

# Exercise 1

a)  $x$  : susceptible people (fraction)

$y$  : infected people

People not infected and not susceptible is...? Recovered!

→  $(1 - x - y)$  is the fraction of recovered people

→  $\alpha$  is the rate at which recovered become susceptible again (to re-infection)

b)

$$\rightarrow \frac{dx}{dt} = 0 \quad y = \frac{-\alpha + \alpha x}{-\alpha - x\beta}$$

$$\alpha - \alpha x - \alpha y - xy\beta = 0$$

$$-\alpha y - xy\beta = -\alpha + \alpha x$$

$$y(-\alpha - x\beta) = -\alpha + \alpha x$$

$$\rightarrow \frac{dy}{dt} = 0 \quad y = 0 \quad \text{or} \quad x = \frac{1}{R_0}$$

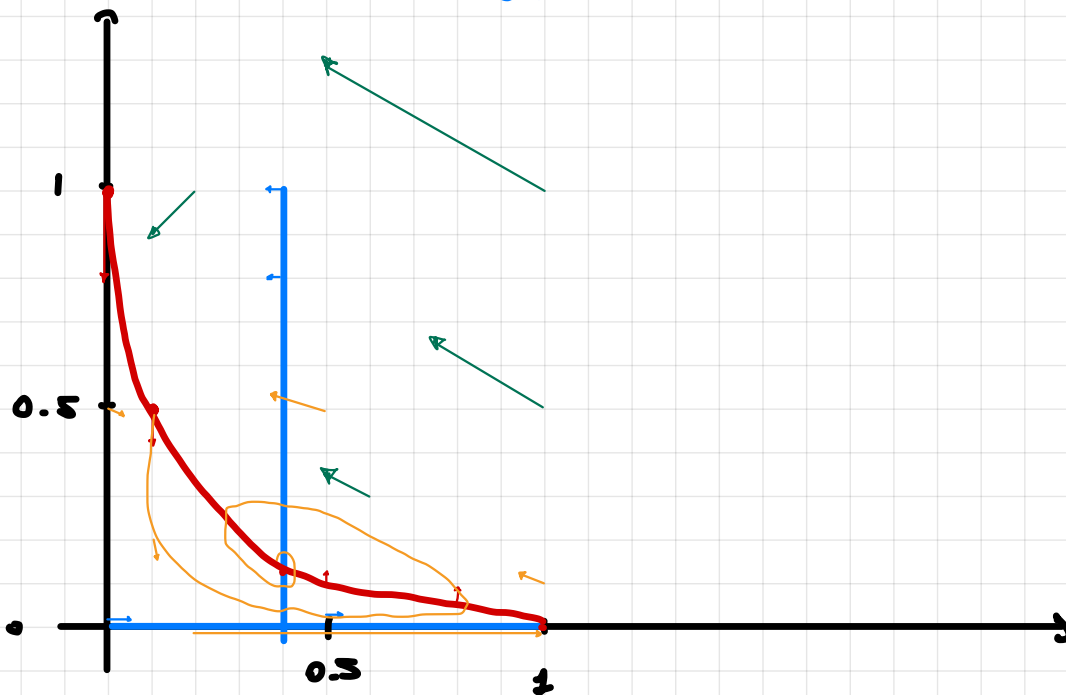
$$\left| \left( x - \frac{1}{R_0} \right) y \beta = 0 \right.$$

$$y = \frac{-\alpha + \alpha x}{-\alpha - x \beta}$$

c)

$$\frac{dx}{dt} = 0$$

$$\frac{dy}{dt} = 0$$



$$d) \quad (0, 0.5) \rightarrow (0.05, -0.2)$$

$$(0.1, 0.2) \rightarrow (0.05, -0.06)$$

$$(0.5, 0.5) \rightarrow (-0.25, 0.05)$$

$$(0.9, 0.1) \rightarrow (-0.09, 0.05)$$

8) <sup>↖ where's e?!</sup> For  $(x, y) = (0.5, 0.5)$  we have

$$\left( \frac{dx}{dt}, \frac{dy}{dt} \right) = (-0.25, 0.05)$$

We can use  $\Delta t = 0.5$  which gives

$$(\Delta x, \Delta y) = (-0.125, 0.025)$$

See plot!

9)  $\frac{dy}{dt}$  at  $(0, 1) = -0.4$   $\Delta y = -0.2$

$\frac{dy}{dt}$  at  $(0.1, 0.5) = -0.15$   $\Delta y = -0.075$

$\frac{dy}{dt}$  at  $(0.5, 0.1) = 0.01$   $\Delta y = 0.005$

$\frac{dy}{dt}$  at  $(0.8, 0.05) = 0.02$   $\Delta y = 0.01$

$$\frac{dx}{dt} \text{ at } (0, 0) = 0.1 \quad \Delta x = 0.05$$

$$\frac{dx}{dt} \text{ at } (0, 0.5) = 0.05 \quad \Delta x = 0.025$$

$$\frac{dx}{dt} \text{ at } (0.4, 1) = -0.44 \quad \Delta x = -0.22$$

$$\frac{dx}{dt} \text{ at } (0.4, 0.8) = -0.38 \quad \Delta x = -0.19$$

(See plot!)

$$h) \quad (x, y) \quad \left( \frac{dx}{dt}, \frac{dy}{dt} \right) \quad (\Delta x, \Delta y)$$

(See plot!)

$(1, 1)$	$(-1.1, 0.6)$	$(-0.55, 0.3)$
$(1, 0.5)$	$(-0.55, 0.3)$	$(-0.27, 0.15)$
$(0.7, 0.3)$	$(-0.21, 0.05)$	$(-0.1, 0.045)$
$(0.2, 1)$	$(-0.22, -0.2)$	$(-0.1, -0.1)$

i) See plot!

j) See plot!

k)

→ Recalling that  $dx/dt = 0$  at

$$y = \frac{-\alpha + \alpha x}{-\alpha - x\beta}$$

When  $\alpha \rightarrow 0$  we have

$$y \sim \frac{\alpha x}{-x\beta} \approx -\frac{\alpha}{\beta}$$

The  $x$  nullcline flattens slightly above the  $y$  nullcline.

→ Flow arrows get more negative  $x$  components. People recover as quickly?

UNSURE!

→ At the limit  $\alpha \rightarrow 0$   $x$  nullcline flattens on  $y$  nullcline. Fixed points everywhere!

## Exercise 2

2)

$$\rightarrow h_i(t) = \sum_j w_{ij} s_j(t) :$$

$$= \sum_j \frac{1}{(1-d^2)N} \sum_{\mu} (\xi_i^{\mu}) (\xi_j^{\mu} - d) s_j(t)$$

$$= \sum_{\mu} \frac{1}{(1-d^2)N} \xi_i^{\mu} \sum_j (\xi_j^{\mu} - d) s_j(t)$$

$$= \frac{1}{N} \sum_{\mu} \xi_i^{\mu} m^{\mu}(t)$$

$$\dots = 0.5 \left[ 1 + \rho \left( \frac{1}{N} \sum_{\mu} \xi_i^{\mu} m^{\mu}(t) \right) \right]$$

$$\rightarrow \frac{1}{N(1-d^2)} \sum_j (\xi_j^{\mu} - d) \xi_j^{\mu} =$$

$$= \frac{1}{N(1-d^2)} \sum_j (1 - d \sum_j^N) =$$

$$= \frac{1}{N(1-d^2)} \left( N - d \sum_j^N \right) \quad \textcircled{=}$$

$$\sum_j^N = \left( (d+1) N/2 - (N - (d+1) \frac{N}{2}) \right)$$

$$\swarrow +1$$

$$\searrow -1$$

$$= (d+1)N - N = dN$$

$$\textcircled{=} \frac{1}{N(1-d^2)} \left( N - d^2 N \right) = 1$$

b) Overlap is a dot-product-like measure of how close the current activation is to each of the memorized patterns.

c)

$$\begin{aligned}
 P[S_i(1) = 1] &= 0.5 \left[ 1 + g \left( \bigcup_{\mu} \sum \xi_i^{\mu} m^{\mu}(t) \right) \right] = \\
 &= 0.5 \left[ 1 + g \left( \bigcup \xi_i^4 M \right) \right] = \\
 &= 0.5 \left[ 1 + g(-\sqrt{M}) \right]
 \end{aligned}$$

For  $\bigcup = 1, M = 0.2$

$$\begin{aligned}
 P[S_i(1) = 1] &= 0.5 \left[ 1 + g(-0.2) \right] = \\
 &= 0.5 \left[ 1 - 0.8 \right] = 0.1
 \end{aligned}$$

d) With  $\bigcup = 1, M = 0.2$  we have

$$P[S_i(1)] = \xi_i^4 = 0.9$$



$$m^4(z) = \frac{1}{N(1-z^2)} \left( \sum_{j \in \text{match}} (z_j^4 - z) z_j^4 - \sum_{j \in \text{mismatch}} (z_j^4 - z) z_j^4 \right)$$

$$= \frac{|\text{match}|}{N} - \frac{|\text{mismatch}|}{N}$$

$$\langle m^4(z) \rangle = 0.9 - 0.1 = 0.8$$

$$\langle m^2(z) \rangle = 0 \quad \leftarrow \begin{array}{l} \text{Matches balance out,} \\ \text{mismatches balance out} \end{array}$$

e) Yes. The number of matches / mismatches between  $S(z)$  and  $\tilde{z}^4$  is binomially distributed. As  $N \rightarrow \infty$ , standard deviation grows with  $\sqrt{N}$  only, and fraction of matches (which determines  $m$ ) converges to having 0 standard deviation.

f)  $\rightarrow$  We have

$$P[S_i(t+1) = \tilde{z}_i^4] = 0.5 \left[ 1 + g \left( \sum m^4(t) \right) \right]$$

So we have

$$\begin{aligned} m^4(t+1) &= 2 \left( 0.5 \left( 1 + g \left( 5m^4(t) \right) \right) \right) - 1 \\ &= 1 + g \left( m^4(t) \right) - 1 = g \left( m^4(t) \right) \end{aligned}$$

Which we can solve for  $m = m^4(\infty)$  by

$$m = g(m)$$

$$m = 1$$

Find overlap with pattern 4 is 1.

Find overlap with pattern 0 is 0.

$$\rightarrow m^4(0) = 0.2$$

$$m^4(1) = 0.8$$

$$m^4(2) = 1$$

Takes two timesteps  
to converge! Cool!

$$g) \quad w_{ij} = \frac{1}{(1-d^2)N} \sum_{\mu} (\xi_i^{\mu+1}) (\xi_j^{\mu} - d)$$

$$h_i(t) = \sum_{\mu} \xi_i^{\mu+1} m^{\mu}(t)$$

So essentially  $\tau$  rotate pattern at every timestep!

$$m^{i+1}(t+1) = g(m^i(t))$$

And with  $m^4(0) = 0.5$  we get

$$m^1(1) = 0$$

$$m^5(1) = 1$$

$$m^4(1) = 0$$

$$m^6(1) = 0$$

## Exercise 3

a)  $\rightarrow h_0$  is the neuron's resting potential

$\rightarrow J$  represents how tightly neurons are connected to each other

b) ?  $\ddot{n}$  FIND OUT!

c) We want  $\frac{dh}{dt} = 0$   
 $\swarrow$   
 $h = g(h)$

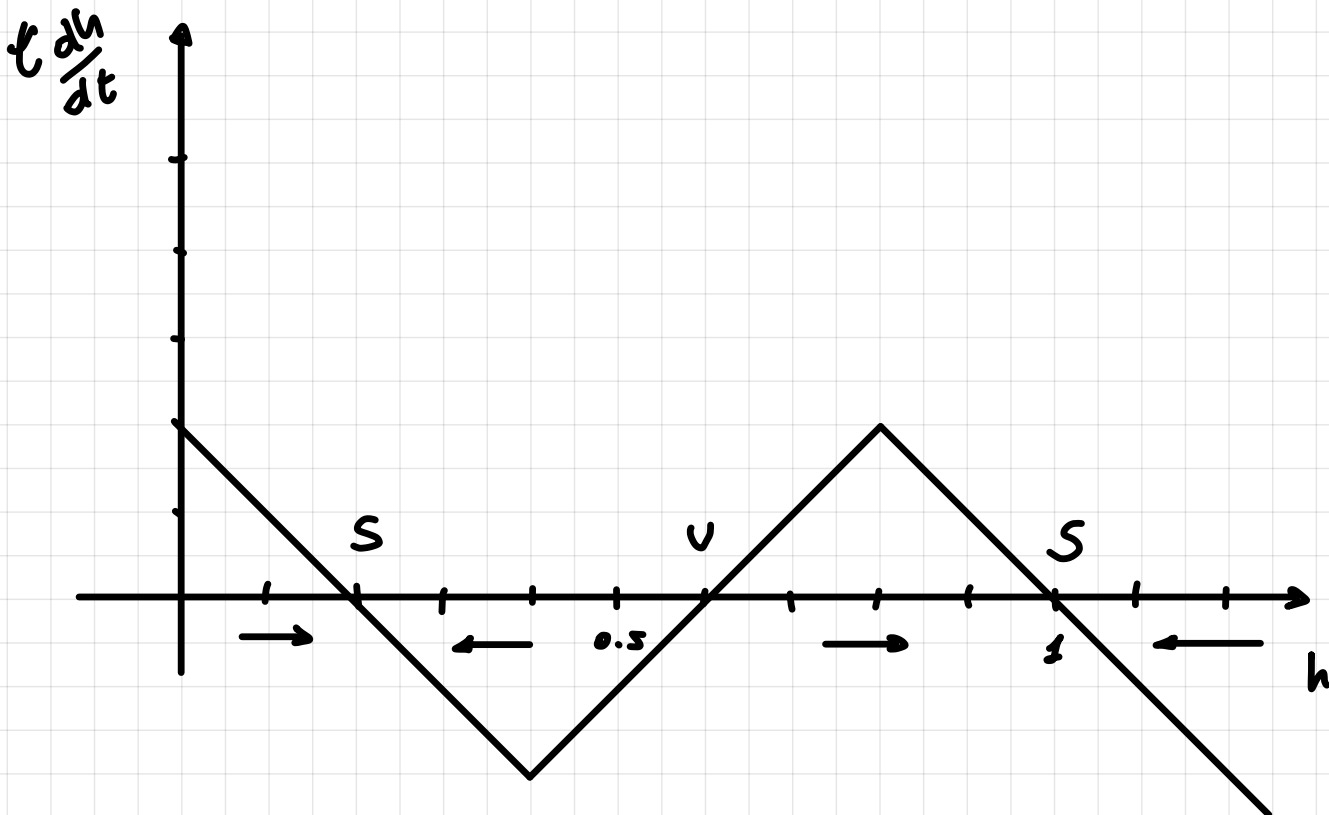
1st branch:  $h = 0.2$

3rd branch:  $h = 1$

2nd branch:  
 $\left\{ \begin{array}{l} h = 2h - 0.6 \\ h = 0.6 \end{array} \right.$

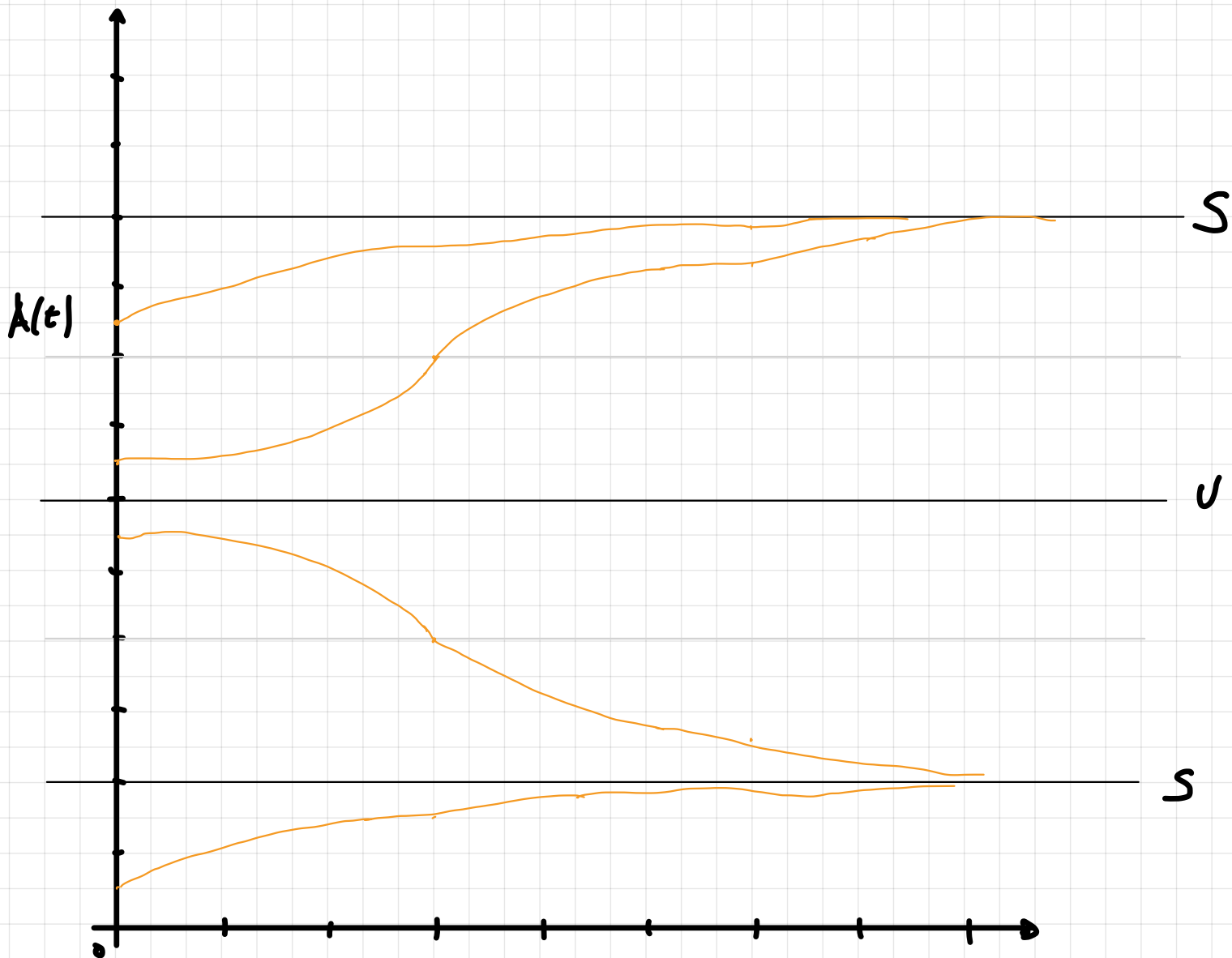
$$d) \quad \tau \frac{dh}{dt} = g(h) - h$$

$$= \begin{cases} 0.2 - h & h < 0.4 \\ h - 0.6 & h \in [0.4, 0.8] \\ 1 - h & h > 0.8 \end{cases}$$



e) Everything converges exponentially to / from the relevant fixed point by  $e^{\pm t/\tau}$

So we need to divide / multiply the distance by 3 every 3 timesteps



## Exercise 4

$$d(t) = \sigma(t) + \sigma(t - t_0)$$

$$\sigma(t) = \begin{cases} b e^{-t/\tau_m} & \text{if } t > 0 \\ 0 & \text{otherwise} \end{cases}$$

d) Each spike doublet elicits in total

$$\begin{aligned} \int_0^{\infty} d(t) dt &= 2 \int_0^{\infty} \sigma(t) dt : \\ &= 2b \int_0^{\infty} e^{-t/\tau_m} dt = 2b\tau_m \end{aligned}$$

So the expected voltage is

$$\langle v \rangle = \frac{2b\tau_m V_0 K}{\text{Single doublet} \quad \text{Doublets per second per neuron} \quad \text{Number of neurons}}$$

CORRECT!  
NUMERICALLY  
VERIFIED!

b1)

$$\langle v \rangle = 2.2 \text{ mV} \cdot 5 \text{ ms} \cdot 0.5 \text{ Hz} \cdot 1000 =$$

$$= 2.2 \text{ mV} \cdot \cancel{5 \text{ ms}} \cdot 5 \cdot 10^{-4} / \cancel{\text{ms}} \cdot 1000 =$$

$$= (2.2 \cdot 5 \cdot 0.5) \text{ mV} = 10 \text{ mV}$$

b2) No, the voltage will hover around a half / a third of the threshold. If  $k$  was smaller it could fire by fluctuation!

c) Mean input potential would double  
(the time integral of each spike would halve, but we would have four times as many).