

Closed-loop adaptation of brain-machine interfaces using error-related potentials and reinforcement learning

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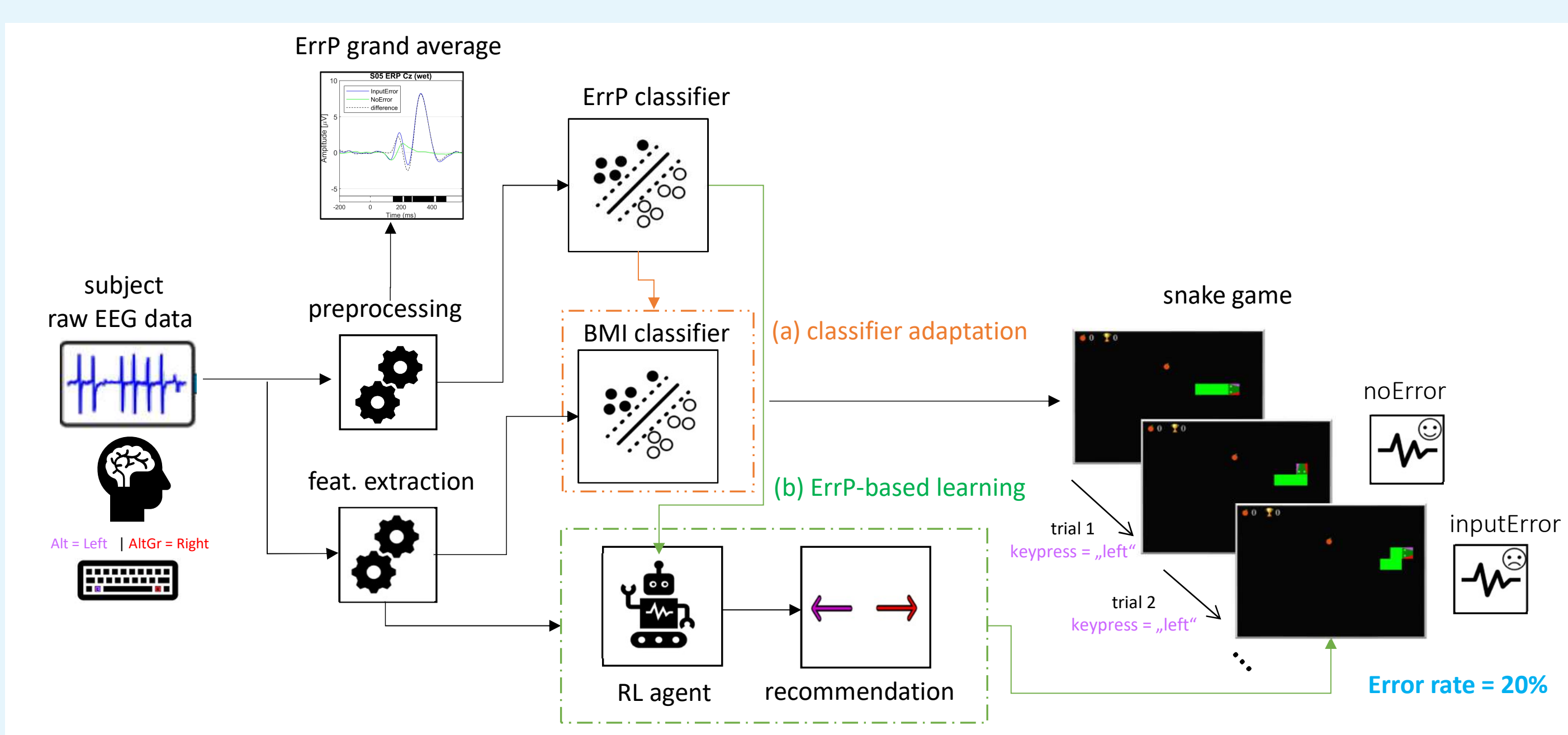
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Motivation

Neurorehabilitation devices can be used to help patients restore the lost mobility of upper-body limbs caused, e.g., by a spinal cord injury or a stroke. Given the long calibration sessions and decrease in decoding performance, the development of systems that adapt to the individual patient's needs is of particular relevance. Error-related potentials (ErrPs) are elicited in the human brain as a

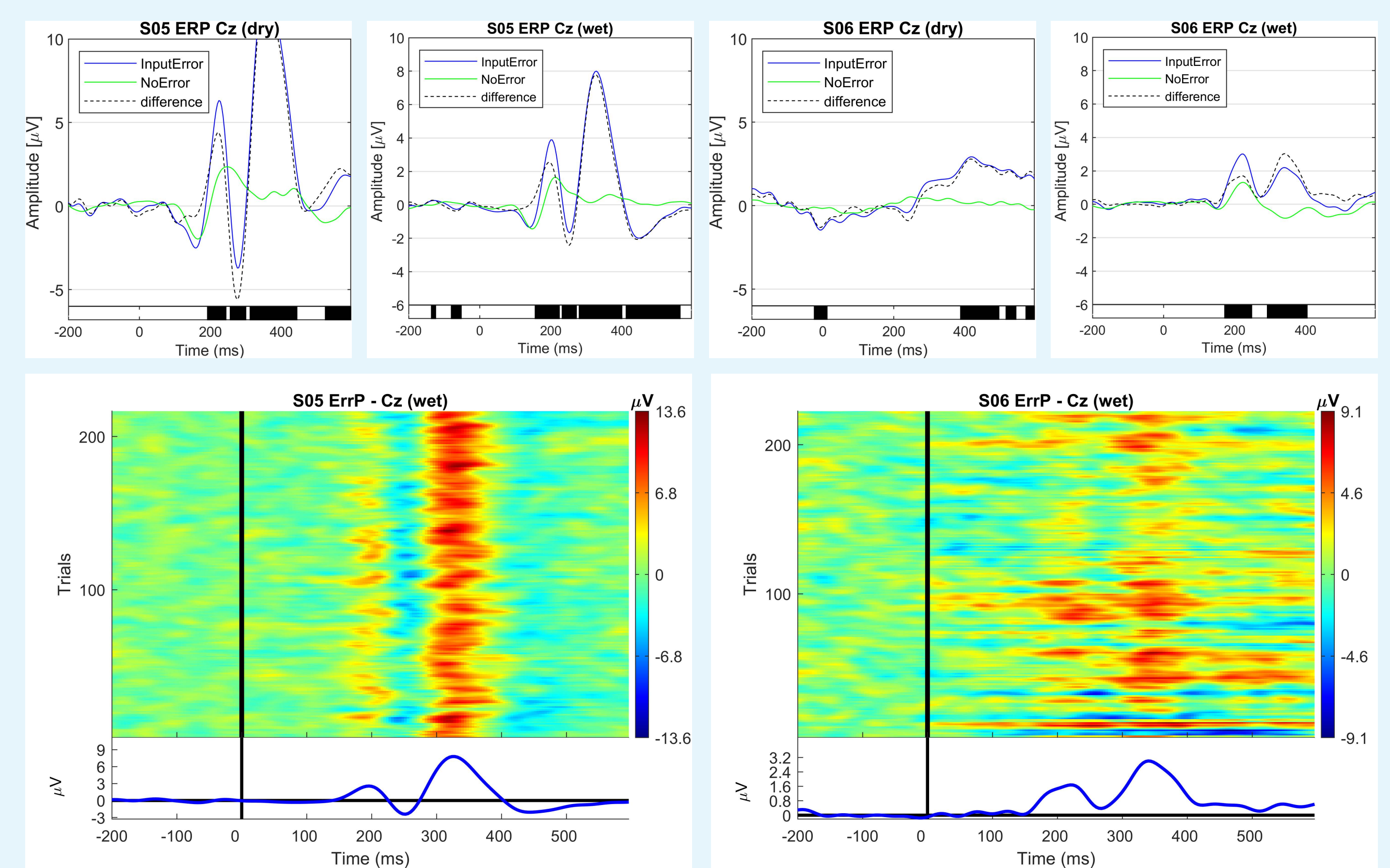
consequence of both self-made and external system errors [1]. Different experimental paradigms can be used to generate such signals, and the ErrPs provide a convenient source of feedback to improve brain-machine interfaces (BMIs) with no extra workload for the subject [2]. However, their effective use depends on their accurate detection, which is a limitation in current ErrP-based BMIs.

Experimental Setup



- **EEG devices:** Brain Products Xpress Twist: 32 active dry channels | Neuroelectrics Enobio: 8 gel-based channels
- **Subjects:** 6 participants (3 male, 30.2 ± 4.6)
- **Datasets:** 2 sessions of 10 runs with 120 trials each

ErrPs signal variability



Single-trial analysis of the measured ErrPs highlights both within- and inter-subjects variability in the signal.

Data Preprocessing

1. Artifact rejection (continuous data) rejection: bad channels and eyes blinks
2. Downsampling to 250 Hz
3. Notch and [1-20] Hz bandpass FIR filters
4. Automatic artifact
5. Common-average re-referencing
6. Baseline correction [-0.2-0]s

Outlook

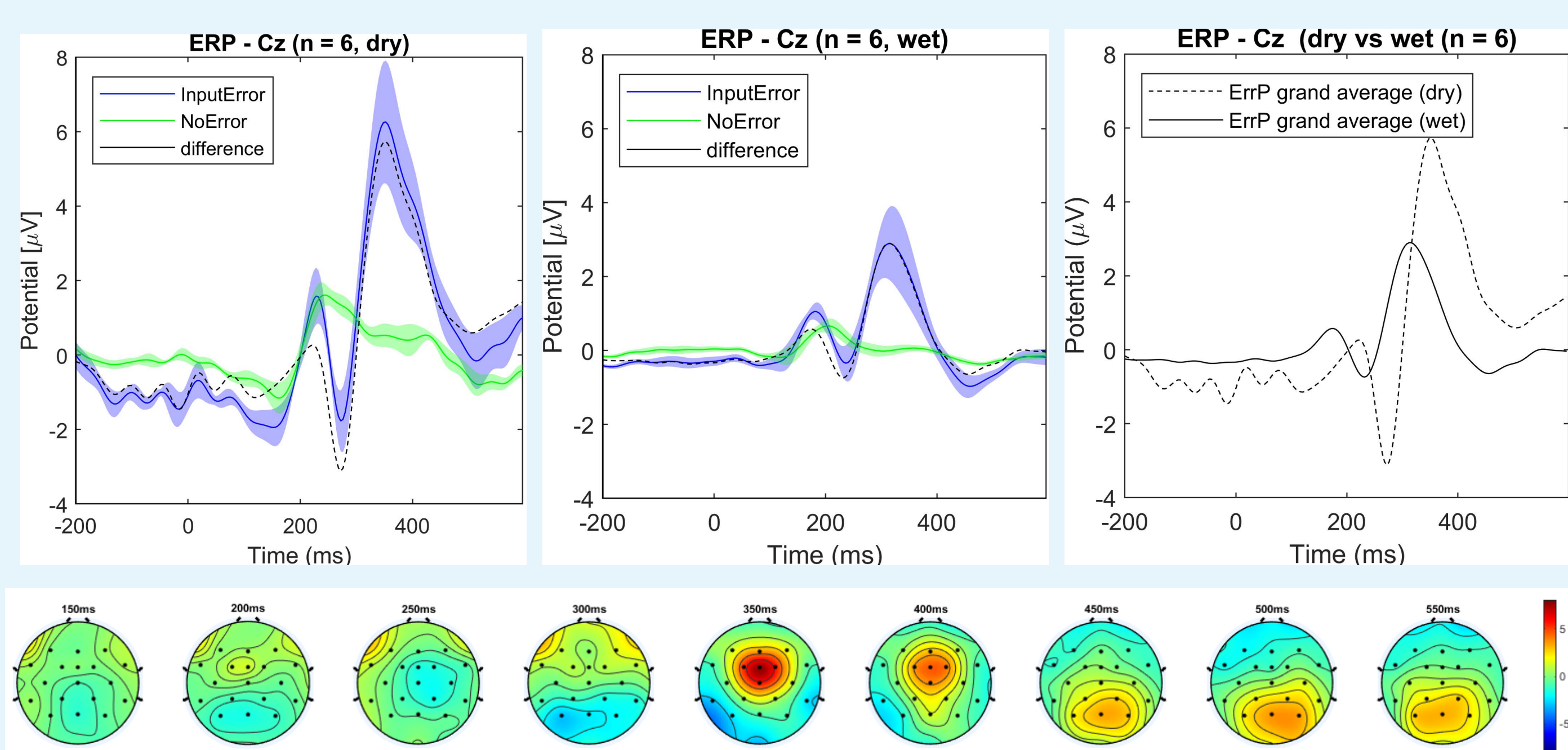
The specified protocol can be used to elicit a prominent interaction ErrPs but variability among subjects is still present. Factors affecting the occurrence of a significant ErrP signal have to be better understood, as well as the influence in the adaptation framework.

References

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- [3] P. W. Ferrez and J. d. R. Millán, "Error-Related EEG Potentials Generated During Simulated Brain-Computer Interaction," *IEEE Transactions on Biomedical Engineering*, vol. 55, no. 3, Mar. 2008

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Results



Overall grand averages at Cz: dry (left) and wet (right). ErrPs difference (error minus correct) grand averages at Cz combined (right). Measured ErrPs present waveshapes consistent with other studies [3].