carrying out an information-processing task must be understood.
Nervous systems are difficult to observe and manipulate
1 mm² of cortex contains 100,000 neurons
Anatomy of a synapse
Are there principles?

“God is a hacker”
– Francis Crick

“Individual nerve cells were formerly thought to be unreliable... This was quite wrong, and we now realise their apparently erratic behavior was caused by our ignorance, not the neuron’s incompetence.”
– H.B. Barlow (1972)
Otto Lilienthal experiments with flight (1890’s)

Der Vogelflug als Grundlage der Fliegekunst (1889)
Mr. Horace Lytle, President, The J. Horace Lytle Company, Dayton, Ohio.

Dear Mr. Lytle:-

Your letter of November 26th was duly received, but having become buried among other papers, it has just come to my attention again.

I can not think of any part bird flight had in the development of human flight excepting as an inspiration. Although we intently watched birds fly in a hope of learning something from them I can not think of anything that was first learned in that way. After we had thought out certain principles, we then watched the bird to see whether it used the same principles. In a few cases we did detect the same thing in the bird's flight.

Learning the secret of flight from a bird was a good deal like learning the secret of magic from a magician. After you once know the trick and know what to look for you see things that you did not notice when you did not know exactly what to look for.

Sincerely yours,

Orville Wright
Principles of optics govern the design of eyes.
squares of perpendicular distance, and least absolute deviations, the third of which is more robust against outliers (13). The standard deviations for the slope and intercept were estimated directly for the first method and by bootstrap for the last two methods (14). Bootstrap may help detect outliers in the data because, when they are left out from a same-size resample, the correlation coefficient often increases, which could be exploited to improve estimation. Systematic bias caused by outliers was not detected in Fig. 2.

3. Theory of Scaling

Our analysis rests on two assumptions. First, we assume that each small piece of cortex of unit area, regardless of its thickness and the overall brain size, sends and receives about the same total cross-sectional area of long-distance connection fibers to and from other cortical regions. Second, we assume that the global geometry of the cortex minimizes the average length of the long-distance fibers. The second assumption follows from Ramon y Cajal's principle for conservation of space, conduction time, and cellular materials (Chap. V in ref. 15). This principle has been explored more recently as the principle of minimal axon length (16–18). Consistent with previous observations on the basic uniformity of the cortex (19–21), the first assumption is supported loosely by the evidence that the total number of neurons beneath a unit cortical surface area is about $10^5$ across different cortical regions for several species, from mouse to human (22) (after shrinkage correction). But there are exceptions, including the higher density in striate cortex of primates (22, 23), the lower density in dolphin cortex (24), and the variability observed in cat cortex (25). The number of axons leaving or entering the gray–white boundary per unit cortical area should be comparable.

Fig. 2. Cortical white and gray matter volumes of various species ($n$) are related by a power law that spans five to six orders of magnitude. Most data points are based on measurement of a single adult animal. The line is the least squares fit, with a slope around 1.23 (mean $\pm$ 0.01 (mean $\pm$ SD). The average and median deviations of the white matter volumes from the regression line are, respectively, 18% and 13% on a linear scale.

Sources of data: If the same species appeared in more than one source below, the one mentioned earlier was used. All 38 species in table 2 in ref. 3 were taken, including 23 primates, 2 tree shrews, and 13 insectivores. Another 11 species were taken from table 2 in ref. 8, including 3 primates, 2 carnivores, 4 ungulates, and 2 rodents. Five additional species came from table 1 in ref. 11, including 1 elephant and 4 cetaceans. The data point for the mouse ($G = 112 \text{ mm}^3$ and $W = 13 \text{ mm}^3$) was based on ref. 30, and that for the rat ($G = 425 \text{ mm}^3$ and $W = 59 \text{ mm}^3$) was measured from the serial sections in a stereotaxic atlas (42). The estimates for the fisherman bat (Noctilio leporinus, $G = 329 \text{ mm}^3$ and $W = 43 \text{ mm}^3$) and the flying fox (Pteropus lylei, $G = 2,083 \text{ mm}^3$ and $W = 341 \text{ mm}^3$) were based on refs. 43 and 44, with the ratios of white and gray matters estimated roughly from the section photographs in the papers. The sea lion data (Zalophus californianus, $G = 113,200 \text{ mm}^3$ and $W = 56,100 \text{ mm}^3$) were measured from the serial sections at the website given in the legend to Fig. 1, with shrinkage correction.
**Principles**

Compute with chemistry

Compute directly with analog primitives

Combine analog and pulsatile processing

Sparsify

Send only what is needed

Send at the lowest acceptable rate

Minimize wire

Make neural components irreducibly small

Complicate

Adapt, match, learn, and forget
Computational principles

- Efficient coding
- Unsupervised learning
- Bayesian inference
- Dynamical systems
- Prediction
- High-dimensional vector arithmetic
- Computing with waves
THAT'S THE WHOLE PROBLEM WITH SCIENCE. YOU'VE GOT A BUNCH OF EMPIRICISTS TRYING TO DESCRIBE THINGS OF UNIMAGINABLE WONDER.