

Simulations of Monocular Deprivation in Mouse

Brian Blais
Scott Kuindersma



Bryant
UNIVERSITY

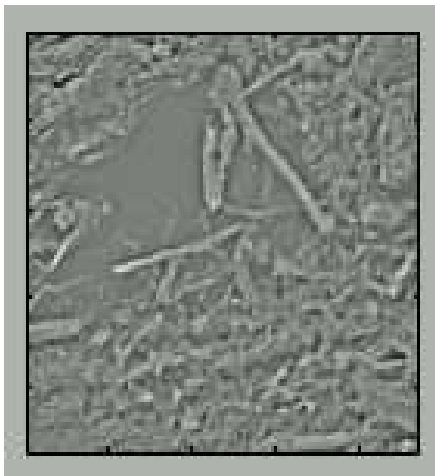
Mouse Simulations

- larger retinal filter
- contralateral bias

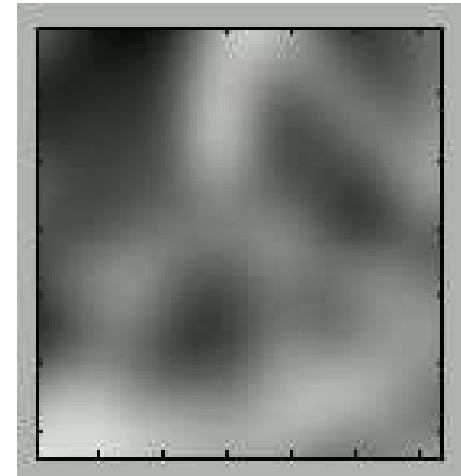
The inputs for the normal rearing (NR) simulations were a set of 12 natural images with the retinal filter (Difference of Gaussian) increased by a factor of 10.



No Filter



Cat
DoG 1:3



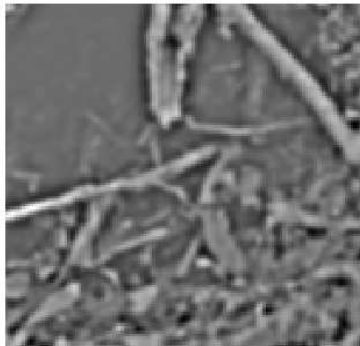
Mouse
DoG 10:30

Two methods for contralateral bias

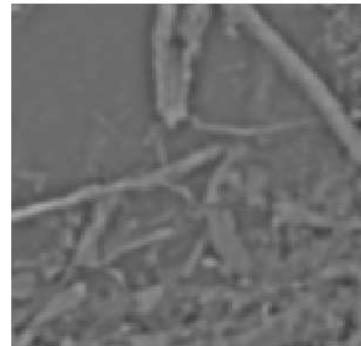
- **Alter image intensity**
- Reduce number of weights

Different image sets were used for the left and right input channels. The right (ipsilateral) channel image intensity was reduced by a factor of 2.5.

Cat
DoG 1:3



Cat with
bias



- Cats do not actually have a bias

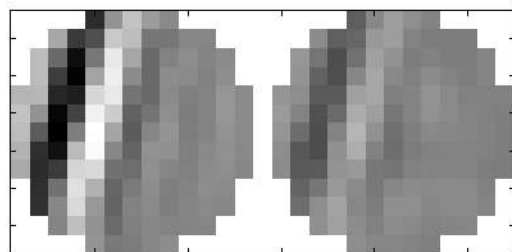


Mouse
DoG 10:30

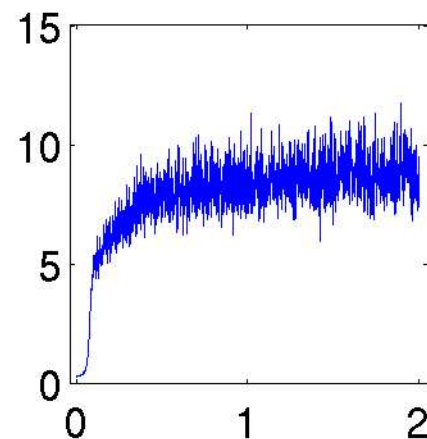


Cat with contralateral bias – NR, intensity induced bias

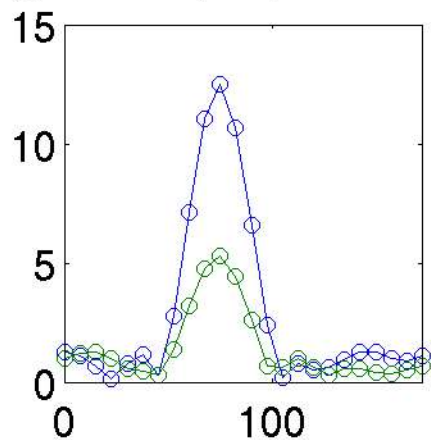
Weights



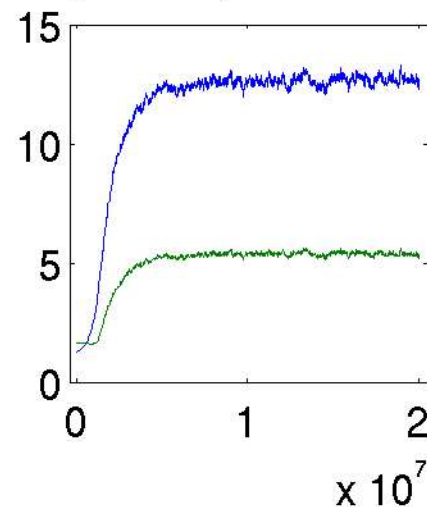
Thresholds (value vs. time)



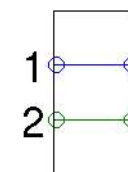
Tuning Curves (response vs. angle)



Responses (value vs. time) $\times 10^7$

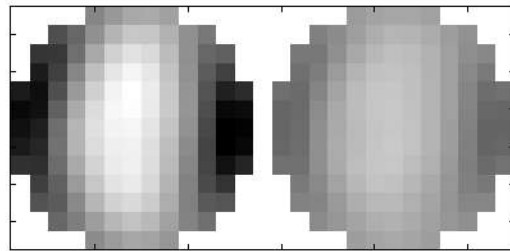


Channel Legend

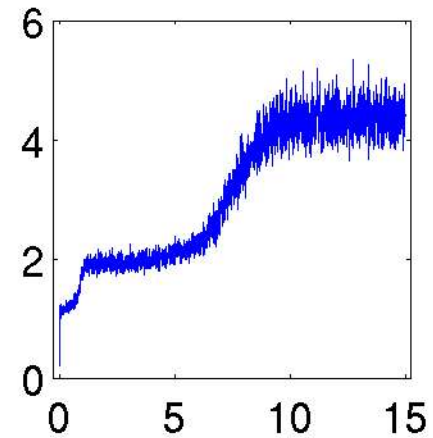


Mouse - NR, intensity induced bias

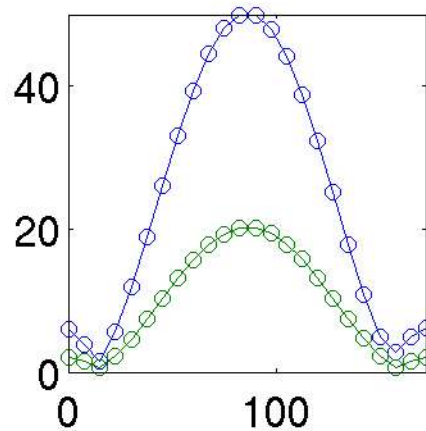
Weights



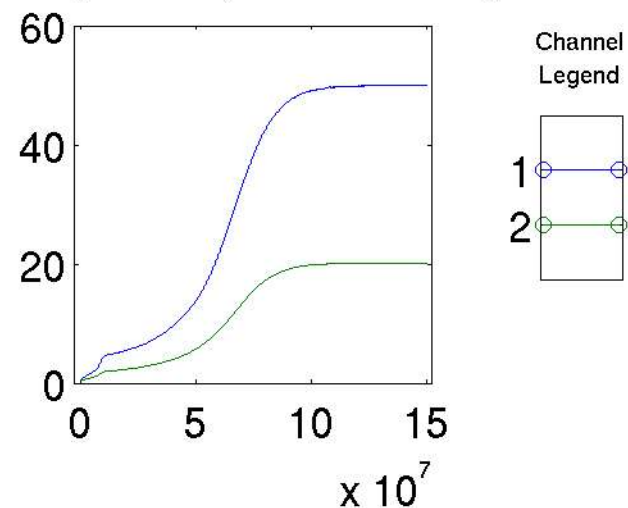
Thresholds (value vs. time)



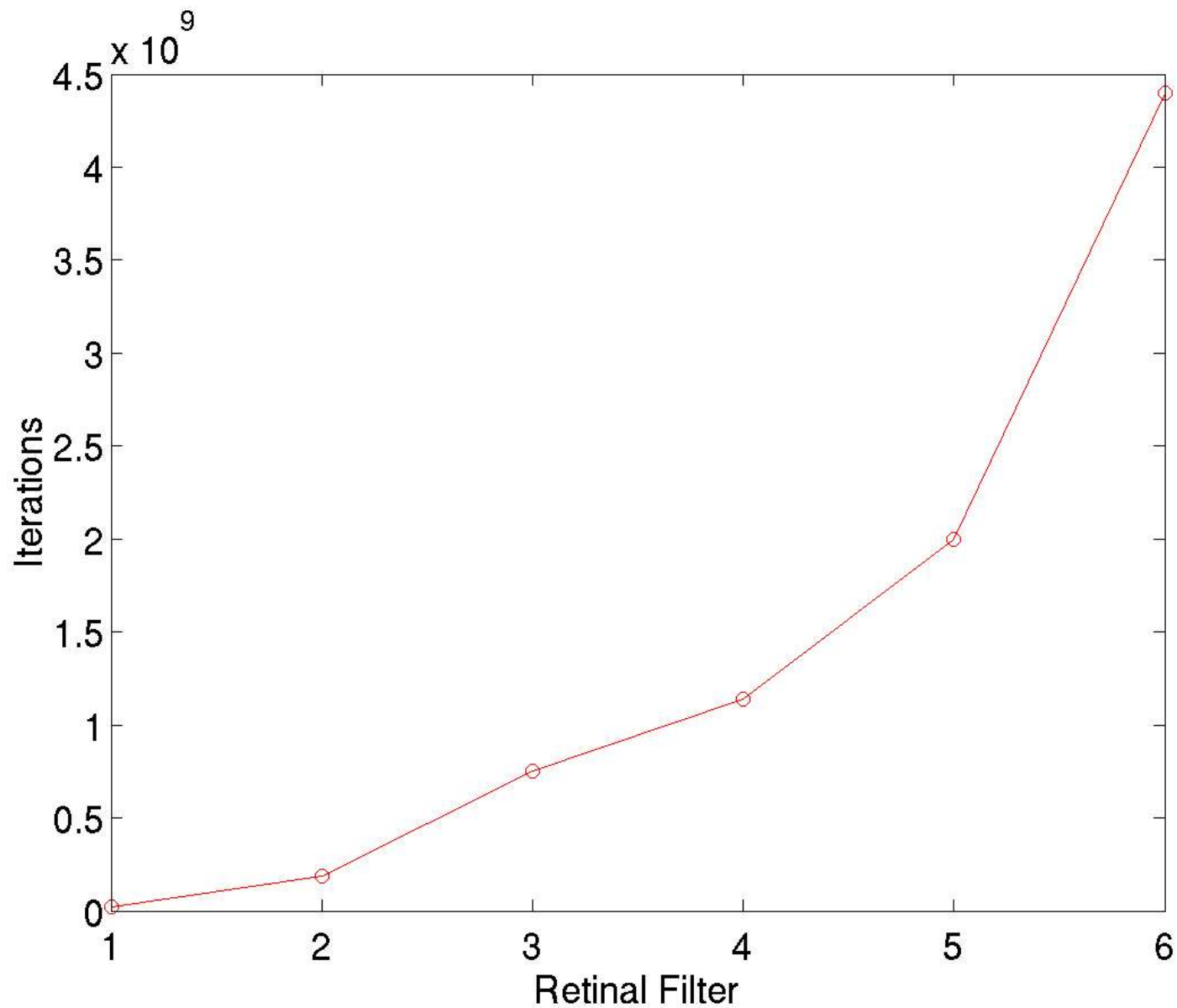
Tuning Curves (response vs. angle)



Responses (value vs. time) $\times 10^7$

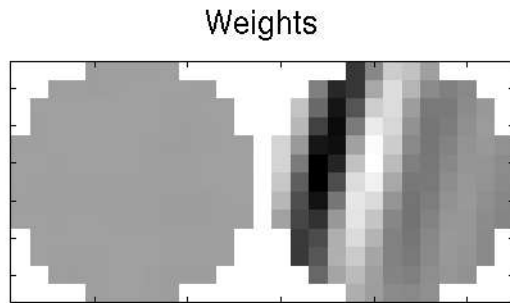


Effect of retinal filter (DoG) on required time for convergence

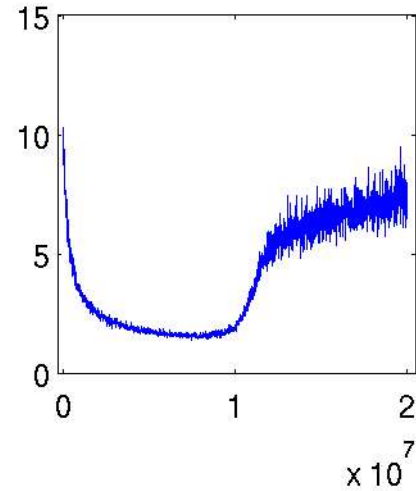


Cat with bias – Monocular Deprivation (MD), High noise (lid suture)

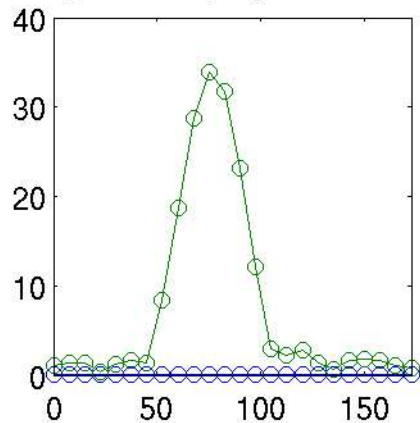
Gaussian SD = 0.8



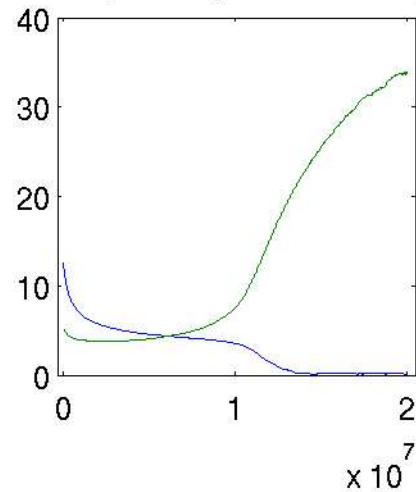
Thresholds (value vs. time)



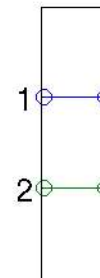
Tuning Curves (response vs. angle)



Responses (value vs. time)

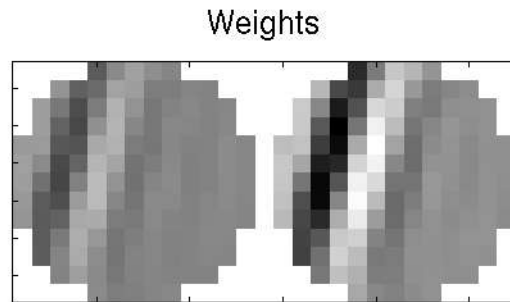


Channel Legend

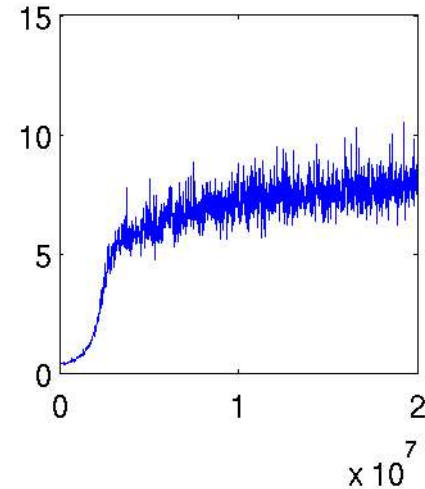


Cat with bias – MD, Low noise (TTX)

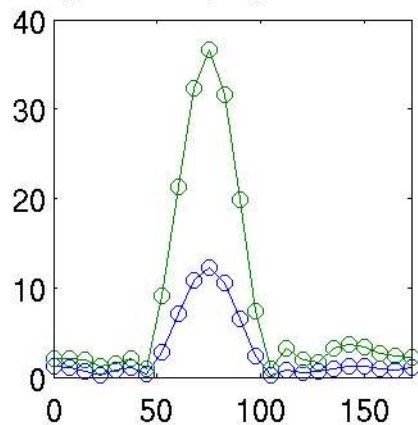
Gaussian SD = 0.01



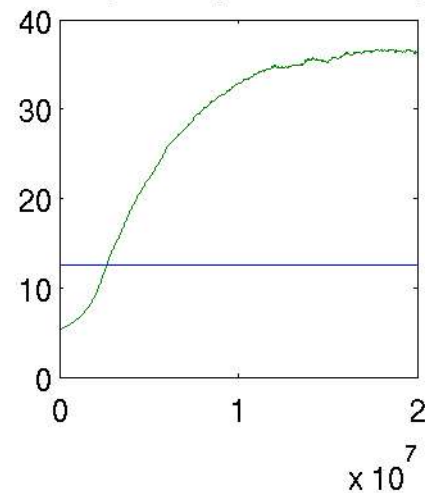
Thresholds (value vs. time)



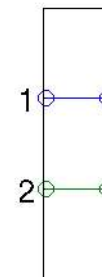
Tuning Curves (response vs. angle)



Responses (value vs. time)



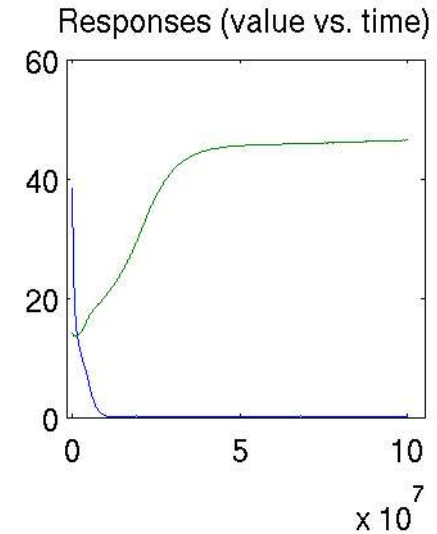
Channel Legend



MD noise dependency

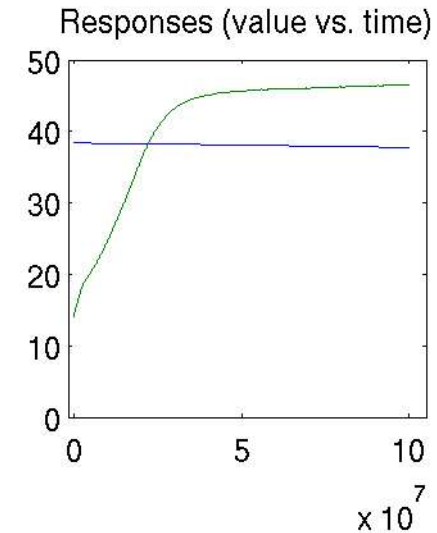
MD, High noise (lid suture):

- rapid response depression
- delayed response potentiation



MD, Low noise (TTX):

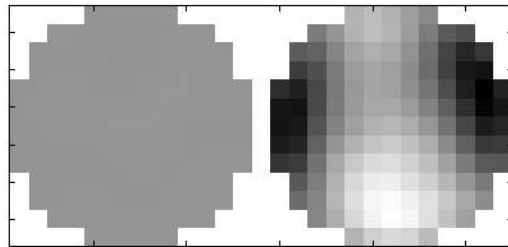
- little or no response depression
- rapid response potentiation



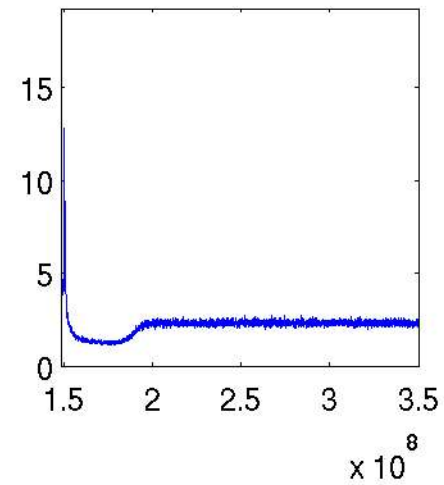
Mouse – MD, High noise (lid suture)

Gaussian SD = 0.8

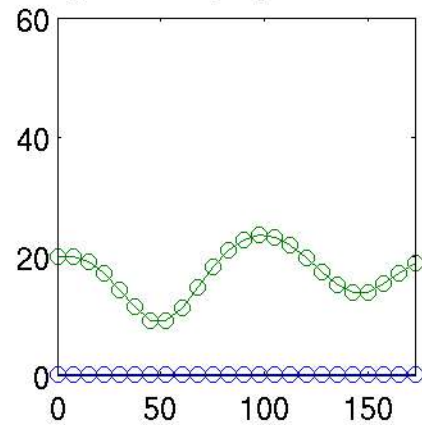
Weights



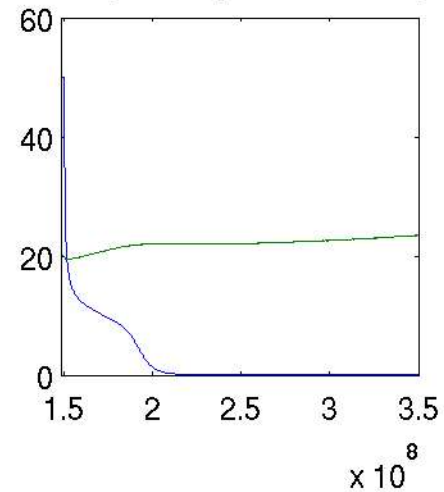
Thresholds (value vs. time)



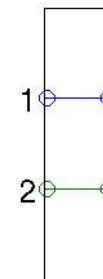
Tuning Curves (response vs. angle)



Responses (value vs. time)



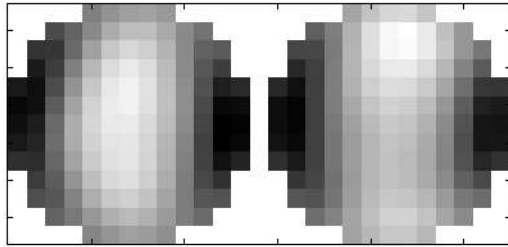
Channel Legend



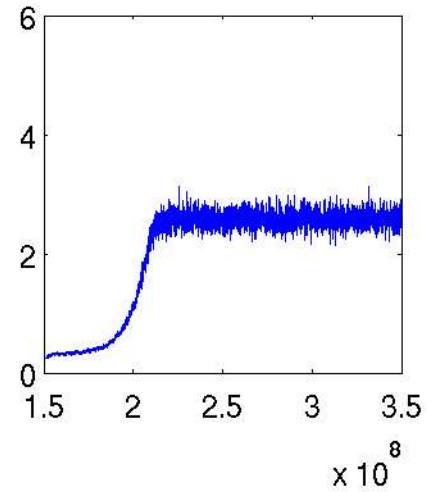
Mouse – MD, Low noise (TTX)

Gaussian SD = 0.01

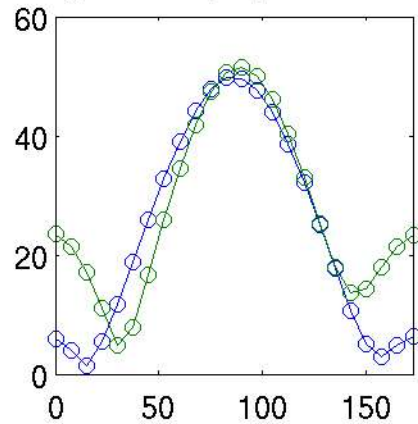
Weights



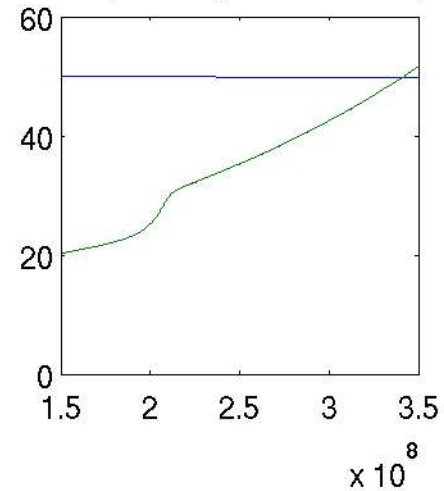
Thresholds (value vs. time)



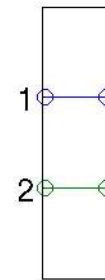
Tuning Curves (response vs. angle)



Responses (value vs. time)



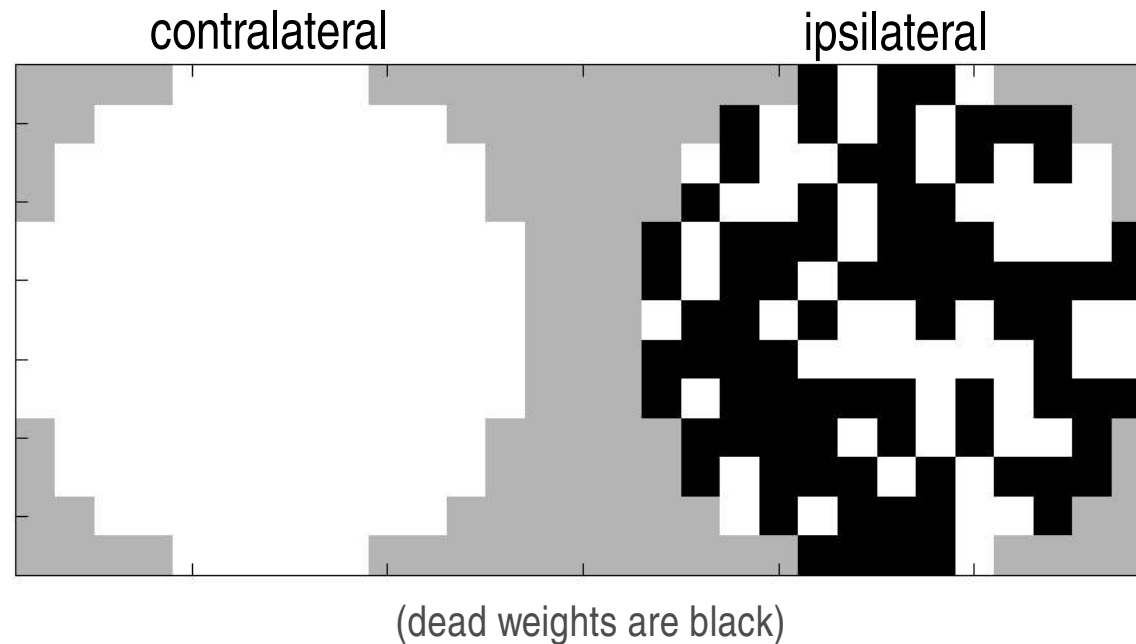
Channel Legend



Two methods for contralateral bias

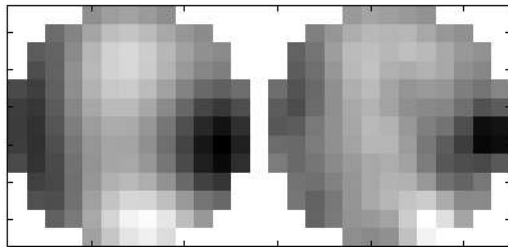
- Alter image intensity
- Reduce number of weights

The number of weights in the right (ipsilateral) channel was reduced by setting a “dead” weight (weight of value 0) every 2.5 weights.

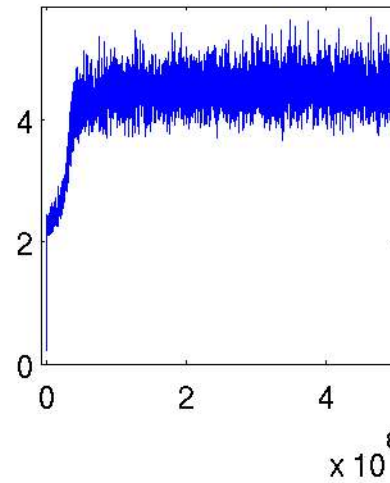


Mouse – NR, Dead weights induced bias

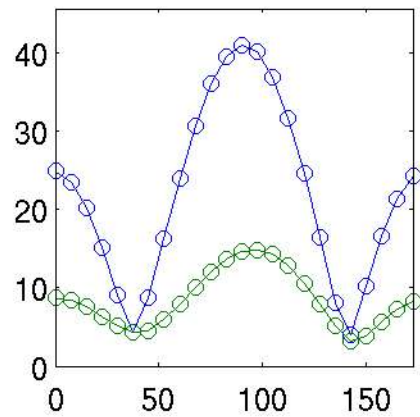
Weights



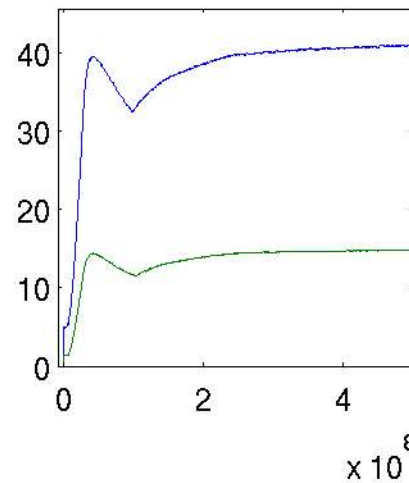
Thresholds (value vs. time)



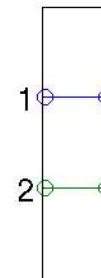
Tuning Curves (response vs. angle)



Responses (value vs. time)

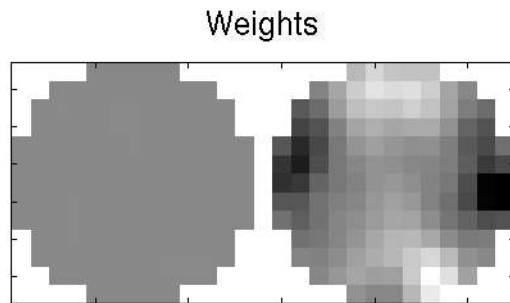


Channel Legend

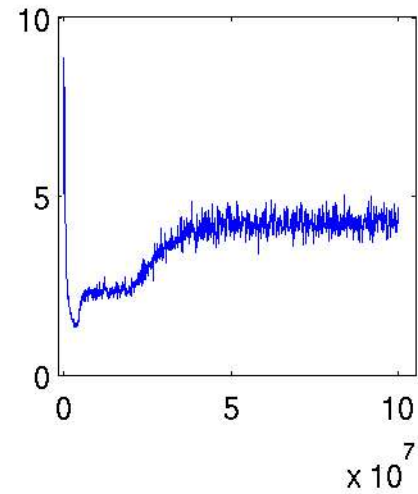


Mouse – MD, High noise (lid suture)

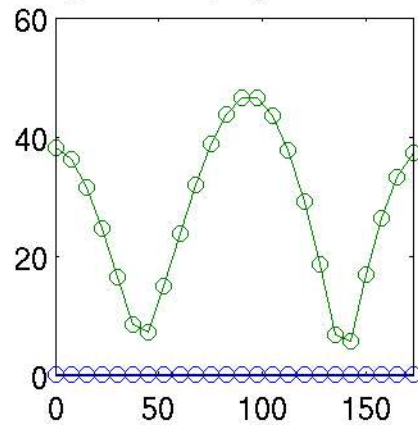
Gaussian SD = 0.8



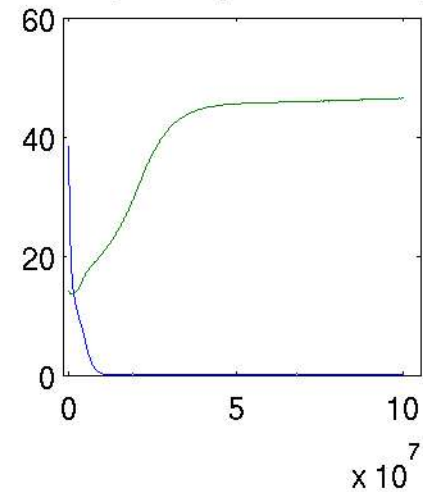
Thresholds (value vs. time)



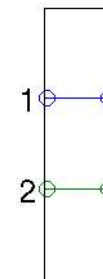
Tuning Curves (response vs. angle)



Responses (value vs. time)



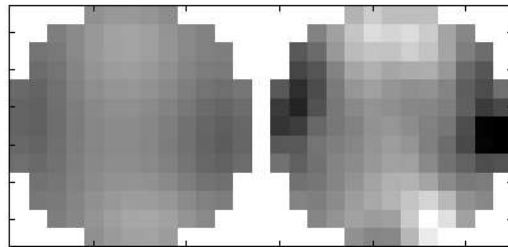
Channel Legend



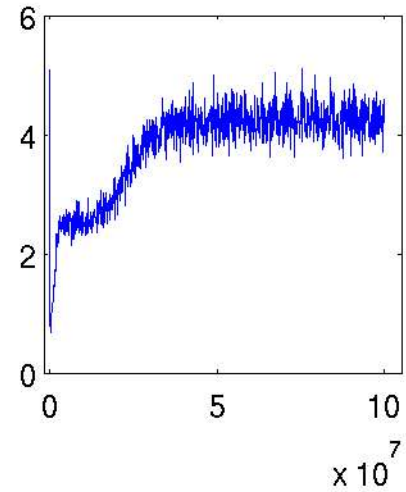
Mouse – MD, Low noise (TTX)

Gaussian SD = 0.01

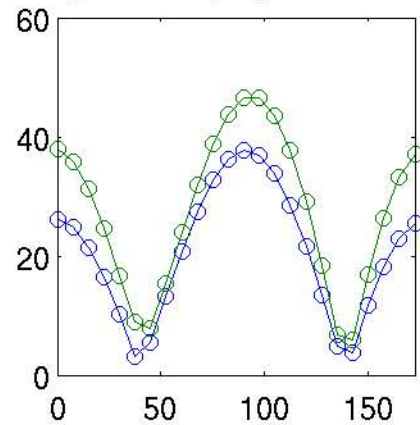
Weights



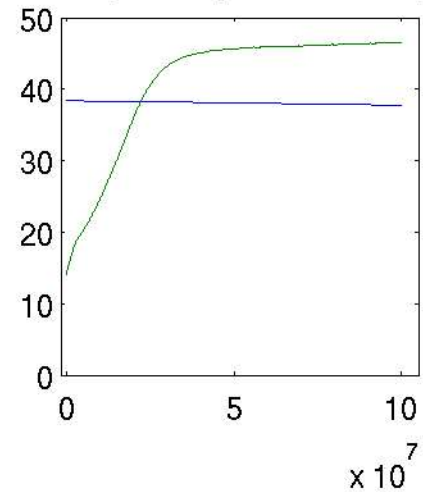
Thresholds (value vs. time)



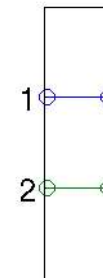
Tuning Curves (response vs. angle)



Responses (value vs. time)

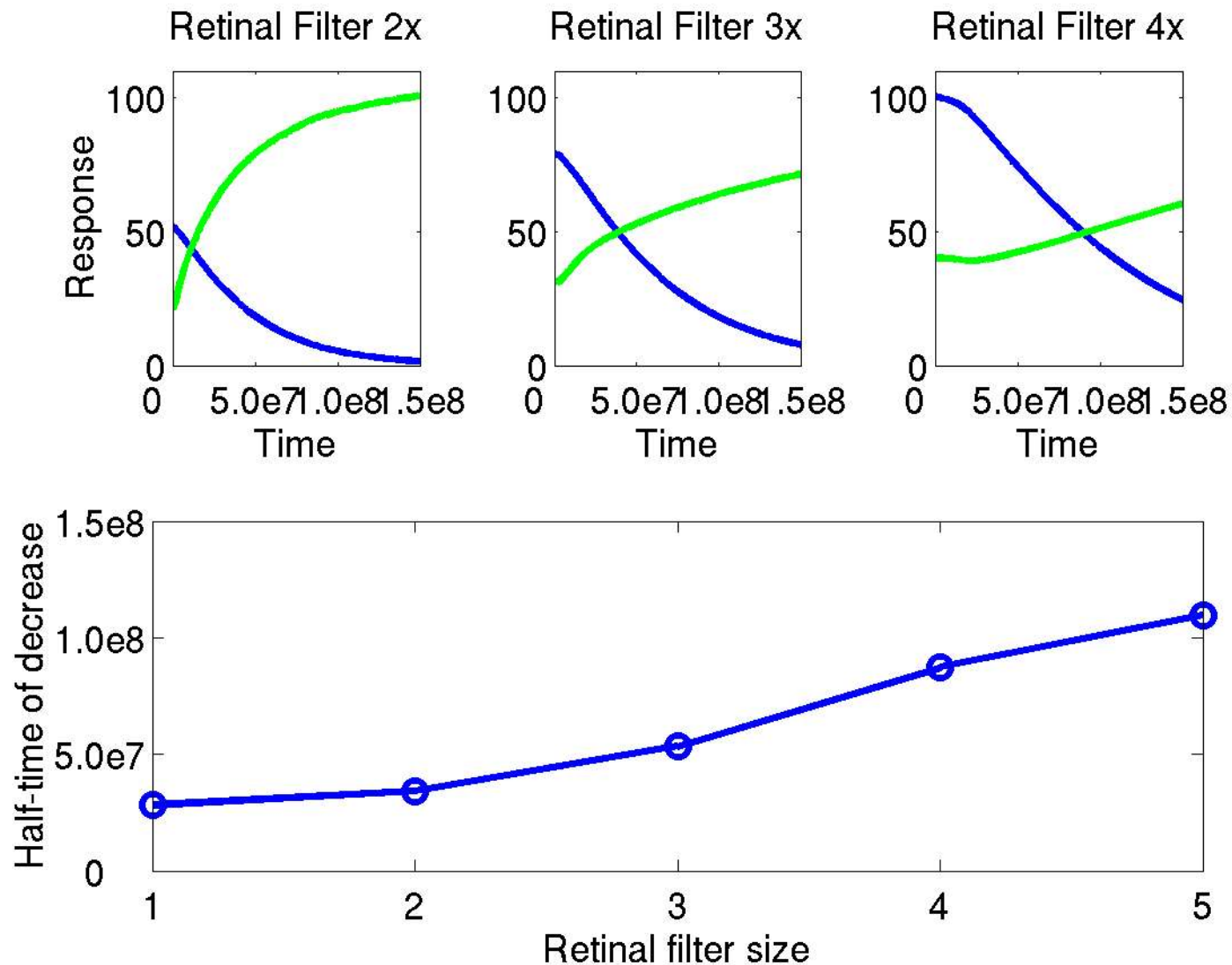


Channel Legend



Effects of retinal filter on MD dynamics

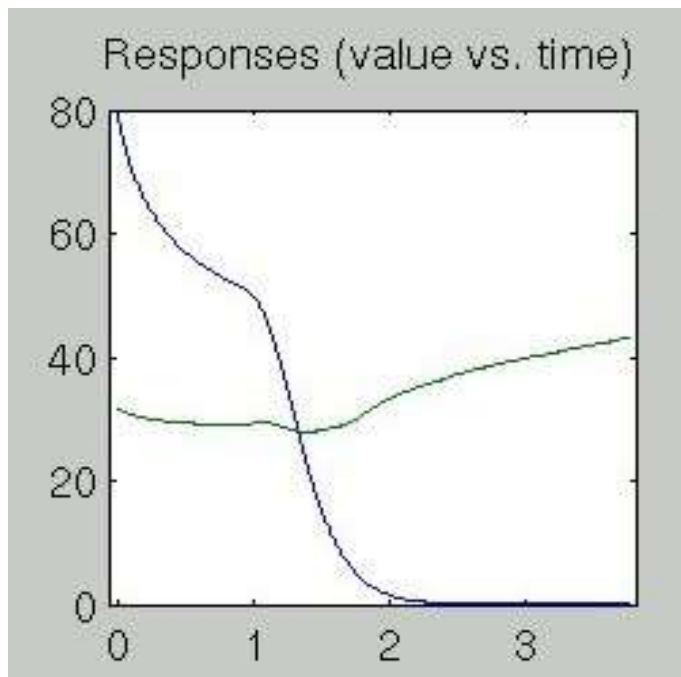
- Increasing retinal filter affects decay time similarly to decreasing noise level.



Mouse MD is similar to Cat RS

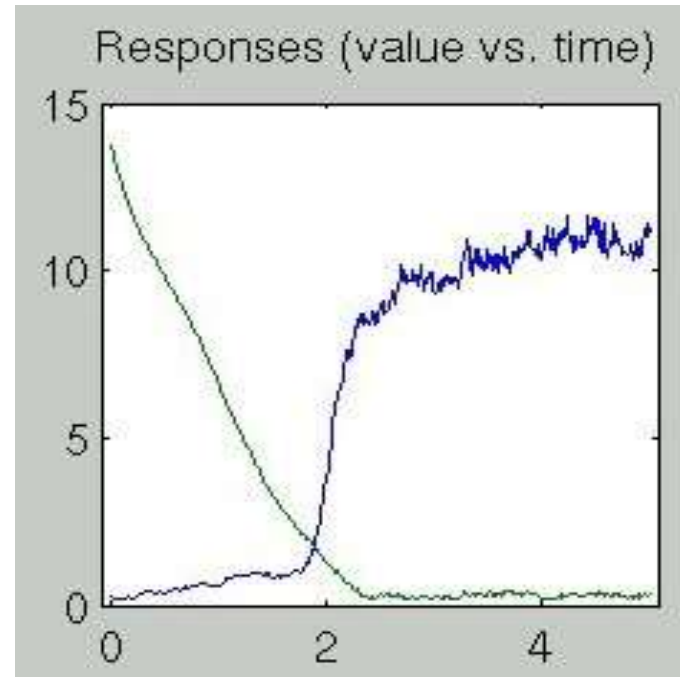
- Pattern input into weaker eye
- Noise input into stronger eye

Mouse MD



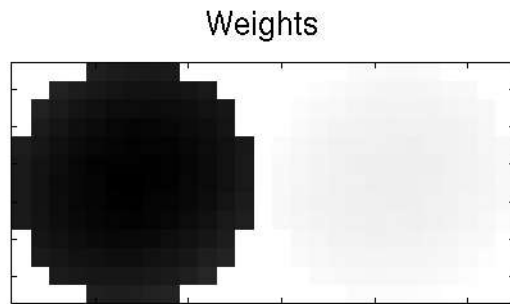
(intensity bias, DoG 3:9)

Cat RS

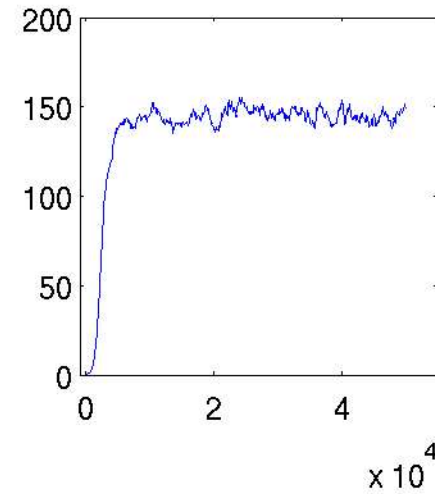


(no bias, DoG 1:3)

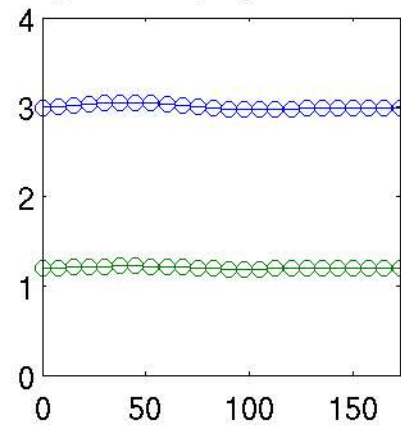
Mouse – NR, Hebbian Learning



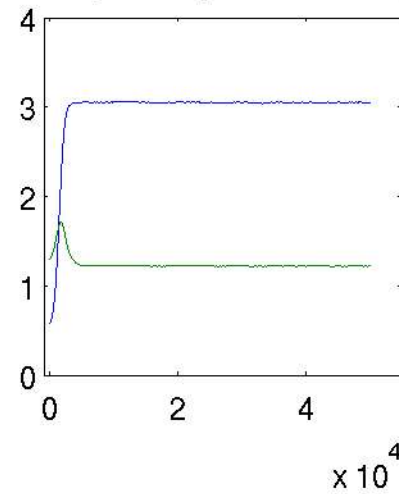
Thresholds (value vs. time)



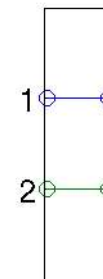
Tuning Curves (response vs. angle)



Responses (value vs. time)

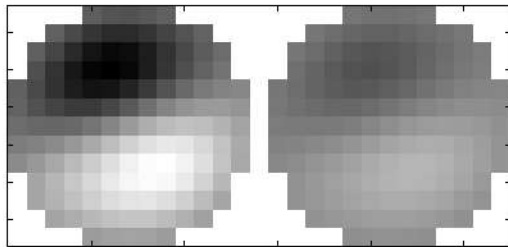


Channel Legend

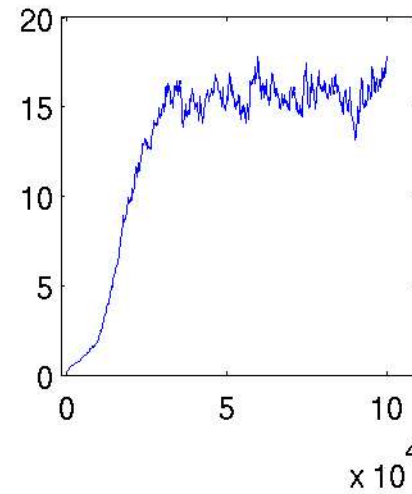


Cat with bias – NR, Hebbian Learning

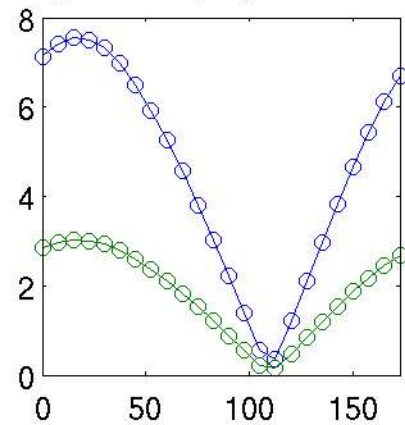
Weights



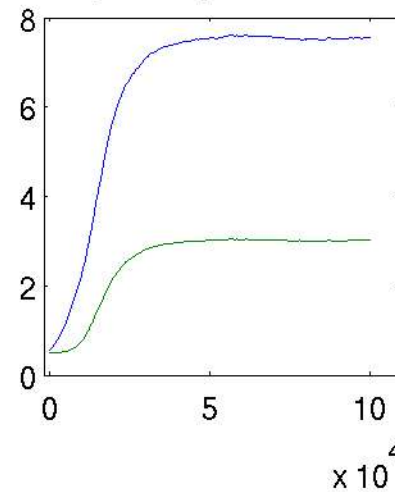
Thresholds (value vs. time)



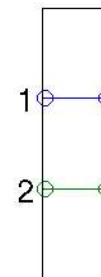
Tuning Curves (response vs. angle)



Responses (value vs. time)

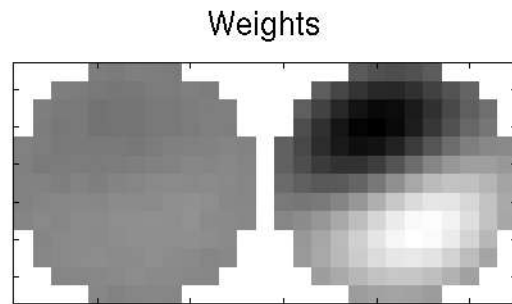


Channel Legend

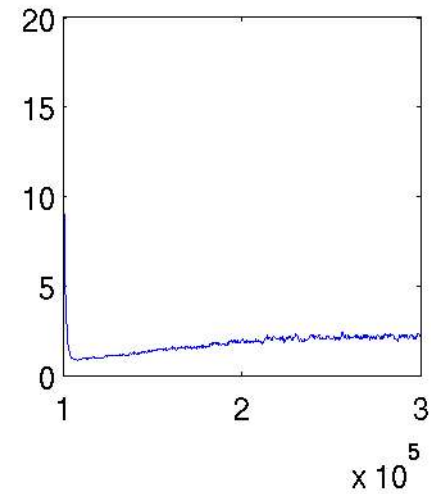


Cat with bias – MD, High noise, Hebbian Learning

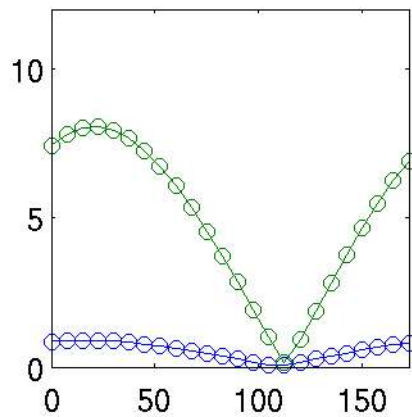
Gaussian SD = 0.8



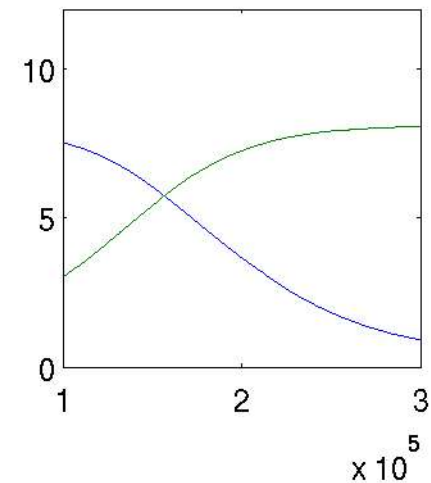
Thresholds (value vs. time)



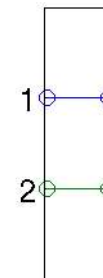
Tuning Curves (response vs. angle)



Responses (value vs. time)

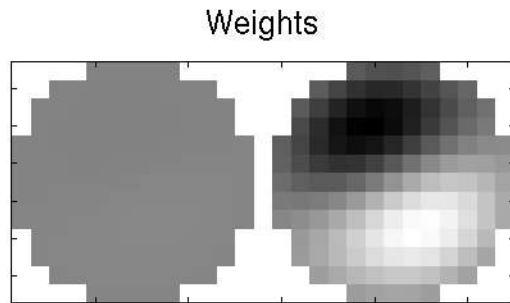


Channel Legend

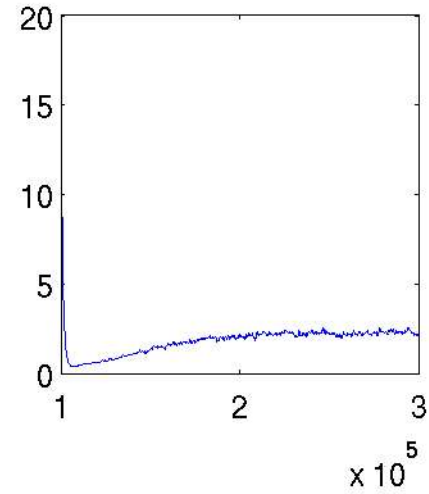


Cat with bias – MD, Low noise, Hebbian Learning

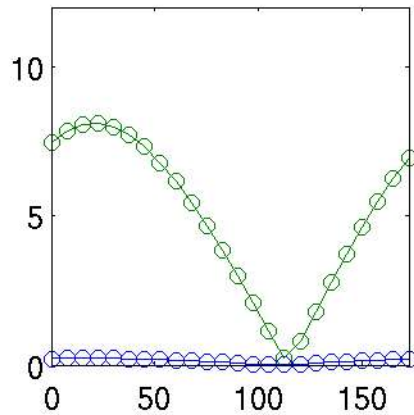
Gaussian SD = 0.01



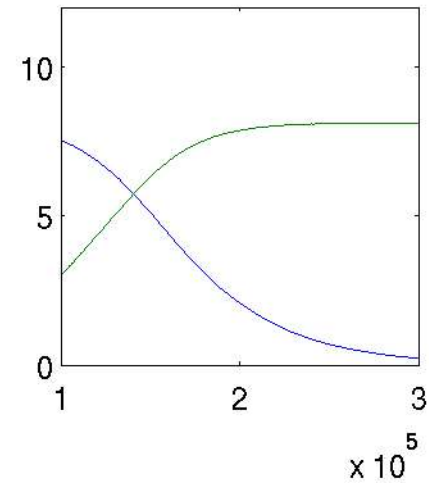
Thresholds (value vs. time)



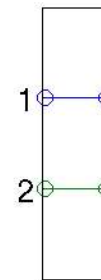
Tuning Curves (response vs. angle)



Responses (value vs. time)



Channel Legend

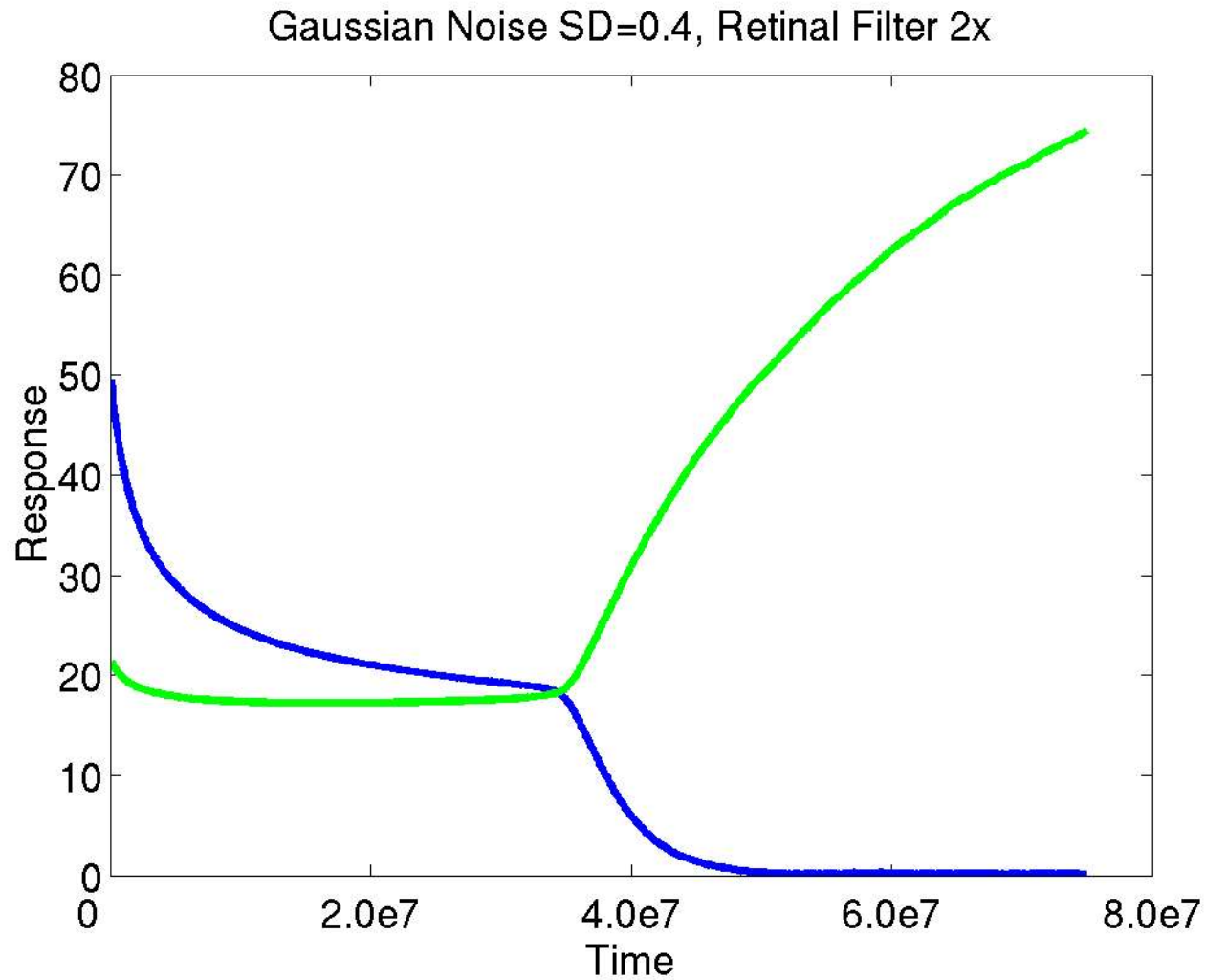


Some current problems with the BCM simulations

- Problem 1: How do we calculate the decay time?
- Problem 2: Dead weights bias instability

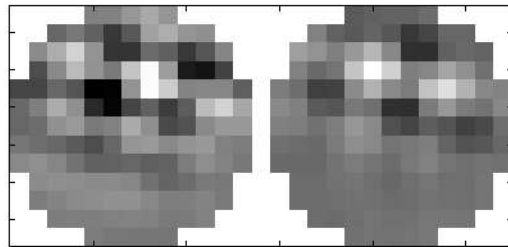
Problem 1: How do we calculate the decay time?

Mouse – NR, BCM

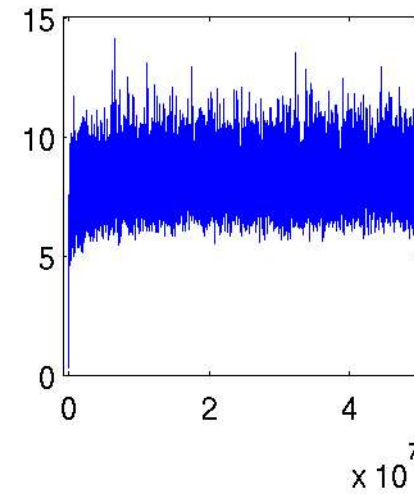


Problem 2: Dead weights instability: Cat with bias – NR

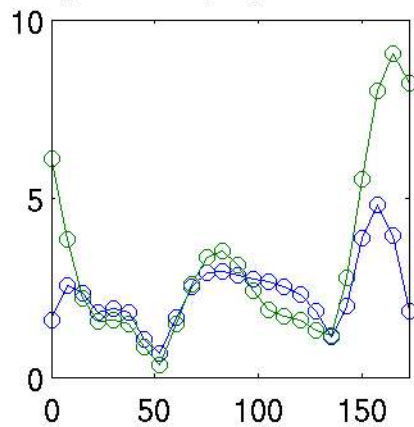
Weights



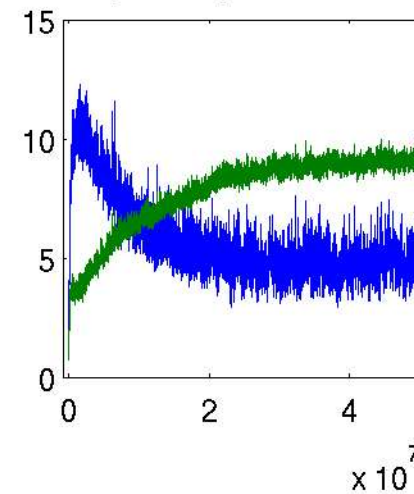
Thresholds (value vs. time)



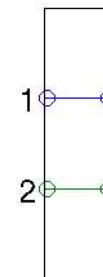
Tuning Curves (response vs. angle)



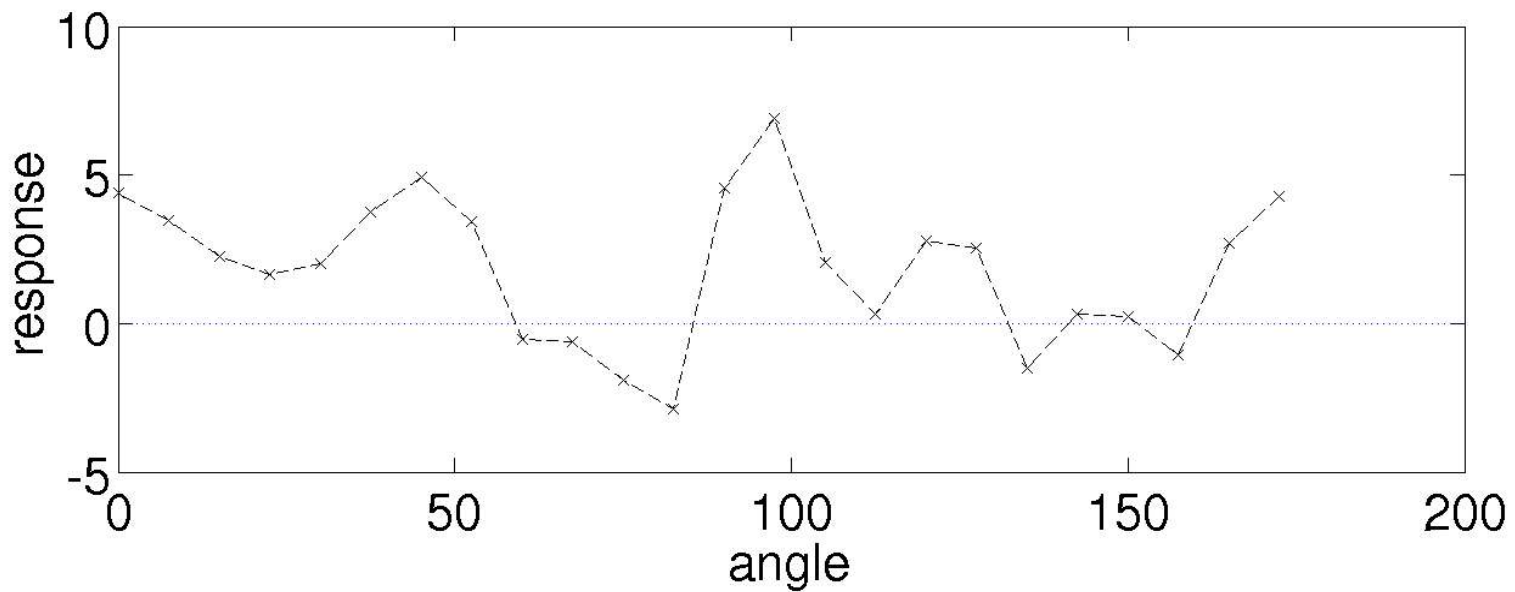
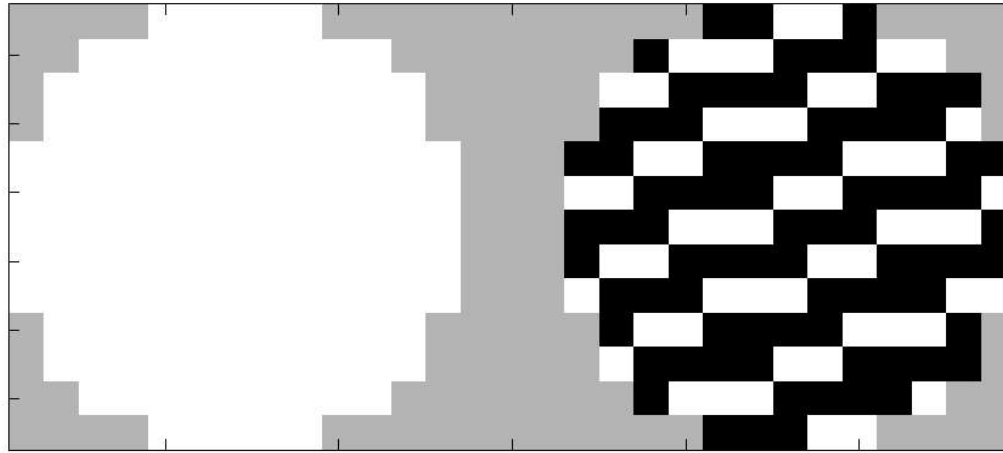
Responses (value vs. time)



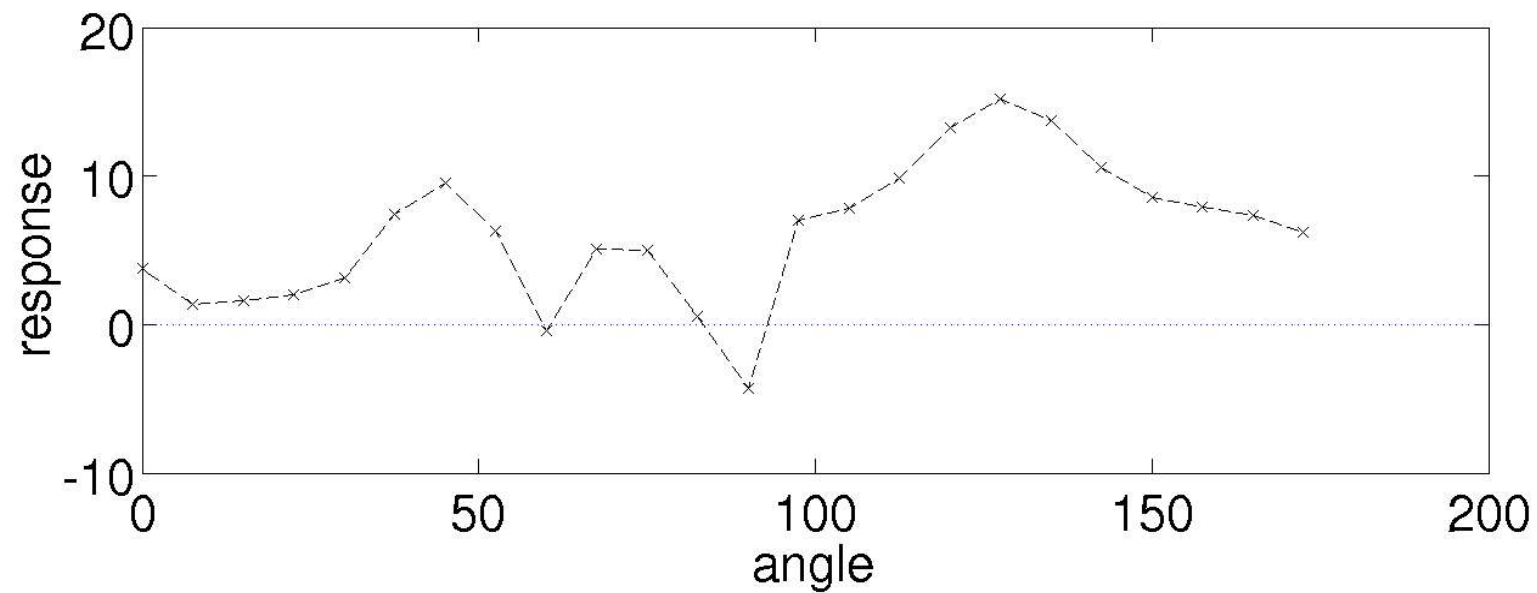
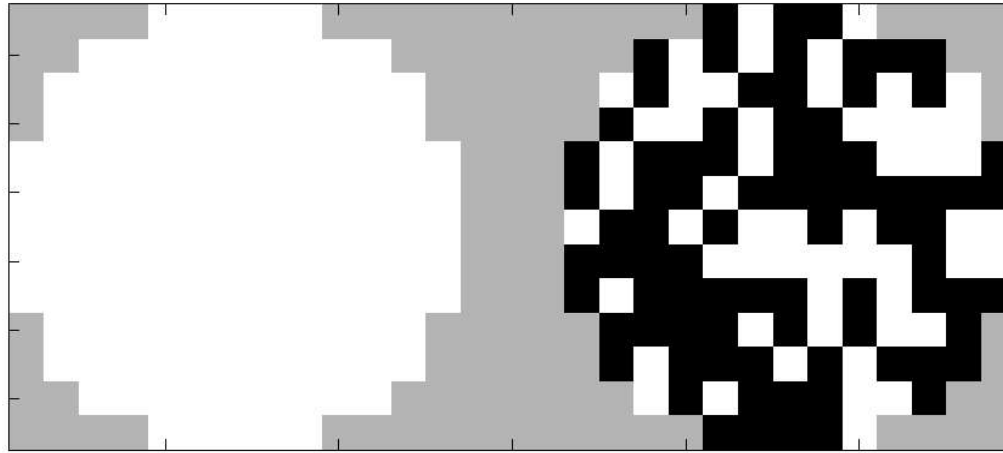
Channel Legend



Dead weights method: Orientation problem



Dead weights method: Orientation problem



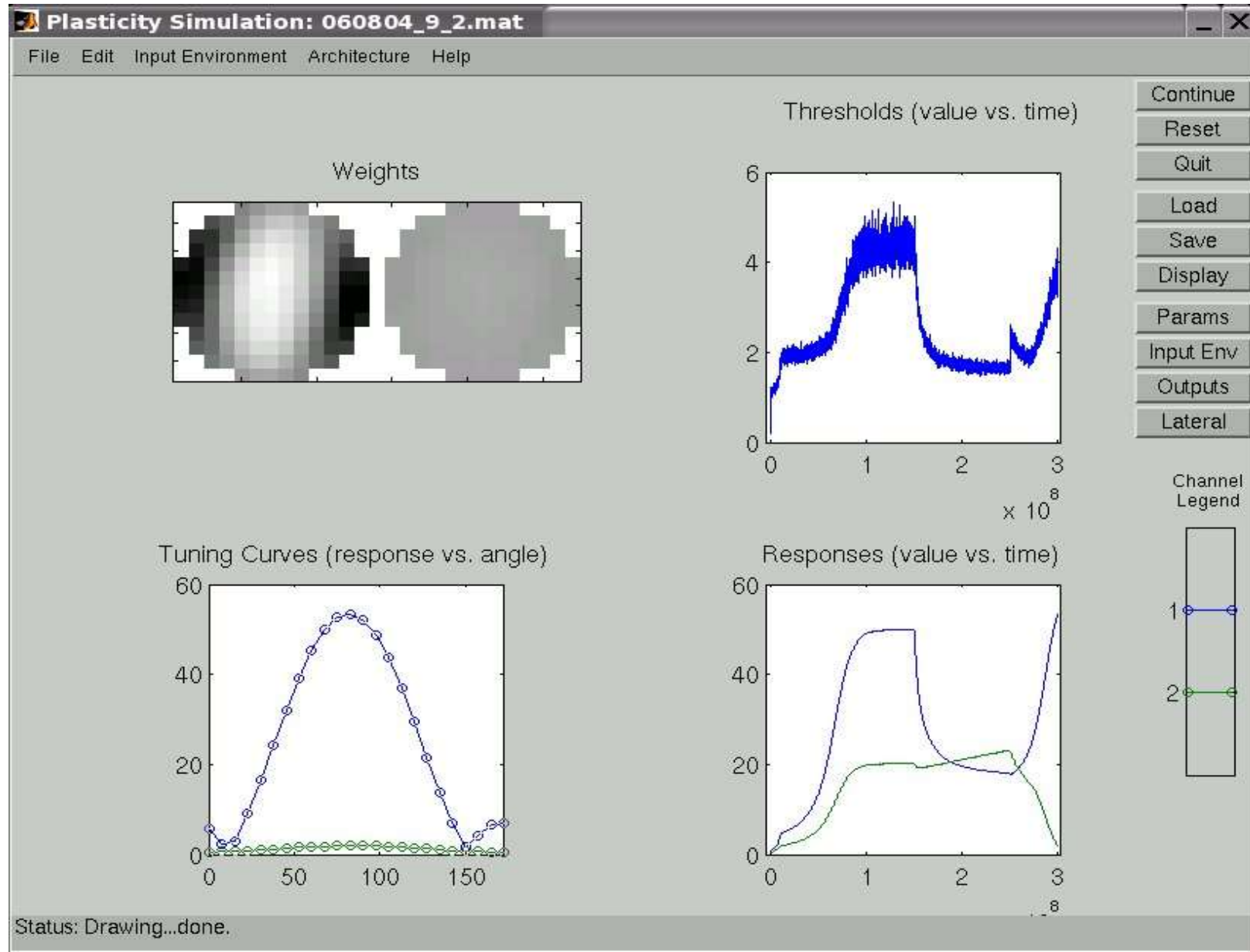
Conclusions

- Contralateral bias can effectively be implemented two ways:
 - Decreased image intensity
 - Decreased number of inputs
- Reproduce the results of Frenkel/Bear with BCM
 - Lid suture
 - Rapid response depression
 - Delayed response potentiation
 - TTX
 - Little or no response depression
 - Rapid response potentiation
- Cannot reproduce results using Hebbian or Kurtosis learning rules

Additional Slides

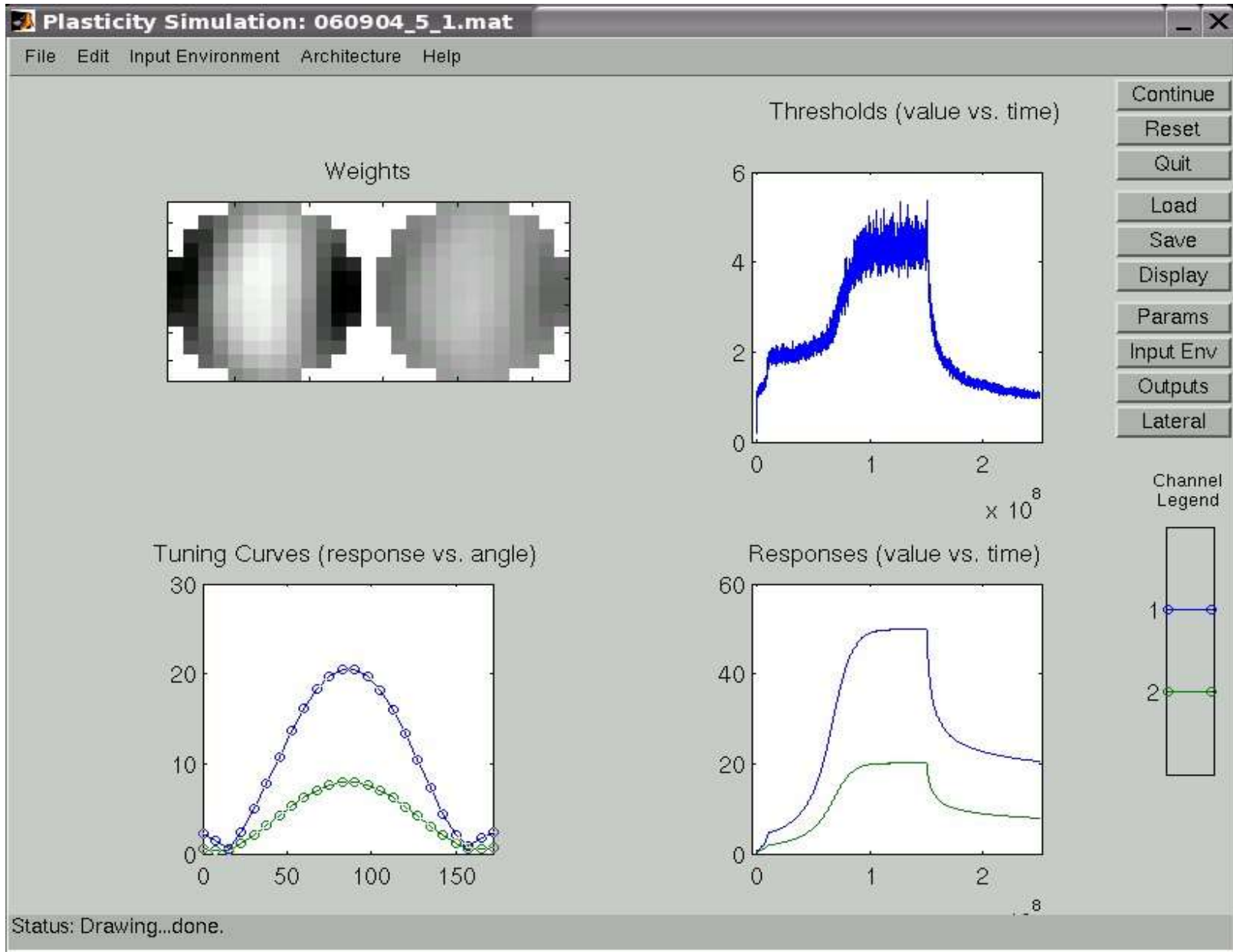
Reverse Suture

$$SD = 0.4$$



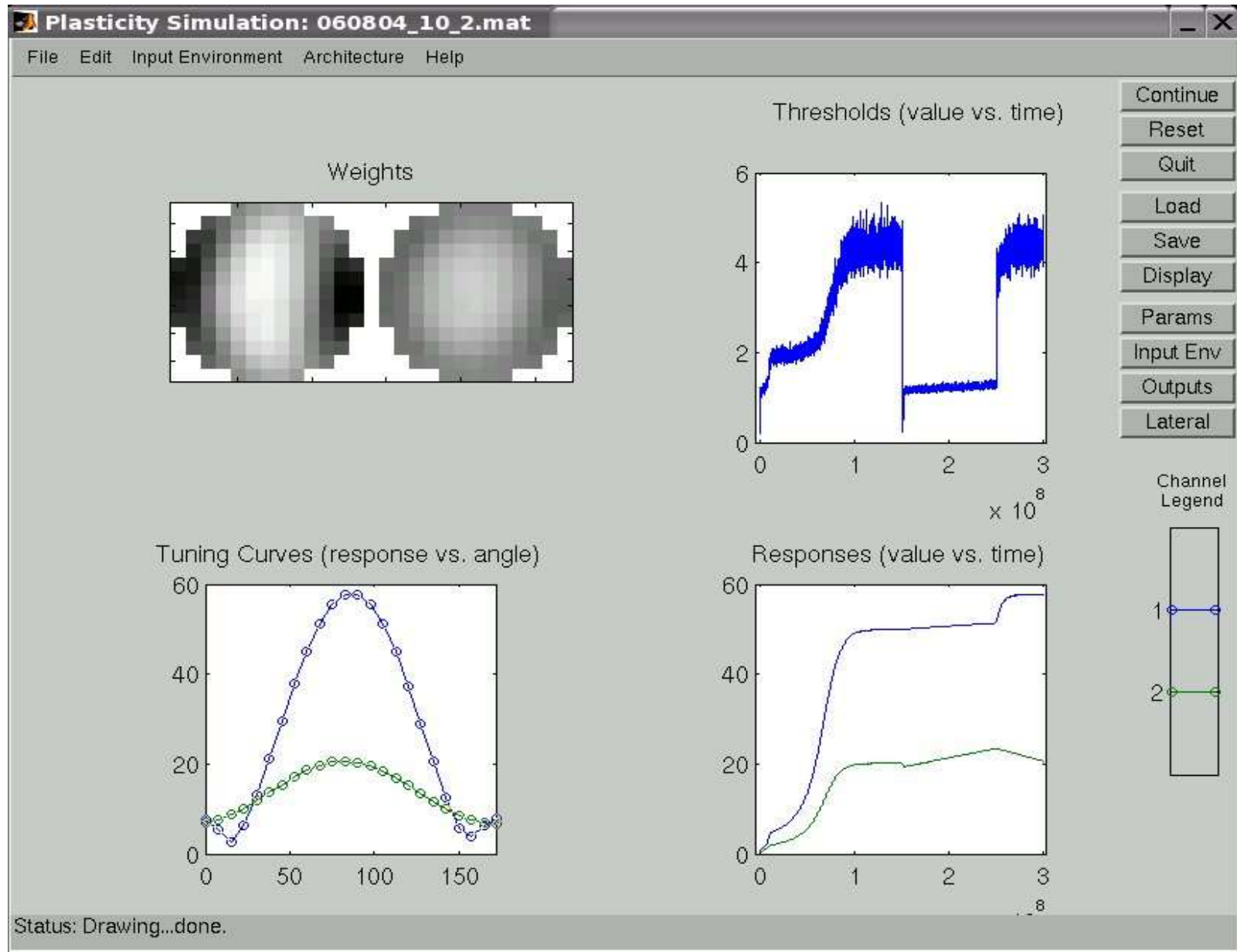
Binocular Deprivation

$$SD = 0.4$$



Reverse Suture

$SD = 0.05$



Binocular Deprivation

SD = 0.05

