Exploring adaptive BCI methods to favor user learning and flow state



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The user changeable nature is an important cause of EEG signal variability. To improve BCI usability, we thus need to cope with the user states and fluctuations, using adaptive methods. We created a framework to formalize and guide the conception and instantiation of adaptive BCI systems. Important aspects that structure this framework include the various factors and time scales that characterize the user and task model (Fig. 1). We illustrate this framework with two concrete and well-known BCI examples.

	\bigcirc	BJECTIVE FUNC	TION	
★FLOW Optimization	◀		\rightarrow S	PEED/ACCURACY
Motor Imagery				Optimization
			(Top-down) control flow	Dage Croeller

 Flow (2) requirements:
 An engaging and ludic environment
 TuxRacer video game



Adaptation of the task difficulty
Background music
Clear goals and immediate audio and visual

feedback



P300-Speller

• Appealing to the **Active Inference**, probabilistic framework (3)



 Inference: Endow the machine with optimal, dynamic perception
 Optimal stopping



Active: Endow the machine with optimal action
 Optimal stopping & flashing

Fig.1. Generic framework presenting the relation between the user, the task and the signal processing pipeline. The user characteristics are arranged according to their degree of changeability in time. [1]

★ Motor Imagery Exp. Design (4)

2 (adapt vs no-adapt) by 2 (music vs no-music) mixed factorial design



Fig.2. Protocol: 10min calibration, 6 x 3min Tux Racer (music or no-music condition), followed by questionnaires: for Flow (EduFlow2), for music (BMRI).

★ P300-Speller Exp. Design (5)

Simulations to evaluate the added value of **optimal** *flashing*



action / perception





Task model

User model -

RESULTS:

Flow correlates with performance offline;
Higher state of Flow in *adapt* and *music* condition;







Fig.3. Machine is endowed with a model of the task and the user

Optimal flashing is the ability to choose those group of items to flash which optimally reveal the target letters

RESULTS:

 Optimal stopping & flashing showed increased accuracy in reduced time

Optimal stopping: 20.1±9 flashes / 80.6% accuracy
Optimal stop & flashing: 15.8±6 flashes / 85.2% accuracy

REFERENCES: [1]Mladenovic et al. (2017) Chapter 33, BCI Handbook:Technological and Theoretical Advances; [2] Csikszentmihályi (1990). Harper & Row.; [3] Friston et al. (2006), J. Physiol.-Paris, vol. 100, no. 1–3, pp. 70–87. [4] Mladenovic et al. (2017) GBCIC; [5] Mladenovic et al. (2017) NaT'17

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