

1 **Spontaneous thought and microstate activity modulation by**
2 **social imitation**

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26 **Abstract**

27 Social imitation increases well-being and closeness by mechanisms that remain poorly understood.
28 We propose that imitation impacts behavioural states in part by modulating post-imitation mind-
29 wandering. The human mind wanders spontaneously and frequently, revisiting the past and
30 imagining the future of self and of others. External and internal factors can influence wandering
31 spontaneous thoughts, whose content predicts subsequent emotional states. In 43 young subjects,
32 we find that imitating the arm movements of an actor alters the dynamics and the content of
33 subsequent resting-state spontaneous thoughts. Imitation-sensitive features of spontaneous
34 thoughts correlate with both behavioural states and salivary oxytocin levels. EEG microstate
35 analysis reveals that global patterns of correlated neuronal activity predict imitation-induced
36 changes in spontaneous thoughts. Thus, imitation can modulate ongoing activity in specific neural
37 networks to change spontaneous thought patterns as a function of oxytocin levels, and to ultimately
38 orchestrate behavioural states.

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50 **Introduction**

51 Humans spend a significant fraction of their waking time defaulting to ‘mind-wandering’, an
52 unconstrained succession of mental states that generate spontaneous thoughts¹⁻⁴. The content of
53 spontaneous thought varies widely within and between individuals, and can be characterized with
54 respect to its dynamics (more or less fragmented), affect (negative, neutral, positive), temporal
55 orientation (remembering or planning), social orientation (self or others), mental modality (visual
56 or verbal), and association with physiological states (sleepiness, stress, etc)^{3, 5-8}. Many of these
57 dimensions can be captured in experimental settings by validated retrospective self-reported
58 questionnaires, like the Amsterdam Resting-State Questionnaire (ARSQ)⁹.

59 Spontaneous thoughts correlate with explicit affective state. Negative mood and spontaneous
60 thoughts interact circularly, as sad spontaneous thoughts tend to be both preceded and succeeded
61 by sad or anxious moods^{7,10}. Socio-temporal features of spontaneous thoughts can also predict
62 subsequent mood, where past- and other-oriented thoughts predict negative affect, but future- and
63 self-oriented thoughts predict improved mood⁷. Other studies, however, found that self-related
64 content of spontaneous thoughts correlates with sad mood and with depressive symptoms¹¹.

65 Spontaneous thoughts can serve important roles in mental well-being, including in generating
66 personal goals¹²⁻¹⁴, in memory consolidation^{15,16}, and in fostering creativity^{14, 16-18}. In other words,
67 spontaneous thought profiles can improve behavioural states and serve as internal tools for
68 cognitive and emotional well-being. Could these tools be leveraged in behavioural interventions?
69 This possibility is bolstered by the documented impact of contextual factors on spontaneous
70 thoughts⁸. However, much more remains to be understood about the nature of contexts capable of
71 changing mind-wandering, and about the mechanisms by which they achieve such changes.

72 Spontaneous thoughts likely arise from task-independent ongoing brain activity^{19,20}. Functional
73 imaging implicates several brain networks in spontaneous thought. Primarily, the default mode
74 network (DMN), which is active during states of rest, and decreases as a consequence of task
75 demands²⁰⁻²² is closely linked to the generation of spontaneous thoughts²³. In addition to the DMN,
76 mind-wandering also engages salience and executive control networks^{21,24,25}. These networks are
77 believed to impose automatic and deliberate constraints on mind-wandering, or in other words they
78 track the ‘wandering path’ from one spontaneous thought to another⁴.

79 The logistics of functional imaging (lying still in a loud scanner) might bias the nature of
80 spontaneous thoughts. A more naturalistic approach to probing the neural substrates of mind-
81 wandering is to perform scalp electroencephalographic (EEG) recordings, and characterize
82 activity-correlated networks using microstate analysis. This analysis identifies transiently (60-120
83 ms) quasi-stable global patterns of scalp potential topographies that are highly reproducible within
84 and across subjects^{26,27}. Between four and seven such states have been identified so far, and have
85 been linked to fMRI resting-state networks^{26,28,29,30,31}. Several studies using simultaneous EEG and
86 fMRI, or direct EEG source localization methodology, support the notion that microstates A and
87 B are associated with primary sensory brain regions like visual, auditory/language cortices, while
88 C, D, and E microstates associate with core regions of the posterior DMN, attention/cognitive
89 control, and salience resting-state networks, respectively^{26,31}. Potential roles of these microstates
90 in visual imagery, cognitive control and planning have been described and also disputed²⁶.
91 However, a distinct pattern of modulations between the posterior DMN C state and attention D
92 state might reinforce their functional meaning. Relaxed, meditative, and hypnotic states entrain
93 longer and more frequent posterior DMN C states, and slower attention D temporal dynamics^{26,32-}
94 ³⁵. Microstate analysis could deepen the understanding of neural substrates for spontaneous
95 thoughts. It could also identify mechanisms by which contextual factors change the pattern and
96 spontaneous thoughts and their effects on behaviour.

97 Social factors represent a major source of context variation in humans, however little is known
98 about the relationship between social contexts and spontaneous thoughts. In the current study, we
99 hypothesize that social contexts can change subsequent resting-state spontaneous thoughts, and
100 microstate activity patterns. To test this, we used a dyadic social imitation task, where subjects
101 follow the arm movements of an actor. We previously showed that this form of social imitation
102 decreases momentary stress, increases well-being and social closeness³⁶. Several mechanisms for
103 these behavioural changes have been proposed, including rewarding effects, and perception-action
104 matching^{36,37}. Hormonal substrates have also been implicated, particularly increased levels of
105 oxytocin^{36,38}. We reasoned that in addition to the above mechanisms, imitation could modify
106 behavioural states indirectly, by changing ongoing brain activity and the pattern of spontaneous
107 thoughts. We used a combination of behavioural, biochemical and physiological measures to show
108 that spontaneous thought patterns and ongoing activity of neural networks are sensitive to social
109 imitation. Our findings indicate a potential mechanism by which social imitation, which is already

110 used in theatre as an exercise that decreases stage fright and increases closeness between actors,
111 could be therapeutic in the general population.

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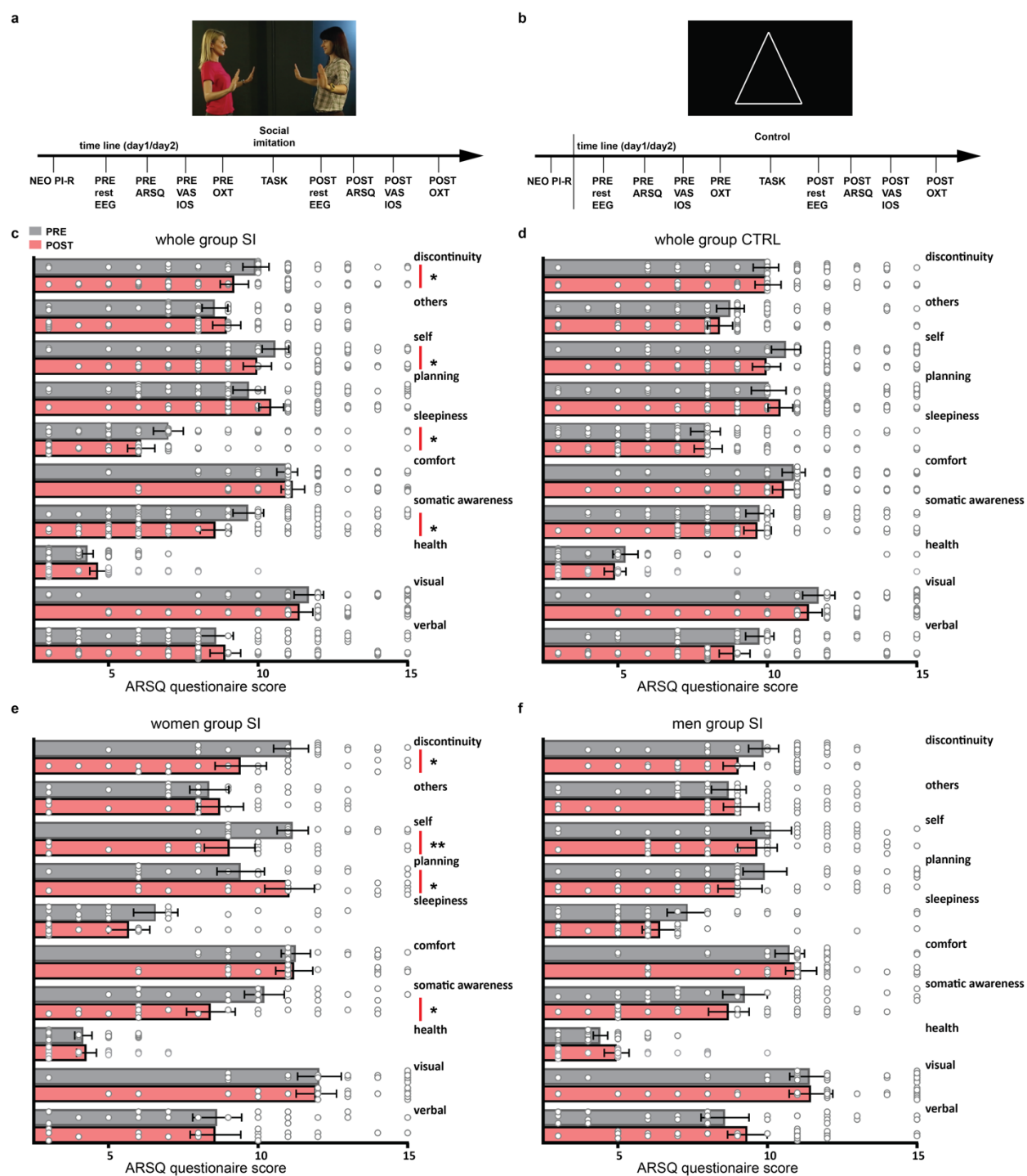
113 **Results**

114 To determine if brief dyadic social imitation (SI) might alter spontaneous thoughts and ongoing
115 activity of neural network in imitators, we probed five-minute resting-state episodes before (PRE)
116 and after (POST) the task (**Fig. 1a**). The SI task consisted of a three-minute interaction where the
117 subject followed the arm movements of an actor, as previously described³⁶. For the control, non-
118 social condition (CTRL), subjects were asked to depict with their arm movements the shape of
119 geometrical figures displayed on a computer screen (**Fig. 1b**). Within each subject, experimental
120 and control conditions were completed on separate days, and in random order.

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122 **Modulation of resting-state spontaneous thoughts by social imitation**

123 To quantify the changes in spontaneous-thoughts during resting-state episodes, we evaluated
124 intrinsic mentation using a standardized test, the ARSQ 2.0 questionnaire⁹ that allowed us to
125 construct a repertoire of the individual subjective mind-wandering experience and their
126 modulations by tasks. In order to investigate if and how SI modulates spontaneous thoughts, we
127 performed three-way repeated measures ANOVA with the ARSQ factors, by time (PRE vs POST)
128 and condition (SI vs CTRL). At the group level, as expected from the structure of the test, we
129 found a main effect of ARSQ factors ($F(3, 378)=38.2, p<0.00001, 0.47 \eta_p^2$). We also found a
130 significant main effect of time ($F(1,42)=4.38, p=0.042, 0.09 \eta_p^2, PRE=9.30\pm 0.17, POST=9.05$
131 $\pm 0.18, N=43$), and a significant three way interaction between time, ARSQ factors and conditions
132 ($F(9,378)=2.03, p=0.03, 0.04 \eta_p^2$). After the SI task, subjects reported less fragmented thoughts,
133 fewer thoughts about themselves and their bodies, and felt less tired (**Fig. 1c**). Post-hoc tests
134 revealed significant changes induced by SI in four different factor categories: *discontinuity of mind*
135 ($PRE= 10.41\pm 2.53, POST=9.20\pm 3.09, p=0.005, N=43$), *self-related thoughts* ($PRE=10.58\pm 2.91,$
136 $POST=9.39\pm 3.4, p=0.0006$), *sleepiness* ($PRE=7.0\pm 3.28, POST=6.09\pm 2.9, p=0.036$) and *somatic*
137 *awareness* ($PRE=9.67\pm 3.34, POST=8.58\pm 3.38, p=0.012$) (**Fig. 1c**). No significant changes were



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Figure 1. SI changes spontaneous thoughts. (a) Experimental design, SI task. (b) Design of control task. (c) Resting-state ARSQ scores before (PRE, gray) and after (POST, red) the SI task. (d) Resting-state ARSQ scores before and after CTRL task. (e) Effects of SI on ARSQ scores in women. (f) Effects of SI on ARSQ scores in men. Errors are SEM, * $p < 0.05$, ** $p < 0.001$.

139 detected after the CTRL conditions (**Fig. 1d**). This indicates that imitating another person, but not
140 an inanimate screen, can lead to changes in subsequent spontaneous thought dynamics (less
141 fragmented), content (less about self), and association with physiological states (less tired).

142 When we separated the group by gender, we found that women account for most of the SI-induced
143 changes in spontaneous thoughts. There was a significant main effect of time PRE-POST
144 ($F(1,18)=4.7$, $p=0.043$, $0.20 \eta_p^2$, $PRE=9.45\pm 0.28$, $POST=9.11 \pm 0.29$), and a significant three way
145 interaction between time, ARSQ factors and conditions ($F(9,162)=2.83$, $p=0.03$, $0.04 \eta_p^2$).
146 Following SI, women reported having significantly less fragmented thoughts (*discontinuity*, $PRE=$
147 11.1 ± 2.53 , $POST=9.42\pm 3.7$, $p=0.004$), thinking less about themselves (*self*, $PRE=11.1\pm 2.26$
148 $POST=9.05\pm 3.6$, $p=0.0003$) and their bodies (*somatic awareness*, $PRE=10.21.67\pm 2.89$,
149 $POST=8.42\pm 3.54$, $p=0.002$) (**Fig. 1e**). Moreover, women reported a significant increase of the
150 *planning* factor ($PRE=9.42\pm 3.45$, $POST=11.06$, $p=0.005$) (**Fig. 1e**). On the contrary, in men we
151 found neither a main effect of time ($F(1,22)=0.6$, $p=0.44$) nor a significant interaction between
152 time, ARSQ factors and conditions ($F(9,198)=1.48$, $p=1.56$) (**Fig. 1f**).

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154 **Associations between spontaneous thoughts and behavioural states**

155 We previously showed that SI but not the CTRL task can increase well-being and social closeness,
156 whereas both SI and CTRL conditions lead to decreased self-reported stress³⁶. To determine if
157 there is an association between these momentary self-reported behavioural states and spontaneous
158 thoughts, we performed rank correlation analyses between behavioural and ARSQ scores across
159 all conditions (**Fig. 2**). In women we found a significant positive correlation between *discontinuity*
160 *of mind* and perceived stress level (Gamma $r=0.36$, $p=0.033$), that was not significant either at the
161 group level or in men (**Fig. 2a**). We also found a positive association between the level of perceived
162 stress and *self*-oriented thoughts (Gamma $r=0.35$, $p=0.001$, **Fig. 2b**) and thoughts about *health*
163 (Gamma $r=0.29$, $p=0.007$, **Fig. 2d**). Stress was negatively associated with thoughts about *comfort*
164 (Gamma $r= -0.31$, $p=0.004$, **Fig. 2d**). The reversed pattern of correlation was found for wellbeing
165 (*health*: Gamma $r= -0.33$, $p= 0.002$; *comfort*: Gamma $r=0.56$, $p<0.0001$, **Fig 2c,d**). Thoughts about
166 future *plans* were positively associated with both closeness (Gamma $r=0.24$, $p=0.02$) and well-
167 being (Gamma $r=0.21$, $p= 0.04$) (**Fig. 2d**). These findings indicate a relationship between

168 behavioural states and the pattern of spontaneous thoughts, both of which change after SI,
 169 particularly in women.

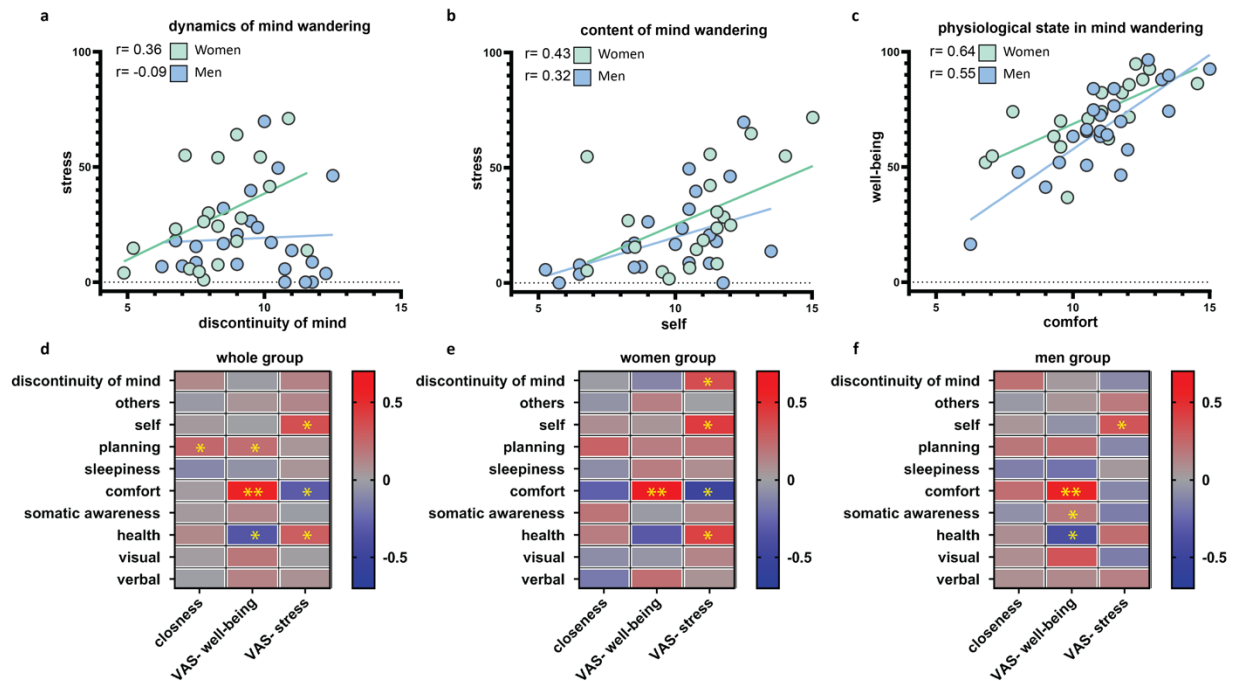


Figure 2. The pattern of spontaneous thoughts associates with subjective behavioral states. (a) Dynamics of spontaneous thought positively associate with self-reported stress levels. (b) Self-centered thought content positively correlates with stress. (c) Perceived states of ‘comfort’ during mind-wandering positively correlate with self-reported well-being. (d) Summary data of correlations (r values) between ARSQ scores and self-reported behavioral state measures for the entire group. (e) Summary of correlations for women. (f) Summary of correlations for men, *p<0.05, **p<0.001.

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172 Associations between spontaneous thoughts and salivary oxytocin

173 Imitation has been linked to increased levels of salivary oxytocin^{36,38}. We previously showed that
 174 this increase is only partially related to the outcome of the SI on self-reported behavioural states³⁶.
 175 In order to find how spontaneous thoughts across conditions relate to salivary oxytocin levels, we
 176 calculated correlations that showed several significant negative associations of salivary OXT
 177 dynamics with *discontinuity of mind* (Gamma, r=-0.25, p= 0.029, **Fig. 3a**), with thoughts about

178 future *planning* (Gamma, $r=-0.25$, $p=0.03$, **Fig. 3b**) and, only in the men group, with *comfort*
 179 (Gamma, $r=-0.25$, $p=0.03$) (**Fig. 3c**). Therefore, while increased salivary oxytocin could represent
 180 a potential mechanism for changing the dynamics of spontaneous thoughts (i.e., less fragmented
 181 thinking), it does not appear to affect their content or their association with physiological states.

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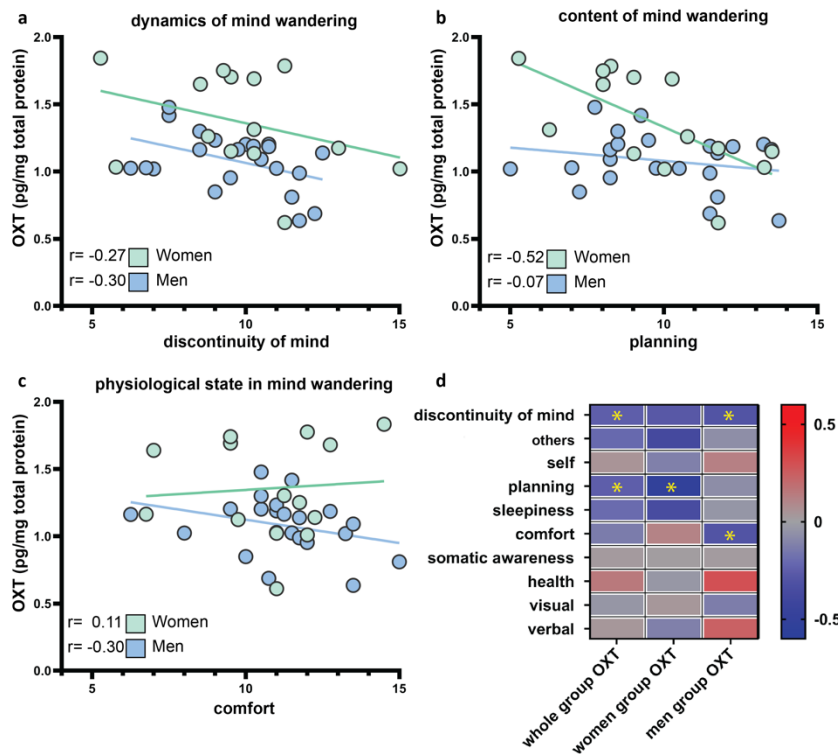


Figure 3. Spontaneous thought patterns associate with salivary OXT levels.

(a) Dynamics of mind-wandering negatively correlate with salivary OXT.

(b) Thoughts about 'planning' negatively correlate with OXT in women.

(c) States of 'comfort' negatively correlate with OXT.

(d) Summary data of correlations between ARSQ scores and salivary OXT, $*p < 0.05$.

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185 Neural correlates for SI-induced changes in resting-state activity

186 To investigate the neural substrate for SI-induced changes in mentation, we recorded EEG activity
 187 during resting-state. Using a set of seven different independent criteria (described in **Methods**),
 188 we determined that five microstates can optimally describe group topographical variability. Across
 189 all conditions (PRE, POST before and after the SI/CTRL) and individuals (N=43), summing more
 190 than 800 individual resting-state dominant topographies, the cluster analysis robustly identified A,

191 B, C, D and E prototypical microstates that explained 81.9 % of variance (**Fig. 4a**). We then
192 analysed SI-induced changes in the duration and occurrence of these EEG microstates.

193

194 **SI modulates the intrinsic temporal dynamics of microstates**

195 To identify how the SI task might modulate intrinsic patterns of resting networks dynamics, we
196 performed repeated-measures three-way ANOVA analysis on the *duration (ms)* and *occurrence*
197 (*Hz*) of microstates, by time (PRE vs POST) and by condition (SI vs CTRL). Separate paired
198 Wilcoxon tests were performed to identify the specific task modulation for each microstate.

199 In the ANOVA analysis for microstate *duration* we found a main effect of time (N=43,
200 PRE=83.8±5.50, POST=85.45±5.55, F(1,42)=10.7, p=0.002, 0.20 η_p^2) and a main effect of
201 microstates (A=75.6±6.71, B=88.5±16.06, C=103.71±18.71, D=76.19±9.84, E=79.16±9.84,
202 F(1,42)=36.53, p<0.00001, 0.46 η_p^2 , **Fig. 4b**). After the SI task, two classes of microstates showed
203 increased *duration*, microstate A (PRE=74.5±6.5, POST=76.5±8.1, p=0.006) and microstate C
204 (PRE=102.6, ±20.3, POST=105±18.7, p=0.04, **Fig. 4b**). We also observed a significant increase
205 in microstate E *duration* after the CTRL task (PRE=77.3±11.3, POST=80±13.3, p=0.006, **Fig. 4c**).
206 Thus, as hypothesized, the SI task modulated a different pattern of network dynamics than the
207 CTRL task.

208 Regarding *occurrence*, we observe neither a significant ANOVA time effect nor a significant time,
209 condition and microstate interaction. However, we found a main effect of microstates
210 (A=1.38±0.41, B=1.95±0.48, C=2.37±0.35, D=1.46±0.69, E=1.49±0.48 F(4,168)=30.7,
211 p<0.00001, 0.42 η_p^2), and a significant time by microstate interaction (A- PRE=1.44±0.37 vs
212 POST=1.49±0.39, B-PRE=1.95±0.47 vs POST=1.91±0.43, C – PRE=2.33±0.35 vs
213 POST=2.35±0.27, D-PRE=1.53±0.62 vs POST=1.45±0.67, E –PRE-1.49±0.49 vs
214 POST=1.52±0.5, F(4,168)=2.47, p=0.04, 0.05 η_p^2). Post-hoc tests showed that SI specifically
215 decreases the occurrence of microstate D (PRE=1.51 ±0.67, POST=1.41±0.72, p=0.012, **Fig. 4b**).

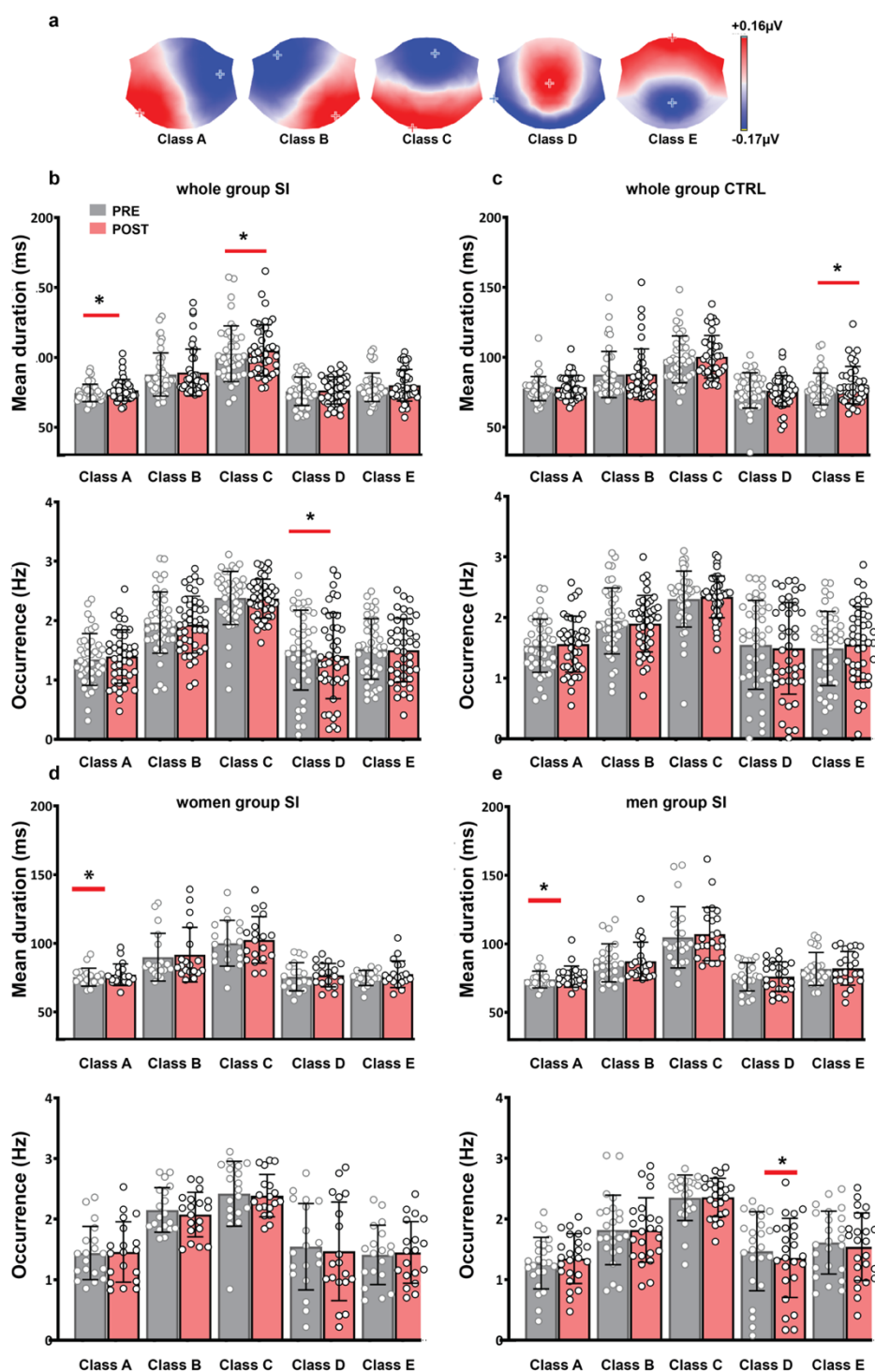


Figure 4. Microstate dynamics following social imitation. (a) Identified classes of microstates. **(b)** SI-induced changes in mean duration (top) and occurrence (bottom) for each class. **(c)** Changes in mean duration (top) and occurrence (bottom) following CTRL task. **(d)** SI-induced changes in class duration (top) and occurrence (bottom) in women. **(e)** SI-induced changes in class duration (top) and occurrence (bottom) in men, * $p < 0.05$.

217 In women, we found a substantial effect of time ($N=19$, $PRE=82.12\pm 4.11$, $POST=85.12\pm 5.10$, F
218 $(1,18)=11.9$, $p=0.002$, $0.39 \eta_p^2$), a main effect of condition ($SI=84.12\pm 4.42$, $CTRL=82.40\pm 4.21$,
219 $F=(1,18)=9.54$, $p=0.006$, $0.34 \eta_p^2$) and a main effect of microstates *duration* ($A=76.24\pm 6.94$,
220 $B=90.78\pm 18.5$, $C=101.24\pm 16.17$, $D=76.2\pm 9.2$, $E=76.15\pm 7.3$ $F(4, 72)=12.57$, $p<0.0001$, $0.41 \eta_p^2$).
221 At the level of paired Wilcoxon analyses, we found a significant POST SI increased *duration* of
222 microstates A ($N=19$, $PRE=75.36\pm 6.47$, $POST=77.13\pm 7.8$, $p=0.04$), and a trend for increased E
223 microstates *duration* ($N=19$, $PRE=74.77\pm 5.48$, $POST=77.52\pm 9.73$, $p=0.07$, **Fig. 4d**). After the
224 CTRL task we found a significant increased E microstates *duration* ($N=19$, $PRE=73.95\pm 9.15$,
225 $POST=76.28\pm 9.36$, $p=0.02$).

226 In men, we found a significant main effect of microstates ($N=24$, $A=75.01\pm 6.51$, $B=86.67\pm 13.59$,
227 $C=105.9\pm 20.26$, $D=76\pm 10.36$, $E=81.89\pm 10.86$, $F(4,88)=31.12$, $p<0.00001$). At the level of paired
228 Wilcoxon analyses, there was a significant difference between PRE and POST SI microstates class
229 A *duration* in men ($N=24$, $PRE=73.96\pm 6$, $POST=76.05\pm 7.73$, $p=0.04$), and also a trend for class
230 C increased duration ($N=24$, $PRE=104.72\pm 22.3$, $POST=107\pm 19.44$ $p=0.063$, **Fig. 4e**). After the
231 CTRL task we found a trend for increased E microstates *duration* ($N=24$, $PRE=79.95\pm 12.42$,
232 $POST=83.01\pm 15.31$, $p=0.09$).

233 With respect to microstate *occurrence* for women and men, we found a main effect of microstates
234 in both women ($A=1.53\pm 0.39$, $B=2.10\pm 0.39$, $C=2.34\pm 0.33$, $D=1.48\pm 0.76$, $E=1.43\pm 0.49$
235 $F(4,72)=12.1$, $p<0.00001$, $0.40 \eta_p^2$, **Fig. 4d**) and men ($A=1.39\pm 0.33$, $B=1.79\pm 0.43$, $C=2.34\pm 0.25$,
236 $D=1.48\pm 0.52$, $E=1.57\pm 0.46$ $F(4,84)=24.5$, $p<0.0001$, $0.53 \eta_p^2$, **Fig. 4e**). The separate paired
237 Wilcoxon analyses by gender showed decreased occurrence for class D only in men
238 ($PRE=1.46\pm 0.65$, $POST=1.35\pm 0.65$, $p=0.005$, **Fig. 4e**).

239

240 **Microstate transition probabilities**

241 In addition to the temporal characteristics for each microstate, we also investigated if SI affects
242 the probability of transitions between microstates. For each subject we computed the number of
243 transitions from each of the five classes to any of the other classes. These values were then
244 normalized to all between-class transitions to obtain the final fractions for each of the twenty

245 transitions pair ($A \rightarrow B$, $A \rightarrow C$, $A \rightarrow D$, $A \rightarrow E$, etc). The repeated three-way ANOVA revealed a
 246 main effect of time (PRE=0.0482±0.0012 vs POST=0.0486±0.0009 $F(1,42)=10.9$, $p=0.001$, 0.20
 247 η_p^2), and a main effect of transition pair $F(19,798)=25.9$, $p<0.00001$, 0.38 η_p^2). In addition, we
 248 identified several two-way significant interactions: the time by transition pair ($F(19,798)=0.001$,
 249 0.05 η_p^2), and the condition by transition ($F(19,798)=1.63$, $p=0.04$, 0.03 η_p^2). The $B \rightarrow A$
 250 (SI=0.045±0.02, CTRL=0.053±0.03, $p=0.048$), $B \rightarrow C$ (SI=0.092±0.04, CTRL=0.081±0.04,
 251 $p=0.01$) and $C \rightarrow B$ transitions (SI=0.091±0.04, CTRL=0.082±0.04, $p=0.02$) differed significantly
 252 between the SI and CTRL conditions. Whereas the $A \rightarrow C$ (PRE=0.052±0.01, POST=0.056±0.02,
 253 $p=0.021$, $C \rightarrow A$ (PRE=0.052±0.01, POST=0.056±0.01, $p=0.021$), $C \rightarrow D$ (PRE=0.059±0.02,
 254 POST=0.056±0.02, $p=0.031$), $C \rightarrow E$ (PRE=0.062±0.03, POST=0.067±0.03, $p=0.0008$) and $E \rightarrow C$
 255 (PRE=0.063±0.03, POST=0.067±0.03, $p=0.0062$) differed significantly between PRE and POST.

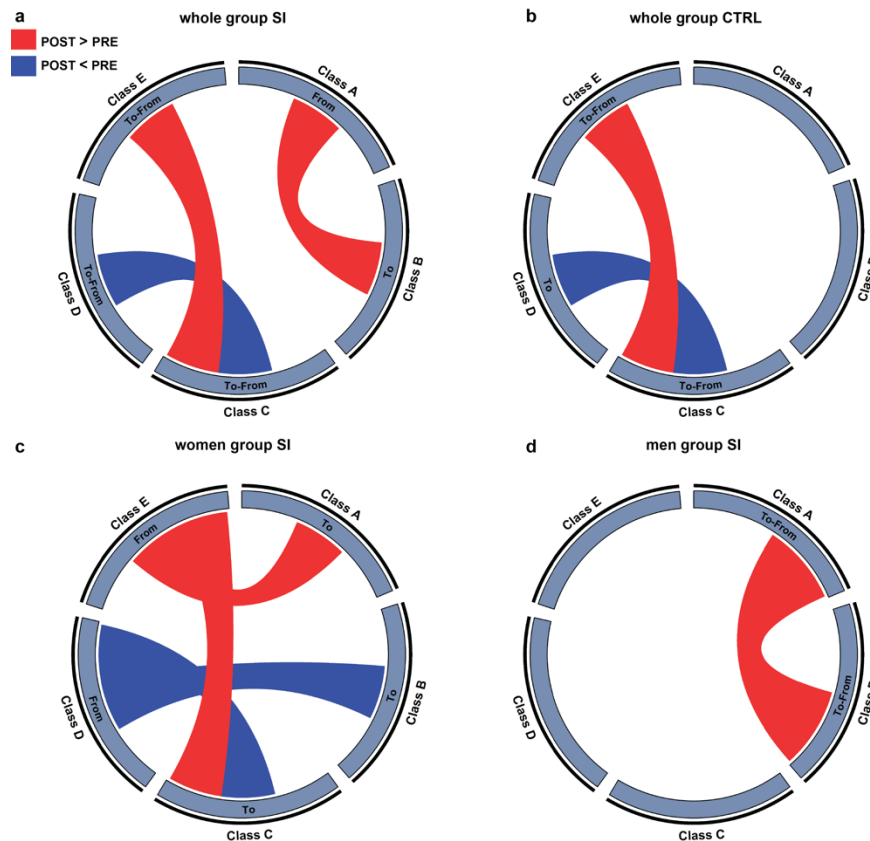


Figure 5. Microstate transition probability following social imitation.

(a) Significant changes in transition probability following SI task.
 (b) Significant changes in transition probability following CTRL task.
 (c) Significant changes in transition probability following SI in women.
 (d) Significant changes in transition probability following SI in men.

259 Separate paired Wilcoxon analyses enabled the identification of transition pairs that showed
 260 significant modulations by the SI task (Fig. 5a-d). There were several microstate pairs that showed
 261 increased transition after the SI: $A \rightarrow B$ (PRE=0.043±0.02, POST=0.046±0.02, $p=0.019$), $C \rightarrow E$

262 (PRE=0.063±0.04, POST=0.069±0.04, p=0.034) and E→C (PRE=0.067±0.04, POST=0.07±0.03,
263 p=0.034). We also noted decreased frequency of transitions between C and D microstates: C→D
264 (PRE=0.058±0.02, POST=0.056±0.02, p=0.04), D→C (PRE=0.058±0.02, POST=0.055±0.02,
265 p=0.031) (**Fig. 5a**). However, we also found significant transition modulations following the
266 CTRL task, when C→D decreased (PRE=0.061±0.03, POST=0.057±0.03, p=0.023), and C→E
267 (PRE=0.061±0.03, POST=0.066±0.03, p=0.0063) and E→C (PRE=0.058±0.03,
268 POST=0.065±0.03, p=0.027) increased in frequency (**Fig. 5b**).

269 In gender subgroups we found a main effect of transition pair in both women (F(19, 342)=11.28,
270 p<0.00001, 0.38 η_p^2) and men (F(19, 437)=20.62, p<0.00001, 0.47 η_p^2), and a main effect of time
271 in both women (PRE=0.048±0.001, POST=0.044±0.001, F(1,18)=6.97, p=0.01, 0.27 η_p^2) and men
272 (PRE=0.048±0.001, POST=0.044±0.001 – F (1,23)= 5.28, 0.031, 0.18 η_p^2). Additionally, in men
273 we found a main effect of condition (SI=0.0483±0.001, CTRL=0.0488±0.0007, F(1,23)=4.30,
274 p=0.049, 0.15 η_p^2), and a significant two way (time by transition) interaction (F(1,437)=1.67,
275 p=0.037, 0.06 η_p^2 , A→C PRE=0.051±0.01, POST=0.056±0.01, p=0.02, C→E PRE=0.069±0.03,
276 POST=0.037±0.01, p=0.01). Finally, in the men group we also found a three-way interaction, for
277 time by condition by microstate transition (F(19, 437)=1.62, p=0.04, 0.06 η_p^2 , **Fig. 5c,d**).

278

279 **Spontaneous thoughts and microstate SI modulation**

280 To investigate how the observed SI-induced changes in microstates might reflect changes in the
281 pattern of spontaneous thought, we conducted a partial least squares (PLS) analysis, a multivariate
282 analysis that finds latent variables (LVs) which optimally link two sets of data by maximizing the
283 covariance between the two. We found a significant LV (permuted p=0.012, 52% of covariance
284 explained) for microstate *duration* (**Fig. 6a,b**). Changes in C and B microstate *duration* were
285 positively correlated with thoughts about future plans (*planning*), and negatively associated with
286 SI modulations of thoughts about *self* and thoughts about others. Microstates D and E showed the
287 reversed patterns of associations (**Fig. 6c-e**).

288 Results on *occurrence* modulations association with ARSQ change scores showed a significant
289 LV (p=0.004, 54% explained covariance). Occurrence of microstates A and B was positively

290 associated with *planning* and *verbal* thoughts. On the contrary, *occurrence* of microstate E was
291 negatively associated with these aspects of spontaneous thoughts (**Fig. S1**).

292 We performed a separate PLS analysis using the SI dynamics of transition probabilities and SI
293 dynamics of ARSQ scores in order to explore possible associations (**Fig. 6f-i**). We found a
294 significant LV ($p=0.0009$, 53% explained covariance). For most transitions to A, B, C or D
295 microstates we found a negative association with *discontinuity of mind*, *others*, *self*, *sleepiness*,
296 *comfort* and a positive association with *planning*, *health* and *verbal* thoughts. From most
297 microstates to E microstate, and for the E to D transitions, we found the reversed pattern of
298 associations (**Fig 6j**). There were no associations found with *somatic awareness* and with *visual*
299 thoughts.

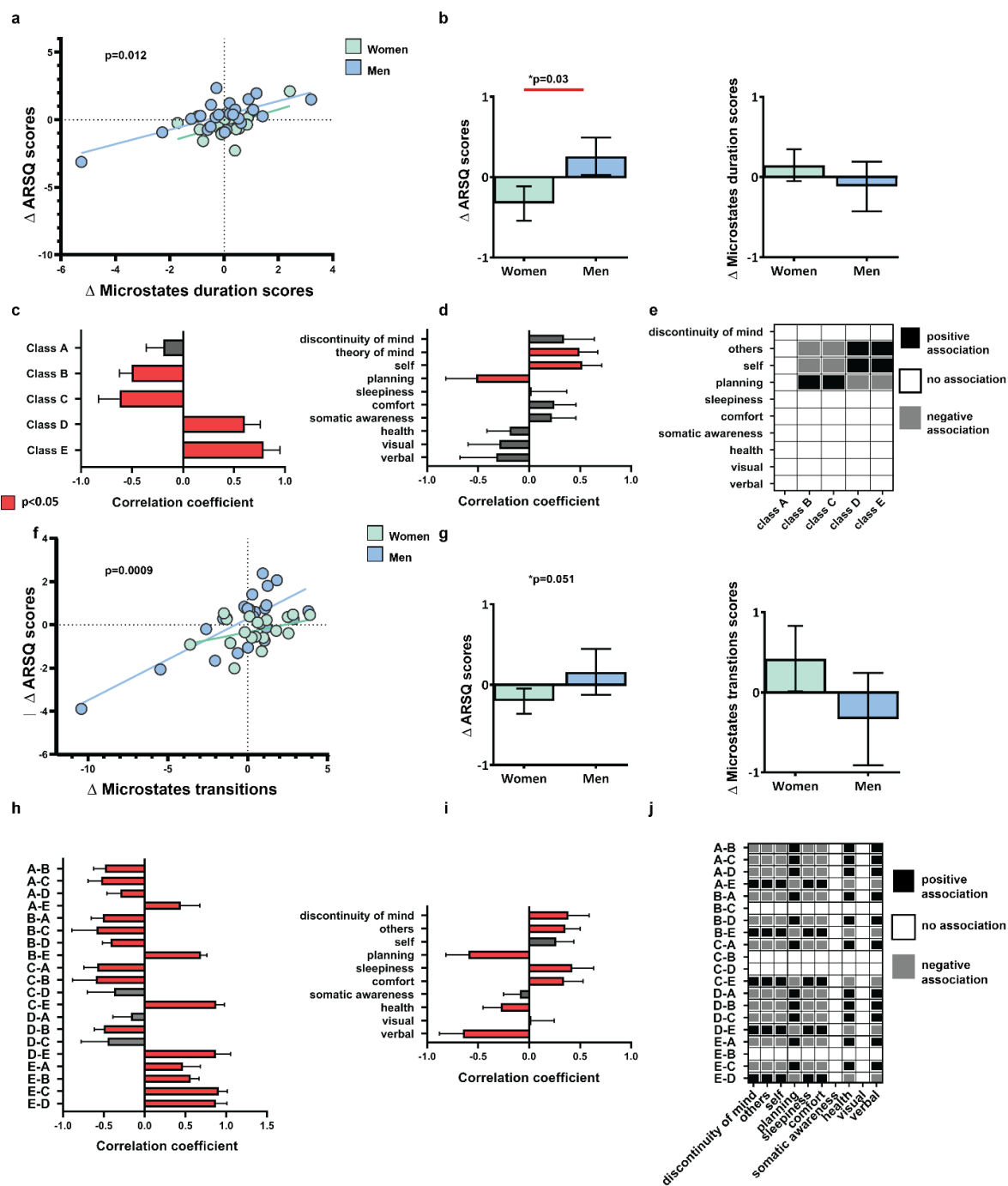
300 These data indicate that the *duration* and *occurrence* of certain microstates could predict aspects
301 of spontaneous thought content, whereas microstate *transition probability* could, in addition to
302 content, also predict spontaneous thought dynamics, modality, and association with physiological
303 states.

304

305 **Associations between microstate dynamics and salivary OXT following SI**

306 To investigate if salivary OXT might point to a potential mechanism for how SI modulates
307 microstate temporal features and transition probability, we tested if the dynamics of salivary
308 oxytocin (Δ OXT) correlate with the dynamics of microstates duration, occurrence and transition
309 probability (**Fig. 7**). There was no significant correlation between changes in OXT and either the
310 *duration* or the *occurrence* of microstate classes (**Fig. 7c**). Similarly, in the whole group analysis
311 there was no significant association between changes in OXT and transition probability (**Fig. 7b**).
312 However, in women we found that Δ OXT negatively correlated with $\Delta A \rightarrow C$ (Gamma $r=-0.47$,
313 $p=0.01$), and positively correlated with $\Delta B \rightarrow E$ (Gamma $r=0.45$, $p=0.02$). In men we found a
314 positive association of Δ OXT with $\Delta D \rightarrow B$ (Gamma $r=0.32$, $p=0.03$), as well as with $\Delta B \rightarrow D$
315 (Gamma $r=0.32$, $p=0.03$).

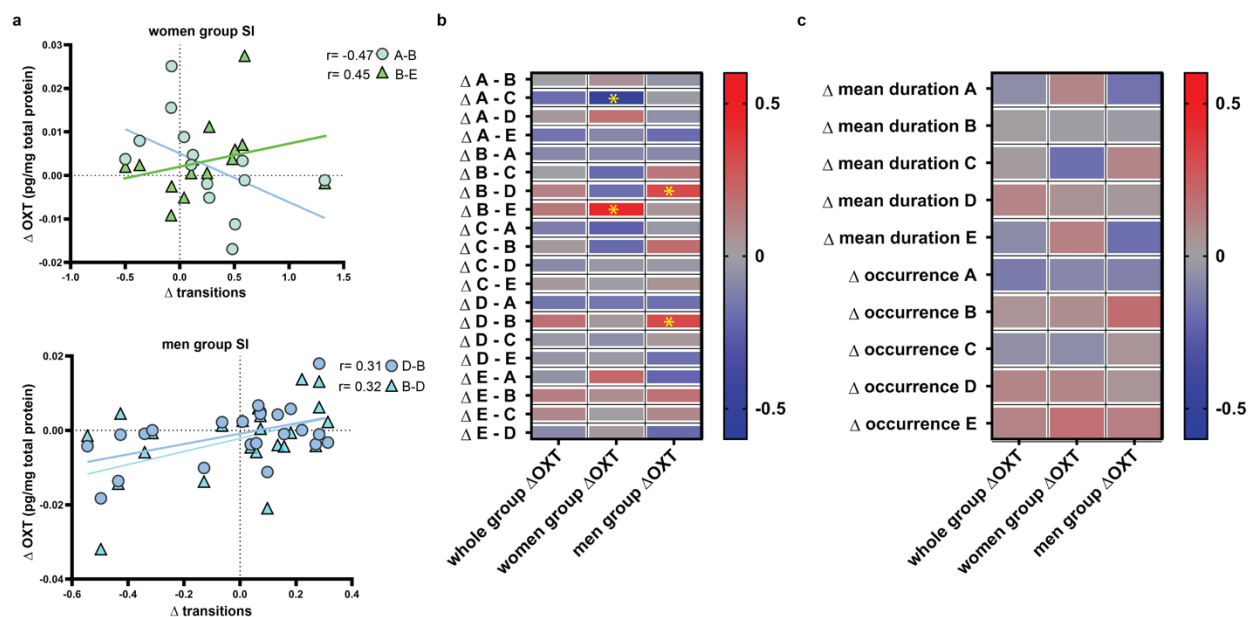
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317

Figure 6. Association between microstate dynamics and spontaneous thoughts.

(a) Correlation between individual-specific changes in ARSQ scores and changes in microstate duration. (b) Group differences in composite scores of ARSQ and microstate duration change. (c) Correlations between original and composite microstate duration scores (d) Correlations between original and composite ARSQ scores (e) Specific associations between microstate duration and spontaneous thoughts (f) Correlation between individual-specific changes in ARSQ scores and changes in microstate transitions. (g) Group differences in composite scores of ARSQ and microstate transitions change. (h) Correlations between original and composite microstate transition scores (i) Correlations between original and composite ARSQ scores (k) Specific associations between microstate transitions and spontaneous thoughts.



318

Figure 7. Association between changes in OXT and microstate temporal dynamics. (a) Example correlations between SI-specific changes in microstate transitions, and in salivary OXT (b) Summary of results for associations between changes in microstate transitions, and changes in salivary OXT following SI (c) Summary of results for associations between changes in microstate duration, occurrence, and changes in salivary oxytocin following SI.

319

320 Associations between personality traits and temporal dynamics of EEG microstates

321 Responses to SI depend in part on personality traits (Papasteri, Sofonea et al. 2020). Similarly,
 322 personality traits impact spontaneous thought patterns (Diaz et al., 2014). Here, we investigated if
 323 SI modulations of spontaneous thoughts vary as a function of personality traits as measured by the
 324 NEO PI-R inventory. We found a significant LV ($p=0.001$, explaining 63% of covariance),
 325 revealing associations between personality traits, and SI modulation of ARSQ scores (Fig. 2a-e).
 326 *Neuroticism*, certain aspects of *openness* and *modesty* traits were positively associated with
 327 increases following SI of thoughts about self, of verbal thoughts and awareness of own
 328 physiological state during resting (Fig. S2e). *Extraversion* and *conscientiousness* traits were
 329 positively correlated with SI modulation of thoughts about planning and of visual imagery (Fig.
 330 S2e). These findings suggest that SI could be more beneficial for those with higher scores on
 331 extraversion and conscientiousness.

332 If changes in neural activity captured by changes in microstate properties serve as mechanisms for
333 the effects of SI, we would expect to see associations between personality traits and microstate
334 dynamics. Therefore, we conducted PLS analyses testing for associations between personality
335 traits and microstates temporal dynamics. We found one LV (LV- $p=0.007$, 68% of covariance
336 explained) showing that temporal dynamics of microstates following SI varied as a function of
337 personality traits (**Fig. S3a**). Changes in microstates C, A, and B *duration* were positively related
338 to *extraversion* and *conscientiousness* traits. The inverse pattern of association was present for
339 microstate E. No significant LV was found explaining the *occurrence* changes as a function of
340 personality traits. In summary, the significant changes in microstates *duration* following SI were
341 more pronounced in those with higher scores in extraversion, conscientiousness and/or lower
342 neuroticism, modesty and openness scores.

343

344 **Discussion**

345 Imitation has been shown to have powerful effects on behavioural and emotional states in human
346 and non-human primates^{36,39}. So far, the search for a mechanism focused on changes in neural
347 activity induces during imitation³⁷. Here, we rely on previous findings linking mind-wandering to
348 emotional states, to investigate if imitation might also act by changing the pattern of spontaneous
349 thoughts. We find that imitation can indeed impact the dynamics, content and association with
350 physiological states of mind-wandering episodes (**Fig. 1**), potentially by changing the pattern of
351 correlated neuronal activity (**Fig. 4, 5**). Further investigation is needed to determine if reported
352 changes in spontaneous thoughts causally relate to changes in behavioural states and in salivary
353 oxytocin.

354 **Association between spontaneous thoughts and behavioural state**

355 We find that certain imitation-sensitive features of spontaneous thought dynamics and content
356 positively associate with improvements in stress, well-being and closeness to the imitated person
357 (**Fig. 1,2**). This finding adds the social implication to the reported effects of mind-wandering. Our
358 findings on stress and well-being seem to contradict prior research. Previous studies report that
359 negative mood and mind-wandering reinforce each other: spontaneous thoughts lead to sadness,

360 which then leads to more mind-wandering^{7,10,11}. The apparent differences in our findings, namely
361 the fact that mind-wandering can lead to improved behavioural states, could result from probing
362 stress levels instead of affect valence, or from focusing on mind-wandering episodes around a
363 generally agreeable task. The later possibility would support the notion that contexts have a
364 substantial impact on spontaneous thoughts and on their ability to impact behaviour. The reverse
365 sequence could also be true, where contexts impact behavioural states, which in turn modulate
366 spontaneous thoughts. In this manuscript we do not establish a causal link, and further
367 investigations will be needed to establish if such a link exists.

368 **Association with changes in OXT**

369 The hormone OXT, known to play an important role in social bonding when acting centrally⁴⁰, has
370 been shown to increase following social imitation^{36,38}. However, in previous work, we showed that
371 while salivary OXT can predict well-being following SI, it does not associate with increased
372 closeness for imitated partner³⁶. This is consistent with our current findings where we report a
373 negative association between salivary OXT and *discontinuity of mind*, a feature of mind-wandering
374 that positively correlates with stress, and that decreases following imitation. One possibility is that
375 increased OXT after SI acts on the dynamics of spontaneous thoughts to decreased stress levels.
376 Also consistent with our previous report³⁶, salivary OXT negatively associate with thoughts about
377 planning, a correlate of increased social closeness following imitation. All results regarding
378 salivary OXT should be interpreted with caution, as the levels of OXT in any peripheral fluid do
379 not accurately reflect central actions of OXT^{41,42}. Although the relationship between oxytocin and
380 spontaneous thoughts has not been reported before, other hormones have been linked to mind-
381 wandering. Changes in the level of the stress-related hormones cortisol and alpha-amylase have
382 been associated with negative spontaneous thoughts⁴³.

383 **Association with microstate dynamics**

384 Our results show that changes in neural activity captured by microstate analysis could potentially
385 explain changes in spontaneous thoughts and well-being following imitation. We find that the
386 *duration* of microstate C, generally linked to activity in the DMN, increased after imitation (**Fig.**
387 **4**), and is positively associated with thoughts about planning, which also increase after imitation,
388 and positively correlate with social closeness and well-being. Although we also observed changes

389 in microstate D *occurrence* after imitation, they seemed to have less predictive potential for
390 changes in spontaneous thought pattern and in behavioural states. However, the observed SI
391 modulation of both microstate C and D are in line with previous findings on changes during
392 relaxing behavioural states^{32,33,34,35}. Changes in microstate transition probability following
393 imitation were associated not only with changes in the content of spontaneous thought but also in
394 their dynamics (**Fig. 6**), and in most cases were consistent with the effects of imitation on
395 spontaneous thoughts. Additionally, we found a distinct pattern of association between SI
396 modulation on microstate transitions and OXT (**Fig. 7**) in men compared to women. This supports
397 previous data on the sexually dimorphic action of OXT, and on the role of microstate dynamics
398 underlying different patterns of information processing^{44,45}. In agreement with previous finding
399 that salivary OXT increases in women but not men following SI³⁶, we further show that only in
400 women the change in OXT is associated with slower transitions from microstate A (generally
401 linked to activity in language/auditory cortices) to microstate C (associated to posterior DMN
402 activations)^{26,31}.

403 Could ongoing, stimulus- and task-independent brain activity impact spontaneous thoughts and
404 possibly subsequent behavioural states? The impact of ongoing neuronal activity on task-related
405 performances has been studied extensively in both animal models and humans^{46,47}, yet the
406 association between ongoing activity patterns and spontaneous thoughts has been more difficult to
407 establish, due to the fleeting nature and unpredictability of spontaneous thoughts. In addition to
408 our current findings linking ongoing patterns of microstate activity with dynamics and content of
409 spontaneous thoughts, previous work showed that patients suffering from neuropsychiatric
410 disorders with abnormal mind-wandering patterns have significant differences in microstate
411 dynamics⁴⁸⁻⁵⁴.

412 Taken together, our findings highlight the effect of social imitation on spontaneous thoughts, and
413 the potential neural and biochemical substrates involved. Our findings should fuel research into
414 potential therapies using social context-dependent changes in spontaneous thoughts.

415

416

417 **Methods**

418 *Ethical considerations*

419 All methods and experiments have been approved by The Ethics Committee of National University
420 for Theatre and Film I.L. Caragiale Bucharest, and followed the guidelines of the Declaration of
421 Helsinki. All participants provided written informed consent for their participation. The instructor
422 and subject shown in **Figure 1a**, both gave their written consent to have their faces shown in the
423 manuscript.

424 *Experimental paradigm*

425 This study investigates social imitation (SI) task modulations of resting-state EEG, OXT and
426 behavioural states. Data from each participant was assessed across the different tasks (SI vs CTRL)
427 on two separate days in a counterbalanced, randomized fashion, with at least one week delay
428 between the two experimental days.

429 As depicted in Figure 1, during the SI task participants were asked to imitate in real time the
430 stereotyped, geometrical arm movements produced by an instructor (e.g. draw a circle in the air).
431 The SI task lasted for 3 minutes during which the instructor provided positive verbal cues (e.g.
432 “very well”) and positive non-verbal cues (e.g. smile, nodding yes). During CTRL task no social
433 interaction was available and participants were asked to imitate the geometrical movements
434 displayed on a screen (3 minutes).

435 *Participants and data collection*

436 Participants in the study were recruited through advertisement within the University of Bucharest,
437 University of Theatre and Film and the CINETic’s Research Centre website
438 <https://cinetic.arts.ro/en/met-2/>. The exclusion criteria included neurologic and psychiatric
439 symptoms. Of the 65 subjects that were initially recruited, several (N=7) were excluded from the
440 analysis due to movement contamination of the EEG data and several (N=15) dropped out of the
441 experiment before the collection of the second day of the experiment. The remaining dataset thus
442 included 43 participants (24 men, 19 women), mean age=25.7 (age range: 20-42), s.d.=5.1.
443 (Women mean age=25.57, s.d.=5.3; Men mean age=25.9, s.d.=5.1). The sample group included in

444 the correlation analyses between OXT or personality traits and EEG temporal dynamics was
445 composed of 37 participants (18 men and 19 women). Six participants were excluded from these
446 analyses due to missing OXT and NEO-PI-R data.

447 EEG data were acquired in a dimly light room using a 128-channel ANT Neuro waveguard™
448 system (<https://www.ant-neuro.com/>) sampled online at 1kHz with a Cz reference. Participants
449 sat in a comfortable, upright position and were instructed to stay awake, as calm as possible, to
450 keep their eyes closed and to relax for five minutes without falling asleep.

451 Salivary oxytocin (OXT) were collected using specially designed tubes (Salivette®, Sarstedt).
452 Participants were instructed to saturate the saliva swab by holding for at least 2 minutes in the
453 mouth. The tubes were then centrifuged at 1000 x g, at 4° C, for 25 minutes and the samples were
454 aliquoted in 1,5 ml Eppendorf vials and stored at -80° C prior to analysis. Oxytocin was measured
455 by radioimmunoassay (RIA) at RIAGnosis, Munich, Germany, while total proteins were measured
456 at National Institute of Endocrinology “C. I. Parhon”, Bucharest, Romania. Salivary total protein
457 was used to normalize the concentration of salivary oxytocin levels, since its concentration can
458 vary significantly with saliva viscosity.

459 Self-reported data were collected in the same room as the EEG data and consisted of several pen-
460 paper tests. The inclusion of others in the self (IOS)⁵⁵ a single item measure of closeness composed
461 of two Venn-like circles varying in their degree of overlap. Participants were instructed to select
462 the diagram reflecting the relationship with the instructor that guided the participants throughout
463 the social interaction task. Two visual analog scales (VAS) assessed the self-reported level of stress
464 and well-being using a 10-cm unmarked scale ranging from 0 “no stress”/”worst unimaginable
465 well-being” up to 100 “most stress ever”/”perfect well-being”.

466 The Neo Personality Inventory –Revised (NEO PI-R) is a 240 item personality inventory with
467 cross-culturally established psychometric properties and validity⁵⁶ that assesses the Big Five
468 Model domains of Neuroticism, Extraversion, Openness to Experience, Agreeableness and
469 Conscientiousness with the 6 factors for each domain. Participants responded on a 5-point Likert
470 scale ranging from 0 (strongly disagree) to 4 (strongly agree).

471 The Amsterdam Resting-State Questionnaire 2.0 is a self-report questionnaire that quantifies mind
472 wandering along ten model: Discontinuity of Mind, Theory of Mind, Self, Planning, Sleepiness,
473 Comfort, Somatic Awareness, Health Concern, Visual Thought, Verbal Thought⁹. Participants had
474 to responded to a total of 30 items using a 5-point Likert Scale from “Completely Disagree” to
475 “Completely Agree”.

476 *EEG data processing*

477 The EEG datasets were band-pass filtered offline between 1 and 40 Hz with an additional notch at
478 50 Hz. EEG periods of movement contamination or other artifacts were marked and excluded from
479 the analyses. In order to remove the oculomotor artifacts such as saccades and eye blinks as well
480 as the cardiac artefacts (ECG), we applied the Infomax-based Independent Component
481 Analysis (ICA)⁵⁷. Bad or noisy electrodes were interpolated using a 3-D spherical spline⁵⁸, and
482 were recomputed to the common average-reference. The data were then down-sampled to 125 Hz
483 for further analysis.

484 The local maxima of the Global Field Power (GFP) show an optimal signal to noise-ratio in the
485 EEG^{58,59}. The EEG signal was extracted at the corresponding time frame of GFP peaks and only
486 the time points of GFP peaks were submitted to a modified k-means cluster analysis^{59,60} in order
487 to identify the most representative classes of stable topographies.

488 The k-means clustering was performed in two steps: first, at the individual level, and, in a second
489 step, at the group level by clustering all individual dominant topographies with varying number of
490 clusters. In order to determine the optimal number of clusters at the individual and the group level,
491 we used the criteria implemented in Cartool (a free academic software developed by Denis Brunet;
492 cartoolcommunity.unige.ch), based on seven maximally independent criteria: Davies and Bouldin,
493 Gamma, Silhouette, Dunn Robust, Point-Biserial, Krzanowski-Lai Index, and Cross-Validation
494 ^{31,59,61}.

495 In the first part of the microstate analysis, only GFP peaks were submitted to the k-means
496 clustering. However, in the second part of the analysis, during the fitting process of the microstates,
497 the entire EEG of participants was used, excluding only the marked artifact epochs. A temporal
498 smoothing (window half-size 3 (24 ms), Besag factor of 10 and a rejection of small time frames

499 (when <3 , i.e. 24 ms) was applied⁵⁹. Subsequently, in order to quantify the temporal parameters of
500 microstates, every time point of the individual data was assigned to the microstate cluster with
501 which it correlated best⁶⁰. A 0.5 correlation coefficient threshold was used in order to exclude
502 transient periods of noise in the data. These periods were not labelled and were excluded from the
503 analysis.

504 This fitting process enabled the determination of the duration and the occurrence of each microstate
505 in each subject. The *duration* represents the average amount of time (in ms) that a given microstate
506 map was present without interruptions, i.e. the duration during which the subject remained in a
507 certain state. The duration is one of the most commonly used parameters of the temporal structure
508 of microstates and has repeatedly been shown to be associated with different vigilance conditions
509 and symptoms of neuropsychiatric disorders⁶². The mean *occurrence* of a microstate is
510 independent of the duration. It indicates the rate at which a given microstate occurred, i.e. how
511 many times per second the brain enters a certain state.

512 In addition to these two temporal parameters for each microstate, we also analysed the between
513 microstate transition dynamics: for each subject and transition pair we computed the number of
514 transitions and normalized them by all between-class transitions^{45,63}.

515 The free academic software Cartool (cartoolcommunity.unige.ch) was used for the EEG data
516 processing and microstate analysis⁶¹.

517 *Statistical analyses*

518 In order to assess the specific modulations of spontaneous thoughts after the SI we utilized a three-
519 way repeated measures ANOVA with three factors: the ARSQ factors (10), time (PRE-POST),
520 and task (SI or CTRL). The averaged ARSQ scores across the different experimental conditions
521 (PRE-SI, POST-SI, PRE-CTRL, POST-CTRL) were averaged for each individual and correlation
522 analyses to investigate associations with behavioural states (stress, well-being and IOS) and
523 salivary OXT. We used Gamma non-parametric correlation analyses which can inform us on the
524 strength of their association disregarding possible outliers and tied ranks in the data.

525 Similarly, for both microstate temporal parameters (duration and occurrence) and transition
526 probabilities for each directional transition pair (20). Separate three-way repeated measures

527 ANOVA's with time (PRE- POST), classes of microstates (5)/transition pairs (20), and task (SI or
528 CTRL) as within factors was performed. We performed non-parametrical Wilcoxon paired test to
529 further assess main drivers of significant main effects. Additionally, to assess the degree of gender
530 differences of these effects we performed separate statistical analyses on men and women
531 subgroups.

532 In order to identify robust patterns of correlations between the behavioural measures and the
533 temporal dynamics of the EEG microstates, we used a multivariate approach called partial least
534 squares (PLS)⁶⁴. PLS is a multivariate data-driven statistical technique that maximizes the
535 covariance between two matrices by deriving *latent variables* (LVs), which are optimal linear
536 combinations of the original matrices⁶⁵. PLS is a powerful technique for relating two sets of data
537 (e.g., neuroimaging and behavioral data), even if these data show autocorrelation or
538 multicollinearity⁶⁴.

539 Each LV is characterized by a distinct EEG microstate pattern (called EEG *loadings*) and a distinct
540 behavioral profile (called behavioral *loadings*). By linearly projecting the EEG and behavioral
541 measures of each participant onto their respective loadings, we obtained individual-specific EEG
542 microstates and behavioral *composite scores* for each LV. PLS seeks to find loadings that
543 maximize across-participant covariance between the EEG microstates parameters and behavioral
544 composite scores. The number of significant LVs was determined by a permutation test (1000
545 permutations). The p-values (from the permutation) for the first five LVs were corrected for
546 multiple comparisons using a FDR of $q < 0.05$. To interpret the LVs, we computed Pearson's
547 correlations between the original EEG data and EEG composite scores, as well as between the
548 original behavioral measures and behavioral composite scores for each LV. A large positive (or
549 negative) correlation for a particular behavioral measure for a given LV indicates greater
550 importance of the behavioral measure for the LV. Similarly, a large positive (or negative)
551 correlation for a particular EEG microstate parameter for a given LV indicates greater importance
552 of the EEG microstate parameter for the LV. To estimate confidence intervals for these
553 correlations, we applied a bootstrapping procedure that generated 500 samples from subjects' data.
554 Z-scores were calculated by dividing each correlation coefficient by its bootstrap-estimated
555 standard deviation. The z-scores were converted to p-values and FDR-corrected ($p < 0.05$)⁶⁶.

556

557

558 **AUTHOR CONTRIBUTIONS**

559 All authors contributed to the design of the experiment and interpretation of results. MIT (first
560 author) analyzed all data, in collaboration with CP and VK. MIT, AS, RB conducted the
561 experiments and data collection. MIT and IC wrote the manuscript, with feedback from all authors.

562

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570

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576

577 **CONFLICT OF INTEREST**

578 The authors declare that the research was conducted in the absence of any commercial or financial
579 relationships that could be construed as a potential conflict of interest.

580

581

582

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