

# The main challenges of Brain-computer interfaces: reliability and usability



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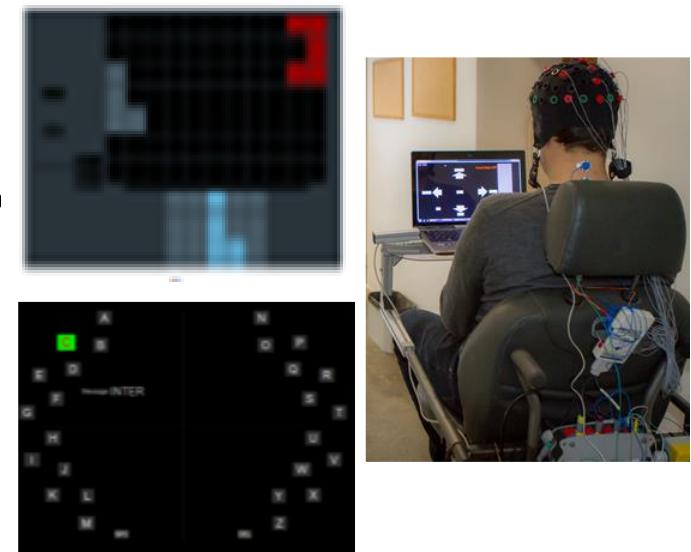
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# Context

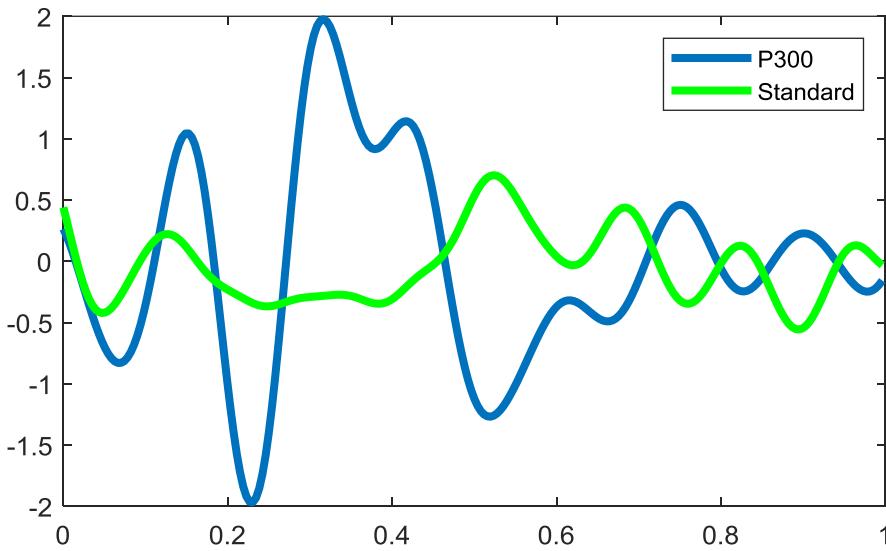
Brain-computer interface (BCI) is an emerging technology with successful applications for both healthy and disabled individuals.

Major EEG-based BCIs challenges:

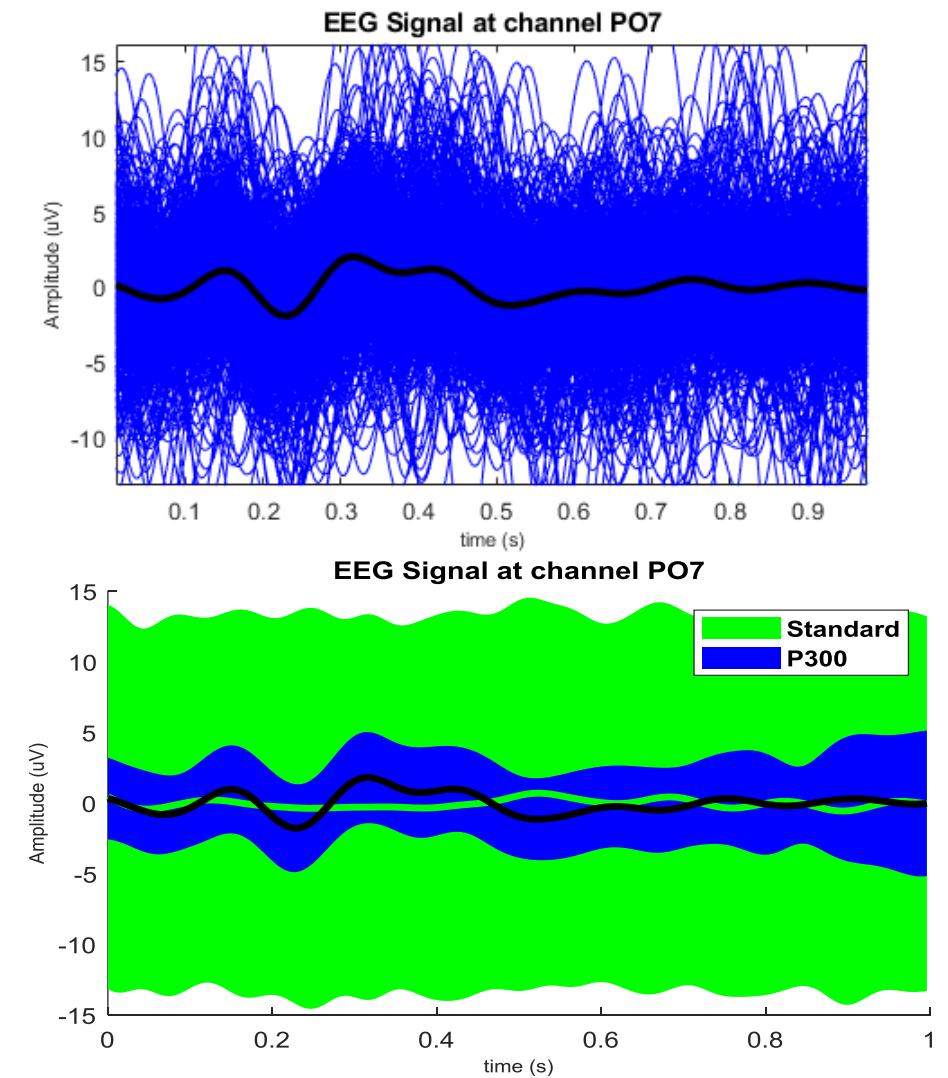
- EEG signal has low Signal-to-noise ratio (SNR);
- Great variability within and across subjects;
- High level of concentration and attention;
- Low reliability and usability.



# Context



Current BCIs systems have low reliability in recognizing the subject's intent.



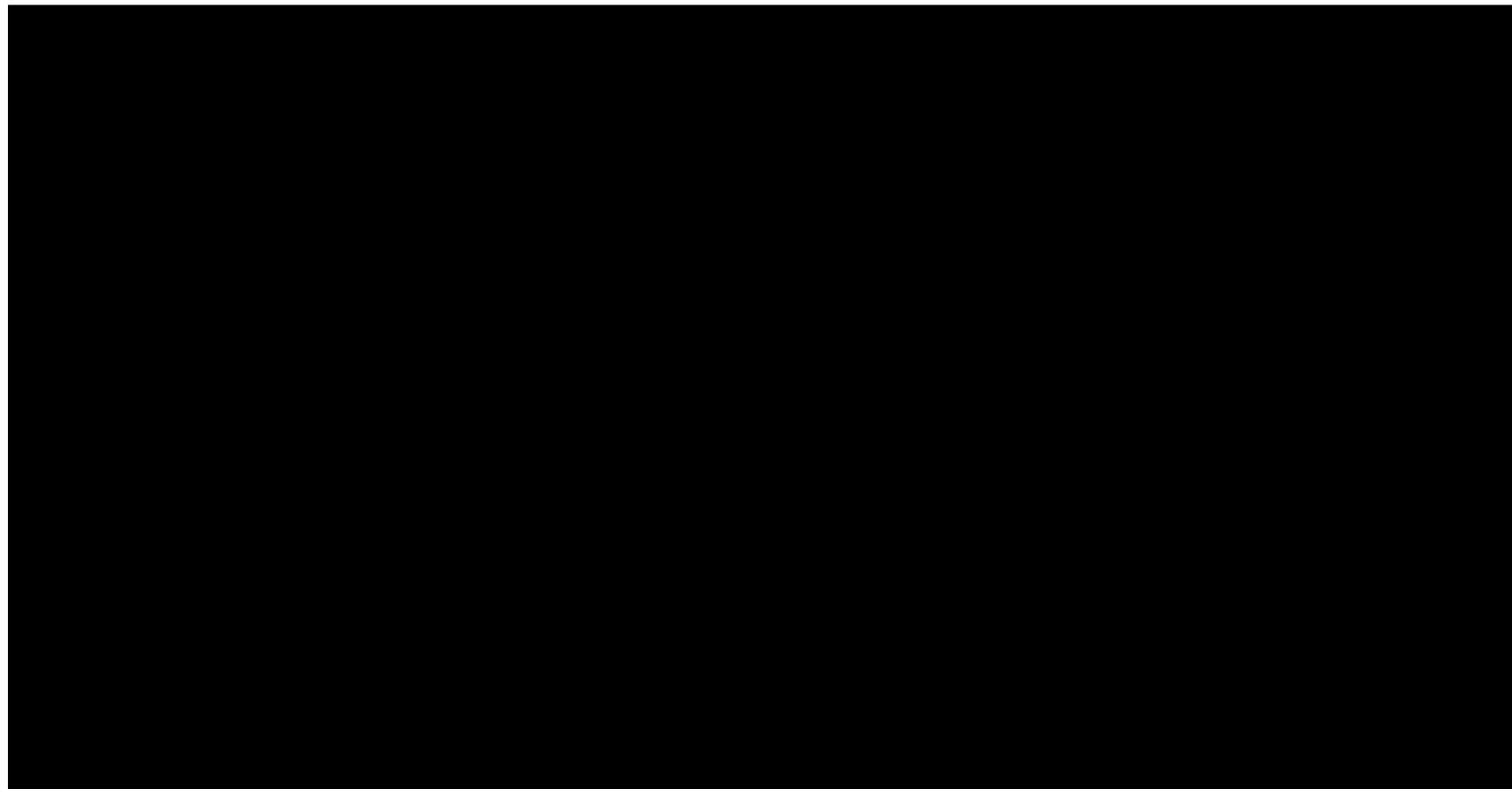
# Error-Related Potential (ErrP): Double Detection Approach

The ErrP is an event-related potential generated naturally in the brain when a person perceives a mistake.

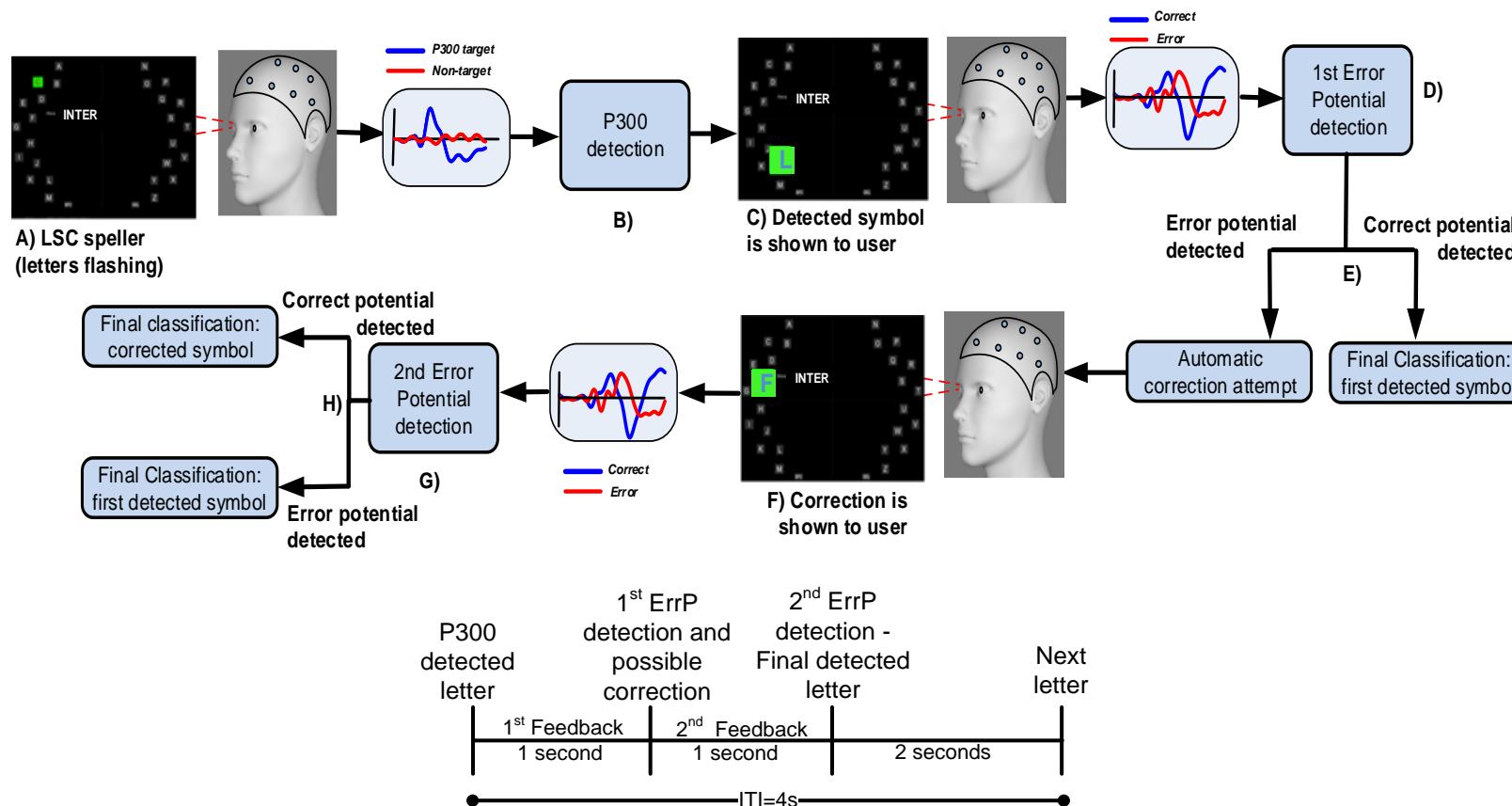
The main goal of this approach:

- Automatic detection of error-related potentials (ErrPs) in order to increase the reliability and transfer rates of brain-machine interfaces.
- The use of ErrPs in a closed-loop, allowing the user to change or confirm system decisions.

# On-line operation: demonstrative video



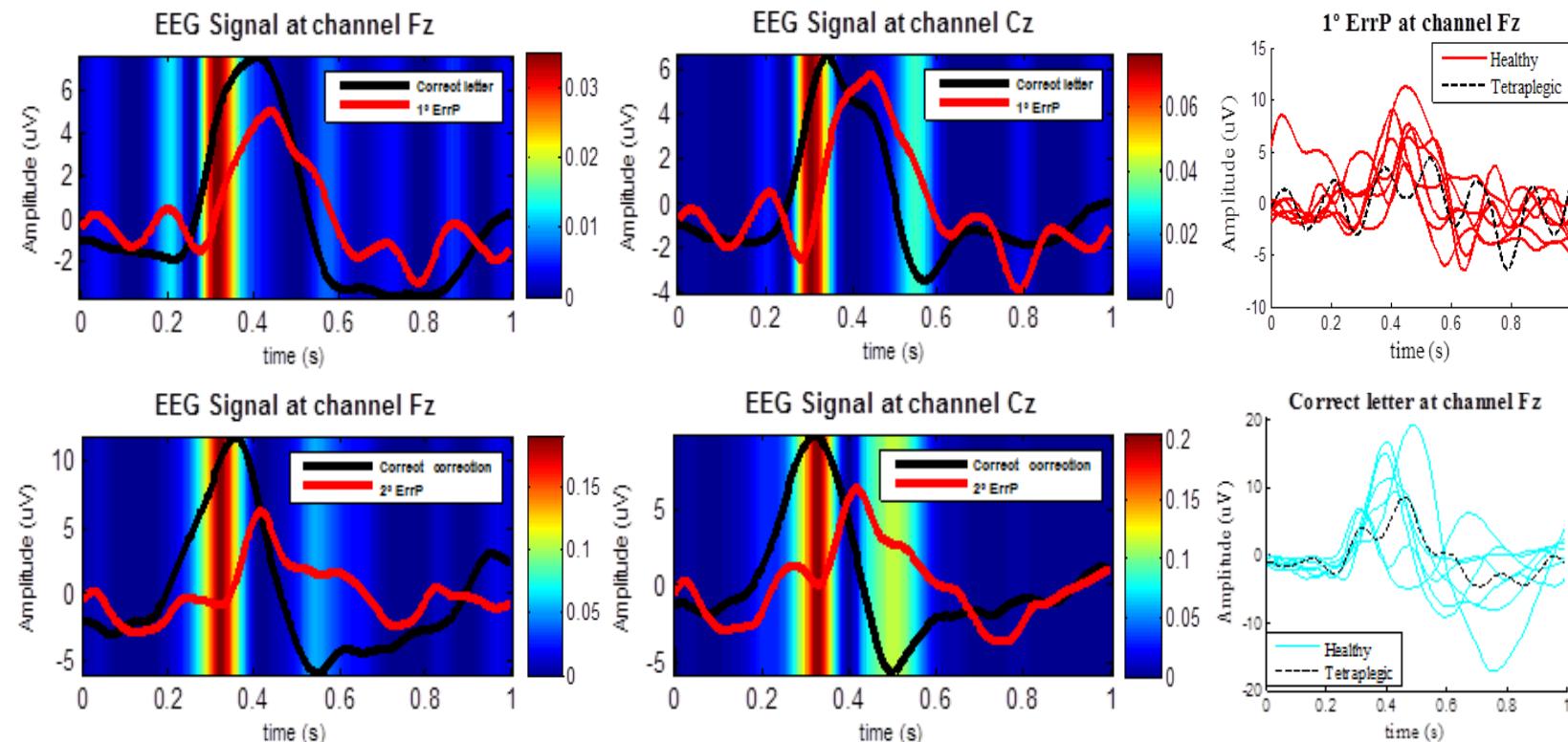
# Double ErrP Detection Approach



## Double ErrP detection:

- Use the information of 1st ErrP to perform the automatic correction;
- Use the information of 2nd ErrP to confirm the correction.

# Evoked potentials after correct and wrong feedbacks



- The positive feedback is different from the ErrP.
- The waveform of 1st and 2nd ErrP is slightly different.

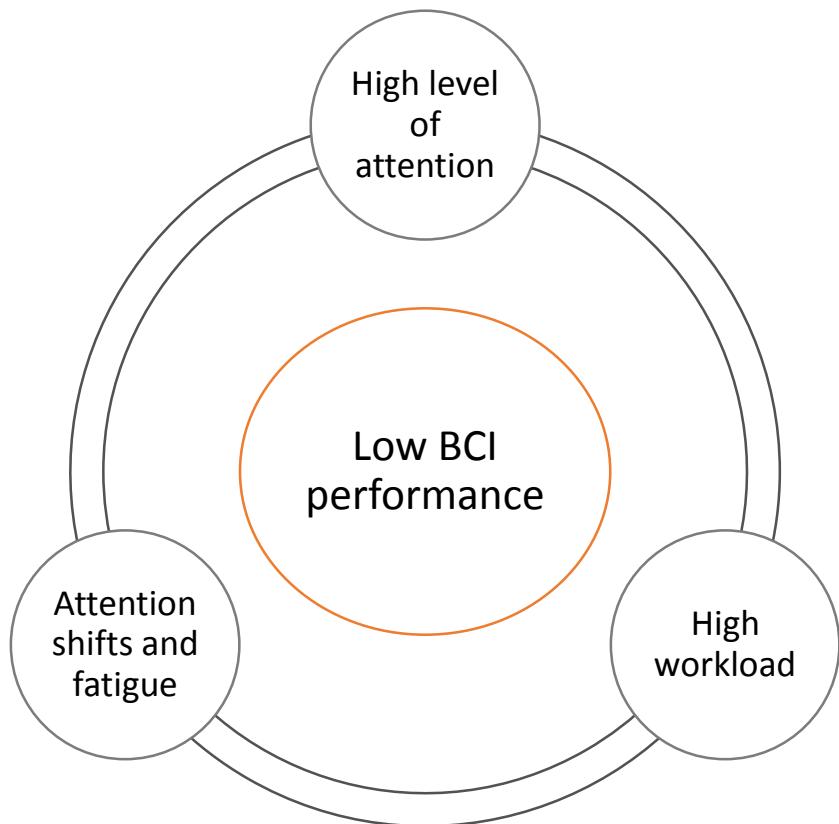
# Online classification performance

	Pre-Acc (%)	Post-Acc (%)	Pre-ITR (bpm)	Post-ITR (bpm)	Pre-B <sub>ECS</sub> (bpt)	Post- B <sub>ECS</sub> (bpt)	Pre-eSPM	Post-eSPM	Acc-ErrP <sub>1</sub> (%)	Acc-ErrP <sub>2</sub> (%)	N <sub>rep</sub>	
Subjects	S1	<b>81,6</b>	<b>92,8</b>	9,87	12,45	3,00	3,57	1,92	2,6	90,8	86,1	7
	S2	93,2	94,7	15,96	16,49	4,10	4,20	3,34	3,46	96,3	81,3	5
	S3	93,2	96,3	15,96	17,05	4,10	4,25	3,34	3,59	89,5	90,9	5
	S4	84,7	89,5	11,81	13,03	3,30	3,60	2,37	2,69	75,8	80,0	6
	S5	73,7	79,6	8,30	9,47	2,25	2,66	1,44	1,8	92,1	78,6	7
	S6	95,8	95,8	14,85	14,85	4,35	4,35	3,12	3,12	76,3	92,2	6
	S9	<b>81,1</b>	<b>90,5</b>	17,02	20,73	2,95	3,50	3,3	4,3	96,8	85,0	3
	P1	<b>75,0</b>	<b>79,6</b>	<b>8,55</b>	<b>9,47</b>	<b>2,38</b>	<b>2,63</b>	<b>1,52</b>	<b>1,8</b>	<b>89,5</b>	<b>84,8</b>	7
	Mean	<b>84,8</b>	<b>89,9</b>	<b>12,79</b>	<b>14,19</b>	<b>3,31</b>	<b>3,60</b>	<b>2,54</b>	<b>2,92</b>	<b>88,4</b>	<b>84,8</b>	<b>5,80</b>
Difference			<b>5,1</b>		<b>1,4</b>		<b>0,29</b>		<b>0,38</b>			

- The proposed approach increased the accuracy of 5.1%.
- The 1<sup>st</sup> and 2<sup>nd</sup> ErrP were successfully detected at a single level.

# Usability of BCI

# Towards User-friendly Brain-controlled Wheelchair

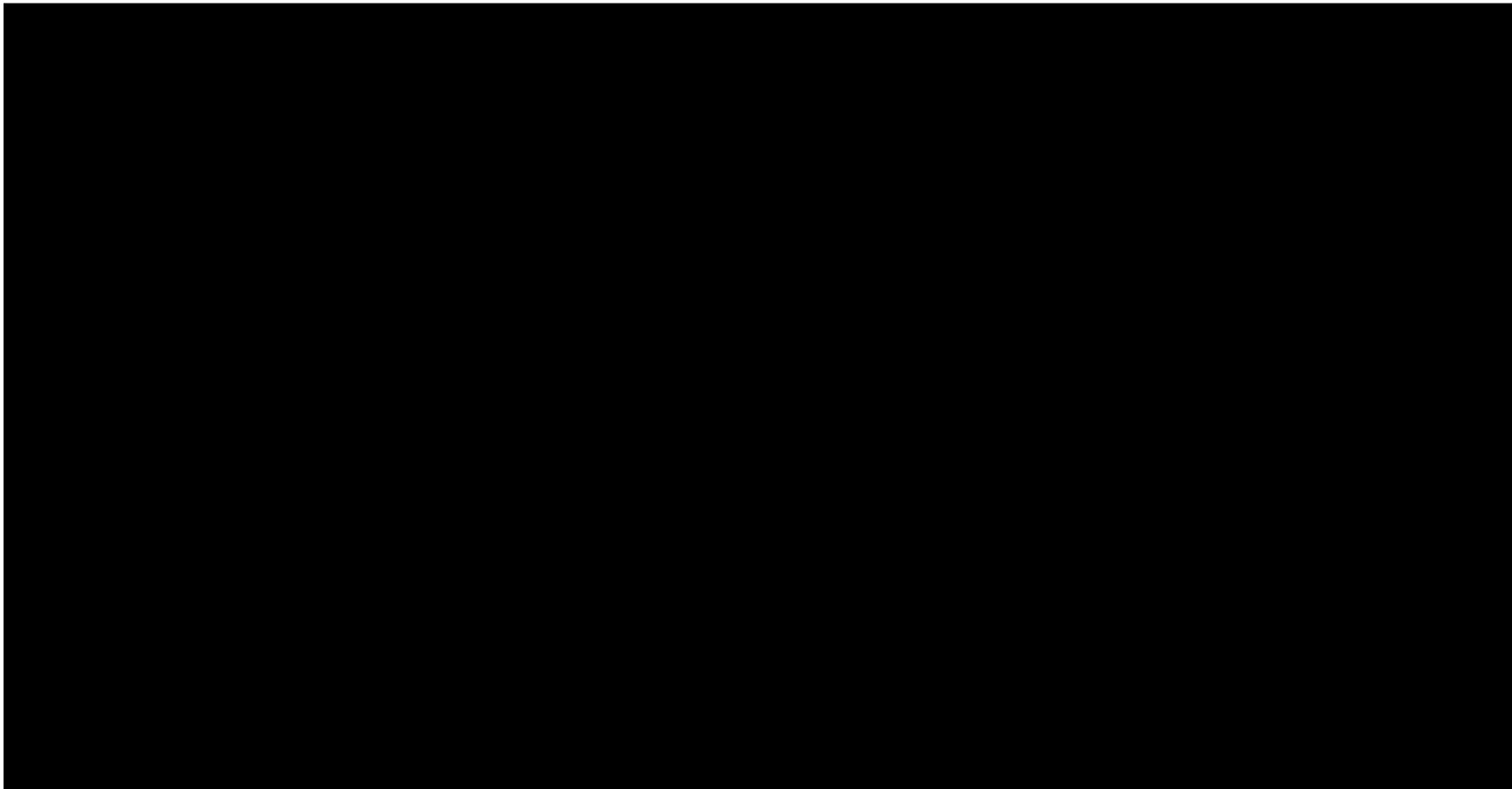


- We propose a minimum-effort brain-controlled wheelchair that combines:
  1. collaborative control;
  2. self-paced control;
  3. dynamic time trial.

# Towards User-friendly Brain-controlled Wheelchair

- **Self-paced control:** allows the user to issue a command only if he/she desires and the BCI detects whether the user is in a control state or in a non-control state.
- **Self-paced approach with static trial:** uses a fixed number of stimuli sequences.
- **Dynamic time trial approach:** automatically sets the number of stimuli sequences according to user's context awareness.
- **Collaborative-control:** validates user's command according to the navigation context.

# Demonstrative video



# Online BCW performance

Task1: self-paced mode with static trial				Task3: non-self-paced mode with static trial-time			
Subjects	CC	NCC	TT	Acc_BCI (%)	Acc_BCW (%)	Acc_BCI CC (%)	Acc_BCW (%)
S1	8	61	7,1	95,7	100,0	85	75,3
S2	12	72	5,9	97,6	100,0	73	97,3
S3	11	58	7,1	100,0	100,0	65	93,8
S5	10	78	7,1	100,0	100,0	71	94,4
S6	10	66	7,1	97,4	100,0	85	91,8
S8	7	57	8,4	96,9	100,0	62	82,3
S9	14	49	7,1	92,1	100,0	71	88,7
Average	<b>10,3</b>	<b>63,0</b>	<b>7,1</b>	<b>97,1</b>	<b>100,0</b>	<b>73,1</b>	<b>89,1</b>
P1	14	63	7,1	92,2	98,7	-	-
P3	23	104	7,1	94,5	100,0	-	-
P4	13	85	7,1	93,9	100,0	-	-
P7	13	68	7,1	91,4	98,8	-	-
P8	17	77	5,9	98,9	100,0	-	-
P9	12	66	7,1	97,4	100,0	-	-
Average	<b>15,3</b>	<b>77,2</b>	<b>6,9</b>	<b>94,7</b>	<b>99,6</b>		

➤ The self-paced control enormously reduced the number of control commands, the user's workload, and consequently, it increased the BCI accuracy in 8.0%.

# Online BCW performance

Task2: self-paced mode with dynamic trial (DTT)

Subjects	CC	NCC	TT(Max)	TT(Min)	TT(Mean)	Acc_BCI (%)	Acc_BCW (%)
S1	7	27	5,9	4,7	5,5	97,1	97,1
S2	11	31	8,4	3,5	4,9	100,0	100,0
S3	14	27	9,6	5,9	6,7	100,0	100,0
S5	10	31	5,9	4,7	5,8	97,6	100,0
S6	19	30	9,6	5,9	6,4	89,8	100,0
S8	10	23	10,8	5,9	8,0	93,9	100,0
S9	41	23	8,4	4,7	6,3	82,8	96,9
Average	16	27,4	8,4	5,0	6,2	94,5	99,1
P1	9	33	7,1	7,1	7,1	95,2	100,0
P3	36	36	9,6	4,7	6,8	81,9	98,6
P4	9	31	9,6	5,9	7,1	97,5	100,0
P7	12	79	5,9	4,7	5,7	94,5	98,9
P8	13	50	4,7	3,5	4,4	100,0	100,0
P9	8	31	7,4	5,2	6,2	87,2	100,0
Average	14,5	43,3	7,4	5,2	6,2	92,7	99,6

- With the DTT approach, BCI speed vs. user's performance is adjusted online.
- DTT reduced the time to select a command in about 1 s, although it decreased the BCI accuracy in about 2%.

## Conclusion

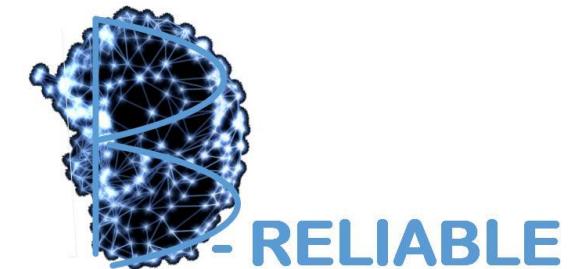
- Low reliability and usability are the major drawbacks of current BCIs, which limit their use outside laboratories.
- Using the proposed double ErrP detection there was an improvement in the BCI online accuracy around 5%.
- The proposed approach that combines collaborative control, self-paced control, and dynamic-time commands was extremely effective with overall accuracy greater than 99% for both able-bodied and motor disabled participants.

# Publications

- Cruz, A., Pires, G., Lopes, A., Carona, C., & Nunes, U. J. (2021). A self-paced BCI with a collaborative controller for highly reliable wheelchair driving: Experimental tests with physically disabled individuals. *IEEE Transactions on Human-Machine Systems*, 51(2), 109-119.
- Cruz, A., Pires, G., Lopes, A. C., & Nunes, U. J. (2019, July). Detection of stressful situations using GSR while driving a BCI-controlled wheelchair. In 2019 41st Annual international conference of the IEEE engineering in medicine and biology society (EMBC) (pp. 1651-1656). IEEE.
- Cruz, A., Pires, G., & Nunes, U. J. (2017). Double ErrP detection for automatic error correction in an ERP-based BCI speller. *IEEE transactions on neural systems and rehabilitation engineering*, 26(1), 26-36.

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# Thank you for your attention!