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1
2 Abbreviated Title: Peak Alpha Frequency and Pain Sensitivity

3 **Cerebral Peak Alpha Frequency Predicts Individual Differences in Pain Sensitivity**

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14 *Keywords:* Pain, Peak Alpha Frequency, EEG

Abstract

The identification of neurobiological markers that predict individual predisposition to pain are not only important for development of effective pain treatments, but would also yield a more complete understanding of how pain is implemented in the brain. In the current study using electroencephalography (EEG), we investigated the relationship between the peak frequency of alpha activity over sensorimotor cortex and pain intensity during capsaicin-heat pain (C-HP), a prolonged pain model known to induce spinal central sensitization in primates. We found that peak alpha frequency (PAF) recorded during a pain-free period preceding the induction of prolonged pain correlated with subsequent pain intensity reports: slower peak frequency at pain-free state was associated with higher pain during the prolonged pain condition. Moreover, the degree to which PAF decreased between pain-free and prolonged pain states was correlated with pain intensity. These two metrics were statistically uncorrelated and in combination were able to account for 50% of the variability in pain intensity. Altogether, our findings suggest that pain-free state PAF over relevant sensory systems could serve as a marker of individual predisposition to prolonged pain. Moreover, slowing of PAF in response to prolonged pain could represent an objective marker for subjective pain intensity. Our findings potentially lead the way for investigations in clinical populations in which alpha oscillations, and the brain areas contributing to their generation are used in identifying and formulating treatment strategies for patients more likely to develop chronic pain.

Highlights

Relationship between EEG peak alpha frequency and prolonged pain is examined

PAF during pain-free state correlated with prolonged pain intensity 40 minutes later

PAF change from pain-free to prolonged pain correlated with reported pain intensity

PAF and PAF changes could represent distinct mechanisms predicting pain sensitivity

36 Introduction

37 Pain is a salient, multidimensional experience that varies widely between individuals in both intensity and duration.
38 Identifying biomarkers that can determine individual susceptibility for the development of chronic pain is a fundamental
39 step for improved pain treatments. One approach to this problem has been to investigate the role that neural
40 oscillations like the alpha rhythm play in the individual pain experience (Peng et al., 2015; Ploner, Sorg, Gross, 2016).
41 The alpha rhythm represents the predominant oscillatory activity in the EEG which is chiefly observed in primary sensory
42 regions (e.g. vision, auditory). Although previously considered a signature of cortical “idling,” significant evidence now
43 suggests that alpha activity plays a top-down role in gating information in sensory cortices depending on task demands
44 (Foxe et al., 1998; Foxe & Snyder, 2011; Jensen & Mazaheri, 2010; Klimesch, 2012, Pfurtscheller et al., 1996).
45 The peak frequency of alpha activity (i.e the frequency within the 8-12Hz, that has the maximal power) has been found
46 to change across the life span, increasing from childhood to adulthood, and subsequently decreasing with age (Aurlien et
47 al., 2004; Lindsley, 1939, Hashemi et al., 2016; Bazanova & Vernon, 2014). There is evidence that the frequency of alpha
48 activity is positively correlated to measures such as working performance (reviewed in Klimesch, 1999). More recently, it
49 has been demonstrated that individuals with higher alpha frequencies in the occipital cortex are able to perceive visual
50 information with a finer temporal resolution (Samaha et al., 2015). Peak alpha frequency has been found to be reliable
51 in test-retest studies (Grandy et al., 2013), and appears to be a heritable phenotypic trait (Posthuma et al., 2001; Smit et
52 al., 2006). Taken together, these studies suggest that peak alpha frequency (PAF) could be viewed as a ‘state’ variable
53 with its subtle fluctuations within an individual reflecting shifts in the excitability of the underlying cortex and its
54 capacity to process information. Alternatively PAF can be viewed as a ‘trait’ variable with its variability across individuals
55 reflecting cognitive ability.

56 In recent years, the variability of alpha frequency has been studied in the context of characterizing disease states in
57 clinical populations, and the subjective experience of pain in the typical population. In patients suffering from central,
58 visceral, and neuropathic pain conditions, PAF was slowed relative to matched, healthy controls (Sarnthein et al., 2005;
59 Walton et al., 2010; de Vries et al., 2013, Lim et al., 2016). It has been hypothesized, that the slowing of PAF and that the

60 increased power of slower alpha rhythms (8-9.5 Hz) contributes to the generation of pathological pain, perhaps
61 reflecting thalamocortical dysrhythmia (Llinas et al., 2005).

62 In contrast to the slowing of PAF associated with chronic pain, exposure to acute, painful stimuli in healthy subjects has
63 been found to increase the frequency of alpha activity (Nir, et al 2010). Furthermore, PAF collected from healthy
64 individuals either during or, perhaps more importantly, prior to stimulation were positively correlated with pain intensity
65 (Nir et al., 2010), suggesting that PAF reflects processes related to both ongoing pain and individual vulnerability.

66 These findings together suggest a rather complex relationship between types of pain and variations in PAF: transient
67 acute pain, increases alpha frequency in the healthy population, whereas alpha frequency is slowed down in patients
68 with chronic pain. The slowing of alpha frequency in chronic pain populations could reflect changes in the brain's neural
69 architecture brought about by the constant experience of pain. Supporting this view is a finding that PAF had an inverse
70 relationship with duration of chronic pancreatitis (de Vries et al., 2013). An alternative explanation could be that
71 individuals with slower alpha frequency are more prone to develop chronic pain. Why some people will go on to develop
72 chronic pain following an injury that would normally heal and not lead to persistent pain remains a major question in the
73 field, and cerebral functional connectivity might be one way to predict this transition from acute to chronic pain (Baliki
74 et al., 2012).

75 Here we investigated the relationship between PAF and sensitivity to prolonged pain. The prolonged pain model we
76 used – the capsaicin-heat pain model – lasts for hours to days and recapitulates cardinal sensory aspects of chronic
77 neuropathic pain (Culp et al., 1989; LaMotte RH, et al,1992; Baron 2009; Lotsch et al., 2015). The prolonged pain model
78 might thus be more similar to chronic pain – or the early transition period from acute to chronic pain – than acute pain,
79 where there is no central sensitization, and the pain disappears as soon as the stimulus is removed. The personal
80 experience of pain is highly variable among individuals even if the underlying noxious stimulation is similar. The
81 objective of our study was to systematically investigate the relationship between PAF prior to and during prolonged pain
82 and the subjective experience of pain. We recorded EEG activity during pain-free and prolonged pain states, which
83 allowed us to determine the relationship of PAF and pain intensity, as well as how PAF shifts (i.e. change in PAF between

84 states) relate to individual pain intensity. We tested the hypothesis that PAF slowing reflects the intensity of prolonged
85 pain.

86 Materials and Methods

87 *Participants*

88 Forty-four pain-free, neurotypical adult participants (22 males, mean age = 28.4, age range = 19 – 42) took part in the
89 experiment. Twenty-seven participants were randomly assigned to the Pain group (would be administered topical
90 capsaicin), while seventeen were assigned to the Non-Pain group (not administered topical capsaicin). The Non-Pain
91 group served as a control to confirm that prolonged pain was a result of the capsaicin application and not only the warm
92 thermode, as well as to control for effects of ongoing stimulation and attention. More participants were assigned to the
93 capsaicin group to account for the variability in response to topical capsaicin (Liu et al., 1998). This study was approved
94 by the University of Maryland, Baltimore Institutional Review Board, and informed written consent was obtained from
95 each participant prior to any study procedures.

96 *EEG*

97 Scalp EEG was collected from an EEG cap housing a 64 channel Brain Vision actiCAP system (Brain Products GmbH,
98 Munich, Germany) labeled in accord with an extended international 10–20 system (Oostenveld and Praamstra, 2001). All
99 electrodes were referenced online to an electrode placed on the right earlobe and a common ground set at the FPz site.
100 Electrode impedances were maintained below 5k Ω throughout the experiment. Brain activity was continuously
101 recorded within .01 to 100 Hz bandpass filter, and with a digital sampling rate of 1000 Hz. The EEG signal was amplified
102 and digitized using a BrainAmp DC amplifier (Brain Products GmbH, Munich, Germany) linked to Brain Vision Recorder
103 software (version 2.1, Brain Products GmbH, Munich, Germany).

104 *Prolonged pain induced by the Capsaicin-Heat Pain model*

105 Thermal stimuli were delivered to the volar surface of participant's left forearm using a thermal-contact heat stimulator
106 (30 × 30 mm Medoc Pathway ATS Peltier device; Medoc Advanced Medical Systems Ltd., Ramat Yishai, Israel). Prior to
107 the beginning of the experiment all participants underwent a brief sensory testing session in which they were asked to
108 report when they felt a change in temperature (for warmth detection threshold (WDT)) or when the temperature first

109 became painful (heat pain threshold (HPT)). For WDT and HPT three and four trials were presented, respectively, and
110 the average across trials, rounded down to the nearest integer, was used.

111 Prolonged pain was modelled following a procedure modified from previous studies (Anderson et al., 2002). We applied
112 ~1g 10% capsaicin paste (Professional Arts Pharmacy, Baltimore, MD) topically to the volar surface of the left forearm,
113 fixing it in place with a Tegaderm bandage. After 15 fifteen minutes of exposure, we placed the thermode over top of
114 the Tegaderm bandage at a temperature that was greater than the WDT and at least 1°C below the HPT. We term this
115 model the capsaicin-heat pain model (C-HP).

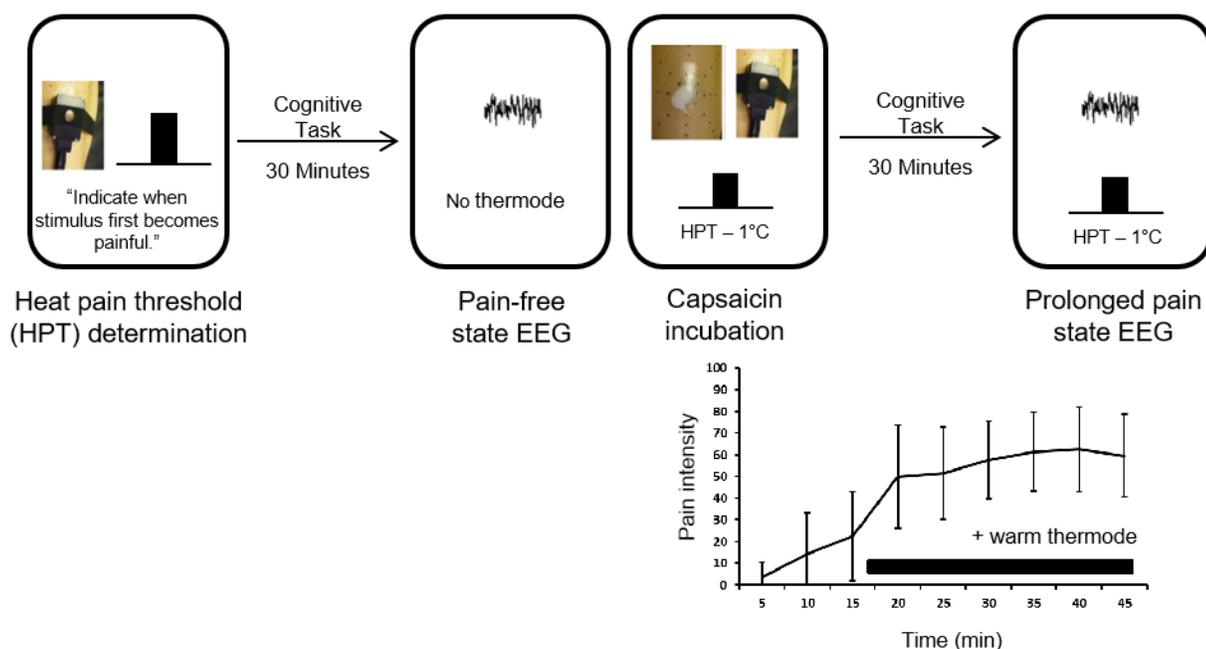
116 To ensure that the capsaicin produced a stable, long-lasting pain, participants were asked to provide pain intensity
117 ratings every minute for the first five minutes following thermode placement. The thermode temperature was adjusted
118 during this time to achieve a consistent pain intensity above 20 on a 0-100 point scale (i.e. if pain was intolerable, the
119 temperature was lowered slightly, and if there was no pain, the temperature was increased closer to the HPT). Once this
120 five minute period elapsed, the temperature was held in place for 25 minutes. Participants were asked to rate pain
121 intensity every 5 minutes. This procedure does not cause lasting tissue damage (Moritz and Henriques, 1947). Previous
122 work has found that topical capsaicin evokes no pain or hypersensitivity in some participants (Liu et al., 1998; Walls et
123 al., 2017). Therefore, we excluded participants who did not develop moderate pain, which we set at a reported pain
124 intensity level of 20 (details of the scale provided below).

125 ***Procedure***

126 A summary of the order of procedures is described in Figure 1. Once the EEG set-up was complete, participants were
127 seated in a comfortable chair and underwent a brief sensory testing session to establish their individual HPT.

128 Participants were then trained on and performed a simple cognitive task which will be detailed elsewhere. The total
129 duration of this task was approximately thirty minutes. While performing this task, participants rate their current pain
130 intensity every five minutes on a 0-100 scale, with the anchors 0, not at all painful and 100, most intense pain
131 imaginable. In total participants provided six pain intensity ratings during this testing session. Ratings were always given
132 during a rest period. At the conclusion of this testing session, and immediately following the final pain intensity rating,
133 all lights in the testing room were turned off and participants were instructed to close their eyes, remain still, and relax

134 without falling asleep. Continuous EEG was recorded during this pain-free resting state for three minutes in both the
135 Pain and Non-Pain groups.



136

137 Figure 1. Outline of the experimental procedure. Participants first underwent sensory testing to determine their
138 Heat Pain Threshold (HPT). After a 30 minute cognitive task, EEG was collected while participants completed a 3-
139 minute eyes closed session in the absence of any thermal stimulus (pain-free state). Next, capsaicin was applied
140 (Pain group) to the forearm and a temperature no more than one degree below their HPT was introduced
141 fifteen minutes later. Five minutes later, when pain in response to the model has stabilized, the same cognitive
142 task from earlier in the experiment was repeated. Following this task, EEG was collected while participants
143 completed a 3-minute eyes closed session in the presence of capsaicin and warm thermode (prolonged pain
144 state). Subjects in the Non-Pain group underwent identical procedures, but without capsaicin application.

145 After finishing this pain-free state EEG recording, the lights in the testing room were turned on, capsaicin was applied to
146 the participant's left forearm, as described above, and the thermode was placed directly on top of the capsaicin
147 application. During this incubation period participants were instructed to relax without falling asleep. The thermode was
148 kept at 32°C, and participants provided a pain intensity rating every three minutes over a total of fifteen minutes. For

149 participants in the Non-Pain group, this process was identical, including thermode placement, except there was no
150 capsaicin application.

151 Following this incubation period, the thermode temperature increased to a warm temperature 3°C below the previously
152 determined HPT. Every minute, for the next five minutes, participants were asked to provide a pain intensity rating. If
153 the participant did not report feeling any sensation from the capsaicin, the temperature was adjusted in 1°C increments
154 with the requirement that the final testing temperature be at least 1°C below their HPT. For Non-Pain group
155 participants, adjustments were only made to lower the temperature in the event that pain was reported. When this five
156 minute period had elapsed, the full twenty-five minute cognitive task from earlier in the experiment was performed
157 once more. As before, participants were asked to provide a total of 6 pain intensity ratings during this testing.
158 Immediately after the last rating was provided, a three minute “stimulation” resting state EEG was collected. For the
159 Pain group, this “prolonged pain” resting state was collected with the capsaicin and warm thermode placed on the
160 forearm. For the Non-Pain Group, this “nonpainful warmth” resting state was collected with the warm thermode placed
161 on the forearm without capsaicin.

162 ***Data Processing***

163 The primary data of interest in this study were the within-subject resting state EEG acquired prior to and during
164 prolonged capsaicin pain. For the primary set of analyses the preprocessing of EEG data was done using EEGLAB 13.6.5b
165 (Delorme and Makeig, 2004) using an approach similar to that used previously (Scheeringa et al., 2011a; Scheeringa et
166 al., 2011b). Here, the first step involved band-pass filtering the EEG between 5 and 16 Hz using the function ‘eegnewfilt’
167 after which Infomax (extended) independent component analysis (ICA) was performed (Bell and Sejnowski, 1995). It
168 should be noted that the ICA was performed on resting state EEG data combined across the pain-free and prolonged
169 pain states. The obtained unmixing matrix was applied to the unfiltered data resulting in components that retained
170 broadband spectral content. A Fourier transform was done on the time series of each component to obtain a frequency-
171 power spectra for each component. Next for each participant we visually inspected the frequency-spectra of the
172 components, and identified components that had a clear alpha peak (8-14 Hz) and a scalp topography that suggested a

173 source predominately over the sensorimotor cortices. This component is referred to as the “central component” for the
174 remainder of the manuscript.

175 **Quantification of PAF**

176 The frequency decomposition of the sensorimotor component data was done using the routines in FieldTrip (Oostenveld
177 et al., 2011). The data was segmented into 5-second epochs and power spectral density in the 2-40 Hz range was derived
178 for each epoch in 0.2 Hz bins using the ‘ft_freqanalysis_mtmfft’ function. A Hanning taper was applied to the data prior
179 to calculating the spectra to reduce any edge artifacts (Mazaheri et al., 2010; Mazaheri et al., 2009; Mazaheri 2014).
180 The peak alpha frequency for each 5 second epoch was estimated using a center of gravity (CoG) method (Jann et al.,
181 2012; Jann et al., 2010; Klimesch, Schimke, & Pfurtscheller, 1993). We defined CoG as follows:

$$182 \text{ CoG} = \frac{\sum_{i=1}^n f_i * a_i}{\sum_{i=1}^n a_i}$$

183 where f_i is the i th frequency bin including and above 9 Hz, n is the number of frequency bins between 9 and 11 Hz, and a_i
184 the spectral amplitude for f_i . PAF, as well as power at the PAF bin (PAF Power), were estimated for the central alpha
185 components for every 5 second epoch and then averaged.

186 **Statistical analysis**

187 We first investigated whether capsaicin led to heightened pain intensity using an independent samples t-test. We
188 determined average pain intensity ratings to capsaicin for each participant by averaging the six ratings during the
189 prolonged pain state. Average pain intensity ratings were compared between Pain and Non-Pain groups using an
190 independent samples t-test. This test was performed separately for the whole sample and the sample that excluded
191 subjects in the Non-Pain group who developed pain and subjects in the Pain group who had <20/100 pain.

192 In order to investigate if central component PAF during pain-free and prolonged pain states were related to pain
193 intensity, we correlated each Pain group participant’s central component PAF during the pain-free state (i.e. before the
194 administration of capsaicin) and during prolonged pain with their averaged pain intensity. In order to account for the
195 possibility that the relationship between PAF and pain intensity ratings could be confounded by the temperature of the
196 thermal device, we performed a partial correlation between PAF and pain controlling for thermode temperature. Due to

197 technical error, thermode temperatures were missing for two participants in the Pain Group and one participant in the
198 Non-Pain Group.

199 For all correlational analyses, Pearson's correlation coefficients were used to test the relationship between variables.
200 Analyses were also conducted using Spearman's rank order correlations, but these did not change any of the results and
201 are therefore not reported.

202 As an additional test to investigate whether alpha frequency was related to pain sensitivity, we separated our Pain group
203 participants into "high" and "low" pain sensitive groups by performing a median split based on pain intensity. Here, a
204 2x2 Repeated Measures ANOVA with group (high pain sensitive vs low pain sensitive vs Non-Pain) x state (pain-free vs
205 prolonged pain state) serving as between- and within-subject factors, respectively, was used to assess how central PAF
206 differed amongst groups and how it changes in response to C-HP.

207 Next, we investigated if changes in central PAF from baseline to prolonged pain state were related to the pain intensity
208 reported by the participants. This PAF shift (Δ PAF) was calculated by subtracting pain-free state PAF from the prolonged
209 pain state PAF. We then correlated Δ PAF with pain intensity, and, as above, we also performed a partial correlation to
210 control for the impact of thermode temperature.

211 Hierarchical multiple regression was used to test the independent contributions of baseline resting state PAF and Δ PAF.
212 In this model, pain intensity was the dependent variable and baseline resting state PAF and Δ PAF were the independent
213 variables entered sequentially in the model.

214 We followed this multiple regression with a leave one out regression approach to formally evaluate the ability of
215 baseline PAF and Δ PAF to predict C-HP model sensitivity. To do so, we generated a series of regression models using
216 central baseline PAF and central Δ PAF from all but one Pain group individual. The resulting model intercept and
217 unstandardized beta coefficients were used to generate a pain prediction for the single individual withheld from model
218 building. This procedure was repeated iteratively so that each individual served as the test participant for exactly one
219 regression model. The accuracy of these pain predictions were then tested by calculating the Pearson correlation
220 between actual pain intensity and the pain intensity predicted by the leave one out models. To test the significance of

221 this prediction, the aforementioned procedure was repeated 10,000 times using randomly shuffled pain and PAF
222 measures to bootstrap a null distribution of r values. The 95% of the null distribution was used as a significance cutoff
223 for assessing the predictive ability of PAF and Δ PAF. To ensure that results generalized beyond this maximally sized
224 training set, we repeated the above analysis with training set sizes ranging from 3 individuals to 19 individuals. For each
225 training set size, a separate regression model was generated for each possible unique combination of a given training
226 size and the overall correlation between all predictions and observed pain intensity was assessed with a Pearson
227 correlation.

228 Results

229 *Pain Intensity and the C-HP model*

230 Prolonged pain was evoked using C-HP model on the forearm. Six participants in the Pain group were excluded for failing
231 to develop moderate pain to the capsaicin (consistent with previous observations that about 25% of people are
232 insensitive to capsaicin (Liu et al., 1998; Walls et al. 2017) and three participants in the Non-Pain group were excluded
233 for developing pain that was rated as greater than 10 on average. For the remaining 21 participants in the Pain group,
234 mean pain intensity was 56.01 (s.d. ± 16.96). For the Non-Pain group, which underwent identical procedures without
235 capsaicin exposure, mean pain was 1.99 (s.d. ± 2.68). As a manipulation check, an independent samples t-test comparing
236 these two groups confirmed that the presence of capsaicin led to heightened pain in response to a warm stimulus, $t(36)$
237 = 11.86, $p < .01$. (This test was also performed for the entire sample (i.e. including subjects who did not respond to the
238 C-HP model and subjects who reported pain with just the warm stimulus): $t(42) = 6.78$, $p < .01$). This difference appears
239 to be a result of the capsaicin rather the heat stimulus given that applied temperatures were not significantly different
240 between the group (Pain Group: mean = 38.52, std = 2.71, range = 32-41; Non-Pain group: mean = 38.25, std = 1.57,
241 range = 37-41; $t(33) = .36$, $p = .72$). Furthermore, there was no difference between the groups in terms of HPT (Pain
242 Group: mean = 43.67, std = 2.22, range = 39 – 47; Non-Pain group: mean = 43.52, std = 2.74, range = 39 – 50; $t(36) = .86$,
243 $p = .17$) or difference between HPT and thermode temperature (Pain Group: mean = 5.21, std = 2.16, range = 1-9; Non-

244 Pain Group: mean = 5.44, std = 2.13, range = 2-9; $t(33) = .75, p = .31$.In addition, there was no relationship between
245 thermode temperature and pain intensity in the Pain group ($r = -0.25, p = 0.30$) or Non-Pain group ($r = -.02, p = .94$).

246 ***PAF at pain-free and prolonged pain states correlated with pain intensity***

247 The topography of the central alpha component used in our analysis, averaged across Pain group participants can be
248 seen in Figure 2A.

249 We first set out to investigate if central component PAF recorded during the pain-free state correlated with pain
250 intensity. We found that pain-free state central component PAF correlated negatively with pain intensity ($r = -.57, p =$
251 $.01$); that is, the lower an individual's average central PAF, the greater their pain (Figure 2B). This provides initial
252 evidence that an individual's central PAF in the absence of a noxious stimulus may play a role in determining an
253 individual's vulnerability to a prolonged pain. There was not a significant relationship between the pain-free state power
254 estimate of the central component PAF (PAF power) and subsequent pain intensity ratings ($r = .23, p = .32$).

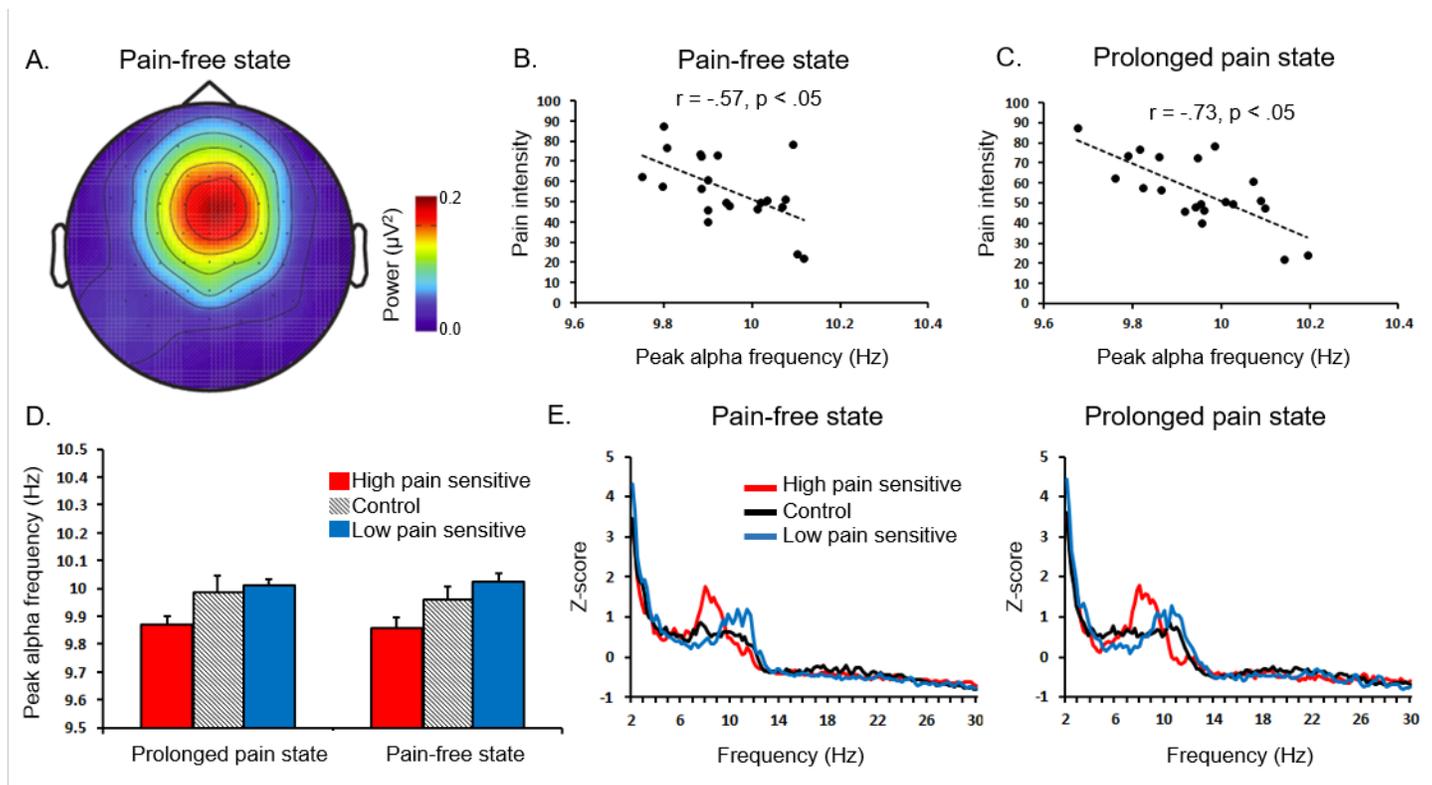
255 Next, we assessed whether central component PAF during the prolonged pain state was related to pain intensity. We
256 found central PAF during prolonged pain correlated negatively with pain intensity ($r = -.73, p < .01$); i.e., slower PAF was
257 associated with greater pain intensity (Figure 2C). The relationship between prolonged pain state central component
258 PAF and pain intensity remained significant when controlling for thermode temperature using a partial correlation ($r = -$
259 $.72, p < .01$), suggesting that this relationship is driven by factors other than the magnitude of the sensory stimulus
260 alone. Again we did not observe a significant relationship between central component PAF power during prolonged pain
261 and pain intensity ($r = 0.10, p = .67$), highlighting the importance of PAF rather than PAF power in prolonged pain.

262 ***PAF can distinguish between high and low pain sensitive individuals***

263 The foregoing correlations suggest that the frequency of central alpha activity at baseline and during pain is related to
264 the pain intensity an individual experiences. To investigate this relationship further we performed a median split of our
265 Pain group participants into high and low pain sensitivity groups based on their reported pain intensity.

266 The difference in central PAF between Non-Pain (control), high pain sensitive, and low pain sensitive groups was
267 statistically assessed using a 2x2 Repeated Measures ANOVA with group (controls vs high pain sensitive vs low pain

268 sensitive) x state (pain-free vs prolonged pain state) serving as between- and within-subject factors. The main effect of
 269 group was significant, $F(2,32) = 3.48, p = .04$. As can be seen qualitatively in Figure 2D, the low pain sensitive group
 270 displayed the fastest central PAF across both states, the high pain sensitive group displayed the slowest central PAF
 271 across both states, and the control group displayed PAF somewhere in between the two; this last observation likely
 272 reflects that the Non-pain group contains some combination of high and low pain sensitive individuals. Critically, neither
 273 the main effect of state $F(2,32) = .127, p = .72$, nor the group x state interaction $F(2,32) = .397, p = .68$ were significant.



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Figure 2. The relationship between PAF and prolonged pain. (A) The topography of the 'central' alpha component selected for peak frequency analysis averaged across Pain group participants during the pain-free state. (B) Central component PAF during the pain-free state was plotted against future pain-intensity ratings (pain during the prolonged pain state). There was a negative correlation between PAF and pain intensity. (C) Central component PAF during the prolonged pain state and pain intensity, showing a similar negative relationship. (D-E) Pain group subjects were divided into low- and high-pain sensitive groups based on a median split of pain intensity ratings in response to the capsaicin-heat pain model. (D) High pain sensitive subjects demonstrated significantly slower central PAF across both pain-free and prolonged pain states than low pain

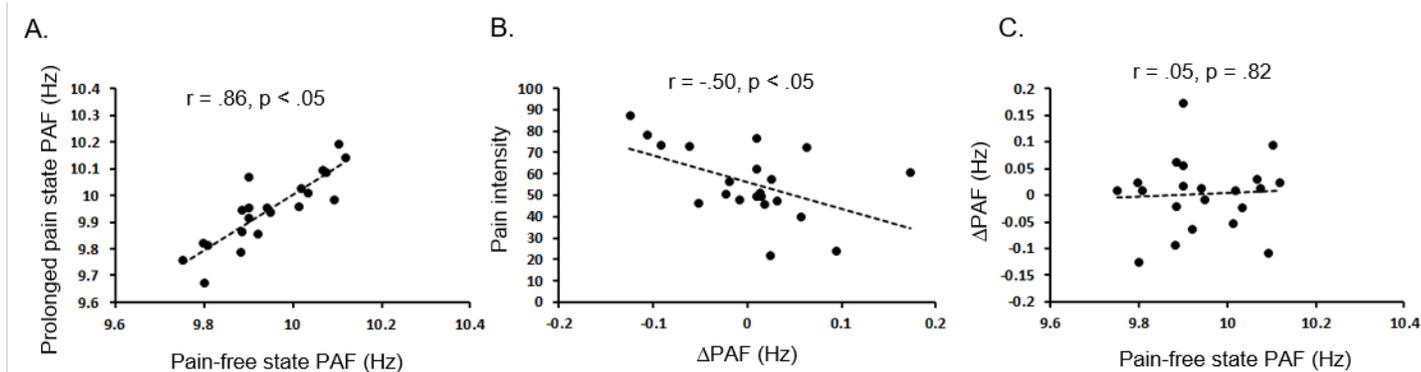
283 sensitive subjects. Error bars reflect \pm SEM. (E) High pain sensitive subjects show a selective increase in power at
284 slower alpha frequencies relative to low pain sensitive subjects. The frequency spectra was normalized across
285 participants by transforming the data into z-scores from the total mean amplitude of the frequency spectra in
286 each 5 second epoch.

287 Bonferroni corrected pair-wise comparisons revealed a significant difference in PAF between high and low pain sensitive
288 groups in the pain-free state, $p = .026$. Visual inspection of the central component power spectra revealed differences
289 between groups were largely restricted to the alpha frequency domain, further highlighting the specific importance of
290 alpha in our model of prolonged pain (Figure 2E).

291 ***PAF shift from pain-free to prolonged pain states (Δ PAF) was associated with pain intensity***

292 Central component PAF in the pain-free and prolonged pain states were strongly correlated ($r = 0.86$, $p < .05$, Figure 3A).
293 While this suggests PAF is largely stationary, it does not rule out the possibility that small changes in PAF also play a role
294 in the experience of pain.

295 To investigate this we calculated the PAF shift (Δ PAF) as the difference between central alpha component PAF during
296 prolonged pain and pain-free states). Δ PAF negatively correlated with pain intensity ($r = -0.50$, $p = .02$, Figure 3B),
297 indicating that PAF slowing is associated with increased pain. The average, absolute PAF shift across individuals was .05
298 Hz (s.d. = .05).



299
300 Figure 3. The relationship between PAF shifts (Δ PAF) from pain-free to prolonged pain states and pain intensity.

301 (A) Central component PAF at pain-free state was highly correlated with central component PAF during

302 prolonged pain, suggesting PAF is a relatively stable measure. (B) Δ PAF correlated with pain intensity. I.e.,
303 individuals whose PAF slowed during the prolonged pain state relative to pain-free state reported greater pain
304 intensity. (C) There was no relationship between an individual's pain-free state PAF and Δ PAF, suggesting that
305 these two metrics independently predict pain sensitivity.

306 ***PAF and Δ PAF provide distinct information about pain intensity***

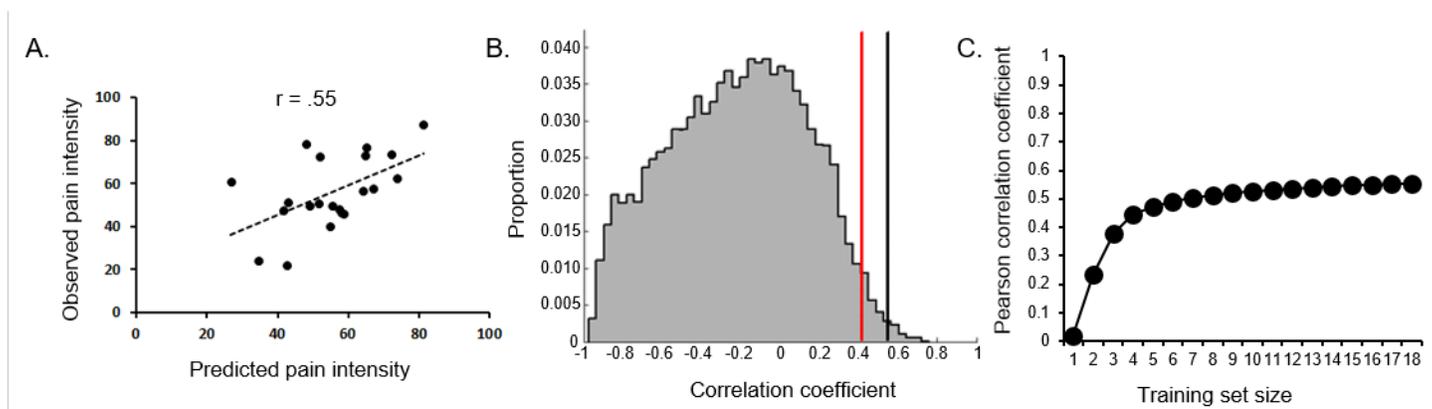
307 Despite showing quantitatively similar relationships to pain intensity, central component Δ PAF and pain-free state
308 central component PAF were uncorrelated ($r = .05, p = .82$, Figure 3C), suggesting that pain-free state PAF and Δ PAF
309 represent distinct elements of pain sensitivity.

310 To formally test the degree to which pain-free state central PAF and central Δ PAF independently predict pain sensitivity,
311 we performed a hierarchical regression using pain sensitivity as the dependent variable and pain-free state, central
312 component PAF and central component Δ PAF as independent variables entered first and second, respectively, into the
313 model. The full regression model significantly predicted pain intensity ($F(2,18) = 10.72, p < .01$) with an adjusted R^2 of
314 $.493$, indicating that pain-free state central PAF and Δ PAF accounted for nearly 50% of the variance in pain intensity.
315 Importantly, addition of pain-free state PAF ($\beta = -.543, p < .01$) and Δ PAF ($\beta = -.47, p < .01$) each yielded significant
316 changes to the R^2 of the regression model (Pain-free state $\Delta R^2 = .323, \Delta F = 9.065, p < .01$; Shift $\Delta R^2 = .221, \Delta F = 8.70, p <$
317 $.01$). Taken together, this analysis provides evidence that PAF characteristic to an individual, indexed by pain-free state
318 central component PAF, and the extent to which PAF is modulated by prolonged pain, indexed by central component
319 Δ PAF, are distinct mechanisms whose action play an important role in determining pain sensitivity.

320 ***PAF and Δ PAF can be used to predict pain intensity***

321 To further assess the robustness of our finding that pain-free state central component PAF and its changes in response
322 to the C-HP model are predictive of pain sensitivity, we performed a leave one out regression analysis. In brief, we
323 generated a series of regression models using pain-free state PAF and Δ PAF from 20 of the 21 individuals (training set)
324 and then used the resulting model to generate a pain prediction for the withheld test individual. Each individual served
325 as the test for exactly one regression model, yielding a total of 21 regression models and 21 predictions. The Pearson
326 correlation between predicted pain intensity and actual pain intensity was $r = .55$ (Figure 4A). This observed relationship

327 surpassed the 95th percentile of a null distribution of r values generated using permuted PAF measures and pain
328 intensity ($r = .22$), indicating that the two PAF measures can be used to predict pain intensities at a level greater than
329 chance (Figure 4B).



330

331 Figure 4. Individual pain sensitivity can be predicted. (A) Correlation between actual pain intensity and the pain
332 intensity predicted by the leave one out regression approach using pain-free state central component PAF and
333 Δ PAF. (B) Histogram of correlation values for a null distribution of pain and PAF indices. Correlation values were
334 obtained by randomly assigning PAF indices to pain intensity and then performing the same leave one out
335 approach as before. The red line indicates the 95th percentile of the null distribution and the black line indicates
336 the correlation value obtained in the actual leave out approach. (C) Correlation between predicted and observed
337 pain scores obtained using a regression approach with a range of training set sizes ranging from three to twenty
338 individuals. The model stabilizes with a training set of about 6, supporting the robustness of the prediction.

339 To ensure that the apparent ability of pain-free state central component PAF and central component Δ PAF to predict
340 pain intensity was not specific to this leave one out approach, we repeated the above analysis with training set sizes that
341 ranged from 3 individuals to 20. Within a training set size, separate regression models were generated for all the unique
342 combinations of participants; models were then evaluated together as the Pearson correlation between all predicted
343 pain intensity and all observed pain intensity. As can be seen in Figure 4C, prediction became stable around a training
344 set size of 6 ($r = .49$) and increased a relatively small amount to the maximum training size of 20 (.55). This suggests that
345 our ability to predict future pain intensity from pain-free state PAF and Δ PAF to predict pain intensity is robust and not
346 altered by the cross-validation procedures we employed.

347 Discussion

348 The personal experience of pain is highly variable, even when the underlying tissue damage is identical. While previous
349 research has found some genetic and psychological factors influencing pain susceptibility, methods to reliably predict
350 pain intensity consequent to medical intervention are lacking. Here we report that the peak alpha frequency and its
351 shifts over time, measured using EEG, were negatively related to the subjective pain intensity experience during induced
352 prolonged pain. Specifically, slower PAF during the pain-free state and a shift to slower PAF (Δ PAF) during the prolonged
353 pain state were independently associated with higher pain intensity. Using these two metrics, we could predict
354 individual pain sensitivity. These observations taken together suggest that PAF could represent a brain biomarker of an
355 individual's predisposition to pain, which would have useful clinical applications.

356 PAF has previously been suggested as a putative biomarker for individual differences in the experience of pain (Nir et al.,
357 2010; Bazanova & Vernon, 2014). For healthy individuals, acute pain intensity is related to faster PAF both before and
358 during exposure to a noxious stimulus. In contrast, studies of chronic pain conditions have repeatedly demonstrated
359 slowing of PAF, but little is known about whether this change reflects disease severity, symptom severity, individual
360 vulnerabilities, or an interaction amongst the three. In the current study, we tested the hypothesis that PAF slowing
361 reflects the intensity of prolonged pain by measuring PAF from healthy individuals in response to the capsaicin-heat pain
362 model, which involves central sensitization (LaMotte RH, et al.1992; Lotsch J, et al 2015). In support of this hypothesis,
363 we demonstrated that PAF recorded from central components during pain-free or prolonged pain states are inversely
364 related to pain intensity. Also in support of our hypothesis, we found an inverse relationship between Δ PAF and
365 prolonged pain intensity, suggesting that slowing of the alpha rhythm promotes prolonged pain intensity.

366 Our finding that PAF recorded during pain-free and prolonged pain states are inversely related to pain intensity is
367 notable for two reasons. First, the direction of this relationship is distinct from what has been previously reported for
368 acute phasic pain (Nir et al., 2010; Nir et al., 2012), but consistent with reports of in chronic pain (Sarnthein et al., 2005;
369 de Vries et al., 2013). This likely reflects the different nature of the prolonged pain model compared to acute phasic
370 pain, with the CH-P model capturing at least some aspects of chronic pain (e.g. central sensitization), or the early

371 transition period to chronic pain (long lasting pain with peripheral nerve damage). Second, the ability of PAF recorded
372 during the pain-free state to predict future prolonged pain intensity indicates that PAF indexes mechanisms that
373 generate individual susceptibility sensitivity to prolonged pain. Our median split analyses provide strong support for this
374 interpretation: the most sensitive individuals demonstrated PAF that were, on average, slower both before and during
375 the pain state. In contrast, individuals with faster pain-free state PAF had a relatively less painful subsequent pain
376 experience. We believe the median split analysis might have clinical relevance, since given identical injuries some
377 individuals will develop persistent pain, while others will heal and be pain free. Taken together, we believe these
378 findings suggest not only that PAF can predict the magnitude of future, prolonged pain but may also set the stage for
379 PAF as a biomarker for distinguishing healthy and pathological pain. One intriguing implication of our findings is that the
380 slowing of alpha frequency observed in chronic pain patients is not solely a reflection of the changes in the brain
381 brought about by the constant experience of pain, but that slower alpha frequency might have represented sensitivity to
382 develop chronic pain in the future.

383 We also observed that across individuals, changes in alpha frequency in the prolonged pain state relative to the pain free
384 state (Δ PAF), were inversely related to the subjective pain experienced. This is the first study to our knowledge
385 demonstrate a relationship between Δ PAF and pain. The magnitude of Δ PAF was small (~ 0.05 Hz) and future
386 investigations are needed to determine how these shifts represent meaningful changes in behavior. We here speculate
387 that the slowing of PAF reflects a maladaptive change in the alpha state leading these individual to experience more
388 pain. Conversely, the stability or increasing of PAF might reflect an adaptive response leading to pain resiliency.

389 An important result from the current study was that Δ PAF is independent of pain-free state PAF. This finding suggests a
390 potential new avenue for future pain treatments that use pain-free state PAF to identify high-risk individuals and
391 generate interventions that aim to prevent injury induced changes in PAF. In fact, we believe that the current findings
392 position PAF as a promising biomarker for treating and evaluating pain. Post-operative pain can sometimes lead to
393 chronic pain, and one of the best predictors of chronicity is pain intensity immediately following surgery (Katz et al.,
394 1996). Thus, by predicting pain sensitivity following surgery with a simple metric such as alpha activity, patients at
395 greater risk of developing chronic pain could be identified before the procedure begins, and appropriate measures could

396 be taken (e.g. pre- and post-operative pain management, or in some cases avoiding surgical interventions). Shifting PAF
397 through transcranial alternating current stimulation (tACS) has been shown to affect perceptual ability (Samaha et al.
398 2015; Cecere et al., 2015) and similar approaches could be used to modulate PAF for prophylactic and interventional
399 pain treatments.

400 Although it is tempting to speculate that the central independent component indexes this cortical hyper-excitability, the
401 precise anatomical localization identity of the neural substrate giving rise to this component cannot be stated with any
402 certainty. Inferring the location of EEG dipoles is always hazardous as different combinations of generators can give rise
403 to the same apparent source (the so called “inverse problem” of EEG). For example, while 8-14 Hz “mu” rhythms
404 originating from somatosensory cortex are modulated by painful stimulation (Ploner et al., 2006) combined EEG-fMRI
405 studies have also suggested a coupling between scalp recorded alpha power and blood-oxygenation levels in the
406 anterior cingulate cortex (Goldman et al., 2002). At present, both neural sources seem like equally good candidates for
407 generating the independent component used in this study. Ultimately, future studies incorporating techniques, such as
408 fMRI, that are better equipped to resolve the spatial identity of the currently sample source will be needed to fully
409 resolve this question.

410 It is important to acknowledge that the current study cannot determine whether PAF or PAF changes index the actual
411 experience of pain as opposed to any process that may co-vary with it, such as the salience of the stimulus or the
412 attention an individual pays to it. Importantly, our finding that PAF measured before capsaicin administration can
413 reliably predict pain sensitivity provides some evidence that PAF does not index these confounding factors directly.
414 Along similar lines, the pain intensity in our study and the Nir et al. (2010) study was relatively well matched, suggesting
415 that potentially confounding factors such as stimulus saliency should be even across the studies and unable to account
416 for the difference in findings.

417 In summary, we provide novel data supporting the hypothesis that slowing of PAF is associated with prolonged pain
418 intensity. These results extend previous findings that linked PAF and chronic neuropathic pain conditions, and suggest
419 that slowing of PAF can be used as a potential marker of prolonged pain sensitivity, as well as a possible mechanism for
420 understanding transitions from acute to chronic pain. The distinct mechanism we identified – PAF and Δ PAF – could

421 provide a number of innovative approaches for understanding, diagnosing, and treating chronic pain. Finally, slow alpha
422 rhythms appear to have a specific relationship to prolonged pain and interventions that directly manipulate these
423 rhythms may represent a viable means to prevent the transition from acute to chronic pain. Future work directly
424 elucidating the neural mechanisms underlying our observation could offer new fundamental insights into how changes
425 in neural oscillations shape the pain experience.

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