



Computations in human sensorimotor control: Modularity & Active Sensing

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- Neural control of voluntary movement
 - Hierarchical organisation of the motor system
 - Variability in the motor system

- Computational Framework
 - Dimensionality reduction
 - Task encoding/decoding – functional characterization

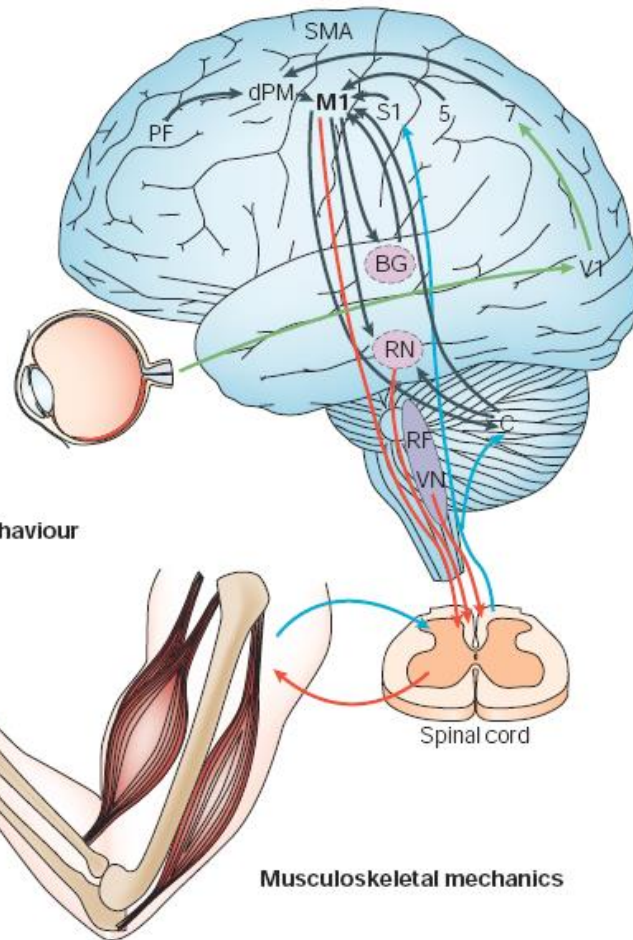
- Space-by-time modularity in muscle activity

- Neural correlates of active sensing behaviour

Hierarchical organisation of human motor system



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from Scott, 2004, Nat Rev Neurosci

Motor behaviour reflects the combined action of the **neural circuit** that controls movement and the **mechanical properties** of the limb.

Three (or more!) levels of the CNS

Spinal cord: integrates sensory feedback with descending commands, generates stereotypical motor patterns.

Brainstem: postural control, refines motor patterns

Cerebral cortex: supports a large and adaptable motor repertoire

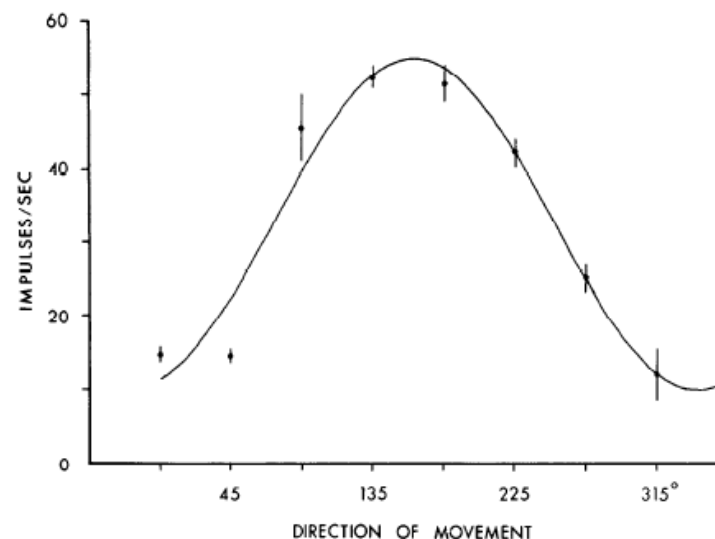
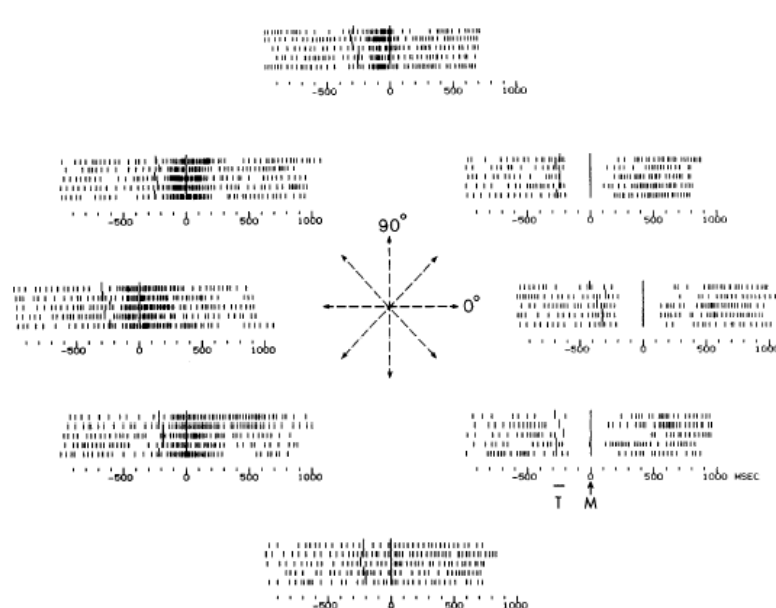


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The Journal of Neuroscience
Vol. 2, No. 11, pp. 1527-1537
November 1982

ON THE RELATIONS BETWEEN THE DIRECTION OF TWO-DIMENSIONAL ARM MOVEMENTS AND CELL DISCHARGE IN PRIMATE MOTOR CORTEX¹

APOSTOLOS P. GEORGOPOULOS,² JOHN F. KALASKA,³ ROBERTO CAMINITI,⁴ AND JOE T. MASSEY⁵



Directional tuning of cell discharges

PCA110.S01

S1A

A small number of task parameters is controlled at the motor cortical level

Lower level – Spinal Cord

Bizzi's seminal work at MIT

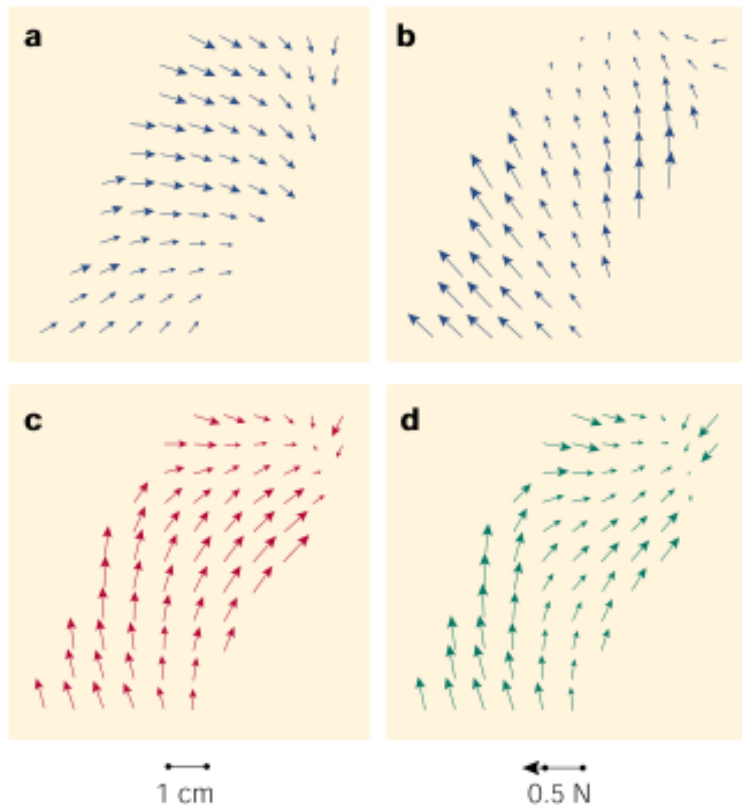
Linear Combinations of Primitives in Vertebrate Motor Control



F. A. Mussa-Ivaldi; S. F. Giszter; E. Bizzi

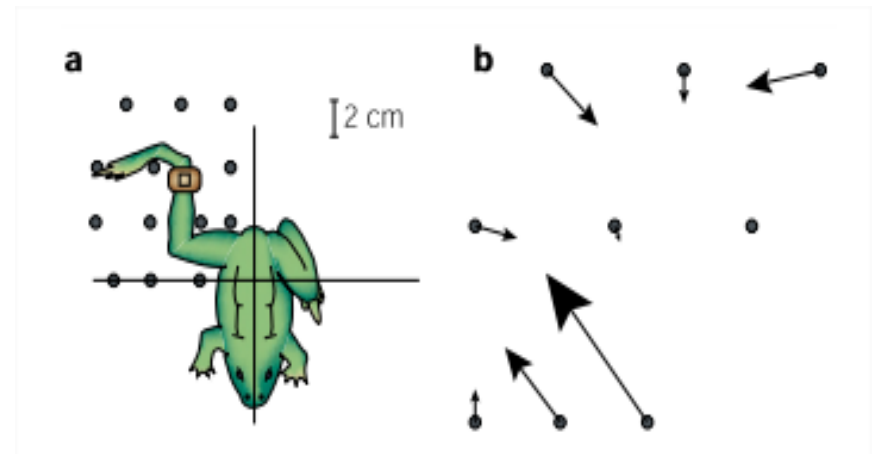
Proceedings of the National Academy of Sciences of the United States of America, Vol. 91, No. 16 (Aug. 2, 1994), 7534-7538.

Stimulation in site a Stimulation in site b



Theoretical summation

Stimulation in sites a+b



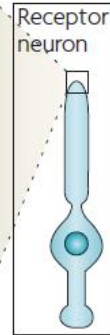
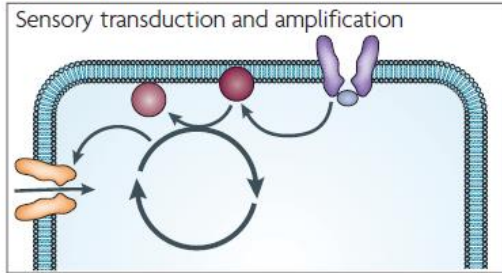
Force fields produced by micro-stimulation in one location of the spinal cord of frogs

Noise in the nervous system

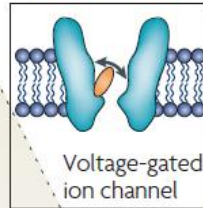


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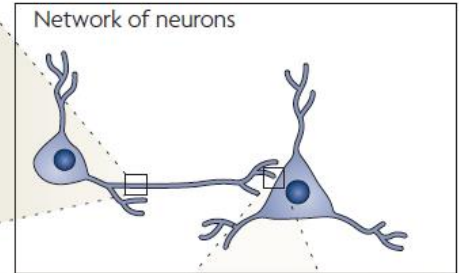
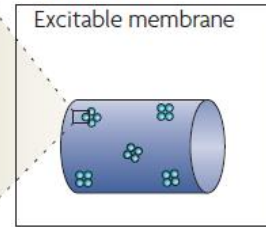
a Sensory noise



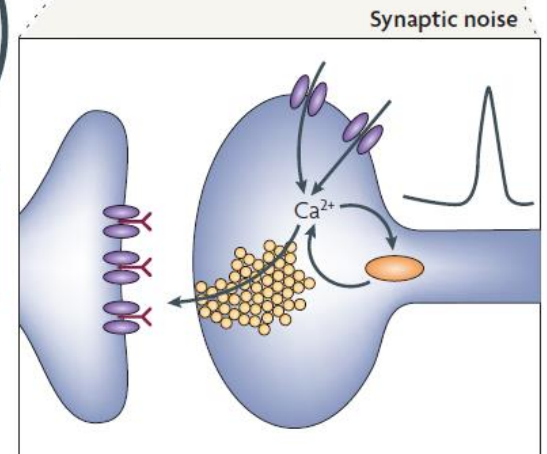
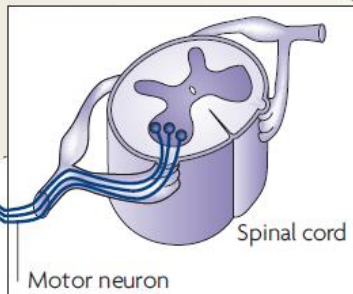
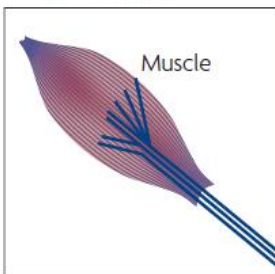
b Cellular noise



Electrical noise



c Motor noise



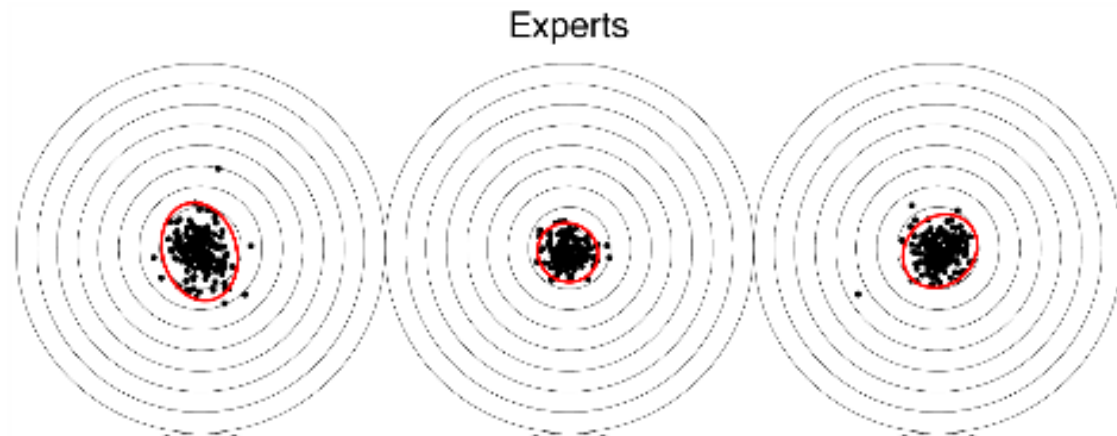
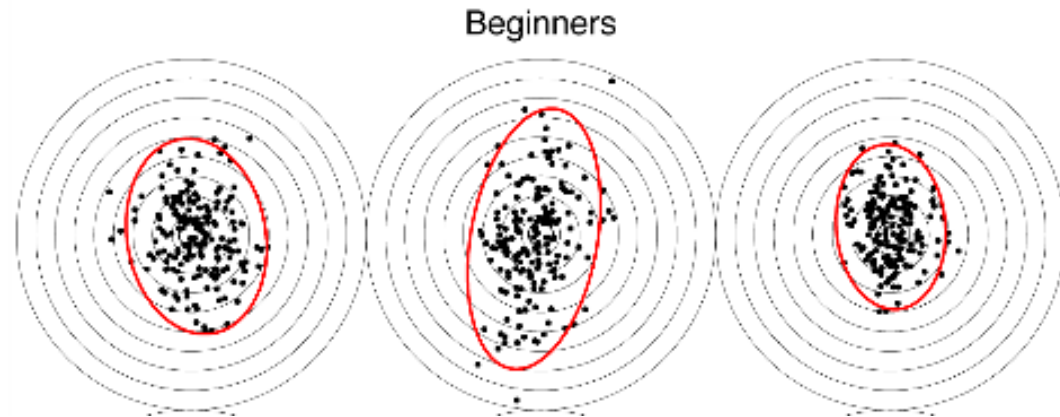
Noise leads to inherent variability in motor behaviours

from Faisal et al. 2009, Nat Rev Neurosci

Sensorimotor variability in humans

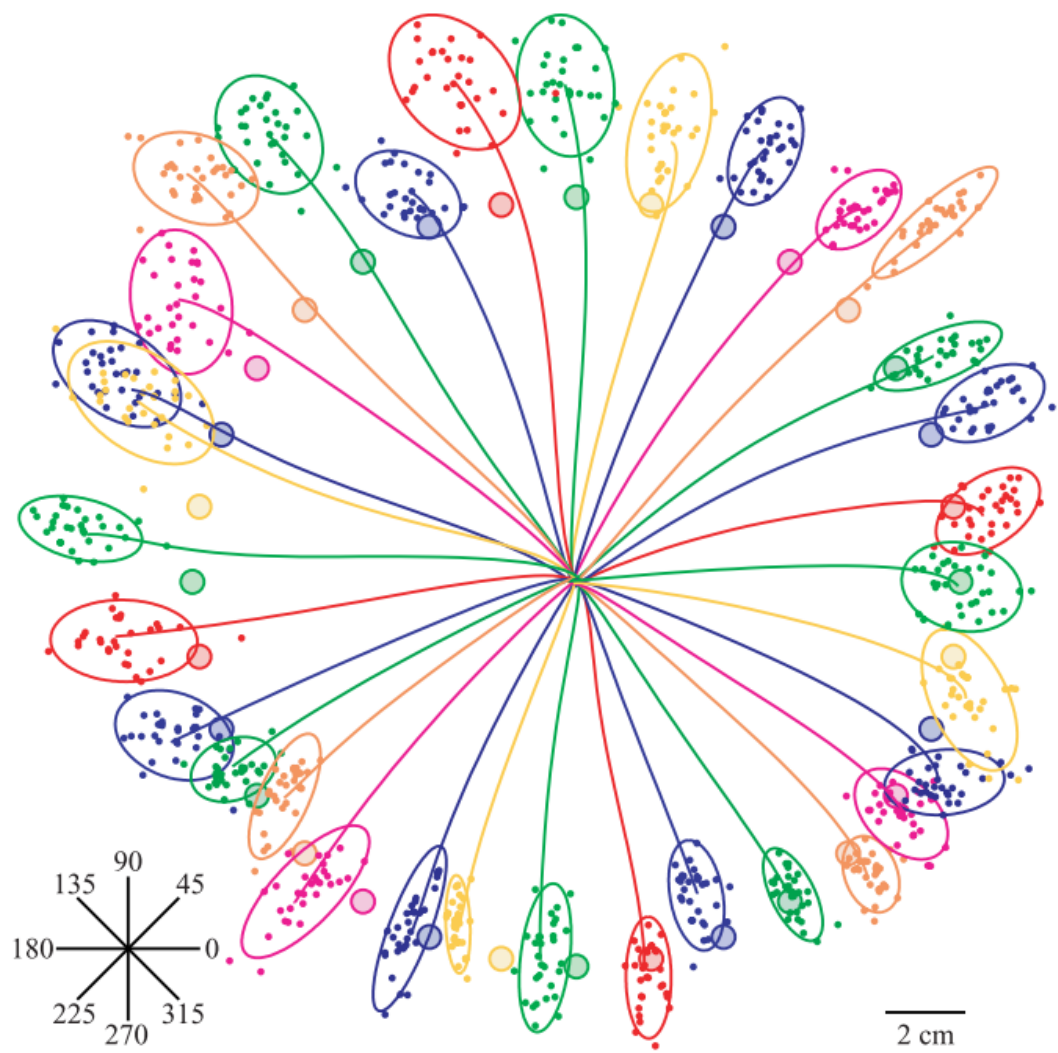
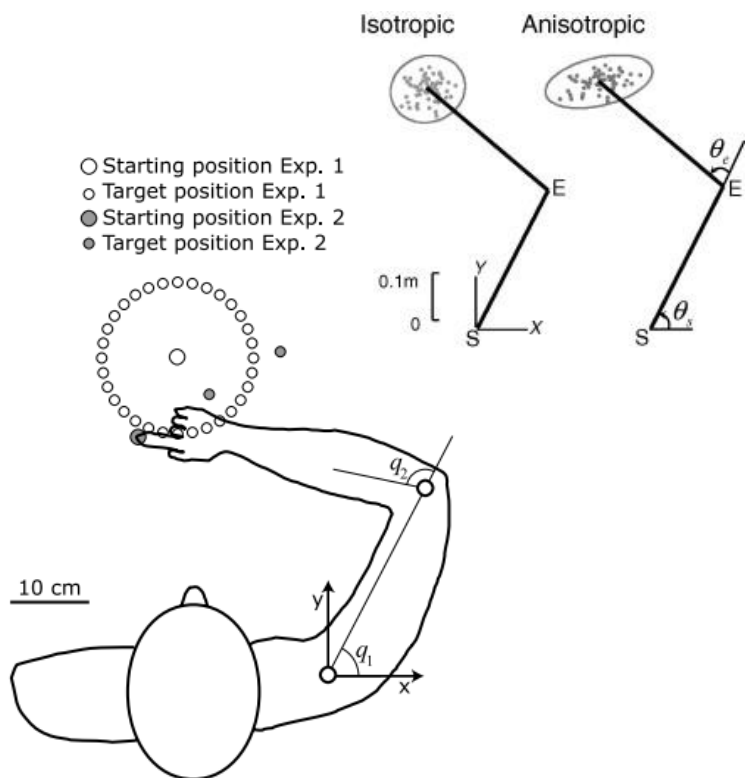


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Variability in arm reaching

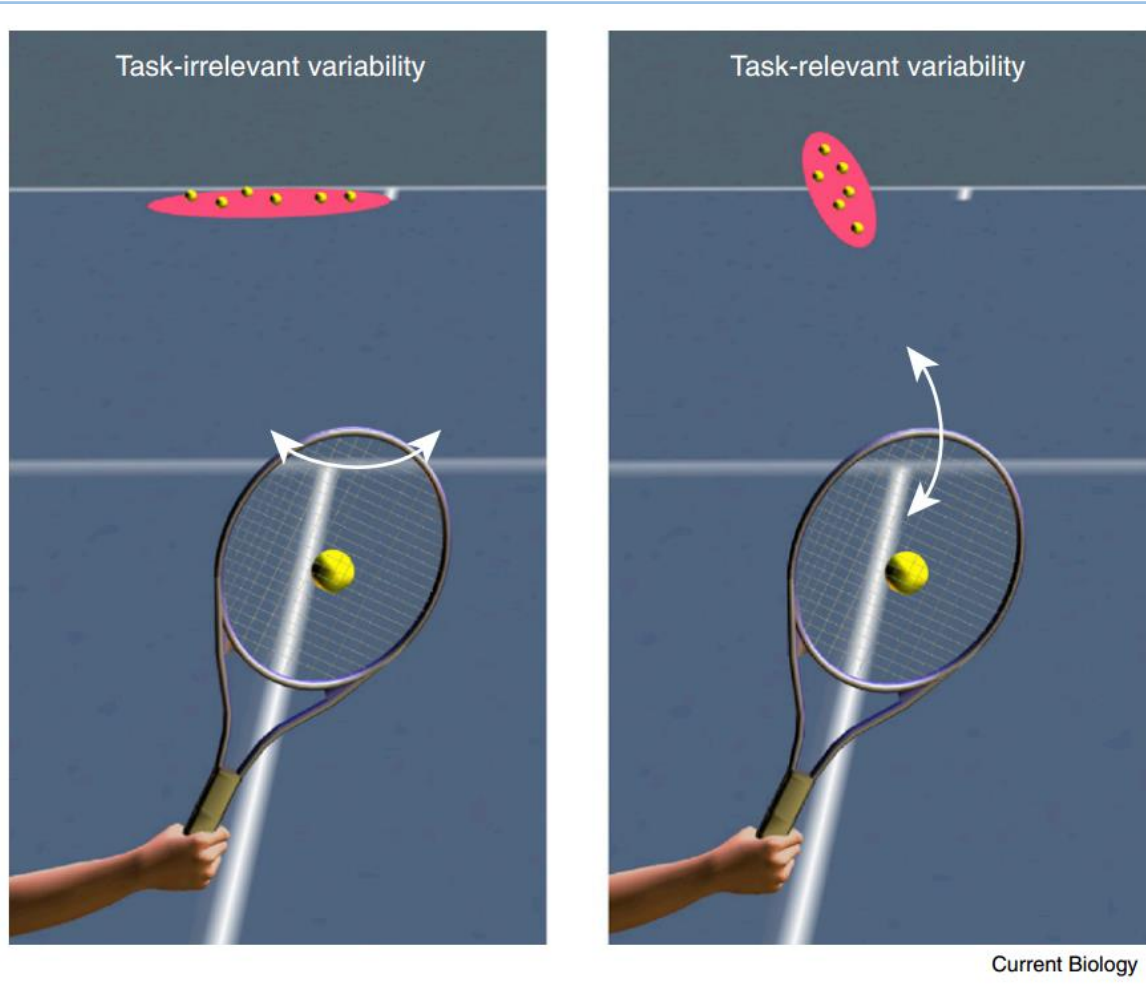


from van Beers et al, 2004, J Neurophysiol

Task-relevant variability



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Miminal intervention principle: control of task-relevant errors

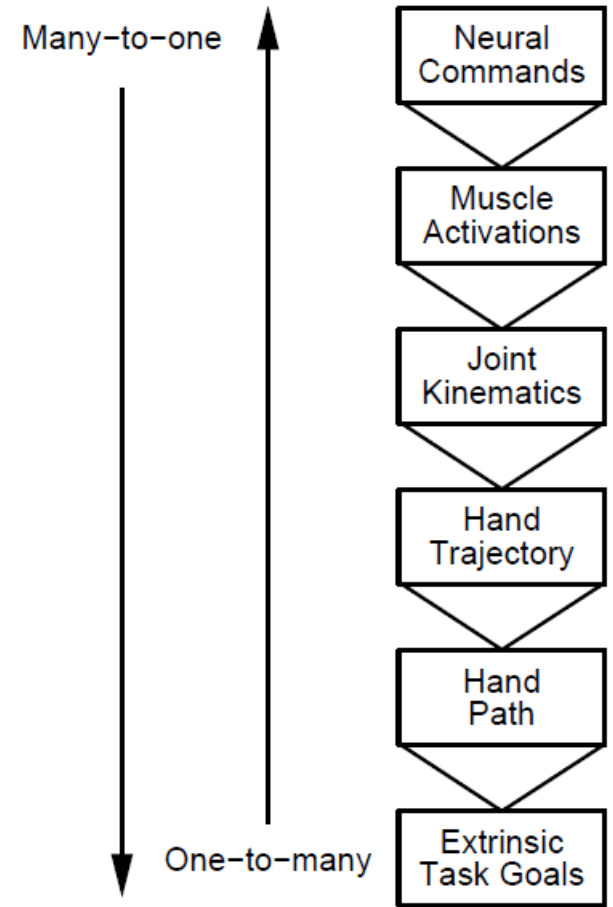
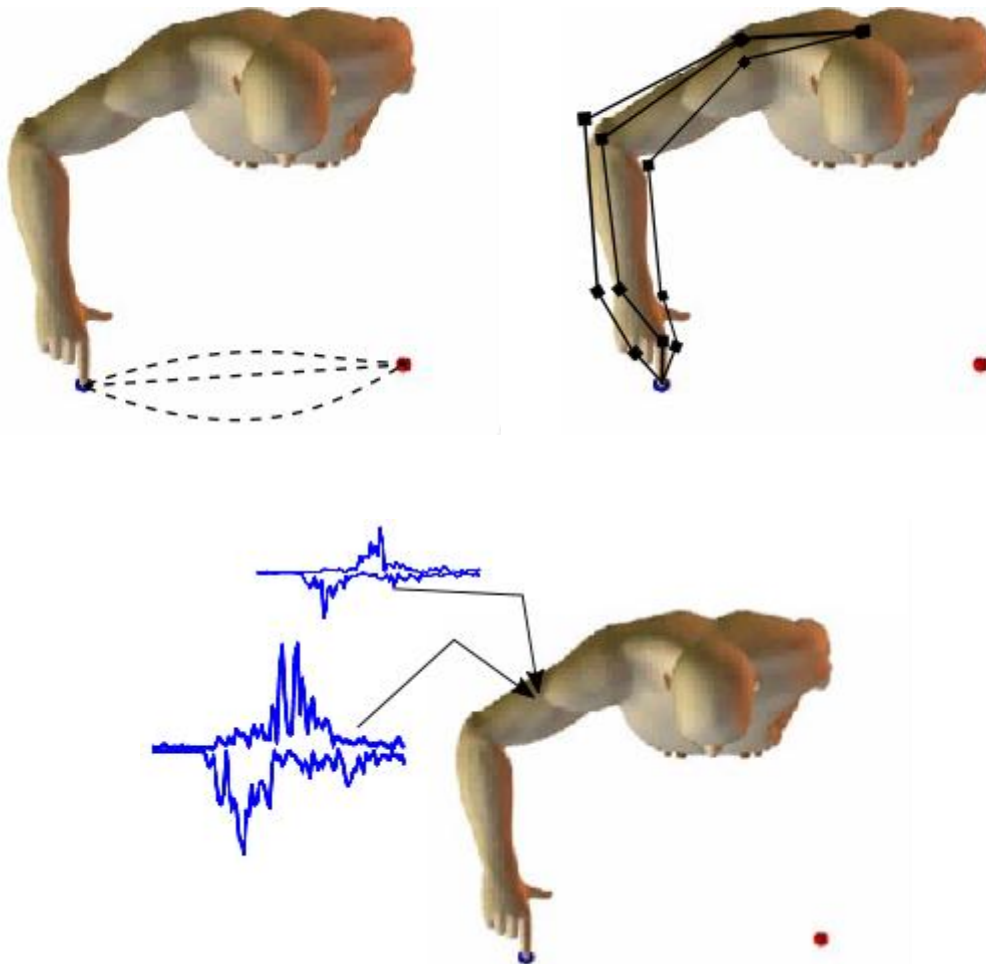
Example :
pointing to a line



Degrees-of-freedom problem



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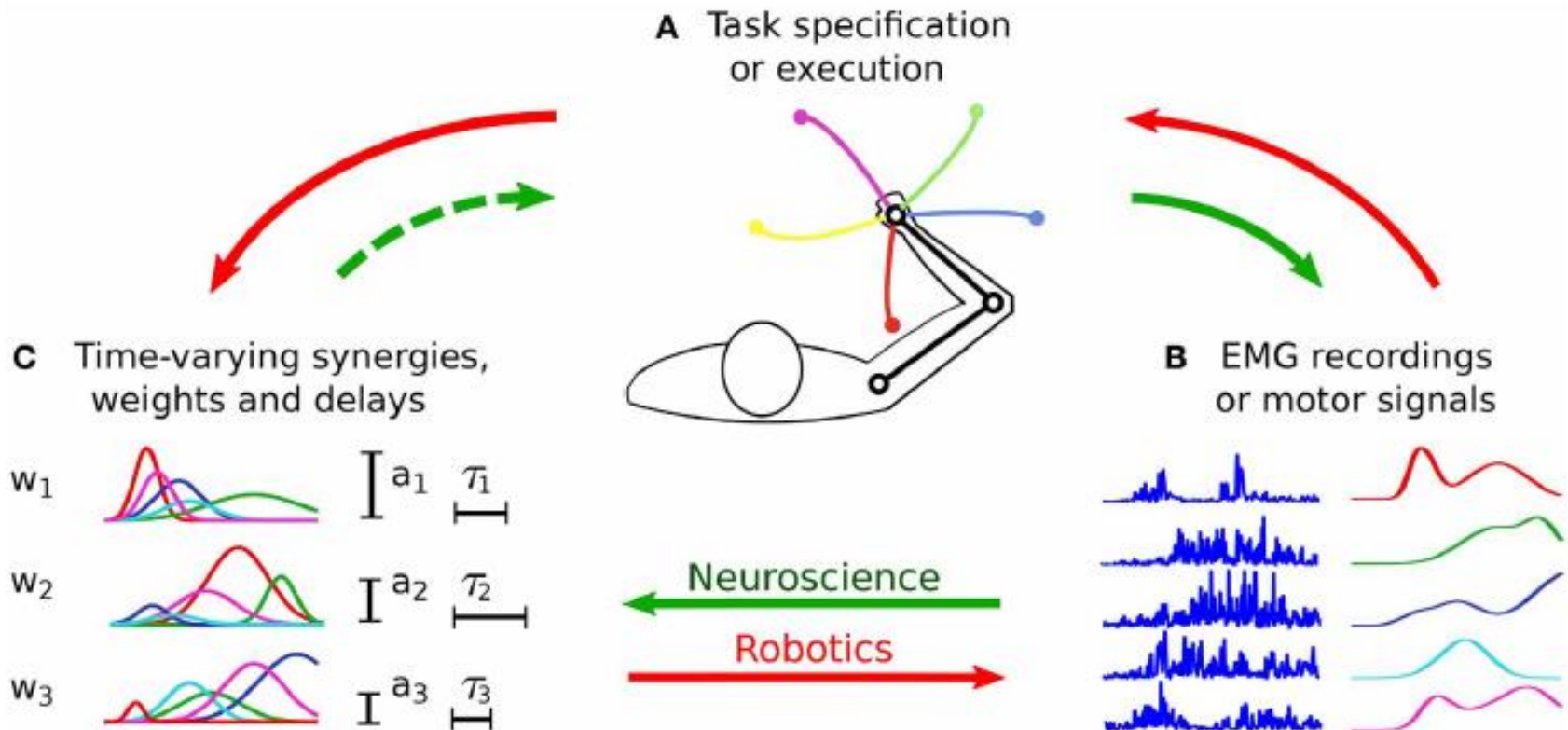


from Wolpert, 1997, Trends Cogn Sci

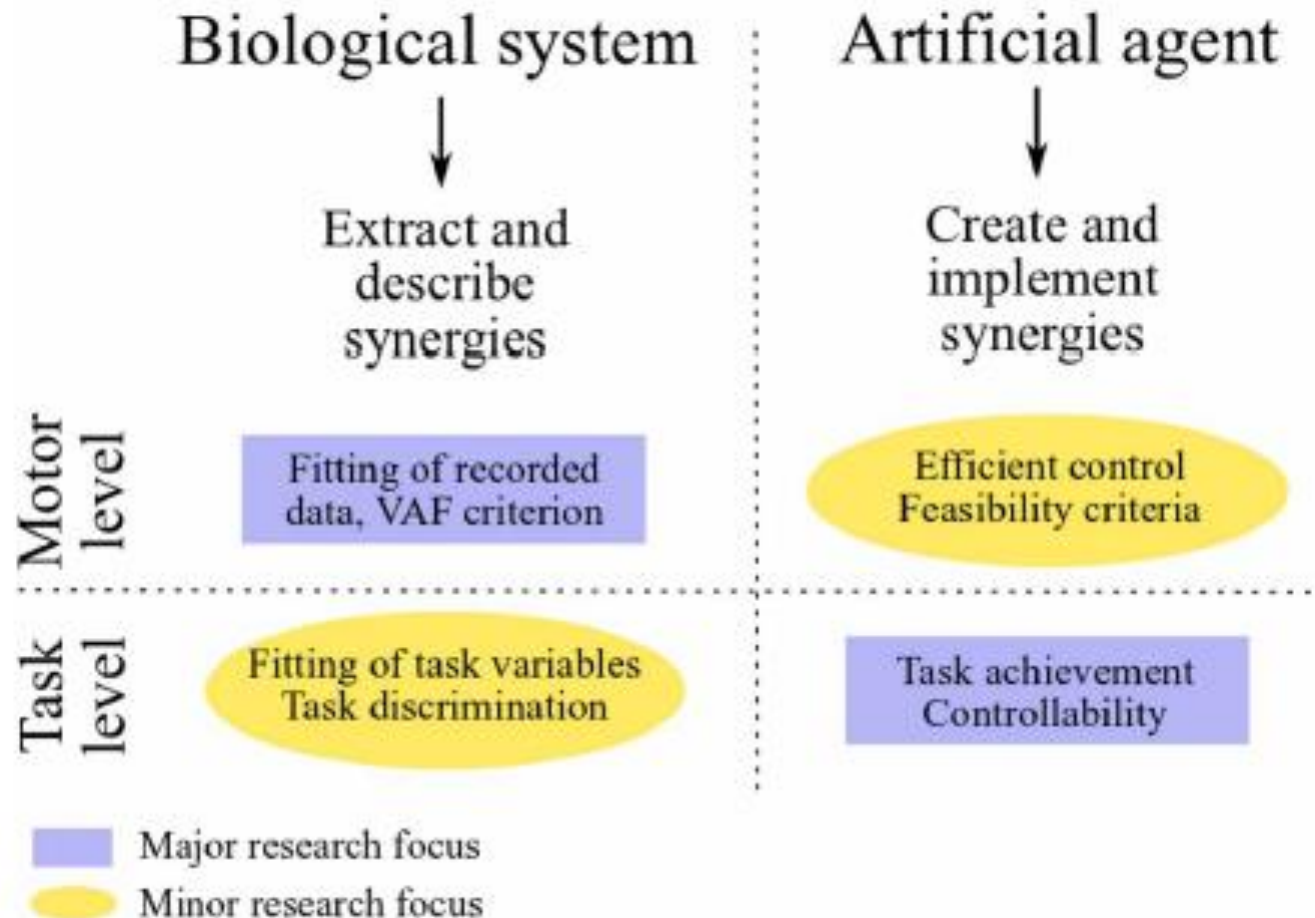
Parallels with robotics research



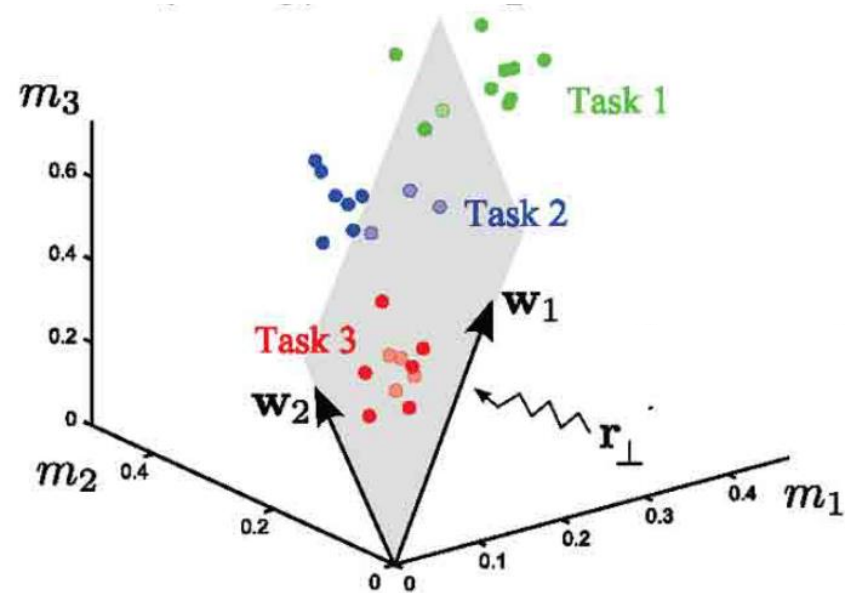
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Parallels with robotics research

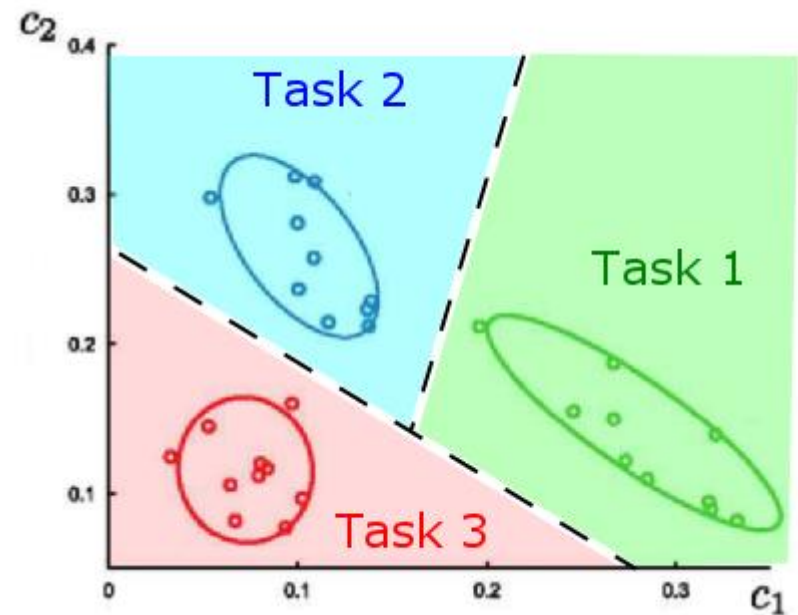


Population activity space



Low-dimensional representation of data

Modular space

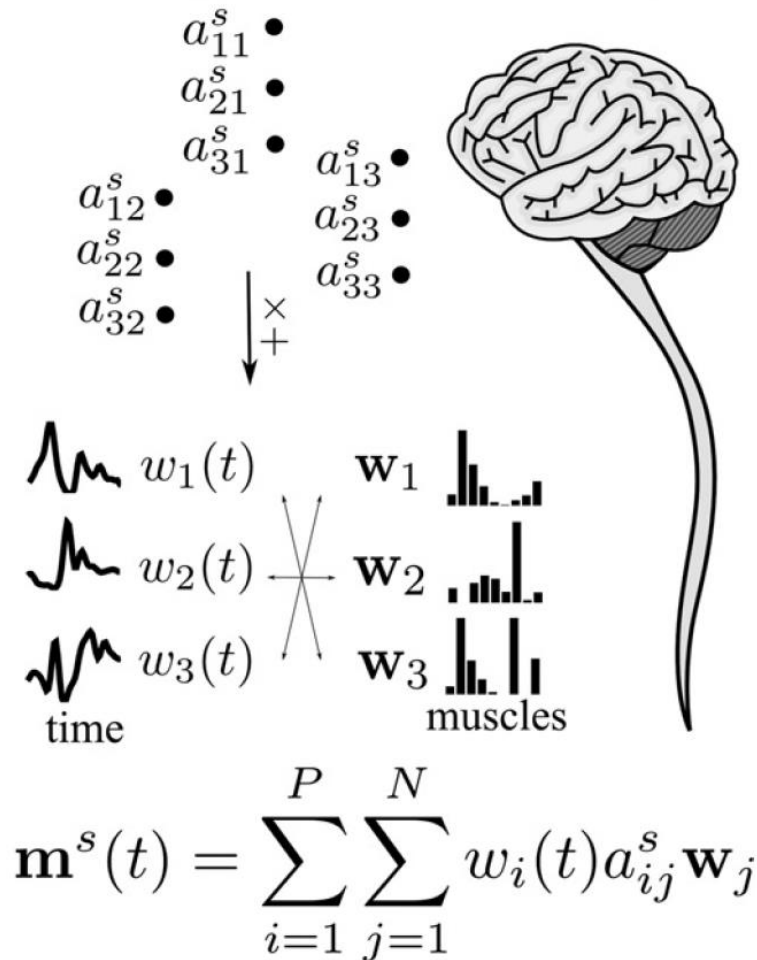


Behavioural relevance of representation

A unifying model of modularity



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Invariant temporal & spatial modules

Trial-by-trial variability is accounted for by scalar coefficients which combine the appropriate modules to perform the task at hand

Low-dimensional representation of muscle activity in single trials

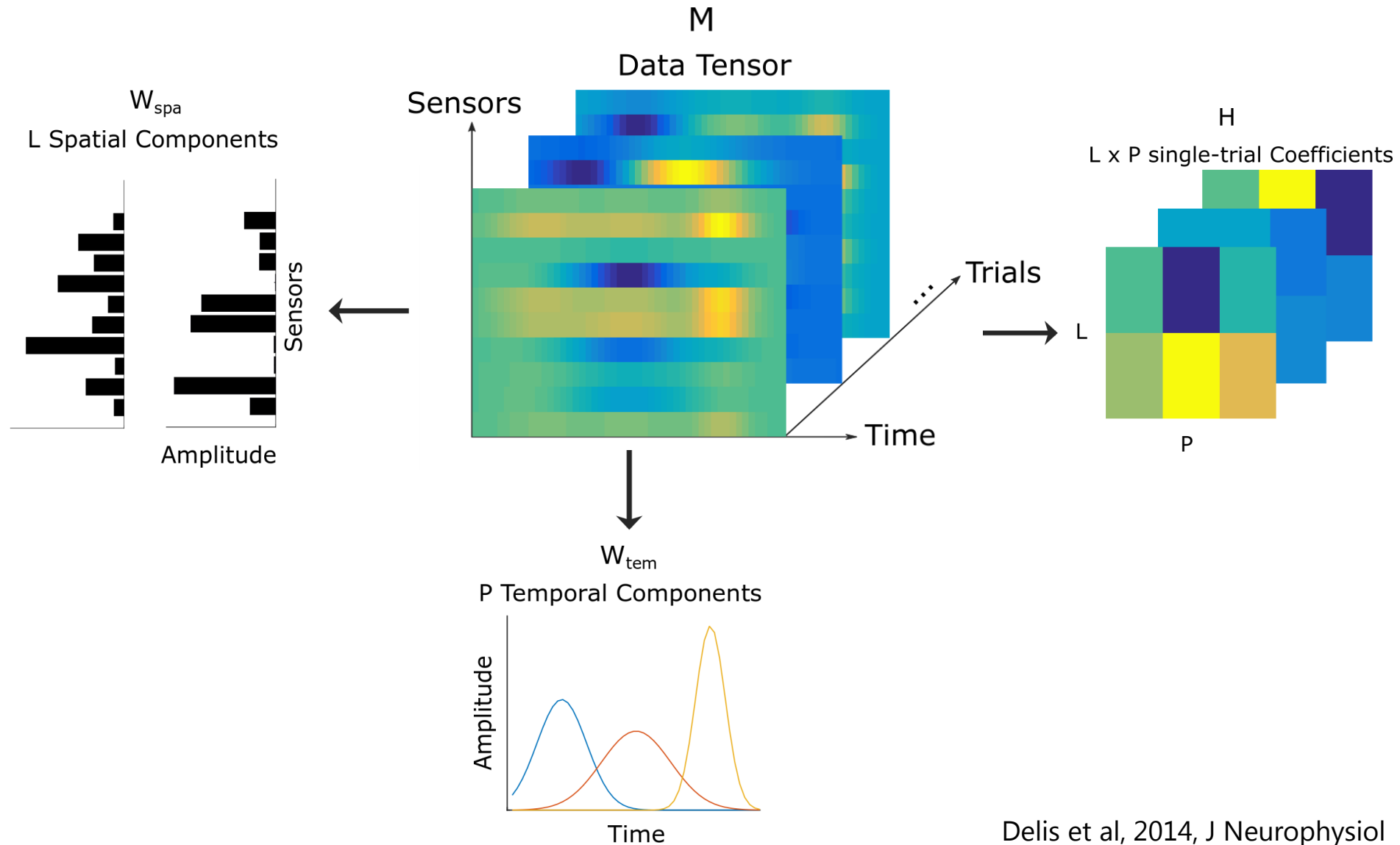
Encompasses existing modularity models

Best trade-off between data approximation, dimensionality and task discrimination

Space-by-time decomposition



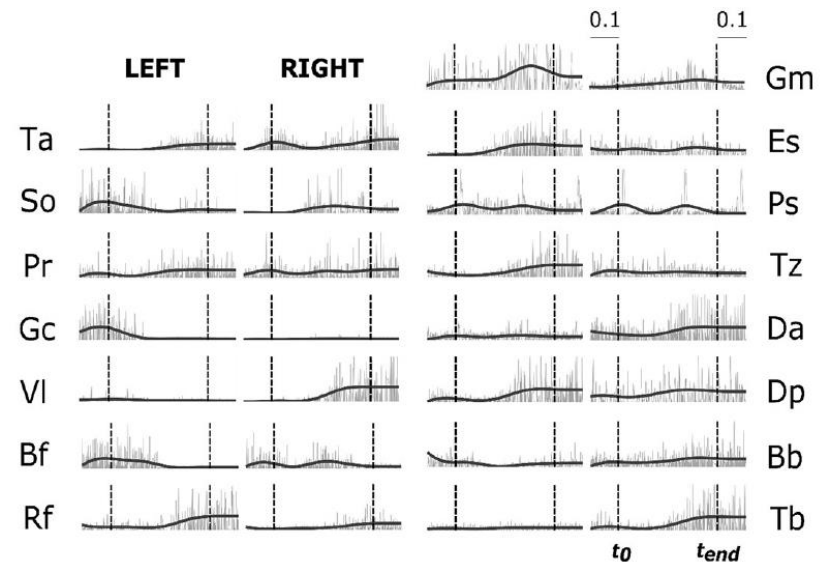
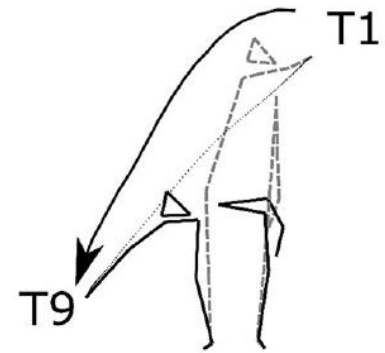
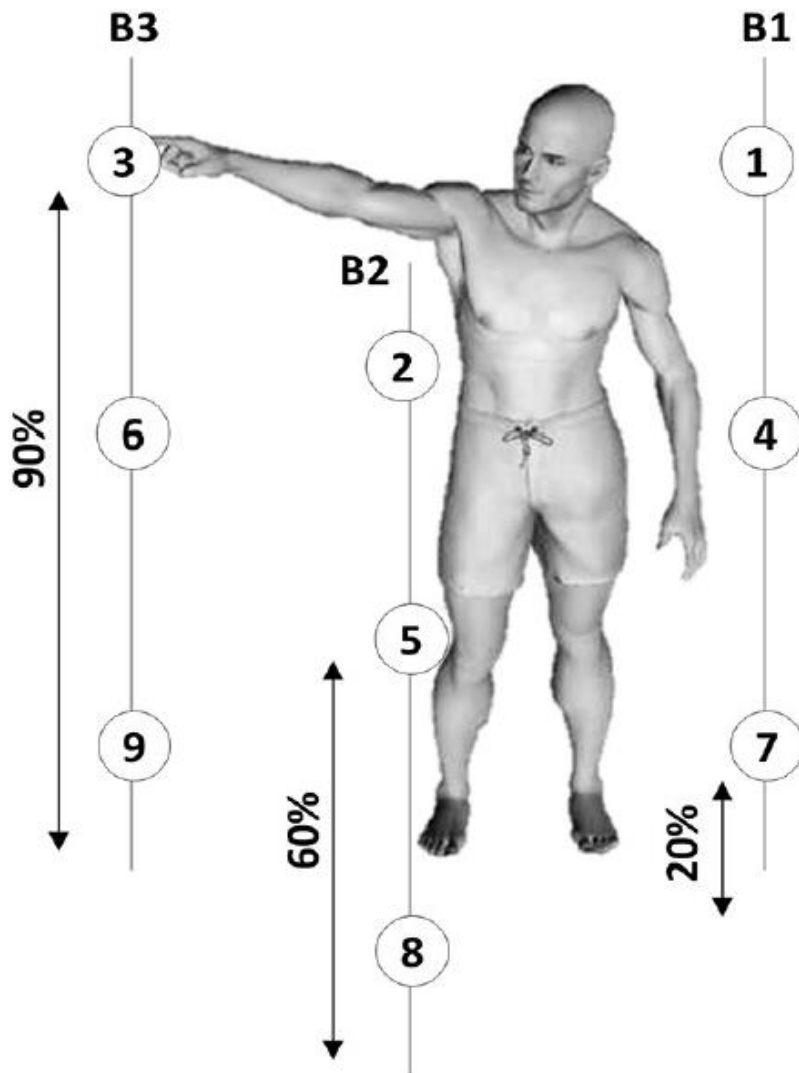
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Whole-body reaching task



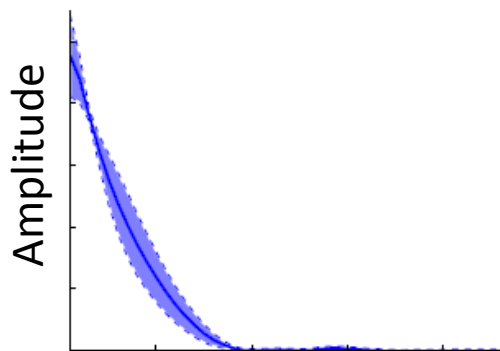
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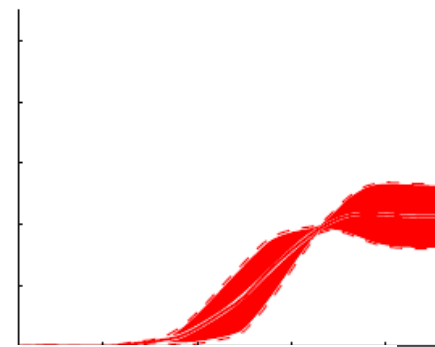
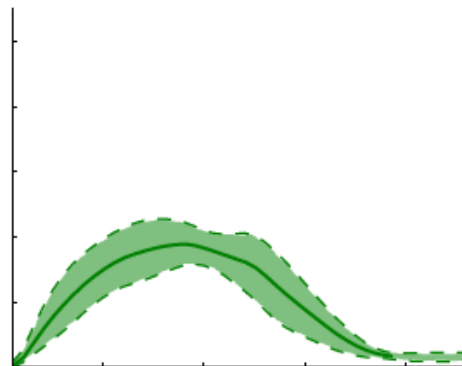


Temporal characterization of movement

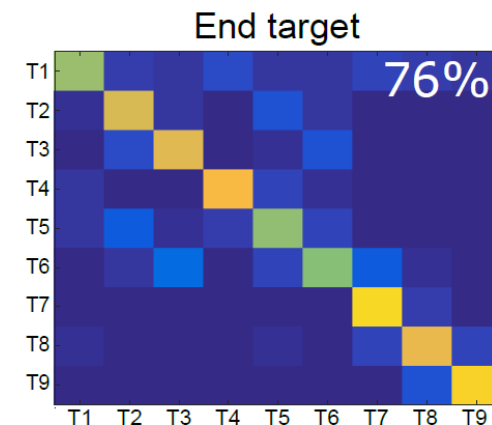
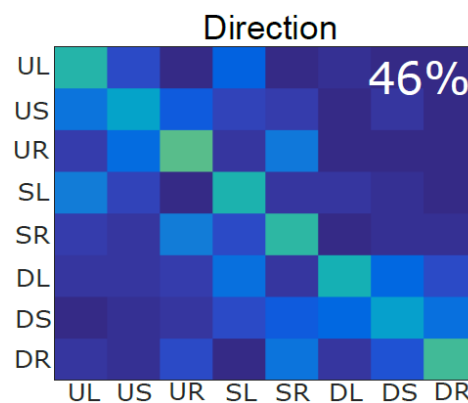
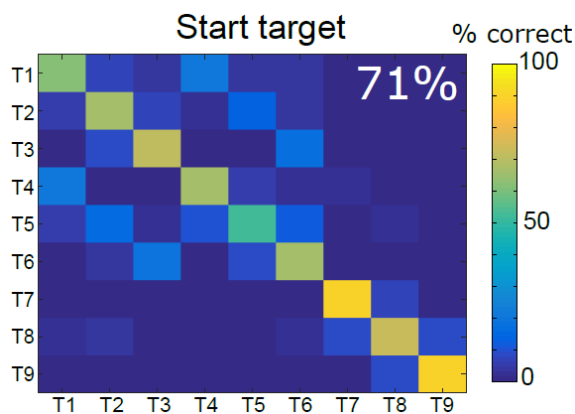
Temporal Modules



Time



Temporal Movement Features

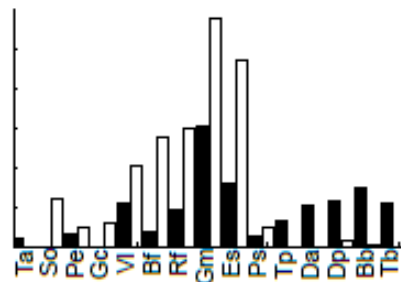
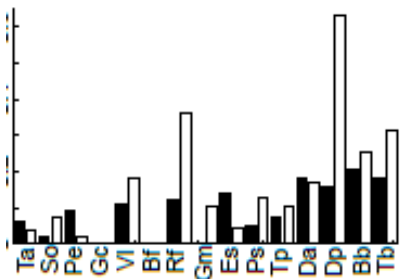
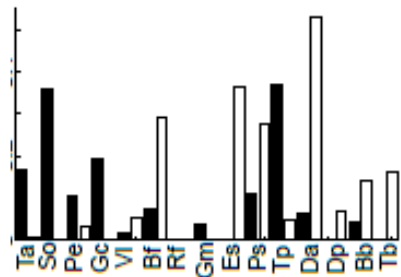


Spatial characterization of movement



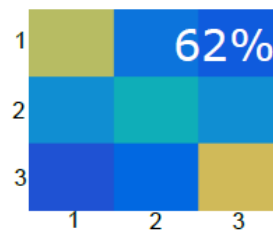
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Spatial Modules

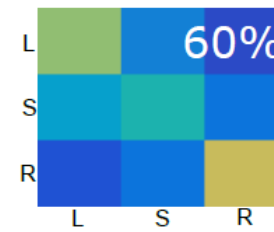


Spatial Movement Features

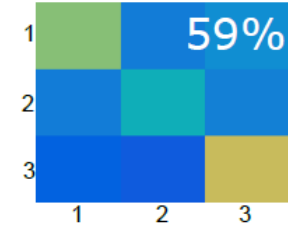
Start bar



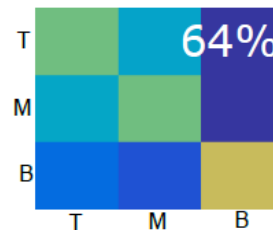
Horizontal direction



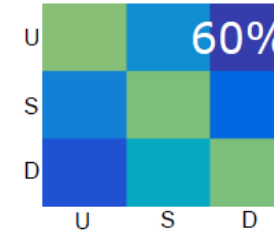
End bar



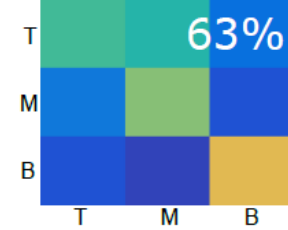
Start height



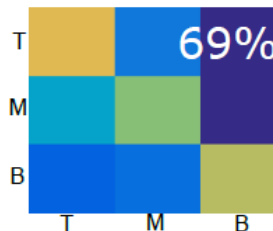
Vertical direction



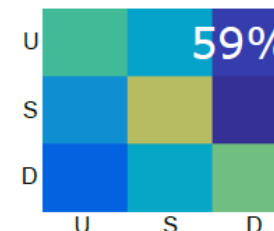
End height



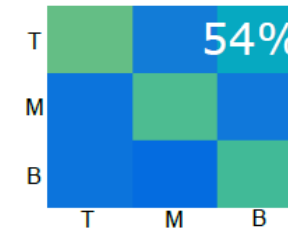
Start height



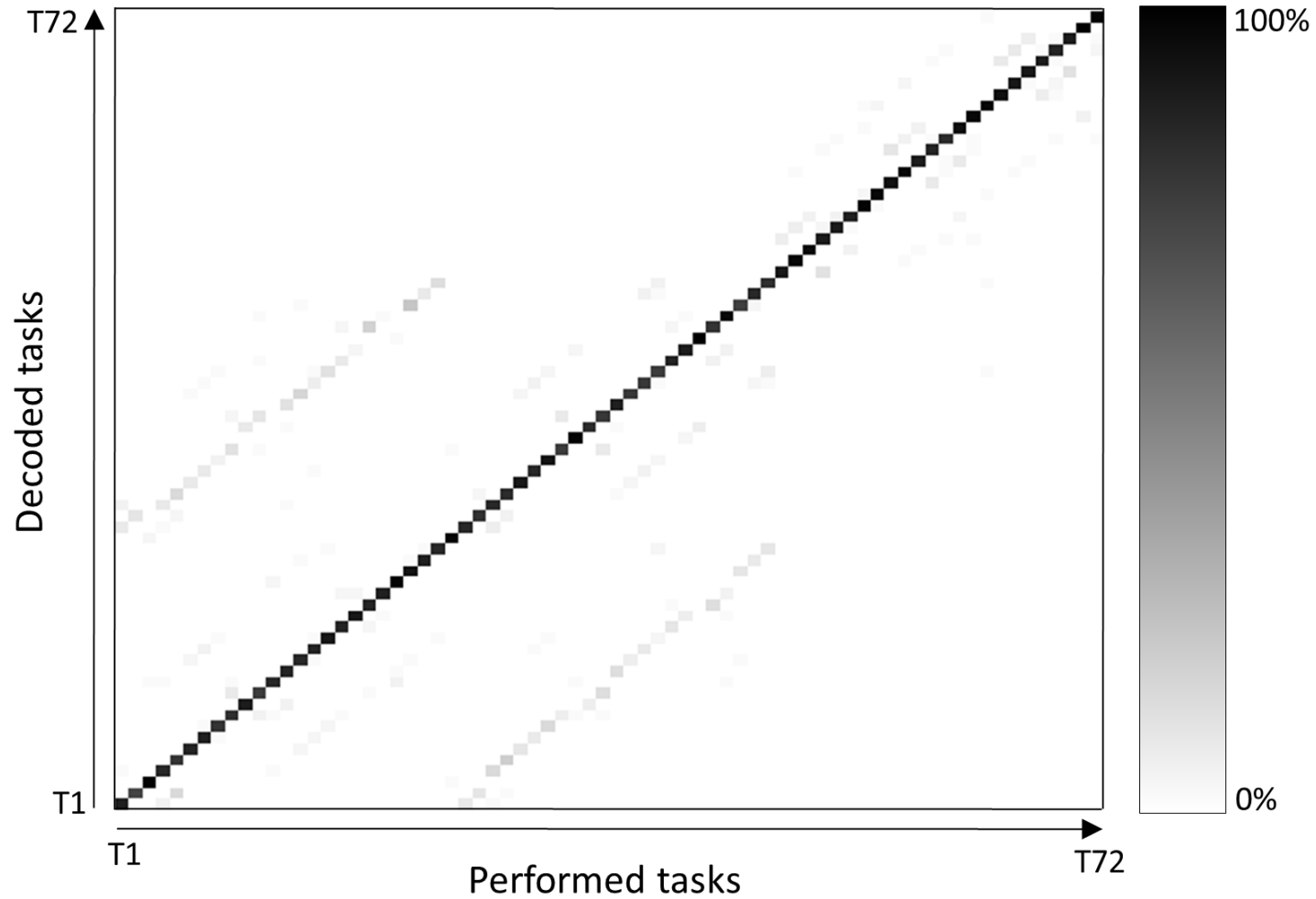
Vertical direction



End height



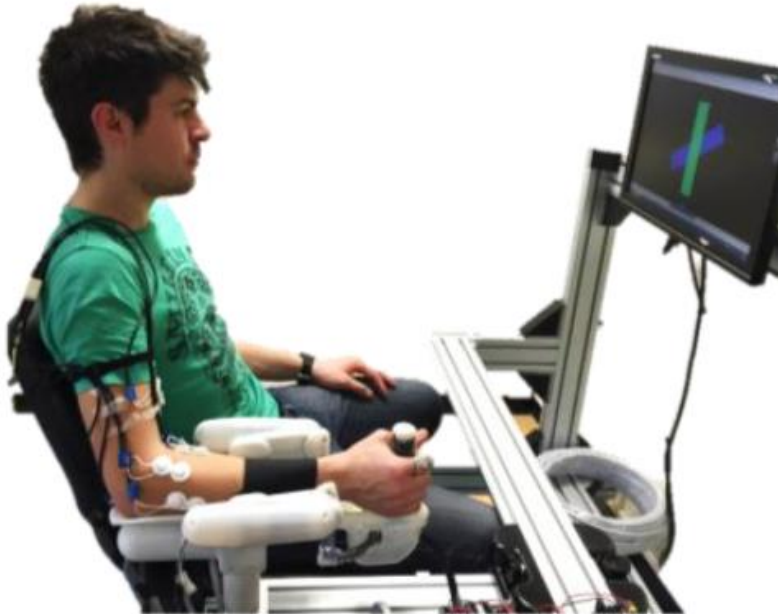
Task decoding



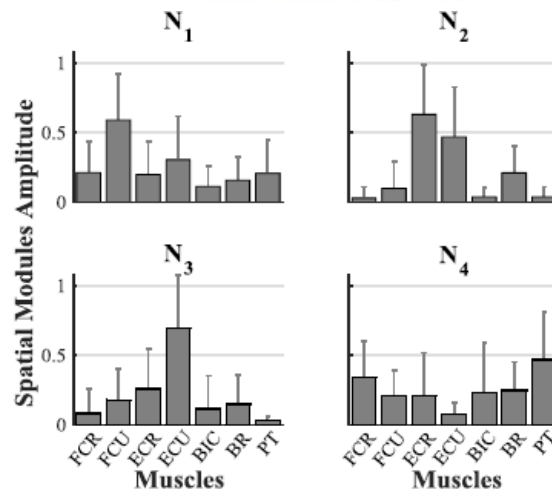
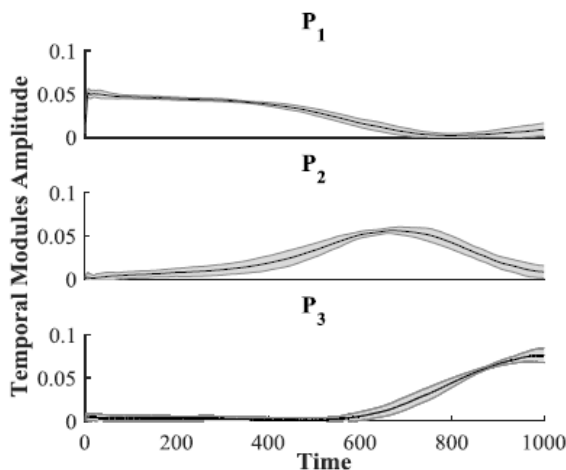
Application in rehabilitation



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- **First step:** Characterization of EMG patterns in healthy individuals
- **Future work:** Application to patient populations
 - Compare with healthy patterns → neural underpinnings of disorder
 - Use robot to provide assistive/resistive forces



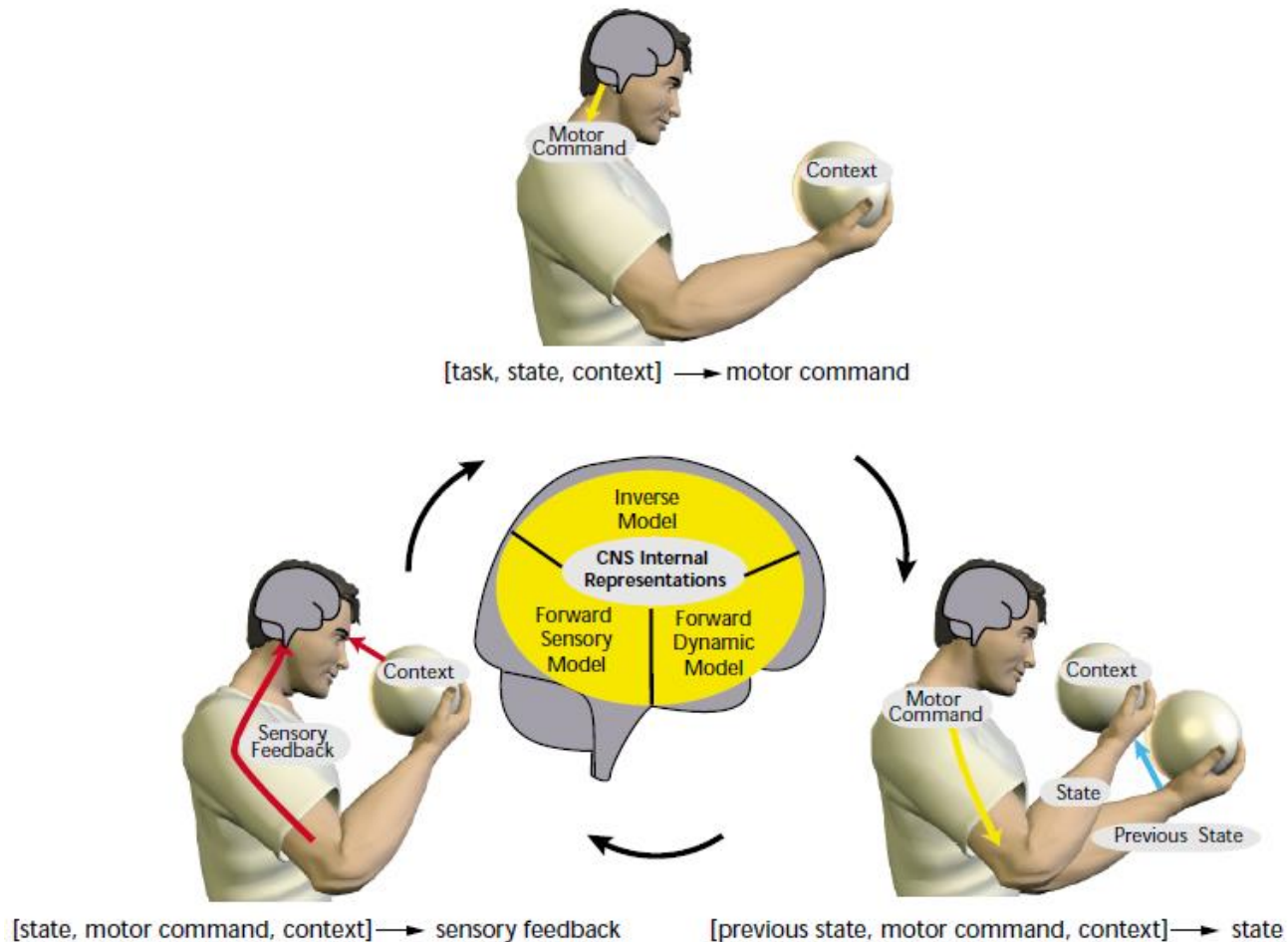
Advantages

- Few (~5) training trials suffice to learn the synergies
- Few single-trial parameters to control movement (independent of number of muscles or time points)

The sensory-motor loop



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from Wolpert & Ghahramani, 2000, Nat Neurosci

Active sensing



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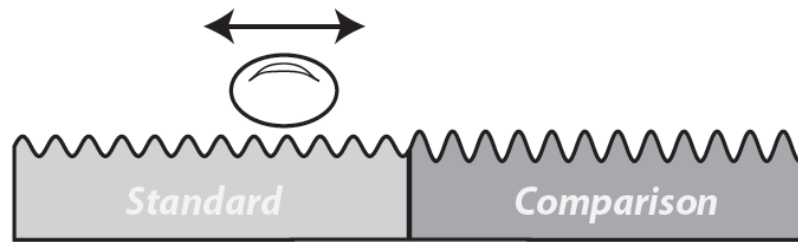
Active tactile decision-making task



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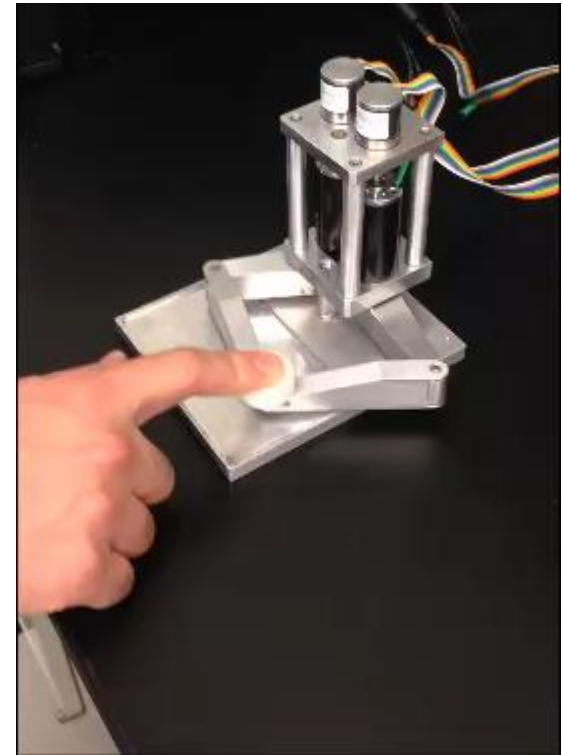
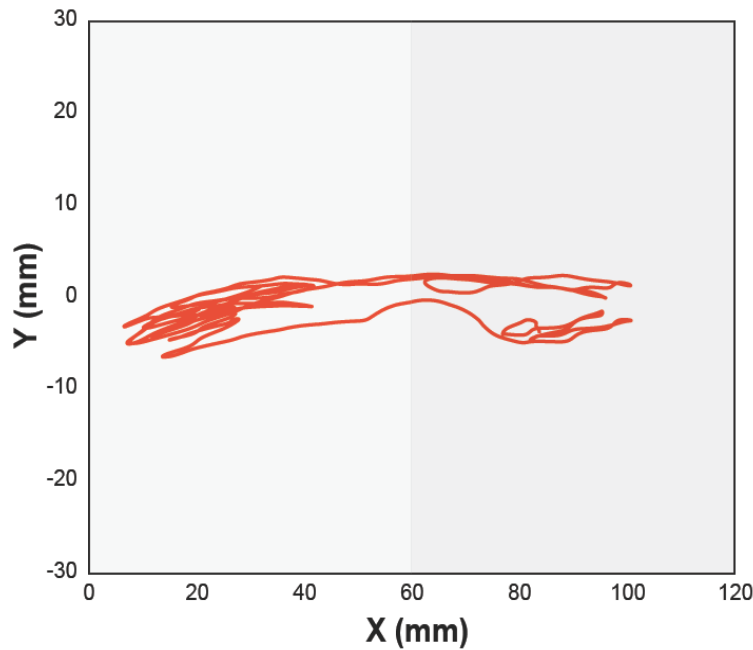


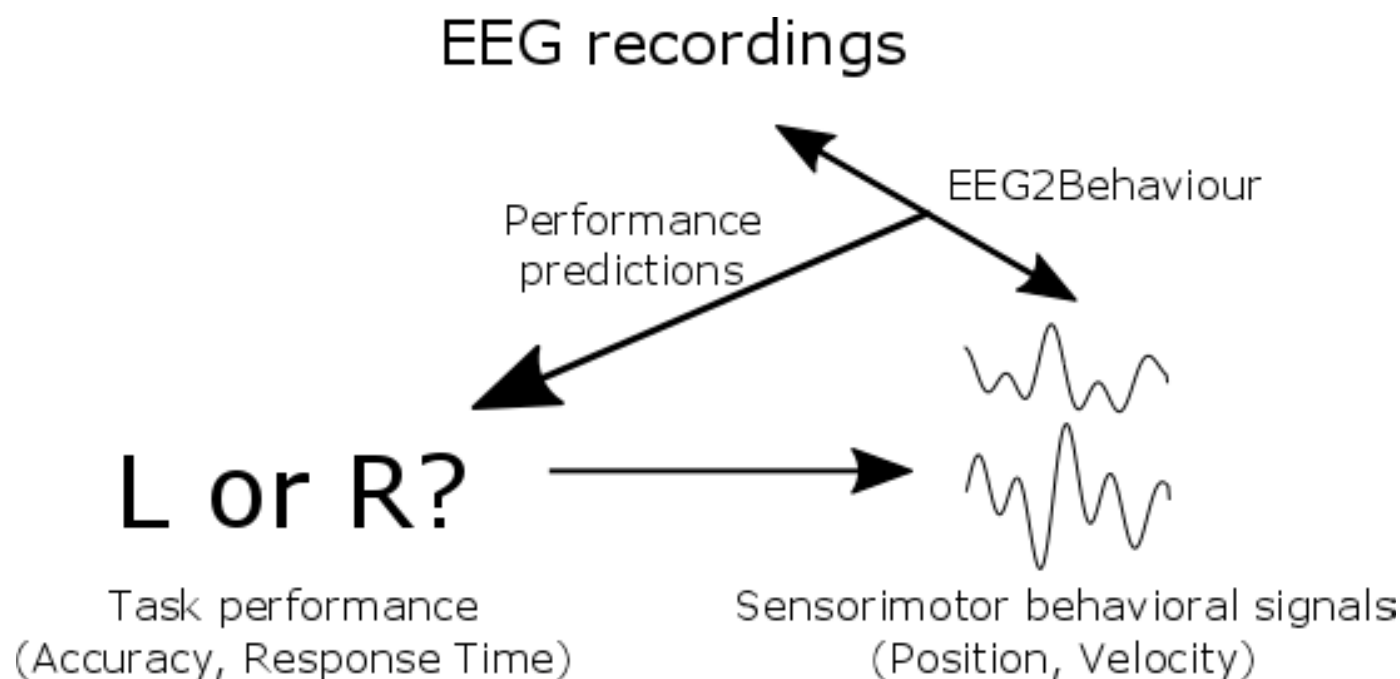
Haptic device



Virtual surface

2AFC task

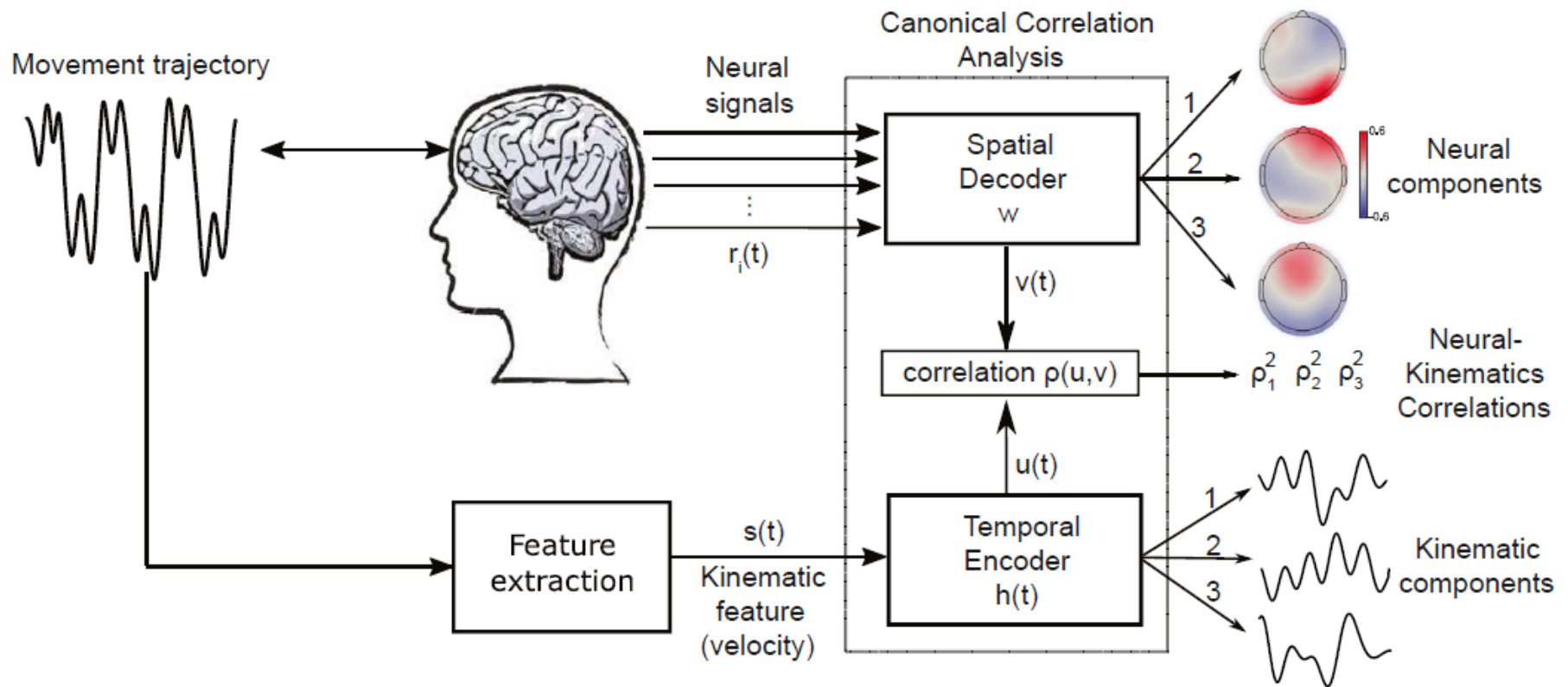




EEG2Beh for multimodal dimensionality reduction



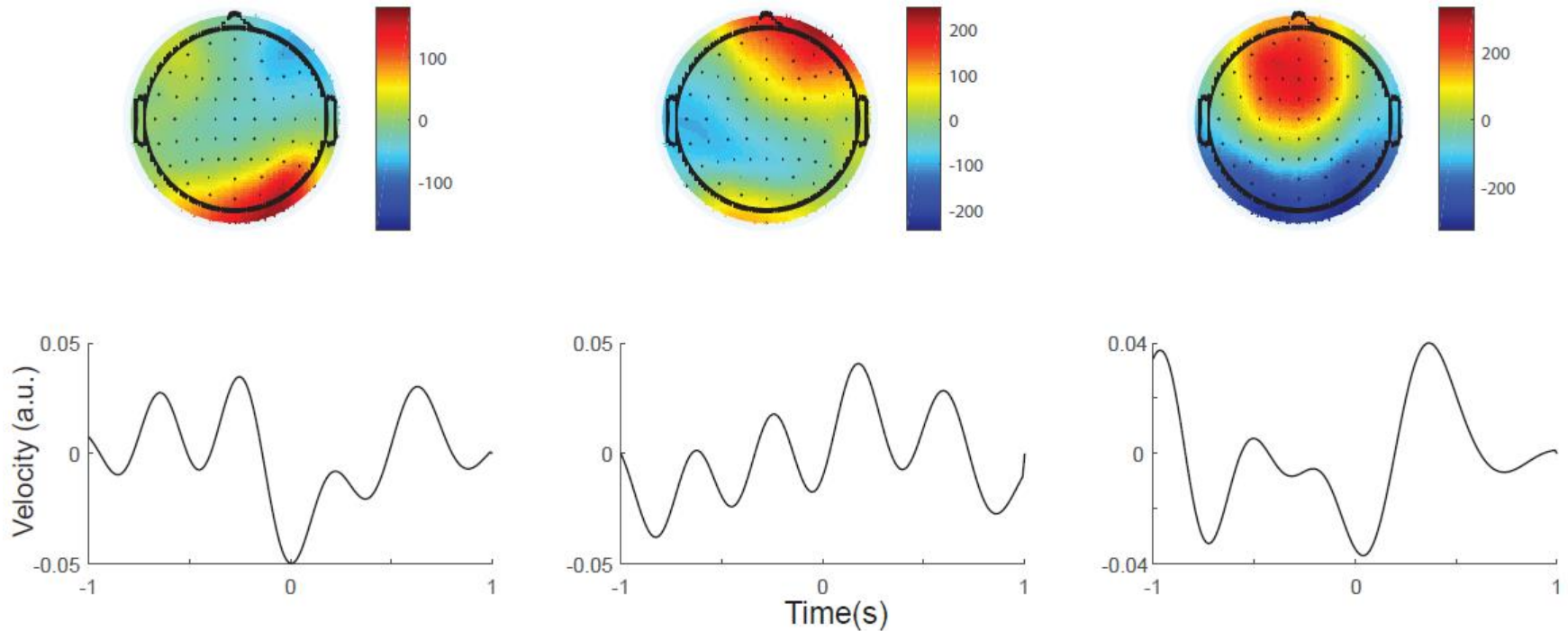
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Three brain-movement couplings



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Hypothesis: Active sensing behaviour, and the underlying neural activity, provide a window into the processes leading to decision formation

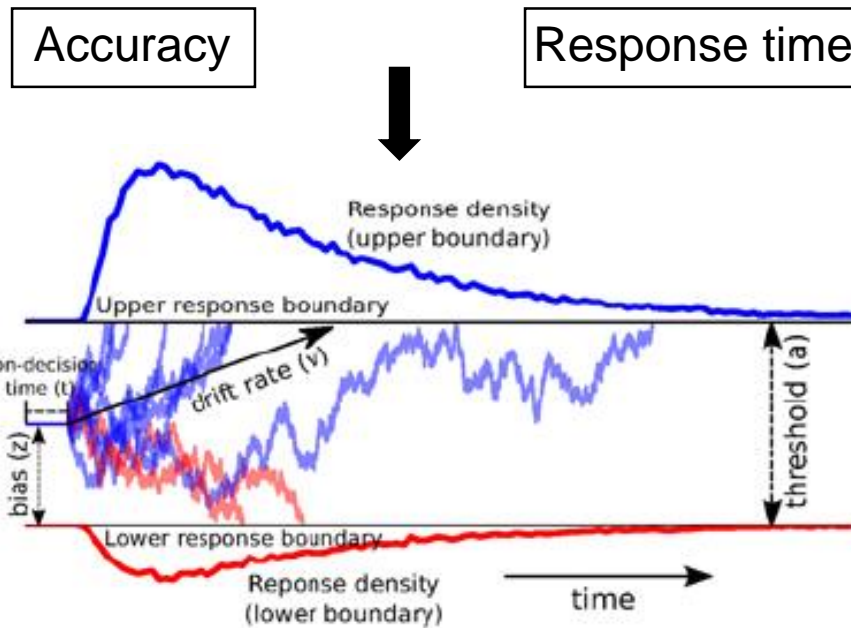
Question: Are the involved processes modulated by the brain-movement couplings?

Modelling decision-making performance



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Performance measures



Constituent processes

Drift rate δ

Evidence integration rate

Non-decision time τ

Sensory processing & motor response

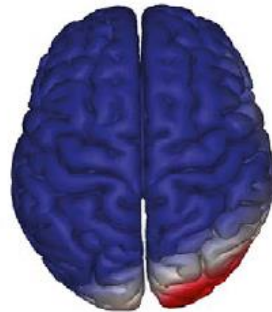
Decision threshold α

Amount of evidence required

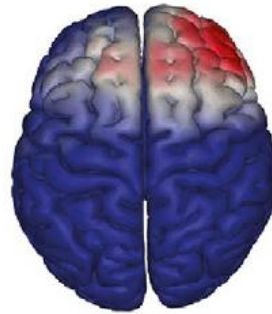
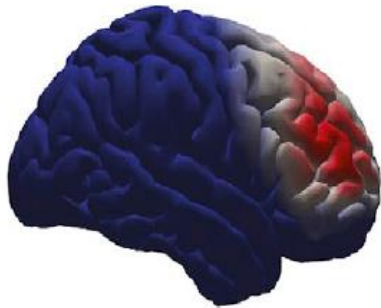
Neural correlates of distinct processes



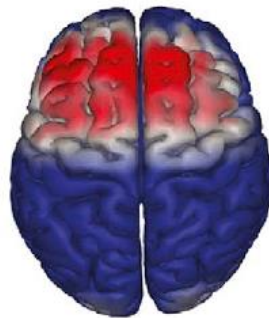
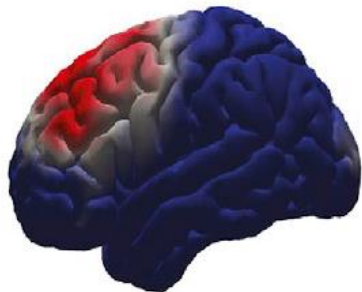
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Sensory processing



Evidence accumulation

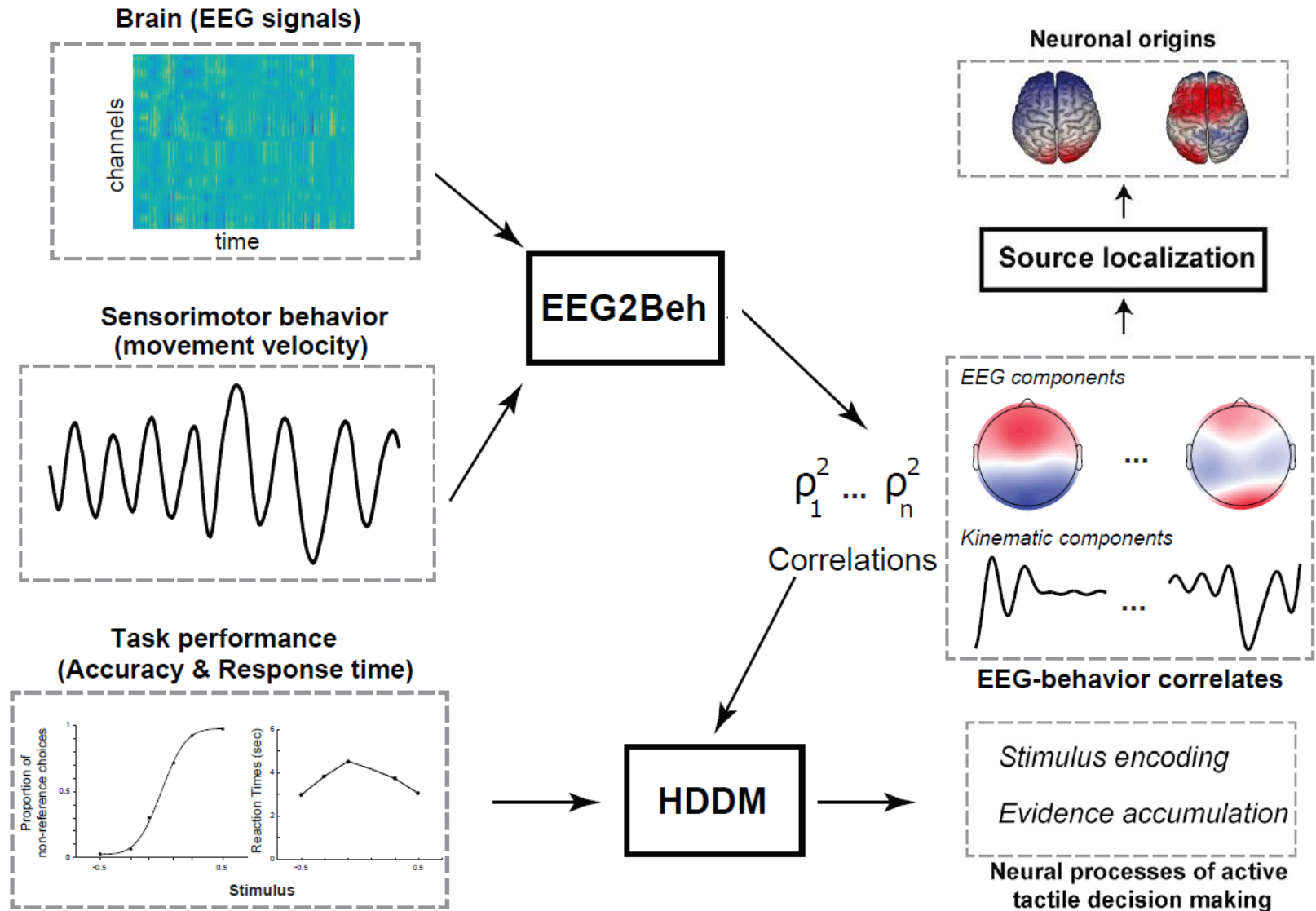


(Movement planning/execution)

Computational framework



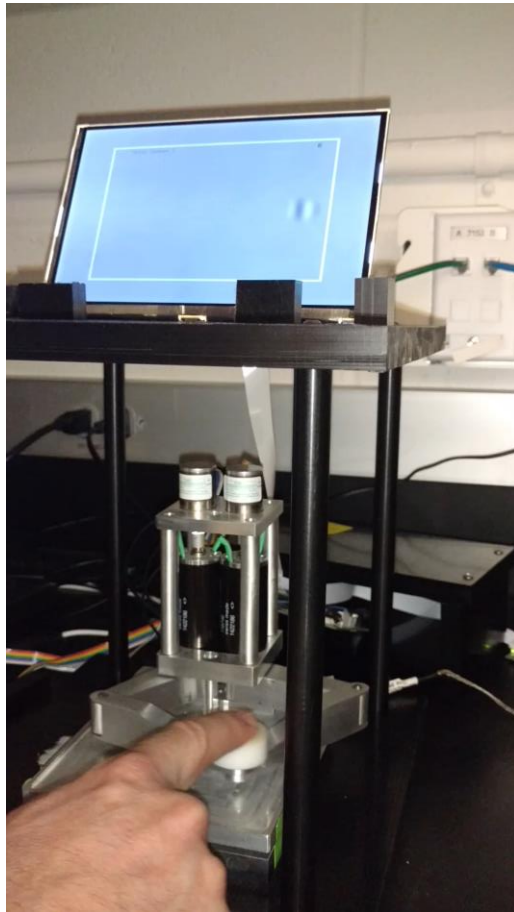
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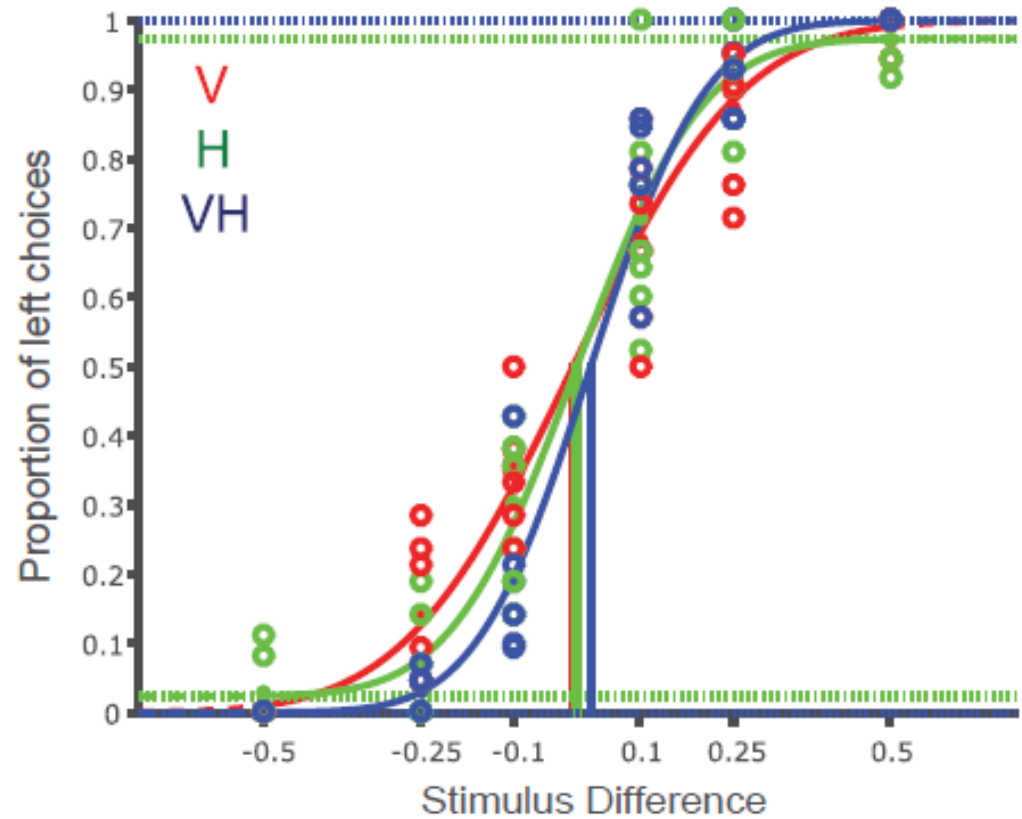
Multisensory task



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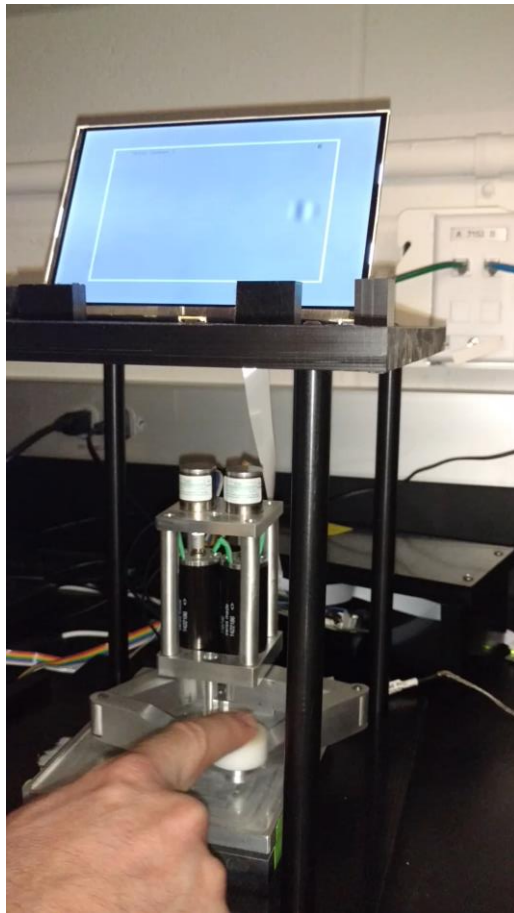
Higher Accuracy in VH



Multisensory task

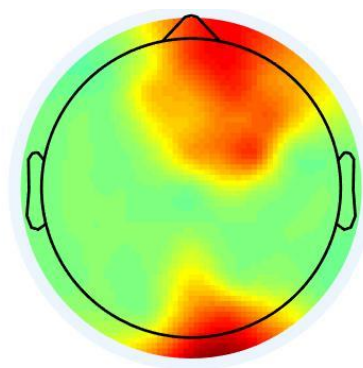


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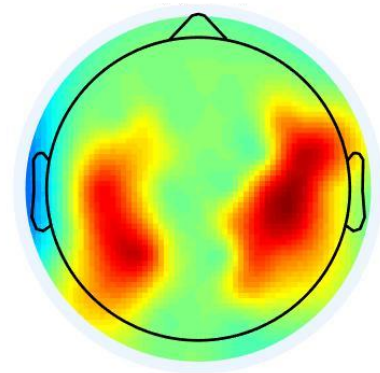
Mutual Information Framework

$$I_{int} = I(VH; K) - I(V; K) - I(H; K)$$



$$I_{int} < 0$$

Redundancy



$$I_{int} > 0$$

Synergy

Acknowledgements



Stefano Panzeri



Thierry Pozzo



Bastien Berret



Pauline Hilt



Marianna Semprini



Paul Sajda



Qi Wang



Robin Ince



Jacek Dmochowski



Thank you!
&
Questions?

