

1 Proactive inhibition is marked by differences in the pattern of motor cortex activity during  
2 movement preparation and execution

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## 10 Key points

- 11 • We sought to investigate how the motor cortex prepares and executes equivalent  
12 movements made under different contexts
- 13 • We probed two different neuronal inputs to corticospinal neurons by activating the  
14 motor cortex with postero-anterior (PA) or antero-posterior (AP) TMS pulses prior to  
15 either simple reaction time movements or movements requiring proactive inhibition
- 16 • PA and AP responses represent separate axes of a state space upon which activity  
17 unfolds during movement preparation and execution
- 18 • The balance between PA and AP networks evolves differently when proactive  
19 inhibition is required versus when it is not. Thus, the evolution of activity in the motor  
20 cortex follows different trajectories in the two tasks, even though the final movement  
21 is identical

## 22 Abstract

23 Successful human behaviour relies on the ability to flexibly alter movements depending on  
24 the context in which they are made. One such context-dependent modulation is proactive  
25 inhibition, a type of behavioural inhibition used when anticipating the need to stop or change  
26 movements. We investigated how the motor cortex might prepare and execute movements  
27 made under different contexts. We used transcranial magnetic stimulation (TMS) in different  
28 coil orientations (PA: postero-anterior and AP: antero-posterior flowing currents) and pulse  
29 widths (120  $\mu$ s and 30  $\mu$ s) to probe the excitability of different inputs to corticospinal neurons  
30 whilst participants performed two reaction time tasks: a simple reaction time task and a stop-

31 signal task requiring proactive inhibition. We took inspiration from state space models to  
32 assess whether the pattern of motor cortex activity changed due to proactive inhibition (PA  
33 and AP neuronal circuits represent the x and y axes of a state space upon which motor cortex  
34 activity unfolds during motor preparation and execution). We found that the rise in motor  
35 cortex excitability was delayed when proactive inhibition was required. State space  
36 visualisations showed altered patterns of motor cortex activity (combined PA<sub>120</sub> and AP<sub>30</sub>  
37 activity) during proactive inhibition, despite adjusting for reaction time. Overall, we show  
38 that the pattern of neural activity generated by the motor cortex during movement preparation  
39 and execution is dependent upon the context under which the movement is to be made.

40 Keywords: proactive inhibition, transcranial magnetic stimulation, motor preparation, motor  
41 execution, dynamical systems, motor control

## 42 News and noteworthy

43 Using directional TMS, we find that the human motor cortex flexibly changes its pattern of  
44 neural activity depending on the context in which a movement is due to be made.  
45 Interestingly, this occurs despite adjusting for reaction time. We also show that state space  
46 and dynamical systems models of movement can be non-invasively visualised in humans  
47 using TMS, thereby offering a novel method to study these powerful models in humans.  
48

## 49 Introduction

50 Imagine accelerating a car from a stationary position. The way you prepare to accelerate will  
51 be different on a quiet road compared with a road outside a school. Your motor system is  
52 capable of generating the necessary muscle forces to accelerate the car at the same speed in  
53 different contexts; around a school it is much more likely that a child will deter your path,  
54 necessitating a sudden stop in motor output. Consequently, your brain must generate the same  
55 motor output in order to move the car but under two very different contexts. The ability to  
56 enact this context-dependent modulation of movement is paramount to normal human  
57 functioning and in this example is called proactive inhibition – a prospective and goal-  
58 oriented type of behavioural inhibition concerned with anticipation (Jahanshahi et al. 2015;  
59 Jahanshahi and Rothwell 2017).

60 But how does the brain do this? On one hand, the motor cortex might prepare and execute  
61 movements in the same way irrespective of the context, and a separate input to the motor  
62 system may modulate or cancel the ongoing movement when required (e.g. the inhibitory  
63 hyperdirect pathway during sudden stopping [Nambu *et al.*, 2002; Frank *et al.*, 2007]). More  
64 specifically, this predicts that the pattern of neural activity during movement preparation and  
65 execution will not differ across contexts. On the other hand, the motor cortex might prepare  
66 and execute movements in a fundamentally different way, based on the context in which they  
67 are to take place. In this model, the pattern of neural activity during preparation and  
68 execution will change across different contexts.

69 To investigate this problem, we used two behavioural tasks: a Go-only simple reaction time  
70 task, which required participants to respond with button presses to a simple go cue, and a  
71 stop-signal task, which had the same format as the Go-only task except that a minority of  
72 trials could be followed by a stop-signal, requiring participants to abort their response.  
73 Importantly, participants have slower responses in the latter task due to the anticipation of  
74 having to stop (proactive inhibition). Essentially, we asked participants to make the same  
75 movement (finger flexion) in different contexts: one where they know they might have to  
76 stop on some trials and another where they do not have to stop and must respond on every  
77 trial.

78 During the tasks, we stimulated the motor cortex with transcranial magnetic stimulation  
79 (TMS), a non-invasive brain stimulation tool that can activate underlying cortical neurons in  
80 a focal manner. If applied over M1, TMS results in muscular contractions called motor-  
81 evoked potentials (MEPs), the amplitude of which reflects excitability of the corticospinal-

82 muscular connection. By applying TMS at different intervals during the tasks, we were able  
83 to probe motor cortex excitability during movement preparation and execution. In particular,  
84 we were interested in the *pattern* of motor cortex excitability during different phases of  
85 movement, described in the next paragraph. To this end, we applied TMS in two different  
86 coil orientations (postero-anterior [PA] and antero-posterior [AP] flowing current), which are  
87 known to activate two (largely separate) populations of cortical motor neurons that have a  
88 common output (Di Lazzaro et al. 2012; Lazzaro et al. 2001; Mills et al. 1992). Recent work  
89 has expanded on this distinction, finding that altering TMS pulse width (120  $\mu$ s and 30  $\mu$ s)  
90 can further differentiate PA and AP neuronal inputs (Casula et al. 2018; D'Ostilio et al. 2016;  
91 Hannah et al. 2020; Hannah and Rothwell 2017). PA and AP inputs activate separate cortical  
92 circuits (Ni et al. 2011; Volz et al. 2015) and are behaviourally separable, given that they are  
93 differentially modulated during movement preparation (Hannah et al. 2017), behavioural  
94 plasticity (Hamada et al. 2014) and motor learning (Spampinato et al. 2020).

95 Patterns of neural activity are typically visualised using state space models that treat each  
96 neuron's activity as an individual axis in multi-dimensional space (Figure 1A). A point in this  
97 space determines the state of neural population activity at a particular time. By plotting these  
98 points throughout time, a trajectory is drawn, which determines the change of neural  
99 population state across time (Vyas et al. 2020). These states and trajectories reflect important  
100 features of movement dynamics and behaviour such as parsing motor preparation and  
101 execution into two discrete processes with independent, putative dynamics. Inspired by this,  
102 we sought to use TMS to visualise corticospinal excitability through the state space  
103 framework. PA and AP inputs activate separate cortical inputs to a common motor output; we  
104 therefore treated their respective activities as x and y dimensions on a 2D plane that  
105 represents the dimensions of a state space upon which motor cortex activity unfolds during  
106 motor preparation and execution (Figure 1B). Specifically, the pattern of motor cortex  
107 excitability would be manifested as the activities in PA<sub>120</sub> and AP<sub>30</sub> networks.

108

## 109 Methods

### 110 Participants

111 16 healthy volunteers (9 males, 16 right-handed) aged 19-33 (mean age 24.65, SD 4.13)  
112 participated. The study was approved by the UCL Ethics Committee and informed consent  
113 was obtained from all participants. The study was performed in accordance with the

114 Declaration of Helsinki. None of the participants had contraindications to TMS, which was  
115 assessed by a TMS screening questionnaire based on the one published by Keel and  
116 colleagues (Keel et al. 2001).

#### 117 Electromyography recordings

118 Throughout the experiment, participants were seated comfortably in a non-reclining chair,  
119 with their right index finger rested over the 'M' key on a keyboard. Their forearms were  
120 supported using a cushion. Electromyographic (EMG) activity was recorded from the right  
121 first dorsal interosseous (FDI) muscle using 19 x 38 mm surface electrodes (Ambu  
122 WhiteSensor 40713) arranged in a belly-tendon montage, with a sensor area of 77 mm<sup>2</sup>. The  
123 raw signals were amplified, and a bandpass filter was also applied (20 Hz to 2 kHz,  
124 Digitimer, Welwyn Garden City, United Kingdom). Signals were digitised at 5 kHz (CED  
125 Power 1401; Cambridge Electronic Design, Cambridge, United Kingdom) and data were  
126 stored on a computer for offline analysis (Signal version 5.10, Cambridge Electronic Design,  
127 United Kingdom).

#### 128 Transcranial magnetic stimulation

129 MEPs in the right FDI muscle were evoked using a controllable TMS (cTMS) device  
130 (cTMS3, Rogue Research Inc., Canada), connected to a standard figure-of-eight coil (wing  
131 diameter 70 mm, Magstim, United Kingdom). The hotspot was identified as the area on the  
132 scalp where the largest and most stable MEPs could be obtained for the right FDI muscle,  
133 using a suprathreshold TMS pulse. The hotspot was marked on the participant's scalp using a  
134 coloured pencil that was removed after the experiment had concluded. Importantly, hotspots  
135 were found separately for PA and AP coil orientations since they have distinct anatomical  
136 bases. We delivered monophasic TMS pulses in two ways. With the coil held approximately  
137 perpendicular to the presumed central sulcus and tangentially to the skull, TMS was given  
138 either with the coil handle pointing backwards for PA stimulation at 120 µs pulse width  
139 (PA<sub>120</sub>) or with the coil handle pointing forwards for AP stimulation at 30 µs pulse width  
140 (AP<sub>30</sub>).

#### 141 Stop-signal task and Go-only simple reaction time task

142 Participants were asked to perform two blocks of the stop-signal task (SST) and two blocks  
143 of a simple reaction time (Go-only) task, which were driven by custom-made MATLAB  
144 (MathWorks) scripts using Psychtoolbox (Brainard 1997). In the Go-only task (Figure 2),  
145 trials began with the presentation of a white fixation cross on a black background. 500 ms

146 later, a go cue (right arrow) was presented, which instructed participants to press the ‘M’ key  
147 on the keyboard as fast as possible with their right index finger (go trial, n=105). On fifteen  
148 trials, only a fixation cross was displayed. These served as catch trials and were randomly  
149 presented throughout the block. In essence, the Go-only task was a simple reaction time task  
150 that required less proactive control than the SST. For the SST (Figure 2), go (n=105) and  
151 catch (n=15) trials were presented as in the Go-only task. However, the SST included stop  
152 trials (n=35), whereby a stop signal (red cross) appeared above the go cue at a variable delay  
153 after the go cue, instructing participants to abort their motor responses. This variable delay is  
154 known as the stop signal delay (SSD) and can range from 100-250 ms in 50 ms time-steps  
155 (100, 150, 200 and 250 ms). Changing the SSD changes the difficulty of stopping when a  
156 stop signal is shown: short SSDs are easy to stop to, whereas longer SSDs make stopping  
157 more difficult. The SSD was initially set at 150 ms and changed on a trial-by-trial basis,  
158 depending on the outcome of the previous stop trial (dynamic tracking algorithm): if the  
159 participant successfully prevented their button press on a stop trial, the next stop trial would  
160 have its SSD set 50 ms later, whereas if the participant failed to stop, the next stop trial would  
161 have its SSD set 50 ms earlier (Verbruggen et al. 2019; Verbruggen and Logan 2009a,  
162 2009b). The dynamic tracking algorithm has been shown to reliably induce a convergence  
163 onto 50% successful inhibition across participants and hence ensures similar task  
164 performance across participants. Effectively, participants were responding on go trials as in  
165 the Go-only task, except with the prior knowledge that they might have to stop in anticipation  
166 of a stop signal – they were responding with restraint (Jahfari et al. 2010) and employing  
167 proactive inhibition. The order of trials was pseudorandomised, such that one in every four  
168 trials contained a stop signal. Inter-trial interval was set to 1750 ms.

169 The main behavioural measure of interest was the response delay effect (RDE) – a reaction  
170 time measure of proactive inhibition. This was calculated as the difference in reaction time on  
171 go trials in the SST and Go-only task. Other behavioural measures collected included Go  
172 reaction time (reaction time on go trials), Stop Respond reaction time (reaction time on failed  
173 stop trials), average SSD and p(inhibit) (proportion of correct stop trials on the SST). We also  
174 calculated the stop-signal reaction time (SSRT) using the mean method (Verbruggen and  
175 Logan 2009b) (mean go reaction time – mean SSD). The SSRT is a measure of reactive  
176 inhibition.

177 Integration of TMS with the stop-signal and Go-only simple reaction time tasks

178 TMS was given on all trials, in all blocks, to the M1 representation for the right FDI muscle,  
179 at an intensity required to produce a test MEP of 0.5 mV peak-to-peak amplitude; a test  
180 stimulus of 0.5 mV was chosen to limit the effect of TMS on reaction time, whilst still being  
181 able to capture the dynamic range of responses during movement. During go trials, one TMS  
182 pulse was given randomly at one of seven time points (at the go cue and 50, 100, 150, 200,  
183 250 and 300 ms after the go cue). 15 MEPs were taken at each time point. In the 15 baseline  
184 trials, TMS was given 1000 ms after presentation of the fixation cross (white cross) to assess  
185 CSE at rest.

186 Experimental protocol

187 Participants were first shown the two types of task that they would need to complete. They  
188 then performed a truncated version of each block so that they understood/learned the task  
189 prior to the collection of data used in this study. The order of task and TMS combination was  
190 randomised using a random number generator, after which participants completed each block  
191 in turn. Breaks were permitted between blocks.

192

193 Data analysis

194 Data handling and analyses were performed using custom-made scripts in MATLAB  
195 (MathWorks, version 2017a). We used the lme4 package (Bates et al. 2015) in R (version  
196 1.1.463) to run linear mixed models and post-hoc t-tests. All statistics were conducted as per  
197 a within-subject design given that the simulated Go-only and SST vs Go-only trajectories  
198 were derived from the same participants. **Data normality was assessed by visualising QQ-**  
199 **plots of residuals. Where data deviated from normality, log-transformations were applied to**  
200 **dependent variables prior to statistical analyses.**

201 Trials with reaction times exceeding 1000 ms were classed as omission errors due to lapses in  
202 concentration. The magnitude of proactive inhibition was determined as the reaction time  
203 difference on go trials between the SST and Go-only task, also known as the RDE. We used a  
204 linear mixed model with COIL ORIENTATION (PA<sub>120</sub>, AP<sub>30</sub>) and TASK (SST, Go-only) as  
205 fixed effects and modelled participant identity as a random intercept effect. Post-hoc paired t-  
206 tests with Tukey's correction were used to interrogate significant interactions.

207 MEPs were pre-processed using visual inspection. Trials where TMS arrived during or after  
208 the EMG burst were excluded from analysis. We used a linear mixed model to assess how

209 CSE changed after presentation of the go cue, for different coil orientations and under  
210 different tasks. For the cue-locked analysis, COIL ORIENTATION (PA<sub>120</sub>, AP<sub>30</sub>), TASK  
211 (SST, Go-only) and TIME POINT (Cue, 50, 100, 150, 200, 250, 300 ms) were treated as  
212 fixed effects, whereas the participant identity was chosen as a random intercept effect. We  
213 also represented CSE between stopping conditions and inputs from the viewpoint of  
214 movement execution. To do this, we calculated the time between TMS delivery and response,  
215 then grouped MEPs according to 50 ms time bins from the response (300-350, 250-300, 200-  
216 250, 150-200, 100-150 and 50-100 ms before movement). Similar to the cue-locked analysis  
217 outlined above, we performed a linear mixed model with the same factors as above, except  
218 that the factor TIME POINT represented time periods preceding the response (300-350, 250-  
219 300, 200-250, 150-200, 100-150 and 50-100 ms before movement). We included all  
220 interactions between our main factors in our models given that we were interested in each of  
221 their effects. Tukey corrected post-hoc paired t-tests were used to further investigate any  
222 significant interaction effects from the linear mixed models.

223 We took inspiration from state space modelling to visualise the pattern of neural activity  
224 during movement preparation and execution in the SST and Go-only tasks. To do so, we  
225 treated PA<sub>120</sub> and AP<sub>30</sub> inputs as dimensions on a 2D plane by plotting normalised to baseline  
226 PA<sub>120</sub> MEPs (x-axis) against normalised to baseline AP<sub>30</sub> MEPs (y-axis). Points on the plot  
227 show PA<sub>120</sub> and AP<sub>30</sub> MEPs at each time point for cue-locked (Cue, 50, 100, 150, 200, 250,  
228 300 ms) and response-locked (300-350, 250-300, 200-250, 150-200, 100-150 and 50-100 ms  
229 before movement) analyses. By treating each point in space as a vector (x co-ordinate =  
230 normalised PA<sub>120</sub> amplitude, y co-ordinate = normalised AP<sub>30</sub> amplitude), we calculated two  
231 measures of distance to help further understand the differences between trajectories in Go-  
232 only and SST conditions. Euclidean distances are sensitive to the magnitude of each vector,  
233 whereas cosine distances are insensitive to magnitude and quantify distance in terms of the  
234 difference in angles from the origin between two vectors. In brief, Euclidean distances signify  
235 the differences in the magnitude, whereas cosine distances represent difference in the relative  
236 weighting of PA<sub>120</sub>/AP<sub>30</sub> inputs. We used a method described by Ames et al. to test whether  
237 the trajectories taken differed significantly when proactive inhibition was used in the SST  
238 (Ames et al. 2014). First, simulated trajectories were first drawn (via bootstrapping with  
239 replacement) from the Go-only task. Each of these trajectories represent the neural  
240 trajectories that would occur by chance if there was no effect of proactive inhibition. Next,  
241 the distance between the simulated Go-only and original Go-only trajectories were then



242 calculated for each time point. These distances represent the distances if there was no effect  
 243 of proactive inhibition. Finally, we statistically compared the simulated distances with the  
 244 distances calculated between the SST and Go-only task using a linear mixed model for each  
 245 time point. Fixed effects in the linear mixed model included COIL ORIENTATION TIME  
 246 and ANALYSIS TYPE (real, simulated), with participant identity modelled as a random  
 247 intercept effect.

## 248 Results

### 249 Physiological measurements

250 No significant differences were found between the amplitudes of the baseline MEPs across  
 251 sessions or between PA<sub>120</sub> and AP<sub>30</sub> conditions. As expected, AP<sub>30</sub> TMS test stimulus (mean:  
 252 82.0%, SD: 10.0% of maximum stimulator output) intensities were higher than those for  
 253 PA<sub>120</sub> (mean: 29.6%, SD: 3.1% of maximum stimulator output) stimulation. Baseline MEPs  
 254 could not be elicited for three participants. Consequently, 16 participants provided data for  
 255 PA<sub>120</sub> TMS, 13 for AP<sub>30</sub> TMS. We note that the differences in stimulator intensities used  
 256 between PA<sub>120</sub> and AP<sub>30</sub> conditions cannot be interpreted as absolute differences given that  
 257 maximum stimulator output varies as a function of pulse width.

### 258 Behavioural measures

259 Behavioural measurements in the SST and Go-only simple reaction time task are shown in  
 260 Table 1. In the SST, the dynamic tracking algorithm correctly resulted in a convergence of  
 261 successful inhibition in approximately 50% of trials. The linear mixed model showed  
 262 significant effects of COIL ORIENTATION ( $p = 0.028$ ,  $F(1,44.49) = 5.21$ ) and TASK ( $p <$   
 263  $0.001$ ,  $F(1,40.74) = 79.85$ ) but no significant interaction ( $p = 0.224$ ,  $F(1,40.74) = 1.52$ ). This  
 264 meant that reaction times were slightly slower for AP<sub>30</sub> TMS trials than PA<sub>120</sub> trials, and  
 265 faster in the Go-only task than SST. This reaction time difference due to anticipatory slowing  
 266 (RDE) is the behavioural manifestation of proactive inhibition.

Measure	Measure description	SST		Go-only	
		PA <sub>120</sub>	AP <sub>30</sub>	PA <sub>120</sub>	AP <sub>30</sub>
Go reaction time	RT to go stimulus in the critical direction	391.55 (35.01)	402.36 (44.42)	288.31 (32.12)	324.15 (52.28)
P(inhibit)	% correct inhibition	50.54 (7.36)	56.70		

			(11.30)		
Stop Respond	RT on failed stop trials	287.84 (33.13)	319.69 (47.90)		
Go omission	% of omissions	0.36 (0.68)	0.44 (0.84)	0.36 (0.84)	0.66 (0.98)
Stop signal delay	Delay between go and stop trials	167.05 (25.42)	185.29 (31.52)		
SSRT	Calculated time to abort response	224.50 (27.75)	216.98 (32.59)		

267 Table 1: Behavioural measurements from the SST and Go-only simple reaction time tasks.

268 The table shows the behavioural measures from the SST, Go-only simple reaction time task.

269 Measures are accompanied by SD in brackets. Reaction times are given in ms. SSRT = stop

270 signal reaction time

271 Evolution of corticospinal excitability in stop-signal and Go-only simple reaction time tasks

272 Preparation of movement: cue-locked analysis

273 The SST was used to probe the temporal dynamics of CSE changes during which proactive

274 inhibition is implemented (Rawji et al. 2020a). This was compared to the same TMS timings

275 in a task where less proactive inhibition should be employed during the Go-only simple

276 reaction time task. A linear mixed model showed significant effects for COIL

277 ORIENTATION ( $p < 0.001$ ,  $F(1,324) = 14.66$ ), TASK ( $p < 0.001$ ,  $F(1,324) = 48.30$ ) and

278 TIME POINT ( $p < 0.001$ ,  $F(6,324) = 53.71$ ). Of note, there was a significant TASK\*TIME

279 POINT ( $p < 0.001$ ,  $F(6,324) = 4.44$ ) interaction. We used post-hoc t-tests to investigate this

280 interaction (full interaction comparisons are shown in the supplementary material). On go

281 trials within the SST, the main rise in excitability, indexed by the timepoint at which CSE

282 became significantly greater than CSE at the cue, occurred later than on Go-only trials for

283 both (Go-only: 150 ms,  $p < 0.001$ ,  $t = 6.19$ ; SST: 200 ms,  $p = 0.002$ ,  $t = 4.29$ ). These results

284 are summarised by plots in the top row of Figure 3.

285 Execution of movement: response-locked analysis

286 To assess CSE from the perspective of movement execution, we realigned the data to the

287 time of the response onset, thereby performing a response-locked analysis (Figure 3, bottom

288 row). A linear mixed model did not find any statistically significant effects of COIL

289 ORIENTATION ( $p = 0.272$ ,  $F(1,266.05) = 1.21$ ) or TASK ( $p = 0.149$ ,  $F(1,266.04) = 2.10$ )  
290 and as expected revealed a significant effect of TIME ( $p < 0.001$ ,  $F(5,266.06) = 92.98$ ). There  
291 were no statistically significant interactions. From this, it appears that CSE preceding a  
292 response did not differ between coil orientations or task.

293

#### 294 Assessing the pattern of neural activity during movement preparation and execution

295 We next assessed if the pattern of neural activity during movement preparation and execution  
296 changed as a function of proactive inhibition. An important distinction between the previous  
297 analysis and the subsequent one is how the factor COIL ORIENTATION is considered: in  
298 the former, there are two groups (PA<sub>120</sub> and AP<sub>30</sub>) with corresponding CSEs for each group,  
299 which allows for the effect of COIL ORIENTATION to be assessed. However, patterns of  
300 neural activity are assessed by considering M1 CSE as the combination of PA<sub>120</sub> and AP<sub>30</sub>  
301 inputs, thereby compressing the factor COIL ORIENTATION into one group. This means  
302 that each value/dependent variable becomes a vector with the first value being PA<sub>120</sub> MEP  
303 amplitude and the second value being AP<sub>30</sub> MEP amplitude. Accordingly, this vector-based  
304 analysis requires a different analytical approach than that described in the previous section.

305 We plotted normalised to baseline PA<sub>120</sub> and AP<sub>30</sub> MEPs against each other for each  
306 timepoint in the cue and response-locked analyses (Figure 4, top row). The resultant  
307 trajectories show how population-level activity within M1 evolves during movement  
308 preparation and execution. The cue-locked analysis shows that M1 population activity  
309 evolves within the same subspace early after cue presentation (bottom left). Approximately  
310 150 ms later, activity increases in both neuronal inputs and occupies a separate space at the  
311 end of movement (top right). Notably, the trajectories taken varied between the tasks given  
312 significant differences in Euclidean distances (TIME: [ $p < 0.001$ ,  $F(6,156) = 42.14$ ];  
313 ANALYSIS TYPE: [ $p < 0.001$ ,  $F(1,156) = 45.89$ ; TIME\*ANALYSIS TYPE: [ $p = 0.741$ ,  
314  $F(6,156) = 0.59$ ]). Cosine distances did not significantly differ by ANALYSIS TYPE ( $p =$   
315  $0.466$ ,  $F(1,156) = 0.53$ ), TIME ( $p = 0.22$ ,  $F(6,156) = 1.39$ ) and there was no interaction ( $p =$   
316  $0.100$ ,  $F(6,156) = 1.81$ ).

317 The response-locked analyses similarly showed a difference in trajectories, despite correcting  
318 for reaction time. We found significant differences in Euclidean distances when proactive  
319 inhibition was required ( $p < 0.001$ ,  $F(1,116.26) = 29.06$ ) and varied significantly over time ( $p$   
320  $< 0.001$ ,  $F(5,116.69) = 18.67$ ), without a significant interaction ( $p = 0.419$ ,  $F(1,116.28) =$

321 1.00). A significant difference in cosine distance was seen in the SST ( $p = 0.007$ ,  $F(1,116.87)$   
322  $= 7.56$ ) but there was no significant effect of time nor interaction. It appears that when  
323 viewing the patterns of neural activity, behaviourally equivalent responses are prepared and  
324 executed differently by M1 population activity, dependent on the requirements for proactive  
325 inhibition.  
326

## 327 Discussion

328 We sought to understand how M1 would prepare and execute the same movement when  
329 made under different contexts. More specifically, we used two reaction time tasks requiring  
330 an identical response, which differed in their task instructions, extent of motor preparation  
331 and anticipatory slowing, such that participants employed greater proactive inhibition in the  
332 SST. By applying TMS, we observed that, relative to the time of onset of the go cue, CSE  
333 increased later during the go trials of the SST compared with the Go-only task (Figure 3, cue-  
334 locked analysis). From the viewpoint of motor execution (Figure 3, response-locked  
335 analysis), the rate of increase in CSE was the same in both tasks. Overall, this suggests that  
336 proactive inhibition delays the initial rise in CSE after a cue is presented, rather than causing  
337 a slower rise of CSE to a notional “threshold” for movement (Rawji et al. 2020b) – a feature  
338 also found when macaque monkeys delay saccades during a visual SST (Pouget et al. 2011).  
339 We placed particular interest on how patterns of neural activity might be differentially  
340 represented during proactive inhibition and used state space models to visualise this. In doing  
341 so, we found that proactive inhibition was marked by a difference in the trajectories taken  
342 preceding movement and by the relative weighting in  $PA_{120}$  and  $AP_{30}$  inputs (Figure 4).

## 343 A dynamical systems view of proactive inhibition

344 The dynamical systems view of motor control is becoming a popular way in which to  
345 visualise neural activity during movement (Shenoy et al. 2013; Vyas et al. 2020). This  
346 proposes that, instead of representing explicit features of the movement (such as direction or  
347 velocity), M1 activity during motor preparation sets the initial state of a dynamical system,  
348 that evolves into the desired movement (Churchland et al. 2010) upon the receipt of some  
349 trigger to move (Kaufman et al. 2016). Consequently, neural activity during movement  
350 preparation and execution reflects the transition from one state to the next under some  
351 dynamical rule, and hence not all M1 activity need represent movement-related activity.

352 Crucially, the dynamical system arises as an interplay between populations of neurons during  
353 motor preparation and execution and is not appreciated from the single-neuron perspective,  
354 which has traditionally driven theories of motor control.

355 The dynamical systems view uses state space models to visualise population-level neural  
356 activity. Whilst largely successful, the investigation of dynamical systems in humans has  
357 been limited by the difficulty and infeasibility of large-scale single-neuron recordings. A  
358 recent study has attempted to overcome this by using an innovative experimental design to  
359 show that choice traces similar to those seen in non-human primates using dynamical systems  
360 models (Mante et al. 2013) can be visualised in premotor cortex using  
361 magnetoencephalography (Takagi et al. 2021). The current study similarly shows that M1  
362 activity can also be interpreted in a dynamical systems framework using TMS. The  
363 dynamical systems approach is appropriate to interpret TMS results, given that they both rely  
364 on population-level activity. By plotting  $PA_{120}$  CSE against  $AP_{30}$  CSE, we visualise how M1  
365 *population* activity evolves throughout time and through movement preparation and  
366 execution (Figure 4). Akin to the findings using a dynamical systems approach, we see that  
367 activity during movement preparation evolves in a particular, confined subspace for  
368 approximately 150 ms after cue presentation. That is, activity occupies the bottom left of the  
369 cue-locked plot for 150 ms (similarly, activity occupies the bottom left of the response-  
370 locked plot 300-350 ms prior to movement). Following this, M1 population CSE shows a  
371 large increase upon receipt of a trigger for movement execution, to a different area in the  
372 subspace (Figure 4, top right of the cue-locked and response-locked plots).

373 An important observation is that the trajectory taken (Figure 4) during movement differs  
374 between tasks. More specifically, the absolute magnitude (Euclidean distance) and balance  
375 within  $PA_{120}$  and  $AP_{30}$  inputs (cosine distance) differ when proactive inhibition is called  
376 upon. Overall, visualisation through this framework shows that the pattern of neural activity  
377 within M1 differs depending on the context in which the movement is due to be made – a  
378 feature not apparent from conventional analyses as in Figure 3.

### 379 Context-dependent modulation of movement

380 The Introduction outlined that context-dependent modulation of movement might be  
381 implemented in a number of ways: in one, the motor command output by M1 is the same  
382 irrespective of the context, and context-dependent modulation of movement comes about by  
383 modulation of this descending output. In another, M1 might change its descending motor

384 command depending on the context. The present study provides evidence for the latter  
385 hypothesis, showing that the pattern of neural activity within M1 is dependent on the context  
386 in which that movement will be performed. But how does the brain generate this context-  
387 relevant activity? In dynamical systems models of movement, neural activity occupies a state  
388 at the end of preparation that is relevant to the upcoming movement, which then evolves into  
389 the movement upon receipt of some trigger. Context-dependent modulation of movement  
390 may arise secondary to: 1) setting of an alternative preparatory state (Driscoll et al. 2018;  
391 Remington et al. 2018), 2) difference in the trigger that causes evolution of preparatory to  
392 movement activity, or 3) both.

393 The ability to shift the distribution of CSE between neuronal inputs may allow for  
394 qualitatively equivalent movements to be performed in a variety of ways depending on task-  
395 specific goals (Aberbach et al. 2021). In fact, a recent study by Baudry and Duchateau has  
396 shown that CSE rise time prior to EMG onset occurs 100 ms earlier for smooth ramp than  
397 ballistic contractions of tibialis anterior (akin to the response-locked analysis in the present  
398 study). They proposed that the difference in rise time was related to differences in how short-  
399 interval intracortical inhibition (SICI) changed prior to EMG onset: a sharp decrease in SICI  
400 was observed 200-100 ms prior to EMG onset for ballistic contractions, whereas SICI  
401 decreased smoothly over time for ramp contractions (Baudry and Duchateau 2021). This  
402 suggests that intracortical dynamics can be flexibly adapted depending on how the movement  
403 is made. The present study adds to this idea and is the first visualisation of CSE during  
404 movement preparation and execution represented as the interplay between different neuronal  
405 inputs in humans depending on movement context.

## 406 Limitations

407 The pseudorandom design of the task meant that participants could develop expectancy and  
408 learn to anticipate the stop signal, which could potentially confound measures of response  
409 inhibition. However, we observed that participants successfully inhibited their responses on  
410 approximately 50% of stop trials, showing that participants correctly engaged with task  
411 demands.

412 In light of the equivalence shown by the response-locked analyses, we concluded that  
413 participants were delaying their trigger to move. Variants of the SST have shown that  
414 proactive inhibition can sometimes be mediated by alterations in the threshold before which  
415 movement is triggered (Rawji et al. 2020a, 2020b). Given these apparent differences in the

416 strategy used to mediate proactive inhibition, it may be the case that our findings are a feature  
417 of task design and differ when proactive inhibition is mediated differently. The movement  
418 (right finger button press) was known throughout the experiment and did not change,  
419 meaning that the same movement was prepared in both tasks, on all trials. Consequently, the  
420 similarity in response-locked CSE profiles (Figure 3) may be so because the movement to be  
421 prepared is the same in both tasks (although this would not account for differences in the  
422 population-activity analysis in Figure 4). Future studies should aim to change the way in  
423 which the same movement is prepared to help establish whether movement execution CSE is  
424 dependent on motor preparation.

425 Button presses involve flexion of the metacarpophalangeal joint in which FDI is a synergist  
426 with flexor indicis profundus (De Almeida et al. 2021). We chose to record from FDI for two  
427 reasons: first, it has a lower threshold to TMS, making it a more pleasant experience for  
428 participants, and second, because it is easier to isolate the activity of FDI with surface EMG  
429 than the deep flexor muscle. However, it would have been useful to have additional  
430 confirmation of the results from other muscles involved in the movement. Nevertheless,  
431 previous studies have shown FDI CSE modulation during movement preparation and  
432 execution of button presses (Ibáñez et al. 2019; Klein-Flugge et al. 2013; Klein-Flugge and  
433 Bestmann 2012).

434 We noted a statistically significant effect of coil orientation on reaction time; on average,  
435 reaction times for AP<sub>30</sub> TMS were 11 ms longer in the Go-only task and 36 ms longer in the  
436 SST. Whilst this would not affect the results of the cue-locked analyses (as MEPs were  
437 aligned to the cue), they could confound the response-locked analysis, since MEPs were  
438 aligned to the response. The difference in reaction time due to coil orientation has been  
439 difficult to resolve given the interactions of pulse width and TMS intensity with the need to  
440 evoke the same amplitude MEP for both orientations. Nevertheless, we think that the effect  
441 of reaction time prolongation on our results may be limited since bin widths were 50 ms wide  
442 – larger than the reaction time differences observed.

## 443 Conclusions

444 We set out to investigate how the motor cortex would prepare and execute equivalent  
445 movement when made under different contexts. When proactive inhibition was required,  
446 movements were prepared by delaying the rise in CSE. Using state space models and  
447 directional TMS, we show that proactive inhibition might operate by altering the pattern of

448 activity used by M1 to execute a movement, indexed by different trajectories prior to  
449 movement onset or setting of an alternative initial state before the rise in CSE leading up to  
450 movement.

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## 456 Author contributions

457 All authors contributed to the design of the study and were involved in the drafting and  
458 revisions of the manuscript. VR and SM collected the data. All authors have approved this  
459 manuscript.

## 460 Conflicts of interest

461 The authors have no conflicts of interest.

## 462 Data availability

463 The data and code used to produce these findings will be posted on the first author's GitHub  
464 page.

465 Supplemental data - <https://doi.org/10.6084/m9.figshare.19193396.v1>

466

## 467 References

468 **Aberbach S, Buaron B, Mudrik L, Mukamel R.** Same Action, Different Meaning: Neural  
469 substrates of Semantic Goal Representation. *Neuroscience*.

470 **Ames KC, Ryu SI, Shenoy KV.** Neural Dynamics of Reaching following Incorrect or  
471 Absent Motor Preparation. *Neuron* 81: 438–451, 2014.

472 **Bates D, Mächler M, Bolker B, Walker S.** Fitting Linear Mixed-Effects Models Using  
473 lme4. *J Stat Soft* 67: 1–48, 2015.



- 474 **Baudry S, Duchateau J.** Changes in corticospinal excitability during the preparation phase  
475 of ballistic and ramp contractions. *J Physiol* 599: 1551–1566, 2021.
- 476 **Brainard H David.** The Psycholysics Toolbox. *Spatial Vision* 10: 433–436, 1997.
- 477 **Casula EP, Rocchi L, Hannah R, Rothwell JC.** Effects of pulse width, waveform and  
478 current direction in the cortex: A combined cTMS-EEG study. *Brain Stimulation* 11: 1063–  
479 1070, 2018.
- 480 **Churchland MM, Cunningham JP, Kaufman MT, Ryu SI, Shenoy KV.** Cortical  
481 Preparatory Activity: Representation of Movement or First Cog in a Dynamical Machine?  
482 *Neuron* 68: 387–400, 2010.
- 483 **De Almeida Y-K, Krebs M, Braun M, Dap F, Dautel G, Athlani L.** Innervation and  
484 vascular supply of the first dorsal interosseous muscle and palmar interosseous muscle of the  
485 index: An anatomic descriptive study. *Morphologie* S1286011520301296, 2021.
- 486 **Di Lazzaro V, Profice P, Ranieri F, Capone F, Dileone M, Oliviero A, Pilato F.** I-wave  
487 origin and modulation. *Brain Stimulation* 5: 512–525, 2012.
- 488 **D’Ostilio K, Goetz SM, Hannah R, Ciocca M, Chieffo R, Chen JCA, Peterchev AV,  
489 Rothwell JC.** Effect of coil orientation on strength-duration time constant and I-wave  
490 activation with controllable pulse parameter transcranial magnetic stimulation. *Clinical  
491 Neurophysiology* 127: 675–683, 2016.
- 492 **Driscoll LN, Golub MD, Sussillo D.** Computation through Cortical Dynamics. *Neuron* 98:  
493 873–875, 2018.
- 494 **Frank MJ, Samanta J, Moustafa AA, Sherman SJ.** Hold your horses: Impulsivity, deep  
495 brain stimulation, and medication in Parkinsonism. *Science* 318: 1309–1312, 2007.
- 496 **Hamada M, Galea JM, Di Lazzaro V, Mazzone P, Ziemann U, Rothwell JC.** Two  
497 Distinct Interneuron Circuits in Human Motor Cortex Are Linked to Different Subsets of  
498 Physiological and Behavioral Plasticity. *Journal of Neuroscience* 34: 12837–12849, 2014.
- 499 **Hannah R, Cavanagh SE, Tremblay S, Simeoni S, Rothwell JC.** Selective suppression of  
500 local interneuron circuits in human motor cortex contributes to movement preparation. *The  
501 Journal of Neuroscience* 2869–17, 2017.
- 502 **Hannah R, Rocchi L, Tremblay S, Wilson E, Rothwell JC.** Pulse width biases the balance  
503 of excitation and inhibition recruited by transcranial magnetic stimulation. *Brain Stimulation*  
504 13: 536–538, 2020.
- 505 **Hannah R, Rothwell JC.** Pulse Duration as Well as Current Direction Determines the  
506 Specificity of Transcranial Magnetic Stimulation of Motor Cortex during Contraction. *Brain  
507 Stimulation* 10: 106–115, 2017.
- 508 **Ibáñez J, Hannah R, Rocchi L, Rothwell J.** Premovement Suppression of Corticospinal  
509 Excitability may be a Necessary Part of Movement Preparation. *Cerebral Cortex* 1–14, 2019.

- 510 **Jahanshahi M, Obeso I, Rothwell JC, Obeso JA.** A fronto-striato-subthalamic-pallidal  
511 network for goal-directed and habitual inhibition. *Nature Reviews Neuroscience* 16: 719–732,  
512 2015.
- 513 **Jahanshahi M, Rothwell JC.** Inhibitory dysfunction contributes to some of the motor and  
514 non-motor symptoms of movement disorders and psychiatric disorders. *Philosophical*  
515 *Transactions of the Royal Society B: Biological Sciences* 372, 2017.
- 516 **Jahfari S, Stinear CM, Claffey M, Verbruggen F, Aron AR.** Responding with restraint:  
517 What are the neurocognitive mechanisms? *Journal of Cognitive Neuroscience* 22: 1479–  
518 1492, 2010.
- 519 **Kaufman MT, Seely JS, Sussillo D, Ryu SI, Shenoy KV, Churchland MM.** The Largest  
520 Response Component in the Motor Cortex Reflects Movement Timing but Not Movement  
521 Type. *eneuro* 3: ENEURO.0085-16.2016, 2016.
- 522 **Keel JC, Smith MJ, Wassermann EM.** A safety screening questionnaire for transcranial  
523 magnetic stimulation. *Clinical Neurophysiology* 112: 720, 2001.
- 524 **Klein-Flugge MC, Bestmann S.** Time-Dependent Changes in Human Corticospinal  
525 Excitability Reveal Value-Based Competition for Action during Decision Processing.  
526 *Journal of Neuroscience* 32: 8373–8382, 2012.
- 527 **Klein-Flugge MC, Nobbs D, Pitcher JB, Bestmann S.** Variability of Human Corticospinal  
528 Excitability Tracks the State of Action Preparation. *Journal of Neuroscience* 33: 5564–5572,  
529 2013.
- 530 **Lazzaro VD, Oliviero A, Saturno E, Pilato F, Insola A, Mazzone P, Profice P, Tonali P,**  
531 **Rothwell J.** The effect on corticospinal volleys of reversing the direction of current induced  
532 in the motor cortex by transcranial magnetic stimulation. *Experimental Brain Research* 138:  
533 268–273, 2001.
- 534 **Mante V, Sussillo D, Shenoy KV, Newsome WT.** Context-dependent computation by  
535 recurrent dynamics in prefrontal cortex. *Nature* 503: 78–84, 2013.
- 536 **Mills KR, Boniface SJ, Schubert M.** Magnetic brain stimulation with a double coil: the  
537 importance of coil orientation. *Electroencephalography and Clinical Neurophysiology/*  
538 *Evoked Potentials* 85: 17–21, 1992.
- 539 **Nambu A, Tokuno H, Takada M.** Functional significance of the  
540 corticoÁ/subthalamoÁ/pallidal ‘hyperdirect’ pathway. *Neuroscience Research* 7, 2002.
- 541 **Ni Z, Charab S, Gunraj C, Nelson AJ, Udupa K, Yeh I-J, Chen R.** Transcranial Magnetic  
542 Stimulation in Different Current Directions Activates Separate Cortical Circuits. *Journal of*  
543 *Neurophysiology* 105: 749–756, 2011.
- 544 **Pouget P, Logan GD, Palmeri TJ, Boucher L, Pare M, Schall JD.** Neural Basis of  
545 Adaptive Response Time Adjustment during Saccade Countermanding. *Journal of*  
546 *Neuroscience* 31: 12604–12612, 2011.

- 547 **Rawji V, Modi S, Latorre A, Rocchi L, Hockey L, Bhatia K, Joyce E, Rothwell JC,**  
548 **Jahanshahi M.** Impaired automatic but intact volitional inhibition in primary tic disorders.  
549 *Brain : a journal of neurology* 143: 906–919, 2020a.
- 550 **Rawji V, Rocchi L, Foltynie T, Rothwell JC, Jahanshahi M.** Ropinirole, a dopamine  
551 agonist with high D3 affinity, reduces proactive inhibition: A double-blind, placebo-  
552 controlled study in healthy adults. *Neuropharmacology* 179: 108278, 2020b.
- 553 **Remington ED, Narain D, Hosseini EA, Jazayeri M.** Flexible Sensorimotor Computations  
554 through Rapid Reconfiguration of Cortical Dynamics. *Neuron* 98: 1005-1019.e5, 2018.
- 555 **Shenoy KV, Sahani M, Churchland MM.** Cortical Control of Arm Movements: A  
556 Dynamical Systems Perspective. *Annu Rev Neurosci* 36: 337–359, 2013.
- 557 **Spampinato DA, Celnik PA, Rothwell JC.** Cerebellar–Motor Cortex Connectivity: One or  
558 Two Different Networks? *J Neurosci* 40: 4230–4239, 2020.
- 559 **Takagi Y, Hunt LT, Woolrich MW, Behrens TE, Klein-Flügge MC.** Adapting non-  
560 invasive human recordings along multiple task-axes shows unfolding of spontaneous and  
561 over-trained choice. *eLife* 10: e60988, 2021.
- 562 **Verbruggen F, Aron AR, Band GP, Beste C, Bissett PG, Brockett AT, Brown JW,**  
563 **Chamberlain SR, Chambers CD, Colonius H, Colzato LS, Corneil BD, Coxon JP,**  
564 **Dupuis A, Eagle DM, Garavan H, Greenhouse I, Heathcote A, Huster RJ, Jahfari S,**  
565 **Kenemans JL, Leunissen I, Logan GD, Matzke D, Morein-Zamir S, Murthy A, Li C-**  
566 **SR, Paré M, Poldrack RA, Ridderinkhof KR, Robbins TW, Roesch M, Rubia K,**  
567 **Schachar RJ, Schall JD, Stock A-K, Swann NC, Thakkar KN, van der Molen MW,**  
568 **Vermeylen L, Vink M, Wessel JR, Whelan R, Zandbelt BB, Boehler CN.** A consensus  
569 guide to capturing the ability to inhibit actions and impulsive behaviors in the stop-signal  
570 task. *eLife* 8: 1–26, 2019.
- 571 **Verbruggen F, Logan GD.** Proactive Adjustments of Response Strategies in the Stop-Signal  
572 Paradigm. *Journal of Experimental Psychology: Human Perception and Performance* 35:  
573 835–854, 2009a.
- 574 **Verbruggen F, Logan GD.** Models of response inhibition in the stop-signal and stop-change  
575 paradigms. *Neuroscience and biobehavioral reviews* 33: 647–61, 2009b.
- 576 **Volz LJ, Hamada M, Rothwell JC, Grefkes C.** What Makes the Muscle Twitch: Motor  
577 System Connectivity and TMS-Induced Activity. *Cerebral Cortex* 25: 2346–2353, 2015.
- 578 **Vyas S, Golub MD, Sussillo D, Shenoy KV.** Computation Through Neural Population  
579 Dynamics. *Annu Rev Neurosci* 43: 249–275, 2020.

580

581 Figure 1: Neural activity during proactive inhibition visualised through state space models.  
582 Dynamical systems models (Figure 1A) posit that neural activity exists in a multi-  
583 dimensional space with each neuron’s activity representing one axis of this space. The figure  
584 shows a 3-dimensional space made up of the activity of three individual neurons. Activity of

585 each of the neurons at any point in time can be plotted in this space. At the end of  
586 preparation, population activity occupies a subspace of this overall space, denoted by the blue  
587 circle (s). Upon the receipt of a trigger to move, population activity then evolves into the  
588 movement along a trajectory (t). Proactive inhibition might be employed by varying the  
589 trajectory of the movement (t') or by setting a different initial state (s') at the end of  
590 preparation and before the receipt of the trigger to move. This idea can be modelled as  
591 activities in PA and AP networks (Figure 1B) measured by their corresponding MEPs during  
592 movement preparation and execution. The similarity/difference between trajectories can be  
593 measured using distance metrics. Two measures can be used: the distance between two  
594 trajectories at a given time point (Euclidean distance –  $d$ ); or the angle between the two  
595 vectors drawn from the origin to each of the two points (cosine distance –  $\theta$ ).

596

597 Figure 2: The Stop-signal and Go-only tasks.

598 SST: Go trials consisted of presentation of a fixation cross, followed by a go cue (right  
599 arrow) 500 ms later. In 25% of trials, the right arrow was followed by a stop-signal (red  
600 cross) at one of four SSDs (100, 150, 200 or 250 ms after the arrow). Participants attempted  
601 to abort their button press on presentation of a stop-signal. Failure to do so resulted in the  
602 next stop-signal having a shorter SSD (-50 ms) whereas successful stopping led to the next  
603 SSD becoming longer (+50 ms). PA<sub>120</sub> or AP<sub>30</sub> TMS was delivered on go trials at one of  
604 seven time points (counterbalanced and randomised), or 1000 ms after presentation of the  
605 fixation cross (white cross) where no signals are shown (baseline trial). Go-only task:  
606 comprised of go and catch trials only; TMS was delivered at the same timepoints described  
607 above.

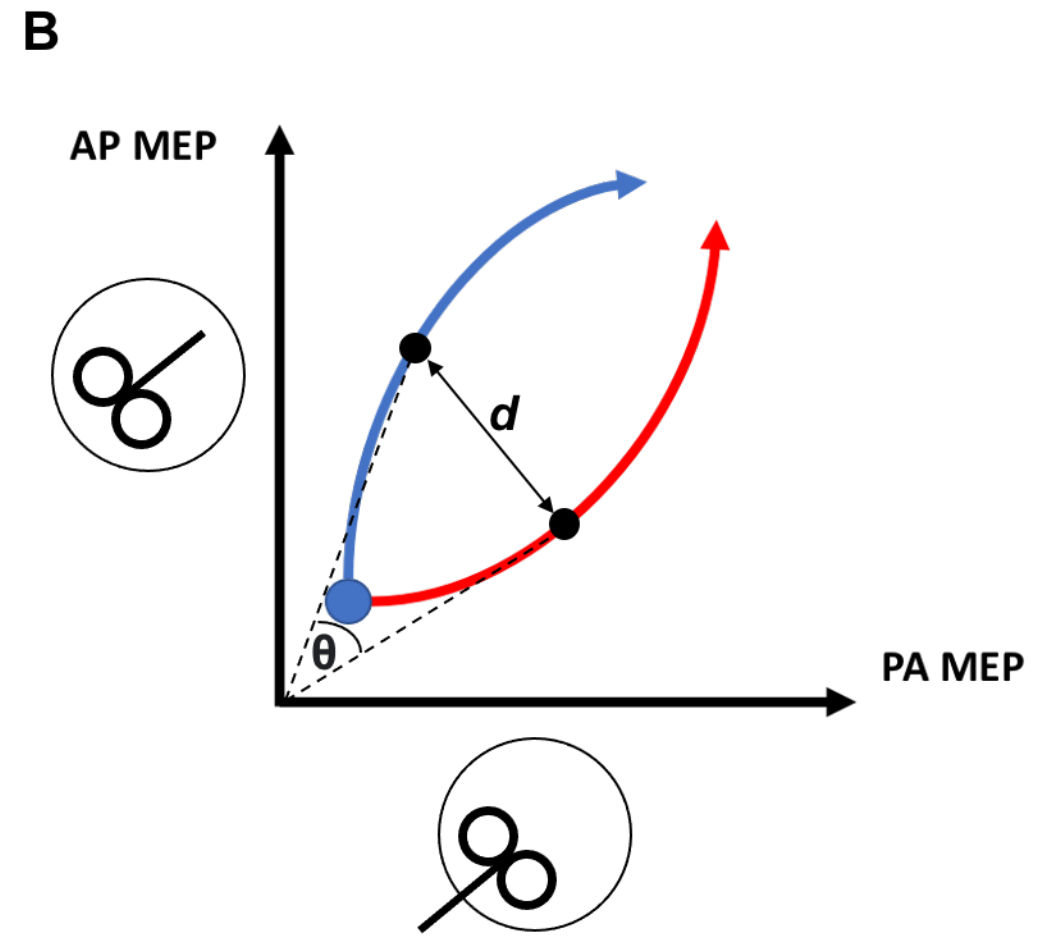
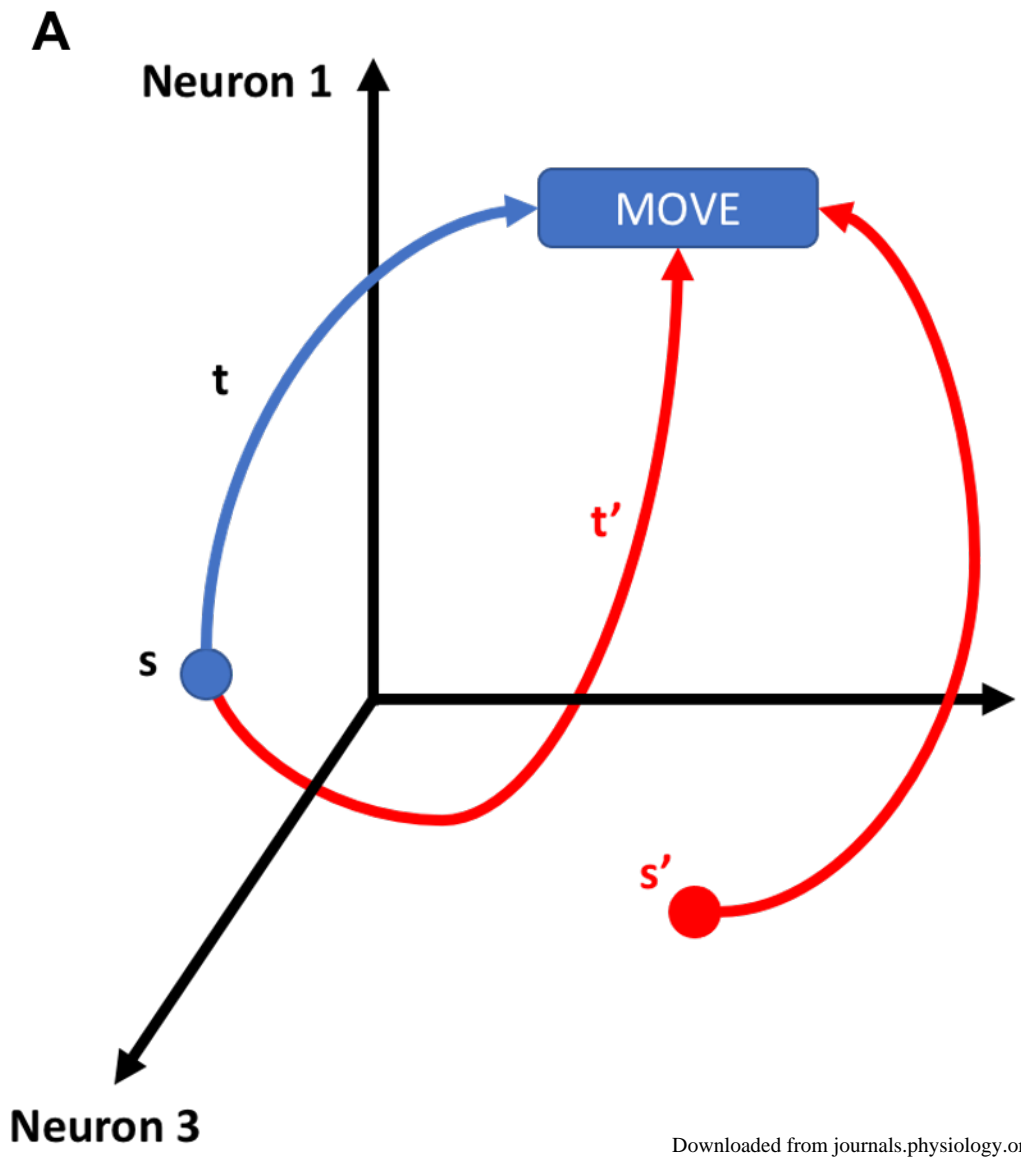
608

609 Figure 3: Corticospinal excitability changes during the SST and Go-only task for AP<sub>30</sub> and  
610 PA<sub>120</sub> TMS.

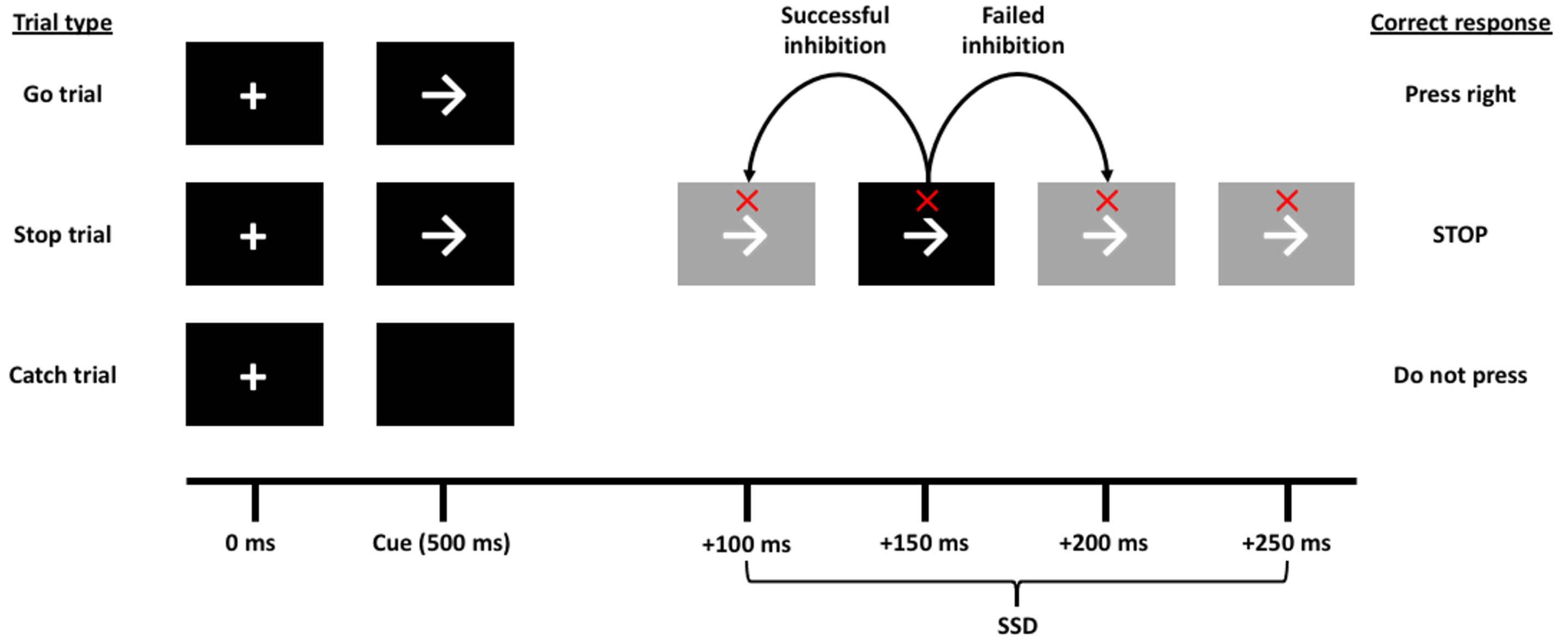
611 Top row: MEPs are taken on go trials at various times after the go cue has been presented, for  
612 the Go-only simple reaction time task and SST. Bottom row: MEPs are sorted into 50 ms bins  
613 prior the response. MEP values are normalised to baseline MEP value. Graphs represent  
614 responses evoked using PA<sub>120</sub> TMS (left column) and AP<sub>30</sub> TMS (right column). Error bars  
615 represent  $\pm$ SEM, Go-only task: blue line, SST: red line.

616

617 Figure 4: Motor cortex population-level activity during movement preparation and execution.  
618 Top row: Motor cortex population-level activity is represented as a combination between  
619  $PA_{120}$  and  $AP_{30}$  inputs. Each plot shows the trajectory taken by this population activity  
620 throughout movement. A: Cue-locked analysis: activity starts at the cue (shown by stars).  
621 Activity then progresses over time, with each marker (circle) representing a time-point (Cue,  
622 50, 100, 150, 200, 250 and 300 ms). This is shown for the SST (red line) and Go-only task  
623 (blue line). B: Response-locked analysis: stars represent population activity 50-100 ms prior  
624 to movement. Working backwards, time-points are as shown in the bottom row for Figure 3.  
625 Dashed,  $x=y$  lines represent balanced  $PA_{120}$  and  $AP_{30}$  CSE. Euclidian (middle row) and  
626 cosine (bottom row) distances are shown for our data and from bootstrapped simulated data  
627 neural trajectories drawn from the same distribution as Go-only data. Smaller values indicate  
628 greater similarity between corresponding time points.

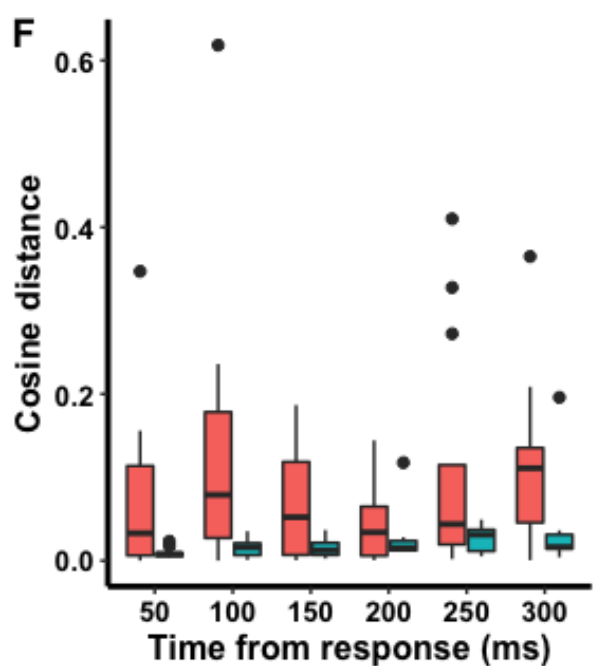
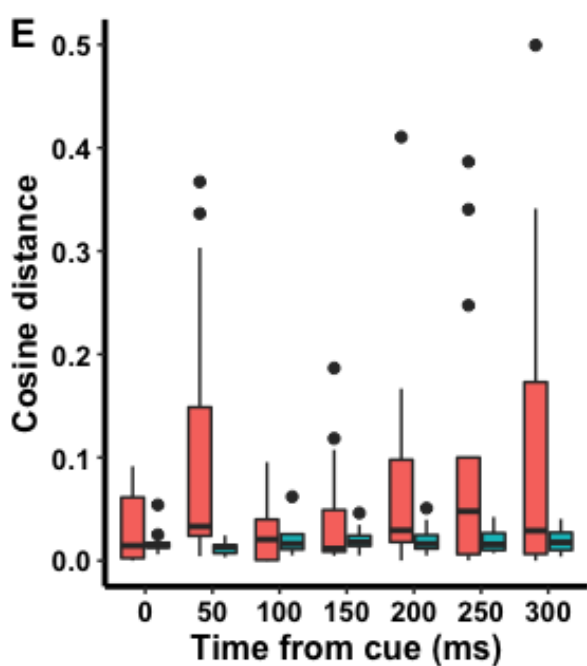
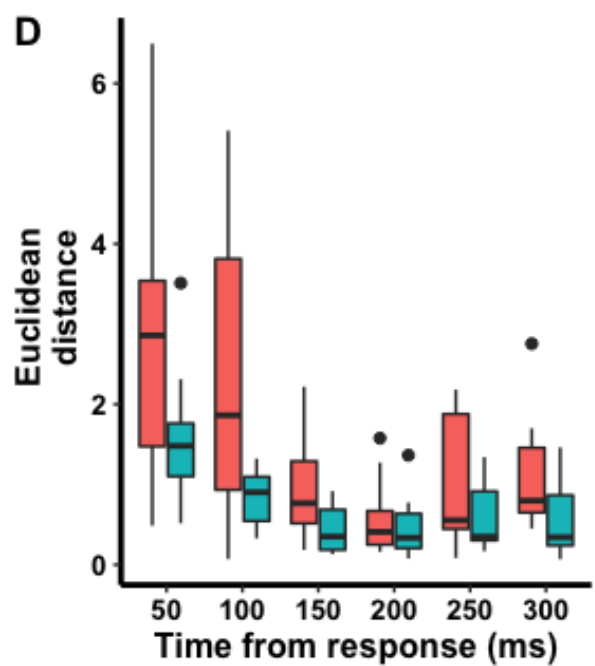
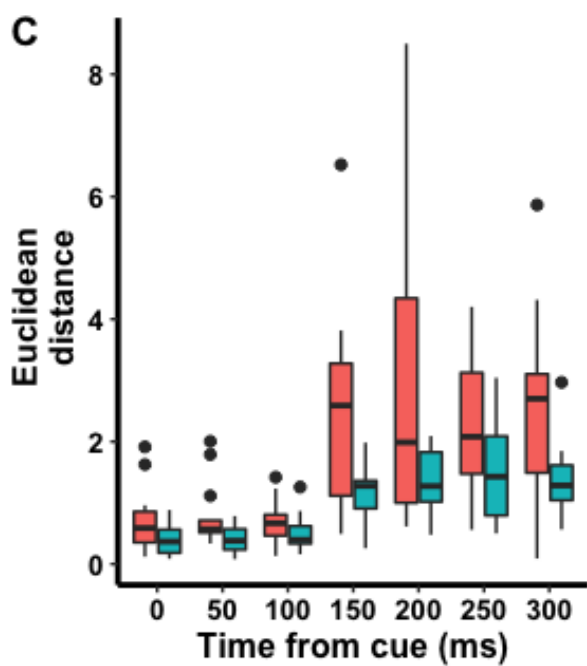
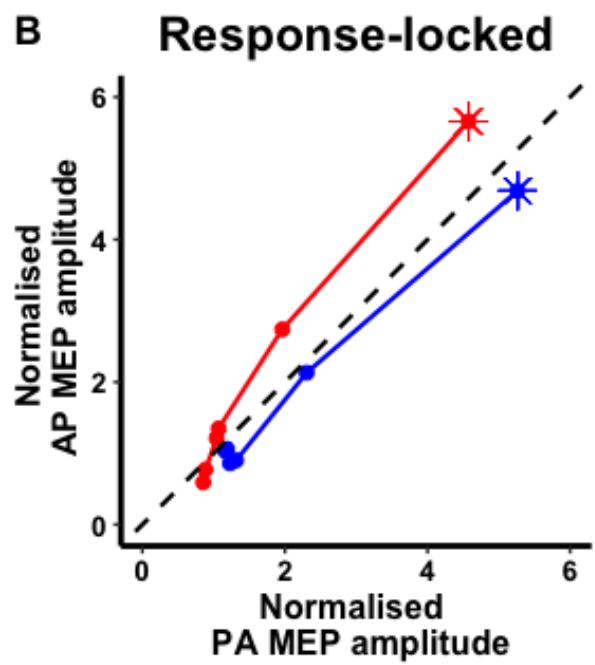
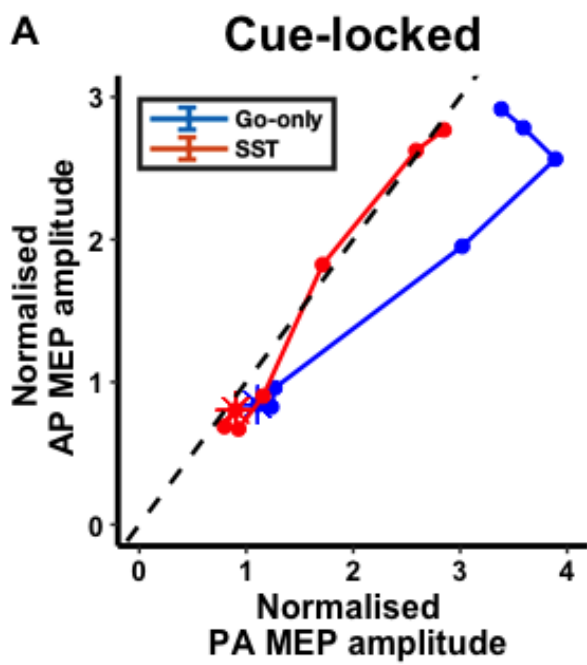


# Stop-signal task

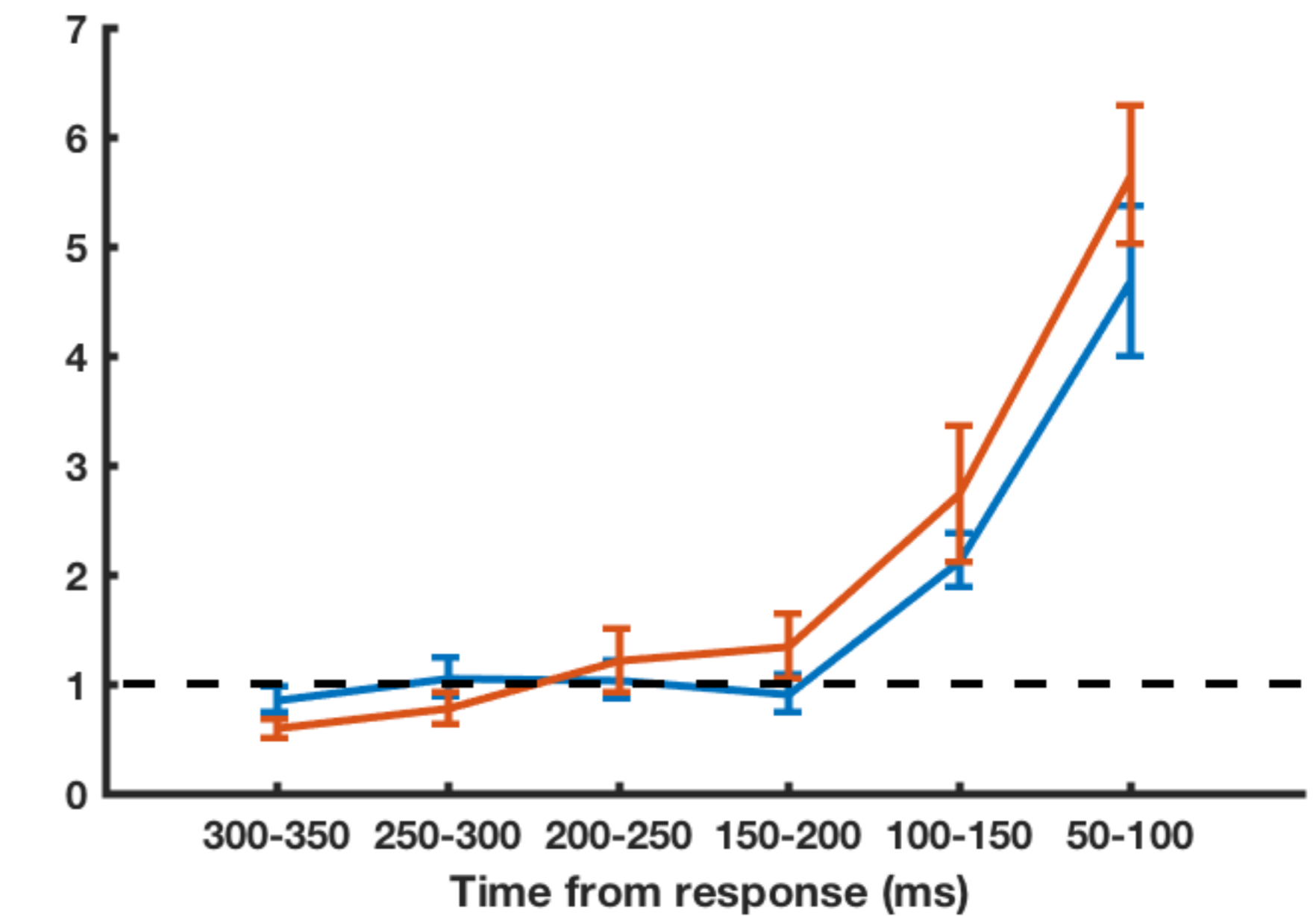
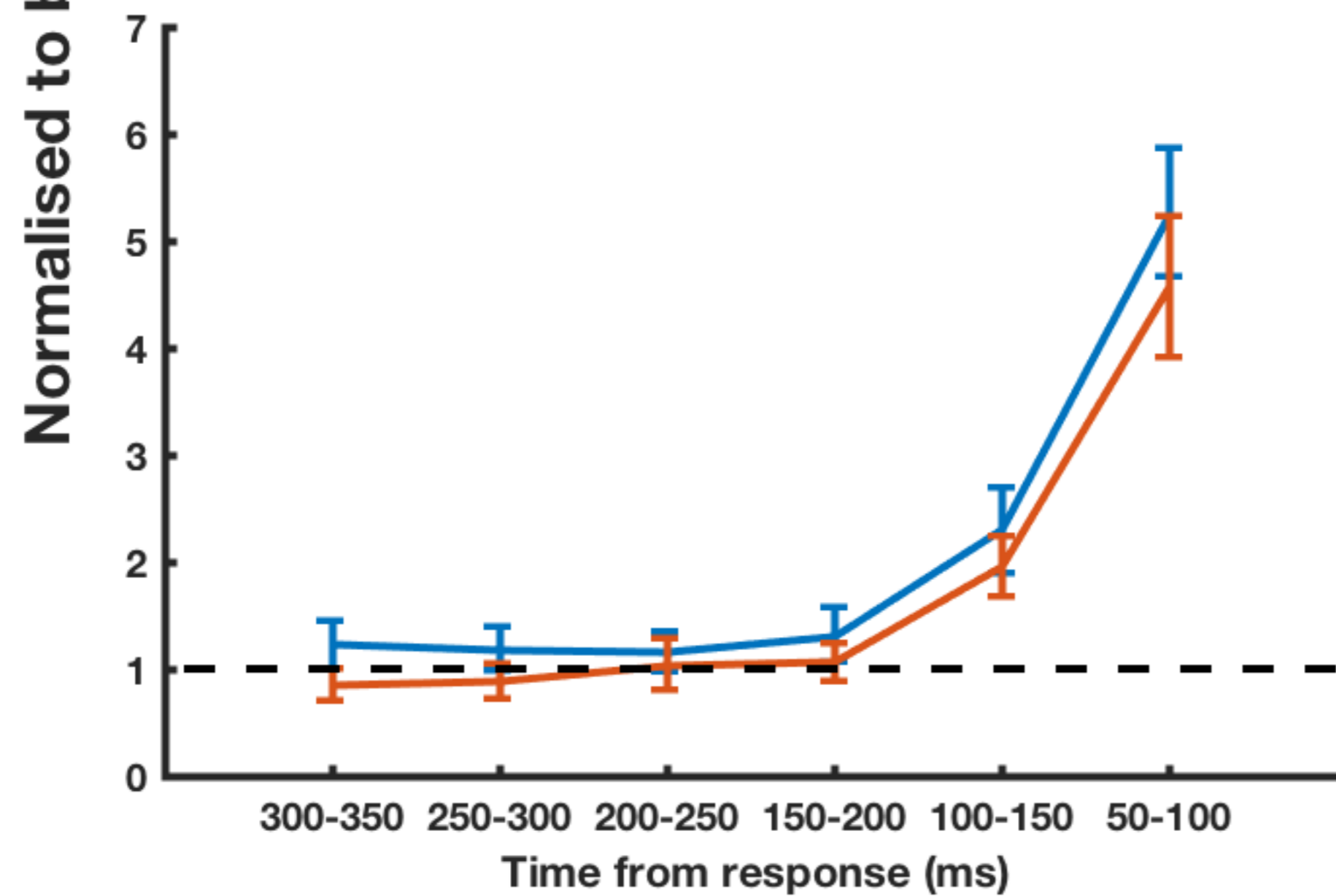
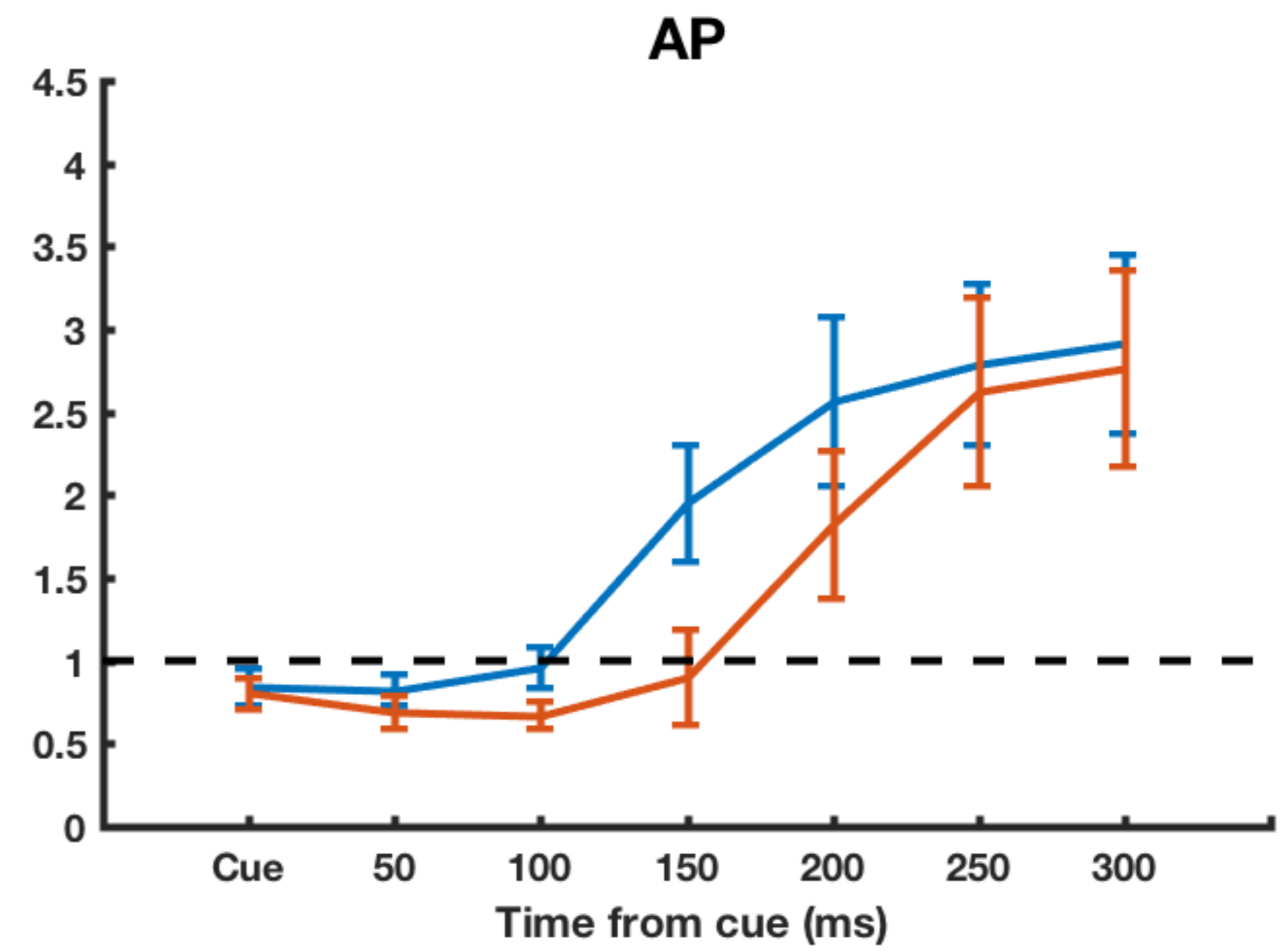
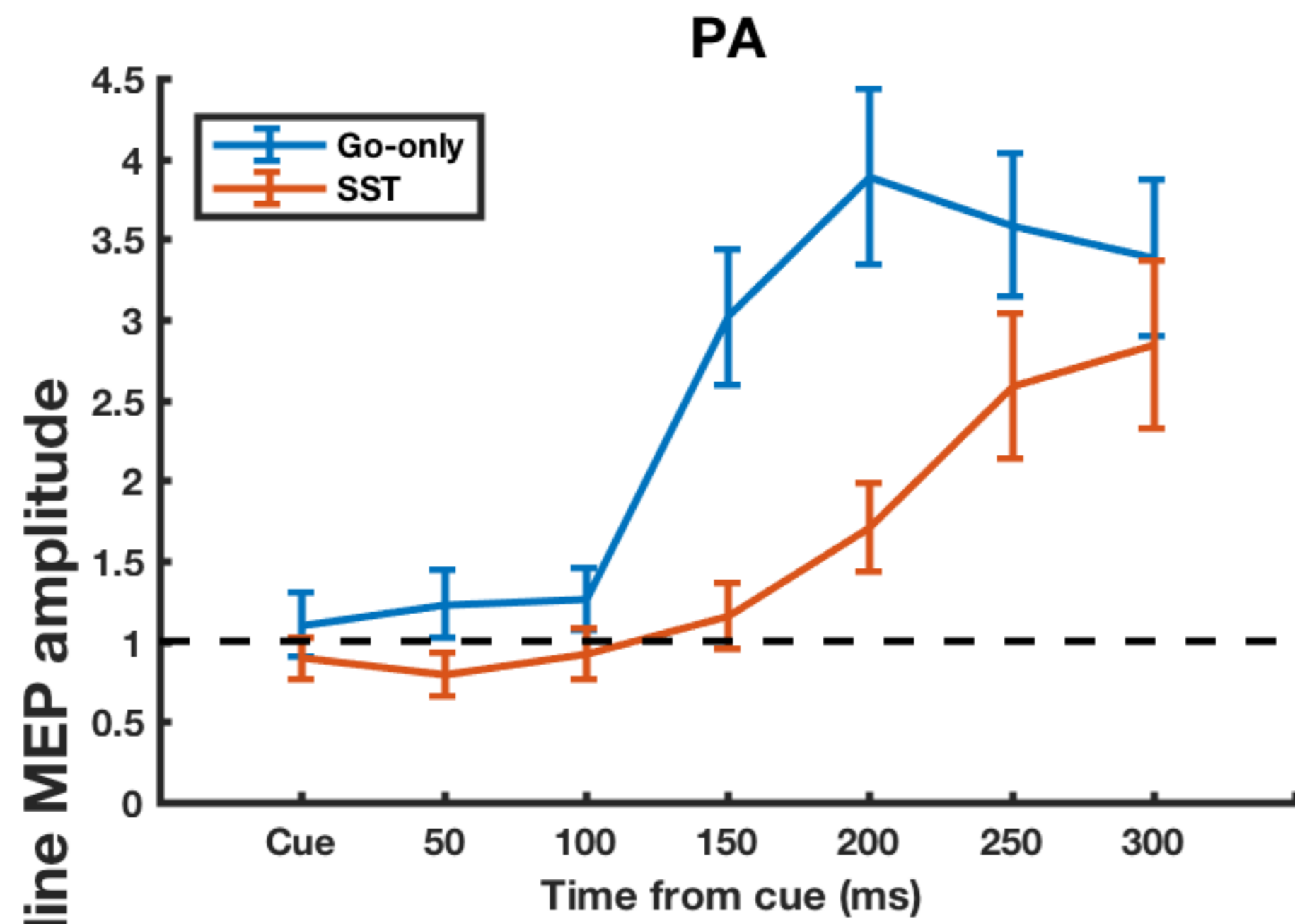


# Go-only task









Measure	Measure description	SST		Go-only	
		PA <sub>120</sub>	AP <sub>30</sub>	PA <sub>120</sub>	AP <sub>30</sub>
Go reaction time	RT to go stimulus in the critical direction	391.55 (35.01)	402.36 (44.42)	288.31 (32.12)	324.15 (52.28)
P(inhibit)	% correct inhibition	50.54 (7.36)	56.70 (11.30)		
Stop Respond	RT on failed stop trials	287.84 (33.13)	319.69 (47.90)		
Go omission	% of omissions	0.36 (0.68)	0.44 (0.84)	0.36 (0.84)	0.66 (0.98)
Stop signal delay	Delay between go and stop trials	167.05 (25.42)	185.29 (31.52)		
SSRT	Calculated time to abort response	224.50 (27.75)	216.98 (32.59)		

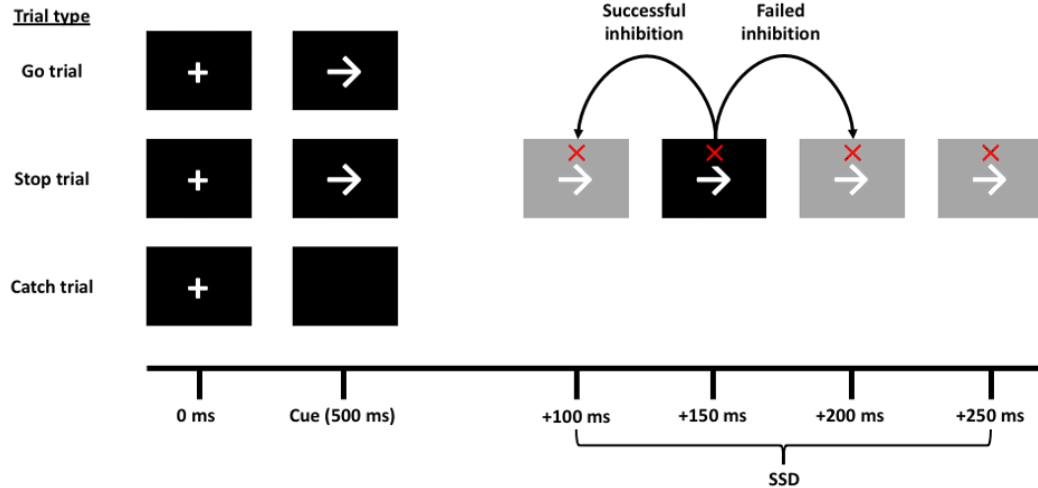
Table 1: Behavioural measurements from the SST and Go-only simple reaction time tasks.

The table shows the behavioural measures from the SST, Go-only simple reaction time task. Measures are accompanied by SD in brackets. Reaction times are given in ms. SSRT = stop signal reaction time

# Proactive inhibition is marked by differences in the pattern of motor cortex activity during movement preparation and execution

We use directional TMS during the stop-signal and Go-only tasks to investigate how motor cortex activity changes during proactive inhibition

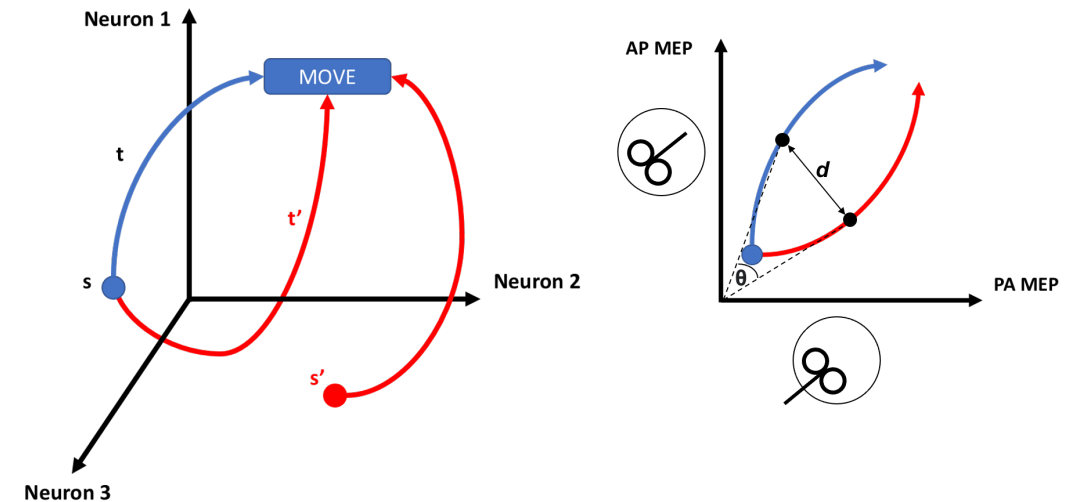
## Stop-signal task



## Go-only task



State-space models hypothesise the pattern of activity during movement preparation and execution will differ between tasks



We find the pattern of activity during preparation and execution changes when proactive inhibition is required

