CISC 3250 example exam questions

This gives a sense of the type of questions that can be included. Further math topics covered in the homeworks (but not in this practice set) may appear on the exam as well.

We use the voltage model: $\tau \frac{dv}{dt} = -(v(t) - E_L) + RI(t)$ With voltage reseting to E_L once it reaches the threshold v_{thresh}, as specified by $v(t^f + \delta) = E_L$

Let us say E_L =-60mV, τ =1, and v_{thresh} =-35mV.

Furthermore, we define the axon voltage at time t=0ms to be: v(0)=-40mV

Presuming a constant input RI(t)=10mV, how does the axon voltage change from time t=0ms to t=1,000 ms? For example, does the voltage spike, stay the same, plateau at another voltage value? Explain why, using the above equation(s) for the voltage model.

Answer: $\frac{dv}{dt} = -(v(t) - E_L) + RI(t) = -(-40 - (-60)) + 10 = -20 + 10 = -10$ at t=0ms v(10ms)=v(0ms)+-10x0.01=-40-.1=-40.1Initially, the -(v(t)-E_L) term is larger than the RI term, so the voltage drop from -40 mV. When the voltage reaches -50 mV, it plateaus as the -(v(t)-E_L) term -(-50--60)=-10 cancels at the RI(t)=10 term.

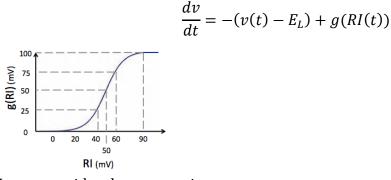
Presuming the input rises to RI(t)=300 starting at 1,001 ms, what is the axon voltage from time t=1,001ms to t=2,000ms? Does the voltage spike, stay the same, plateau at another voltage value? Explain why, using the above equation(s) for the voltage model.

Presuming a constant positive input RI(t) over a long period of time, what is the minimum input value to produce spiking?

Let us assume τ =1000 (and all other parameters are the same as initially described). How does the change in τ change the minimum input value required to produce spiking, if at all?

Answer: The change in τ has no effect on the minimum RI to cause spiking. Explain a biological mechanism by which spiking information is communicated from the axon of neuron A to the dendrite of neuron B.

We can replace the RI(t) term from the equation above with g(RI(t)), where g() is the sigmoid function we learned in class, now scaled to have a maximum output of 100mV, rather than just 1:



Let us consider three scenarios: 1: RI is a constant 60 mV 2: RI is a constant 120 mV 3: RI is a constant 240 mV

What is the change in spiking frequency from scenario 1 to scenario 2 (increase, decrease, stay roughly the same)?

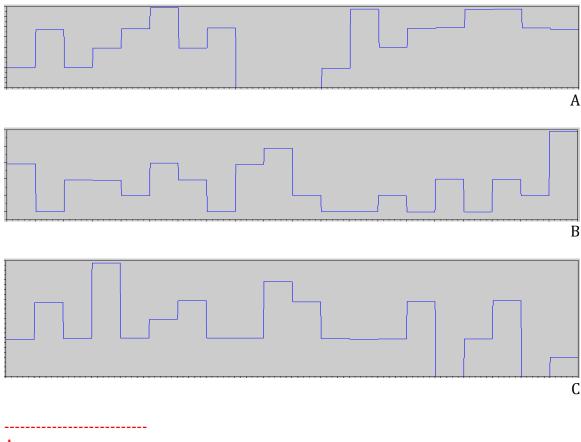
What is the change in spiking frequency from scenario 2 to scenario 3?

Answer:

Spiking frequency stays roughly the same between scenario 2 and scenario 3 (both RI values produce roughly 100mV from the sigmoid function.

Which rate plot describes the spiking pattern shown here? *Rate plots have time on the x axis and rate over fixed time window on the y axis.*

	1 11 1	1 1 1 1 11 11	11 11 1	1 11 100	111		1.11
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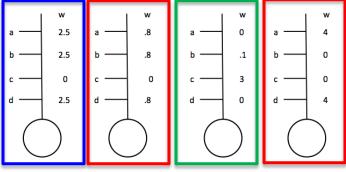


Answer:

Plot A (note: if I give a question like this on the exam, it will be a bit more challenging --- here you can just see the pause in spiking corresponds to the dip in spike rate).

The following neuron computes the weighted sum of its inputs as $h = \sum_j w_j r_j$. It then output a 1 if h>2 and otherwise outputs a 0. Each input provides a rate r_j of either 0 or 1 (no values in between).

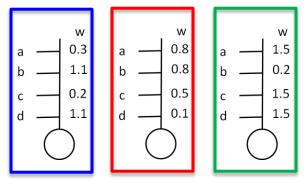
For each set of weights, say if the neuron performs generalization/pattern completion, and if so, name the inputs that are tied together in the pattern.



The second neuron (w=.8, .8, 0, .8), there is not generalization --- it requires multiple simultaneous inputs to fire.

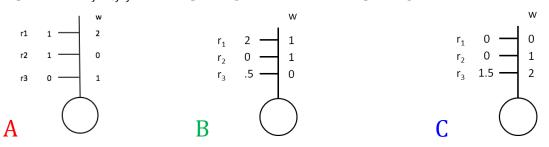
The right-most neuron (w=4, 0, 0, 4), there is generalization --- between inputs a and d.

For each set of weights, say if the neuron performs template recognition, and if so, name the inputs that are tied together in the pattern.



Performs template recognition, using inputs a, b, and c.

Presume the following neuron's output is computed by taking the weighted sum of inputs $h = \sum_{i} w_{ii} r_{i}$, and outputting 2 if h>1.5 and outputting 0 otherwise.



NEURON A

Provide new weights using learning rule $\Delta w_{ij} = \epsilon r_i r_j$ and $\epsilon(w) = \begin{cases} 1 & \text{if } w \ge 0 \\ -1 & \text{if } w < 0 \end{cases}$ for all weight values

Answer: rout=gthresh(1x2+1x0+0x1)=gthresh(2)=2

 $\Delta w_{out1} = 1 \times 2 \times 1 = 2$; after learning: $w_{out1}^{new} = w_{out1} + \Delta w_{out1} = 2 + 2 = 4$

 $\Delta w_{out2} = 1 \times 2 \times 1 = 2$; after learning: $w_{out2}^{new} = w_{out2} + \Delta w_{out2} = 0 + 2 = 2$

wout3 does not change, since r₃=0, so wout3^{new}=wout3=1

Provide new weights using learning rule above, followed by normalization of weights

Answer:

Taking previous answer, we now sum the new weights $\sum_{j} w_{outj} = 4 + 2 + 1 = 7$ and divide all the weights by this sum

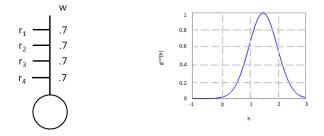
 $W_{out1}^{norm} = 4/7 \approx .57$ $W_{out2}^{norm} = 2/7 = .29$ $W_{out3}^{norm} = 1/7 \approx .14$

Provide new weights using Willshaw learning rule.

Answer: $\Delta w_j = r_{out}(r_j - w_j)$ $\Delta w_{out1} = 2 \times (1 - 2) = -2; \text{ after learning: } w_{out1}^{new} = w_{out1} + \Delta w_{out1} = 2 - 2 = 0$ $\Delta w_{out2} = 2 \times (1 - 0) = 2; \text{ after learning: } w_{out2}^{new} = w_{out2} + \Delta w_{out2} = 0 + 2 = 2$ $\Delta w_{out3} = 2 \times (0 - 1) = -2; \text{ after learning: } w_{out3}^{new} = w_{out3} + \Delta w_{out3} = 0 - 2 = -2$

can use a calculator, can estimate to 1 decimal place

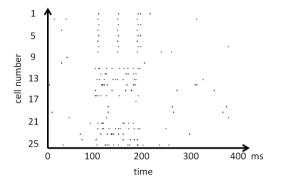
The neuron 5 takes four inputs r_1 , r_2 , r_3 , and r_4 . Presume each of the four inputs is either 0 or 1. Further presume neuron 5 takes the weighted sum of its inputs and then has an output determined by $g^{rad}(h)$ below.



Propose two combinations of inputs that will produce an output great than 0.8.

Propose two combinations of inputs that will produce an output less than 0.1.

To the right is a spikegram recording of a group of neurons in an animal's olfactory bulb. (This is actually fake data for the sake of testing concepts from class.) At some point, in time an odor is presented to the animal and at some later point the odor is removed.



Some of the recorded neurons respond through change is spiking rate. Which are these neurons?

Neurons 11-16 and neurons 21-25. (You can eye-ball your answers here for the purposes of this question based on the y axis labels.)

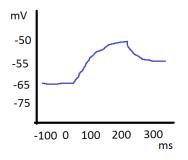
For these neurons, estimate the spike rate (per second) while the smell is presented. Roughly 80 spikes for 11 neurons -> roughly 7.5 spikes per neuron Altered spiking activity lasts 100ms (or .1 s), so **75 spikes/second** (again, you can eye-ball. I would give credit anywhere between 85 and 65 spikes/second)

Some of the recorded neurons respond through change in temporal coding. Which are these neurons?

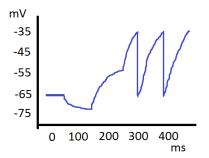
At what time is the smell presented? At what time is it taken away?

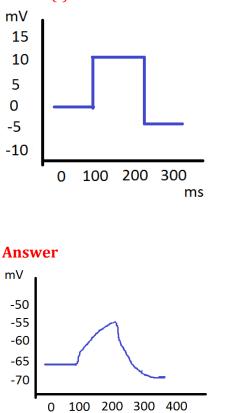
Assume E_L=-65mV, v_{thresh}=-35mV, v_{reset}=-65mV, and τ =.05

Draw RI(t) that will produce the resulting v(t)









Draw v(t) that results from the following RI(t)

Matlab commands:

Matrix1 is defined with the following content:

ms

10 1 12 6	20	20	30	0	
1	1	0	0	1 5	
12	-5	-8	9	5	
6	0	0	2	4	

What is the value of variable x at the end of each chunk (or line) of code: Chunk 1: x=Matrix1(2:4,5);

Chunk 2: val=find(Matrix1(2,:)==0); x=Matrix1(1,val);

[20, 30]

Chunk 3: x=min(Matrix1(:,2)); What is the value of variable v at the end of each chunk of code: Chunk 1: v(1)=3;

for i=1:5
 v(i+1) = v(i)*2;
end

Chunk 2: v=[4,5,8,10,12]; for k=3:5 v(k)=v(k)-2; end;

[4,5,6,8,10]

Chunk 3: v=4:2:10; v(4)=15;

Convert the following to an equivalent for loop:

```
v(1) = -70;

v(2) = v(1) + (-1*(v(1) - 70) + 50) * 0.1;

v(3) = v(2) + (-1*(v(2) - 70) + 50) * 0.1;

v(4) = v(3) + (-1*(v(3) - 70) + 50) * 0.1;

v(5) = v(4) + (-1*(v(4) - 70) + 50) * 0.1;

v(6) = v(5) + (-1*(v(5) - 70) + 50) * 0.1;

v(7) = v(6) + (-1*(v(6) - 70) + 50) * 0.1;

v(1) = -70;

for i=2:7

v(i) = v(i-1) + (-1*(v(i-1) - 70) + 50) * 0.1;

end
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