Check for updates

OPEN ACCESS

EDITED BY César O. Tapia-Fonllem, University of Sonora, Mexico

REVIEWED BY

Agnieszka Anna Olszewska-Guizzo, NeuroLandscape Foundation, Poland Norma Isabel Rodelo Morales, University of Sonora, Mexico

*CORRESPONDENCE

Giuseppe Alessio Carbone i giuseppealessio.carbone@unier.it Angelo Panno i angelo.panno@unier.it

SPECIALTY SECTION

This article was submitted to Environmental Psychology, a section of the journal Frontiers in Psychology

RECEIVED 21 February 2023 ACCEPTED 30 March 2023 PUBLISHED 20 April 2023

CITATION

Imperatori C, Massullo C, De Rossi E, Carbone GA, Theodorou A, Scopelliti M, Romano L, Del Gatto C, Allegrini G, Carrus G and Panno A (2023) Exposure to nature is associated with decreased functional connectivity within the distress network: A resting state EEG study. *Front. Psychol.* 14:1171215. doi: 10.3389/fpsyg.2023.1171215

COPYRIGHT

© 2023 Imperatori, Massullo, De Rossi, Carbone, Theodorou, Scopelliti, Romano, Del Gatto, Allegrini, Carrus and Panno. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

Exposure to nature is associated with decreased functional connectivity within the distress network: A resting state EEG study

Claudio Imperatori¹, Chiara Massullo², Elena De Rossi¹, Giuseppe Alessio Carbone^{1,3}*, Annalisa Theodorou², Massimiliano Scopelliti⁴, Luciano Romano¹, Claudia Del Gatto¹, Giorgia Allegrini¹, Giuseppe Carrus² and Angelo Panno¹*

¹Cognitive and Clinical Psychology Laboratory, Department of Human Sciences, European University of Rome, Rome, Italy, ²Experimental Psychology Laboratory, Department of Education, Roma Tre University, Rome, Italy, ³Department of Psychology, University of Turin, Turin, Italy, ⁴Department of Human Studies, Libera Università Maria SS. Assunta (LUMSA), Rome, Italy

Introduction: Despite the well-established evidence supporting the restorative potential of nature exposure, the neurophysiological underpinnings of the restorative cognitive/emotional effect of nature are not yet fully understood. The main purpose of the current study was to investigate the association between exposure to nature and electroencephalography (EEG) functional connectivity in the distress network.

Methods: Fifty-three individuals (11 men and 42 women; mean age 21.38 ± 1.54 years) were randomly assigned to two groups: (i) a green group and (ii) a gray group. A slideshow consisting of images depicting natural and urban scenarios were, respectively, presented to the green and the gray group. Before and after the slideshow, 5min resting state (RS) EEG recordings were performed. The exact low-resolution electromagnetic tomography (eLORETA) software was used to execute all EEG analyses.

Results: Compared to the gray group, the green group showed a significant increase in positive emotions ($F_{1:50}$ =9.50 p=0.003) and in the subjective experience of being full of energy and alive ($F_{1:50}$ =4.72 p=0.035). Furthermore, as compared to urban pictures, the exposure to natural images was associated with a decrease of delta functional connectivity in the distress network, specifically between the left insula and left subgenual anterior cingulate cortex (T=-3.70, p=0.023).

Discussion: Our results would seem to be in accordance with previous neurophysiological studies suggesting that experiencing natural environments is associated with brain functional dynamics linked to emotional restorative processes.

KEYWORDS

nature exposure, contact with nature, restorativeness, simulated nature, EEG, functional connectivity, distress network, eLORETA

Introduction

The literature on contact with nature has consistently shown that mere exposure to certain types of natural environments may cause beneficial effects on humans (e.g., Bowler et al., 2010; Hartig et al., 2014; Houlden et al., 2018; Beute et al., 2020; Gladwin et al., 2022). One of the most important changes that have been observed in people when they are exposed to nature is

psychological restoration and stress reduction (Yao et al., 2021). In this regard, two prominent theories have been used to explain these effects. First, the Stress Reduction Theory (SRT; Ulrich, 1993) postulates that nature provokes automatic psychophysiological responses that modulate stress. The authors support the idea that living in urban areas may cause stress of which we are not fully aware, and, accordingly, nature can reduce it outside the consciousness. Second, the Attention Restoration Theory (ART; Kaplan and Kaplan, 1989) that sustains how nature restores voluntary attention, namely a type of attention requiring active effort for the individual to focus on an object meanwhile suppressing all the distracting stimuli. This occurs because the natural environment stimulates a type of involuntary and effortless attention, allowing for our cognitive functions to restore. Both the SRT and the ART see the changes observed in individuals as evolutionary responses to environments that can guarantee high rates of survival (Bratman et al., 2012). In this regard, Kaplan (1995) has convincingly discussed the potential integration of the two theories and related processes. Interestingly, these restorative effects have been proved to be comparable, although not completely overlapping, both in natural (outdoors) and simulated (indoors) natural environments, including virtual ones (Kjellgren and Buhrkall, 2010; Bratman et al., 2012; Pasca et al., 2022; Spano et al., 2022; Theodorou et al., 2023a,b).

Despite the well-established evidence supporting the restorative potential of nature exposure, the neurophysiological underpinnings of the restorative cognitive/emotional effect of nature are not yet fully understood. Two recent systematic reviews and meta-analyses that focused on both self-reported measures and psychophysiological bio-signals (i.e., systolic and diastolic blood pressure, heart rate variability, and cortisol levels) showed that exposure to natural environments is associated with stress reduction and/or enhanced relaxation (Mygind et al., 2021; Yao et al., 2021).

Similar findings have been also reported in studies investigating brain functional dynamics by means of different neuroimaging techniques (Norwood et al., 2019). In the same way, it has been detected that the exposure to urban stimuli is related with the activation of brain areas and/or large-scale brain networks linked to stress and negative affect (Ancora et al., 2022).

Particularly, the electroencephalography (EEG) is widely used for the investigation of brain activity related to the exposure of natural settings (Grassini et al., 2022), providing an important source of information on changes in neurophysiological dynamics associated with stress-related processes (Alonso et al., 2015; Giannakakis et al., 2019; Saeed et al., 2020; Massullo et al., 2022). More specifically, compared to the traditional visual EEG scoring, through quantitative EEG (qEEG; i.e., numerical computations of parameters from the EEG data) it is possible to estimate different parameters, such as EEG power spectra (i.e., the power distribution of EEG series in the frequency domain), that offer a significant source of information in terms of brain activity (Thatcher, 2010; Lehembre et al., 2012).

In this research field, previous EEG studies showed that experiencing natural environments is associated with several neurophysiological patterns linked to stress reduction, including increased alpha (Ulrich, 1981; Chang et al., 2008; Grassini et al., 2019, 2022; Sahni and Kumar, 2020; Koivisto et al., 2022; Olszewska-Guizzo et al., 2022b) and theta power (Olszewska-Guizzo et al., 2022b), as well as decreased beta activity (Grassini et al., 2019).

Of relevance, an important advance in the analysis of EEG signal is the investigation of functional connectivity referring to the calculation of "various measures of neural dynamics to functional brain state, determined by behavior, cognition, or neuropathology" (Srinivasan et al., 2007). Particularly, EEG connectivity-based measures, such as EEG coherence, provide a valuable estimation of the functional interactions between brain structures operating in specific frequency bands, offering significant information about network dynamics and functional integration across neural systems (Srinivasan et al., 2007; Whitton et al., 2018).

To the best of our knowledge, only one study has investigated the association between EEG functional connectivity and exposure to nature. Chen et al. (2020) showed that, as compared to a nonrestorative scenario (i.e., a traffic island), a restorative one (i.e., a wooded garden) was associated with a significant increase of theta connectivity. Therefore, in order to extend previous EEG findings, the main aim of the current study was to investigate the association between exposure to nature and EEG functional connectivity among brain nodes involved in emotional distress (i.e., the distress network; De Ridder et al., 2011; Pattyn et al., 2018). According to the view that natural environments can lead to stress reduction (Norwood et al., 2019; Mygind et al., 2021; Yao et al., 2021), we hypothesized that compared to images depicting urban spaces (i.e., gray condition), pictures showing green spaces (i.e., green condition) would be associated with a decrease of functional connectivity within the distress network.

Materials and methods

Participants

An *a priori* power analysis was conducted using G*Power 3.1 software (Faul et al., 2009). It revealed that, given a probability level of 0.05, a sample size of 52 was required to provide a satisfactory statistical power $(1-\beta=80\%)$ with large effect size (i.e., d = 0.80) in a two-sided test. Given the novelty of the study, this effect size was estimated according to a previous EEG connectivity report in this research field (Chen et al., 2020).

Participants were recruited in the campus of the European University of Rome through advertising. The enrollment lasted from November 2021 to May 2022. The following inclusion criteria were applied: right-handedness [i.e., Laterality Quotient > of 60 according to the Edinburgh Handedness Inventory–short form (EHI–SF; Veale, 2014)], self-reported lifetime negative anamnesis of neurological diagnosis, normal or corrected-to-normal vision, absence of use of active drugs (including psychotropic medications) on the central nervous system in the 2 weeks before the study. A checklist with dichotomous items was administered to assess inclusion criteria as well as socio-demographic data (e.g., age, educational attainment) and clinical data (e.g., tobacco and alcohol use in the last 6 months).

Eighty-seven university students were assessed for eligibility after providing their informed consent. Fifty-eight participants fulfilled the inclusion criteria and were enrolled in the present research. Five participants (8.62%) had no suitable EEG recordings (i.e., bad EEG signal), thus we have excluded such participants from the analyses. Therefore, the final sample consisted of 53 participants (11 men and 42 women; mean age 21.38 ± 1.54 years). Participants received course credits for their participation and were not aware about the experimental hypothesis. This research was approved by the European University's ethics review board (Prot. N. 11/2021) and was performed according to the Helsinki declaration standards.

Study design and procedures

In line with previous experimental research on exposure to natural and built environments (e.g., Chen et al., 2020), the current study was performed in the following three consecutive time-points.

Pre-stimuli assessment (T0)

After providing the written informed consent, all participants underwent a 5 min resting state (RS) EEG recording (T0-RS-EEG condition) and were administered the state version of the Positive and Negative Affect Schedule (PANAS; Watson et al., 1988), and the state version of the Subjective Vitality Scale (SVS-S; Ryan and Frederick, 1997).

Stimuli administration

After the pre-assessment phase, participants were randomly assigned to the two groups using the online software Research Randomizer:¹ (i) the green group and (ii) the gray group. A slideshow consisting of images depicting a natural and an urban scenarios were, respectively, presented to the green and the gray group. More specifically, 20 natural and urban digitized color pictures were used. Images were presented for 12 s and were preceded by a fixation cross placed centrally on the screen with a duration of 1 s. Participants were instructed to freely explore each picture.

Post-stimuli assessment (T1)

Immediately at the end of the slideshow, all participants underwent another 5 min RS EEG recording (T1-RS-EEG condition) and were asked to complete again the state version of the PANAS and the SVS-S. All participants also completed the Brief Symptoms Inventory (BSI; Derogatis and Melisaratos, 1983) and were asked to evaluate the pleasantness (i.e., *"To what extent did you like the images shown?"*), the quality (i.e., *"How would you rate the quality of the images shown?"*), the lightness (i.e., *"How would you rate the lightness of the images shown?"*), and the familiarity (i.e., *"How familiar are you with the place you saw in the pictures?"*) of the stimuli using a 10-point visual analog scale. These measures have been frequently used in earlier research on exposure to natural and built environments (e.g., Carrus et al., 2013).

EEG recordings and analysis

All EEG recordings were performed in an eye-closed RS condition with participants sitting on a comfortable chair (Massullo et al., 2020). EEG data were recorded using a 31-electrode cap placed according to the International 10–20 System and acquired using Micromed System Plus digital EEGraph (Micromed© S.p.A., Mogliano Veneto, TV, Italy). In order to avoid the possible effect of alcohol and caffeine on EEG data, participants were asked to refrain from them immediately before their EEG recordings (i.e., at least 4h). All EEG recordings lasted 5 min and were carried out with the following parameters: 256 Hz sampling rate, and A/D conversion performed at 16 bits. The reference electrodes were placed on the right mastoid. Impedances were kept below $5 k\Omega$ before starting EEG recording and checked again at the end of EEG session (i.e., T0 and T1) for each participant.

EEG recordings were offline filtered 1-40 Hz (with values corresponding, respectively, to high and low frequency band-pass filters) using the EEGLAB toolbox for MatLab v.2022.1 (Delorme and Makeig, 2004). Successively, the average reference was computed and a first visual inspection of the most noticeable and severe artifacts was carried out. Furthermore, in order to remove the main electrical, muscular, and visual artifacts, an independent component analysis (ICA) based on the infomax decomposition algorithm ("runica" tool of EEGLAB), was applied to all EEG channels. After the ICA decomposition, bad components were removed using the MARA EEGLAB plug-in and by visual inspection (Winkler et al., 2011, 2014). Subsequently, the identification of artifact-contaminated channels was mainly carried out in two ways: firstsly, the data spectrum of the channels was observed using the Plot \rightarrow Channel spectra and maps tool; secondly, the EEGLAB's "automatic channel rejection" function was further used to identify channels with kurtosis above five z-scores. Finally, a three-dimensional spherical spline interpolation was performed on the most artifact-contaminated channels (Perrin et al., 1989; Ferree, 2006).

The minimum length of the artifact-free EEG recording included in the analysis was 200 s (even if not consecutive) for each participant for each condition. Artifact-free data were then fragmented into epochs of 4 s for the EEG coherence analysis (Miljevic et al., 2022).

The exact Low-Resolution Brain Electromagnetic Tomography software (eLORETA; Pascual-Marqui et al., 2011) was used for all EEG analyses (including epochs fragmentation). This is a validated computer program for the investigation of large-scale brain networks functional connectivity (Liu et al., 2018). The eLORETA is considered one of the most widely adopted systems among the brain source localization procedures (Jatoi et al., 2014; Halder et al., 2019; Asadzadeh et al., 2020) able to detect electrocortical activity with rigorous localization capacity in terms of cortical, limbic (e.g., the hippocampus and the amygdaloid complex) and para-limbic (e.g., sta) have been applied (e.g., Mulert et al., 2004; Zotev and Bodurka, 2020).

In the current study, in order to evaluate the connectivity in the distress network, eight Regions of Interest (ROIs; Table 1) were pre-defined according to a previous LORETA study (Pattyn et al., 2018) choosing the "ROI-maker#2 method" (i.e., based on the Brodmann's areas) available in the eLORETA software.

The connectivity analysis was performed by computing the Lagged Phase Synchronization (LPS; Pascual-Marqui et al., 2011), one of the main neurophysiological indices used for the assessment of brain functional connectivity (Olbrich et al., 2014; Hata et al., 2016; Imperatori et al., 2017). The LPS reduces the artifacts (e.g., the volume-conduction) by removing the instantaneous zero-lag contribution (Hata et al., 2016), and it calculates the similarity between signals in the frequency domain, based on normalized Fourier transform (performed through the eLORETA), with values ranging from 0 (i.e., no synchronization) to 1 (i.e., the maximum synchronization).

¹ https://www.randomizer.org/

TABLE 1 eLORETA Montreal Neurological Institute (MNI) coordinates of the distress network.

| Regions of interest | Brodmann areas | eLORETA MNI coordinates* | | |
|---------------------|-------------------|-----------------------------|-----|-----|
| | | x | У | z |
| Left insula | 13 | -40 | -10 | 10 |
| Right insula | 13 | 40 | -5 | 10 |
| Left amygdala | 34+35 | -20 | -10 | -20 |
| Right amygdala | 34+35 | 20 | -10 | -20 |
| Left sgACC | 25 | -10 | 20 | -15 |
| Right sgACC | 25 | 5 | 15 | -15 |
| Left dACC | 24+32 | -10 | 15 | 30 |
| Right dACC | 24+32 | 5 | 15 | 30 |

MNI, montreal neurological institute; eLORETA, exact low-resolution brain electromagnetic tomography software; sgACC, subgenual anterior cingulate cortex; and dACC, dorsal anterior cingulate cortex.

*eLORETA ROIs coordinates reconstruction.

According to previous studies (e.g., Canuet et al., 2011; de la Salle et al., 2016; Kitaura et al., 2017), in order to perform the source reconstruction of the brain networks, the "single nearest voxel" option (i.e., each ROI consisting of a single voxel, the closest to each seed) was selected in the eLORETA software. In the current study, the following four frequency bands were considered (Mazza et al., 2014): delta (0.5–4 Hz), theta (4.5–7.5 Hz), alpha (8–13 Hz), and beta (13.5–30 Hz).

Self-report questionnaires

The PANAS (Watson et al., 1988) is a 20-item self-report questionnaire assessing both negative (e.g., scared, distressed) and positive affect (e.g., strong, enthusiastic). In the current study, the "state" (i.e., "*at this moment*") Italian version of the PANAS (Terraciano et al., 2003) was used. At T0, the Cronbach's alpha values were 0.81 for the positive and 0.86 for the negative subscale. At T1, the Cronbach's alpha values were 0.90 for both positive and negative subscale.

The SVS-S (Ryan and Frederick, 1997) is a seven-item self-report questionnaire assessing subjective experience of being full of energy and alive. Each item is rated on a seven-point Likert scale ranging from "not at all" to "very true," with higher scores indicating greater subjective vitality. In the current study, the Italian version of the SVS-S was used, and the Cronbach's alpha values were 0.81 and 0.86 for T0 and T1, respectively.

The BSI (Derogatis and Melisaratos, 1983) is a self-report questionnaire widely used to assess psychological symptoms (e.g., anxiety and depressive symptoms). It is composed of 53 items rated on a five-point Likert scale (0–4), with higher scores indicating more severe self-reported symptoms. The BSI provides a Global Severity Index (GSI), which is designed to measure overall psychopathological distress. This questionnaire is characterized by good internal consistency and test–retest reliability (Derogatis and Melisaratos, 1983). In the current study, the Italian version of the scale (Adawi et al., 2019) was used and the Cronbach's alpha was 0.96 for the GSI.

Statistical analysis

In order to investigate the association between nature exposure and EEG functional connectivity in the distress network, the following comparisons were performed: (i) T0-RS-EEG green group vs. T0-RS-EEG gray group, and (ii) T1-RS-EEG green group vs. T1-RS-EEG gray group. Comparisons were performed using the statistical non-parametric mapping methodology provided by the eLORETA software (Nichols and Holmes, 2002). This approach is based on Fisher's permutation test and used a nonparametric randomization procedure in order to perform the correction of significance for multiple testing. Briefly (for more details see Kitaura et al., 2017; Hata et al., 2019), this procedure computes 5,000 data randomizations to determine the critical probability threshold of T-values corresponding to a statistically corrected (i.e., after the multiple comparisons among each ROIs in each frequencies) p values (p=0.05 and p=0.01). The eLORETA software also computed the effect size thresholds for T-statistics corresponding to Cohen's d values (Cohen, 1988): small = 0.2, medium = 0.5, and large = 0.8.

Two-way chi-squared and univariate ANOVAs were used to analyze differences between groups at T0, respectively for dichotomous and dimensional measures. According to the recommendations for pre-post study designs (Senn, 2006), questionnaire data were analyzed using analysis of covariance (ANCOVA) with group (green vs. gray) as a between-subject factor, value at T1 as dependent variables, and value at T0 as a covariate. In order to determine effect sizes, Cohen's d and Cohen's d_{ppc2} (Morris, 2008; equation number 8) values were calculated for ANOVAs and ANCOVA analyses, respectively. Chi-squared, ANOVA and ANCOVA analyses were performed using SPSS (version 18.0).

Results

For all study participants the qualitative visual evaluation of the EEG recordings (i.e., T0-RS-EEG and T1-RS-EEG) showed no relevant evidence of unusual neurophysiological patterns, such as epileptic discharges and/or focal delta slow waves. In the T0-RS-EEG condition the average epochs analyzed for the green group and the gray group was 70.54 ± 6.62 and 68.56 ± 9.28, respectively ($F_{1;51}$ = 0.797 p = 0.376). In the T1-RS-EEG condition, the average epochs analyzed for green group and gray group was 70.88 ± 8.16 and 66.96 ± 7.17, respectively ($F_{1;51}$ = 3.466, p = 0.068).

The two groups did not significantly differ in age and in other socio-demographic variables. At T1, a significant group main effect was observed for PANAS positive dimension and SVS-S total score. Specifically, compared to the gray group, the green group showed a significant increase of positive emotions (green group: 28.62 ± 8.70 vs. gray group: 24.70 ± 7.41 ; $F_{1;50} = 9.50$; p = 0.003) and SVS-S total score (green group: 27.42 ± 8.05 vs. gray group: 23.15 ± 7.48 ; $F_{1;50} = 4.72$; p = 0.035). Detailed descriptive and F statistics are reported in Table 2.

Functional connectivity results

The effect sizes for *T*-threshold were 1.428, 3.571, and 5.713, corresponding, respectively, to small, medium, and large effect sizes.

| Variable | Green group (N=26) | Gray group (<i>N</i> =27) | Test statistics | ρ | Effect size |
|----------------------------------|-----------------------|-------------------------------|---------------------------|---------|---------------------|
| Age: M±DS | 21.15 ± 1.43 | 21.59 ± 1.62 | $F_{1;51} = 1.084$ | 0.303 | |
| Women: <i>N</i> (%) | 21 (80.8%) | 21 (77.8%) | $\chi^2 = 0.072$ | 0.788 | |
| Tobacco use: N (%) | 12 (46.2%) | 10 (37%) | $\chi^2 = 0.453$ | 0.501 | |
| Alcohol use: $N(\%)$ | 23 (88.5%) | 24 (.9%) | $\chi^2 = 0.002$ | 0.961 | |
| BSI-GSI: M ± DS | 0.85 ± 0.68 | 0.89 ± 0.75 | $F_{1;51} = 0.047$ | 0.829 | <i>d</i> = 0.056 |
| Stimuli pleasantness: M \pm DS | 7.46 ± 1.45 | 3.33 ± 1.47 | $F_{1;51} = 106.14$ | < 0.001 | d = -2.828 |
| Stimuli quality: M±DS | 7.19±1.86 | 5.04 ± 1.99 | $F_{1;51} = 16.602$ | < 0.001 | d=-1.116 |
| Stimuli lightness: M±DS | 7.69 ± 1.78 | 5.37 ± 2.20 | $F_{1;51} = 17.693$ | < 0.001 | d = -1.157 |
| Stimuli familiarity: M±DS | 8.19±1.44 | 5.85 ± 2.49 | $F_{1;51} = 17.336$ | < 0.001 | d = -1.145 |
| PANAS-N T0: M±DS | 18.00 ± 5.38 | 15.19±5.95 | F _{1,51} = 3.255 | 0.077 | d = -0.495 |
| PANAS-P T0: M±DS | 29.15±6.82 | 30.15±5.13 | $F_{1;51} = 0.362$ | 0.550 | <i>d</i> = 0.166 |
| SVS-ST0: M±DS | 26.77±7.03 | 25.56 ± 6.17 | $F_{1;51} = 0.447$ | 0.507 | d = -0.183 |
| PANAS-NT1: M±DS | 15.15 ± 6.34 | 13.48 ± 4.85 | $F_{1;50} = 0.018$ | 0.893 | $d_{ppc2} = -0.198$ |
| PANAS-P T1: M±DS | 28.62 ± 8.70 | 24.70 ± 7.41 | $F_{1;50} = 9.501$ | 0.003 | $d_{ppc2} = 0.805$ |
| SVS-ST1: M±DS | 27.42 ± 8.05 | 23.15 ± 7.48 | $F_{1;50} = 4.722$ | 0.035 | $d_{ppc2} = 0.456$ |

TABLE 2 Detailed descriptive and F statistics.

M, mean; N, number; SD, standard deviation; BSI, brief symptoms inventory; GSI, global severity index; PANAS-N, positive and negative affect schedule-negative affect; PANAS-P, positive and negative affect schedule-positive affect; SVS-S, subjective vitality scales-state version; T0, pre-stimuli assessment; and T1, post-stimuli assessment. Means and SDs for T1 are not adjusted for the covariates. In bold significant values.

At T0, the thresholds for significance (corrected for multiple testing) were $T=\pm 3.462$ corresponding to p < 0.05, and $T=\pm 3.883$, corresponding to p < 0.01. In this condition, no significant differences were observed between groups.

At T1, the thresholds for significance (corrected for multiple testing) were $T=\pm 3.463$ corresponding to p < 0.05, and $T=\pm 3.969$ corresponding to p < 0.01. In this condition, significant modifications were observed in delta band. Compared to the gray group, participants in the green group showed a decreased delta LPS between the left insula and left subgenual anterior cingulate cortex (sgACC; T=-3.70, p=0.023; Figure 1). No significant differences were observed in the other frequency bands.

Discussion

The main aim of the present study was to investigate the association between the exposure to nature and EEG functional connectivity in the distress network. Our results showed that the exposure to natural images, compared to urban pictures, was associated with a decrease of delta functional connectivity in the distress network, specifically between the left insula and left sgACC. Moreover, as compared to gray stimuli, the exposure to green spaces was also associated with a significant increase of positive emotions and the subjective experience of being full of energy and alive.

Our results would seem to be in accordance with previous neurophysiological studies suggesting that experiencing natural environments is associated with brain functional dynamics linked to emotional restorative processes and stress reduction (Grassini et al., 2019, 2022; Norwood et al., 2019; Sahni and Kumar, 2020). Indeed, both the insula and the sgACC are known to be brain regions



commonly involved in emotional distress (Li et al., 2018; Pattyn et al., 2018). More specifically, while the insula plays a crucial for the integration of interoceptive information and global-emotional processing (e.g., emotional salience detection) and awareness (Duerden et al., 2013), the sgACC, given its connections with subcortical and cortical structures, is critically involved in regulating visceral and autonomic responses to stressful events, assigning emotional valence to internal and external stimuli, and emotional expression (Pizzagalli, 2011; Rolls, 2019). These two brain areas are strongly and reciprocally connected (Simmons et al., 2013) and abnormal increased resting state insula-sgACC connectivity has been

suggested as a neurophysiological pattern associated to negative emotions in response to stress (Shao et al., 2018). Consistently, it has been reported that adaptive response to stress (i.e., high level of resilience) was associated with reduced sgACC-insula RS connectivity immediately following stress induction (Shao et al., 2018).

The involvement of these structures has been also documented by several studies measuring the effect of environment on brain activity. For example, the decrease of subgenual prefrontal cortex activity has been detected (Bratman et al., 2015) following the exposure to a brief nature experience (i.e., a 90-min walk in a natural setting). It has also been shown that the limbic and paralimbic areas (including the insula) were more active in uncomfortable urban environments reflecting more negative emotional stimulation (Kim and Jeong, 2014). Furthermore, previous EEG studies showed that unpleasant pictures elicited increased delta coherence over long-range circuits of the brain (Klados et al., 2009; Guntekin et al., 2017).

Intriguingly, the decrease of delta connectivity between the insula and the sgACC was detected in the left hemisphere. Although the right hemisphere dominance in emotional processing is generally recognized, the traditional division of cognitive functions according to the left–right brain dichotomy seems to have become obsolete (Stankovic, 2021). Moreover, the modern view of brain supports a dynamic large-scale brain networks model underlying different components of emotional experience (i.e., generation, perception, and regulation; Morawetz et al., 2020; Palomero-Gallagher and Amunts, 2022). For example, it has been proposed (Stankovic, 2021) that the right-biased lateralization of emotional experience can be changed rapidly due to strong environmental input such as acute stress.

Thus, considering the above mentioned findings, the current neurophysiological pattern (i.e., decreased delta coherence between the left insula and the left sgACC) might reflect the neural underpinning of the processes related to emotional restorative effects of interacting with nature. Obviously, the other side of the coin could be considered, and both are not mutually exclusive: urban scenarios can induce brain responses associated with negative emotion and stress that may be restored through the exposure to natural environments (Ancora et al., 2022).

Of relevance, no significant associations were detected in other frequency bands (e.g., alpha and theta) previously associated with nature exposure and stress reduction (Ulrich, 1981; Chang et al., 2008; Grassini et al., 2019, 2022; Sahni and Kumar, 2020; Koivisto et al., 2022). The discrepancies between previous reports and the current research may be explained by several differences in study designs and methods. More specifically, our study differs from these previous studies by investigating, for the first time, EEG functional connectivity in a pre-defined brain network (i.e., the distress network). Indeed, while EEG power reflects the amount of activity in certain frequency bands (Dressler et al., 2004), EEG connectivity provides "information on the degree of synchrony of brain activity at different locations for each frequency, independent of power" (Bowyer, 2016). Furthermore, compared to previous studies that evaluated the modifications of EEG parameters during the view of nature-related stimuli, we have investigated EEG data during the RS condition, which is thought to reflect the neurophysiological base rhythm providing potentially valuable information on the functional dynamic interactions between brain structures (Raichle, 2011). Thus, the current findings should be considered specific to the eyes-closed RS condition and future EEG connectivity investigations should be performed also during different task-related conditions.

The present findings should be considered while taking into account several limitations. First, our study provides an image that is informative of the dynamics that take place right after exposure to nature, while in future research it could be interesting to investigate if the changes observed in the neural correlates would last and, if so, for how long. These findings could also contribute to designing interventions aimed at restoration through nature exposure, providing protocols for the duration of the intervention (e.g., a certain amount of time every day for several days/weeks) and testing the short-and long-term consequences. This issue have been previously investigated by a handful of studies (e.g., Barton and Pretty, 2010; Shanahan et al., 2016), and further research is needed. Secondly, although an a priori power analysis was performed according to a previous EEG coherence study, the sample size was adequate for detect a large effect size. Moreover, notwithstanding university undergraduate students represent an easily accessible population, a potential source of bias should be considered (Medin et al., 2017).

Thirdly, we employed two photo slideshows showing green vs. gray spaces and found significant results in line with previous studies. Nevertheless, it would be interesting to test the replicability of the findings in outdoor nature including blue or white spaces, to see if there would be any differences in the distress network activity observed. Moreover, future studies investigating differences in neurophysiological underpinnings between the exposure to urban green spaces vs. wild green spaces are needed. Furthermore, other stress-related psychophysiological markers (e.g., the cortisol levels) in relation to nature exposure were not evaluated, and future studies would need to fill in such a gap. Lastly, while the eLORETA software is a reliable tool to investigate brain connectivity, it has an intrinsic limit in space resolution. Moreover, although 31 electrodes might be appropriate for investigating brain networks, it is known that the spatial localization accuracy of EEG increases with high-density (e.g., 256 electrodes) recording approaches (Dattola et al., 2020). Thus, our results should be treated with caution and should be deeply investigated in further studies combining multimodal neuroimaging data (i.e., high-density EEG/fMRI) in larger representative samples in order to better understand the neural dynamics underlying the restorative effect of nature.

Despite these limitations, we believe that our results have relevant theoretical and practical implications. From a theoretical point of view, these findings add up to all previous research that suggests psychophysiological restoration induced by nature exposure (Berto, 2014). Importantly, they are in line with both SRT and ART, and the integrative framework between these two theories (Ulrich, 1981; Kaplan and Kaplan, 1989; Ulrich et al., 1991; Kaplan, 1995). Indeed, the decreased delta coherence we found between the left insula and the left sgACC is consistent with the SRT postulating that nature exposure reduces physiological stress. Moreover, this neurophysiological pattern suggests decreased effort in voluntary attention as theorized in the ART.

Starting from these findings, it would be interesting to investigate the individual differences that are related to different brain dynamics in the distress network, when exposing people to nature, such as environmental identity, connectedness to nature, sensation seeking, venturesomeness, and preferences for a specific natural environment. In this regard, future studies could examine the psychophysiological dynamics in response to different types of natural environments (e.g., sea, lake, desert, mountain, and arctic) or different types of green space. In this regard, a recent systematic review (Beute et al., 2020) detected a general beneficial relation between several green space types (e.g., park, garden, and forest) and mental health, suggesting the need of further investigations focused on underlying pathways of the association between specific green spaces (e.g., types and characteristics) and individual well-being. For example a recent multimodal EEG and functional near-infrared spectroscopy study (Olszewska-Guizzo et al., 2022a), showed that the passive exposure to a "therapeutic garden" (compared to a "residential green space" and to a "busy downtown") was associated with an improvement of mood status and a modification of brain activity (e.g., it had a moderating effect on frontal alpha asymmetry values) in both healthy and depressed individuals.

Furthermore, it could be relevant to study how the humannature interaction might provide restoration by observing the distress network activity (e.g., hiking, gardening, and swimming in a natural area; Bratman et al., 2019). Lastly, in recent years growing attention was directed to multisensorial experiences (i.e., indoor nature experience in which the visual stimulus is teamed with auditory and/or olfactory stimuli) and virtual reality experiences (i.e., nature exposure through head-mounted displays that can guarantee a high level of immersion to the user). Thus, observing the distress network activity also in multisensorial and virtual reality nature experiences could contribute to extend this new and expanding line of research.

As mentioned before, if future studies will confirm our results, they might be relevant to build up applied psychological interventions across different fields. However, importantly, the research should first test the duration of such beneficial effects. For instance, in clinical populations (e.g., with mental disorders or general medical patients), the exposure to nature could be important in alleviating both psychological and physical symptomatology (Collado et al., 2017). Occupational psychologists might use such interventions in order to restore employees' cognitive resources after exposure to intensive workload in the workplace. In the general population, the exposure to nature might be employed in individuals undergoing periods characterized by high levels of stress and/or negative mood. In all these cases, the research focusing on the distress network activity could offer additional insights into the dynamics involved in the restoration and help to tailor better-suited interventions for all circumstances.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

References

Adawi, M., Zerbetto, R., Re, T. S., Bisharat, B., Mahamid, M., Amital, H., et al. (2019). Psychometric properties of the brief symptom inventory in nomophobic subjects: insights from preliminary confirmatory factor, exploratory factor, and clustering analyses in a sample of healthy Italian volunteers. *Psychol. Res. Behav. Manag.* 12, 145–154. doi: 10.2147/PRBM.S173282

Alonso, J. F., Romero, S., Ballester, M. R., Antonijoan, R. M., and Mananas, M. A. (2015). Stress assessment based on EEG univariate features and functional connectivity measures. *Physiol. Meas.* 36, 1351–1365. doi: 10.1088/0967-3334/36/7/1351

Ancora, L. A., Blanco-Mora, D. A., Alves, I., Bonifacio, A., Morgado, P., and Miranda, B. (2022). Cities and neuroscience research: a systematic literature review. *Front. Psychol.* 13:983352. doi: 10.3389/fpsyt.2022.983352

Ethics statement

The studies involving human participants were reviewed and approved by European University's ethics review board (Prot. N. 11/2021). The patients/participants provided their written informed consent to participate in this study.

Author contributions

CI: study design, formal analysis, data interpretation, supervision, and writing—original draft. CM, EDR, and GAC: data acquisition, formal analysis, data interpretation, and writing—review and editing. AT and LR: data acquisition, resources, and writing—review and editing. MS: resources, supervision, and writing—review and editing. CDG: data interpretation, supervision, and writing—review and editing. GA: data acquisition, data interpretation, and writing—review and editing. GC: project administration, funding acquisition, and writing—review and editing. AP: conceptualization, study design, funding acquisition, project administration, supervision, and writing—original draft. All authors contributed to the article and approved the submitted version.

Funding

This work was part of the project named "Establishing Urban FORest based solutions in Changing Cities (EUFORICC)" financially supported by the Ministry of Education, University and Research (MIUR) of Italy (PRIN 20173RRN2S).

Conflict of interest

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

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Asadzadeh, S., Yousefi Rezaii, T., Beheshti, S., Delpak, A., and Meshgini, S. (2020). A systematic review of EEG source localization techniques and their applications on diagnosis of brain abnormalities. *J. Neurosci. Methods* 339:108740. doi: 10.1016/j.jneumeth.2020.108740

Barton, J., and Pretty, J. (2010). What is the best dose of nature and green exercise for improving mental health? A multi-study analysis. *Environ. Sci. Technol.* 44, 3947–3955. doi: 10.1021/es903183r

Berto, R. (2014). The role of nature in coping with psycho-physiological stress: a literature review on restorativeness. *Behav. Sci.* 4, 394–409. doi: 10.3390/bs4040394

Beute, F., Andreucci, M. B., Lammel, A., Davies, Z., Glanville, J., Keune, H., et al. (2020). *Types characteristics of urban peri-urban green spaces having an impact on human mental health wellbeing: a systematic review*. Wallingford, United Kingdom: EKLIPSE

Expert Working Group, UK Centre for Ecology & Hydrology. Available at: https:// eklipse.eu/wp-content/uploads/website_db/Request/Mental_Health/EKLIPSE_ HealthReport-Green_Final-v2-Digital.pdf

Bowler, D. E., Buyung-Ali, L. M., Knight, T. M., and Pullin, A. S. (2010). A systematic review of evidence for the added benefits to health of exposure to natural environments. *BMC Public Health* 10:456. doi: 10.1186/1471-2458-10-456

Bowyer, S. M. (2016). Coherence a measure of the brain networks: past and present. *Neuropsychiatr. Electrophysiol.* 2, 1–12. doi: 10.1186/s40810-015-0015-7

Bratman, G. N., Anderson, C. B., Berman, M. G., Cochran, B., De Vries, S., Flanders, J., et al. (2019). Nature and mental health: an ecosystem service perspective. *Sci. Adv.* 5:eaax0903. doi: 10.1126/sciadv.aax0903

Bratman, G. N., Hamilton, J. P., and Daily, G. C. (2012). The impacts of nature experience on human cognitive function and mental health. *Ann. N. Y. Acad. Sci.* 1249, 118–136. doi: 10.1111/j.1749-6632.2011.06400.x

Bratman, G. N., Hamilton, J. P., Hahn, K. S., Daily, G. C., and Gross, J. J. (2015). Nature experience reduces rumination and subgenual prefrontal cortex activation. *Proc. Natl. Acad. Sci. U. S. A.* 112, 8567–8572. doi: 10.1073/pnas.1510459112

Canuet, L., Ishii, R., Pascual-Marqui, R. D., Iwase, M., Kurimoto, R., Aoki, Y., et al. (2011). Resting-state EEG source localization and functional connectivity in schizophrenia-like psychosis of epilepsy. *PLoS One* 6:e27863. doi: 10.1371/journal.pone.0027863

Carrus, G., Lafortezza, R., Colangelo, G., Dentamaro, I., Scopelliti, M., and Sanesi, G. (2013). Relations between naturalness and perceived restorativeness of different urban green spaces. *Psyecology* 4, 227–244. doi: 10.1174/217119713807749869

Chang, C. Y., Hammitt, W. E., Chen, P. K., Machnik, L., and Su, W. C. (2008). Psychophysiological responses and restorative values of natural environments in Taiwan. *Landsc. Urban Plan.* 85, 79–84. doi: 10.1016/j.landurbplan.2007.09.010

Chen, Z., He, Y., and Yu, Y. (2020). Attention restoration during environmental exposure via alpha-theta oscillations and synchronization. *J. Environ. Psychol.* 68:101406. doi: 10.1016/j.jenvp.2020.101406

Cohen, J. (1988). Statistical Power Analysis for the Behavioral Sciences (2nd Edn.). Hillsdale, NJ: Erlbaum.

Collado, S., Staats, H., Corraliza, J. A., and Hartig, T. (2017). "Restorative environments and health" in *Handbook of Environmental Psychology and Quality of Life Research*. eds. G. Fleury-Bahi, E. Pol and O. Navarro (Cham: Springer International Publishing), 127–148.

Dattola, S., La Foresta, F., Bonanno, L., De Salvo, S., Mammone, N., Marino, S., et al. (2020). "Effect of sensor density on eLORETA source localization accuracy" in *Neural Approaches to Dynamics of Signal Exchanges*. eds. A. Esposito, M. Faundez-Zanuy, F. C. Morabito and E. Pasero (Singapore: Springer Nature), 403–414.

De La Salle, S., Choueiry, J., Shah, D., Bowers, H., Mcintosh, J., Ilivitsky, V., et al. (2016). Effects of ketamine on resting-state EEG activity and their relationship to perceptual/dissociative symptoms in healthy humans. *Front. Pharmacol.* 7:348. doi: 10.3389/fphar.2016.00348

De Ridder, D., Vanneste, S., and Congedo, M. (2011). The distressed brain: a group blind source separation analysis on tinnitus. *PLoS One* 6:e24273. doi: 10.1371/journal. pone.0024273

Delorme, A., and Makeig, S. (2004). EEGLAB: an open source toolbox for analysis of single-trial EEG dynamics including independent component analysis. *J. Neurosci. Methods* 134, 9–21. doi: 10.1016/j.jneumeth.2003.10.009

Derogatis, L. R., and Melisaratos, N. (1983). The brief symptom inventory: an introductory report. *Psychol. Med.* 13, 595–605. doi: 10.1017/S0033291700048017

Dressler, O., Schneider, G., Stockmanns, G., and Kochs, E. F. (2004). Awareness and the EEG power spectrum: analysis of frequencies. *Br. J. Anaesth.* 93, 806–809. doi: 10.1093/bja/aeh270

Duerden, E. G., Arsalidou, M., Lee, M., and Taylor, M. J. (2013). Lateralization of affective processing in the insula. *NeuroImage* 78, 159–175. doi: 10.1016/j. neuroimage.2013.04.014

Faul, F., Erdfelder, E., Buchner, A., and Lang, A. G. (2009). Statistical power analyses using G*power 3.1: tests for correlation and regression analyses. *Behav. Res. Methods* 41, 1149–1160. doi: 10.3758/BRM.41.4.1149

Ferree, T. C. (2006). Spherical splines and average referencing in scalp electroencephalography. *Brain Topogr.* 19, 43–52. doi: 10.1007/s10548-006-0011-0

Giannakakis, G., Grigoriadis, D., Giannakaki, K., Simantiraki, O., Roniotis, A., and Tsiknakis, M. (2019). Review on psychological stress detection using biosignals. *IEEE Trans. Affect. Comput.* 13, 440–460. doi: 10.1109/TAFFC.2019.2927337

Gladwin, T. E., Markwell, N., and Panno, A. (2022). Do semantic vectors contain traces of Biophilic connections between nature and mental health? *Ecopsychology* 15, 37–44. doi: 10.1089/eco.2022.0036

Grassini, S., Revonsuo, A., Castellotti, S., Petrizzo, I., Benedetti, V., and Koivisto, M. (2019). Processing of natural scenery is associated with lower attentional and cognitive load compared with urban ones. *J. Environ. Psychol.* 62, 1–11. doi: 10.1016/j.jenvp.2019.01.007

Grassini, S., Segurini, G. V., and Koivisto, M. (2022). Watching nature videos promotes physiological restoration: evidence from the modulation of alpha waves in electroencephalography. *Front. Psychol.* 13:2865. doi: 10.3389/fpsyg.2022.871143

Guntekin, B., Femir, B., Golbasi, B. T., Tulay, E., and Basar, E. (2017). Affective pictures processing is reflected by an increased long-distance EEG connectivity. *Cogn. Neurodyn.* 11, 355–367. doi: 10.1007/s11571-017-9439-z

Halder, T., Talwar, S., Jaiswal, A.K., and Banerjee, A. (2019). Performance evaluation of inverse methods for identification and characterization of oscillatory brain sources: Ground truth validation & empirical evidences. bioRxiv [Preprint]. doi: 10.1101/395780

Hartig, T., Mitchell, R., De Vries, S., and Frumkin, H. (2014). Nature and health. Annu. Rev. Public Health 35, 207–228. doi: 10.1146/annurev-publhealth-032013-182443

Hata, M., Hayashi, N., Ishii, R., Canuet, L., Pascual-Marqui, R. D., Aoki, Y., et al. (2019). Short-term meditation modulates EEG activity in subjects with post-traumatic residual disabilities. *Clin. Neurophysiol. Pract.* 4, 30–36. doi: 10.1016/j.cnp.2019.01.003

Hata, M., Kazui, H., Tanaka, T., Ishii, R., Canuet, L., Pascual-Marqui, R. D., et al. (2016). Functional connectivity assessed by resting state EEG correlates with cognitive decline of Alzheimer's disease–an eLORETA study. *Clin. Neurophysiol.* 127, 1269–1278. doi: 10.1016/j.clinph.2015.10.030

Houlden, V., Weich, S., Porto De Albuquerque, J., Jarvis, S., and Rees, K. (2018). The relationship between greenspace and the mental wellbeing of adults: a systematic review. *PLoS One* 13:e0203000. doi: 10.1371/journal.pone.0203000

Imperatori, C., Della Marca, G., Amoroso, N., Maestoso, G., Valenti, E. M., Massullo, C., et al. (2017). Alpha/theta Neurofeedback increases Mentalization and default mode network connectivity in a non-clinical sample. *Brain Topogr.* 30, 822–831. doi: 10.1007/s10548-017-0593-8

Jatoi, M. A., Kamel, N., Malik, A. S., and Faye, I. (2014). EEG based brain source localization comparison of sLORETA and eLORETA. *Australas. Phys. Eng. Sci. Med.* 37, 713–721. doi: 10.1007/s13246-014-0308-3

Kaplan, S. (1995). The restorative benefits of nature: toward an integrative framework. *J. Environ. Psychol.* 15, 169–182. doi: 10.1016/0272-4944(95)90001-2

Kaplan, R., and Kaplan, S. (1989). The Experience of Nature: A Psychological Perspective. Cambridge: Cambridge university press.

Kim, G. W., and Jeong, G. W. (2014). Brain activation patterns associated with the human comfortability of residential environments: 3.0-T functional MRI. *Neuroreport* 25, 915–920. doi: 10.1097/WNR.0000000000205

Kitaura, Y., Nishida, K., Yoshimura, M., Mii, H., Katsura, K., Ueda, S., et al. (2017). Functional localization and effective connectivity of cortical theta and alpha oscillatory activity during an attention task. *Clin. Neurophysiol. Pract.* 2, 193–200. doi: 10.1016/j. cnp.2017.09.002

Kjellgren, A., and Buhrkall, H. (2010). A comparison of the restorative effect of a natural environment with that of a simulated natural environment. *J. Environ. Psychol.* 30, 464–472. doi: 10.1016/j.jenvp.2010.01.011

Klados, M. A., Frantzidis, C., Vivas, A. B., Papadelis, C., Lithari, C., Pappas, C., et al. (2009). A framework combining delta event-related oscillations (EROs) and synchronisation effects (ERD/ERS) to study emotional processing. *Comput. Intell. Neurosci.* 2009;549419. doi: 10.1155/2009/549419

Koivisto, M., Jalava, E., Kuusisto, L., Railo, H., and Grassini, S. (2022). Top-down processing and nature connectedness predict psychological and physiological effects of nature. *Environ. Behav.* 54, 917–945. doi: 10.1177/00139165221107535

Lehembre, R., Gosseries, O., Lugo, Z., Jedidi, Z., Chatelle, C., Sadzot, B., et al. (2012). Electrophysiological investigations of brain function in coma, vegetative and minimally conscious patients. *Arch. Ital. Biol.* 150, 122–139. doi: 10.4449/aib.v150i2.1374

Li, B. J., Friston, K., Mody, M., Wang, H. N., Lu, H. B., and Hu, D. W. (2018). A brain network model for depression: from symptom understanding to disease intervention. *CNS Neurosci. Ther.* 24, 1004–1019. doi: 10.1111/cns.12998

Liu, Q., Ganzetti, M., Wenderoth, N., and Mantini, D. (2018). Detecting large-scale brain networks using EEG: impact of electrode density, head modeling and source localization. *Front. Neuroinform.* 12:4. doi: 10.3389/fninf.2018.00004

Massullo, C., Bersani, F. S., Carbone, G. A., Panno, A., Farina, B., Murillo-Rodriguez, E., et al. (2022). Decreased resting state inter-and intra-network functional connectivity is associated with perceived stress in a sample of university students: an eLORETA study. *Neuropsychobiology* 81, 286–295. doi: 10.1159/000521565

Massullo, C., Carbone, G. A., Farina, B., Panno, A., Capriotti, C., Giacchini, M., et al. (2020). Dysregulated brain salience within a triple network model in high trait anxiety individuals: a pilot EEG functional connectivity study. *Int. J. Psychophysiol.* 157, 61–69. doi: 10.1016/j.ijpsycho.2020.09.002

Mazza, M., Losurdo, A., Testani, E., Marano, G., Di Nicola, M., Dittoni, S., et al. (2014). Polysomnographic findings in a cohort of chronic insomnia patients with benzodiazepine abuse. *J. Clin. Sleep Med.* 10, 35–42. doi: 10.5664/jcsm.3354

Medin, D., Ojalehto, B., Marin, A., and Bang, M. (2017). Systems of (non-) diversity. *Nat. Hum. Behav.* 1:0088. doi: 10.1038/s41562-017-0088

Miljevic, A., Bailey, N. W., Vila-Rodriguez, F., Herring, S. E., and Fitzgerald, P. B. (2022). Electroencephalographic connectivity: a fundamental guide and checklist for optimal study design and evaluation. *Biol. Psychiatry Cogn. Neurosci. Neuroimag.* 7, 546–554. doi: 10.1016/j.bpsc.2021.10.017

Morawetz, C., Riedel, M. C., Salo, T., Berboth, S., Eickhoff, S. B., Laird, A. R., et al. (2020). Multiple large-scale neural networks underlying emotion regulation. *Neurosci. Biobehav. Rev.* 116, 382–395. doi: 10.1016/j.neubiorev.2020.07.001

Morris, S. B. (2008). Estimating effect sizes from pretest-posttest-control group designs. Organ. Res. Methods 11, 364–386. doi: 10.1177/1094428106291059

Mulert, C., Jager, L., Schmitt, R., Bussfeld, P., Pogarell, O., Moller, H. J., et al. (2004). Integration of fMRI and simultaneous EEG: towards a comprehensive understanding of localization and time-course of brain activity in target detection. *NeuroImage* 22, 83–94. doi: 10.1016/j.neuroimage.2003.10.051

Mygind, L., Kjeldsted, E., Hartmeyer, R., Mygind, E., Stevenson, M. P., Quintana, D. S., et al. (2021). Effects of public green space on acute psychophysiological stress response: a systematic review and meta-analysis of the experimental and quasi-experimental evidence. *Environ. Behav.* 53, 184–226. doi: 10.1177/0013916519873376

Nichols, T. E., and Holmes, A. P. (2002). Nonparametric permutation tests for functional neuroimaging: a primer with examples. *Hum. Brain Mapp.* 15, 1–25. doi: 10.1002/hbm.1058

Norwood, M. F., Lakhani, A., Maujean, A., Zeeman, H., Creux, O., and Kendall, E. (2019). Brain activity, underlying mood and the environment: a systematic review. *J. Environ. Psychol.* 65:101321. doi: 10.1016/j.jenvp.2019.101321

Olbrich, S., Tränkner, A., Chittka, T., Hegerl, U., and Schönknecht, P. (2014). Functional connectivity in major depression: increased phase synchronization between frontal cortical EEG-source estimates. *Psychiatry Res.* 222, 91–99. doi: 10.1016/j. pscychresns.2014.02.010

Olszewska-Guizzo, A., Fogel, A., Escoffier, N., Sia, A., Nakazawa, K., Kumagai, A., et al. (2022a). Therapeutic garden with contemplative features induces desirable changes in mood and brain activity in depressed adults. *Front. Psychol.* 13:757056. doi: 10.3389/fpsyt.2022.757056

Olszewska-Guizzo, A., Sia, A., Fogel, A., and Ho, R. (2022b). Features of urban green spaces associated with positive emotions, mindfulness and relaxation. *Sci. Rep.* 12:20695. doi: 10.1038/s41598-022-24637-0

Palomero-Gallagher, N., and Amunts, K. (2022). A short review on emotion processing: a lateralized network of neuronal networks. *Brain Struct. Funct.* 227, 673–684. doi: 10.1007/s00429-021-02331-7

Pasca, L., Carrus, G., Loureiro, A., Navarro, O., Panno, A., Tapia Follen, C., et al. (2022). Connectedness and well-being in simulated nature. *Appl. Psychol. Health Well Being* 14, 397–412. doi: 10.1111/aphw.12309

Pascual-Marqui, R. D., Lehmann, D., Koukkou, M., Kochi, K., Anderer, P., Saletu, B., et al. (2011). Assessing interactions in the brain with exact low-resolution electromagnetic tomography. *Philos. Trans. A. Math. Phys. Eng. Sci.* 369, 3768–3784. doi: 10.1098/rsta.2011.0081

Pattyn, T., Vanneste, S., De Ridder, D., Van Rompaey, V., Veltman, D. J., Van De Heyning, P., et al. (2018). Differential electrophysiological correlates of panic disorder in non-pulsatile tinnitus. *J. Psychosom. Res.* 109, 57–62. doi: 10.1016/j. jpsychores.2018.03.168

Perrin, F., Pernier, J., Bertrand, O., and Echallier, J. F. (1989). Spherical splines for scalp potential and current density mapping. *Electroencephalogr. Clin. Neurophysiol.* 72, 184–187. doi: 10.1016/0013-4694(89)90180-6

Pizzagalli, D. A. (2011). Frontocingulate dysfunction in depression: toward biomarkers of treatment response. *Neuropsychopharmacology* 36, 183–206. doi: 10.1038/npp.2010.166

Raichle, M. E. (2011). The restless brain. Brain. Connect 1, 3-12. doi: 10.1089/ brain.2011.0019

Rolls, E. T. (2019). The cingulate cortex and limbic systems for emotion, action, and memory. *Brain Struct. Funct.* 224, 3001–3018. doi: 10.1007/s00429-019-01945-2

Ryan, R. M., and Frederick, C. (1997). On energy, personality, and health: subjective vitality as a dynamic reflection of well-being. *J. Pers.* 65, 529–565. doi: 10.1111/j.1467-6494.1997.tb00326.x

Saeed, S. M. U., Anwar, S. M., Khalid, H., Majid, M., and Bagci, A. U. (2020). EEG based classification of long-term stress using psychological labeling. *Sensors* 20:1886. doi: 10.3390/s20071886

Sahni, P., and Kumar, J. (2020). Effect of nature experience on Fronto-parietal correlates of neurocognitive processes involved in directed attention: an ERP study. *Ann. Neurosci.* 27, 136–147. doi: 10.1177/0972753121990143

Senn, S. (2006). Change from baseline and analysis of covariance revisited. *Stat. Med.* 25, 4334–4344. doi: 10.1002/sim.2682

Shanahan, D. F., Bush, R., Gaston, K. J., Lin, B. B., Dean, J., Barber, E., et al. (2016). Health benefits from nature experiences depend on dose. *Sci. Rep.* 6:28551. doi: 10.1038/ srep28551

Shao, R., Lau, W. K. W., Leung, M. K., and Lee, T. M. C. (2018). Subgenual anterior cingulate-insula resting-state connectivity as a neural correlate to trait and state stress resilience. *Brain Cogn.* 124, 73–81. doi: 10.1016/j.bandc.2018.05.001

Simmons, W. K., Avery, J. A., Barcalow, J. C., Bodurka, J., Drevets, W. C., and Bellgowan, P. (2013). Keeping the body in mind: insula functional organization and functional connectivity integrate interoceptive, exteroceptive, and emotional awareness. *Hum. Brain Mapp.* 34, 2944–2958. doi: 10.1002/hbm.22113

Spano, G., Theodorou, A., Reese, G., Carrus, G., Sanesi, G., and Panno, A. (2022). Virtual nature and psychological outcomes: a systematic review. OSF [Preprint]. doi: 10.31219/osf.io/8ux9a

Srinivasan, R., Winter, W. R., Ding, J., and Nunez, P. L. (2007). EEG and MEG coherence: measures of functional connectivity at distinct spatial scales of neocortical dynamics. *J. Neurosci. Methods* 166, 41–52. doi: 10.1016/j.jneumeth.2007.06.026

Stankovic, M. (2021). A conceptual critique of brain lateralization models in emotional face perception: toward a hemispheric functional-equivalence (HFE) model. *Int. J. Psychophysiol.* 160, 57–70. doi: 10.1016/j.ijpsycho.2020.11.001

Terraciano, A., Mccrae, R. R., and Costa, P. T. J. (2003). Factorial and construct validity of the Italian positive and negative affect schedule. *Eur. J. Psychol. Assess.* 19, 131–141. doi: 10.1027//1015-5759.19.2.131

Thatcher, R. W. (2010). Validity and reliability of quantitative electroencephalography (qEEG). *J. Neurother.* 14, 122–152. doi: 10.1080/10874201003773500

Theodorou, A., Romano, L., Bratman, G. N., Carbone, G. A., Rodelli, R., Casagrande, G., et al. (2023a). Different types of virtual natural environments enhance subjective vitality through restorativeness. *J. Environ. Psychol.* 87:101981. doi: 10.1016/j. jenvp.2023.101981

Theodorou, A., Spano, G., Bratman, G. N., Monneron, K., Sanesi, G., Carrus, G., et al. (2023b). Emotion regulation and virtual nature: cognitive reappraisal as an individuallevel moderator for impacts on subjective vitality. *Sci. Rep.* 13:5028. doi: 10.1038/ s41598-023-30287-7

Ulrich, R. S. (1981). Natural versus urban scenes: some psychophysiological effects. *Environ. Behav.* 13, 523–556.

Ulrich, R. S. (1993). "Biophilia, biophobia and natural landscapes" in *The Biophilia Hypothesis*. eds. S. R. Kellert and E. O. Wilson (Washington, DC: Island Press), 73–137.

Ulrich, R. S., Simons, R. F., Losito, B. D., Fiorito, E., Miles, M. A., and Zelson, M. (1991). Stress recovery during exposure to natural and urban environments. *J. Environ. Psychol.* 11, 201–230. doi: 10.1016/S0272-4944(05)80184-7

Veale, J. F. (2014). Edinburgh handedness inventory-short form: a revised version based on confirmatory factor analysis. *Laterality* 19, 164–177. doi: 10.1080/1357650X.2013.783045

Watson, D., Clark, L. A., and Tellegen, A. (1988). Development and validation of brief measures of positive and negative affect: the PANAS scales. *J. Pers. Soc. Psychol.* 54, 1063–1070. doi: 10.1037//0022-3514.54.6.1063

Whitton, A. E., Deccy, S., Ironside, M. L., Kumar, P., Beltzer, M., and Pizzagalli, D. A. (2018). Electroencephalography source functional connectivity reveals abnormal high-frequency communication among large-scale functional networks in depression. *Biol. Psychiatry Cogn. Neurosci. Neuroimag.* 3, 50–58. doi: 10.1016/j.bpsc.2017.07.001

Winkler, I., Brandl, S., Horn, F., Waldburger, E., Allefeld, C., and Tangermann, M. (2014). Robust artifactual independent component classification for BCI practitioners. *J. Neural Eng.* 11:035013. doi: 10.1088/1741-2560/11/3/035013

Winkler, I., Haufe, S., and Tangermann, M. (2011). Automatic classification of artifactual ICA-components for artifact removal in EEG signals. *Behav. Brain Funct.* 7:30. doi: 10.1186/1744-9081-7-30

Yao, W., Zhang, X., and Gong, Q. (2021). The effect of exposure to the natural environment on stress reduction: a meta-analysis. *Urban For. Urban Green.* 57:126932. doi: 10.1016/j.ufug.2020.126932

Zotev, V., and Bodurka, J. (2020). Effects of simultaneous real-time fMRI and EEG neurofeedback in major depressive disorder evaluated with brain electromagnetic tomography. *Neuroimag. Clin.* 28:102459. doi: 10.1016/j.nicl.2020.102459