

# ReLU Recurrent Neural Networks

Daniel and David

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# Outline

① Background

② Fixed Points (Hahnloser)

③ Hand Built Networks

④ Trained Networks

- Short term memory
- RDM

⑤ Small Networks

- 1-D Networks
- 2-D Networks

# Table of Contents

1 Background

2 Fixed Points (Hahnloser)

3 Hand Built Networks

4 Trained Networks

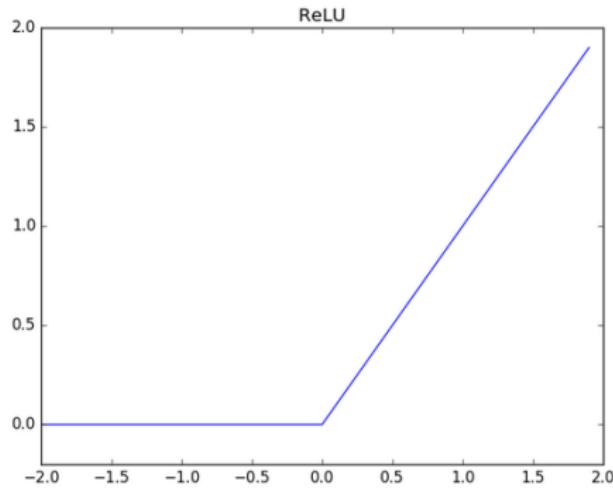
- Short term memory
- RDM

5 Small Networks

- 1-D Networks
- 2-D Networks

# ReLU

$$f(x) = \max(x, 0)$$



# Network Equations

Continuous

$$\tau \dot{x} = -x + W_{rec}f(x) + b_{rec} + W_{in}u + Noise$$

# Network Equations

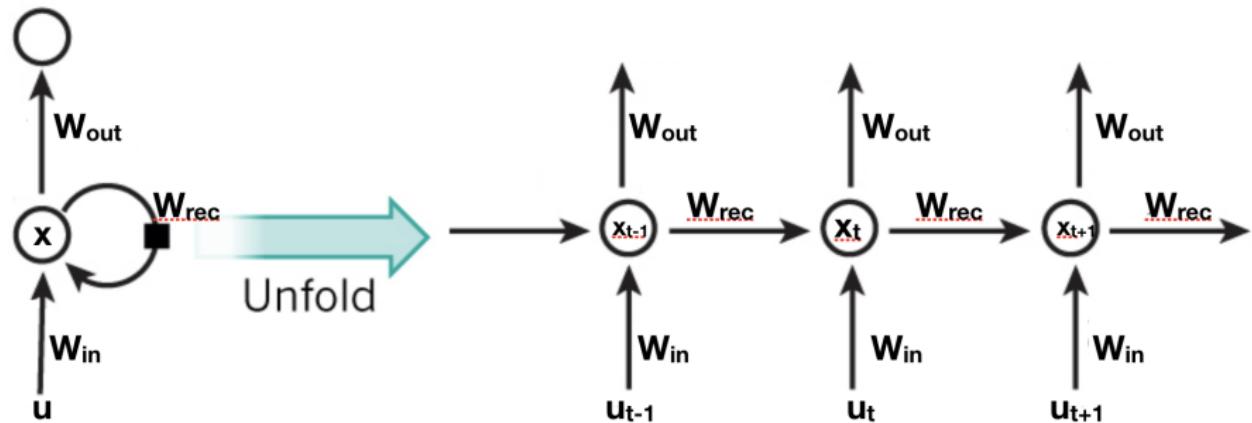
Continuous

$$\tau \dot{x} = -x + W_{rec}f(x) + b_{rec} + W_{in}u + Noise$$

Discrete

$$x_{t+1} = \left(1 - \frac{dt}{\tau}\right)x + \left(\frac{dt}{\tau}\right)(W_{rec}f(x_t) + b_{rec} + W_{in}u_t) + Noise$$

# Network Visualization



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# Hahnloser, 1997

Simplified model:

$$\tau \dot{x} = -x + W_{rec}f(x) + J$$

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Note

$$\begin{pmatrix} w & w & w \\ w & w & w \\ w & w & w \end{pmatrix} \begin{pmatrix} x \\ 0 \\ x \end{pmatrix} = \begin{pmatrix} w & 0 & w \\ w & 0 & w \\ w & 0 & w \end{pmatrix} \begin{pmatrix} x \\ x \\ x \end{pmatrix}$$

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So, define  $W^+$  such that

$$W^+x = W_{rec} \max(x, 0) = W_{rec}f(x)$$

# Hahnloser, 1997

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So, define  $W^+$  such that

$$W^+x = W_{rec} \max(x, 0) = W_{rec}f(x)$$

The region where  $W^+$  is the recurrent matrix is called a *partition*.

# Hahnloser, 1997

At a fixed point

$$\begin{aligned}0 &= -x + W^+x + J \\x - W^+x &= J \\x &= (1 - W^+)^{-1}J\end{aligned}$$

# Hahnloser, 1997

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Subtelty:  $x$  must be in the partition defined by  $W^+$

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**3 Hand Built Networks**

4 Trained Networks

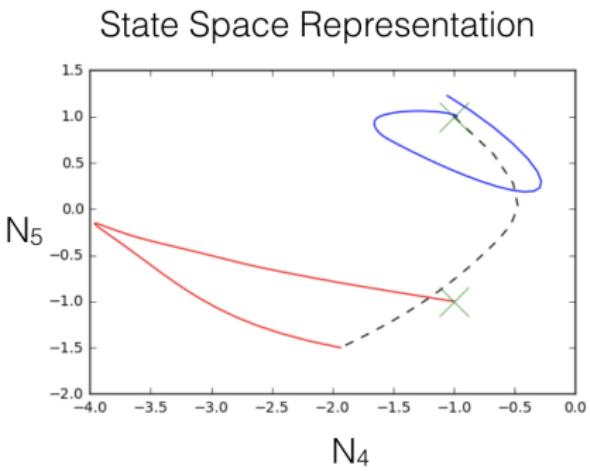
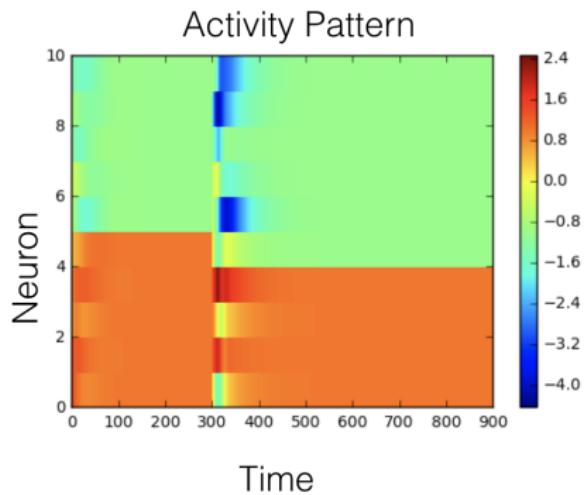
- Short term memory
- RDM

5 Small Networks

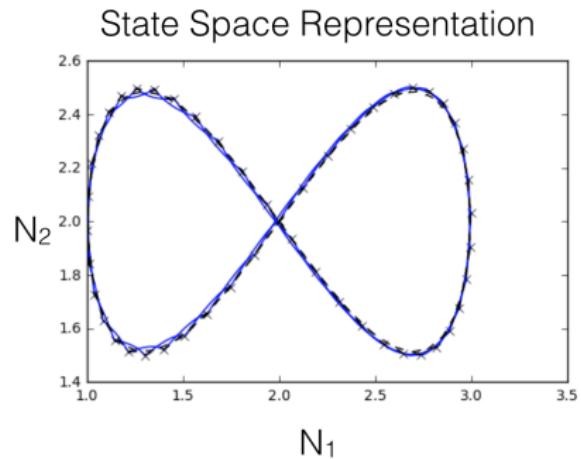
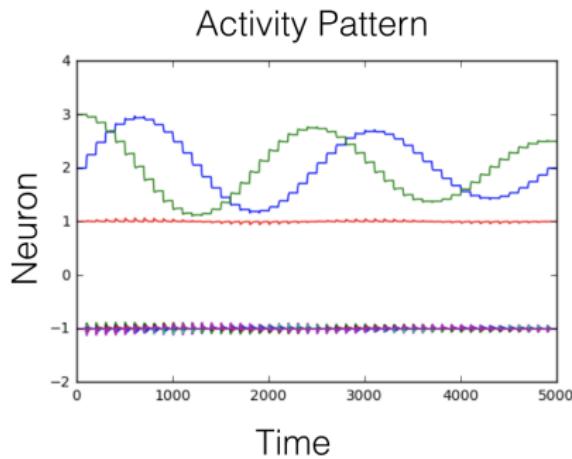
- 1-D Networks
- 2-D Networks

# Hand Built Networks

$$x = (1 - W^+)^{-1} J$$

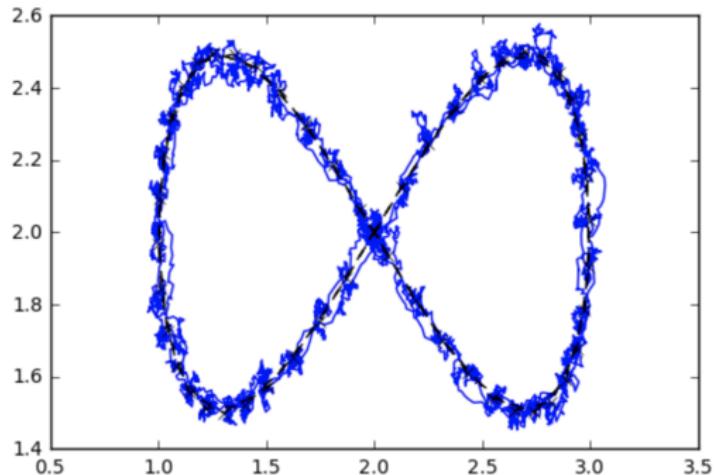


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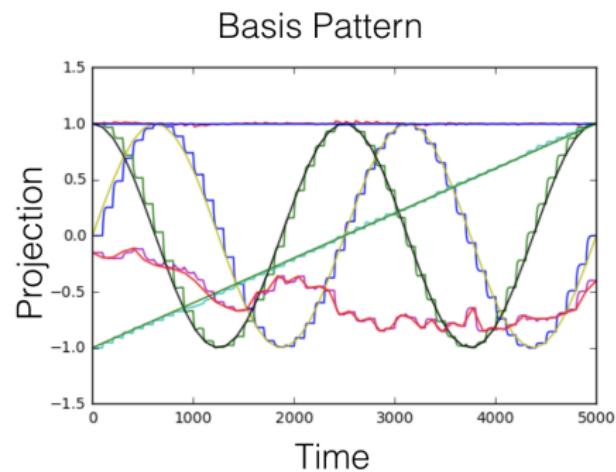
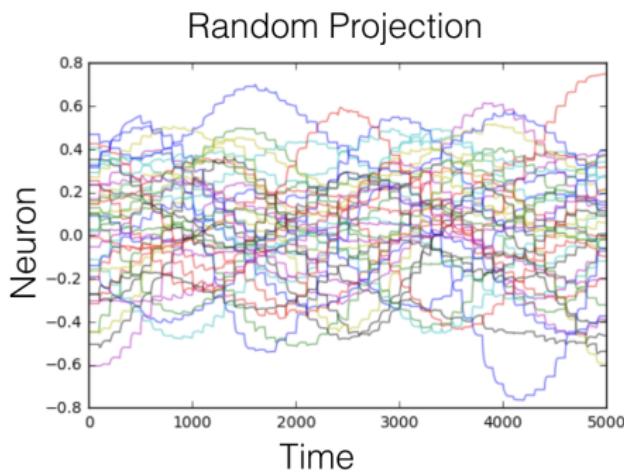


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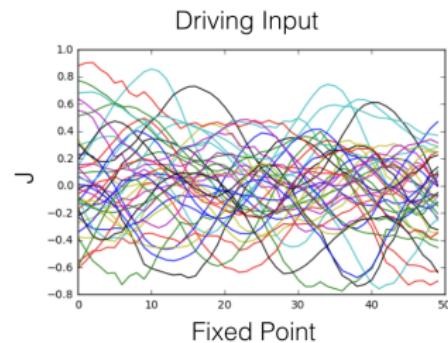
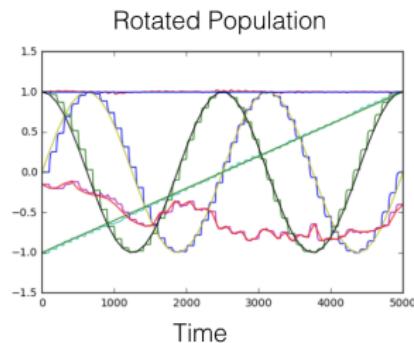
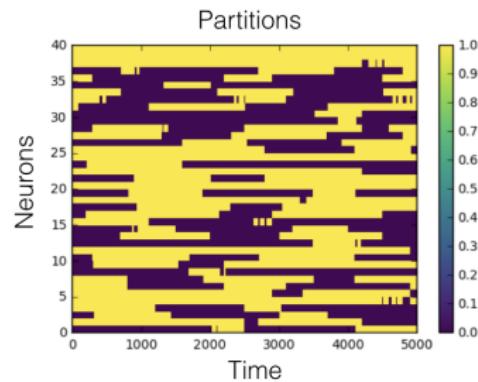
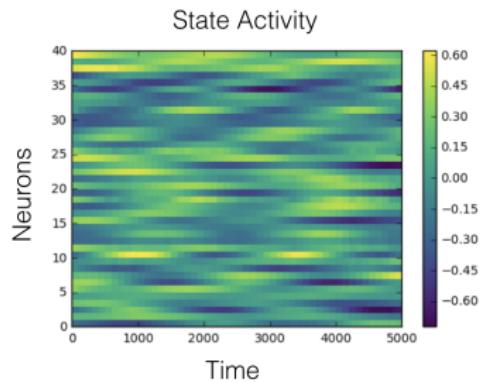
Noisy ReLU Networks



# Hand Built Networks



# Hand Built Networks



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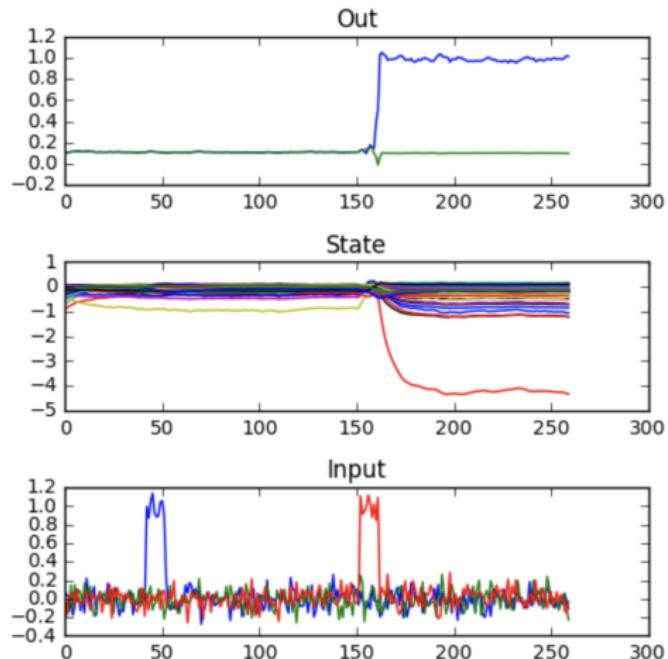
4 Trained Networks

- Short term memory
- RDM

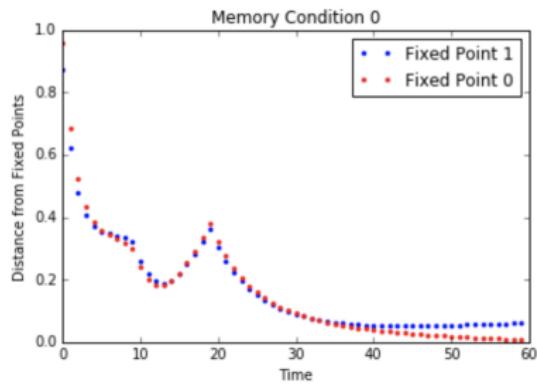
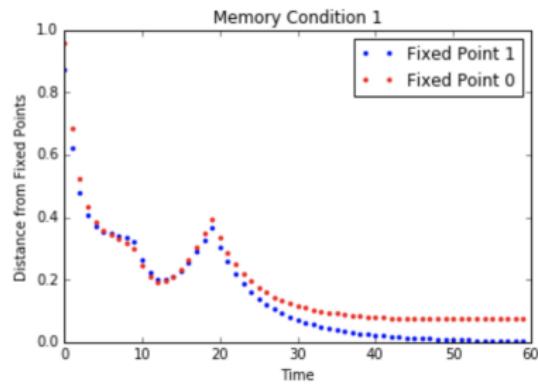
5 Small Networks

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- 2-D Networks

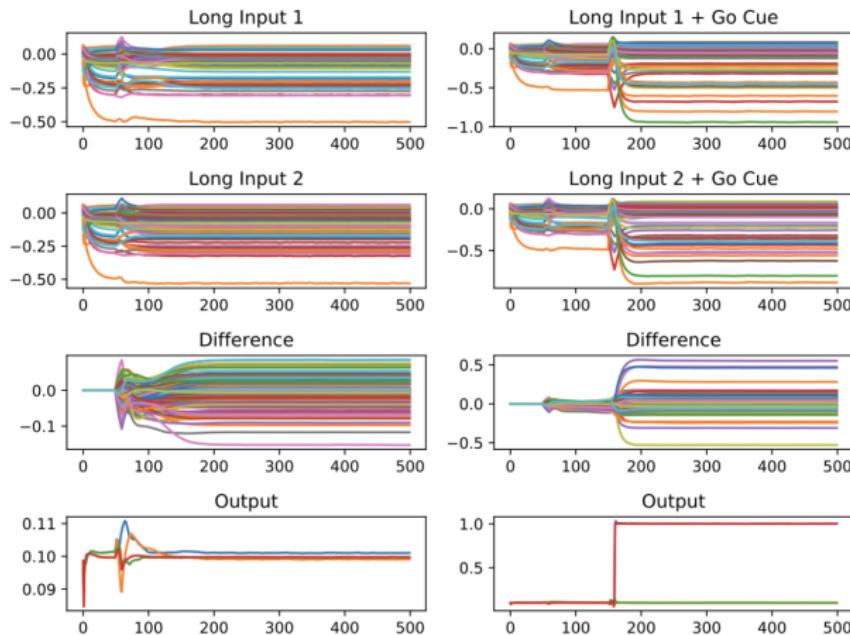
# Memory Task



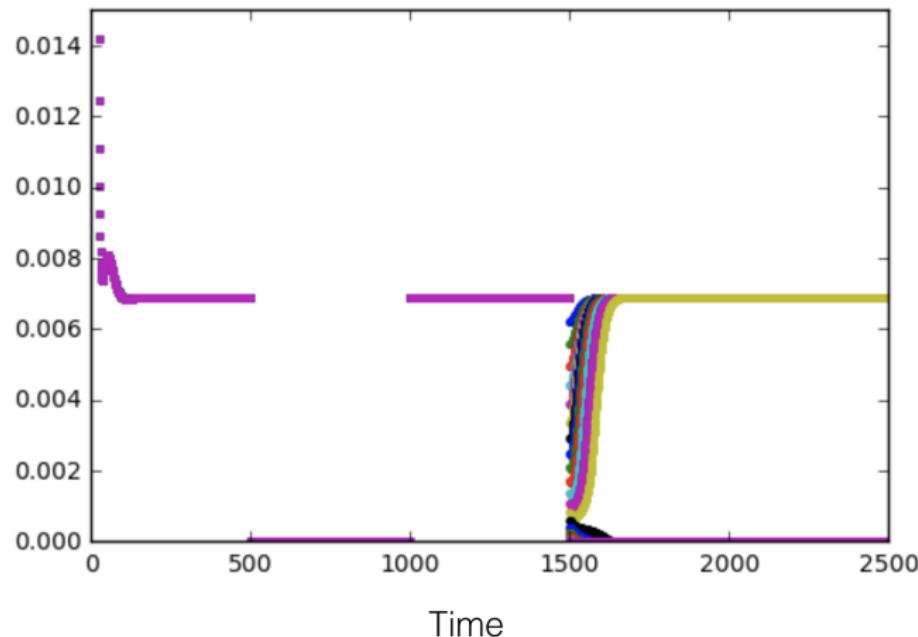
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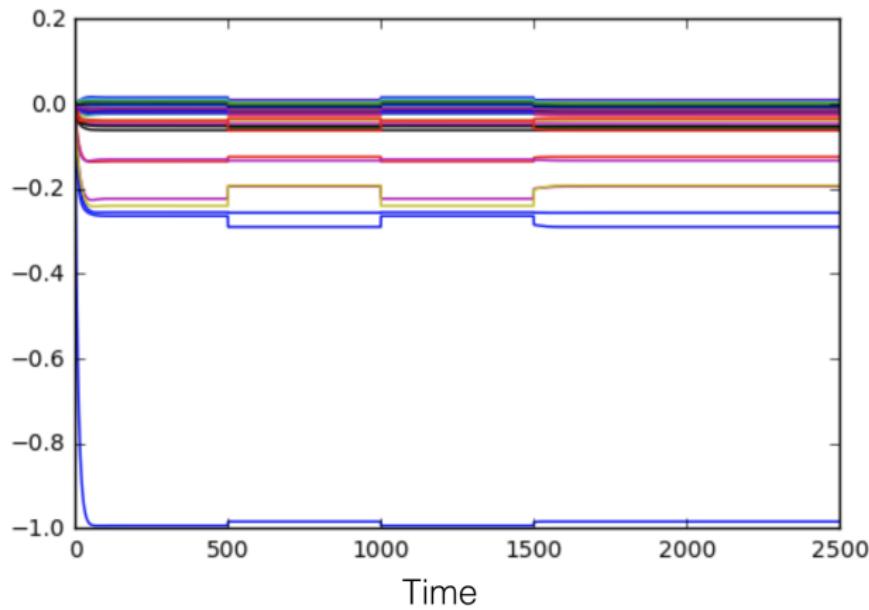
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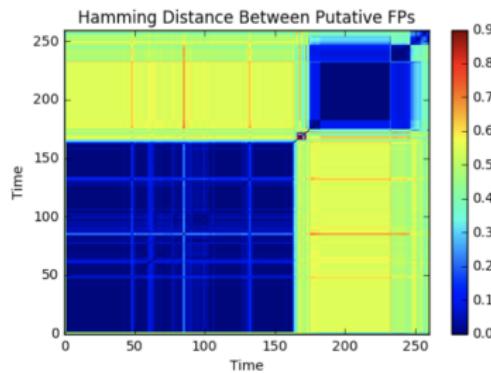
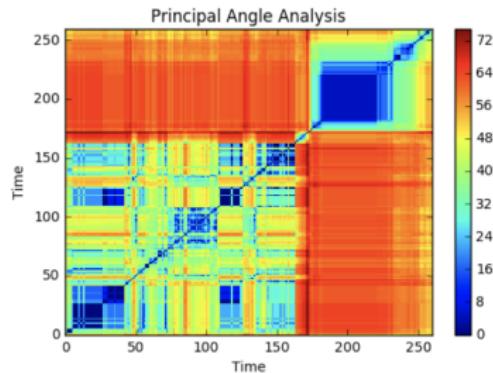
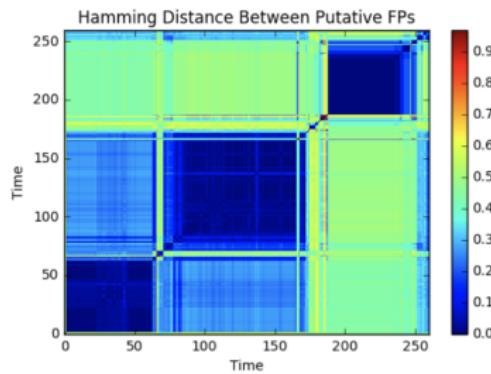
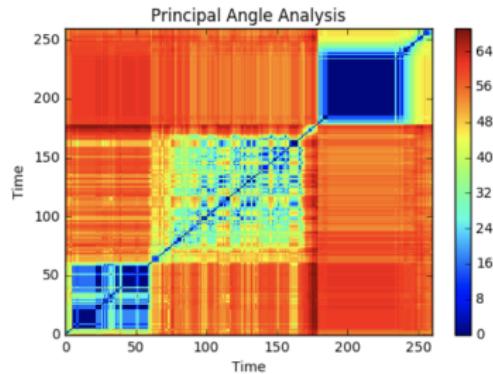
# Memory Task



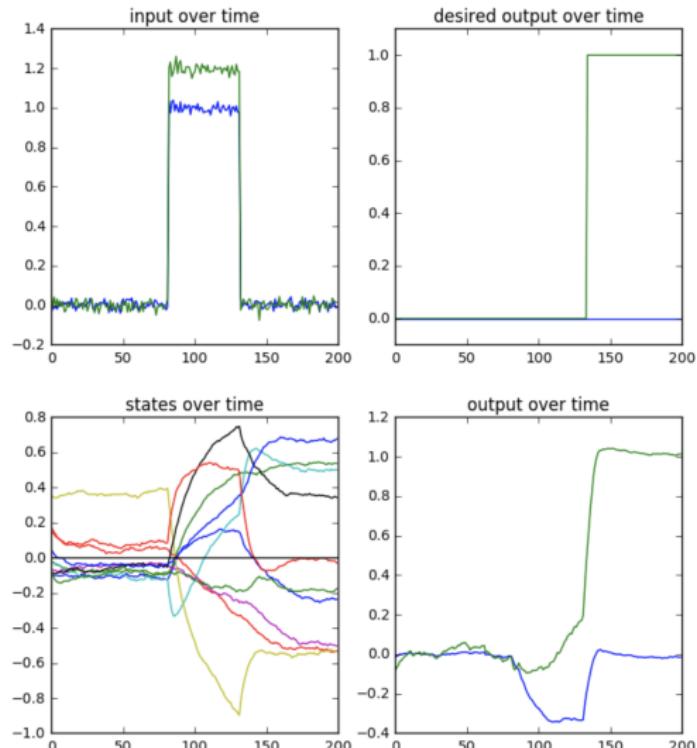
# Memory Task



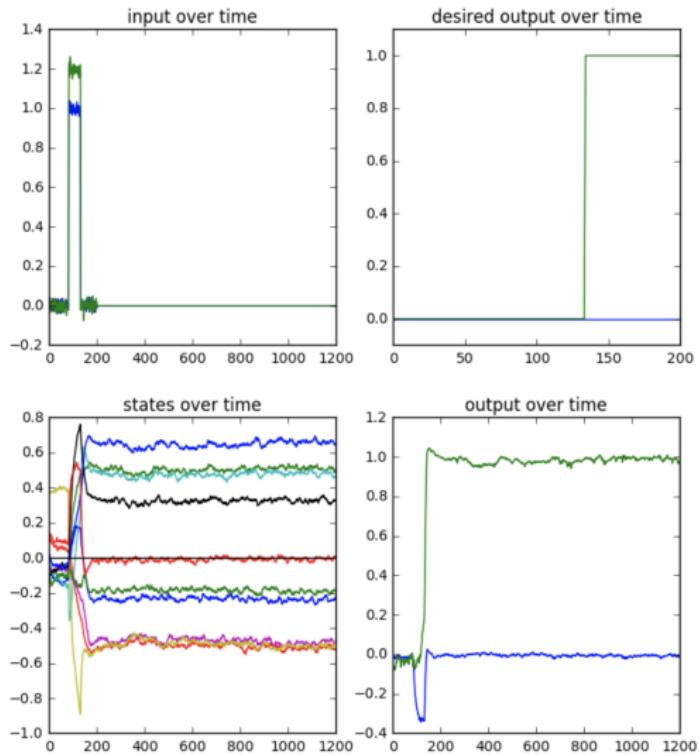
# Memory Task



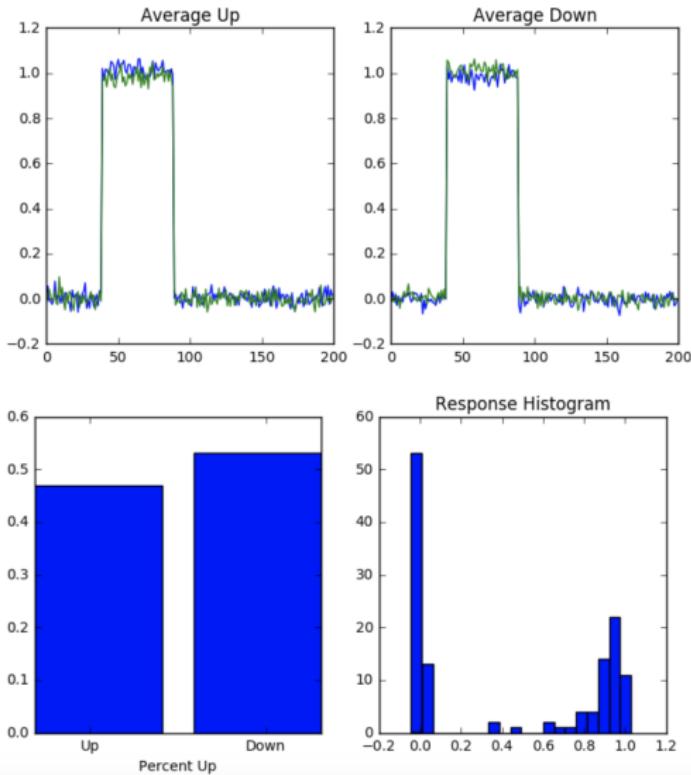
# RDM Task



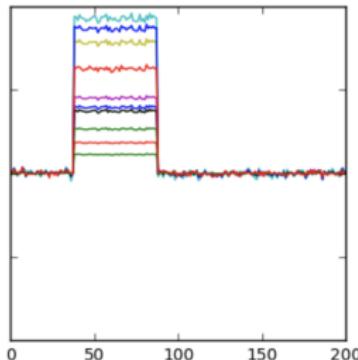
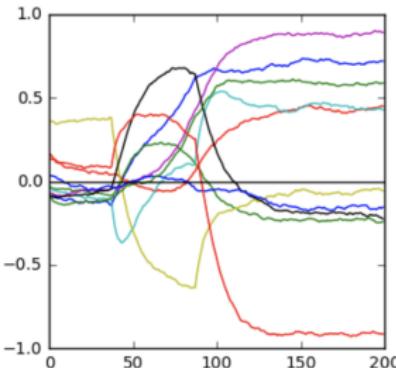
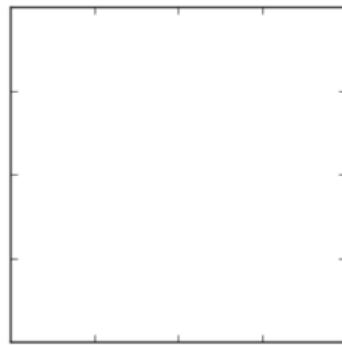
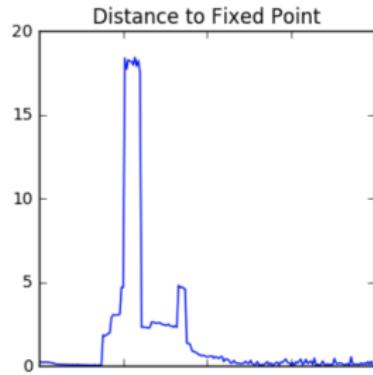
# Long Time Scales



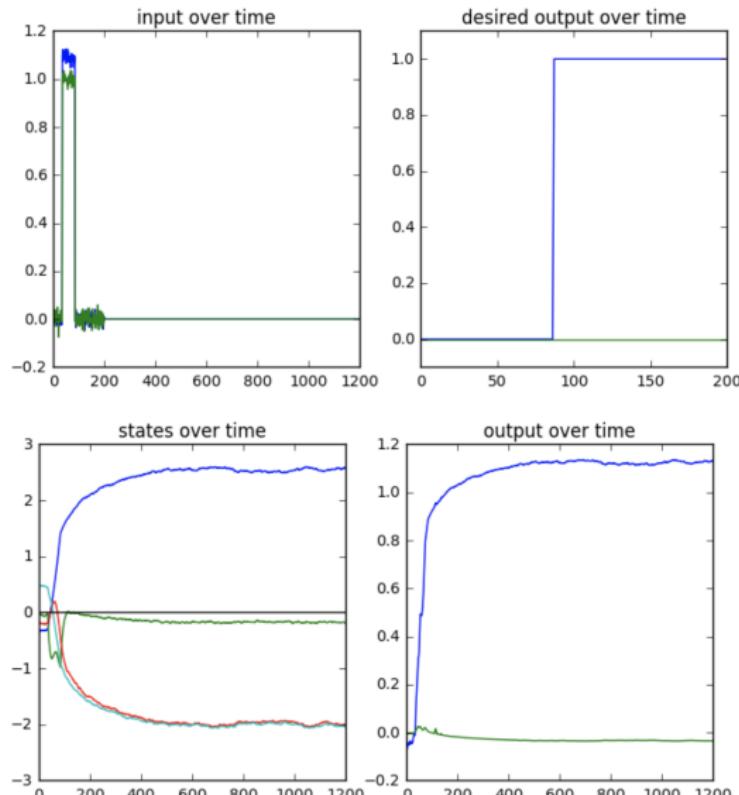
# Random Inputs



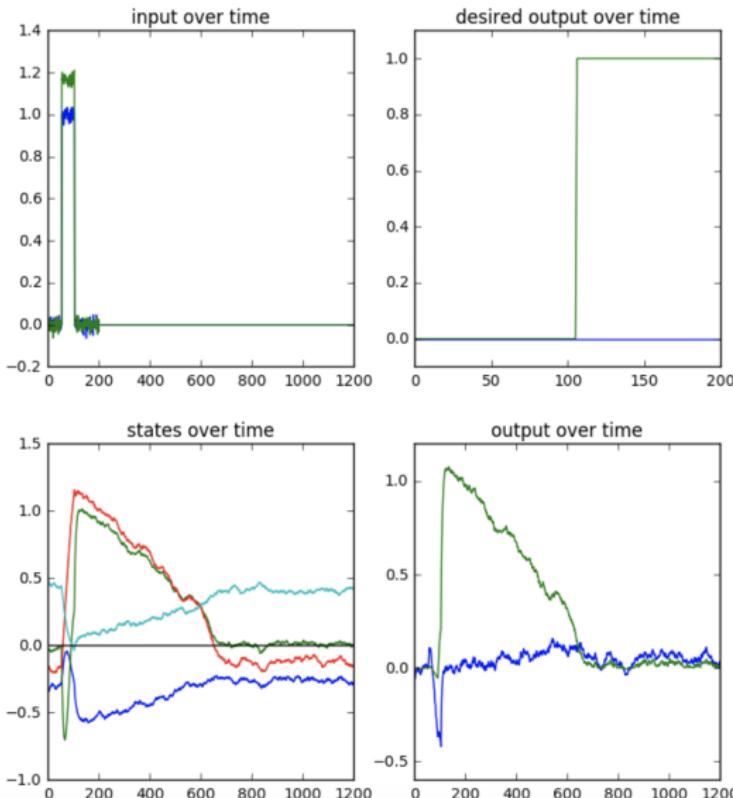
# Distance to Fixed Point



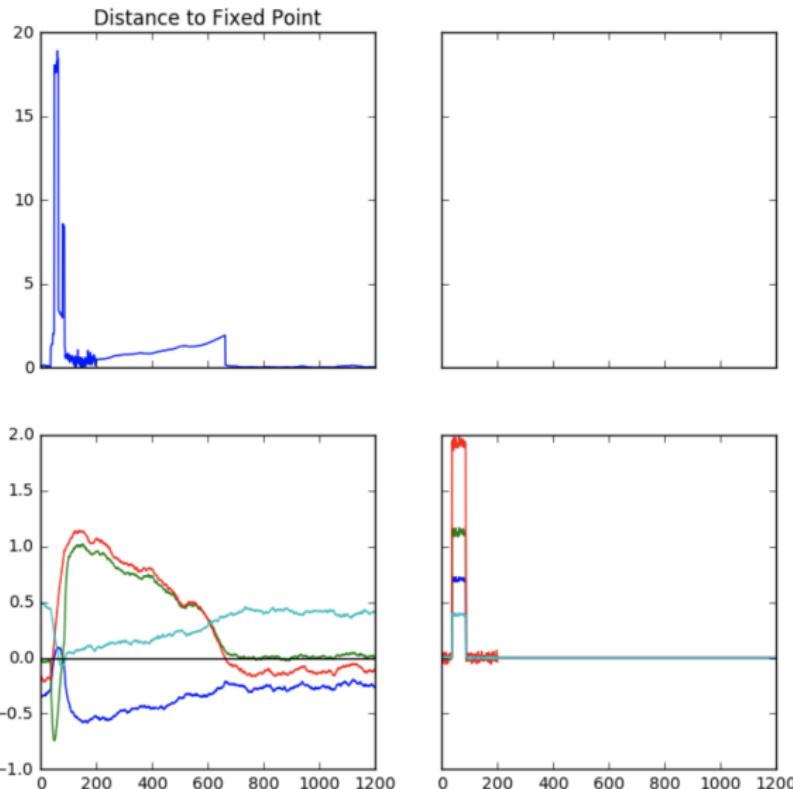
# 4 Neuron Model



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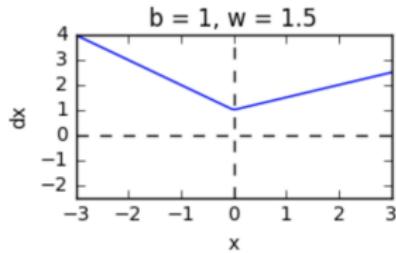
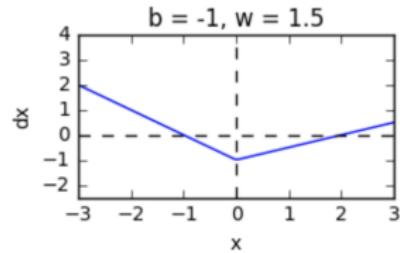
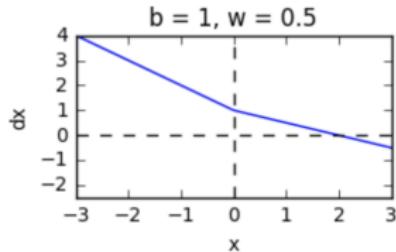
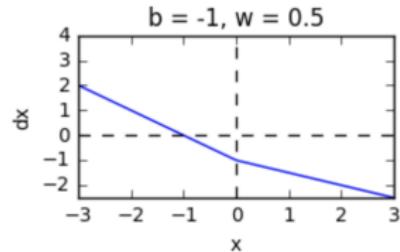
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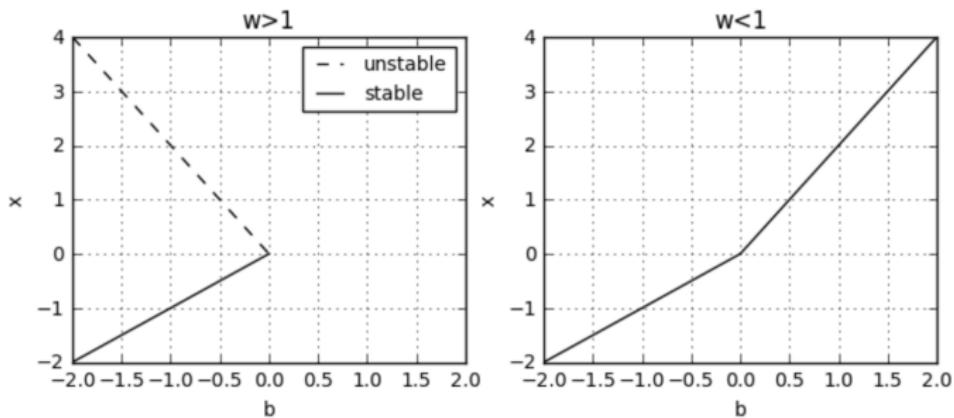
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# 1-D ReLU Networks

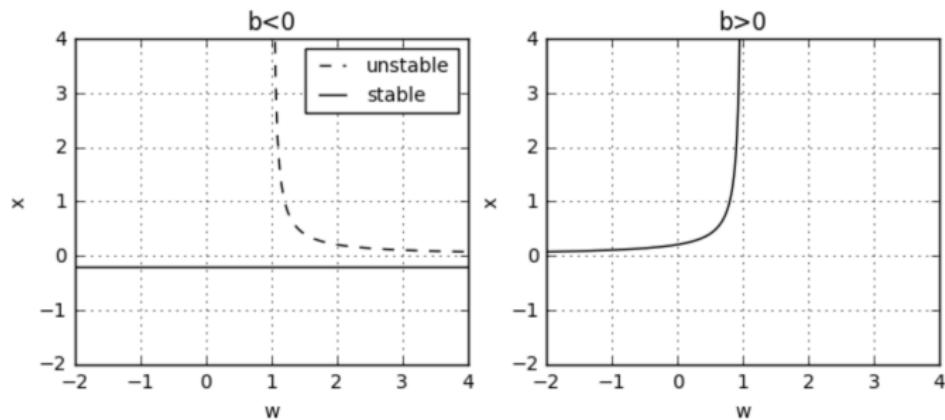
$$\dot{x} = -x + wf(x) + b$$
$$f(x) = \max(x, 0)$$



# 1-D ReLU Networks

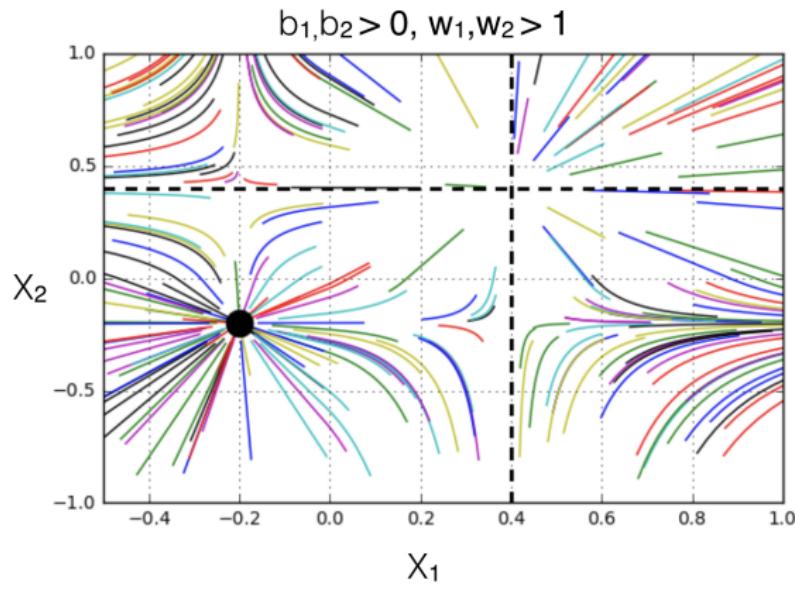


# 1-D ReLU Networks



# 2-D ReLU Networks

$$\dot{x}_i = -x_i + \sum_{j=1}^n w_{i,j} f(x_j) + b_i$$



# 2-D ReLU Networks

