

Since January 2020 Elsevier has created a COVID-19 resource centre with free information in English and Mandarin on the novel coronavirus COVID-19. The COVID-19 resource centre is hosted on Elsevier Connect, the company's public news and information website.

Elsevier hereby grants permission to make all its COVID-19-related research that is available on the COVID-19 resource centre - including this research content - immediately available in PubMed Central and other publicly funded repositories, such as the WHO COVID database with rights for unrestricted research re-use and analyses in any form or by any means with acknowledgement of the original source. These permissions are granted for free by Elsevier for as long as the COVID-19 resource centre remains active. Clinical Neurophysiology 140 (2022) 126-135



Contents lists available at ScienceDirect

# **Clinical Neurophysiology**



journal homepage: www.elsevier.com/locate/clinph

# High-density EEG sleep correlates of cognitive and affective impairment at 12-month follow-up after COVID-19



Maria Rubega<sup>a</sup>, Luciana Ciringione<sup>a</sup>, Margherita Bertuccelli<sup>a,b</sup>, Matilde Paramento<sup>c</sup>, Giovanni Sparacino<sup>c</sup>, Andrea Vianello<sup>d</sup>, Stefano Masiero<sup>a,b</sup>, Antonino Vallesi<sup>a,b</sup>, Emanuela Formaggio<sup>a,b</sup>, Alessandra Del Felice<sup>a,b,\*</sup>

<sup>a</sup> Department of Neuroscience, Section of Rehabilitation, University of Padova, via Giustiniani, 3, Padova 35128, Italy

<sup>b</sup> Padova Neuroscience Center, University of Padova, via Orus 2/B, Padova 35129, Italy

<sup>c</sup> Department of Information Engineering, University of Padova, via Gradenigo 6/B, Padova 35131, Italy

<sup>d</sup> Department of Cardio-Thoraco-Vascular Sciences and Public Health, University of Padova, via Giustiniani, 2, Padova 35128, Italy

#### ARTICLE INFO

Article history: Accepted 31 May 2022 Available online 15 June 2022

Keywords: Electroencephalography Nap Neuropsychological assessment Iong COVID-19

# HIGHLIGHTS

- Sleep spindles may be a biomarker of psycho-affective symptoms in long COVID-19.
- Survivors perceive a decreased quality of life and post-traumatic stress symptoms .
- Post-traumatic stress symptoms may crucially contribute to cognitive impairment.

# ABSTRACT

*Objective:* To disentangle the pathophysiology of cognitive/affective impairment in Coronavirus Disease-2019 (COVID-19), we studied long-term cognitive and affective sequelae and sleep high-density electroencephalography (EEG) at 12-month follow-up in people with a previous hospital admission for acute COVID-19.

*Methods:* People discharged from an intensive care unit (ICU) and a sub-intensive ward (nonICU) between March and May 2020 were contacted between March and June 2021. Participants underwent cognitive, psychological, and sleep assessment. High-density EEG recording was acquired during a nap. Slow and fast spindles density/amplitude/frequency and source reconstruction in brain gray matter were extracted. The relationship between psychological and cognitive findings was explored with Pearson correlation.

*Results:* We enrolled 33 participants (17 nonICU) and 12 controls. We observed a lower Physical Quality of Life index, higher post-traumatic stress disorder (PTSD) score, and a worse executive function performance in nonICU participants. Higher PTSD and Beck Depression Inventory scores correlated with lower executive performance. The same group showed a reorganization of spindle cortical generators.

*Conclusions:* Our results show executive and psycho-affective deficits and spindle alterations in COVID-19 survivors – especially in nonICU participants – after 12 months from discharge.

*Significance:* These findings may be suggestive of a crucial contribution of stress experienced during hospital admission on long-term cognitive functioning.

© 2022 International Federation of Clinical Neurophysiology. Published by Elsevier B.V. All rights

reserved.

\* Corresponding author at: Department of Neuroscience, University of Padova, Via Giustiniani, 3, 35128 Padova, Italy.

*E-mail addresses*: maria.rubega@unipd.it (M. Rubega), luciana.ciringione@unipd.it (L. Ciringione), margherita.bertuccelli@phd.unipd.it (M. Bertuccelli), matilde. paramento@studenti.unipd.it (M. Paramento), gianni@dei.unipd.it (G. Sparacino), andrea.vianello.1@unipd.it (A. Vianello), stef.masiero@unipd.it (S. Masiero), antonino. vallesi@unipd.it (A. Vallesi), emanuela.formaggio@unipd.it (E. Formaggio), alessandra.delfelice@unipd.it (A. Del Felice).

*Abbreviations:* MEG, Magnetoencephalography; EEG, Electroencefalography; PTSD, Post Traumatic Stress Disorder; BDI, Beck Depression Inventory; COVID-19, Coronavirus Disease 19; ICU, Intensive Care Unit; CTRL, Controls; MoCA, Montreal Cognitive Assessment; OSAS, Obstructive Sleep Apnea Syndrome; FAB, Frontal Assessment Battery; DigitFW, Digit Span forward; DigitBW, Digit Span backward; TMT-A, Trail Making Test - version A; TMT-B, Trail Making Test - version B; PCS-12, physical component summary scores; MCS-12, mental component summary scores; PSQI, Pittsburgh Sleep Quality Index; EHI, Edinburgh Handedness Inventory; NREM, Non-Rapid-Eye-Movement; REM, Rapid-Eye-Movement; SWS, Slow Wave Sleep; AASM, American Academy of Sleep Medicine; Stroop\_e, Stroop task, errors interference (Stroop\_e); Stroop\_i, Stroop task, time interference (Stroop\_i); LORETA, Low resolution electromagnetic tomography.

# 1. Introduction

Since the beginning of the severe acute respiratory syndromerelated coronavirus-2 (SARS-CoV-2) pandemic, almost 400 million cases of infection have been reported world-wide, with 5.5 million associated deaths (https://covid19.who.int/). Beyond symptoms of the acute disease, which include respiratory tract involvement, gastrointestinal symptoms, anosmia and dysgeusia (Docherty et al., 2020; Petrilli et al., 2019) and neurological signs (Mao et al., 2020; Helms et al., 2020), a still not fully recognized syndrome persisting after 12 weeks from hospital discharge named long or post Coronavirus disease 2019 (COVID-19) has been described (Shah et al., 2021; Nalbandian et al., 2021). Among long COVID-19 symptoms, fatigue, post-exertion malaise, and attention, working memory and executive dysfunctions are most commonly reported (Ziauddeen et al., 2021; Graham et al., 2021; Mao et al., 2020; Helms et al., 2020). COVID-19 follow-up studies have described in up to 40% of COVID-19 survivors concurrent affective disturbances, including depression, post-traumatic stress disorders, anxiety, and non-restorative sleep (Nalbandian et al., 2021).

Consensus is still missing on the real clinical entity of long COVID-19. The Post-hospitalisation COVID-19 (PHOSP-COVID) study described 4 symptom clusters, with 3 clusters characterized by mental-health and physical impairments to varying degrees, but few or no cognitive difficulties and a fourth with minimal mentalhealth and physical impairments, but severe cognitive problems (Evans et al., 2021). Recently, the lack of a clear-cut correlation between cognitive complaints and performance, as well as a relation of cognitive performance with psychological distress, has further questioned the real nature of the reported cognitive component of long COVID-19 (Whiteside et al., 2022).

A biomarker of cognitive dysfunction may help in shedding light on these open issues. Neurophysiology may be a tool in this quest. We know that specific cognitive functions are encoded by neurophysiological signals detected by scalp electroencephalographic (EEG) recordings. Among these neural activities, sleep spindles - i.e., bursts of neural oscillatory activity generated in thalamocortical systems during stage 2 Non-rapid eye movement (NREM) sleep (Steriade, 2005) - are considered the scalp equivalent of hippocampal ripples, which encode memory and are markers of neural plasticity (LaFortune et al., 2014). Sleep spindles consist of brief bursts ([0.5, 2] s) of rhythmic activity distinguished in slow spindles ([8, 12] Hz), with a prevalent frontal distribution, and fast spindles ([13, 15] Hz), with a more centro-parietal distribution. Fast and slow frequency types (Cox et al., 2017) display a different coupling to sleep slow oscillations (Mölle et al., 2011: Cox et al., 2014): in the temporal domain, fast spindles are timelocked to the depolarizing up-state of slow oscillations, whereas slow spindles to the up-to-down-state transition of the slow oscillations (Klinzing et al., 2016). These differences in coupling behavior may translate to different functional roles for fast and slow spindle types (Schabus et al., 2006; Cowan et al., 2020; Fernandez and Lüthi, 2020), with fast spindles apparently more involved in motor memory encoding (Tamaki et al., 2009) and slow spindles in declarative memory (Lustenberger et al., 2015). In fact, a definite role of these neurophysiological correlates is far from being agreed upon. Other metrics, such as spindle density, relate to verbal learning, visual attention, and verbal fluency performance in healthy middle-age and older individuals (LaFortune et al., 2014). An impairment of spindle density and localization is also reported in psycho-affective disturbances (Shao et al., 2022). In major depression, sleep spindles have an increased density and distribution over frontal and parietal cortices (Plante et al., 2013). In post-traumatic stress disorder (PTSD), fast sleep spindles apparently increase peak frequencies (Denis et al., 2021). The study of cortical generators of neural activity has further extended our knowledge of the neurophysiology of cognition and sleep. Several Magnetoencephalography (MEG) and EEG studies investigated the cortical distribution of spindles, suggesting that two distinct forms of spindles (i.e., slow and fast spindles) propagate to the cortex through different underlying neuronal circuits (Urakami, 2008; Manshanden et al., 2002; Ishii et al., 2003; Dehghani et al., 2010) with different functional significance (Del Felice et al., 2013; Del Felice et al., 2014). Executive and memory functions, reportedly affected in long COVID-19, are processed in neural networks that encompass the prefrontal cortex, but also engage the parietal cortex (Rabinovici et al., 2015), anatomical areas that are part of the cortico-thalamo-cortical circuit of spindle cortical generators.

Our hypothesis is that features of sleep spindles may provide a biomarker of physiopathology of cognitive/affective symptoms in long COVID-19. The aim of this study is to investigate sleep spindle patterns and the relation between spindles and cognitive/affective symptoms in people with previous COVID-19 infection at 12month follow-up after hospital discharge.

#### 2. Methods

#### 2.1. Experimental protocol

People discharged from the Intensive Care Units (ICU) and medical wards (nonICU) of the Teaching Hospital in Padova from March 2020 to May 2020 were contacted by phone between March and June 2021 by a researcher presenting the study. We also enrolled age and sex-matched healthy controls (CTRL). Exclusion criteria were: age < 18 years, previous diagnosis of cognitive impairment, previous diagnosis of neurological disorder, on drugs altering sleep architecture (e.g., benzodiazepines) or with a known diagnosis of Obstructive Sleep Apneas (OSAS).

Participants were asked to wake up at 5 AM and to avoid the intake of stimulant substances (e.g., coffee, coke, tea). The experimental procedure was divided in two stages: 1) the neuropsychological assessment administered by a trained neuropsychologist and 2) the nap EEG recording. Nap sleep recording was performed at about 1.30 PM. The time of sleep was either at least 60 min of sleep or 1 h and 30 min of time in bed, whichever came first.

The local Ethics Committee approved the study at Padova (4932/AO/20). It was conducted in compliance with the Declaration of Helsinki, the Good Clinical Practice guidelines, and local regulatory requirements. Informed consent was obtained from all participants.

#### 2.2. Neuropsychological assessment

#### 2.2.1. Neuropsychological evaluation

Neuropsychological assessment was performed testing general and specific cognitive domains. Scores were adjusted for age, sex, and education (when applicable). The Montreal Cognitive Assessment (MoCA) (Nasreddine et al., 2005; Santangelo et al., 2015) was performed as a measure of global cognitive assessment; a score up to 26 indicates poor overall cognitive functioning.

The Frontal Assessment Battery (FAB) (Dubois et al., 2000) and the Stroop task (Caffarra et al., 2002) were used to assess executive functioning. A FAB score up to 13.50 indicates poor executive functioning. For the Stroop task, errors interference (Stroop\_e) and time interference (Stroop\_i) were considered.

Digit Span forward (DigitFW) and backward (DigitBW) (Monaco et al., 2013) and Rey Auditory Verbal Learning test (Carlesimo et al., 1995; Caltagirone et al., 1995) (immediate memory – Rey\_i; delayed recall – Rey\_d) were performed to assess short-term memory, working memory, and long-term memory. The Trail Making Test (version A – TMT-A – and B – TMT-B) (Giovagnoli et al., 1996) and Digit Symbol Modalities Test (Symbol) (Nocentini et al., 2006) assessed focused selective and divided attention.

#### 2.3. Psychological evaluation

The Beck Depression Inventory (BDI) (Beck et al., 1996) was administered to assess depressive symptoms; the COVID-19 post-traumatic stress disorder (PTSD) scale to assess PTSD symptoms related to COVID-19 (Forte et al., 2020); a 12-item short-form health survey to assess the quality of life – i.e., physical and mental component summary scores, PCS-12 and MCS-12 (Pezzilli et al., 2006).

## 2.3.0.1. Subjective sleep assessment

The Epworth Sleepiness Scale (Vignatelli et al., 2003) was performed to test subjective sleepiness; Pittsburgh Sleep Quality Index (PSQI) (Curcio et al., 2013) was used to measure the major components of subjective sleep – i.e., quality, latency, duration, efficiency, disturbance, medication, daytime dysfunction.

Eventually, participants' handedness dominance in everyday activities was defined through Edinburgh Handedness Inventory (EHI) – i.e., a 10-item self-response inventory.

#### 2.4. Nap EEG assessment

#### 2.4.1. EEG recording

High-density EEG recordings were acquired at the Padova Neuroscience Center inside a dimly lit sound-attenuated and electrically shielded room using the Geodesic Sensor Net system with 256 electrodes (Electrical Geodesic, Inc, Eugene, OR). Electrodeskin impedances were maintained <  $50k\Omega$ . The recordings were sampled at 250 Hz, referenced to Cz. As a measure of electrooculogram activity, electrodes on the left and right eye canthus were considered (electrodes 10 and 248 referring to Sensor layout for 256-channel Hydrocel Geodesic Sensor Net). As a measure of muscular activity, electrodes in the area of the masseter were selected (electrodes 231, 249).

# 2.4.2. EEG pre-processing

EEG data were zero-phase-high-pass-filtered above 1 Hz through a 4th order Butterworth filter and then zero-phase-low-pass-filtered below 40 Hz through a 4th order Butterworth filter avoiding phase distortion. Channels in the cheeks and in the neck were discarded (204 channels left). Noisy channels were identified by visual inspection and interpolated using the nearest-neighbor spline method (average percentage of channels interpolated: 2%). Individual epochs containing non-stereotyped artifacts, eye movements were also identified by visual inspection and removed from further analysis (average percentage of epochs removed: 1%).

#### 2.4.3. Sleep scoring and spindles detection

Sleep scoring (Wake; Non-Rapid-Eye-Movement - NREM Sleep, Stage 1 – N1; NREM Sleep, Stage 2 – N2; Slow-Wave Sleep – SWS; and Rapid-Eye-Movement – REM) was visually performed on 30-s EEG epochs from C3 in line with American Academy of Sleep Medicine (AASM) guidelines (Berry et al., 2012).

During NREM sleep, sleep spindles were automatically scored according to AASM guidelines on monopolar montage, considering frontal, central, and occipital leads (F3, C3, O1, F4, C4, O2), referenced to contralateral mastoids (93 as A1, 201 as A2). Then, frontal (F3 and F4) and central (C3 and C4) sleep spindles were automatically detected using an algorithm implemented by Wonambi 6.12 (https://wonambi-python.github.io/) (Lacourse et al., 2019) – Fig. 1a. This method detects spindles using a combination of 4 parameters related to sigma power (we refer the reader to

(Lacourse et al., 2019) for more details). Based on the peak frequency, we divided sleep spindles into slow ([9-12] Hz) and fast spindles ([12-16] Hz) – Fig. 1b. We computed the density (i.e., number of spindles per 60-s epoch), duration (s), peak frequency (Hz), amplitude/topography (mV) of the spindles.

#### 2.4.4. Spindles localization

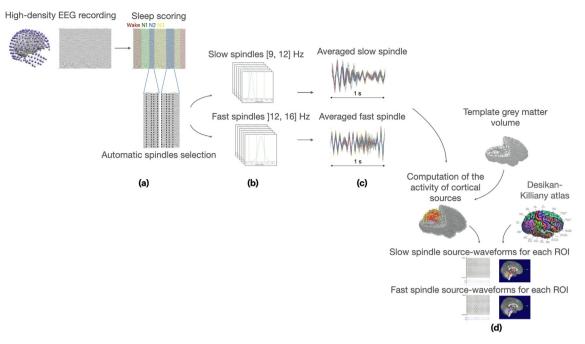
Slow and fast spindles containing non-stereotyped artifacts or muscle activity were detected and excluded from the following analysis - Fig. 1c-d. The remaining slow and fast spindles were averaged to obtain the two inputs for the source reconstruction i.e., one time-series representing the averaging of slow spindles and one time-series representing the averaging of fast spindles for each participant – Fig. 1c. We applied the LORETA algorithm implemented in Cartool [cartoolcommunity.unige.ch] to compute the source reconstruction of the averaged spindles (Del Felice et al., 2014) taking into account the participant's age to calibrate the skull conductivity (Michel and Brunet, 2019) - Fig. 1d. The method restricts the solution space to the gray matter of the brain. The grav matter was then parcellated using the 83 cortical regions of the Desikan-Killiany (DK) atlas (Desikan et al., 2006). The dipoles in each region-of-interest (ROI) were represented with one unique time-series by applying a singular-value decomposition (Rubega et al., 2019; Glomb et al., 2020a, 2020b; Pascucci et al., 2020). The amplitude at the onset of the source-waveforms representing the sleep spindles were extracted and compared between the three groups (i.e., CTRL, nonICU participants and ICU participants).

#### 2.5. Statistical analyses

After testing data distribution with a Shapiro-Wilks normality test, either Kruskal-Wallis test or ANOVA were performed among the three groups (CTRL, ICU, nonICU). Subsequently, based on the test results, either a two-sided Wilcoxon rank sum test or a twosample t-test was performed on: 1) cognitive and psychological tests scores; 2) slow and fast spindle density/frequency/amplitude; 3) spindle onset amplitude values for each ROI; comparing CTRL vs ICU, CTRL vs nonICU, and ICU vs nonICU to test for significant differences. Considering the sample size and the exploratory nature of the study, no corrections for multiple comparisons were performed in hypothesis testing. We assessed the relationship between psychological and cognitive results through Pearson correlation coefficient (*r*). *r* measured the linear correlation between each combination of neuropsychological tests scores ( $r \in [-1, 1]$ ). Then, multiple linear regression modeled the linear relationship between the explanatory variables and response variables. We used as explanatory variables the psychological scores and as response variables the cognitive ones and vice versa.

#### 3. Results

Ninety-one COVID-19 survivors were contacted. Twelve were unreachable, 17 could not participate because of severe disabling long COVID-19 physical symptoms (mainly shortness of breath, myalgia and fatigue), 17 refused to participate because of fear of contagion and safety concerns, 3 could not be recruited for sleep disorder comorbidities (i.e., insomnia, obstructive sleep apnoea syndrome) or for taking medications that alter sleep architecture (e.g., benzodiazepines, antidepressants), and 9 dropped out. Eventually, 33 COVID-19 participants (27% female (F), F age interval: [50, 69] y, male (M) age interval: [49, 80] y) and 12 age and sexmatched control participants (33% F, F: [50, 63] y, M: [49, 76] y) were enrolled. The COVID-19 cohort consisted of Intensive Care Unit (ICU) discharged participants (N = 16, 25% F, F: [64, 69] y,



**Fig. 1. Overview of the analysis pipeline.** (a) Electroencephalography (EEG) was recorded during a nap in COVID-19 survivors and age-sex matched controls. (b) Slow (i.e., [9–12] Hz) and fast (i.e.,]12–16] Hz) spindles were identified and (c) averaged obtaining one time-series (for each electrode) representing the slow spindles and one representing the fast spindles for each participant. (d) The cortical grey matter (e.g., considering age, sex, cortical thickness) was reconstructed and the brain parcelled from cortical regions-of-interest (ROI) of Desikan–Killiany atlas. For each participant, sleep grapho-elements cortical generators were computed with Electrical Source Imaging (ESI).

M: [55, 80] y) and participants (nonICU) admitted to a sub intensive medical ward (N = 17, 29% F, F: [53, 62] y, M: [50, 78] y). Among these participants, 7 reported persistence fatigue (4 nonICU).

All participants admitted to ICU suffered from Acute Respiratory Distress Syndrome (ARDS), requiring invasive ventilation and in most cases also sedation and curarization. This was never the case in the medical ward (i.e., nonICU participants).

#### 3.1. Neuropsychological assessment

COVID-19 survivors reported worse Physical Quality of Life compared to controls (CTRL) (median<sub>CTRL</sub> = 53.07 vs median<sub>ICU</sub> = 46.19 and median<sub>nonICU</sub> = 43.33,  $p_{value} < 0.04$ ) – Fig. 2a. Post-Traumatic Stress Disorder – Fig. 2b (CTRL vs nonICU:  $p_{value} = 0.06$ ) – and Beck Depression Inventory – Fig. 2c – scores showed a decreasing trend from the nonICU group, ICU group and CTRL. The cognitive assessment revealed a trend towards worse performance in executive functions in COVID-19 survivors, in particular in nonICU in the Trail Making Test (TMT-B, CTRL vs nonICU:  $p_{value} = 0.04 - Fig. 2d$ ; TMT-B-A, CTRL vs nonICU:  $p_{value} = 0.08 - Fig. 2e$ ).

Pearson correlation coefficient (r) reports a statistically significant linear correlation in the nonICU participants between:

• PTSD and DigitBW (r = -0.6,  $p_{value} = 0.0033$ ), TMT-B (r = 0.59,  $p_{value} = 0.01185$ ), TMT B-A (r = 0.56,  $p_{value} = 0.01185$ ) – i.e., a higher likelihood of PTSD is correlated to a worse performance in DigitBW (i.e., lower number of digits that the subject repeated correctly) and TMT-B (i.e., more time needed to complete the task and higher number of errors),

• BDI and MoCA (r = -0.55,  $p_{value} = 0.02$ ) – i.e., a higher score in BDI (i.e., higher risk of depression) is correlated to a lower score in MoCA (i.e., higher risk of cognitive impairment), Fig. 2f.

In ICU participants, a negative linear correlation emerged between PTSD and PCS-12 (r = -0.53,  $p_{value} = 0.03467$ ) – i.e.,

higher likelihood of PTSD is correlated to a worse perception of physical health, Fig. 2g.

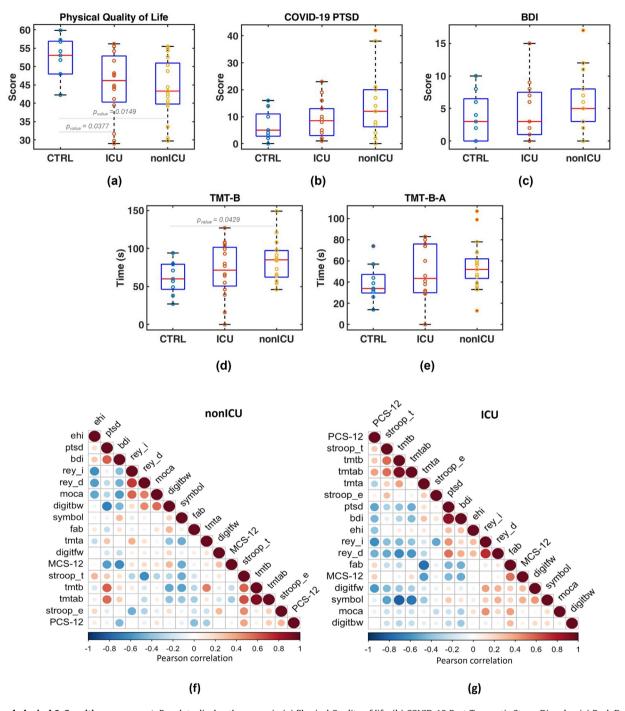
Multiple linear regression analyses highlighted that nonICU lower scores in cognitive tasks evaluating executive function and working memory can be predicted by psychological scores (i.e., PTSD and BDI) – Table 1.

See Supplementary Material (*Supplementary\_material\_neuropsy. pdf*) for all the neuropsychological test scores.

## 3.2. Nap EEG assessment

Two ICU and seven nonICU participants did not fall asleep during the recording. Results report EEG data from 14/16 (88%) ICU, 10/17 (59%) nonICU and 12/12 (100%) CTRL.

Slow and fast spindle density did not differ among the three groups. Slow spindles frequency was decreased in ICU participants compared to controls (pvalue = 0.0477) - Fig. 3i. Descriptively, in CTRL, slow and fast spindles showed different EEG topographical distributions - Fig. 3 - with slow spindles more frontal and fast spindles more centro-parietal. In COVID-19 survivors the physiological sleep spindles distribution is not clear-cut. Slow spindles shift to more posterior regions (central in nonICU - Fig. 3b - and temporal in ICU participants - Fig. 3c); fast spindles tend to become more anterior in ICU participants - Fig. 3f. In CTRL, slow spindle onsets were generated in the frontal cortices - Fig. 4a and fast spindle onsets in the central brain regions - Fig. 4d. In COVID-19 survivors, slow spindle onsets shifted to centraltemporal regions in nonICU - Fig. 4b - and ICU participants -Fig. 4c -, whereas fast spindle onset shifted to tempo-parietal brain regions in nonICU - Fig. 4e - and to frontal brain regions in ICU -Fig. 4f. Indeed, we observed statistically significant differences in slow and fast EEG source-waveforms spindle amplitude onset among the three groups. All statistically significant amplitude differences are reported in Fig. 4g, h, i, l, m, n. The differences in fast spindle sources between CTRL and nonICU is worthy of note -



**Fig. 2. Psychological & Cognitive assessment.** Boxplots display the scores in (a) Physical Quality of life, (b) COVID-19 Post-Traumatic Stress Disorder, (c) Beck Depression Inventory; (d) Trail Making Test-B and (e) the difference between Trail Making Test-B and A performance. Blue circles stand for controls (CTRL) scores, orange ones for intensive care unit (ICU) participants and yellow ones for participants admitted to sub-intensive ward (nonICU). Physical quality of Life resulted significantly lower in nonICU vs CTRL (p<sub>value</sub> = 0.0149), in nonICU vs ICU (p<sub>value</sub> = 0.0377) and in ICU vs CTRL (p<sub>value</sub> = 0.0377). Non-ICU participants scored significantly lower than CTRL (p<sub>value</sub> = 0.0429) in TMT-B. Panels (f) and (g) display Pearson Correlation Coefficient between psychological score (i.e., *Beck Depression Inventory: BDI; Post-traumatic stress disorder scale: PTSD; Physical and mental component summary scores: PCS-12 and MCS-12*); cognitive performance (i.e., *Montreal Cognitive Assessment: MoCA; Frontal Assessment Batery: FAB; Stroop task: Stroop\_t; Stroop\_e; Digit Span forward and backward: DigitFW, DigitBW; Rey Auditory Verbal Learning test: Rey\_i, Rey\_d; Trail Making Task: TMT A; TMT B, TMT B-A; Digit Symbol Modalities Test: Symbol) and participants' handedness dominance (Edinburgh Handedness Inventory: EHI). Circle dimension and color identify the value of the Pearson correlation coefficient, i.e., largest dark red circle indicates maximum positive correlation (1) and largest dark blue circle represents maximum negative correlation (-1) between variables in the two axes.* 

#### M. Rubega, L. Ciringione, M. Bertuccelli et al.

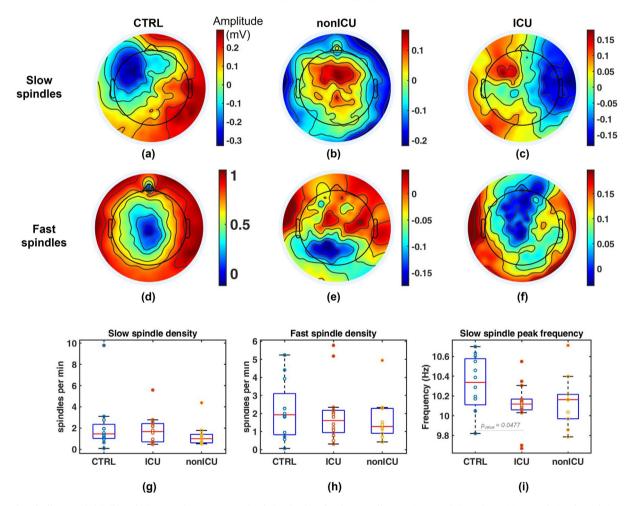
#### Table 1

Multiple linear regression in nonICU participants' scores.

Predictor	Predicted variable	<i>Pr</i> (>t)*	Adjusted r <sup>2**</sup>	<b>F-statistics</b>
PTSD	DigitBW	0.02998	0.3363	$3.703 \ (p_{value} = 0.03998)$
PTSD	TMT-B	0.02	0.2536	$2.812 (p_{value} = 0.08088)$
BDI	MoCa	0.0304	0.2248	$2.546 (p_{value} = 0.1012)$

\* Pr(>t) stands for the probability of the value of the t-distribution; \*\*r is the Pearson correlation coefficient.

Post-traumatic stress disorder: PTSD; Beck Depression Inventory: BDI; Digit Span backward: DigitBW; The Trail Making Test (version B): TMT-B; Montreal Cognitive Assessment: MoCA.



Spindle topography

**Fig. 3. Scalp spindles.** Panels (a), (b) and (c) report the average amplitude (mV) values for slow spindle onset in controls (CTRL), participants admitted to sub-intensive ward (nonlCU) and intensive care unit (ICU), respectively, visualized as a topography. Panels (d), (e) and (f) report the average values for fast spindle onset in CTRL, nonICU and ICU, respectively, visualized as a topography. Boxplots display (g) slow spindle density (h) fast spindle density and (i) slow spindle peak frequency.

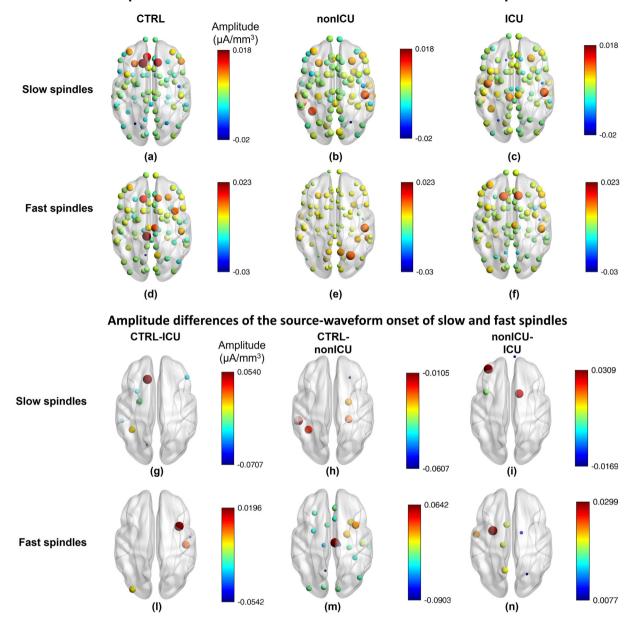
Fig. 4m. Frontal and tempo-parietal brain regions have higher amplitude in nonICU than in CTRL at fast spindle onset.

See Supplementary Material (*Supplementary\_material\_EEGspin dles.pdf*) for the EEG results not discussed in the manuscript because of the lack of statistically significant differences among groups.

# 4. Discussion

We report the presence of executive performance impairments, reduced physical quality of life, and symptoms of PTSD at 12 months after discharge in people previously admitted to hospital for COVID-19 infection. The neurophysiological correlate of these findings – for the first time, investigated through highdensity EEG – is a modification of cortical generators of sleep spindles, with generators of slow spindles shifting towards more posterior cortical regions and fast spindle generators shifting anteriorly in COVID-19 participants compared to CTRL. These phenomena are more evident in people discharged from medical wards than to ICUs. The identification of neurophysiological abnormalities – i.e., sleep spindles cortical generators – may provide additional information on the still non fully elucidated clinical picture of long COVID-19.

Reports on effects on cognition after COVID-19 infection emerged in the early days of the pandemic, although a clear-cut definition of these symptoms has not yet been defined, with the



# Amplitude of the source-waveform onset of slow and fast spindles

**Fig. 4. Amplitude of the source-waveform onset of slow and fast spindles.** Panels (a), (b) and (c) report the average values ( $\mu$ A/mm<sup>3</sup>) for slow spindles in controls (CTRL), participants admitted to sub-intensive ward (nonICU) and intensive care unit (ICU). Panels (d), (e) and (f) report the average values for fast spindles in CTRL, nonICU and ICU participants. Panels (g), (h) and (i) report amplitude differences of the source-waveform onset of slow spindles and panels (1), (m) and (n) of fast spindles between (g-l) nonICU and ICU; (h-m) CTRL and ICU; (i-n) CTRL and nonICU: Only regions-of-interest (ROIs) in which the differences between the two groups were statistically significant ( $p_{value} < 0.05$ ) are reported. Node dimension and color (centered in each ROI) identify the value of the spindle onset amplitude ( $\mu$ A/mm<sup>3</sup>) of the reconstructed source-waveforms. The brain networks were visualized with the BrainNet Viewer (Xia et al., 2013).

umbrella definition of *brain fog.* In fact, the PHOSP-COVID study identified only a fraction of long COVID-19 survivors with clinically relevant cognitive problems (Evans et al., 2021). The systematic inclusion of neurophysiological biomarkers in our dataset aimed at providing quantifiable data to support this research. The combination of comprehensive neuropsychological and psycho-affective testing and advanced EEG analysis aims at disentangling the physiopathology of cognitive impairment in long COVID-19.

Sleep spindles are a robust biomarker of cognition. They have widely been studied as the scalp biomarkers of hippocampal ripples encoding memory (Cairney et al., 2018), with a hierarchical frequency nestling paced by the de- and hyper-polarizing slow oscillations (Staresina et al., 2015). Slow oscillation dynamics and

sleep spindle density, topography and distribution of cortical generators have largely been related to visuo-motor (mainly fast spindles) and verbal (mainly slow spindles) memory formation (Debarnot et al., 2011). In mild cognitive impairment, sleep spindles are decreased (D'Atri et al., 2021; D'Rozario et al., 2020) or shorter (Lam et al., 2021); in memory impairment in temporal epilepsies, cortical source generators of sleep spindles shift their physiological location (Del Felice et al., 2013). Our focus on these biomarkers is based on specific changes described also in depression and PTSD. In major depression, sleep spindles increase density and shift toward frontal and parietal cortices (Plante et al., 2013; Ferrarelli et al., 2019). In PTSD, fast sleep spindles increase peak frequencies (Denis et al., 2021). Indeed, we observed a shift of fast spindles toward parietal regions in people discharged from non-ICU and towards anterior areas in people discharged from ICU. The shift of generators has already been described in association with affective, i.e., major depression, PTSD, schizophrenia among others (Wang et al., 2020; Ferrarelli et al., 2019; Plante et al., 2013, De Maertelaer et al., 1987), and neuropsychological abnormalities, i.e., memory (Del Felice et al., 2013-2015).

We observed a lower physical quality of life in COVID-19 survivors, suggesting the perception of physical symptoms persistence (Alkodaymi et al., 2022) even after 12 months from acute infection resolution. Indeed, long COVID-19 is defined as the persistence of at least a symptom for more than 12 weeks from acute infection, but long-term follow-up data are just starting to be reported, suggesting a multifaceted and more prolonged picture. PTSD scores were significantly higher in nonICU admitted participants. This finding is more striking considering the growing wealth of evidence of long-term harmful effects after ICU discharge, termed post-intensive care syndrome (PICS). PICS encompasses multi-domain dysfunctions ranging from reduced physical performance, cognitive impairments, and failed social construction (Needham et al., 2012; Hanifa et al., 2018; Torres et al., 2017). These symptoms persist for at least three months (Hanifa et al., 2018) and up to 5 years (Torres et al., 2017) from discharge. In fact, clinical practice suggests that this counterintuitive finding may be based on the higher levels of anxiety and stress experienced by conscious patients, especially during non-invasive high flow ventilation. Emerging data demonstrated that up to 30% of COVID-19 survivors experience PTSD (Greene et al., 2022): the later the diagnosis and appropriate care is put in place, the worse the outcome and prognosis. Lastly, we observed higher depression scores in our nonICU participants, although this finding remains only a non-significant trend: this subclinical symptom may well be considered a reactive phenomenon.

We clinically ruled out the presence of hypoxic encephalopathy among our participants: all of them had a negative neurological examination and were fully independent. In fact, hypoxic encephalopathy may have only occurred in severe ARDS not adequately supported in ICU, with catastrophic neurological sequelae. In fact, the role of microvascular bleedings and endothelial dysfunction has been widely reported in COVID-19 infections and is at the moment of this writing considered the leading cause of cerebral involvement. Whereas cases of hypoxic encephalopathy have been described, these are extremely rare and possibly related to a therapeutic failure (Popescu C, 2021).

Our data report an involvement of executive functions more evident in nonICU participants than ICU. On the one hand, memory impairment – a predominant symptom in the so-called brain fog clinical picture – appears at a subclinical level in our cohort. A possible explanation may be the high prevalence of PTSD: PTSD is characterized, among other symptoms, by poor cognitive flexibility, processing speed, controlled and sustained attention, higher response inhibition, and recall memory (Schultebraucks et al., 2022), all traits observed in our cohort. On the other hand, processing speed and executive/working memory are the most commonly impaired domains in depression (Varghese et al., 2022).

In contrast to previous findings, we did not observe a relation between disease severity and worse cognitive performance (Cristillo et al., 2022). Although our definition of disease severity was based on the admission ward (ICU vs medical wards) and we did not factor in inflammatory markers or other biochemical indexes, we observed higher depression and PTSD scores in people discharged from medical wards. We would have expected a combined effect of ICU admission and long COVID-19 symptoms. Indeed, in our cohort the worst cognitive and psycho-affective performance has been observed in nonICU admitted participants, suggesting a long-term persistence of non-physical symptoms. Our finding has inconsistently been replicated in previous studies on neuropsychological and neurological sequelae in long COVID-19 (Premraj et al., 2022), with a reportedly higher association of fatigue, anxiety, depression in ICU admitted persons than non-ICU participants. These data rarely refer to follow-ups longer than 6 months, with sporadic correlation of psycho-affective symptoms with cognitive impairment. Our findings suggest a strong effect of stress experienced during admission, a well-known determinant of worse cognitive performance.

We may speculate on a correlation among psycho-affective and neuropsychological impairments and sleep spindle generators by inferring from correlation between these deficits. We observed lower memory and executive function performance in subjects with higher PTSD scores; spindle generators shift may signal impaired neural circuitries regulating these functions. While we have no clear hypothesis on the direction of the shift, we observed that fast spindles, contributing to memory consolidation, may have an altered function when appearing in areas remote from physiologically occurring ones. In addition, a competitive phenomenon may take place, in which either fast or slow spindles may be generated in a given cortical region. In the specific case, anterior fast spindles are not effective as they should be and at the same time prevent memory consolidation by naturally occurring slow spindles.

To sum up, we could observe a distribution of spindles generators more in line with depressive and PTSD symptoms. A reflection of these altered distributions is an impairment of the physiological function of sleep spindles, mainly memory consolidation and for slow spindles embedded in slow waves also a decrease of executive functions. We did not observe a clear reduction of spindle density as described in cognitive impairment, suggesting a functional nature of the observed neuropsychological symptoms.

A major limitation of our study is the lack of inclusion of people with likely more severe symptoms, who refused to reach the laboratory for assessment. This inclusion bias may probably have contributed to our finding of mostly subclinical impairments, likely related to PTSD and depression. The lack of inclusion of a high number of people for fear of contagion may indirectly reflect high levels of anxiety possibly related to long COVID-19. Since recruitment and recordings were performed during the second lockdown in Italy (i.e., Veneto region shut schools and barred residents from leaving their homes except for work, health or basic needs, and among commercial activities, only supermarkets and pharmacies stayed open), this might explain the high refusal rate. Type I errors are conceptual and it is not possible to ascertain if they have occurred after statistical testing and we could not prevent type I errors from occurring by increasing the sample size (Sedgwick, 2014). Another limitation of statistical analysis is the lack of correction for multiple comparisons. Eventually, the missing information about the socioeconomic status of included participants may have affected symptoms severity. The information of state and trait anxiety status would also have supported a more comprehensive evaluation, which is strongly suggested by the observation of a lower percentage of nonICU falling asleep despite deprivation.

#### 4.1. Conclusion

To sum up, based on clinical neuropsychological and neurophysiological features in a cohort of long COVID-19 participants, we observed a persistence of PTSD symptoms and reduced physical performance, which are likely to impact on cognitive performance. The neurophysiological sleep signatures suggest a more likely psycho-affective nature of the subtle impairment in cognition. The finding that participants admitted to medical wards experience worse stress and long-term psycho-affective symptoms is of note for the development of clinical pathways that incorporate care of these issues.

# Funding

This work was supported by University of Padova crowdfunding 2020 – Emergenza COVID-19 (LC); by the Department of excellence 2018–2022 initiative of the Italian Ministry of education (MIUR) awarded to the Department of Neuroscience – University of Padova (EF, AV); by REACT EU-PON *Ricerca e Innovazione* 2014–2020, DM 1062/2021 (MR) and by the Italian Ministry of Health – *Progetto Giovani Ricercatori* (AV). ADF also received support for other research from the Italian Ministry for Foreign Affairs and International Collaborations (*Progetti Grande Rilevanza* 2019–2022) and H2020-MSCA-RISE-2017, grant number 778043.

# 6. Disclosure Statement

Financial Disclosure: none. Non-financial Disclosure: none.

# 7. Ethics approval

The experimental protocol was approved by the Regional Ethics Committee (37053/2020). The study was carried out in accordance with the relevant guidelines and regulations. This is an observational non interventional study (no need for trial registration).

# 8. Consent to participate and publication

Informed consent was obtained from all participants.

# 9. Availability of data, materials and code

Anonymous Data and Code will be made available upon reasonable request to the Corresponding Author. None of the experiments was preregistered.

# 10. Authors' contribution

ADF, MR and EF designed the project. ADF granted financial support; ADF, MR, MB and EF conceived the experiment; AV, ADF, LC and MR recruited the participants; LC, MR, EF, MB recorded the data; LC ran neuropsychological assessments and scoring and scored sleep; MR analyzed EEG data; EF and GS supervised data analysis; MR, MP, LC and MB run the statistics; MR drafted the manuscript; ADF and EF contributed to manuscript editing; SM, AV, and GS reviewed the manuscript. All authors critically discussed the results.

# **Declaration of Competing Interest**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

# Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.clinph.2022.05.017.

#### References

- Alkodaymi MS, Omrani OA, Fawzy NA, Shaar BA, Almamlouk R, Riaz M, et al. Prevalence of post-acute COVID-19 syndrome symptoms at different follow- up periods: A systematic review and meta-analysis. Clin Microbiol Infect 2022;28 (5):657–66.
- Beck AT, Steer RA, Brown GK. Beck depression inventory (BDI-II). London, UK: Pearson; 1996. p. s15327752jpa6703\_13.
- Berry RB, Brooks R, Gamaldo CE, Harding SM, Marcus C, Vaughn BV. The AASM manual for the scoring of sleep and associated events. Rules, Terminology and Technical Specifications 2012;176.
- Caffarra P, Vezzadini G, Dieci F, Zonato F, Venneri A. Una versione abbreviata del test di Stroop: dati normativi nella popolazione italiana. Riv Neurol 2002;12 (4):111–5.
- Cairney SA, Guttesen AáV, El Marj N, Staresina BP. Memory consolidation is linked to spindle-mediated information processing during sleep. Curr Biol 2018;28 (6):948–954.e4.
- Caltagirone C, Gainotti G, Carlesimo GA, Parnetti L. Batteria per la valutazione del deterioramento mentale: I. Arch psicol Neurol psichiatr: Descrizione di uno strumento di diagnosi neuropsicologica; 1995.
- Carlesimo GA, Caltagirone C, Gainotti G, Nocentini U. Batteria per la valutazione del deterioramento mentale: II. Arch psicol Neurol psichiatr: Standardizzazione e affidabilità diagnostica nell'identificazione di pazienti affetti da sindrome demenziale; 1995.
- Cowan E, Liu A, Henin S, Kothare S, Devinsky O, Davachi L. Sleep spindles promote the restructuring of memory representations in ventromedial prefrontal cortex through enhanced hippocampal–cortical functional connectivity. J Neurosci 2020;40(9):1909–19.
- Cox R, Schapiro AC, Manoach DS, Stickgold R. Individual differences in frequency and topography of slow and fast sleep spindles. Front Hum Neurosci 2017;11:433.
- Cox R, van Driel J, de Boer M, Talamini LM. Slow oscillations during sleep coordinate interregional communication in cortical networks. J Neurosci 2014;34 (50):16890–901.
- Cristillo V, Pilotto A, Cotti Piccinelli S, Bonzi G, Canale A, Gipponi S, et al. Premorbid vulnerability and disease severity impact on Long-COVID cognitive impairment. Aging Clin Exp Res 2022;34(1):257–60.
- Curcio G, Tempesta D, Scarlata S, Marzano C, Moroni F, Rossini PM, et al. Validity of the Italian version of the Pittsburgh sleep quality index (PSQI). Neurol Sci 2013;34(4):511–9.
- D'Atri A, Scarpelli S, Gorgoni M, Truglia I, Lauri G, Cordone S, et al. EEG alterations during wake and sleep in mild cognitive impairment and Alzheimer's disease. iScience 2021;24(4):102386. <u>https://doi.org/10.1016/j.isci.2021.102386</u>.
   D'Rozario AL, Chapman JL, Phillips CL, Palmer JR, Hoyos CM, Mowszowski L, et al.
- D'Rozario AL, Chapman JL, Phillips CL, Palmer JR, Hoyos CM, Mowszowski L, et al. Objective measurement of sleep in mild cognitive impairment: A systematic review and meta-analysis. Sleep Med Rev 2020;52:101308. <u>https://doi.org/ 10.1016/j.smrv.2020.101308</u>.
- De Maertelaer V, Hoffman O, Lemaire M, Mendlewicz J. Sleep spindle activity changes in patients with affective disorders. Sleep 1987;10(5):443–51.
- Debarnot U, Castellani E, Valenza G, Sebastiani L, Guillot A. Daytime naps improve motor imagery learning. CABN 2011;11(4):541–50.
- Dehghani N, Cash SS, Chen CC, Hagler DJ, Huang M, Dale AM, et al. Divergent cortical gener- ators of MEG and EEG during human sleep spindles suggested by distributed source modeling. PLoS One 2010;5(7):e11454. <u>https://doi.org/ 10.1371/journal.pone.0011454.00110.1371/journal.pone.0011454.g00310.1371/ journal.pone.0011454.g00210.1371/journal.pone.0011454.g00310.1371/ journal.pone.0011454.g00610.1371/journal.pone.0011454.g00510.1371/ journal.pone.0011454.g00610.1371/journal.pone.0011454.s00110.1371/ journal.pone.0011454.g00610.1371/journal.pone.0011454.s00110.1371/</u>
- Del Felice A, Arcaro C, Storti SF, Fiaschi A, Manganotti P. Electrical source imaging of sleep spindles. Clin EEG Neurosci 2014;45(3):184–92.
- Del Felice A, Arcaro C, Storti SF, Fiaschi A, Manganotti P. Slow spindles' cortical generators overlap with the epileptogenic zone in temporal epileptic patients: an electrical source imaging study. Clin Neurophysiol 2013;124(12):2336–44.
- Denis D, Bottary R, Cunningham TJ, Zeng S, Daffre C, Oliver KL, et al. Sleep Power Spectral Density and Spindles in PTSD and Their Relationship to Symptom Severity. Front Psychiatry 2021;12. <u>https://doi.org/10.3389/fpsyt.2021.766647</u>.
- Desikan RS, Ségonne F, Fischl B, Quinn BT, Dickerson BC, Blacker D, et al. An automated labeling system for subdividing the human cerebral cortex on MRI scans into gyral based regions of interest. Neuroimage 2006;31(3):968–80.
- Docherty AB, Harrison EM, Green CA, Hardwick HE, Pius R, Norman L, et al. Features of 20 133 UK patients in hospital with covid-19 using the ISARIC WHO Clinical Characterisation Protocol: prospective observational cohort study. BMJ 2020;369.
- Dubois B, Slachevsky A, Litvan I, Pillon B. The FAB: a frontal assessment battery at bedside. Neurology 2000;55(11):1621–6.
- Evans RA, McAuley H, Harrison EM, Shikotra A, Singapuri A, Sereno M, et al. Physical, cognitive, and mental health impacts of COVID-19 after hospitalisation (PHOSP-COVID): a UK multicentre, prospective cohort study. Lancet Respir Med 2021;9(11):1275–87.

#### M. Rubega, L. Ciringione, M. Bertuccelli et al.

Fernandez LMJ, Lüthi A. Sleep spindles: mechanisms and functions. Physiol Rev 2020;100(2):805–68.

- Ferrarelli F, Kaskie R, Laxminarayan S, Ramakrishnan S, Reifman J, Germain A. An increase in sleep slow waves predicts better working memory performance in healthy individuals. Neuroimage 2019;191:1–9.
- Forte G, Favieri F, Tambelli R, Casagrande M. COVID-19 pandemic in the Italian population: validation of a post-traumatic stress disorder questionnaire and prevalence of PTSD symptomatology. Int J Environ Res 2020;17(11):4151.
- Giovagnoli AR, Del Pesce M, Mascheroni S, Simoncelli M, Laiacona M, Capitani E. Trail making test: normative values from 287 normal adult controls. Ital J Neurol Sci 1996;17(4):305–9.
- Glomb K, Mullier E, Carboni M, Rubega M, Iannotti G, Tourbier S, et al. Using structural connectivity to augment community structure in EEG functional connectivity. Netw Neurosci 2020a;4(3):761–87.
- Glomb K, Rué Queralt J, Pascucci D, Defferrard M, Tourbier S, Carboni M, et al. Connectome spectral analysis to track EEG task dynamics on a subsecond scale. NeuroImage 2020b;221:117137. <u>https://doi.org/10.1016/j.</u> neuroimage.2020.117137.
- Graham EL, Clark JR, Orban ZS, Lim PH, Szymanski AL, Taylor C, DiBiase RM, Jia DT, Balabanov R, Ho SU, Batra A, Liotta EM, Koralnik IJ. Persistent neurologic symptoms and cognitive dysfunction in non-hospitalized Covid-19 "long haulers". Ann Clin Transl Neurol 2021;8(5):1073–85.
- Greene T, El-Leithy S, Billings Jo, Albert I, Birch J, Campbell M, et al. Anticipating PTSD in severe COVID survivors: the case for screen-and-treat. Eur J Psychotraumatol 2022;13(1). <u>https://doi.org/10.1080/20008198.2021.1959707</u>.
- Hanifa ALB, Glæemose AO, Laursen BS. Picking up the pieces: Qualitative evaluation of follow-up con- sultations post intensive care admission. Intensive Crit Care Nurs 2018;48:85–91.
- Helms J, Kremer S, Merdji H, Clere-Jehl R, Schenck M, Kummerlen C, et al. Neurologic features in severe SARS-CoV-2 infection. N Engl J Med 2020;382 (23):2268–70.
- Ishii R, Dziewas R, Chau W, Sörös P, Okamoto H, Gunji A, et al. Current source density distribution of sleep spindles in humans as found by synthetic aperture magnetometry. Neurosci Lett 2003;340(1):25–8.
- Klinzing JG, Mölle M, Weber F, Supp G, Hipp JF, Engel AK, et al. Spindle activity phase-locked to sleep slow oscillations. Neuroimage 2016;134:607–16.
- Lacourse K, Delfrate J, Beaudry J, Peppard P, Warby SC. A sleep spindle detection algorithm that emulates human expert spindle scoring. J Neurosci Methods 2019;316:3–11.
- Lafortune M, Gagnon J-F, Martin N, Latreille V, Dubé J, Bouchard M, et al. Sleep spindles and rapid eye movement sleep as predictors of next morning cognitive performance in healthy middle-aged and older participants. J Sleep Res 2014;23 (2):159–67.
- Lam A, Haroutonian C, Grummitt L, Ireland C, Grunstein RR, Duffy S, et al. Sleepdependent memory in older people with and without MCI: The Relevance of Sleep Microarchitecture, OSA, Hippocampal Subfields, and Episodic Memory. Cereb Cortex 2021;31(6):2993–3005.
- Lustenberger C, Wehrle F, Tüshaus L, Achermann P, Huber R. The multidimensional aspects of sleep spindles and their relationship to word-pair memory consolidation. Sleep 2015;38(7):1093–103.
- Manshanden I, De Munck JC, Simon NR, Lopes da Silva FH. Source localization of MEG sleep spindles and the relation to sources of alpha band rhythms. Clin Neurophysiol 2002;113(12):1937–47.
- Mao L, Jin H, Wang M, Hu Yu, Chen S, He Q, et al. Neurologic manifestations of hospitalized patients with coronavirus disease 2019 in Wuhan. China JAMA Neurol 2020;77(6):683. <u>https://doi.org/10.1001/jamaneurol.2020.1127</u>.
- Michel CM, Brunet D. EEG source imaging: a practical review of the analysis steps. Front Neurol 2019;10:325.
- Mölle M, Bergmann TO, Marshall L, Born J. Fast and slow spindles during the sleep slow oscillation: disparate coalescence and engagement in memory processing. Sleep 2011;34(10):1411–21.
- Monaco M, Costa A, Caltagirone C, Carlesimo GA. Forward and backward span for verbal and visuo-spatial data: standardization and normative data from an Italian adult population. Neurol Sci 2013;34(5):749–54.
- Nalbandian A, Sehgal K, Gupta A, Madhavan MV, McGroder C, Stevens JS, et al. Postacute COVID-19 syndrome. Nat Med 2021;27(4):601–15.
- Nasreddine ZS, Phillips NA, Bédirian V, Charbonneau S, Whitehead V, Collin I, et al. The Montreal Cognitive Assessment, MoCA: a brief screening tool for mild cognitive impairment. J Am Geriatr Soc 2005;53(4):695–9.
- Needham DM, Davidson J, Cohen H, Hopkins RO, Weinert C, Wunsch H, et al. Improving long-term outcomes after discharge from intensive care unit: report from a stakeholders' conference. Crit Care Med 2012;40(2):502–9.
- Nocentini U, Giordano A, Vincenzo SD, Panella M, Pasqualetti P. The Symbol Digit Modalities Test-Oral version: Italian normative data. Funct Neurol 2006;21 (2):93–6.
- Pascucci D, Rubega M, Plomp G, Battaglia FP. Modeling time-varying brain networks with a self-tuning optimized Kalman filter. PLoS Comput Biol 2020;16(8): e1007566. https://doi.org/10.1371/journal.pcbi.1007566.g00210.1371/journal. pcbi.1007566.g00310.1371/journal.pcbi.1007566.g00210.1371/journal.

Clinical Neurophysiology 140 (2022) 126–135

pcbi.1007566.s00110.1371/journal.pcbi.1007566.s00210.1371/journal. pcbi.1007566.r00110.1371/journal.pcbi.1007566.r00210.1371/journal. pcbi.1007566.r00310.1371/journal.pcbi.1007566.r004.

- Petrilli CM, Jones SA, Yang J, Rajagopalan H, O'Donnell L, Chernyak Y, et al. Factors associated with hospital admission and critical illness among 5279 people with coronavirus disease, in New York City: prospective cohort study. BMJ 2019;2020:369.
- Pezzilli R, Morselli-Labate AM, Frulloni L, Cavestro GM, Ferri B, Comparato G, et al. The quality of life in patients with chronic pancreatitis evaluated using the SF-12 questionnaire: a comparative study with the SF-36 questionnaire. Dig Liver Dis 2006;38(2):109–15.
- Plante DT, Goldstein MR, Landsness EC, Peterson MJ, Riedner BA, Ferrarelli F, et al. Topographic and sex-related differences in sleep spindles in major depressive disorder: a high-density EEG investigation. J Affect Disord 2013;146(1):120–5.
- Popescu. Hypoxic-Ischemic Injury of Basal Ganglia Associated with the COVID-19 Infection: Case Report.. ase Rep Neurol 2021;13(3). <u>https://doi.org/10.1159/ 000518506</u>.
- Premraj L, Kannapadi NV, Briggs J, Seal SM, Battaglini D, Fanning J, et al. Mid and long-term neurological and neuropsychiatric manifestations of post-COVID-19 syndrome: A meta-analysis. J Neurol Sci 2022;434:120162. <u>https://doi.org/ 10.1016/j.jns.2022.120162</u>.
- Rabinovici GD, Stephens ML, Possin KL. Executive dysfunction. Continuum (Minneap Minn) 2015;21(3):646–59.
- Rubega M, Carboni M, Seeber M, Pascucci D, Tourbier S, Toscano G, et al. Estimating EEG Source Dipole Orientation Based on Singular-value Decomposition for Connectivity Analysis. Brain Topogr 2019;32(4):704–19.
- Santangelo G, Siciliano M, Pedone R, Vitale C, Falco F, Bisogno R, et al. Normative data for the Montreal Cognitive Assessment in an Italian population sample. Neurol Sci 2015;36(4):585–91.
- Schabus M, Hödlmoser K, Gruber G, Sauter C, Anderer P, Klösch G, et al. Sleep spindle-related activity in the human EEG and its relation to general cognitive and learning abilities. Eur J Neurosci 2006;23(7):1738–46.
- Schultebraucks K, Ben-Zion Z, Admon R, Keynan JN, Liberzon I, Hendler T, Shalev AY. Assessment of early neurocognitive functioning increases the accuracy of predicting chronic PTSD risk. Mol Psychiatry 2022;27(4):2247–54.
- Sedgwick P. Pitfalls of statistical hypothesis testing: type I and type II errors. BMJ 2014;349(jul03 1):g4287.
- Shah W, Hillman T, Playford ED, Hishmeh L. Managing the long term effects of covid-19: summary of NICE, SIGN, and RCGP rapid guideline. BMJ 2021;372.
- Shao Y, Zou G, Tabarak S, Chen J, Gao X, Yao P, et al. Spindle-related brain activation in patients with insomnia disorder: An EEG-fMRI study. Brain Imaging Behav 2022;16(2):659–70.
- Staresina BP, Bergmann TO, Bonnefond M, van der Meij R, Jensen O, et al. Hierarchical nesting of slow oscillations, spindles and ripples in the human hippocampus during sleep. Nat Neurosci 2015;18(11):1679–86.
- Steriade M. Sleep, epilepsy and thalamic reticular inhibitory neurons. Trends Neurosci 2005;28(6):317-24.
- Tamaki M, Matsuoka T, Nittono H, Hori T. Activation of fast sleep spindles at the premotor cortex and parietal areas contributes to motor learning: a study using sLORETA. Clin Neurophysiol 2009;120(5):878–86.
- Torres J, Carvalho D, Molinos E, Vales C, Ferreira A, Dias CC, et al. The impact of the patient post-intensive care syndrome components upon caregiver burden. Med Intensiva 2017;41(8):454–60.
- Urakami Y. Relationships between sleep spindles and activities of cerebral cortex as determined by simultaneous EEG and MEG recording. J Clin Neurophysiol 2008;25(1):13–24.
- Varghese S, Frey BN, Schneider MA, Kapczinski F, Azevedo Cardoso T. Functional and cognitive impairment in the first episode of depression: A systematic review. Acta Psychiatr Scand 2022;145(2):156–85.
- Vignatelli L, Plazzi G, Barbato A, Ferini-Strambi L, Manni R, Pompei F, et al. Italian version of the Epworth sleepiness scale: external validity. Neurol Sci 2003;23 (6):295–300.
- Wang C, Laxminarayan S, Ramakrishnan S, Dovzhenok A, Cashmere JD, Germain A, Reifman J. Increased oscillatory frequency of sleep spindles in combat-exposed veteran men with post-traumatic stress disorder. Sleep 2020;43(10). <u>https:// doi.org/10.1093/sleep/zsaa064</u>.
- doi.org/10.1093/sleep/zsaa064.
  Whiteside DM, Basso MR, Naini SM, Porter J, Holker E, Waldron EJ, et al. Outcomes in post-acute sequelae of COVID-19 (PASC) at 6 months post-infection Part 1: Cognitive functioning. Clin Neuropsychol 2022;36(4):806–28.
- Xia M, Wang J, He Y, Csermely P. BrainNet Viewer: a network visualization tool for human brain connectomics. PloS One 2013;8(7):e68910. <u>https://doi.org/ 10.1371/journal.pone.0068910.0371/journal.pone.0068910.g00310.1371/</u> journal.pone.0068910.g00210.1371/journal.pone.0068910.g00310.1371/ journal.pone.0068910.g00610.1371/journal.pone.0068910.g00710.1371/ journal.pone.0068910.g00610.1371/journal.pone.0068910.g00710.1371/ journal.pone.0068910.g00810.1371/journal.pone.0068910.g00910.1371/ journal.pone.0068910.g00810.1371/journal.pone.0068910.g00910.1371/ journal.pone.0068910.g00810.1371/journal.pone.0068910.g00910.1371/
- Ziauddeen N, Gurdasani D, O'Hara ME, Hastie C, Roderick P, Yao G. Characteristics of Long Covid: findings from a social media survey. BMJ 2021;A90.