

Perceptive self-organizing maps based on the coupling of neural fields with the BCM learning rule for multi modal association

Mathieu Lefort, Yann Boniface, Bernard Girau

▶ To cite this version:

Mathieu Lefort, Yann Boniface, Bernard Girau. Perceptive self-organizing maps based on the coupling of neural fields with the BCM learning rule for multi modal association. 2011. inria-00593567

HAL Id: inria-00593567 https://inria.hal.science/inria-00593567

Preprint submitted on 16 May 2011

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Perceptive self-organizing maps based on the coupling of neural fields with the BCM learning rule for multi modal association

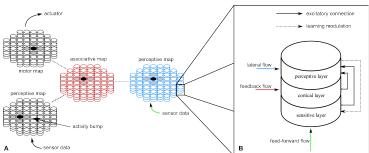
Mathieu Lefort¹, Yann Boniface¹, Bernard Girau¹

¹ LORIA, Campus Scientifique, BP 239, 54506 Vandoeuvre-lès-Nancy Cedex, France E-mail: mathieu.lefort@loria.fr

To move in a complex and changing environment, according to the active perception theory, an agent needs to learn the multi modal correlations between its actions and the changes they induce in the environment [1]. Human beings interact with their environment through several distant organs, whose sensory flow processing are influencing on each others as, for example, in the Mc Gurk effect [2]. In the functional view of the cortex, each sensory area processes a specific sensory flow and associative areas integrate these flows in a consistent representation of the world that influences in return uni modal perceptions as for the ventriloquist effect [3]. At a mesoscopic level, the cortex shows a generic structure composed of multi layers cortical columns. We propose a bio inspired model of perceptive map which, using a continuous and unsupervised learning, self-organizes to map a sensory data flow. Thus, using a spatial competition mechanism, a perceptive map provides an activity bump representing the current perception. In a multi modal architecture, by connecting multiple perceptive maps to an associative map with reciprocal spatially constrained connectivity, the spatial localization of each perception have to relax the multi modal constraints(see figure 1 A).

More precisely, each perceptive map is composed of generic cortical columns with computing and learning which are local and decentralized (see figure 1 B). Sensitive layer receives the feed-forward flow coming from a sensor and uses the BCM learning rule [4]. This synaptic learning rule is based on the hebbian one ans is able to autonomously raise a selectivity to one stimulus of the upcoming flow. From a computation point of view, BCM rule uses a sliding threshold between long term potentiation and long term depression, which has been confirmed by biological evidences [5]. The cortical layer receives feedback influence coming from the associative map, whose activity represents the multi modal perception. The perceptive layer is based on the neural fields theory [6] and receives the sensitive activity modulated by the cortical one. Thanks to the competitive mechanism introduces by the lateral connectivity with a difference of Gaussian shape, an activity bump emerges at the map level, where the activity is the most spatially consistent, representing a consensus between the local sensation and the multi modal constraints. To obtain a self-organization of the sensitive layer at the map level, the perceptive layer modulates the BCM activity, so that the spatial consistency of the activity bump is propagated to the selectivity organization. We have also introduced an unlearning term in the BCM learning rule in order to forget the current selectivity if it is not consistent with the received modulation. This unlearning mechanism provides plasticity to the self-organization in order to adapt to the multi modal constraints.

Figure 1: **A** Example of use of our perceptive map model in a multi modal architecture with two sensors and one actuator. **B** Generic structure of one cortical column in a perceptive map.



References

- 1. J.K. O'Regan, A. Noë: A sensorimotor account of vision and visual consciousness. *Behavioral and brain sciences* 2001, **24(05)**:939-973.
- 2. H. Mcgurk, J. Macdonald: Hearing lips and seeing voices. Nature 1976, 264(5588):746-748.
- 3. B. Bonath, T. Noesselt, A. Martinez, J. Mishra, K. Schwiecker, H.J. Heinze, S.A. Hillyard: **Neural basis of the ventriloquist illusion**. *Current Biology* 2007, **17(19)**:1697-1703.
- 4. E.L. Bienenstock, L.N. Cooper, P.W. Munro: Theory for the development of neuron selectivity: orientation specificity and binocular interaction in visual cortex. *Journal of Neuroscience* 1982, **2(1)**: 32.
- 5. M.F. Bear. Mechanism for a sliding synaptic modification threshold. Neuron 1995, 15(1):1-4.
- 6. S. Amari: **Dynamics of pattern formation in lateral-inhibition type neural fields**. *Biological Cybernetics* 1977, **27(2)**:77-87.