SYDE 556/750

Simulating Neurobiological Systems Lecture 6: Recurrent Dynamics

Chris Eliasmith

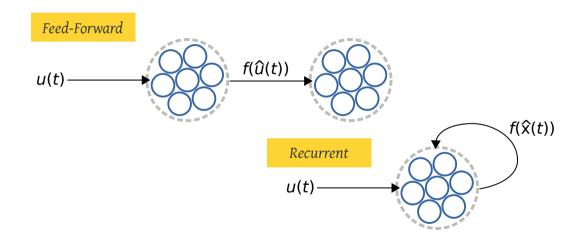
October 7 & 20, 2022

- ► Slide design: Andreas Stöckel
- ► Content: Terry Stewart, Andreas Stöckel, Chris Eliasmith

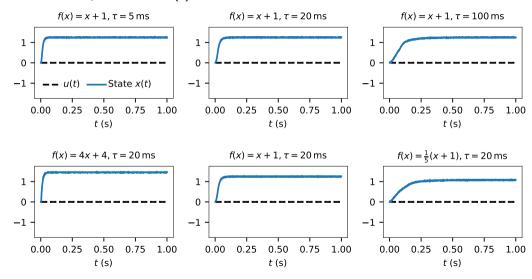




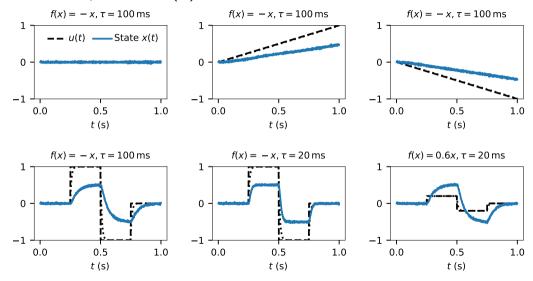
Feed Forward vs. Recurrent Connections



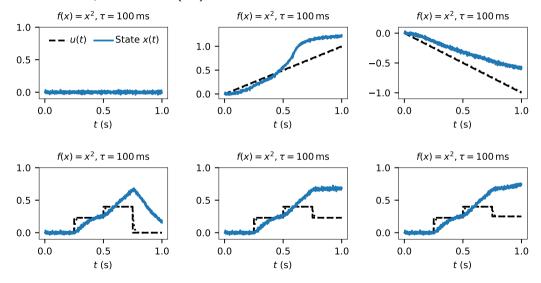
Recurrence Experiments (I)



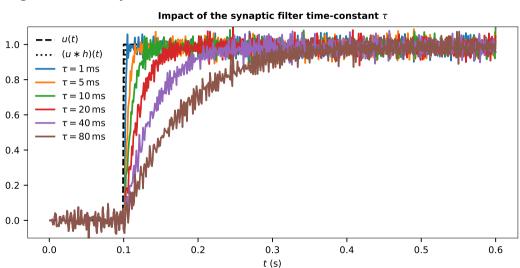
Recurrence Experiments (II)



Recurrence Experiments (III)



Making Sense of Dynamics



Behaviour of a Recurrent Connection

Feed-forward

$$x(t) = g(u(t)) * h(t)$$

Recurrent

$$x(t) = (g(u(t)) + f(x(t))) * h(t)$$

$$X(s) = (G(s) + F(s))H(s)$$

$$X(s) = (G(s) + F(s))\frac{1}{1 + s\tau}$$

$$X(s) + s\tau X(s) = G(s) + F(s)$$

$$s\tau X(s) = G(s) + F(s) - X(s)$$

$$sX(s) = \frac{F(s) - X(s)}{\tau} + \frac{G(s)}{\tau}$$

$$\frac{dx}{dt} = \frac{f(x(t)) - x(t)}{\tau} + \frac{g(u(t))}{\tau}$$

Behaviour of a Recurrent Connection

$$\frac{dx}{dt} = \frac{f(x(t)) - x(t)}{\tau} + \frac{g(u(t))}{\tau}$$

If we want this

$$\frac{dx}{dt} = a(x) + b(u)$$

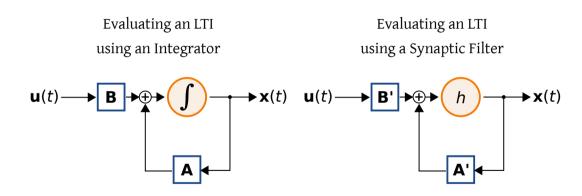
Then we set

$$f(x) = \tau a(x) + x$$
$$g(x) = \tau b(u)$$

And we get

$$\frac{dx}{dt} = \frac{(\tau a(x) + x) - x}{\tau} + \frac{(\tau b(u))}{\tau}$$
$$\frac{dx}{dt} = a(x) + b(u)$$

Implementing Dynamics using a Neural Ensemble



Implementing Dynamical Systems as a Neural Ensemble

LTI System

$$arphi(\mathbf{u}, \mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$

 $arphi'(\mathbf{u}, \mathbf{x}) = \mathbf{A}'\mathbf{x} + \mathbf{B}'\mathbf{u}$
 $\mathbf{A}' = \tau \mathbf{A} + \mathbf{I}$
 $\mathbf{B}' = \tau \mathbf{B}$.

Additive Time-Invariant System

$$\varphi(\mathbf{u}, \mathbf{x}) = f(\mathbf{x}) + g(\mathbf{u})$$
$$\varphi'(\mathbf{u}, \mathbf{x}) = f'(\mathbf{x}) + g'(\mathbf{u})$$
$$f'(\mathbf{x}) = \tau f(\mathbf{x}) + \mathbf{x}$$
$$g'(\mathbf{u}) = \tau g(\mathbf{u})$$

"General" Recipe

Scale the original dynamics by τ , add feedback x

NEF Principle 3: Dynamics

Time-Invariant Dynamical System

$$\frac{\mathrm{d}\mathbf{x}(t)}{\mathrm{d}t} = f(\mathbf{x}(t), \mathbf{u}(t))$$

Linear Time-Invariant (LTI)

Dynamical System

$$\frac{\mathrm{d}\mathbf{x}(t)}{\mathrm{d}t} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$

NEF Principle 3 – Dynamics

Neural dynamics are characterized by considering neural representations as control theoretic state variables. We can use control theory (and dynamical systems theory) to analyse and construct these systems.

Low-pass Filter Example

Desired behaviour

$$\frac{dx}{dt} = \frac{u - x}{\tau_{desired}}$$

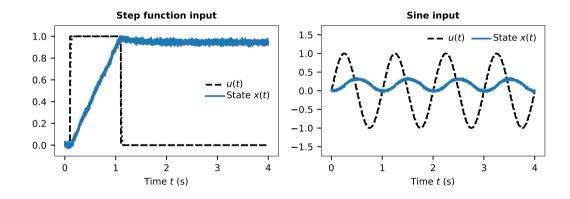
Feed-forward input

$$g(u) = au_{synapse}(rac{u}{ au_{desired}})$$
 $g(u) = rac{ au_{synapse}}{ au_{desired}}u$

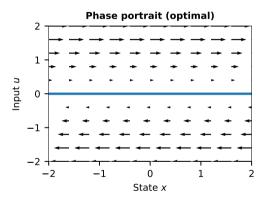
Recurrent

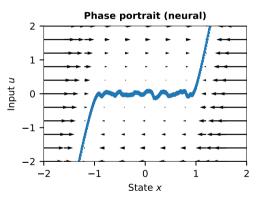
$$f(x) = au_{synapse}(rac{-x}{ au_{desired}}) + x$$
 $f(x) = (1 - rac{ au_{synapse}}{ au_{desired}})x$

Integrator Example (I)

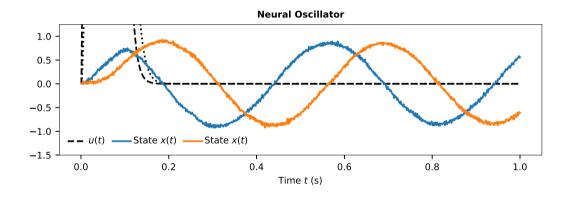


Integrator Example (II)

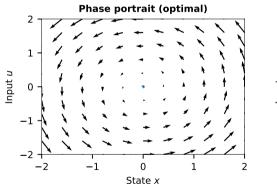


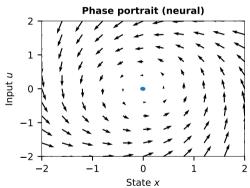


Oscillator Example (I)

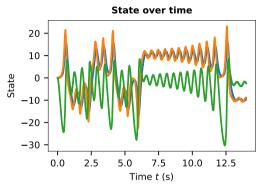


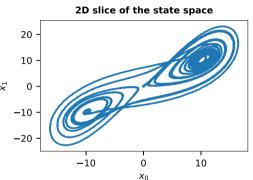
Oscillator Example (II)





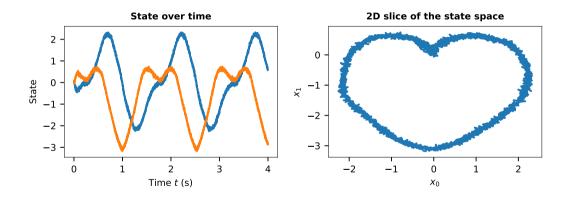
Lorentz Attractor



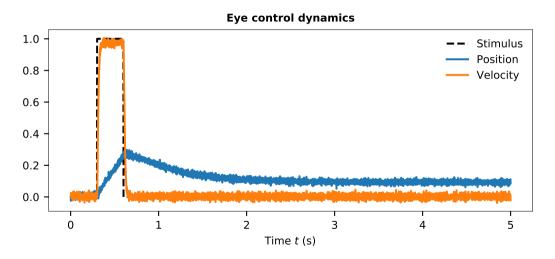


$$\frac{\mathrm{d}\mathbf{x}(t)}{\mathrm{d}t} = \begin{pmatrix} 10x_2(t) - 10x_1(t) \\ -x_1(t)x_3(t) - x_2(t) \\ x_1(t)x_2(t) - \frac{8}{3}(x_3(t) + 28) - 28 \end{pmatrix}$$

Heart Shape



Horizontal Eye Control



Administrative Remarks

- ► Assignment 2 due Monday, October 25
- ► **Project proposals** due Wednesday, October 27

Image sources

Title slide

"The Canada 150 Mosaic Mural" Author: Mosaic Canada Murals.

From Wikimedia.