Computational Neurodynamics

Topic 2 Neurons

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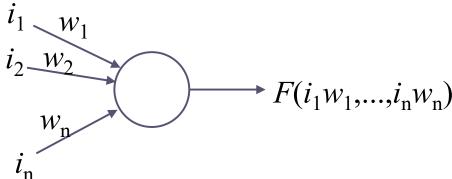
(Slides: Pedro Mediano & Murray Shanahan)

Overview

- Artificial and real neurons
- Axons and dendrites
- Neuron behaviour
- The Hodgkin-Huxley model

Artificial Neurons 1

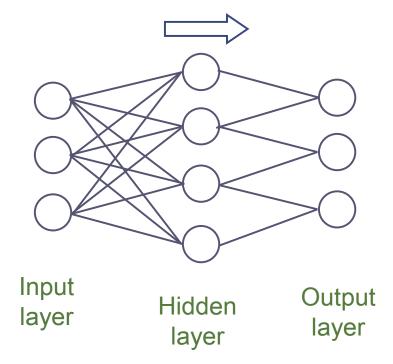
- The kind of artificial neuron traditionally used for neural network applications in computer science is very useful, but has little biological plausibility
- The type of artificial neuron on the right is very common. It computes a weighted sum (F) of its inputs i₁ to i_n



- Each connection has an associated weight (w_i) , and a neuron's output is a function of all its weighted inputs
- Many useful applications have been built out of such simple artificial neurons

Artificial Neurons 2

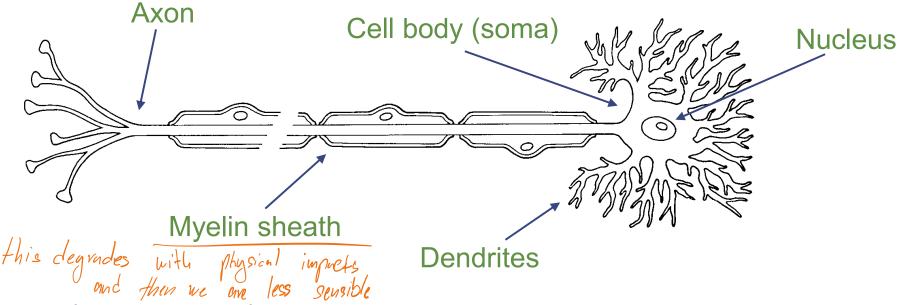
- They are often organised into feedforward networks, comprising an input layer, a hidden layer (or often many hidden layers), and an output layer
- A learning algorithm such as back propagation is applied to train the network



- Some applications use recurrent neural networks (RNNs) with a loop of feedback from the output back the input layer
- But this course is NOT about neural network applications

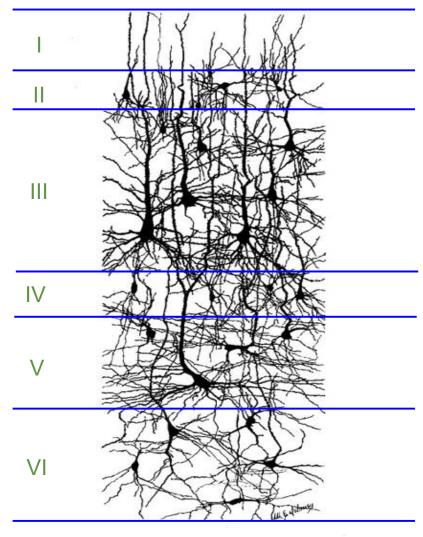
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A real neuron comprises a cell body, a tree of dendrites and an axon. The dendrites carry incoming electrical signals, and the axon delivers the neuron's electrical output



 Long (white matter) axons are covered in a myelin sheath, which increases the speed of electrical conduction

Dendrites and Axons



From Elston (2003), Cereb. Cortex 13:1124-1138

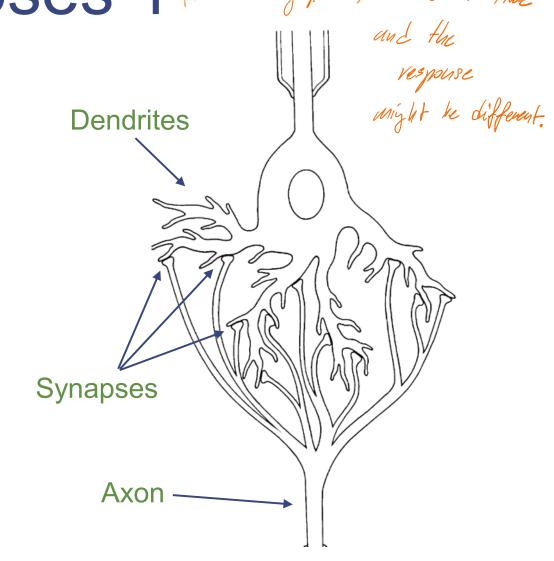
- Human cerebral cortex contains 20 billion neurons, with a variety of morphological (shape) and signalling properties, organised into six layers
- This image is of a "vertical" slice through cortex. Only a few of the densely connected neurons that would be found in an area of this size are shown
- Each neuron's dendritic tree ramifies widely
- A single neuron can project to as many as 10,000 other neurons

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- biological neuron one send signal (with weather power) over larger time, some as

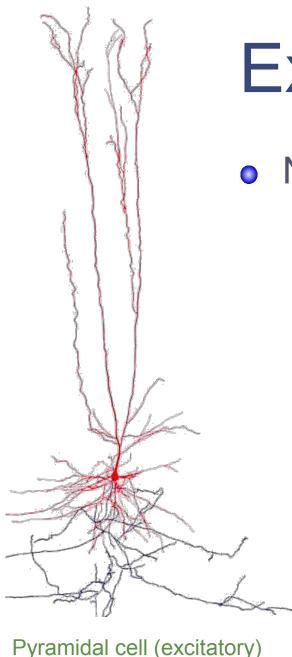
Synapses 1 (with strong power) are short time

- The junctions where axons meet dendrites and signals are transmitted from the former to the latter are called synapses
- Synapses are not direct electrical connections.
 Rather, there is a tiny gap between the axon and the dendrite (the synaptic cleft) in which a complex electrochemical process takes place that allows a signal to be transmitted



Synapses 2

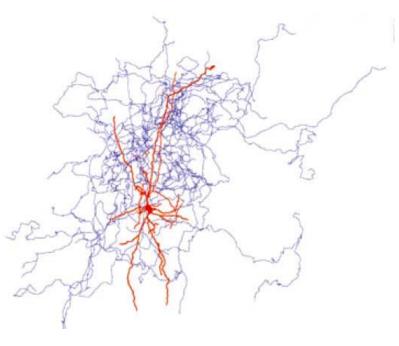
- This process of synaptic transmission is fundamental to the operation of the brain
 - There is a whole soup of electrically significant chemicals in the synaptic cleft. These are called *neurotransmitters*, and include serotonin, dopamine, and adrenaline
 - Many antidepressant drugs work by modifying serotonin uptake
 - Adrenaline influences behaviour in "fight or flight" situations
 - Dopamine is involved in the brain's reward system
- However, in this course we will treat synapses as simple weighted connections, because our focus is dynamics on a larger scale. But it's important to recognise the limitations of this simplification



Excitation and Inhibition

- Neurons fall into two major sub-classes
 - Excitatory neurons increase the activity of the neurons they are connected to
 - Inhibitory neurons decrease the activity of neurons they are connected to
 - Neurons are either excitatory or inhibitory, but not both



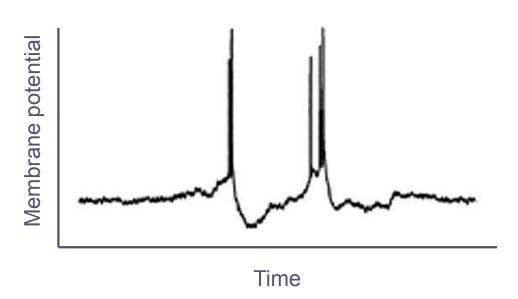


Inhibitory interneuron

Basic Neuron Behaviour

a Sit

- Neurons receive and transmit electrical pulses, or spikes
- Incoming spikes travel along a neuron's dendrites, and cause charge to build up in the body of the neuron. When this charge reaches a threshold, the neuron fires, and sends a spike along its axon

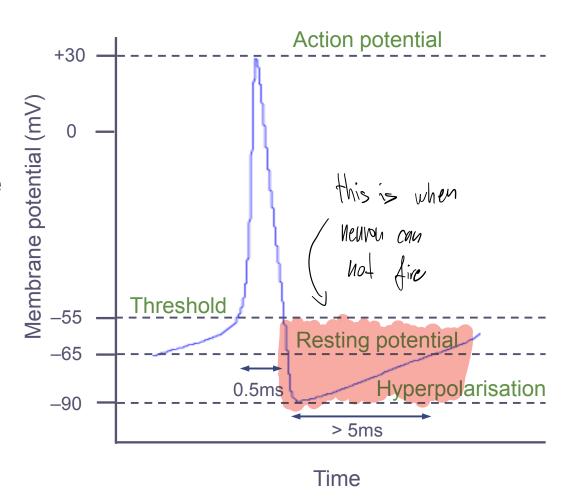


From Hirsch, et al. (2002). J. Physiol 540:335-350

- This plot shows the spiking behaviour of a single neuron recorded in the visual cortex of a cat
- Axons meet dendrites at synapses. The transmission of a signal across a synapse involves a complex electrochemical process which we won't go into

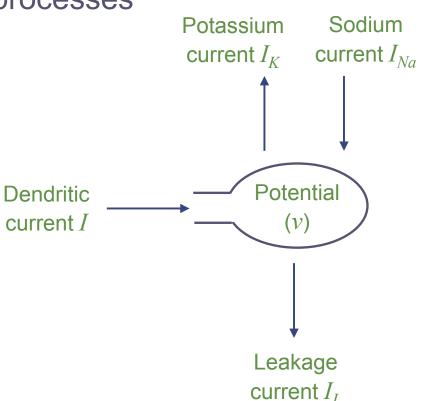
Detailed Neuron Behaviour

- Current flowing into a neuron along its dendrites, causes its membrane potential to increase
- Eventually the membrane potential reaches a threshold and the neuron rapidly depolarises, emitting a spike along its axon
- It then *repolarises*, typically undershooting its resting potential
- This undershoot gives rise to a refractory period, during which the neuron cannot fire again
- When unperturbed, the neuron tends towards a stable resting potential, normally around -65 mV



Towards a Computer Model

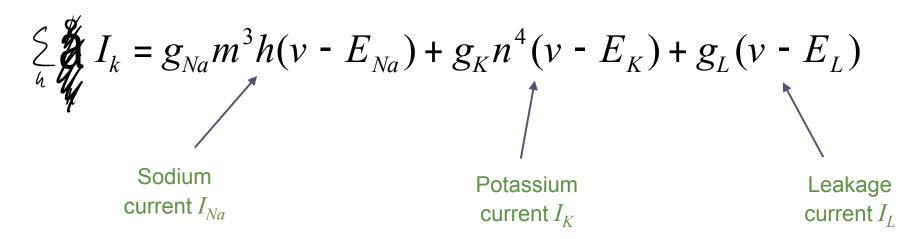
- To see how this behaviour can modeled mathematically, and then simulated on a computer, we need to understand a bit more about the underlying physical processes
- The neuron's potential (v) exhibits its characteristic spiky profile thanks to the interplay of three currents that flow across the neuron's membrane, in addition to the incoming current from its dendrites (I)
 - The potassium current I_K
 - The sodium current I_{Na}
 - The leakage current I_L



More formally, according to the Hodgkin-Huxley model, we have

$$C\frac{dv}{dt} = - \sum_{k=1}^{\infty} I_k + I$$

where C is the capacitance of the neuron (set to 1), and



• The gs and Es are parameters of the model, determined empirically. The following values are the ones reported by Hodgkin and Huxley in their 1952 paper

$g_{Na} = 120$	$E_{Na} = 115$
$g_{K} = 36$	$E_K = -12$
$g_L = 0.3$	$E_L = 10.6$

• Three further differential equations govern the evolution of m, n, and h



 The potassium and sodium currents behave as if gates open and close, allowing strong but brief flows of current, first in (sodium) then out (potassium)

$$\frac{dm}{dt} = a_m(v)(1 - m) - b_m(v)m$$

$$\frac{dn}{dt} = a_n(v)(1 - n) - b_n(v)n \qquad \text{where}$$

$$\frac{dh}{dt} = a_h(v)(1 - h) - b_h(v)h$$

$$a_m = (2.5 - 0.1v)/(e^{(2.5 - 0.1v)} - 1)$$

 $b_m = 4e^{-v/18}$

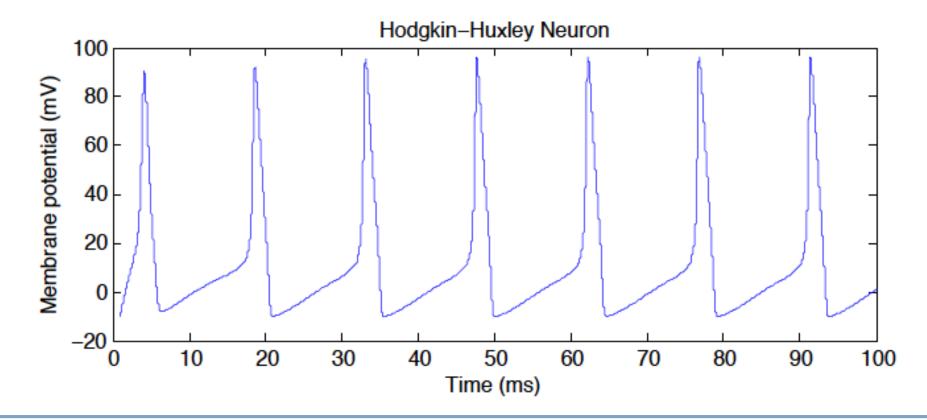
$$a_n = (0.1 - 0.01v)/(e^{(1-0.1v)} - 1)$$

$$b_n = 0.125e^{-v/80}$$

$$a_h = 0.07e^{-v/20}$$

$$b_h = 1/(e^{(3-0.1v)} + 1)$$

 The resulting model accurately reproduces the signalling properties of neurons, and is still the standard mathematical model used today



Related Reading

Trappenberg, T.P. (2010). Fundamentals of Computational Neuroscience. Oxford University Press.