

# Functions of distributed plasticity in a biologically-inspired control system: from electrophysiology to robotics

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## 1 Introduction

We describe here the progress of a multi-disciplinary collaboration to evaluate cerebellar-inspired control architectures for the adaptive control of mobile robots. Our example problem is the calibration of the vestibulo-ocular reflex (VOR), a reflex which stabilises the eyes during head movements, for which the cerebellum is known to be essential.

A puzzling feature of the cerebellar control network for the VOR is that it uses two sites of plasticity, one within the cerebellum itself and one in the brainstem, related to the neurons to which the cerebellum projects. We are plasticity with the following goals:

- to characterise the properties of brainstem plasticity by electrophysiological recording
- to construct models of cerebellum and brainstem plasticity to investigate the computational properties of algorithms with brainstem plasticity
- to implement and evaluate biologically-inspired algorithms for the adaptive control of a robot-mounted camera.

## 4 Learning Control By Decorrelation

Decorrelation is a generic principle of motor skill learning from sensory error: *Errors sensed during motor tasks that are correlated with motor commands indicate inaccurate motor commands.*

Learning is driven by reducing the correlation between errors in motor control and motor commands. The cerebellum can be represented as a discrete-time adaptive filter  $C(z)$ , which relates motor commands  $u(t)$  to cerebellar output  $y(t)$ , i.e.  $y(t) = C(z)u(t)$  (see above right). The decorrelation algorithm updates the weights  $w(t)$  of  $C(z)$ , according to the learning rule

$$w(t+1) = w(t) - K \langle e(t) \rangle u(t)$$

where  $K$  is a gain that affects the rate of learning and  $\langle e(t) \rangle u(t)$  is the correlation between error and motor command.

## 7 In Vitro MVN Neuron Recording

We are investigating the cellular mechanisms which regulate the intrinsic excitability of MVNs, in brain-slice preparations *in vitro*.

Using whole-cell patch clamp recordings from MVNs, we have shown that the intrinsic excitability of the MVN cells can be regulated by patterns of inhibitory inputs applied to the cells (see slide 8).

The increase in excitability is long-lasting, and has been shown to be mimicked by drugs that block specific potassium channels, e.g. paxilline (see slide 8). Drugs which open potassium channels reduce the intrinsic excitability of MVN neurons, indicating that the regulation of potassium channel function is one important way in which the firing rate gain of the MVN neurons may be regulated.

Figure 8. In vitro, 2005. Decorrelation of Selfish Activity: The Plasticity of an Intrinsic Firing Rate in an Adaptively Firing Vestibulo-Ocular Motor Neuron. Nat. Rev. Neurosci. 8, 623-631.

## 10 Eye-Robot: Artificial VOR

A custom built 3 degree of freedom head movement simulator and camera platform rotates in 3 axes (yaw, pitch and roll). The gimbal mounted camera is operated by 3 brushless DC motors each driven with a dsPIC<sup>®</sup> microcontroller including position feedback via incremental encoders.

Communication between the different driver modules is established using CAN bus. The interface is also used to control the Eye-Robot (CAN) operating at 1Mbit/s. This interface is also continuously observe the system state and evaluate the system on a PC. The performance on a PC, the Eye-Robot, is controlled by the Field Programmable Gate Array (FPGA), the sensors and the microcontroller implemented using a spiral peripheral interface (SPI) running at 9 MHz.

## 2 The VOR: A Test-Bed System

The cerebellum is divided into thousands of modules, each with identical micro-circuitry but distinguished by other parts of the brain.

Since the micro-circuitry of the cerebellum is so uniform, the choice of which to investigate the motor control can be made on grounds such as tractability and generic significance.

The VOR is just such a problem. This is a reflex that drives the eyes in the opposite direction to head rotation to stabilise the image on the retina.

## 5 Slip Delay Prevents HF Learning

The retinal slip delay, which is 100 ms, would cause learning in the VOR to be unstable at high frequencies (HF).

- Learning becomes unstable, because with a 100 ms delay the sign of the correlation reverses for motor command signals  $u(t)$  above 2.5 Hz (below), i.e.

$$w(t+1) = w(t) + K \langle e(t) \rangle u(t)$$

- Stable Learning
- Critically Stable
- Unstable Learning

## 8 Possible Plasticity Mechanism

The intrinsic excitability of MVN cells is increased above normal following intermittent inhibitory hyperpolarising current injections which silence the action potential firing of the neurons (below).

Left: Gain plot (current input (pA) v. steady state instantaneous firing rate) of a type B MVN cell. The neuron was hyperpolarised with a current step, causing a transient decrease in firing rate. In the presence of BK, the frequency gain was unaffected (remains parallel) but sensitivity to a given current input.

## 11 Robot Construction

A novel design of a 3-D gyroscope system using three MEMS gyros-capable ADXL500 (Analog Devices<sup>®</sup>) emulates the mammalian vestibular system.

An embedded microcontroller generates realistic emulation of sensor/motor dynamics. The neuro-controller is implemented on the Xilinx Virtex II range of FPGAs.

Top right: BenNUJEY Virtex II 2V6000 FPGAs with their heat-sinks. Left: Gimbal mounted camera with driver circuitry, yaw, pitch and roll control. Bottom right: 7x7 mm 1 axis gyroscope.

## 3 Project Aims and Objectives

### Project Aims

- Elucidate and evaluate the function of distributed plasticity in order to:
  - improve the understanding of the biological motor control system and
  - determine the potential for improving traditional engineering solutions using biologically motivated methods.

### Objectives:

- Investigate rules of plasticity in recording from medial vestibular nucleus (MVN) neurons.
- Determine a computational model of plasticity mechanisms, control algorithms and oculomotor plant dynamics to mimic the biological VOR system.
- Construct a robot to mimic the VOR system.
- Evaluate performance of control in the robot to provide feedback on the algorithm from both a biological and engineering perspective.
- Adjust investigation into plasticity VOR model based on findings from robot implementation.

## 6 Brainstem Plasticity Enables HF Learning

Learning in the cerebellum is stable for motor command signals  $u(t)$  below 2.5Hz.

- Low-pass filtering the motor command signal (cut-off at 2.5Hz) leads to stable learning in the cerebellum (right).
- The high frequency gain is now learnt by the brainstem (below).

The cerebellum provides the teaching signal to the brainstem.

- The correlation between cerebellar output  $y(t)$  and the head velocity signal  $x(t)$  is used to adjust the intrinsic gain  $g(t)$  of the brainstem (see slide 4), i.e.

$$g(t+1) = g(t) - R \langle y(t) \rangle x(t)$$

Where  $R$  is a gain term.

## 9 Signal Transmission Improved

Bode plot showing the response of a model of a type B MVN neuron in the presence of BK. The blockade of BK potassium channels by paxilline did not increase the low-frequency gain of the neuron. However, the corner frequency was increased. This corresponds to an increase in gain at higher frequencies, improving the signal transmission properties of the cell in the affected region.

Proof of Concept: It has been demonstrated that plastic effects on the signal transmission properties of MVN neurons are achievable *in vitro*.

## 12 Virtual Plant

A separate microcontroller drives each DC motor, corresponding to each axis of rotation, in a closed loop scheme. This architecture provides the freedom to add an 'arbitrary' filter in an open-loop manner so that each drive system as a whole can emulate a desired plant model (below).

Right: The response of a model (simulated in Matlab) and the response of the mechatronic eye (including the same plant model) to a given reference signal. The oculomotor plant model utilised in this instance was a first order filter with a time constant 170.5ms.