Understanding impaired learning with enhanced plasticity

based on work in preparation with: T.D. Barbara Nguyen-Vu, Grace Q. Zhao, Han-Mi Lee, Surya Ganguli, Carla J. Shatz, Jennifer L. Raymond

Subhaneil Lahiri

Stanford University, Applied Physics

July 19, 2013

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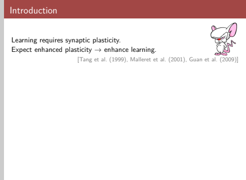
1. Acknowledge Barbara and Grace

Introduction

Learning requires synaptic plasticity.
Expect enhanced plasticity $\rightarrow$ enhance learning.

[Tang et al. (1999), Malleret et al. (2001), Guan et al. (2009)]



1. It does help in some cases
2. Want to understand when and why
3. Depends on circumstance. Rich pattern of behaviour
4. Develop understanding of when and why learning is enhanced/impaired

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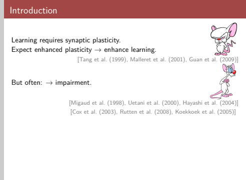
But often: → impairment.

[Migaud et al. (1998), Uetani et al. (2000), Hayashi et al. (2004)]
[Cox et al. (2003), Rutten et al. (2008), Koekkoek et al. (2005)]



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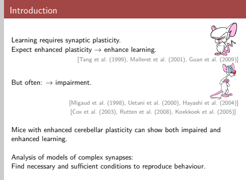
Mice with enhanced cerebellar plasticity can show both impaired and enhanced learning.

Analysis of models of complex synapses:
Find necessary and sufficient conditions to reproduce behaviour.



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Outline

- Cerebellar learning of mice with enhanced plasticity
- Complex synaptic models

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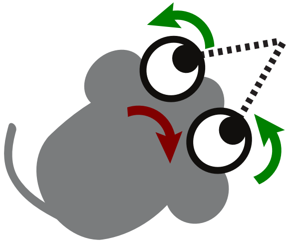
Vestibulo-Occular Reflex

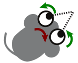


Eye movements compensate for head movements to maintain fixation.

Requires control of VOR gain $= \frac{\text{eye velocity}}{\text{head velocity}}$.

Needs to be adjusted as eye muscles age, etc.

Vestibulo-Occular Reflex



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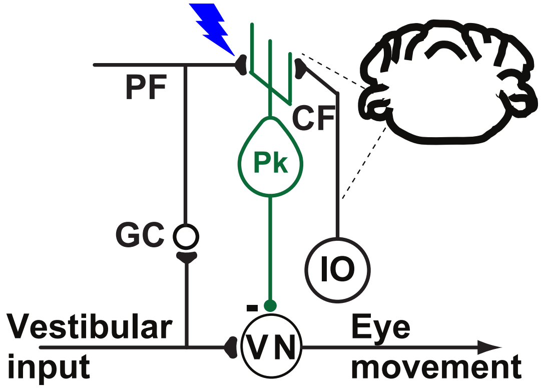
VOR training

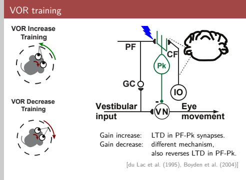
VOR Increase Training

VOR Decrease Training



Gain increase:	LTD in PF-Pk synapses.
Gain decrease:	different mechanism, also reverses LTD in PF-Pk.

[du Lac et al. (1995), Boyden et al. (2004)]

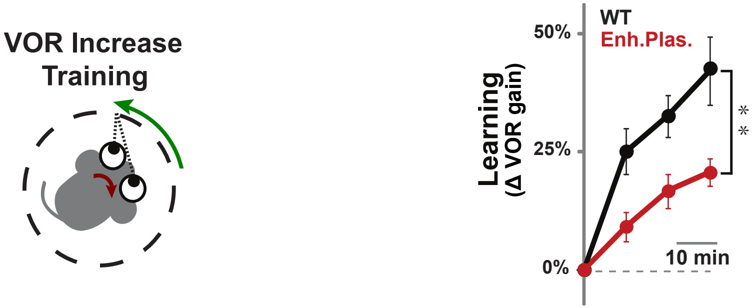


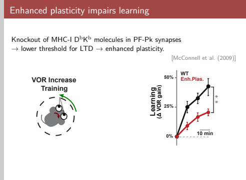
1. trick brain into thinking VOR gain needs adjusting my moving visual stimulu
2. anti-phase → increase gain
3. in phase → decrease gain
4. Gain change involves cerebellum
5. If we enhanced plasticity here: expect enhanced learning

Enhanced plasticity impairs learning

Knockout of MHC-I D^bK^b molecules in PF-Pk synapses
$\rightarrow$ lower threshold for LTD $\rightarrow$ enhanced plasticity.

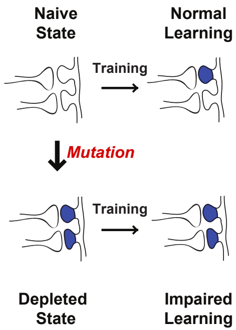
[McConnell et al. (2009)]

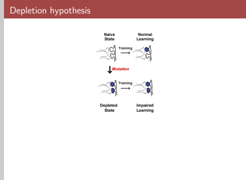




1. Major Histocompatibility Complex - involved in synaptic plasticity (Carla Shatz lab)
2. Easier LTD $\rightarrow$ expect better learning
3. Impairment of learning
4. Looking at change of VOR gain during gain-up training

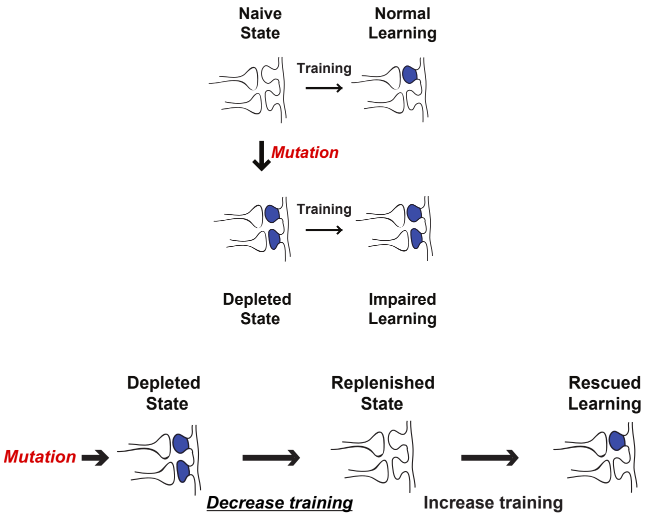
Depletion hypothesis

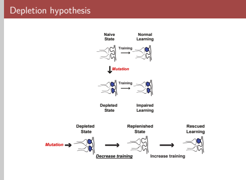




1. Our model: baseline activity → saturation → less depression possible
2. Saturation has to compete with enhanced plsaticity. Which will win?

Depletion hypothesis





1. Our model: baseline activity → saturation → less depression possible
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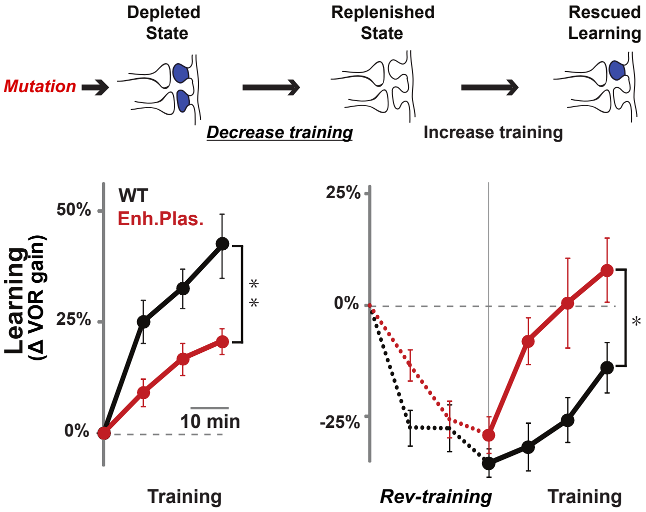
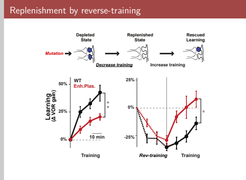
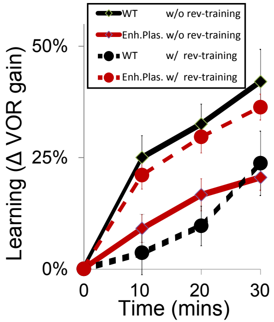
3. Prediction: replenish with rev-training → rescue

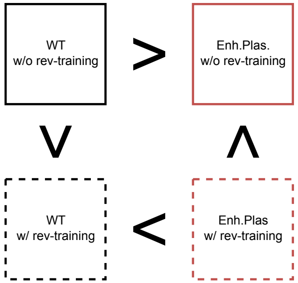
Replenishment by reverse-training





1. precede gain inc training w/ gain dec rev-training: reverses LTD
2. but behaviour from elsewhere → not modelled
3. Focus on gain inc part





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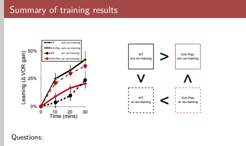

1. Restricted to gain inc for comparison
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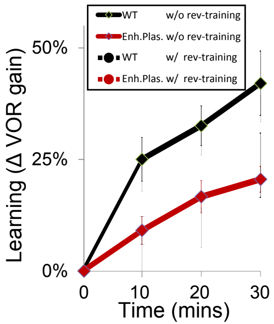
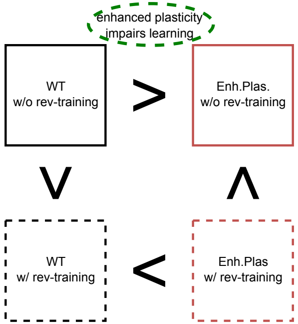
3. Solid: no pre. Dashed: with pre
4. Diagonal comparisons: parameter fitting. Depend on size of KO vs. pretraining

Summary of training results





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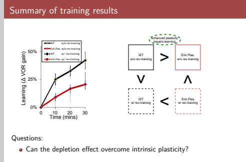
- Can the depletion effect overcome intrinsic plasticity?



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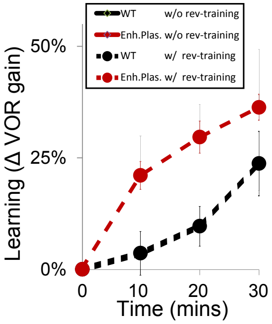

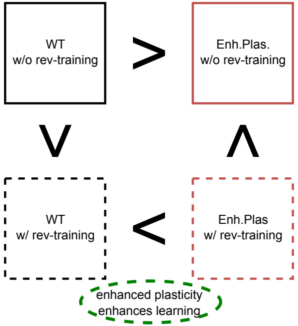

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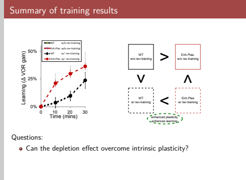

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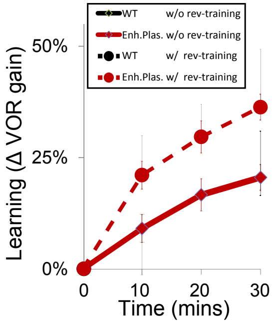

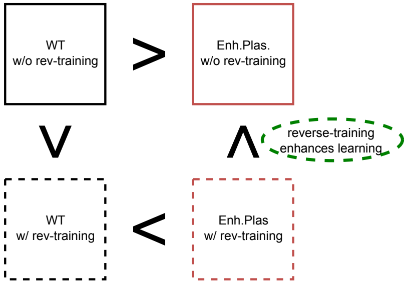

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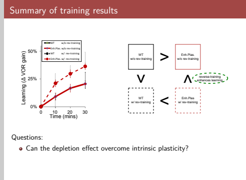

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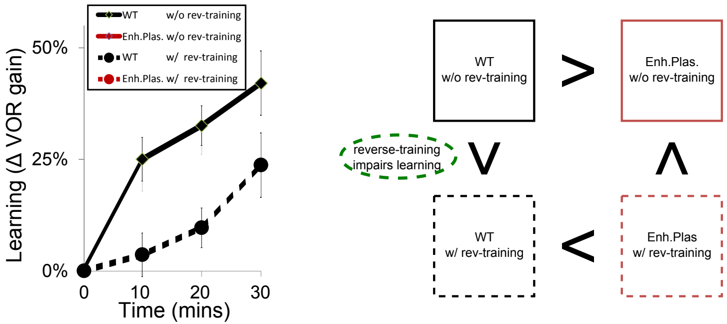

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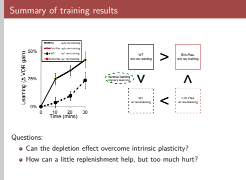

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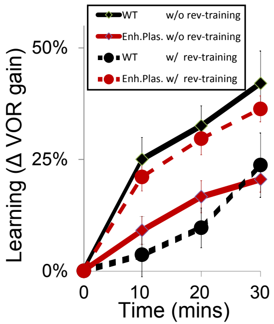

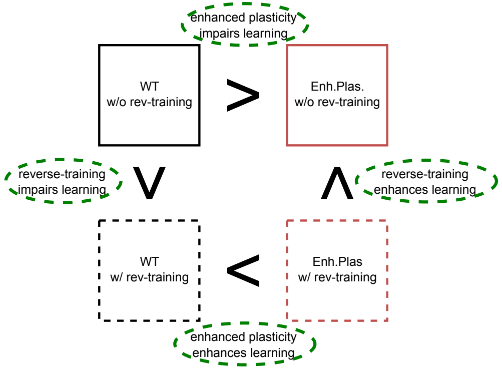

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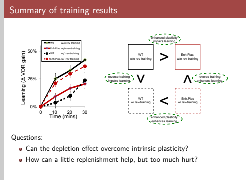

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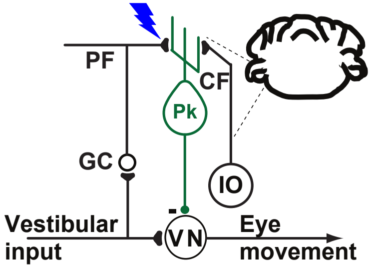
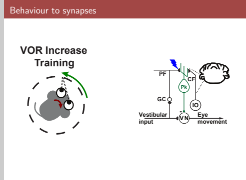
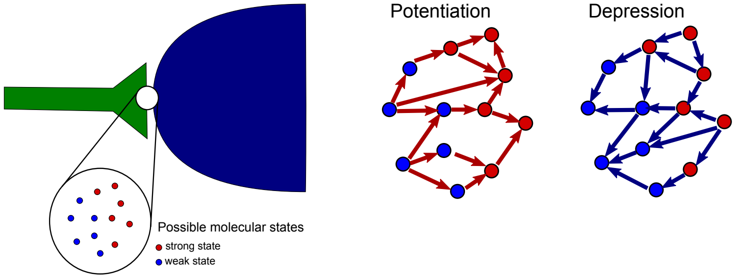
Behaviour to synapses





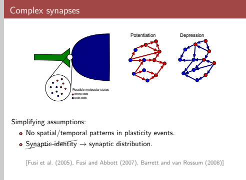
1. Focus on synapses. See if we can understand this behaviour.



Simplifying assumptions:

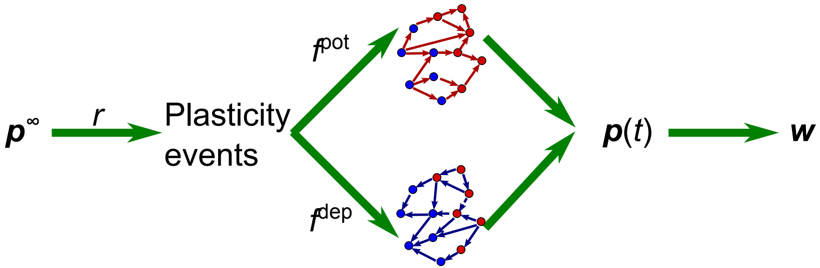
- No spatial/temporal patterns in plasticity events.
- ~~Synaptic identity~~ → synaptic distribution.

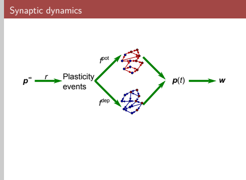
[Fusi et al. (2005), Fusi and Abbott (2007), Barrett and van Rossum (2008)]



1. Not just synaptic weight, internal dynamical system
2. Important for memory: simple synapses – terrible storage, rescued by complexity
3. Multiple functional states w/ different weights
4. Stochastic transitions between states
5. pot/dep occur randomly
6. allows us to concentrate on synapse, not neuron/network
7. This is a question about synaptic populations after all.

Synaptic dynamics





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2. Prior activity puts it in this state. row vec.
3. plasticity events at rate r
4. fraction pot/dep
5. probs changed by Markov matrices, prob $i \to j$
6. Readout: synaptic weight vec when in each state.

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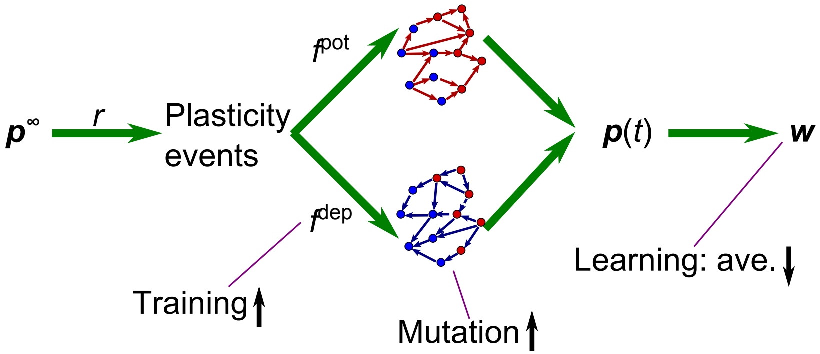

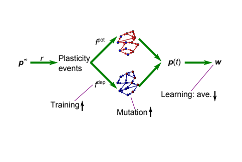

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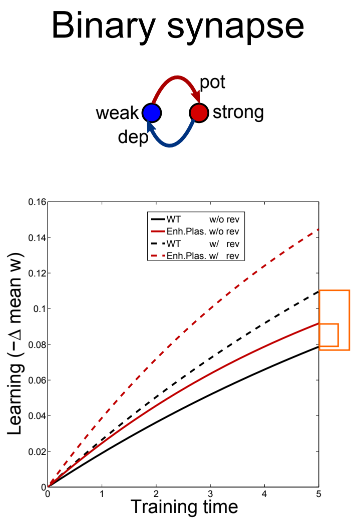
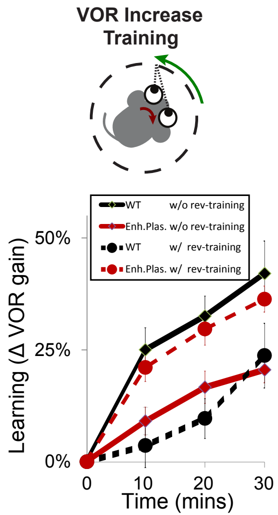
6. Readout: synaptic weight vec when in each state.
7. Mutation: lower threshold → increase transition probs
8. Training: Changes statistics of LTP/LTD. Only parameters we have. Don't care about r.
9. Learning: Only output we have. Don't keep track of synaptic identity.
10. Same PF+CF input → same r, f^{pot}, f^{dep} in each case.
11. Input to Pk, some linear combination of **w**'s.

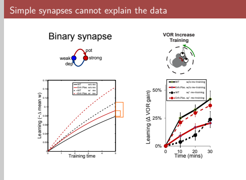




2013-07-19

Impaired learning with enhanced plasticity

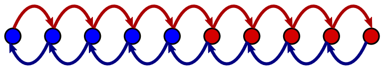
└Simple synapses cannot explain the data

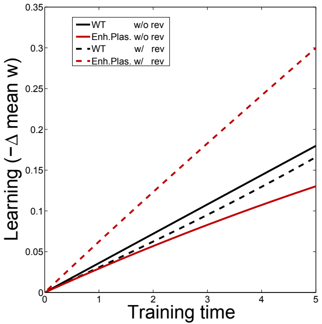


1. Binary fails – mathematical proof for any params
2. Enh.Plas: faster depression wins over bias
3. pre: reduces/reverses bias. always helps.

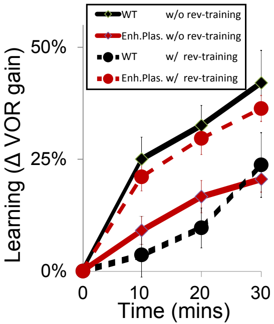
Complex synapses can explain the data

Serial synapse

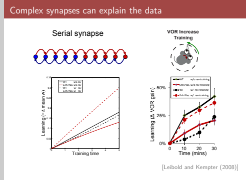




VOR Increase Training

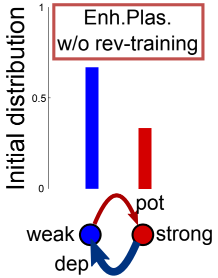


[Leibold and Kempter (2008)]

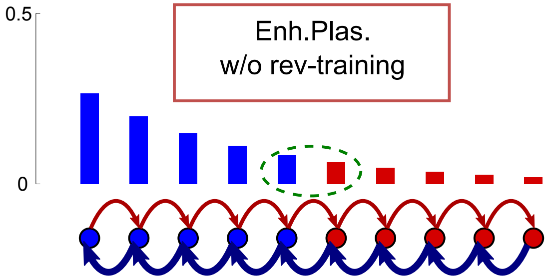


1. Serial: still only two weights. Works.
2. Understand by looking at distributions before training

Enhanced plasticity can enhance or impair learning



Intrinsic plasticity
dominates depletion



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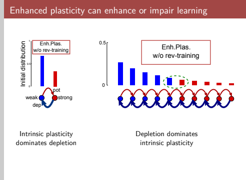
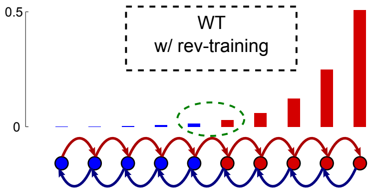


1. Binary: enhanced plasticity → bias
2. Not enough to overcome faster depression
3. Serial: Only get signal from boundary
4. Exponential decay depopulates boundary, enhances effect of bias

Reverse-training can impair or enhance learning



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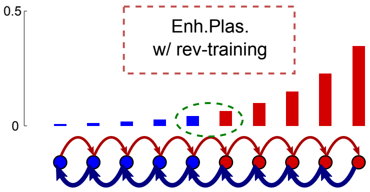

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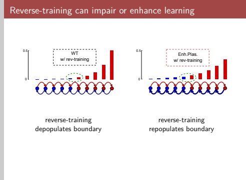
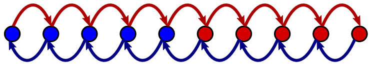


1. rev-training: little repopulates boundary
2. Too much pushes to other side, depopulates boundary
3. this effect is absent in any simple synapse

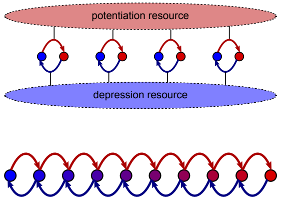
Essential features



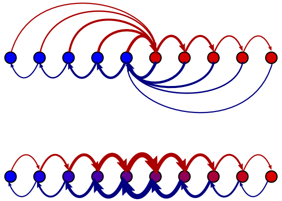
The success of the serial model relies on two features:

- Complexity - needed to amplify the effect of depletion,
- Metaplasticity – repeated potentiation makes subsequent depression harder.

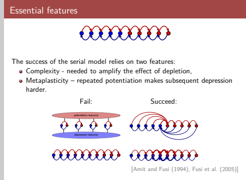
Fail:



Succeed:



[Amit and Fusi (1994), Fusi et al. (2005)]



1. due to exponential decay
2. push away from boundary where signal generated
3. borne out by other models that fail/succeed

Conclusions and further questions

- We find diverse behavioural patterns:
  Enhanced plasticity $\rightarrow$ enhance/impair learning depending on prior experience.
  Reverse-training $\rightarrow$ enhance/impair learning depending on plasticity rates.
- We can explain these behavioural patterns using synaptic models.
- Key required synaptic properties are:
  Synaptic complexity: necessary to amplify depletion.
  Synaptic stubbornness: repeated potentiation makes subsequent depression harder.
- We used behaviour to constrain the dynamics of synaptic plasticity

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What about memory?

- Simple synapses have poor memory storage capacity.
  Synaptic complexity is needed for rescue.

- Trade-off between learning and remembering:
  Too rigid $\to$ difficult to learn new memories.
  Too plastic $\to$ new memories quickly overwrite old.

- Exploring the *entire* space of complex synaptic models
  $\to$ upper bounds on their storage ability
  & the models that saturate them.

[Lahiri and Ganguli (submitted)]

Tradeoff: learning vs. remembering

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