Learning an inverse model for vocal production: toward a bio-inspired model

6th European Birdsong Meeting, April 12-13, 2018, Odense, Denmark

Silvia Pagliarini (with Xavier Hinaut and Arthur Leblois) INRIA Bordeaux Sud-Ouest, Institut des Maladies Neurodégénératives, Université de Bordeaux, FR







WHAT IS SENSORIMOTOR LEARNING?

Control problem which maps a sensory input into a motor output

Basic components:

• Input: sensory stimulus

• Output: reproduction of the stimulus



LEARNING BY IMITATION AND INVERSE MODEL

Imitation: learning from a tutor using a feedback guided error



LEARNING BY IMITATION AND INVERSE MODEL

Imitation: learning from a tutor using a feedback guided error

Inverse model: the aim is to transform a sensory stimulus into the corresponding motor command



A BIOLOGICAL EXAMPLE: SONG LEARNING IN BIRDS

• Comparable learning mechanisms and behavior



A BIOLOGICAL EXAMPLE: SONG LEARNING IN BIRDS

• Comparable learning mechanisms and behavior



Brainard and Doupe, 2002

A BIOLOGICAL EXAMPLE: SONG LEARNING IN BIRDS

• Comparable learning mechanisms and behavior











HAHNLOSER-GANGULI THEORETICAL MODEL



 $\Delta W_t = \eta (M_t - W_{t-1}A_t)A_t^T$

NONLINEAR MODEL INTRODUCTION



 M^st : target motor pattern

 σ : tuning selectivity width

 $||M_{j}^{*} - M_{t}||^{2}$ represents the distance between the target and the random exploration

GANGULI-HAHNLOSER MODEL



 $\Delta W_t = \eta (M_t - W_{t-1}A_t)A_t^T$

NORMALIZATION

Synaptic weights have a maximal value, related to the number of synaptic receptors one neuron is able to produce.

NORMALIZATION

Synaptic weights have a maximal value, related to the number of synaptic receptors one neuron is able to produce.

- Maximal weights normalization $W_{i,j} = rac{W_{i,j}}{<\!W\!>_{col}}$
- Supremum weights normalization $W_{i,j} = egin{cases} W_{i,j} & if & < W >_{col} < 1 \ rac{W_{i,j}}{< W >_{col}} & otherwise \end{cases}$

NORMALIZATION

Synaptic weights have a maximal value, related to the number of synaptic receptors one neuron is able to produce.

- Maximal weights normalization $W_{i,j} = rac{W_{i,j}}{\langle W
 angle_{col}}$
- Supremum weights normalization $W_{i,j} = egin{cases} W_{i,j} & if & < W >_{col} < 1 \ rac{W_{i,j}}{<\!W\!>_{col}} & otherwise \end{cases}$
- Decreasing factor normalization $\Delta W_{i,j} = \eta M_t A_t \left(1 < W >_{col}
 ight)$

NORMALIZED INVERSE MODEL



Normalization applied over the auditory neurons

NORMALIZED INVERSE MODEL



$$\Delta W_{i,j} = \eta M_t A_t igg(1 - < W >_{col} igg)$$

AUDITORY SELECTIVITY EFFECT



VARYING INPUT/OUTPUT DIMENSION



Distance from the motor target at convergence

VARYING INPUT/OUTPUT DIMENSION



Convergence time



- Simple normalization schema are successful in the nonlinear model
- Decreasing tuning selectivity width:
 - convergence time explosion
 - accuracy of learning increases
- Auditory VS motor dimension

• Duration of syllable and feedback delay

- Duration of syllable and feedback delay
- Production of sound

- Duration of syllable and feedback delay
- Production of sound

Enjoy the poster from Xavier Hinaut



- Duration of syllable and feedback delay
- Production of sound
- Make prediction on experimental data

Enjoy the poster from Xavier Hinaut



Thanks for the attention.

$$d_t=rac{||M^*-W_tA^*||}{n_m}$$
 .