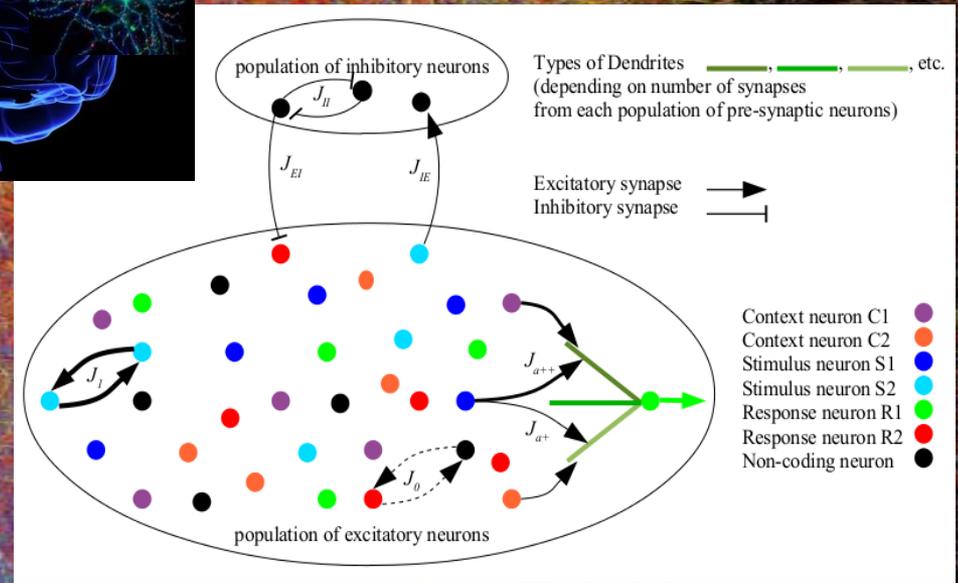


# Learning and processing of non-linearly separable combinations: from dendrites to behavior

Frédéric Lavigne, Francis Avnaïm and Laurent Dumercy



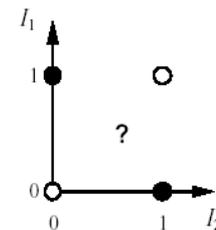
# On-line processing of XOR combinations



<u>Context:</u> stroke	<u>Stimulus:</u> racket forehand side	<u>Response:</u> Backspin or topspin rotation of the ball
forehand	upward	slice
forehand	downward	lift
backhand	upward	lift
backhand	downward	slice

responses cannot be selected based on any single stimulus, but only based on their combinations

Input1	input2	ouput
0	0	0
0	1	1
1	0	1
1	1	0



(c)  $I_1 \text{ xor } I_2$

How non-linearly separable problems are learned by the brain?

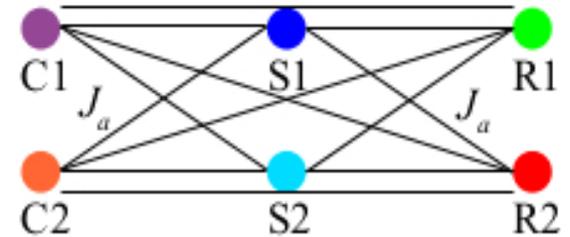
# Failure of hebbian learning of XOR combinations

context	stimulus	→	response
C1	S1	→	R1
C1	S2	→	R2
C2	S1	→	R2
C2	S2	→	R1

Triadic XOR combinations

$$\Rightarrow a_{ij} = q^+ \xi_i \xi_j \Rightarrow$$

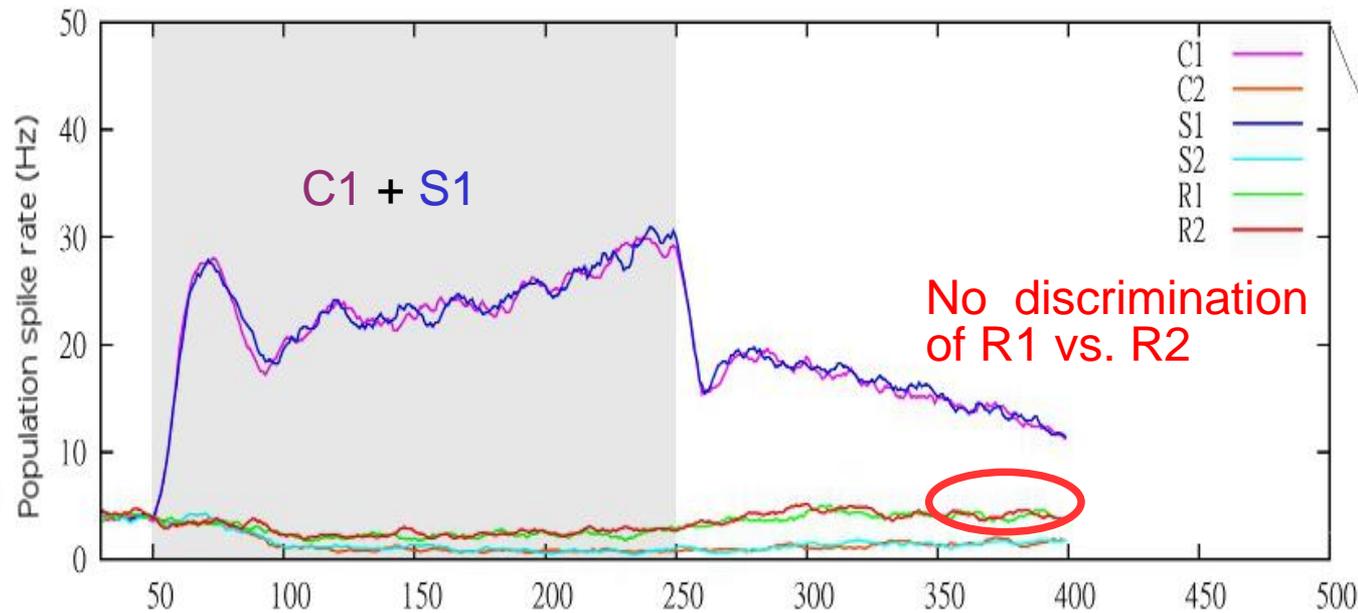
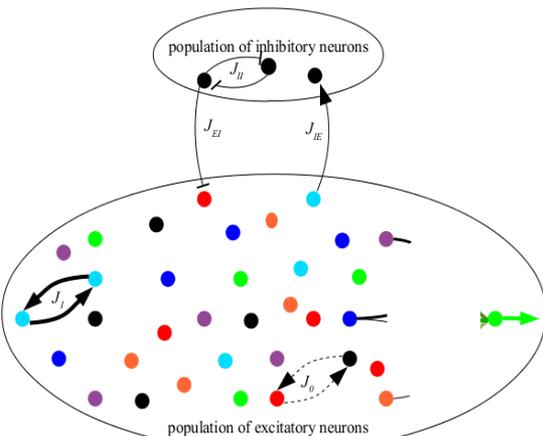
Local Hebbian learning



Pairwise synaptic efficacies

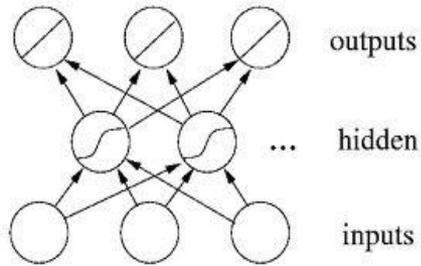
## Computational model of a cortical network

### Hebbian learning



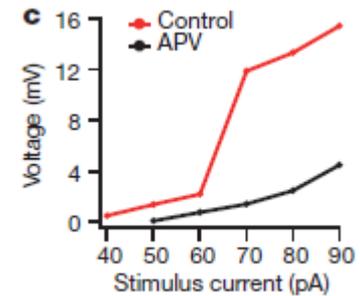
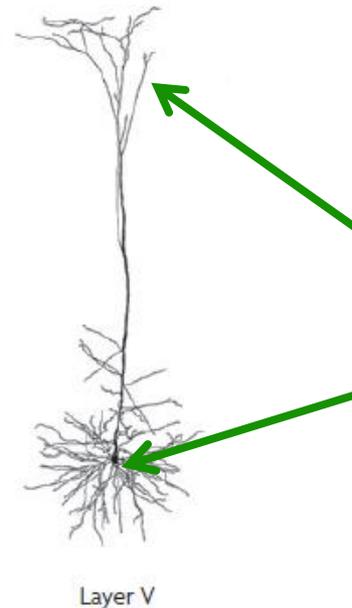
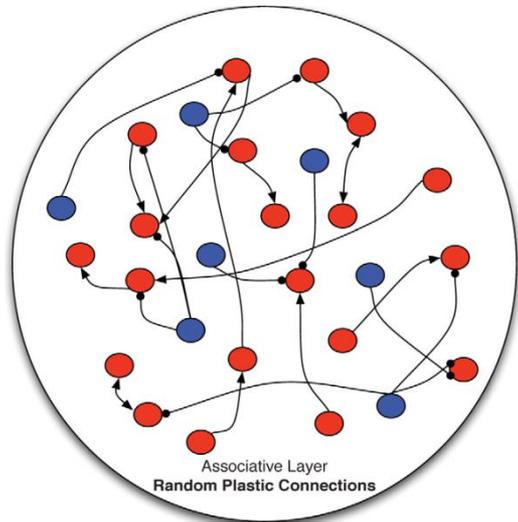
What are learning solutions found by the brain?

# Learning of non-linearly separable combinations



*a priori* wired hidden layer of neurons

increase of network size



Non-linear dendritic integration  
(from Lavzin et al., 2012 Nature;  
Spruston, 2008, Nature)

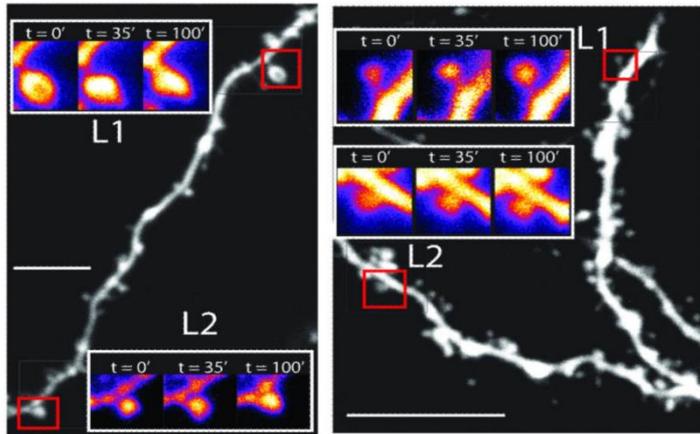
Integration in the cell body

The problem of random connectivity  
in the cerebral cortex (Markram et al., 2005, 2006)

How synaptic efficacies are learned at  
the level of dendrites?

# Intra-dendritic amplification of synaptic potentiation

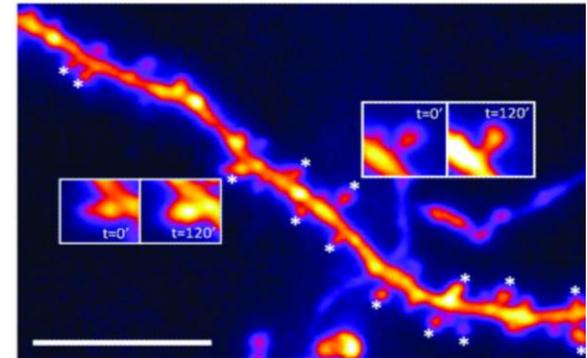
Within dendritic potentiation is stronger



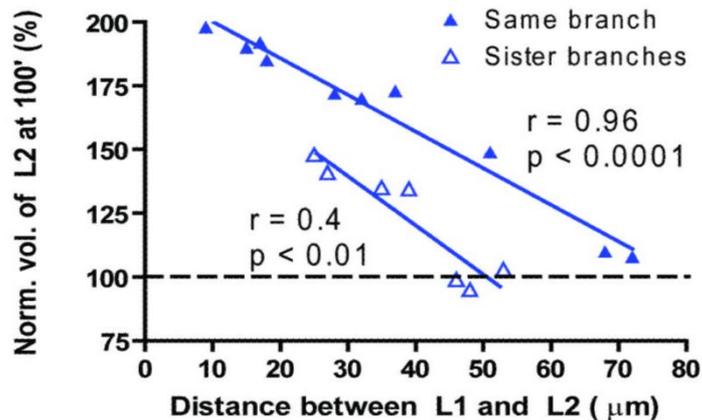
data from Govindarajan et al., 2012, Neuron

see also  
Branco et al., 2010, Science  
Harvey & Svoboda, 2007, Nature

Multi-synaptic potentiation is stronger

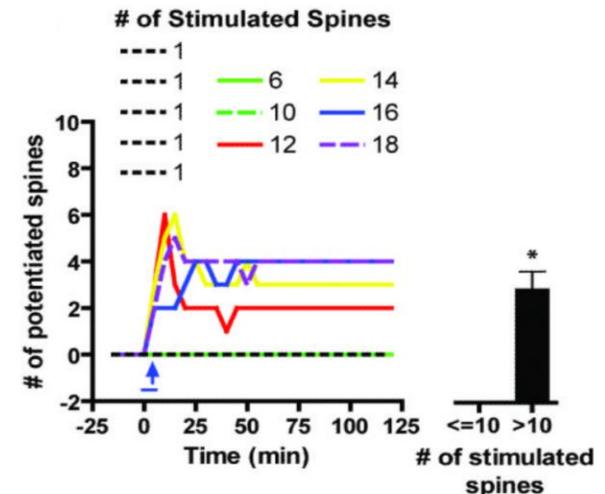


Potentiation if several synapses are active within a same dendrite

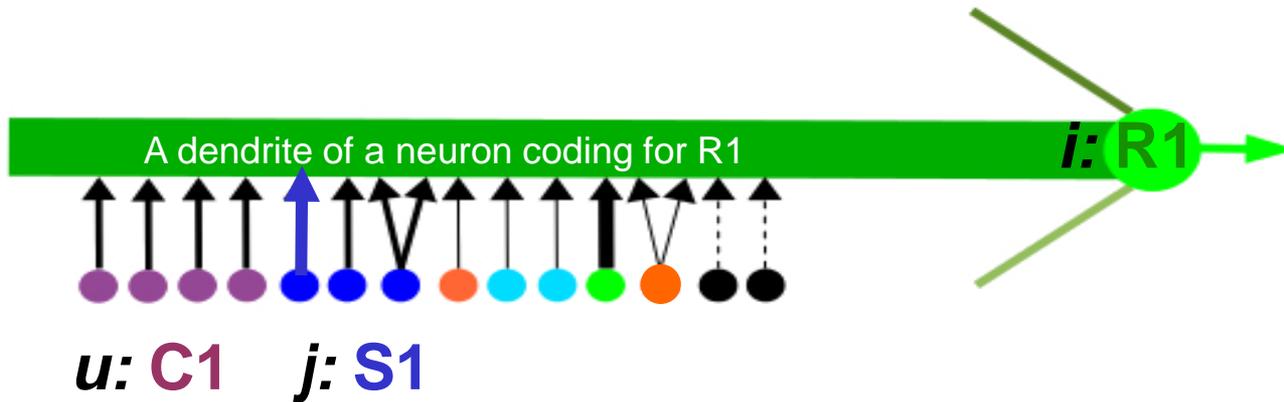


Before learning, connectivity is random and combinations are not known *a priori*.

How to link combinations of items to efficacies within dendrites?



# A solution: Intra-dendritic Inter-synaptic learning



Hebbian learning with pre-synaptic neuron  $j$

Amplification of potentiation inside population  $J$  (minus Hebbian learning of the pre-synaptic neuron)

Amplification of potentiation by other populations  $U$

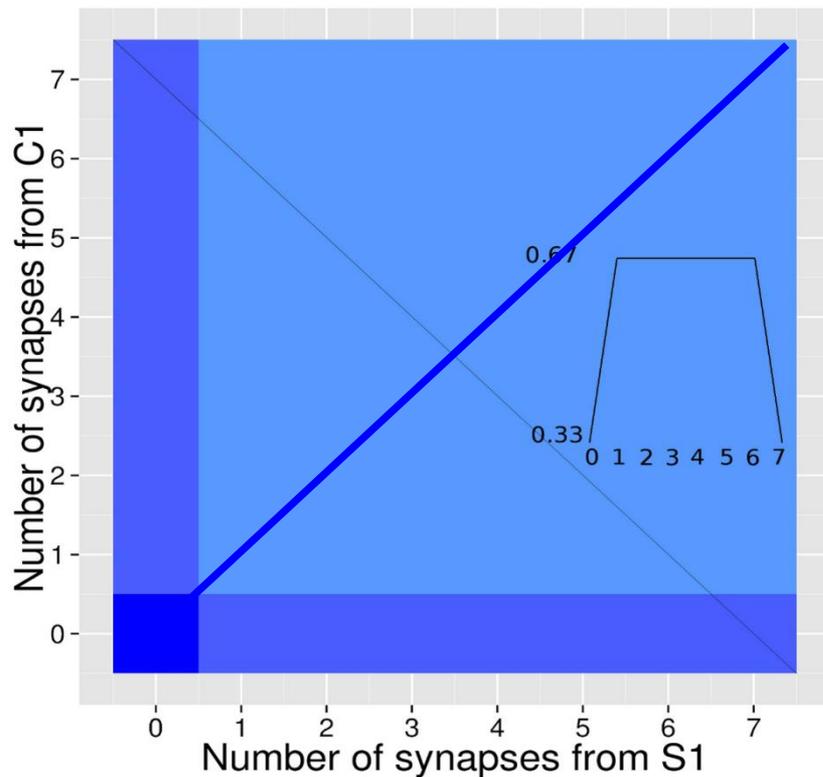
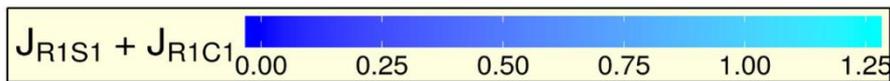
$$a_{ij}(D) = q^+ \zeta_i \zeta_j + q^+ \zeta_i \zeta_j (n_j - 1) + q^+ \zeta_i \zeta_j \left( \sum_{u=1, u \neq j}^g \zeta_{P_u} n_u \right)$$

Inter-Synaptic learning links groups of items to clusters of synapses from neurons coding for several items

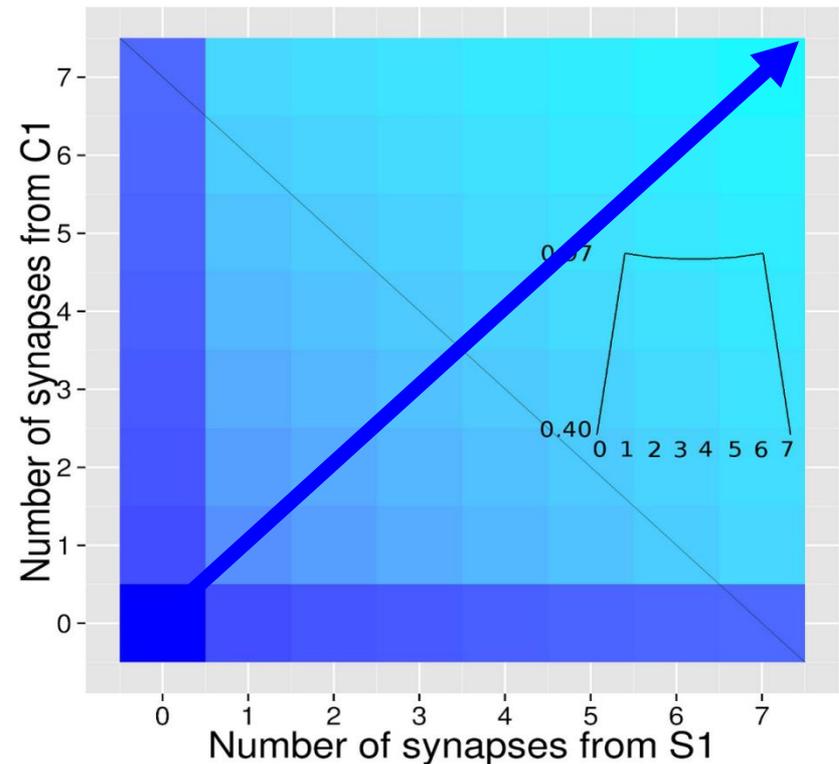
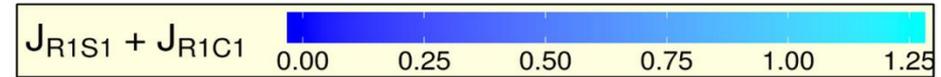
# Amplification of potentiation with increasing number of synapses

Dendrites with different numbers of synapses from C1 and S1  
on a neuron coding for R1 (learned combination C1S1R1)

Hebbian learning



Inter-synaptic learning



Different dendrites learn some combinations better than others

# Network processing of XOR combinations

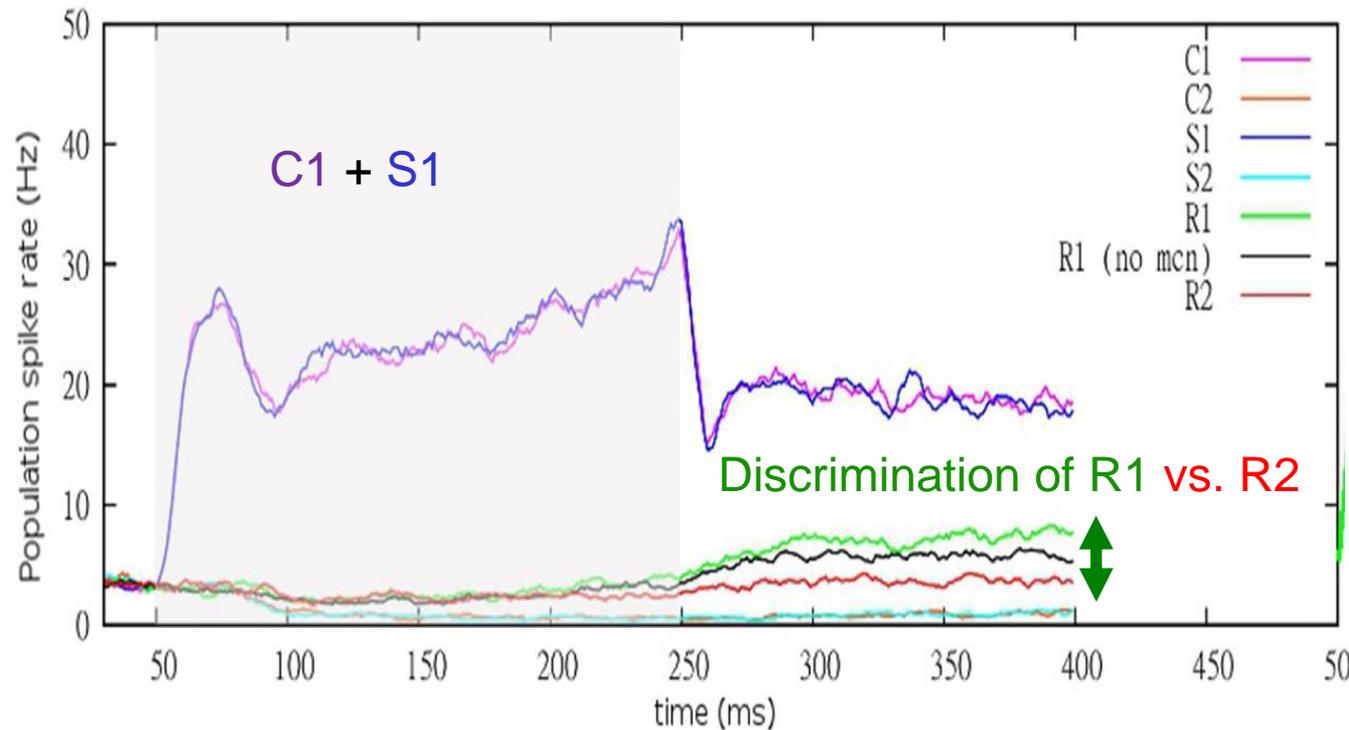
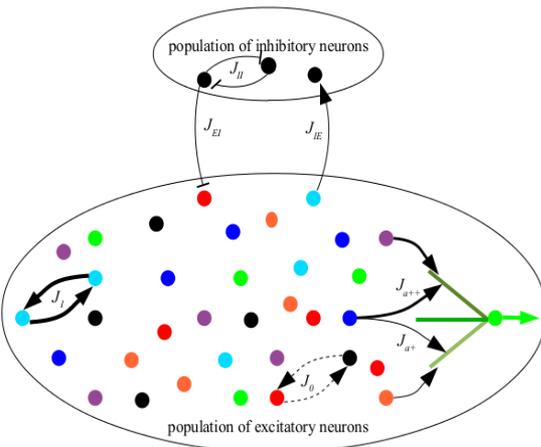
Inter-synaptic learning on random connectivity

5000 neurons with 50 dendrites of 20 synapses

Prospective activity of neurons coding for Response 1 like in electrophysiological studies in monkeys

(Wallis & Miller, 2009, Science)

## Cortical network model With dendrites and Inter-synaptic learning

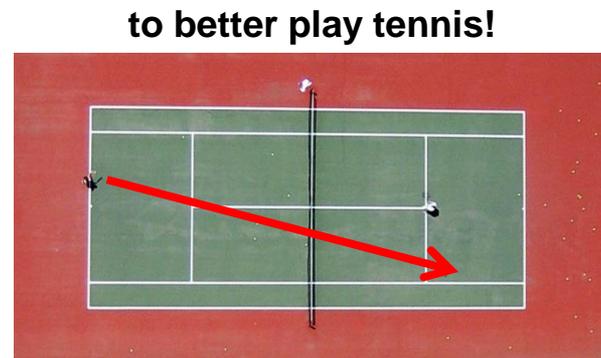
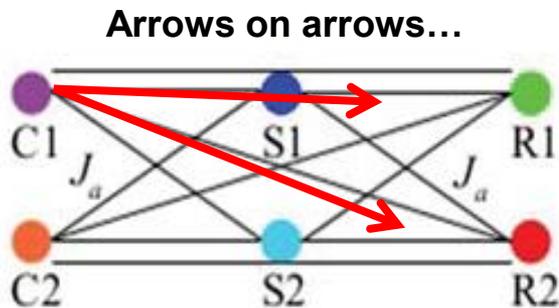
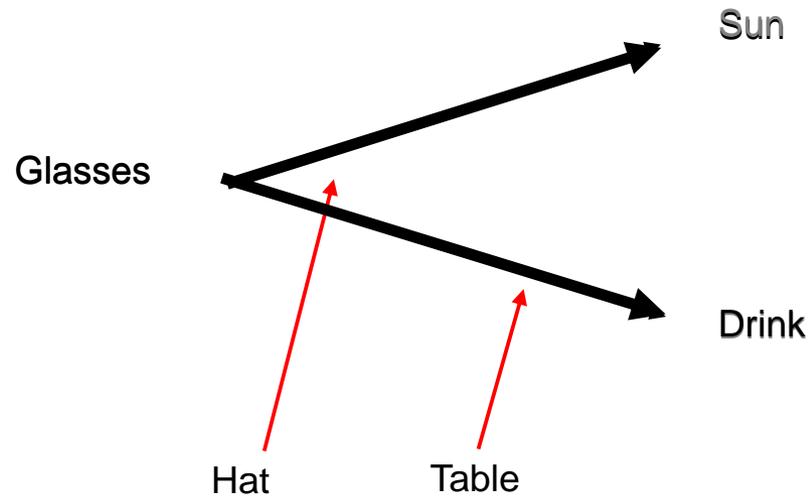


# Conclusion: generalization to conditional activation

Inter Synaptic learning: biologically realistic, on random connectivity, without additional neurons

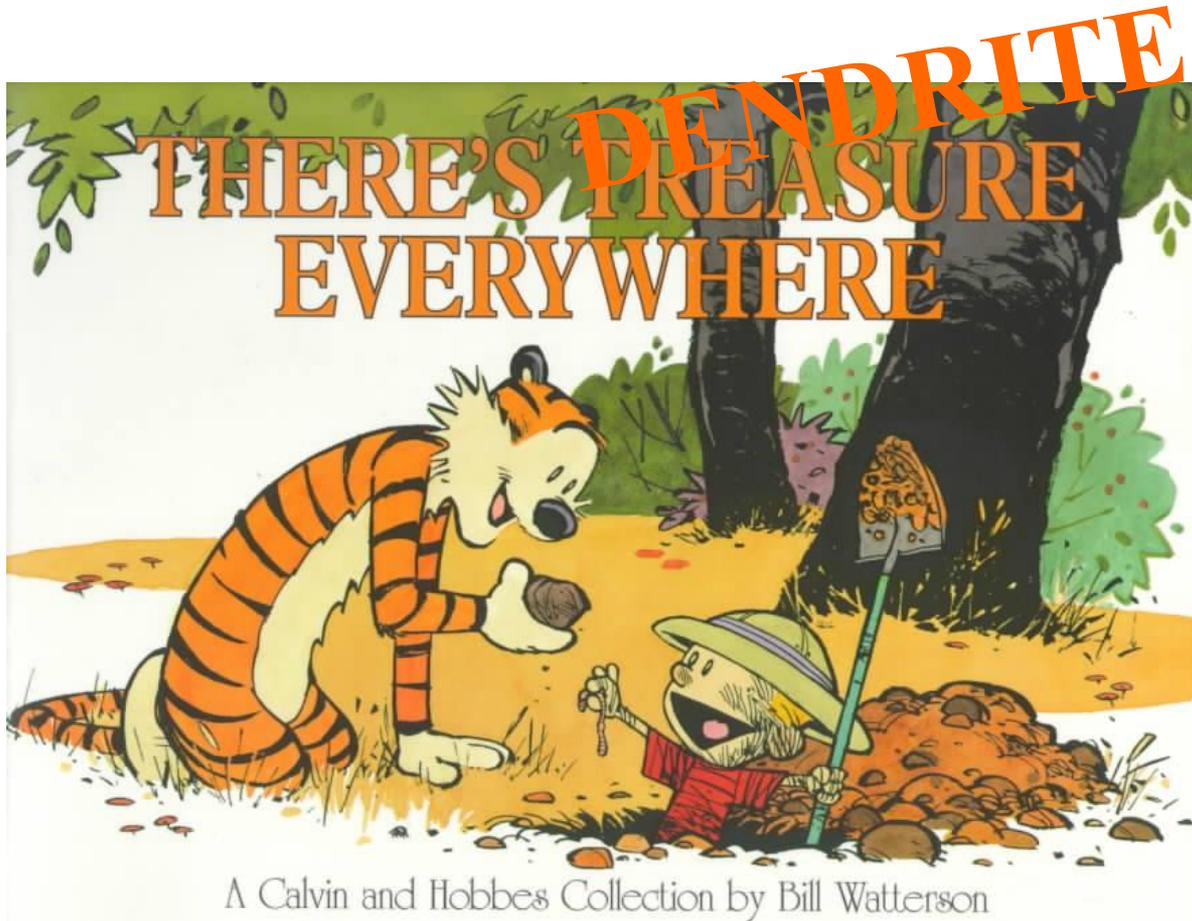
Non-local learning involving several pre-synaptic neurons increases network capacity (Alemi, Baldassi, Brunel & Zecchina, 2015)

General problems of context-dependent conditional activation in cognitive processing (semantic priming, mental arithmetics)



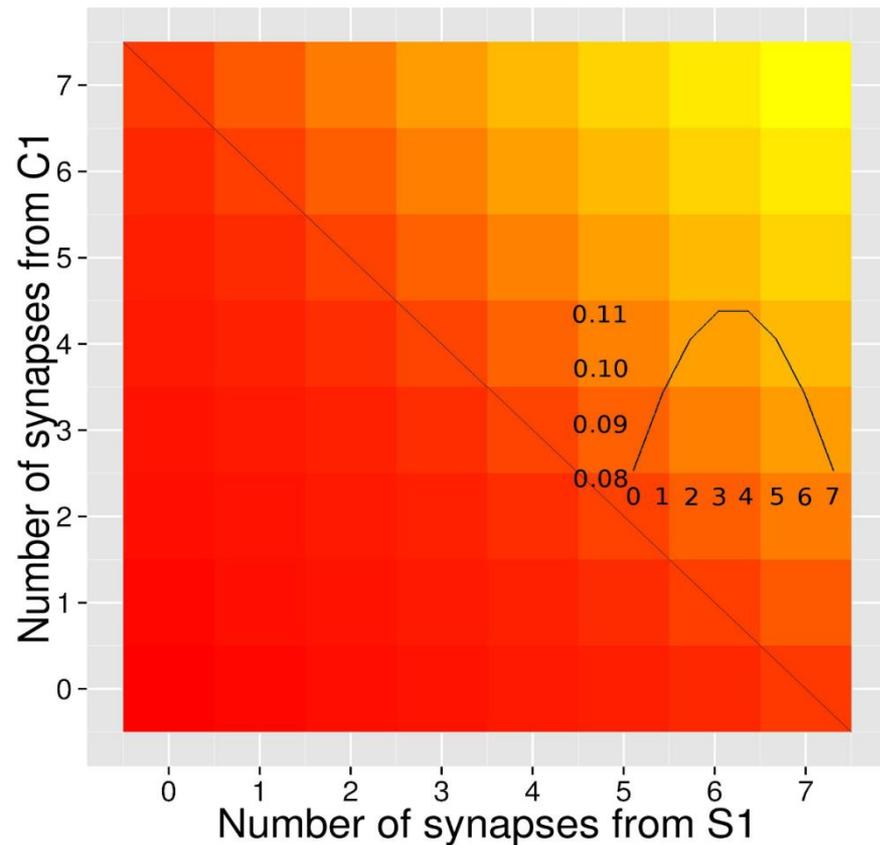
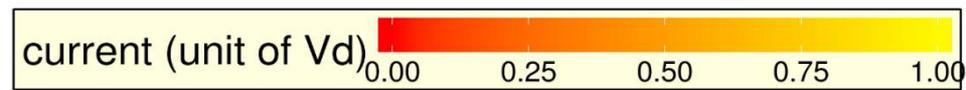
Thank you for your attention

Many thanks to Francis Avnaïm, Nicolas Brunel and Laurent Dumercy

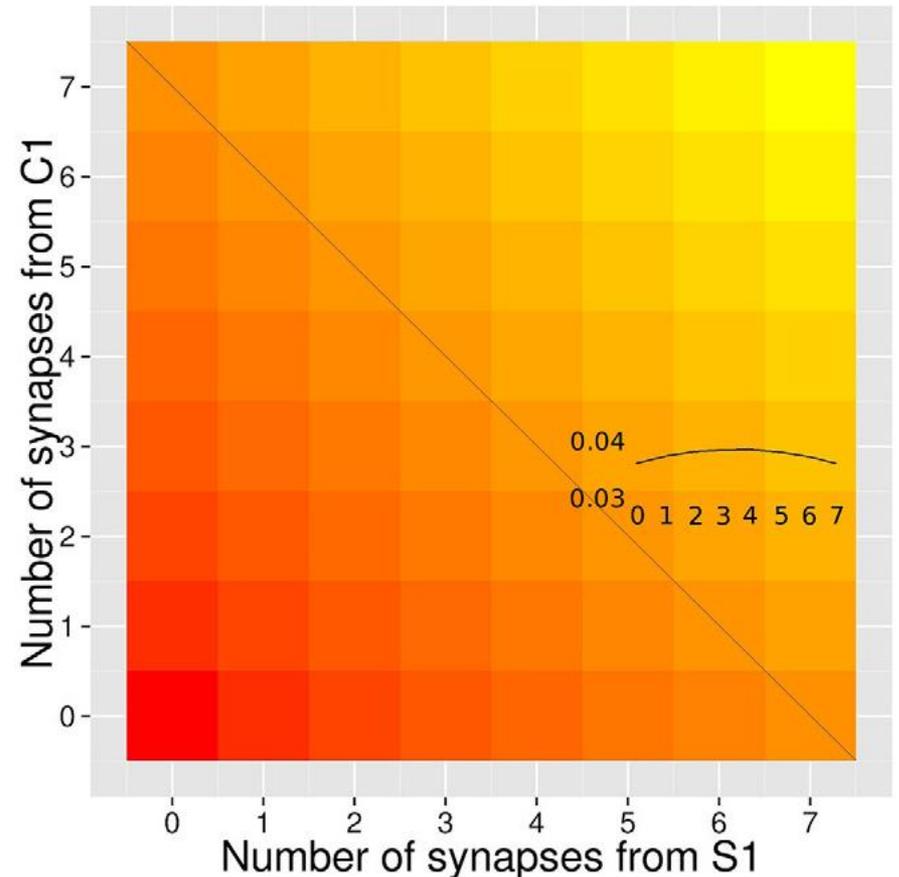
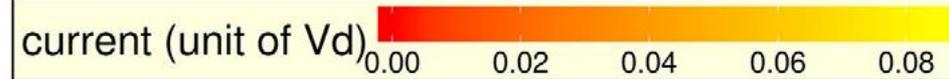


# Amplification of dendritic currents with increasing number of synapses (learned combination **C1S1R1**)

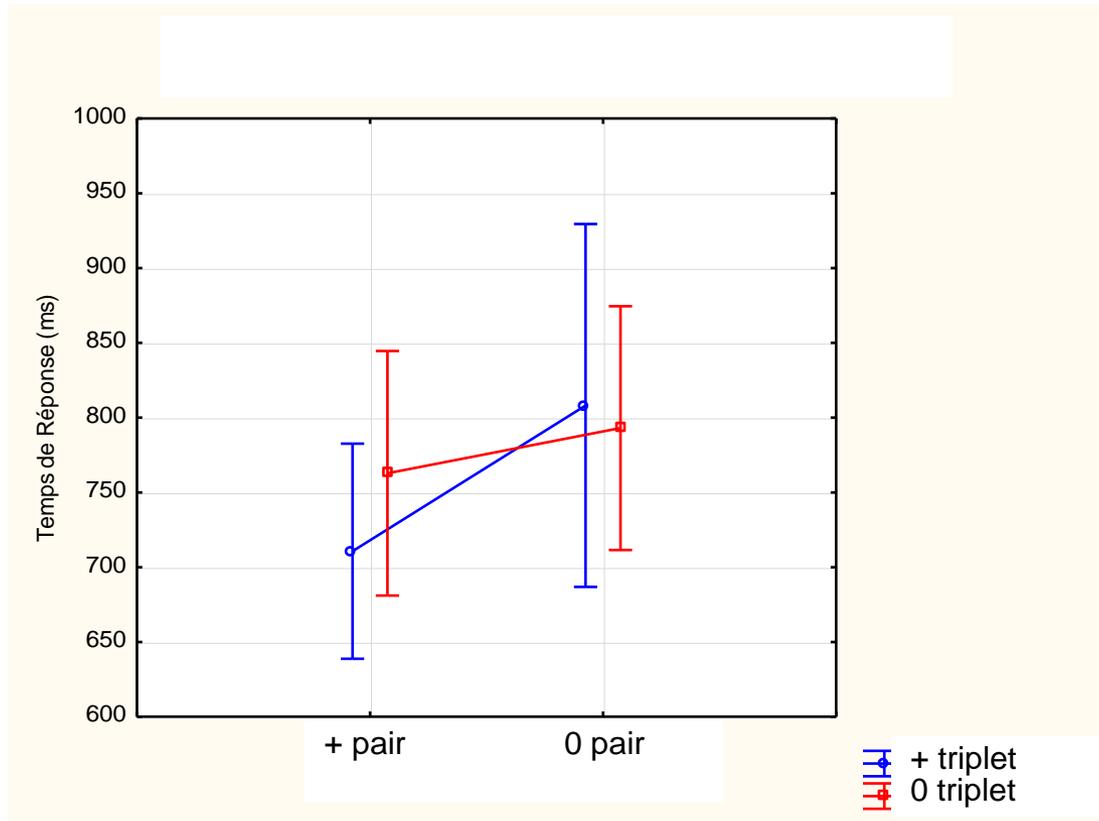
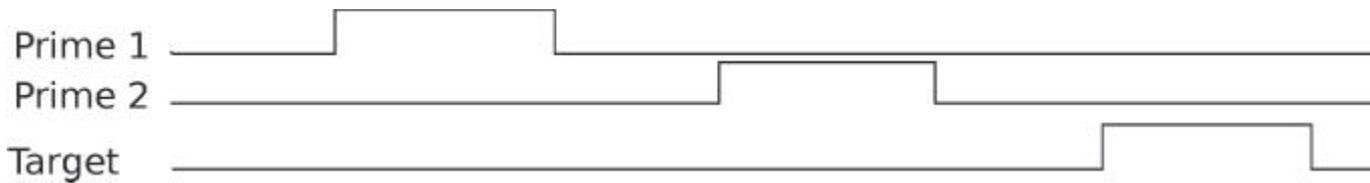
## Linear dendritic integration



## Non-linear dendritic integration



# Au delà des paires: Priming par les triplets



**Pour des paires équivalentes:**

**l'activation d'une cible 2 amorces est plus forte lorsque le triplet est fréquent**

# Processing of XOR combinations

Context (descend/climb)	Stimulus (normal/upside down)	Response (push/pull)
0	0	0
0	1	1
1	0	1
1	1	0

