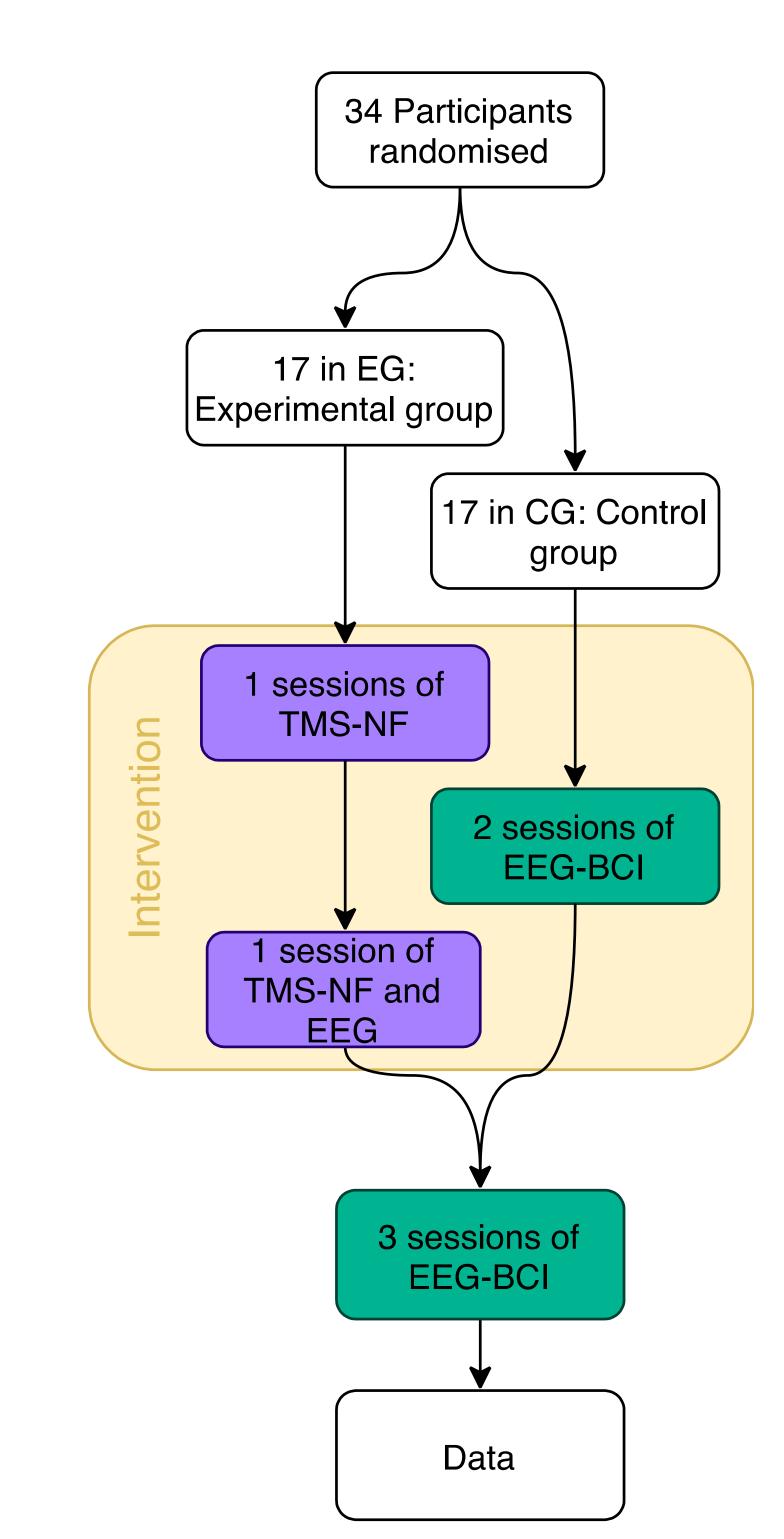
A two stage process for improving Brain-Computer Interface outcomes

BACKGROUND

- Most types of Brain Computer Interfaces (BCI) for neurorehabilitation use Electroencephalography (EEG) to detect brain activation patterns deemed beneficial for recovery
- These systems take many (5-8) sessions before the BCI can be used effectively, as it needs to adapt to individual patterns of brain activity
- Using other brain derived signals such as Motor Evoked Potentials (MEPs), elicited by Transcranial Magnetic Stimulation (TMS) as Neurofeedback (TMS-NF) healthy participants can achieve control over the BCI within 2 days (Ruddy et al., 2018)
- The trained brain states were associated with distinct patterns of neural oscillations within the motor network (Ruddy et al., 2018)
- Here we will test whether it is possible to use a pattern of brain activity learned during TMS-NF to feed into a two-staged BCI that can be used to pre-train participants before using a traditional EEG-BCI system, such that they would learn faster and with greater success



METHOD

- 34 participants are needed
- Based on a sample size calculation assuming a predicted effect size of d=0.97 (MEP Neurofeedback effect size from Ruddy et al., 2018)
- 2 groups matched on motor imagery ability using Motor Imagery Questionnaire (MIQ-RS, Butler et al., 2012)
- Experimental group will have 2 sessions of TMS-NF
- The TMS-NF blocks are made up of 4-6 blocks with 30 trials
- Second TMS-NF session collects TMS Evoked Potentials (TEPs), the brain's EEG response to the TMS pulse
- Control group will have 5 sessions of EEG-BCI
- To test if TMS-NF can pre-train participants before using a EEG-BCI system, *success* rate, learning rate, and classification accuracy between the two groups and their system will be compared

REFERENCES: Ruddy, K., Balsters, J., Mantini, D., Liu, Q., Kassraian-Fard, P., Enz, N., Mihelj, E., Subhash Chander, B., Soekadar, S. R., & Wenderoth, N. (2018). Neural activity related to volitional regulation of cortical excitability. *ELife*, 7, e40843. https://doi.org/10.7554/eLife.40843. https://doi.org/10.7554/eLife.40843 Schalk, G., McFarland, D. J., Hinterberger, T., Birbaumer, N., & Wolpaw, J. R. (2004). BCI2000: A general-purpose brain-computer interface (BCI) system. *IEEE Transactions on Biomedical Engineering*, 51(6), 1034–1043. https://doi.org/10.1109/TBME.2004.827072

C. Simon¹, K. Ruddy¹

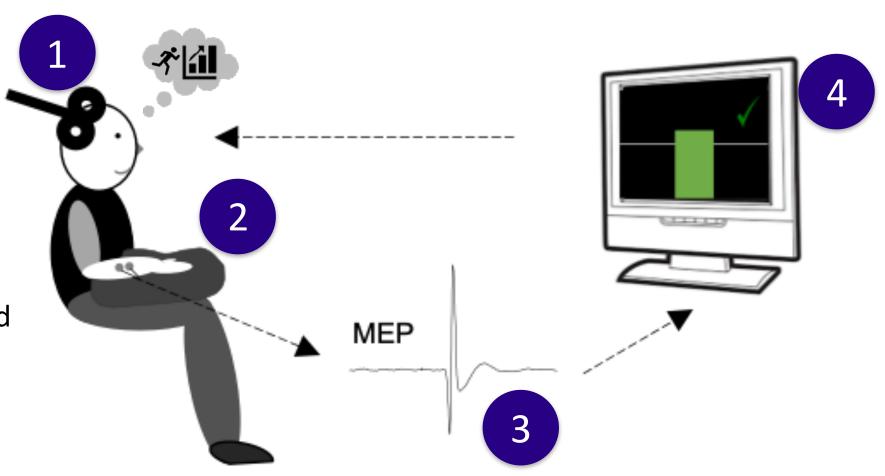
¹Institute of Neuroscience, Trinity College Dublin, Dublin 2, Ireland

csimon@tcd.ie

EXPERIMENTAL GROUP

TMS-NF

- 1. Transcranial Magnetic Stimulation (TMS) evokes Motor Evoked Potential (MEP)
- 2. MEP is recorded by Electromyography
- 3. MEP amplitude is compared to baseline
- 4. Feedback is displayed to participant



TMS-NF control scheme

The experimental group will have two sessions of TMS-NF. The last session of TMS-NF will include record EEG recordings.

Using a Common Spatial Patterns (CSP) approach the state responsible for MEPs with high amplitudes will be determined and used to drive the EEG-BCI sessions for the experimental group.

CONTROL GROUP

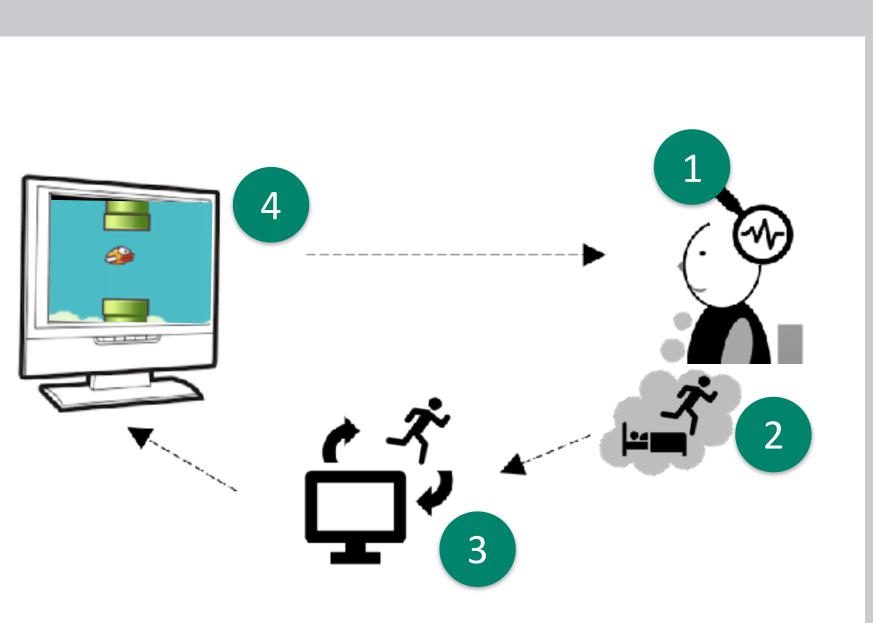
EEG-BCI

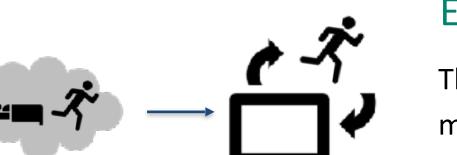
ズ山

1. Acquire EEG signals during rest and motor imagery

(Schalk et al., 2004)

- 2. Differentiate electrophysiological patterns for these two states using CSP
- 3. Display feedback to participant in computer game to train them to achieve control of BCI.



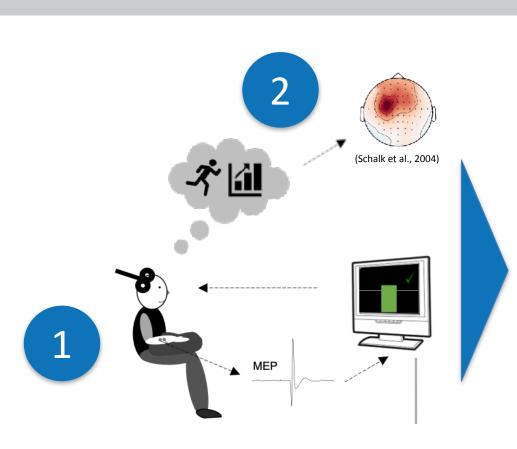


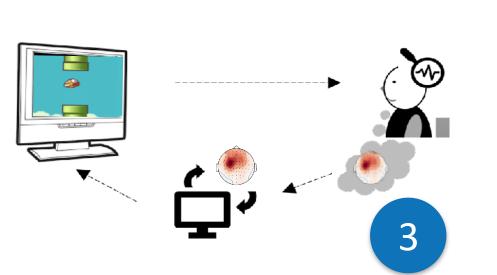
EEG-BCI control scheme

The control group BCI will also use CSP, calibrated by two different mental states- resting EEG and Motor Imagery.

TWO STAGES: COMBINED APPROACH

- 1. Two sessions of TMS-NF
- 2. Acquire CSP of maximal MEP amplitudes
- 3. Use CSP for Motor Imagery BCI





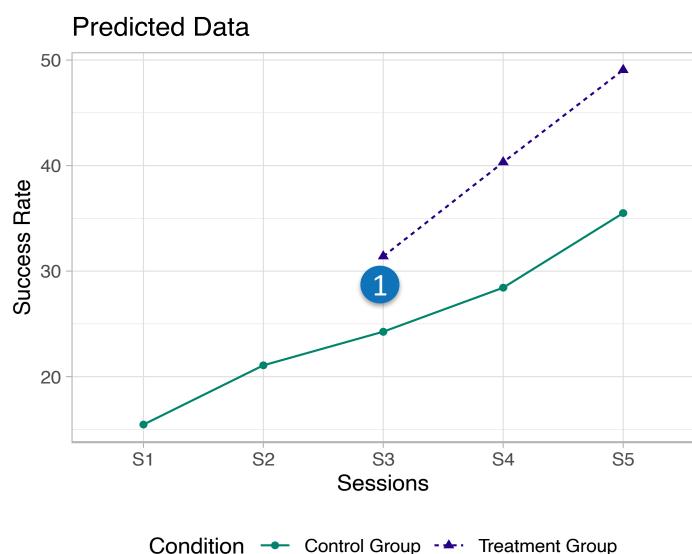
HYPOTHESES

(H1) TMS-NF will be a faster and, or more efficient way to train participants for EEG-BCI

ANALYSIS

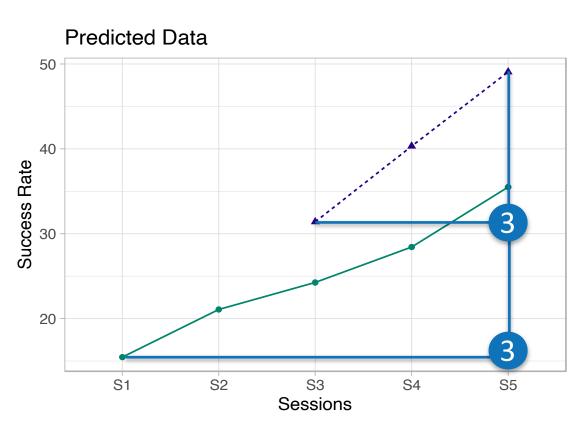


EEG-BCI Success Rate (H1)

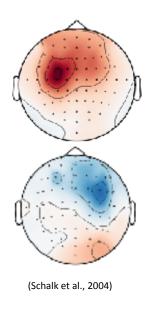


• Comparison 2: EEG-BCI success rate between TG, session 3 and CG session 1

This shows if TMS-NF can be used to pretrain participants for the use of EEG-BCI.



BCI Drivers (H2 & H3)



• Control group and experimental group used different controls for the EEG-BCI

• Differences in topography, oscillations and successful state categorisation between the groups could be related to the perfomance

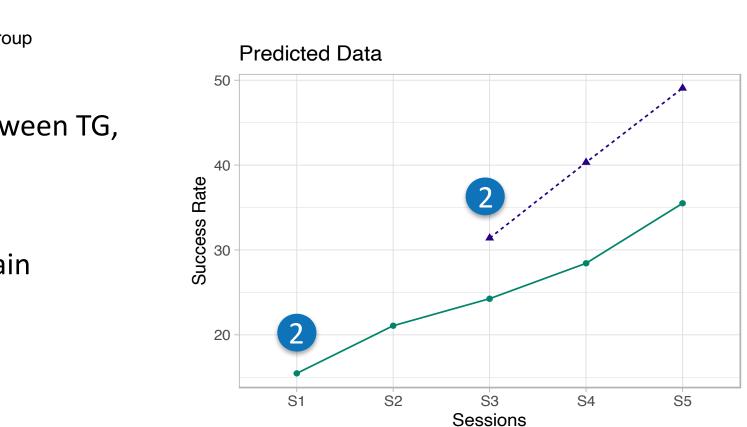
CONCLUSIONS

- injury)



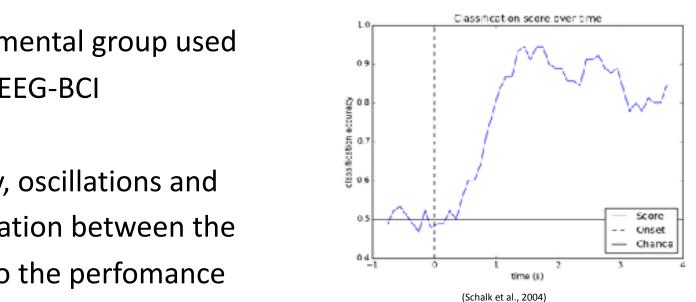
- (H2) CSP can be used to find a state responsible for high MEP amplitudes and can be used as EEG-BCI driver (H3) The groups will show differences in topography or oscillations, and classification accuracy rates
- This section discusses *hypothesised* study outcomes (data collection has not yet occured).
 - To investigate success rates between groups, there are three interesting comparisons:
 - Comparison 1: EEG-BCI success rate between the two groups at Session 3

This shows if two sessions of TMS-NF are more efficient than two traditional EEG-BCI sessions for success rates of EEG-BCI.



 Comparison 3: Slope of learning curve between TG and CG

This shows if TMS-NF helps participants to learn how to use the EEG-BCI faster.



• TMS-NF *could* be an alternative way to train participants for EEG-BCI use. The MEP feedback from TMS-NF could be valuable for participants who have difficulties using traditional EEG-BCI (for example due to a brain

• The quick and successful transition from TMS-NF to EEG-BCI *could* enable a two step programme for BCI using a portable EEG-BCI for homebased, patient driven neurorehabilitation, after TMS-NF training