

The Big Five personality traits and CNS arousal in the resting state

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Abstract

Based on Eysenck's pioneering work, CNS arousal has long been considered an encouraging biological candidate that may explain individual differences in human personality. Yet, results from empirical studies remained inconclusive. Notably, the vast majority of published results have been derived from small samples, and EEG alpha power has usually served as exclusive indicator for CNS arousal. In this study, we selected $N = 468$ individuals of the LIFE-Adult cohort and investigated the associations between the Big Five personality traits and CNS arousal by using the low-resolution electromagnetic tomography-based analysis tool VIGALL. Our analyses revealed that subjects who reported higher levels of extraversion and openness to experience, respectively, exhibited lower levels of CNS arousal in the resting state. Bayesian and frequentist analysis results were especially convincing for openness to experience. Among the lower-order personality traits, we obtained strongest evidence for neuroticism facet 'impulsivity' and reduced CNS arousal. We regard these findings as well in line with the postulations of Eysenck and Zuckerman and consistent with the assumptions of the 'arousal regulation model'. Our results also agree with meta-analytically derived effect sizes in the field of individual differences research, highlighting the need for large studies with at least several hundreds of subjects.

Keywords: Arousal, Big Five, EEG, Resting State, VIGALL, Extraversion, Neuroticism

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1 Introduction

2 Over the past decades, a substantial body of research has focused on the relationship between individual
3 differences in human personality and the underlying biological mechanisms. Aside from a general interest
4 to identify the biological factors that explain the great diversity in human behavior, research in this field
5 has been motivated by theoretical concepts and empirical evidence linking personality traits to mental
6 health outcomes (Maher & Maher, 1994; Strickhouser, Zell, & Krizan, 2017). Beyond this, personality
7 traits have been proposed to constitute vulnerability factors for mental diseases, and affective disorders
8 in particular (Akiskal, Hirschfeld, & Yerevanian, 1983; Barnett et al., 2011; Hensch et al., 2019;
9 Jeronimus, Kotov, Riese, & Ormel, 2016; Klein, Kotov, & Bufferd, 2011). On this account, elucidating
10 the biological basis of personality has not only been argued to provide valuable insights into the etiology
11 of psychiatric diseases, but may also have important implications for identifying at-risk individuals,
12 initiating early preventions, and tailoring treatments.

13 One of the most prominent trait approaches to describe and measure the structure of human
14 personality is the Five-Factor Model (FFM; Goldberg, 1990; McCrae & Costa, 2008). The FFM is a
15 taxonomy that strives for an economic description of the whole range of individual differences in
16 personality by means of five overarching factors. These ‘Big Five’ personality traits encompass openness
17 to experience, conscientiousness, extraversion, agreeableness, and neuroticism. With some limitations,
18 the five-factor structure of personality has been shown to generalize across languages and cultures, and
19 has been argued to be based on innate biological factors (Macdonald, 1998; McCrae et al., 2000; see also
20 De Raad, 1998). In fact, evidence from twin studies and genome-wide complex trait analyses suggests
21 that a substantial proportion of the Big Five variance is accounted for by genetics (Bouchard & McGue,
22 2003; Lo et al., 2017; Vernon, Martin, Schermer, & Mackie, 2008). However, the biological mechanisms
23 that bridge the effects of genetic variation on human personality still remain elusive. In order to provide
24 an explanatory biological basis of human personality, various neuropsychological trait theories have been
25 postulated, with Eysenck’s Arousal-Activation Theory of Extraversion and Neuroticism having attracted
26 particular attention (Brocke & Battmann, 1992; Eysenck, 1967).

27 Eysenck’s Arousal-Activation Theory builds upon the early 1960s’ psychophysiological activation
28 theories, according to which the ascending reticular activation system (ARAS) regulates central nervous
29 system (CNS) arousal (Duffy, 1962; Malmö, 1959). Eysenck distinguishes two components of his
30 conceptual nervous system: the reticulo-cortical brain system (i.e., ARAS) and the reticulo-limbic
31 visceral brain system (VBS; Matthews & Gilliland, 1999). Excitation of the ARAS by incoming stimuli
32 is referred to as ‘arousal’, whereas the excitation of the VBS by emotional stimuli is referred to as
33 ‘activation’. An increase in activation has arousing effects, while arousal may also occur without
34 activation (i.e., a unidirectional relationship). Eysenck postulated that extraverted individuals possess,
35 on average, relatively low habitual levels of CNS arousal in the resting state, which he traces back to a

36 higher ARAS activation threshold (Brocke & Battmann, 1992). As a compensatory mechanism, they
37 engage in arousal-enhancing behavior by seeking human interactions as well as novelty, change and
38 excitement. In comparison, Eysenck describes neurotic individuals as emotionally hypersensitive, which
39 he attributes to a lower activation threshold of the VBS (Brocke & Battmann, 1992). According to
40 Eysenck, individuals with high levels of neuroticism are more susceptible towards stress and show a
41 prolonged autonomic stress response.

42 The Arousal-Activation Theory has served as theoretical framework in numerous empirical studies
43 (Küssner, 2017; Matthews & Gilliland, 1999). Eysenck himself referred to the alpha range of the human
44 Electroencephalogram (EEG) as the standard measure of CNS arousal (Matthews & Gilliland, 1999). In
45 line with Eysenck's postulations, a number of studies demonstrated higher resting-state EEG alpha
46 power (indicating lower CNS arousal) in extraverted relative to introverted individuals (Gale, Coles, &
47 Blaydon, 1969; Gale, Edwards, Morris, Moore, & Forrester, 2001; Hagemann et al., 2009; Smith et al.,
48 1995). Several other studies failed to provide supportive evidence (Beauducel, Brocke, & Leue, 2006;
49 Hagemann et al., 1999; Matthews & Amelang, 1993; Schmidtke & Heller, 2004). In addition, some
50 investigators used EEG beta power as CNS arousal indicator and revealed both supporting and opposing
51 evidence (Gale et al., 1969; Gram, Dunn, & Ellis, 2005; Matthews & Amelang, 1993). In sum, empirical
52 investigations addressing the link between extraversion and CNS arousal have provided only inconsistent
53 evidence for Eysenck's postulations.

54 A few studies also reported on the relationship between neuroticism and arousal. Based on
55 Eysenck's postulations, researchers have argued that the habitual level of arousal may tend to be higher
56 in labile (N-) extraverts and introverts when compared to their stable (N+) counterparts (Brocke, Netter,
57 & Hennig, 2004). Consistent with this assumption, investigations in laboratory settings – including
58 resting-state assessments – have previously been shown to elicit an arousal-enhancing 'first day in lab
59 effect' (Huang et al., 2015) similar to the 'first night' effect in sleep medicine (Hirscher et al., 2015). This
60 may especially affect individuals with high levels of neuroticism, who have been proposed to be more
61 vulnerable towards stress. Notably, enhanced arousal levels in neurotic individuals would also tie in with
62 the substantial genetic overlap demonstrated between neuroticism and major depression (Baselmans et
63 al., 2019; Lo et al., 2017), with the latter having repeatedly been linked to enhanced and 'hyperstable'
64 CNS arousal levels in the resting state (Hegerl, Wilk, Olbrich, Schoenknecht, & Sander, 2012; Sander,
65 Schmidt, Mergl, Schmidt, & Hegerl, 2018; Schmidt et al., 2016, 2017; Ulke et al., 2017; Ulke, Tenke, et
66 al., 2019; Ulke, Wittekind, et al., 2019). Despite these converging lines of research, available EEG studies
67 have not yet provided supportive evidence for an association between CNS arousal and neuroticism (Gale
68 et al., 2001; Hagemann et al., 2009, 1999; Savage, 1964). It should be noted, though, that the vast
69 majority of published results on both neuroticism and extraversion have been derived from small samples
70 with fewer than 100 subjects.

71 In comparison to previous approaches that predominantly used EEG alpha power as exclusive
72 indicator for CNS arousal, above-mentioned studies that demonstrated a link between CNS arousal and
73 depression made use of the Vigilance Algorithm Leipzig (VIGALL), an EEG- and EOG-based analysis
74 tool that utilizes low-resolution electrotopography (Sander, Hensch, Wittekind, Böttger, & Hegerl,
75 2016). VIGALL is typically applied to fifteen to twenty-minute resting-state recordings and incorporates
76 information on the cortical distribution of the frequency bands alpha, delta, and theta. Beyond this,
77 VIGALL features adaptive procedures that account for the individual differences in alpha peak frequency
78 and EEG total power. Primarily, VIGALL was developed for investigating arousal disturbances in
79 psychiatric samples and to objectively test the assumptions of the ‘arousal regulation model of affective
80 disorders and attention-deficit hyperactivity disorder’ (Hegerl & Hensch, 2014). Similar to Eysenck’s
81 theory, the arousal regulation model postulates that depressive- and manic-like behavior partly reflects
82 an autoregulatory attempt to reduce and enhance habitual high and low arousal levels, respectively. A
83 particular emphasis is put on the regulation of arousal, which is postulated to be unstable in clinical
84 syndromes such as ADHD and mania and is expressed, at the behavioral level, in hyperactivity and
85 sensation seeking (similar to the behavior frequently observed in overtired children). Major depression,
86 in contrast, is postulated to be characterized by enhanced and hyperstable arousal, which is behaviorally
87 expressed in avoidance of additional external stimulation. Noteworthy, by applying VIGALL, a number
88 of empirical studies addressing arousal in depressive, bipolar, and ADHD patients have provided
89 supportive evidence for the assumptions of the arousal regulation model (Hegerl et al., 2012; Strauß et
90 al., 2018; Ulke et al., 2017; Ulke, Wittekind, et al., 2019; Wittekind et al., 2016). In addition, VIGALL
91 has been validated in an fMRI and PET study (Guenther et al., 2011; Olbrich et al., 2009), against
92 evoked potentials and parameters of the autonomous nervous system (Huang et al., 2017, 2018; Olbrich
93 et al., 2011), against the Multiple Sleep Latency Test (Olbrich et al., 2015), and in a large study
94 addressing the agreement with subjective ratings (Jawinski et al., 2017). These previous encouraging
95 results raise the question, whether the application of VIGALL may leverage investigations on the role
96 of arousal in human personality.

97 Against this background, we here sought to examine the relationship between the Big Five
98 personality traits and CNS arousal in the resting state by making use of the EEG- and EOG-based
99 analysis tool VIGALL. In accordance with previous theoretical and empirical indications, we
100 hypothesized that CNS arousal is negatively associated with the personality trait extraversion and
101 positively associated with neuroticism. Notably, each Big Five personality trait has been demonstrated
102 to genetically overlap with psychiatric disorders (Lo et al., 2017), and each of the respective psychiatric
103 disorders has been proposed to possess arousal-related pathophysologies (Hegerl & Hensch, 2014). On
104 this account, we here examined the potential associations between CNS arousal and each Big Five
105 personality trait. Given the relatively weak effect sizes in personality and individual differences research
106 (Gignac & Szodorai, 2016; Schäfer & Schwarz, 2019), we considered a sample of several hundreds of

107 participants to derive our estimates. In this vein, we sought to contribute empirical evidence to the so-
108 far unresolved issue of whether basic personality dimensions are reflected in habitual levels of arousal.

109 **Methods and Materials**

110 In the following sections, we report how we determined our sample size, all data exclusions (if any), all
111 manipulations, and all measures in the study (Simmons, Nelson, & Simonsohn, 2012). All analysis scripts
112 have been made publicly available on the repository of the Open Science Framework
113 (<https://doi.org/10.17605/osf.io/ud38w>). The original data will be accessible via the Leipzig Health
114 Atlas (<https://www.health-atlas.de>) upon publication of the peer-reviewed article.

115 **Sample**

116 Participants were drawn from the LIFE-Adult study, a population-based cohort study of 10,000
117 inhabitants of the city of Leipzig, Germany (Loeffler et al., 2015). The scope of LIFE-Adult is to examine
118 prevalences, genetic predispositions, and lifestyle factors of civilization diseases. All subjects underwent
119 a comprehensive medical assessment program and completed various psychological surveys. We
120 considered subjects with available resting-state EEG and NEO Personality Inventory data (562 subjects
121 aged 40-79 years). Of these, we selected subjects who reported no current intake of EEG-affecting drugs
122 and had no prior diagnosis of stroke, multiple sclerosis, Parkinson's disease, epilepsy, skull fracture,
123 cerebral tumor, or meningitis (leaving 533 subjects). Based on a structured clinical interview for DSM-
124 IV axis I disorders, we selected subjects without a history of psychotic disorders or substance dependence,
125 and who were free of current anxiety and affective disorders (leaving 528 subjects). Moreover, EEGs with
126 substantial artifacts ($\geq 15\%$ of all EEG segments) and those showing low-voltage alpha, alpha variant
127 rhythms, or pathological activity were not included. This resulted in $N = 468$ eligible subjects (246
128 females; mean age: 58.5 years). Participants gave written informed consent and received an expense
129 allowance. All procedures were conducted according to the Declaration of Helsinki and were approved
130 by the Ethics Committee of the University of Leipzig (263-2009-14122009).

131 **Questionnaire**

132 Subjects completed the German version of the revised NEO Personality Inventory (NEO-PI-R; Costa &
133 McCrae, 1992; Ostendorf & Angleitner, 2004). The NEO-PI-R is a widely used self-report questionnaire
134 that enables measuring the personality traits neuroticism, extraversion, openness to experience,
135 agreeableness, and conscientiousness. The NEO-PI-R consists of 240 items and ratings are made on a
136 five-point scale ranging from 'strongly disagree' to 'strongly agree'. Item scores are aggregated to the
137 five NEO personality dimensions. The internal consistency (Cronbach's alpha) of the five overarching
138 factors has been reported to range from 0.87 to 0.92 (Ostendorf & Angleitner, 2004). Test-retest
139 reliability (1-month interval) has been reported to range from 0.88 to 0.91. Further, the NEO-PI-R allows
140 to calculate scores for thirty personality facets, six facets per factor. The internal consistency and test-

141 retest reliability of the facets has been reported to range from 0.53 to 0.85 and 0.48 to 0.90, respectively
142 (Ostendorf & Angleitner, 2004). NEO personality dimension and facet scores were transformed into sex-
143 and age-normalized T-scores according to the NEO-PI-R manual.

144 **Physiological data collection and processing**

145 Physiological data collection and processing was carried out as previously described (Jawinski et al.,
146 2019, 2015). EEG assessments were conducted according to a standardized operating procedure.
147 Assessments took place at three time slots: 8:30 am, 11:00 am, and 1:30 pm. During the twenty-minute
148 resting condition, subjects lay on a lounge chair within a light-dimmed sound-attenuated booth. Subjects
149 were instructed to close their eyes, relax and not to fight any potential drowsiness. In order to achieve
150 similar initial levels of arousal activation, all subjects completed a brief arithmetic task immediately
151 before the onset of recording. EEGs were derived from 31 electrode positions according to the extended
152 international 10-20 system. Two bipolar electrodes served to record vertical and horizontal eye
153 movements (EOGs). EEGs were recorded against common average reference with AFz ground. We used
154 a QuickAmp amplifier (Brain Products GmbH, Gilching, Germany) and sampled recordings at 1000 Hz.
155 EEG offline processing was carried out using Brain Vision Analyzer 2.0 (Brain Products GmbH, Gilching,
156 Germany). EEGs were filtered (70 Hz low-pass and 0.5 Hz high-pass with 48 dB/Oct slope, 50 Hz notch)
157 and rectified from eye movement, sweating, cardiac, and muscle artifacts using Independent Component
158 Analysis (ICA). Graph elements (sleep spindles and K-complexes) were manually marked by experienced
159 raters as previously described (Jawinski et al., 2017). Please see the publicly available VIGALL 2.1
160 manual for further preprocessing details (Hegerl et al., 2016).

161 **Assessment of brain arousal**

162 The assessment of brain arousal was carried out as described elsewhere (Jawinski et al., 2019, 2017).
163 EEG-vigilance served as indicator for brain arousal and was measured using the Brain Vision Analyzer
164 add-on VIGALL 2.1 ([https://www.deutsche-depressionshilfe.de/forschungszentrum/aktuellestudien/vig](https://www.deutsche-depressionshilfe.de/forschungszentrum/aktuellestudien/vig-all-vigilance-algorithm-leipzig-2-1)
165 [all-vigilance-algorithm-leipzig-2-1](https://www.deutsche-depressionshilfe.de/forschungszentrum/aktuellestudien/vig-all-vigilance-algorithm-leipzig-2-1); Hegerl et al., 2016). Based on the cortical distribution and spectral
166 composition of EEG activity, VIGALL assigns one of seven EEG-vigilance stages to each one-second
167 EEG segment. EEG-vigilance stages correspond to active wakefulness (stage 0), relaxed wakefulness
168 (stages A1, A2, A3), drowsiness (stages B1, B2/3), and sleep onset (stage C). Notably, stages A1-3 are
169 characterized by predominant alpha activity, which may indicate relatively enhanced CNS arousal during
170 eyes-closed resting-state conditions where stages of drowsiness (delta- and theta-activity) and sleep onset
171 (occurrence of K-complexes and sleep spindles) are frequently observed. Therefore, the range of arousal
172 stages implicated in the present study extends traditional approaches where higher EEG alpha power
173 (relaxed wakefulness) has been used as exclusive indicator for reduced CNS arousal. We transformed
174 assigned EEG-vigilance stages into values ranging from 7 (active wakefulness) to 1 (sleep onset) and
175 calculated three outcome variables: mean vigilance, stability score, and slope index. Variable ‘mean

176 vigilance’ provides an estimate for the average level of EEG-vigilance during rest. The variables ‘stability
177 score’ and ‘slope index’ particularly focus on the dynamics of EEG-vigilance. Lower scores indicate lower
178 average levels and steeper declines, respectively, of EEG-vigilance. All three outcome variables have been
179 validated, have been found test-retest reliable, and have previously been used as default parameters to
180 summarize complex EEG-vigilance time-courses (Huang et al., 2017, 2015; Jawinski et al., 2017;
181 Jawinski, Mauche, et al., 2016; Jawinski, Tegelkamp, et al., 2016).

182 **Statistical analyses**

183 The internal consistency of the NEO personality dimensions and facets was calculated using SPSS
184 Statistics 25.0 (IBM corp.; Armonk, New York, USA). All frequentist analyses were carried out using
185 Matlab R2018a (The MathWorks Inc., Natick, Massachusetts, USA). The nominal level of significance
186 was set at $p < .05$ (two-tailed). Further, p-values were adjusted by applying the False Discovery Rate
187 (FDR) procedure according to Benjamini and Hochberg (1995). Associations with $FDR < 0.05$ were
188 regarded as significant after multiple testing correction. In addition, we sought to derive evidence for
189 the alternate and null hypothesis, respectively, by calculating Bayes factors. Bayes factors reflect the
190 likelihood ratio between the alternate and null hypothesis (BF_{10}). Bayesian analyses were conducted
191 with a moderate symmetrical 1/3 beta prior width using package ‘BayesFactor’ (Morey & Rouder, 2018)
192 for R 4.0.1 (R Core Team, 2017).

193 First, we carried out Spearman correlations between the higher-order NEO personality dimensions
194 (sex- and age-normalized T-scores) and the three EEG-vigilance variables (mean vigilance, stability
195 score, and slope index). Next, we generated a permutation-based quantile-quantile plot (qq-plot) to
196 examine whether the distribution of observed p-values differs from a random p-value distribution under
197 the null hypothesis. On this account, for the set of 15 observed p-values (5 NEO personality dimensions
198 x 3 EEG-vigilance variables), one million sets of 15 expected p-values were derived from correlations
199 after data permutation. Original correlations within the domain of personality traits and the domain of
200 EEG-vigilance variables were preserved, whereas original correlations between these domains were
201 removed through random shuffling. Subsequently, in order to identify facets that particularly contribute
202 to the observed associations, we conducted exploratory Spearman correlations between the thirty NEO
203 personality facets (sex- and age-normalized T-scores) and the three EEG-vigilance variables. By analogy
204 to the higher-order ‘Big Five’ analyses, we also generated a permutation-based qq-plot for the NEO
205 personality facets. Analyses were repeated with sex, age, and daytime of EEG-assessment serving as
206 covariates.

207 **Statistical power**

208 Power analyses were conducted using R package *pwr* (version 1.3-0; Champely, 2020), with effect sizes
209 quantified as Spearman’s rho (r_S). Given $N = 468$ and $\alpha = .05$, power calculations revealed that
210 associations with true effect sizes of $r_S = 0.052$, $r_S = 0.091$, and $r_S = 0.129$ were identified with a chance

211 of 20%, 50% and 80% ($1-\beta$), respectively. After Bonferroni-correction ($\alpha = .0033$; resembling the most
 212 conservative case where the FDR procedure ends at the smallest observed p-value), power calculations
 213 revealed that associations with true effect sizes of $r_S = 0.097$, $r_S = 0.135$, and $r_S = 0.173$ were identified
 214 with a chance of 20%, 50% and 80% ($1-\beta$), respectively. Supplementary Figure S1 shows the probabilities
 215 of associations to reach the threshold of significance, given true effect sizes of up to $r_S = 0.4$.

216 Results

217 The descriptive statistics for the five higher-order NEO personality traits and the three EEG-vigilance
 218 variables are shown in Table 1.

Table 1 Descriptive statistics of NEO-PI-R scores and VIGALL 2.1 variables of EEG-vigilance

N = 468	Cronbach's α	Mean	SD	Q1	Q2	Q3	Min	Max	Skew	Kurt
NEO-PI-R										
Neuroticism	0.906	44.95	8.77	39.00	45.00	52.00	20.00	71.00	-0.23	-0.29
Extraversion	0.899	51.69	9.83	45.00	52.00	58.00	21.00	79.00	-0.02	0.18
Openness	0.868	46.85	8.45	41.00	46.00	51.00	20.00	72.00	0.32	0.32
Conscientiousness	0.836	53.02	8.87	47.00	52.00	59.00	26.00	80.00	0.10	0.09
Agreeableness	0.881	54.73	9.04	48.00	54.00	61.00	30.00	80.00	0.30	0.03
EEG-vigilance										
Mean vigilance	-	5.09	1.06	4.43	5.40	5.89	1.93	6.76	-0.90	-0.03
Stability score	-	9.20	4.19	6.00	9.00	13.00	1.00	14.00	-0.58	-0.94
Slope index	-	-1.52	0.92	-2.26	-1.38	-0.74	-4.26	0.73	-0.57	-0.58

SD: standard deviation, Q1: quartile 1, Q2: quartile 2 (median), Q3: quartile 3, Min: minimum observed value; Max: maximum observed value; Skew: skewness, Kurt: excess kurtosis

219

220 The internal consistency (Cronbach's alpha) of the five NEO personality dimensions ranged between
 221 0.84 and 0.91 and was thus comparable to previous reports (Ostendorf & Angleitner, 2004). The NEO
 222 personality dimensions were significantly intercorrelated (suppl. Table S1), with the strongest correlation
 223 observed between extraversion and openness to experience ($r_S = .477$, $p = 5E-28$). The internal
 224 consistency of the NEO personality facets ranged between 0.46 and 0.83 (Cronbach's alpha and
 225 intercorrelations shown in suppl. Figure S2). Intercorrelations between EEG-vigilance variables reached
 226 $r_S \geq 0.82$ (suppl. Table S2). Regarding covariates, we observed that younger participants and those who
 227 underwent the EEG assessment at later daytime exhibited a lower EEG-vigilance (e.g. mean vigilance;
 228 age: $r_S = .168$, $p = 3E-4$; daytime: $r_S = -.155$, $p = 8E-4$). Although we used sex- and age-normalized T-
 229 scores according to the NEO-PI-R manual, we observed some remaining associations between the NEO
 230 personality traits and both sex and age. Detailed association results between covariates and our outcome
 231 variables are shown in supplementary Table S3.

232 **Big Five personality traits and CNS arousal**

233 Spearman correlations between the five NEO personality dimensions (sex and age-normalized T-scores)

234 and the three EEG-vigilance variables are shown in Table 2.

Table 2 Spearman correlations between NEO personality dimensions (T-Scores) and EEG-vigilance variables

<i>N</i> = 468	Mean vigilance				Stability score				Slope index			
	rho	<i>p</i>	FDR	BF ₁₀	rho	<i>p</i>	FDR	BF ₁₀	rho	<i>p</i>	FDR	BF ₁₀
Neuroticism	-.063	.170	.365	0.27	-.030	.515	.766	0.13	-.002	.972	.972	0.11
Extraversion	-.104	.025*	.082	1.29	-.096	.038*	.095	0.91	-.137	.003*	.023**	8.35
Openness	-.102	.027*	.082	1.20	-.121	.009*	.044**	3.23	-.173	2E-4*	.002**	121.33
Agreeableness	-.027	.561	.766	0.13	-.012	.791	.913	0.11	-.032	.496	.766	0.14
Conscientiousness	-.023	.624	.780	0.12	-.005	.913	.972	0.11	-.039	.399	.748	0.15

FDR: False Discovery Rate according to Benjamini and Hochberg; BF₁₀ Bayes factor showing the likelihood ratio between the alternate and null hypothesis (1/3 beta prior width).

* *p* < .05 (two-sided nominal significance)

** FDR < .05 (p-value corrected for all tested associations using FDR method)

235

236 Analyses revealed six associations with nominal significance (*p* < .05). Of these, three remained

237 significant after multiple testing correction (FDR < .05). We observed EEG-vigilance to be inversely

238 associated with the degree of extraversion (slope index: $r_s = -.137$, $p = .003$, FDR = .023, BF₁₀ = 8.35)

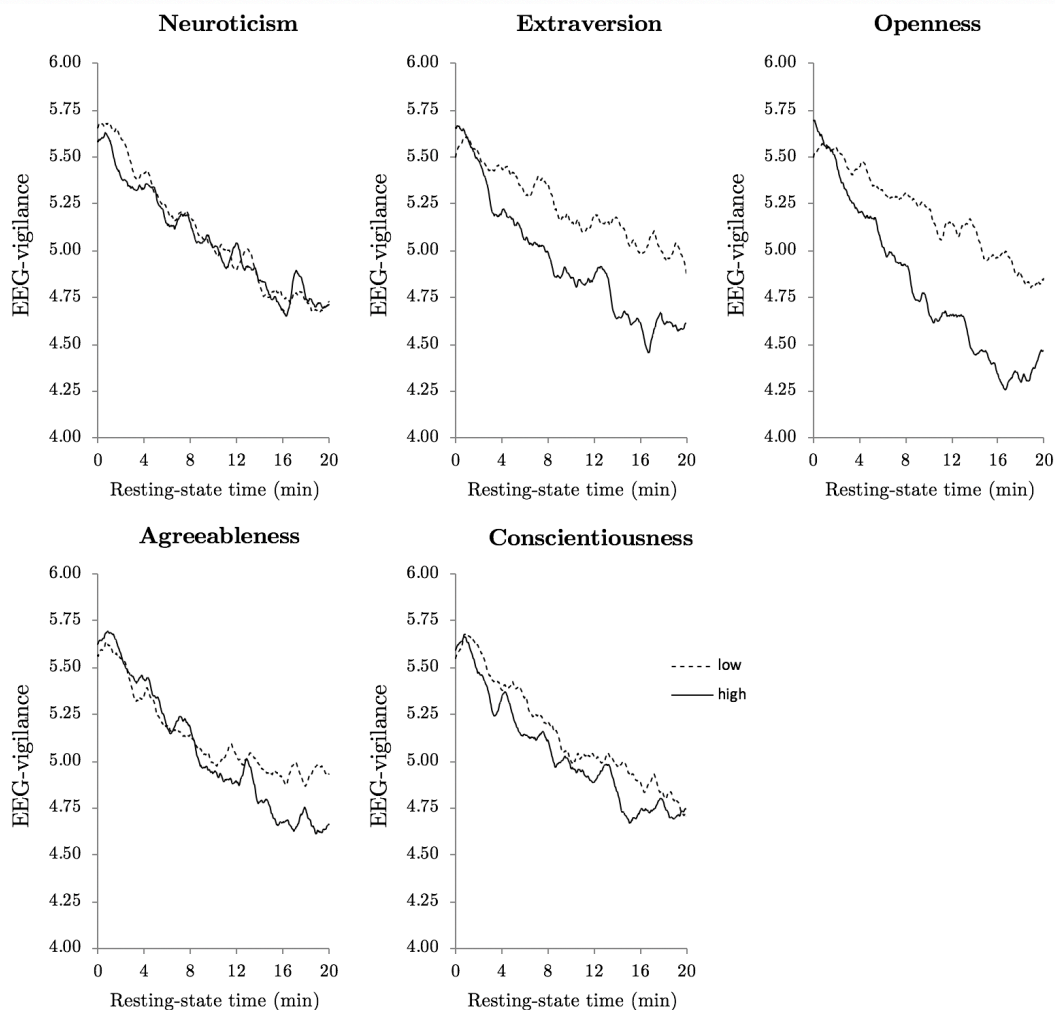
239 and openness to experience (stability score: $r_s = -.121$, $p = .009$, FDR = .044, BF₁₀ = 3.23; slope index:

240 $r_s = -.173$, $p = 2E-4$, FDR = .002, BF₁₀ = 121.33). Subjects who reported higher levels of extraversion

241 and openness to experience, respectively, exhibited lower EEG-vigilance. For illustrative purposes, the

242 time-courses of EEG-vigilance stratified by groups scoring low vs. high on the respective Big Five

243 dimension (lower vs. upper quartile of the ascending distribution) are shown in Figure 1.

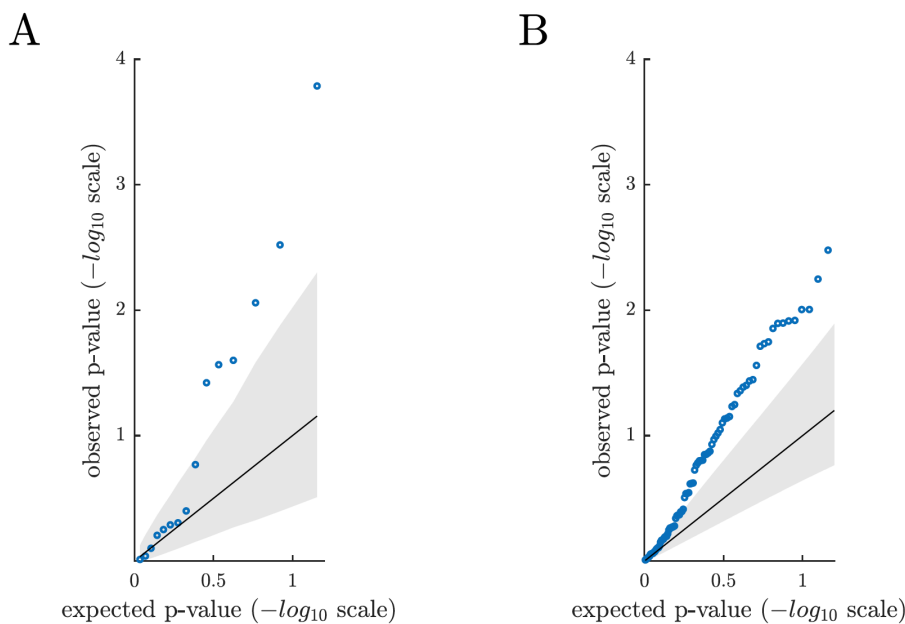


244

245 **Fig. 1** Time-courses of EEG-vigilance during the 20-minute eyes-closed resting-state condition stratified by groups
246 scoring low vs. high on the respective Big Five scale (i.e. subjects with scores in the lower vs. upper quartile of the
247 ascending distribution). Time-courses reflect simple moving averages (SMA), i.e., every data point represents an
248 averaged 61-second interval of EEG-vigilance (data point in time \pm 30 seconds). Statistical analyses revealed
249 significant correlations between EEG-vigilance and both extraversion and openness to experience.
250

251 We repeated our analysis by additionally adjusting correlations by sex, age, and daytime of EEG
252 assessment (suppl. Table S4). This resulted in three associations reaching nominal significance. Of these,
253 one remained significant after multiple testing correction (in this case the FDR corrected p-value is
254 equivalent to the Bonferroni-corrected p-value): Subjects who reported higher levels of openness to
255 experience exhibited lower EEG-vigilance (slope index: $r_S = -.152$, $p = .001$, $FDR = .015$, $BF_{10} = 23.40$).

256 In order to examine whether the distribution of observed p-values differs from a random p-value
257 distribution, we generated a permutation-based qq-plot that takes into account the dependencies between
258 association tests (Fig. 2A).



259

260 **Fig. 2** Permutation-based qq-plot showing the observed p-values from the association analyses (blue circles)
261 plotted against the expected p-values under the null hypothesis. The solid diagonal line represents the mean expected
262 p-values. The lower and upper bound of the grey area represent the 5th and 95th percentile (-log₁₀ scale) of the
263 expected p-values. **A** Results based on the sex- and age-normalized T-scores of the NEO personality traits. **B** Results
264 based on the sex- and age-normalized T-scores of the NEO personality facets. In total, qq-plots suggest that
265 association analyses revealed stronger evidence than expected by chance.
266

267 The qq-plot shows that the six strongest observed p-values exceed the 95th percentile (-log₁₀ scale) of
268 the computed expected p-value distribution. In detail, only 0.7% of the one million sets of expected p-
269 values contained at least six p-values below 0.038 (that is the 6th lowest observed p-value), and 0.2% of
270 the one million sets of expected p-values contained at least one p-value below 2E-4 (that is the lowest
271 observed p-value). Overall, the plot indicates that the distribution of observed p-values differs from a
272 random p-value distribution under the null hypothesis. When additionally adjusting association results
273 by sex, age, and daytime of EEG assessment, only the strongest observed association exceeded the 95th
274 percentile of the expected p-value distribution (suppl. Figure S3A). In this regard, only 1.1% of the one
275 million sets of expected p-values contained one p-value lower than 0.001 (that is the lowest observed p-
276 value).

277 **NEO personality facets and CNS arousal**

278 To further elaborate the nature of the underlying associations, we carried out exploratory Spearman
279 correlations between the 30 NEO facets (sex and age-normalized T-scores) and each of the 3 EEG-
280 vigilance variables. Detailed association results are shown in supplementary Table S5. In total, 24 out
281 of 90 correlations reached the level of nominal significance. The strongest association was observed for
282 neuroticism facet 'impulsiveness' (mean vigilance: $r_s = -.150$, $p = .001$, $BF_{10} = 19.88$). No other
283 neuroticism facet reached nominal significance. Regarding extraversion, we found nominally significant

284 results for the facets ‘warmth’ (slope index: $r_S = -.119$, $p = .010$, $BF_{10} = 2.90$), ‘assertiveness’ (slope
285 index: $r_S = -.109$, $p = .018$, $BF_{10} = 1.73$), ‘activity’ (slope index: $r_S = -.114$, $p = .014$, $BF_{10} = 2.14$), and
286 ‘positive emotions’ (slope index: $r_S = -.116$, $p = .012$, $BF_{10} = 2.43$). Regarding openness to experience,
287 we found significant associations for the facets ‘fantasy’ (slope index: $r_S = -.094$, $p = .041$, $BF_{10} = 0.85$),
288 ‘aesthetics’ (slope index: $r_S = -.137$, $p = .003$, $BF_{10} = 8.60$), ‘feelings’ (slope index: $r_S = -.137$, $p = .003$,
289 $BF_{10} = 8.27$), ‘actions’ (slope index: $r_S = -.109$, $p = .018$, $BF_{10} = 1.68$), and ‘ideas’ (slope index: $r_S = -$
290 $.128$, $p = .006$, $BF_{10} = 4.76$). We also observed nominally significant results for agreeableness facet
291 ‘tender-mindedness’ (slope index: $r_S = -.145$, $p = .002$, $BF_{10} = 14.40$) and conscientiousness facet
292 ‘achievement striving’ (slope index: $r_S = -.135$, $p = .003$, $BF_{10} = 7.65$).

293 By analogy to the NEO personality dimension analyses, we generated a permutation-based qq-
294 plot to examine whether the distribution of observed p-values of the NEO personality facets differs from
295 a random p-value distribution (Fig. 2B). Again, the qq-plot indicates that association analyses revealed
296 stronger evidence than expected by chance. Notably, consistent with the Big Five results, we observed
297 an attenuation of effect sizes when additionally adjusting sex- and age-normalized T-Score correlations
298 by sex, age and daytime of EEG-assessment (suppl. Table S6). The distribution of observed vs. expected
299 p-values after adjusting T-Scores is shown in supplementary Figure S3B, with a large proportion still
300 exceeding the 95th percentile of expected p-values.

301 Discussion

302 In this study, we investigated the association between the Big Five personality traits and CNS arousal
303 in the resting-state by making use of the EEG- and EOG-based analysis tool VIGALL. Our primary
304 analysis suggests that, after multiple testing correction, CNS arousal is negatively associated with the
305 degree of extraversion and openness to experience: Subjects who reported higher levels of extraversion
306 and openness to experience, respectively, showed steeper declines of EEG-vigilance. In addition, when
307 considering all tested associations between the Big Five personality traits and CNS arousal, we observed
308 overall stronger effects than expected by chance. This finding was supported by association results of
309 the thirty NEO personality facets. Facet analyses also revealed that the observed associations of the
310 higher-order Big Five traits and CNS arousal were not driven by a single facet with a distinct, strong
311 effect but rather appeared to arise from a distributed pattern of associations across several facets.
312 Notably, for the majority of nominally significant associations ($p < 0.05$), Bayesian analysis revealed
313 only anecdotal evidence for the alternate hypothesis (BF_{10} ranging between 1 and 3). In addition, when
314 taking into account potential confounders, we observed a general attenuation of effect sizes, with several
315 associations dropping below the nominal and FDR-corrected level of significance. Further, we did not
316 obtain evidence for an association of CNS arousal and neuroticism, a personality trait that we regard as
317 highly plausible candidate for arousal alterations. Overall, across frequentist and Bayesian analyses and

318 irrespective of accounting for potential confounders or not, we obtained the strongest and most
319 compelling evidence for a link between openness to experience and CNS arousal.

320 To our knowledge, this investigation is the largest EEG study so far addressing the link between
321 CNS arousal and the Big Five personality traits. In keeping with this, the statistical power to detect
322 associations in this study was substantially higher when compared to the vast majority of previous
323 investigations, which usually featured a sample size fewer than 100 subjects. Given the present study
324 design and analysis procedure, the achieved statistical power enables to conclude that neuroticism is
325 unlikely to account for more than 4% ($r_s \geq 0.2$) of the variance in CNS arousal, since the probability ($1 -$
326 β) of identifying such an effect at $p < 0.05$ exceeded 99.98% (see suppl. Fig. S1 for a power plot).
327 Similarly, extraversion surpassed the FDR-corrected but not the Bonferroni-adjusted level of significance,
328 with the latter being reached with a probability above 92% given a true effect size of $r \geq 0.20$. Thus, if
329 extraversion and neuroticism are truly associated with CNS arousal, correlations are certainly below r
330 $= 0.20$. This is well in agreement with a study of 708 meta-analytically derived correlations in the field
331 of personality and individual differences research, suggesting a median reported effect size of $r = 0.19$
332 (Gignac & Szodorai, 2016). Notably, when considering preregistered studies only, the median effect size
333 has been reported to be even lower ($r = 0.12$; Schäfer & Schwarz, 2019). Accordingly, to elucidate the
334 biological basis of individual differences in human personality, we believe that there is an urgent need
335 for large (collaborative) studies with at least several hundreds and preferably thousands of subjects.

336 The present study adds empirical results to the ongoing debate of whether extraverted individuals
337 exhibit lower habitual levels of CNS arousal. Consistent with the theoretical assumptions, our primary
338 analyses provided supportive evidence for a negative correlation between extraversion and arousal.
339 Nevertheless, there remain some reservations that we would like to outline. First, although we used sex-
340 and age-normalized T-scores, we still observed associations between the NEO personality scores and
341 both sex and age (suppl. Table S3). After considering sex and age as additional covariates, the observed
342 associations between extraversion and CNS arousal did not remain significant after multiple testing
343 correction. Hence, the present results do not provide stringent support for extraversion to share unique
344 variance with CNS arousal beyond the effects of sex and age. Second, Bayesian analyses provided only
345 anecdotal to moderate evidence for the proposed link. This may partly be explained by the selected
346 priors. Here, we used a symmetrically scaled $1/3$ beta prior, which is the default setting of R package
347 ‘BayesFactor’ (Morey & Rouder, 2018). This prior corresponds to the expectation that with an 80%
348 probability the true effect falls in between $r = -0.5$ and $r = 0.5$. However, in light of the reported effect
349 sizes in the field of individual differences research (Gignac & Szodorai, 2016; Schäfer & Schwarz, 2019),
350 this prior width might still be considered too wide, and resulting Bayes factors may thus show some bias
351 towards favoring the null hypothesis. Notably, the software package JASP, which is widely used in the
352 social and behavioral sciences, implicates an even more naive uniform prior as default setting (JASP

353 Team, 2020). Hence, given the rising popularity of Bayesian analyses in the life sciences, we feel that
354 the selection of adequate prior widths may be one crucial topic of the future scientific debate. Taken
355 together, we here find some evidence supporting Eysenck's postulations concerning the link between
356 extraversion and CNS arousal, but the observed effect strength suggests that even larger sample sizes
357 are required to establish reliable associations that withstand a rigorous control for potential confounders.
358 An elaborated a priori knowledge of the expected effect sizes may further increase the study power.

359 Although we did not obtain evidence for a link between CNS arousal and neuroticism, exploratory
360 analyses revealed indications for a negative association with neuroticism facet 'impulsiveness'. This result
361 was the most compelling among all facet associations and remained significant after multiple-testing
362 correction. Interestingly, impulsiveness showed relatively low correlations with the other neuroticism
363 facets and, in contrast to them, correlated positively with facets of extraversion and openness to
364 experience (suppl. Fig. S2). In this light, impulsiveness may be considered as rather atypical facet of the
365 higher-order trait neuroticism. Notably, the observation of low habitual arousal levels in individuals
366 exhibiting impulsive behavior is well in line with previously proposed concepts (Eysenck, 1967; Hegerl
367 & Hensch, 2014; Zuckerman, 1979).

368 In comparison to neuroticism facet 'impulsiveness', our analyses did not reveal indications for a
369 link between neuroticism facet 'depression' and CNS arousal. By applying the Vigilance Algorithm
370 Leipzig, several previous studies provided supportive evidence for an association between clinical
371 depression and enhanced CNS arousal in the resting state (Hegerl et al., 2012; Sander et al., 2018;
372 Schmidt et al., 2016, 2017; Ulke et al., 2017; Ulke, Tenke, et al., 2019; Ulke, Wittekind, et al., 2019;
373 Wittekind et al., 2016). Although the present study included subjects without a current depression
374 diagnosis, it can be assumed that alterations in CNS arousal occur in both the normal and pathological
375 range of human behavior. This is consistent with the view that personality traits and psychopathology
376 are no distinct entities, but may rather manifest along a common spectrum of functioning (Widiger,
377 2011). This argumentation also ties in with the postulations of the Research Domain Criteria Project
378 (RDoC), according to which mental diseases can be considered to fall along multiple continuous trait
379 dimensions, with traits ranging from normal to the extreme (Cuthbert & Insel, 2013). Intriguingly, the
380 RDoC project considers 'arousal and regulatory systems' as one out of five fundamental domains to
381 describe and classify psychiatric disorders. In the present study, the lack of evidence for an association
382 between CNS arousal and 'depression' may be explained by lower effect sizes among healthy subjects
383 relative to the study of healthy control vs. in-patient samples.

384 Across all analyses, we obtained the strongest evidence for an association between the Big Five
385 personality trait openness to experience and CNS arousal. Bayes factors indicated 'extreme evidence'
386 ($BF_{10} > 100$) and 'strong evidence' (BF_{10} ranging from 10 to 30) for this link, respectively, depending
387 on whether we considered zero-order correlations or whether sex, age, and daytime of EEG assessment

388 served as covariates. Interestingly, previous genetic correlation analyses suggest a positive association
389 between openness to experience and major depression as well as bipolar disorder (Lo et al., 2017). These
390 mental diseases have both been argued to possess arousal-related pathophysiologies (Hegerl & Hensch,
391 2014). However, since manic and depressive-like behavior have been postulated to be linked to habitually
392 low vs. high arousal levels, it remains difficult to deduce the sign of the potential association between
393 openness to experience and CNS arousal. Importantly, previous investigations have shown that openness
394 to experience positively correlates with sensation seeking (Aluja, García, & García, 2003). Further,
395 openness to experience has been reported to positively correlate with extraversion, which is also shown
396 in the present dataset (suppl. Table S1). On this account, we regard the present findings of lower arousal
397 levels in subjects scoring high on openness to experience as consistent with the concepts of Eysenck
398 (1967) and Zuckerman (1979).

399 Our study poses some limitations that need to be addressed. First, our participants were, on
400 average, 58 years old, and reported association strengths may not generalize across other age groups. In
401 particular, we previously observed a general tendency towards stronger effect sizes for arousal
402 associations among the younger age groups (Jawinski et al., 2017). This might be explained by the higher
403 EEG total power in younger adults and a possibly related higher accuracy of EEG-vigilance
404 classifications. Further, we here addressed the relationship between individual differences in personality
405 and habitual levels of CNS arousal by means of an EEG resting-state paradigm. However, one major
406 emphasis of Eysenck's theory is put on the differential performance of extraverts and introverts as a
407 function of arousal-enhancing situational factors (Brocke & Battmann, 1992). Thus, an interesting future
408 direction might be the use of VIGALL in experimental studies with behavioral performance outcomes
409 (e.g., as done previously by Huang et al., 2017) while taking into account the 'hedonic tone of an
410 individual', i.e., the preferred level of excitation. Lastly, it should be noted that the present study sought
411 to answer the question of whether CNS arousal but not the EEG, in general, is predictive for basic
412 personality traits. Stronger associations may be derived by the application of machine learning models
413 trained on the EEG to directly predict human personality (for an example see Li et al., 2019).

414 **Conclusion**

415 To the best of our knowledge, this study is the largest EEG study so far addressing the relationship
416 between the Big Five personality traits and habitual levels of CNS arousal. Concerning Eysenck's Arousal
417 Activation Theory, our results provide some support for extraversion and no support for neuroticism to
418 be linked to CNS arousal. Intriguingly, Bayesian and frequentist analyses revealed convincing evidence
419 for a link between openness to experience and lower levels of CNS arousal. In addition, among the lower-
420 order personality traits, we obtained evidence for neuroticism facet 'impulsivity' and reduced CNS
421 arousal. We regard these findings as well in line with the postulations of Eysenck and Zuckerman and
422 consistent with the assumptions of the arousal regulation model. In total, the present study results agree

423 with meta-analytically derived effect sizes in the field of personality and individual differences research,
424 highlighting the need for large (collaborative) studies with at least several hundreds of subjects.

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432 Conflict of Interest

433 The authors have no financial or competing interests to declare.

434 Supplementary Material

435 Supplementary information is available at bioRxiv online.

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