

1 **Title:** Echolocation reverses information flow in a cortical vocalization network.

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18

19 **Abstract**

20 The mammalian frontal and auditory cortices are important for vocal behaviour. Here, using
21 local field potential recordings, we demonstrate for the first time that the timing and spatial
22 pattern of oscillations in the fronto-auditory cortical network of vocalizing bats (*Carollia*
23 *perspicillata*) predict the purpose of vocalization: echolocation or communication. Transfer
24 entropy analyses revealed predominantly top-down (frontal-to-auditory cortex) information
25 flow during spontaneous activity and pre-vocal periods. The dynamics of information flow
26 depended on the behavioural role of the vocalization and on the timing relative to vocal onset.
27 Remarkably, we observed the emergence of predominantly bottom-up (auditory-to-frontal
28 cortex) information transfer patterns specific echolocation production, leading to self-directed
29 acoustic feedback. Electrical stimulation of frontal areas selectively enhanced responses to
30 echolocation sounds in auditory cortex. These results reveal unique changes in information
31 flow across sensory and frontal cortices, potentially driven by the purpose of the vocalization
32 in a highly vocal mammalian model.

33 **Introduction**

34 Vocal production is a crucial behaviour that underlies the evolutionary success of various
35 animal species. Several cortical and subcortical structures in the mammalian brain support
36 vocalization¹, their activities related to vocal control^{2,3,4}, motor preparation^{2,5,6}, and
37 feedback correction^{7,8}. However, the precise neural dynamics that underpin vocal production,
38 and the nature of long-distance interactions in large-scale neural networks related to vocal
39 utterance, remain poorly understood.

40 The connectivity patterns of the frontal cortex make it a major hub for cognitive control and
41 behavioural coordination^{9,10,11}. Frontal cortical areas are anatomically connected with
42 structures directly involved in vocal production, such as the periaqueductal grey¹² and the
43 dorsal striatum¹³. Experimental evidence demonstrates that the neural activity in frontal
44 regions relates to vocalization^{4,14,15,16}, correlating with the acoustic and behavioural
45 properties of produced calls^{14,16}. Frontal regions are also anatomically and functionally
46 connected with the auditory cortex (AC; ^{17,18,19,20,21,22}), which exhibits suppression to self-
47 produced sounds generated by body movements^{23,24,25,26} or vocalizations^{27,28,29,30}. This
48 suppression is thought to be mediated by preparatory motor signals originating in the motor
49 system (i.e. “corollary discharges” or “efference copies”; ^{31,32,33}). The attenuation of neural
50 responses in AC during vocal production supports precise vocal control by means of feedback
51 mechanisms^{7,8} in coordination with frontal cortical areas^{34,35,36,37}. Although current
52 evidence shows that fronto-auditory circuits are essential for the accurate control of
53 vocalization, neural interactions in these networks remain obscure.

54 In this study, we addressed the neural mechanisms of vocal production in the fronto-auditory
55 system of a highly vocal mammal: the bat *Carollia perspicillata*^{38,39,40,41}. Bats constitute an
56 excellent model to study the underpinnings of vocalization because they rely heavily on vocal
57 behaviour for communication and navigation (i.e. echolocation). Communication and
58 echolocation utterances are vocalized for very different behavioural purposes, and typically
59 differ markedly in their spectrotemporal design⁴⁰. The production of these calls is distinctly
60 controlled at the level of the brainstem⁴², possibly mediated by frontal cortical circuits
61 involving regions such as the anterior cingulate cortex⁴³ and the frontal-auditory field (FAF;
62 ¹⁶).

63 Vocalization circuits were studied by measuring local-field potential (LFP; ⁴⁴) oscillations
64 simultaneously in frontal and auditory cortex regions of vocalizing bats. In frontal and

65 sensory cortices, LFPs are involved in cognitive processes, sensory computations, and
66 interareal communication via phase coherence^{17, 45, 46, 47, 48, 49}. In the FAF, a richly connected
67 auditory region of the bat frontal cortex^{20, 50}, LFP activity predicts vocal output and
68 synchronizes differentially with dorso-striatal oscillations according to vocalization type¹⁶.
69 Oscillations in the bat frontal cortex synchronize across socially interacting bats⁵¹, and seem
70 to be involved in the cognitive aspects of social exchange⁵². The roles of auditory cortical
71 oscillatory activity in vocal production are less clear, although human studies suggest that
72 oscillations mediate synchronization with frontal and motor areas for feedback control^{35, 53, 54}.
73 However, the precise dynamics of information exchange in the fronto-auditory circuit during
74 vocalization are unknown.

75 The goal of this study was to unravel the nature of information exchange in the bat's FAF-AC
76 network, and to understand whether information between these structures flows in accordance
77 with the canonical roles of the frontal cortex for vocal coordination, and of the AC for
78 feedback control. We further aimed to address whether the distinct behavioural contexts of
79 echolocation and communication affect the dynamics of information transfer in the fronto-
80 auditory circuit. We found complex causal interactions (within a transfer entropy framework)
81 between frontal and auditory cortices during spontaneous activity and periods of vocal
82 production. Directed connectivity in the FAF-AC network varied dynamically according to
83 whether animals produced communication or echolocation calls, and to the timing relative to
84 vocal onset. For echolocation the changes were drastic, resulting in a reversal of information
85 flow from pre-vocal to post-vocal periods. Altogether, our data suggest that dynamic
86 information transfer patterns in large-scale networks involved in vocal production, such as the
87 FAF-AC circuit, are shaped by the behavioural consequences of produced calls.

88 **Results**

89 Neural activity was studied in the FAF and AC of *C. perspicillata* bats (3 males) while
90 animals produced self-initiated vocalizations. From a total of 12494 detected vocalizations,
91 138 echolocation and 734 non-specific communication calls were preceded by a period of
92 silence lasting at least 500 ms and were therefore considered for subsequent analyses.
93 Representative echolocation and communication vocalizations are shown in **Fig. 1a**. Overall,
94 the two types of vocalizations did not differ significantly in terms of call length (Wilcoxon
95 rank sum test, $p = 0.78$; **Fig. 1b**), although call length distributions differed significantly (2-
96 sample Kolmogorov-Smirnov test, $p = 7.93 \times 10^{-7}$). There were clear differences in the power

97 spectra of echolocation and communication calls (**Fig. 1c**, left), such that peak frequencies of
98 echolocation utterances were significantly higher than their communication counterparts
99 (Wilcoxon rank sum test, $p = 2.24 \times 10^{-66}$; **Fig. 1c**, right). These differences are consistent with
100 the structure of echolocation and communication sounds in bats (*C. perspicillata*) described in
101 previous studies {Knornschild, 2014 #245}{Porter, 1979 #374}.

102 Oscillations in frontal and auditory cortices predict vocalization type

103 **Figure 1d** illustrates electrophysiological activity recorded simultaneously from FAF and AC
104 at various cortical depths, as the echolocation and communication vocalizations shown in **Fig.**
105 **1a** were produced (see location of recording sites in **Fig. S1**). Single-trial LFP traces revealed
106 conspicuous pre-vocal oscillatory activity in low and high-frequencies, more pronounced in
107 frontal regions, and strongest when animals produced echolocation pulses. Power spectral
108 densities (PSD) obtained from pre-vocal LFP segments (i.e. -500 to 0 ms relative to vocal
109 onset; **Fig. 1f**) indicated low- and high-frequency power increases (relative to a no-
110 vocalization baseline, or “no-voc”) associated with vocal production, particularly in FAF and
111 for electrodes located at depths $> 100 \mu\text{m}$ (**Fig. 1e** illustrates this at $300 \mu\text{m}$; black arrows).
112 Differences in AC across types of vocal outputs were less pronounced and appeared limited to
113 low LFP frequencies (grey arrows in **Fig. 1e**). These pre-vocal spectral patterns were analysed
114 using canonical LFP frequency bands, namely: delta (δ), 1-4 Hz; theta (θ), 4-8 Hz; alpha (α),
115 8-12 Hz; low beta (β_1), 12-20 Hz; high beta (β_2), 20-30 Hz; and three sub-bands of gamma
116 (γ): γ_1 (30-60 Hz), γ_2 (60-120 Hz), and γ_3 (120-200 Hz). Pre-vocal LFP power in each band
117 normalized to no-voc periods on a trial-by-trial basis.

118 There were significant power changes between no-voc and pre-vocal periods across frequency
119 bands (**Fig. 1f, S2**). Notably, the power increase in low- (δ - α) and high-frequency (γ_2) LFP
120 bands of the FAF was different when animals produced echolocation and communication
121 vocalizations, with the highest increase in the pre-vocal echolocation case. The opposite
122 pattern was observed in the AC, where differences between ensuing vocalization types were
123 most prominent in β_1 (but not δ - α or γ) frequencies, and were explained by higher pre-vocal
124 power increase for communication than for echolocation vocalizations (**Fig. 1f**).

125 We addressed whether pre-vocal LFP power in frontal and auditory cortices was a significant
126 predictor of ensuing call type. To this effect, generalized linear models (GLMs) were fit using
127 echolocation and communication pre-vocal power changes as predictors, for all channels (in
128 both structures) and frequency bands. A summary of these models is given in **Fig. 1g** (see

129 outcomes of two representative GLMs illustrated in **Fig. S2**). Low- and high-frequency power
130 increase (mostly in the δ - α and γ_2 bands) in FAF predicted whether animals produced
131 echolocation or communication calls, typically with moderate effect sizes ($p < 0.05$; $R^2_m \geq$
132 0.1), highest in middle-to-deep electrodes (i.e. depths $> 300 \mu\text{m}$; **Fig 1g**, left). In the AC, pre-
133 vocal power predicted ensuing call type mostly in the α - β bands of the spectrum, although
134 more strongly in β_1 frequencies. Moderate effect sizes were also observed ($p < 0.05$; $R^2_m \geq$
135 0.1). Overall, pre-vocal oscillatory power significantly predicted ensuing call type in frontal
136 and auditory cortices with complementary frequency specificity and functionally opposite
137 effects.

138 We evaluated whether differences in the spectral dynamics of pre-vocal LFPs could be
139 explained by differences in the frequency content of echolocation and communication
140 utterances. To that end, communication calls were separated into two groups: high-frequency
141 and low-frequency communication (HF- and LF-communication, respectively). The spectral
142 content of pre-vocal LFPs predicted ensuing call type, even when HF-communication calls
143 were pitched against echolocation utterances ($p < 0.05$; $R^2_m \geq 0.1$; **Fig. S3**). Additionally,
144 pre-vocal spectral differences were considerably less noticeable when comparing HF- vs. LF-
145 communication vocalizations, with even significant models ($p < 0.05$) performing poorly in
146 FAF and AC ($R^2_m < 0.1$; **Fig. S3b, d**). Thus, pre-vocal spectral differences are not fully
147 accounted for by differences in the spectral content of echolocation and communication calls.

148 Directed connectivity in the FAF-AC circuit related to vocal production

149 Oscillations in FAF and AC predict ensuing vocal output with functionally opposite patterns,
150 but how rhythms in this network interact during vocal production remains unknown. In
151 previous work, we reported low-frequency (1-12 Hz) phase coherence in the FAF-AC circuit
152 during spontaneous activity, with emergence of γ -band (> 25 Hz) coherence at the onset of
153 external acoustic stimulation¹⁷. To study FAF-AC oscillatory dynamics during vocal
154 production, we looked beyond phase correlations and examined causal interactions (within a
155 transfer entropy framework) in the fronto-auditory circuit. Causal interactions were quantified
156 using directed phase transfer entropy (dPTE), a metric that measures the degree of preferential
157 information transfer between signals based on phase time series^{55, 56}. dPTE calculations were
158 performed across vocal conditions for all channel pairs, and for frequency bands which most
159 strongly predicted vocalization type: δ , θ , α , β_1 , and γ_2 .

160 Average dPTE connectivity matrices across conditions (echolocation and communication pre-
161 and post-vocal periods, and no-voc segments) are illustrated in **Fig. S4**. dPTE matrices were
162 used as adjacency matrices for directed graphs, which characterized patterns of directional
163 information flow in the FAF-AC network (**Fig. 2**). In each graph, nodes represent adjacent
164 channels pooled according to cortical depth and layer distribution in AC (where layer borders
165 are well-defined anatomically; **Fig. S1**): superficial (*sup*), channels 1-4 (0-150 μm); top-
166 middle (*mid1*), channels 5-8 (200-350 μm); bottom-middle (*mid2*), channels 9-12 (400-550
167 μm); and *deep*, channels 13-16 (600-750 μm). Directed edges were weighted according to a
168 directionality index (DI), obtained from normalizing dPTE values to 0.5 (dPTE = 0.5
169 indicates no preferred direction of information transfer). Only edges with significant DI
170 values, based on bootstrapping, are shown.

171 Upon inspection of the connectivity graphs, we noticed general patterns that entailed strong
172 top-down preferential information transfer (i.e. in the FAF \rightarrow AC direction) during
173 spontaneous activity, pre-vocal periods irrespective of vocalization type, and post-vocal
174 communication periods (**Fig. 2**). Top-down information flow (blue arrows in the figure) was
175 strongest for δ , θ , and γ_2 frequencies, although also occurred sparsely in α and β_1 bands with
176 patterns that depended on ensuing call type. Within FAF and during pre-vocal echolocation
177 periods, information flowed predominantly from deep to superficial layers in δ and β_1
178 frequencies (**Fig. 2b**), and in the opposite direction for α -LFPs, also during no-voc periods
179 (**Fig. 2a, b**). Within AC, information flowed in the superficial to deep direction during no-voc
180 and pre-vocal communication periods in γ_2 frequencies.

181 To our surprise, a predominance of bottom-up information flow (i.e. AC \rightarrow FAF direction)
182 appeared to be specific to post-vocal echolocation periods in the δ and β_1 bands, although
183 bottom-up information transfer did occur in α frequencies during post-vocal communication
184 epochs (**Fig. 2c**). Note that, for echolocation, there was strong to-down information transfer
185 before vocalization onset, particularly in the δ -band (cf. **Fig. 2c** with **Fig. 2b**, top). These
186 results hint toward a pre- to post-vocal reversal of information flow in the FAF-AC network
187 during echolocation, evident in low frequencies of the LFP. Considering within-structure
188 information transfer, patterns were diverse in FAF, consisting of information exchange in the
189 deep-to-superficial (bands: δ , echolocation and communication; α , communication; and β_1 ,
190 both) and the superficial-to-deep (α -band, for echolocation) directions. Within AC,
191 predominant information flow occurred both in the superficial-to-deep (δ -band, echolocation)

192 and in the deep-to-superficial directions (bands: θ , echolocation; α , echolocation and
193 communication; β_1 , echolocation, γ_2 , communication) as well.

194 Taken together, the data in **Fig. 2** illustrate rich patterns of information exchange within and
195 between frontal and auditory cortices. Information transfer patterns depended on whether a
196 vocalization was produced and on its type, either considering within-structure connectivity, or
197 information transfer across regions. Differences in the information flow dynamics of the
198 fronto-auditory circuit, across vocal conditions and call-types, are depicted in **Figures S5** and
199 **S6**, and quantified in detail in the supplementary results.

200 Information flow in the fronto-auditory circuit reverses for echolocation production

201 The contrasts between pre- and post-vocal periods for echolocation (**Fig. 2b, c**) suggest that
202 the preferred directionality of information flow reverses when animals echolocate.
203 Differences in the direction of information transfer between pre-vocal and post-vocal
204 activities were addressed by statistically comparing connectivity graphs associated to each
205 case (**Fig. 3**). Paired statistics were performed for these comparisons (Wilcoxon signed-rank
206 tests, significance when $p < 10^{-4}$; see Methods); edges, representing significant differences in
207 dPTE, were only shown for large effect sizes ($|d| > 0.8$). As expected from the data depicted in
208 **Figure 2**, echolocation-related FAF \rightarrow AC preferred information flow was significantly
209 higher during pre-vocal than post-vocal periods in the δ and θ bands (**Fig. 3a**, top). In γ_2
210 frequencies, the effect was the opposite: FAF \rightarrow AC directionality was highest during post-
211 vocal periods than during pre-vocal ones. Remarkably, AC \rightarrow FAF preferred directionality of
212 information flow was significantly stronger during post-vocal periods in δ and β_1 frequency
213 bands (**Fig. 3a**, top). Within FAF, differences in preferred information flow occurred in
214 frequency bands δ , α , and β_1 . Within AC, differences in dPTE occurred mostly in α and β_1
215 bands (**Fig. 3a**, top). Information flow was strongest in the deep-to-superficial direction
216 during post-vocal periods, and in superficial-to-deep directions during pre-vocal periods. In
217 the case of communication call production (**Fig. 3b**, top), differences in dPTE occurred only
218 in the δ and θ bands. Values were significantly higher (with large effect sizes) in the FAF \rightarrow
219 AC direction during pre-vocal periods.

220 Changes in directional information transfer in the FAF-AC network were quantified by
221 calculating the net information outflow (DI_{net}) from each area. The DI_{net} represents the sum of
222 DI values obtained from outgoing connections per region (e.g. all edges in FAF related to
223 FAF \rightarrow AC connections, representing the net strength of preferential information flow in the

224 fronto-auditory direction). DI_{net} values were used to statistically compare pre- and post-vocal
225 periods in terms of information transfer from one cortical area to another. Considering this
226 metric, significant differences (FDR-corrected Wilcoxon signed-rank tests, $p_{\text{corr}} < 0.05$) with
227 large effect sizes ($|d| > 0.8$) occurred mostly for low and intermediate frequency bands (i.e. δ
228 and β_1) of the LFP. Specifically, for the pre-vocal vs. post-vocal echolocation condition (**Fig.**
229 **3a**, bottom), the information outflow from FAF was significantly higher in the δ band during
230 pre-vocal periods related to echolocation call production ($p_{\text{corr}} = 1.63 \times 10^{-82}$, $d = 3.44$).
231 Notably, the net information outflow from AC was significantly higher when considering
232 post-vocal periods than pre-vocal ones ($p_{\text{corr}} = 3.27 \times 10^{-64}$, $d = -1.5$). In the β_1 frequency range,
233 there were no significant differences with large effect sizes between pre-vocal and post-vocal
234 net information outflow from the FAF. However, DI_{net} values from AC were significantly
235 higher during post-vocal periods ($p_{\text{corr}} = 3.71 \times 10^{-36}$, $d = -0.88$). Pre-vocal vs. post-vocal
236 comparisons of DI_{net} values from FAF and AC related to communication calls revealed
237 significant differences with large effect sizes only for δ frequencies in FAF (**Fig. 3b**, bottom;
238 DI_{net} higher for pre-vocal periods: $p_{\text{corr}} = 1.63 \times 10^{-82}$, $d = 1.39$).

239 The passive listening of echolocation-like or communication sounds did not account for the
240 data above (**Fig. S7-9**), suggesting that mere feedback from the calls was not sufficient to
241 explain echolocation-related, δ -band information flow reversal in the network. Likewise, the
242 production of HF-communication sounds did not account for the patterns observed during
243 echolocation (**Fig. S10**), indicating that the information transfer dynamics for echolocation are
244 not fully explained by the frequency content of the vocalizations themselves. Altogether,
245 these results unveil dynamic changes of predominant connectivity patterns in the FAF-AC
246 network from pre- to post-vocal periods, exhibiting frequency specificity and particularly
247 associated with echolocation production.

248 Electrical stimulation of the FAF enhances auditory cortical responses

249 The data thus far indicate strong top-down modulation in the FAF-AC network, which can
250 nevertheless be significantly altered when animals produce echolocation sounds. However,
251 the dPTE analyses cannot rigorously establish whether FAF activity indeed modulates AC
252 responses. To examine this question, we conducted perturbation experiments of the FAF to
253 evaluate whether manipulations in this region affect auditory cortical responses to external
254 sounds.

255 The FAF was stimulated electrically with biphasic pulse trains (6 pulses/train; pulse interval:
256 500 ms) while simultaneously recording from the AC (n = 20 penetrations; **Fig. 4a**). Electrical
257 stimulation of FAF did not produce detectable artefacts or LFP power changes in AC (**Figs.**
258 **4c** and **S11**), nor did it elicit vocalization production (**Fig. S12**), potentially due to weaker
259 stimulation as compared to previous work with measurable behavioural outputs^{43,57}. Acoustic
260 stimuli were presented after the train (either a distress -a type of communication sound- or an
261 echolocation call; see **Fig. S7b**, and **Fig. 4b**) at different delays. Response strengths in AC to
262 sounds after FAF electrical stimulation (“Estim” condition) were compared to response
263 strengths related to the same sounds, but presented without prior electrical stimulation (“no-
264 Estim” condition). Representative responses to distress and echolocation calls for both
265 conditions and for a delay of 135 ms (in the Estim case) are depicted in **Fig. 4d**.

266 We observed differences between the ERP energy measured in Estim and no-Estim conditions
267 (red and blue, respectively), more evidently when considering AC responses to echolocation
268 sounds (**Fig. 4d**). These differences occurred consistently at a population level. **Figure 4e**
269 depicts response strengths for all AC depths across recordings, related to a distress syllable
270 presented with a delay of 135 ms (in the Estim case; red traces). Response strengths from the
271 no-Estim case are shown for one example iteration out of 500 conducted for comparisons (see
272 Methods). A trend was present, wherein responses in the Estim condition were stronger than
273 those in the no-Estim condition, although without statistically significant differences (**Fig. 5f**;
274 FDR-corrected Wilcoxon signed-rank tests, significance if $p_{\text{corr}} < 0.05$). Overall, when
275 acoustic stimulation was done with a distress syllable, significant differences between Estim
276 and no-Estim occurred in ~40 % of out 500 iterations in total. Such differences were
277 concentrated mostly in middle-to-deep layers (depths > 300 μm) with small to medium values
278 of Cliff’s delta (⁵⁸; **Fig. 4i, j, left**).

279 In terms of AC responses to echolocation sounds (**Fig. 4g**), differences between Estim and no-
280 Estim conditions appeared most prominent in superficial-to-middle layers (depths 50 – 350
281 μm). Responses were significantly stronger in the Estim condition than in the no-Estim
282 condition at depths of 50 – 300 μm (**Fig. 4h**; $p_{\text{corr}} < 0.05$), in particular for a sound-delay of
283 135 ms. Significant differences between Estim and no-Estim conditions were very reliable,
284 observed in up to 90.6 % of the iterations for a delay of 135 ms and a cortical depth of 150
285 μm (**Fig. 4i, right**). At depths ranging 50-350 μm , for the same delay, reliability was larger
286 than 70 %, with medium effect sizes (**Fig. 4j, right**). These data indicate that electrical

287 stimulation on the FAF enhances response strength in the AC, with particularly high
288 reliability when the animals listen to echolocation sounds.

289 **Discussion**

290 In this study, we addressed the dynamics of information exchange between frontal and
291 auditory cortices of vocalizing bats (summarized in **Fig. 5**). Consistent with previous reports ^{4,}
292 ^{14, 16, 35, 46, 59}, we show that neural activity in the frontal cortex predicts vocal outputs. Taken
293 together, the data from this and previous work suggest that oscillations in frontal regions may
294 be instrumental for vocal production. From our perspective, the above is further supported by
295 call-type specific, pre-vocal LFP spectral dynamics and information transfer patterns in the
296 FAF-AC network. The relationship between oscillations and vocal production remains,
297 nevertheless, correlational: our results do not allow to rigorously assert a causal role of LFPs
298 for the initiation or planning of vocal outputs.

299 Neural activity in the AC also relates to vocalization ²⁷, but the involvement of auditory
300 cortical oscillations in vocal production is still to be fully understood. Our results indicate that
301 pre-vocal LFPs in AC, as previously reported with single-unit spiking ^{27, 29, 30}, relate to vocal
302 initiation. We show, for the first time to our knowledge, that pre-vocal oscillatory patterns in
303 AC are call-type specific and, remarkably, complementary to those observed in frontal cortex
304 in frequency and effect (**Fig. 1**). These patterns may be explained by our current
305 understanding of the roles of AC for vocal production. Neuronal activity in the AC is
306 predominantly suppressed during vocalization, with inhibition occurring hundreds of
307 milliseconds prior to call onset ^{8, 27, 28, 60}. Vocalization-related inhibition is mediated by motor
308 control regions, which send a copy of the motor command to the AC as “corollary discharge”
309 or “efferent copy” signals ^{33, 61}. These signals, respectively, have either a general suppressive
310 effect, or carry specific information about the produced sound which potentially facilitates
311 feedback processing ³². Thus, pre-vocal, call-type unspecific power changes in low
312 frequencies could reflect general inhibitory mechanisms in AC consistent with corollary
313 discharges from higher order structures. Directed connectivity analyses support the notion of
314 top-down (FAF → AC) control of pre-vocal low-frequency activity (**Fig. 2**). In contrast, pre-
315 vocal β -band LFPs might constitute oscillatory correlates of efference copies, given the
316 observed call-type specificity. Because FAF → AC causal influences did not equally extend
317 to the β frequencies, pre-vocal β activity in AC might be influenced instead by specialized
318 regions such as the premotor cortex, providing a more specific copy of the motor commands

319 required for vocalization. Channels for motor-auditory communication (see ^{62, 63}) could in fact
320 operate over β frequencies ^{53, 64, 65}.

321 Differences in spectral patterns cannot be solely explained by the distinct frequency content of
322 echolocation and communication calls (**Fig. S3**). However, considering that orofacial
323 movement in primates ^{15, 66} and vocalization-specific movements in bats ⁵⁰ are associated to
324 neural activity in frontal areas, distinct pre-vocal motor related activity for echolocation or
325 communication calling is a plausible explanation for our results. Microstimulation of *C.*
326 *perpscillata*'s FAF can result in motor effects such as pinna and nose-leaf movements, as
327 well as vocalizations (including echolocation-like calls; ⁵⁰). These movements also occur
328 naturally before spontaneous vocalization ⁵⁰, suggesting that the FAF may be involved in the
329 motor aspect of vocal production. Nevertheless, vocalization-specific neural populations in
330 primates coexist with those related to orofacial movements ⁶⁷. Therefore, the vocal-motor
331 explanation does not necessarily entail that the FAF fails to participate in other forms of vocal
332 preparation beyond the orchestration of motor programs.

333 In terms of a cortical network for vocalization, the FAF and AC are engaged in rich
334 information transfer dynamics with functional relationships to vocalization. Moreover,
335 interactions extend to periods of vocal quiescence, when information flows top-down (FAF \rightarrow
336 AC) in low (δ - α) and high (γ_2) frequencies. Low-frequency top-down influences from higher-
337 order structures (like the FAF) modulate neuronal activity in sensory cortices according to
338 cognitive variables such as attention, also during spontaneous activity ^{55, 68, 69}. However,
339 whether and how attentional processes exploit the nature of neural connections in the FAF-
340 AC circuit remains thus far unknown. Our data resonate with the hypothesis of top-down
341 modulation of oscillatory activity in AC, and suggest a strict control of higher-order structures
342 over sensory areas reflected in concurrent LFP activity across regions. Such strong top-down
343 control is supported by the fact that FAF microstimulation enhances auditory cortical
344 responses to sounds (**Fig. 4**).

345 Vocalization-specific changes in power may affect causality estimations, e.g. by creating
346 confounding differences between the vocal conditions studied. However, the dPTE is a
347 causality estimate that shows robustness to the influence of power, noise, and other variables.
348 In our dataset, the pre-vocal δ -band power increase within each region when animals
349 produced echolocation (call-type specific in FAF, unspecific in AC) was nonetheless
350 accompanied by a decrease of interareal dPTE values. In addition, a δ -band power increase of

351 communication pre-vocal LFPs relative to baseline (**Figs. 1, S1**) did not result in significant
352 differences of dPTE values during pre-vocal and spontaneous periods. Thus, changes in
353 causality did not necessarily follow changes in power, as has been reported in previous work
354 {Hillebrand, 2016 #559}.

355 Based on dPTE values associated with spontaneous and pre-vocal activities (**Fig. 2** and **S5**), it
356 appears that as animals prepare the production of an echolocation call, the FAF gradually
357 relinquishes control over the AC in the low-frequency (δ) channel. That is, the top-down
358 control wanes during echolocation pre-vocal periods in δ LFPs. The weakening of preferred
359 top-down information transfer could be taken as a preamble of emerging bottom-up
360 information flow in δ frequencies after an echolocation call is emitted (**Fig. 3**). This fails to
361 happen in the communication case. Echolocation, the predominant strategy for navigation, is
362 essential for bats. After vocalizing an echolocation pulse, the bat auditory system must be
363 ready to process incoming echoes and use them to construct a representation of surrounding
364 objects⁷⁰, potentially involving higher order structures. The observed switch from top-down
365 to bottom-up processing when animals find themselves in echolocation mode could represent
366 the readiness of the bat's auditory machinery for the aforementioned task. Concretely, our
367 data suggest that the former may occur over a continuum encompassing a gradual release of
368 the AC from top-down influences (i.e. from the FAF), which opens the way for auditory-
369 frontal information transfer supporting the processing and integration of incoming echoes. A
370 reversal in information transfer is also visible (albeit weaker and in a different LFP frequency
371 band) during the production of high frequency communication sounds (**Fig. S10**). This result
372 could hint towards smooth transition in the way the FAF-AC network operates, which finds
373 its two extremes in echolocation and low-frequency communication call production.

374 In all, processing feedback information directly related to navigation appears to have a larger
375 weight in the bottom-up processing of acoustic cues resulting from a self-generated sound.
376 Echolocation pulses are produced to generate echoes that must be listened to. Communication
377 calls are often targeted to an audience as means of transmitting internal behavioural
378 information (e.g. distress, aggressive mood, etc.), not aimed at the emitter itself. For the
379 emitter, in such scenario, feedback processing mostly contributes to the adjustment of vocal
380 parameters such as loudness or pitch^{7, 71, 72}. Since in this study animals vocalized without an
381 audience (i.e. they were isolated in the recording chamber), further research could elucidate
382 whether the presence of conspecifics increases bottom-up information transfer when
383 vocalizing communication calls, as animals could expect a response.

384 The reversal of information flow reported in the δ -band when animals echolocate cannot be
385 solely attributed to passively hearing feedback from their own utterances (**Fig. S7-10**): active
386 vocalization seems to be necessary to trigger bottom-up information transfer in the FAF-AC
387 circuit. The current data, together with the fact that passive listening fails to significantly alter
388 low-frequency coherence in the FAF-AC network ¹⁷, indicate that passive listening alone is
389 not sufficient to significantly alter the dynamics of communication between FAF and AC.
390 Likewise, information flow dynamics associated to echolocation calls could not be attributed
391 to their high frequency content alone, since qualitatively similar differences between
392 echolocation and communication calls were observed when considering only HF-
393 communication utterances, particularly for δ -LFPs (**Fig. S10**). Echolocation, therefore,
394 triggers unique patterns of information flow reversal in the fronto-auditory network of *C.*
395 *perspicillata*. Our data indicates that the connectivity in this circuit is shaped by the
396 behavioural implications of the calls produced.

397 The transfer entropy analyses discussed above indicate that the frontal cortex exerts top-down
398 modulation over its auditory cortical counterpart, particularly during spontaneous activity
399 (**Fig. 2**). Consistent with the top-down modulation perspective, electrical perturbation of the
400 FAF enhanced the strength of responses in AC, echoing known effects of frontal stimulation
401 in other mammals ²¹. Given that FAF manipulation most reliably altered responses to
402 echolocation sounds (**Fig. 4**), and that the production of echolocation pulses reverses
403 information flow in the fronto-auditory circuit, the data suggest that dynamic interactions in
404 the bat's FAF-AC network favour echolocation behaviour.

405 **Methods**

406 Animal preparation and surgical procedures

407 The study was conducted on five awake *Carollia perspicillata* bats (one female).
408 Experimental procedures were in compliance with European regulations for animal
409 experimentation and were approved by the Regierungspräsidium Darmstad (experimental
410 permit #FU-1126 and FR-2007). Bats were obtained from a colony at the Goethe University,
411 Frankfurt. Animals used for experiments were kept isolated from the main colony.

412 Prior to surgical procedures, bats were anaesthetized with a mixture of ketamine (10
413 mg*kg⁻¹, Ketavet, Pfizer) and xylazine (38 mg*kg⁻¹, Rompun, Bayer). For surgery and for
414 any subsequent handling of the wounds, a local anaesthetic (ropivacaine hydrochloride, 2

415 mg/ml, Fresenius Kabi, Germany) was applied subcutaneously around the scalp area. A
416 rostral-caudal midline incision was cut, after which muscle and skin tissues were carefully
417 removed in order to expose the skull. A metal rod (ca. 1 cm length, 0.1 cm diameter) was
418 attached to the bone to guarantee head fixation during electrophysiological recordings. The
419 FAF and AC were located by means of well-described landmarks, including the sulcus
420 anterior and prominent blood vessel patterns (see ^{17, 50, 73}). The cortical surface in these
421 regions was exposed by cutting small holes (ca. 1 mm²) with the aid of a scalpel blade on the
422 first day of recordings. In the AC, recordings were made mostly in the high frequency fields
423 ^{17, 50, 73})

424 After surgery, animals were given no less than two days of rest before the onset of
425 experiments. No experiments on a single animal lasted longer than 4 h per day. Water was
426 given to the bats every 1-1.5 h periods, and experiments were halted for the day if the animal
427 showed any sign of discomfort (e.g. excessive movement). Bats were allowed to rest a full
428 day between consecutive experimental sessions.

429 Electrophysiological and acoustic recordings

430 Electrophysiology was performed chronically in fully awake animals, inside a sound-proofed
431 and electrically isolated chamber. Inside the chamber, bats were placed on a custom-made
432 holder which was kept at a constant temperature of 30 °C by means of a heating blanket
433 (Harvard, Homeothermic blanket control unit). Electrophysiological data were acquired from
434 FAF and AC on the left hemisphere, using two 16-channel laminar electrodes (one per
435 structure; Model A1x16, NeuroNexus, MI; 50 µm channel spacing, impedance: 0.5-3 MW per
436 electrode). Probes were carefully inserted into the brain perpendicular to the cortical surface,
437 and lowered with piezo manipulators (one per probe; PM-101, Science 455 products GmbH,
438 Hofheim, Germany) until the top channel was barely visible above the surface of the tissue.
439 The placing and properties of the probes allowed us to record simultaneously at depths
440 ranging from 0-750 µm, spanning all six cortical layers (see ⁷⁴). Probes were connected to a
441 micro-preamplifier (MPA 16, Multichannel Systems, MCS GmbH, Reutlingen, Germany),
442 and acquisition was done with a single, 32-channel portable system with integrated
443 digitization (sampling frequency, 20 kHz; precision, 16 bits) and amplification steps (Multi
444 Channel Systems MCS GmbH, model ME32 System, Germany). Acquisition was online-
445 monitored and stored in a computer using the MC_Rack_Software (Multi Channel Systems
446 MCS GmbH, Reutlingen, Germany; version 4.6.2).

447 Vocal outputs were recorded by means of a microphone (CMPA microphone, Avisoft
448 Bioacoustics, Glienicke, Germany) located 10 cm in front of the animal. Recordings were
449 performed with a sampling rate of 250 kHz and a precision of 16 bits. Vocalizations were
450 amplified (gain = 0.5, Avisoft UltraSoundGate 116Hm mobile recording interface system,
451 Glienicke, Germany) and then stored in the same PC used for electrophysiology.
452 Electrophysiological and acoustic data were aligned using two triggers, an acoustic one (5
453 kHz tone, 10 ms long) presented with a speaker located inside of the chamber (NeoCD 1.0
454 Ribbon Tweeter; Fountek Electronics), and a TTL pulse sent to the recording system for
455 electrophysiology (see above). Note that the onsets of the tones were in synchrony with the
456 TTL pulses registered by the acquisition system for electrophysiology.

457 Acoustic stimulation

458 Two acoustic stimuli were used to evaluate transfer entropy patterns during passive listening.
459 One of them, the high-frequency frequency modulated sound (HF-FM; 2 ms long; downward
460 frequency sweep from 80-50 kHz), mimicked the spectrotemporal structure of echolocation
461 pulses; the other, consisted of a distress syllable (distress, 3.8 ms long) typical of *C.*
462 *perspicillata*'s vocal repertoire. The latter stimulus was embedded in a sequence in which the
463 syllable was presented every 500 ms for 2 seconds (2 Hz rate); other sequences with faster
464 rates were also presented to the animal, but were not considered for this study. Only the first
465 syllable of the 2 Hz sequence was used for analyses. Stimuli for determining frequency tuning
466 consisted of short (10 ms) pure tones at various frequencies (5-90 kHz, in steps of 5 kHz) and
467 levels (15-75 dB SPL, steps of 15 dB). Since the HF-FM and the distress sounds were
468 presented at 70 dB SPL (rms), we focused on the frequency tuning curved obtained with pure
469 tone stimuli presented at 75 dB SPL.

470 The setup for stimulation has been described in previous studies (see ¹⁷). In short, sounds were
471 digital-to-analog converted using a sound card (M2Tech Hi-face DAC, 384 kHz, 32 bit),
472 amplified (Rotel power amplifier, model RB-1050), and presented through a speaker
473 (description above) inside of the chamber. The speaker was located 12 cm away from the bat's
474 right ear, contralateral to the cerebral hemisphere on which electrophysiological recordings
475 were made. Prior to stimulation, sounds were downsampled to 192 kHz and low-pass filtered
476 (80 kHz cut-off) Sound presentation was controlled with custom written Matlab softwares
477 (version 8.6.0.267246 (R2015b), MathWorks, Natick, MA) from the recording computer.

478 Classification of vocal outputs

479 Two sessions of concurrent acoustic recordings (~10 min long) were made per paired
480 penetrations in FAF and AC. Vocalizations were automatically detected based on the acoustic
481 envelope of the recordings. The envelope was z-score normalized to a period of no
482 vocalization (no less than 10 s long), which was manually selected, per file, after visual
483 inspection. If a threshold of 5 standard deviations was crossed, a vocalization occurrence was
484 marked and its start and end times were saved. Given the stereotyped spectral properties of *C.*
485 *perspicillata*'s echolocation calls, a preliminary classification between echolocation and
486 communication utterances was done based on each call's peak frequency (a peak frequency >
487 50 kHz suggested an echolocation vocalization, whereas a peak frequency below 50 kHz
488 suggested a communication call). In addition, vocalizations were labelled as candidates for
489 subsequent analyses if there was a time of silence no shorter than 500 ms prior to call
490 production to ensure no acoustic contamination on the pre-vocal period that could affect LFP
491 measurements in FAF or AC. Finally, echolocation and communication candidate
492 vocalizations were individually and thoroughly examined via visual inspection to validate
493 their classification (echolocation or communication), the absence of acoustic contamination in
494 the 500 ms prior to vocal onset, and the correctness of their start and end time stamps.
495 According to the above, and out of a total of 12494 detected vocalizations, 138 echolocation
496 and 734 communication calls were then used in further analyses.

497 High-frequency communication calls (HF-communication) were selected according the
498 frequency component of the vocalizations. Specifically, an HF-communication call was a
499 communication utterance with more than 50 % of its power in the 50-100 kHz range. HF-
500 communication calls represented 21.12% of the communication calls used (155/734).

501 Extraction of LFP signals and power analyses

502 Data analyses were performed using custom-written scripts in MatLab (versions
503 9.5.0.1298439 (R2018b), and 9.10.0.1684407 (R2021a)), Python (version 2.6 or 3.6), and R
504 (RStudio version 1.3.1073). For extracting LFPs, the raw data were band-pass filtered (zero-
505 phase) between 0.1 and 300 Hz (4th order Butterworth filter; *filtfilt* function, MatLab), after
506 which the signals were downsampled to 1 kHz.

507 All LFP spectral analyses were done using the Chronux toolbox ⁷⁵. Pre-vocal power was
508 calculated with LFP segments spanning -500-0 ms relative to vocal onset, using a TW of 2,

509 and 3 tapers. No-vocalization baseline periods (*no-voc*) with a length of 500 ms were pseudo-
510 randomly selected and their power spectra calculated in order to obtain baseline power values
511 for spontaneous activity. The total number of *no-voc* periods matched the total number of
512 vocalizations ($n = 872$), in a way that the number of selected *no-voc* periods per recording file
513 matched the number of vocalizations found in that particular file. The power of individual
514 frequency bands (i.e. δ , 1-4 Hz; θ , 4-8 Hz; α , 8-12 Hz; β_1 , 12-20 Hz; β_2 , 20-30 Hz; γ_1 , 30-60
515 Hz; γ_2 , 60-120 Hz; γ_3 , 120-200 Hz) was calculated by integration of the power spectral density
516 accordingly for each case. Finally, the increase of pre-vocal power relative to the baseline
517 periods was calculated as follows (per frequency band, on a call-by-call basis):

$$518 \quad \text{Relative power change} = \frac{BP_{pre-voc} - BP_{no-voc}}{BP_{no-voc}} * 100 \quad [1],$$

519 where $BP_{pre-voc}$ is the pre-vocal power (in the case of either an echolocation or communication
520 vocalization) of the given frequency band and a trial (i.e. a specific call), and BP_{no-voc} is the
521 baseline no-voc power associated to the same frequency band and trial.

522 Generalized linear model for vocal output prediction

523 To determine whether pre-vocal power change relative to baseline was able to predict the type
524 of ensuing vocal output, we used a GLM with a logistic link function (i.e. logistic regression).
525 The model analysis was done in Rstudio with the *lme4* package. In brief, logistic regression
526 was used to predict the probability of a binary outcome (0 or 1; communication or
527 echolocation, respectively) based on the pre-vocal power change as the predictor variable.
528 The probabilities are mapped by the inverse logit function (sigmoid):

$$529 \quad \sigma(x) = \frac{1}{1 + \exp(-x)} \quad [2],$$

530 which restricts the model predictions to the interval $[0, 1]$. Because of these properties, a
531 logistic regression with GLMs is well suited to compare data (and thus, evaluate predictions
532 of ensuing vocal-output) on a single-trial basis ⁷⁶.

533 To estimate the effect size of the fitted models, we used the marginal coefficient of
534 determination (R^2_m) with the *MuMIn* package. The R^2_m coefficient quantifies the variance
535 in the dependent variable (echolocation vs. communication vocalization) explained by the
536 predictor variable (i.e. the relative pre-vocal power change). This value is dimensionless and
537 independent of sample size ^{76, 77}, which makes it ideal to compare effect sizes of different

538 models (e.g. across channels and frequency bands, as in **Fig. 1g**). Effect sizes were considered
539 small when $R^2m < 0.1$, medium when $0.1 \leq R^2m < 0.4$, and large when $R^2m \geq 0.4$ ⁷⁶.

540 Directionality analyses

541 Directional connectivity in the FAF-AC network was quantified with the directed phase
542 transfer entropy (dPTE; ⁵⁵), based on the phase transfer entropy (PTE) metric ⁷⁸. PTE is a
543 data-driven, non-parametric directionality index that relates closely to transfer entropy (TE;
544 ⁷⁹), but is based on the phase time-series of the signals under consideration (here, FAF and
545 AC field potentials). PTE is sensitive to information flow present in broad- and narrowband
546 signals, and is in a large degree robust to the effects of, for example, noise, linear mixing, and
547 sample size ^{78, 80}.

548 In terms of TE, a signal X causally influences signal Y (both of them can be considered as
549 phase times series), if the uncertainty about the future of Y can be reduced from knowing both
550 the past of signal X and signal Y, as compared to knowing the past of signal Y alone.

551 Formally, the above can be expressed as follows:

$$552 \quad TE_{xy} = \sum p(Y_{t+\delta}, Y_t, X_t) \log\left(\frac{p(Y_{t+\delta}|Y_t, X_t)}{p(Y_{t+\delta}|Y_t)}\right) \quad [4],$$

553 where δ represents the delay of the information transfer interaction, and TE_{xy} is the transfer
554 entropy between signals X and Y. The estimation of the probabilities for TE quantification
555 requires large computational times and the tuning of various parameters ⁵⁵. PTE, on the other
556 hand, converts the time series into a sequence of symbols (binned-phase time series, see
557 below), and is able to estimate TE on the phase series reducing significantly both processing
558 times and the necessity for parameter fitting ⁷⁸.

559 Phase time series were obtained after filtering the LFP signals in a specific frequency band
560 (e.g. θ , 4-8 Hz) and Hilbert transforming the filtered data. To avoid edge artefacts, the full
561 ~10 minutes recordings were filtered and Hilbert transformed before chunking segments
562 related to individual trials (i.e. pre-voc: -500-0 ms relative to call onset, post-voc: 0-250 ms
563 relative to call onset, or no-voc baseline periods). According to the condition under
564 consideration (echolocation/communication and pre-voc/post-voc, or baseline periods), we
565 selected 50 trials pseudo-randomly and then concatenated them before quantifying directional
566 connectivity. This process was repeated 500 times and the distribution of dPTE values
567 obtained from each repetition used for further analyses. The former resulted in a distribution

568 of 500 dPTE connectivity matrices; the median value across these was used for constructing
569 connectivity graphs (see below).

570 Given the phase of the LFP signals, the PTE was calculated according to equation [4].
571 However, probabilities in this case were estimated by constructing histograms of binned
572 phases⁷⁸ instead of using the full, continuous time series. Following⁸¹, the number of bins in
573 the histograms was set to:

$$574 \quad 3.49 * \mu(\sigma(\phi)) * N_s^{-\frac{1}{3}} \quad [5],$$

575 where m and s represent the mean and standard deviation, respectively, f represents the phase
576 time series, and N_s denotes the number of samples.

577 The prediction delay d was set to $(N_s \times N_{ch}) / N_{+}$ ⁵⁵, where N_s and N_{ch} are the number of
578 samples and channels ($N_{ch} = 32$), respectively. The value of N_{+} corresponds to the number of
579 times the LFP phase changes sign across all channels and times.

580 The dPTE was calculated from the PTE as follows⁵⁵:

$$581 \quad dPTE_{xy} = \frac{PTE_{xy}}{PTE_{xy} + PTE_{yx}} \quad [6]$$

582 With values ranging between 0 and 1, dPTEs > 0.5 indicate information flow preferentially in
583 the $X \rightarrow Y$ direction, dPTE values below 0.5 indicate preferential information flow in the
584 opposite direction, and dPTE = 0.5 indicates no preferred direction of information flow. In
585 other words, dPTE is a metric of preferred directionality between two given signals. Note that
586 the dPTE analysis among a set of electrodes yields a directed connectivity matrix that can be
587 considered as an adjacency matrix of a directed graph (see below). All PTE and dPTE
588 calculations were done with the Brainstorm toolbox in MatLab⁸².

589 Connectivity graphs

590 A graph-theoretic examination of the connectivity patterns was made by constructing directed
591 graphs based on the results obtained from the dPTE analyses (i.e. the median across the 500
592 repetitions; see above). For simplicity, channels in the FAF and AC within a range of 150 μm
593 were grouped as follows (in the FAF, as an example): FAF_{top}, channels 1-4 (0-150 μm);
594 FAF_{mid1}, channels 5-8 (200-350 μm); FAF_{mid2}, channels 9-12 (400-550 μm); FAF_{bottom},
595 channels 13-16 (600-750 μm). A similar grouping was done for electrodes located in AC.

596 These channel groups were considered as the nodes of a directed graph. A directed edge (u, v)
597 between any two nodes then represents a preferential information flow from node u to node v .
598 The weight of the edge was taken as the median dPTE for the channel groups corresponding
599 to the nodes, according to the dPTE connectivity matrices. For instance, if the groups
600 considered were FAF_{top} and AC_{bottom}, then the weight between both nodes was the median of
601 the obtained dPTE values calculated from channels 1-4 in FAF towards channels 13-16 in
602 AC. The weight of an edge was quantified as a directionality index (DI):

$$603 \quad DI = \frac{\text{median}(dPTE_{uv}) - 0.5}{0.5} * 100 \quad [7],$$

604 which expresses, in percentage points, the strength of the preference of information flow in a
605 certain direction. Equation [6] is based on the fact that a dPTE of 0.5 corresponds to no
606 preferred direction of information flow⁵⁵.

607 To statistically validate the directionality shown in the graphs we used a bootstrapping
608 approach. Surrogate adjacency matrices were built for the same channel groups (top, mid1,
609 mid2 and bottom), but electrodes were randomly assigned to each group, independently of
610 their depths or cortical location. This randomization was done independently within each of
611 the 500 dPTE matrices obtained from the main connectivity analysis. Then, an adjacency
612 matrix was obtained from these surrogate data in the same way as described above (i.e. using
613 the median across 500 randomized dPTE matrices). Such a procedure was repeated 10,000
614 times, yielding an equal number of surrogate graphs. An edge in the original graph was kept if
615 the DI of that edge was at least 2.5 standard deviations higher than the mean of the surrogate
616 distribution obtained for that edge (i.e. higher than the 99.38% of the surrogate observations).
617 Edges that did not fulfil this criterion were labelled as non-significant and were therefore not
618 considered for any subsequent analyses.

619 Directionality analyses for passive listening conditions

620 dPTE values and connectivity graphs for passive listening conditions were quantified using
621 the same methodology described for the cases of active vocalization. Analyses based on
622 responses to acoustic stimulation were made on a trial-by-trial basis. Trials were randomly
623 selected 500 times across all penetrations, depending on whether responses to the HF-FM or
624 the distress sound were considered. We ensured that the number of trials chosen for each
625 penetration, in every randomization run, matched the number of vocalizations taken (in a HF-
626 FM/echolocation or distress/communication scheme) from that particular penetration. With

627 this we aimed to avoid possible biases in the comparisons across passive listening and active
628 vocalizations conditions.

629 Electrical stimulation experiments

630 The FAF was electrically stimulated by means of biphasic pulses lasting 410 μs (200 μs per
631 phase, with 10 μs gap between them) and with an amplitude of 2 μA . Electric pulses were
632 delivered by inserting an A16 Neuronexus shank (same used for recordings) into the frontal
633 cortex, using the channels at depths of 350 and 450 μm as stimulating electrodes. These
634 channels were directly connected to the outputs of an A365 stimulus isolator (World Precision
635 Instruments, Friedberg, Germany). Pulse amplitude was selected based on values used in the
636 literature (e.g. ^{83, 84, 85}), and after empirically establishing that electrical artefacts were
637 undetectable in the AC online during recordings. Recordings in AC were conducted with a
638 second A16 shank (as described above); in the AC no electrical stimulation was delivered.

639 Precisely, the electrical and acoustic stimulation protocol was as follows. Six biphasic electric
640 pulses were delivered into FAF with inter-pulse intervals of 500 ms (2 Hz, within the δ -band
641 range). After the electrical pulse train, acoustic stimuli were presented to the bats at given
642 delays relative to the time last electrical pulse (10, 135, 260, 385, 510, 635, 760 and 885 ms;
643 “Estim” condition). Delays were consistently with sampling four different phases per
644 electrical stimulation cycle (period is considered as the inter-pulse interval), for a total of 2
645 cycles after the last electrical pulse delivered (see **Fig. 4a**). Acoustic stimuli consisted of a
646 distress (same syllable used for the passive listening experiments) and an echolocation call
647 (duration, 1.2 ms; spectrum shown in **Fig. 4b**), both presented at 50 dB SPL (rms). All
648 possible combinations of acoustic stimulus type (i.e. distress or echolocation) and sound onset
649 delay were pseudorandomly presented 25 times each. The interval between a trial block (i.e.
650 electric pulse train and acoustic stimulus) was of 2.5 seconds. Sounds were also presented
651 without any prior electrical stimulation (“no-Estim” condition), with a variable inter-stimulus
652 interval between 4 and 16 seconds, a total of 50 times each. Acoustic stimuli were delivered
653 using the same speaker setup described above, but for these experiments a different sound
654 card was used (RME Fireface UC; 16 bit precision, 192 kHz; RME Audio, Haimhausen,
655 Germany).

656 Comparisons of auditory cortical response strength between Estim and no-Estim conditions
657 were made by calculating the energy of the event-related potential (ERP). Specifically, the

658 response strength was calculated as the area under the curve of the absolute value of the
659 Hilbert-transform of the high-frequency component (25 – 80 Hz) of the ERP, for the first 150
660 ms after sound onset. We used the high frequency component of the ERP to avoid biases
661 related with low-frequency pre-stimulus trends in the LFPs. Because the number of trials
662 differed in the Estim and no-Estim conditions (i.e. 25 and 50, respectively), the response
663 strength of the no-Estim condition was calculated using 25 randomly selected trial out of 50.
664 The ERP is sensitive to the number of trials as it is a trial-average response; the number of
665 trials must therefore be equalized. Only then we compared response strengths between Estim
666 and no-Estim conditions, for each channel and delay relative to the last electrical pulse
667 delivered. Statistical comparisons were made with FDR-corrected Wilcoxon signed-rank tests,
668 with an alpha of 0.05. The effect sizes of these comparisons were calculated using the Cliff's
669 delta metric (a non-parametric approach). Effect sizes are considered negligible when Cliff's
670 delta < 0.147, small when $0.33 \leq \text{Cliff's delta} < 0.33$, medium when $0.33 \leq \text{Cliff's delta} <$
671 0.474 , and large when Cliff's delta ≥ 0.474 ⁵⁸. On account of selecting 25 random trials from
672 the no-Estim condition, the above procedures were repeated 500 times, with the aims of
673 testing whether the outcomes of the statistical comparisons (Estim vs. no-Estim response
674 strength) were reliable and independent of the randomized trial selection.

675 Statistical procedures

676 All statistical analyses were made with custom-written MatLab scripts. Paired and unpaired
677 statistical comparisons were performed with Wilcoxon signed-rank and rank sum tests,
678 respectively. These are appropriately indicated in the text, together with sample sizes and p-
679 values. All statistics, unless otherwise noted, were corrected for multiple comparisons with
680 the False Discovery Rate approach, using the Benjamini and Hochberg procedure ⁸⁶. An alpha
681 of 0.05 was set as threshold for statistical significance. The effect size metric used, unless
682 stated otherwise (as in the GLM case), was Cohen's d:

$$683 \quad d = \frac{\mu_{D1} - \mu_{D2}}{\sqrt{\frac{(n_1 - 1)\sigma_{D1}^2 + (n_2 - 1)\sigma_{D2}^2}{n_1 + n_2 - 2}}} \quad [8],$$

684 where D1 and D2 are two distributions, μ represents the mean, σ^2 represents the variance,
685 while n_1 and n_2 are the sample sizes. Effect sizes were considered small when $|d| < 0.5$,
686 medium when $0.5 \leq |d| \leq 0.8$, and large when $|d| > 0.8$ ⁸⁷.

687 To test differences in the connectivity graphs across conditions (e.g. echolocation vs.
688 communication, or passive listening vs. active vocalization), we obtained adjacency matrices
689 for each of the 500 penetrations (one per dPTE connectivity matrix; see above) and compared
690 the distributions using Wilcoxon signed rank tests. Given that the large sample size ($n = 500$
691 here) increases the occurrence of significant outcomes in statistical testing, edges were only
692 shown when comparisons were significant and produced large effect sizes ($|d| > 0.8$).

693 When comparing connectivity graphs between pre-voc and post-voc conditions, we used the
694 exact same trials per repetitions to construct the distribution of dPTE matrices for the pre- and
695 post-voc cases. A certain repetition m for each condition was then treated as paired, and
696 therefore Wilcoxon signed rank tests were used for comparing (as opposed to unpaired
697 statistics above). Again, only edges representing significant differences ($p_{\text{corr}} < 0.05$) with
698 large effect sizes were shown.

699 **Conflict of interests**

700 The authors declare no financial or non-financial conflicts of interest.

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993 **Figure legends**

994 **Fig. 1. Pre-vocal oscillations in frontal and auditory cortices predict ensuing vocal output. (a)**
995 Oscillograms (top) and spectrogram (bottom) of exemplary echolocation (left) and communication
996 calls produced by *C. perspicillata*. **(b)** Cumulative probability distribution of echolocation (blue, n
997 = 138) and communication (orange, n = 734) call lengths. No significance differences were
998 observed (Wilcoxon rank sum test, p = 0.12). **(c)** *(Left)* Normalized average power spectral density
999 (PSD) of echolocation (blue) and communication (orange) calls. *(Right)* Distribution of peak
1000 frequencies of echolocation and communication utterances; communication calls were significantly
1001 higher in frequency than their counterparts (p = 2.24×10^{-66}). **(d)** Single-trial LFPs recorded
1002 simultaneous to echolocation and communication utterances. The vertical red dashed line, at time
1003 0, indicates the moment of vocalization onset. The top 16 traces correspond to LFPs recorded in the
1004 FAF; the bottom 16 LFP traces were recorded from the AC. Auditory cortical layers are marked.
1005 LFP amplitude was normalized within structure independently. **(e)** Average pre-vocal (-500 to 0
1006 ms, relative to call onset) power spectral densities (PSD) at a representative depth (300 μ m) in FAF
1007 and AC. Blue: echolocation; orange: communication; black, dashed: no-voc periods. The difference
1008 between echolocation and communication PSDs is depicted in grey (right). **(f)** Percentage pre-vocal
1009 power change across representative LFP bands (δ , 1-4 Hz; β_1 , 12-20 Hz; γ_2 , 60-120 Hz), relative to
1010 a no-voc baseline, across cortical depths in FAF (top) and AC (bottom). Values related to
1011 echolocation utterances (n = 138) are depicted in blue; those related to communication utterances
1012 (n = 734) are depicted in orange. Data shown as mean \pm sem. **(g)** Pre-vocal power change in frontal
1013 and auditory regions predict vocalization type. Effect size (R^2_m) of GLMs considering all frequency
1014 bands and channels, both in frontal and auditory cortices. Effect sizes were considered small when
1015 $R^2_m < 0.1$, and medium for $R^2_m \geq 0.1$. For illustrative purposes, effect size values from non-
1016 significant models were set to 0.

1017 **Fig. 2. Directed connectivity patterns in the FAF-AC network. (a)** Graph visualization of directed
1018 connectivity between FAF and AC during no-voc periods. Channels in frontal and auditory cortices
1019 were combined into four categories: top (0-150 μ m), mid1 (200-350 μ m), mid2 (400-550 μ m), and
1020 bottom (600-750 μ m). Graph edges are weighted according to the strength of the preferred
1021 directionality (FAF \rightarrow AC in blue; AC \rightarrow FAF in orange; within area directionality in grey). Edges
1022 are only shown if there was significant preferred directionality according to a threshold defined by
1023 bootstrapping. **(b)** Similar to **a**, but directed connectivity was calculated in the pre-vocal
1024 echolocation and communication conditions. **(c)** Same as **b**, with connectivity patterns obtained for
1025 post-vocal echolocation and communication conditions.

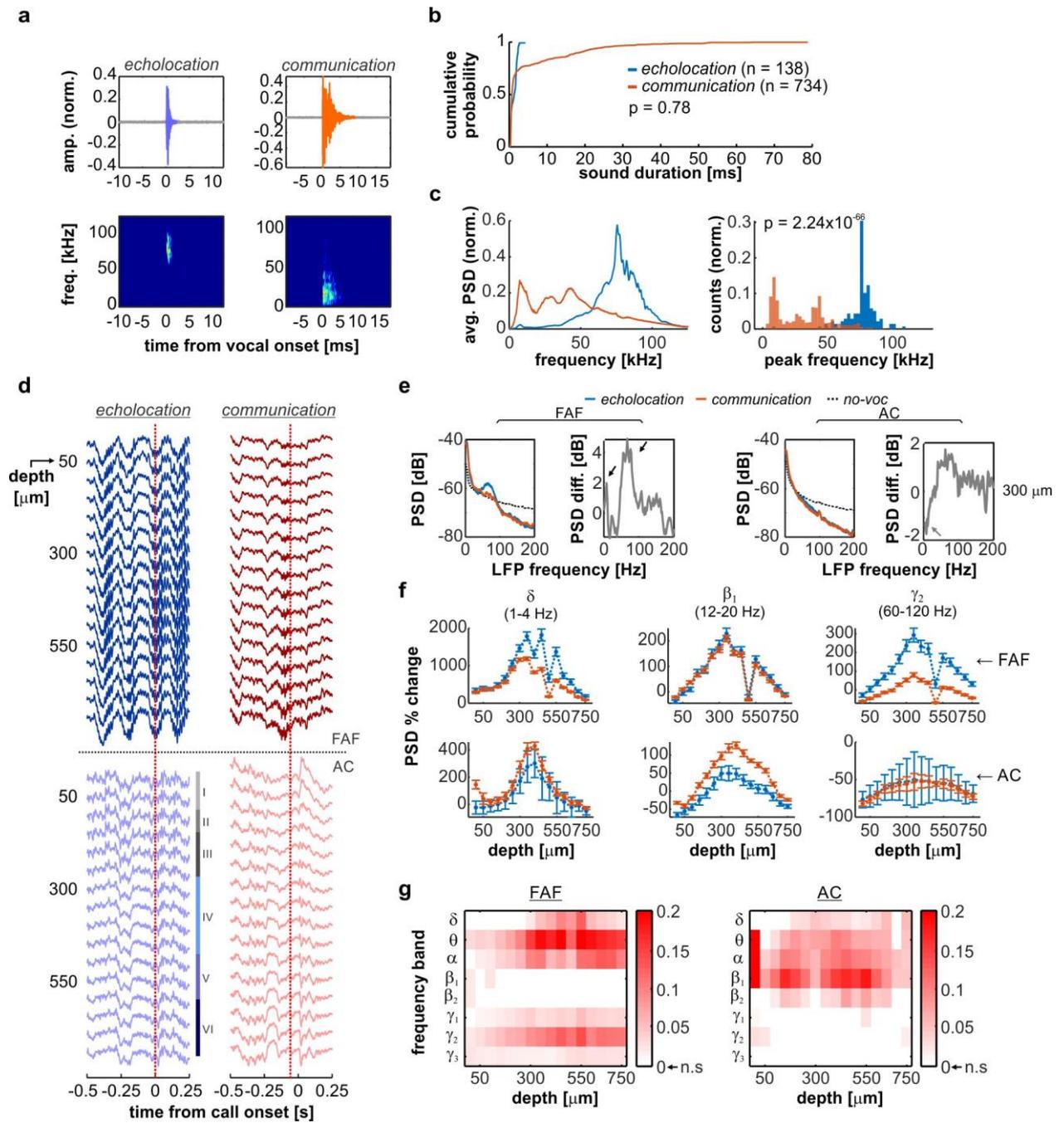
1026 **Fig. 3. Pre-vocal and post-vocal directionality differences in the FAF-AC network. (a)** *(Top)*
1027 Graphs illustrating the differences between pre-vocal and post-vocal directionality, across
1028 frequency bands and during the production of echolocation calls. Edges were shown if three
1029 conditions were met: (i) the differences were significant (FDR-corrected Wilcoxon signed-rank
1030 tests, $p_{\text{corr}} < 0.05$), (ii) the effect size was large ($|d| > 0.8$), and (iii) edges were already significantly
1031 directional (see edges in **Fig. 2**). Edge thickness is weighted according to the effect size of the
1032 comparison. Continuous lines indicate pre-vocal dPTEs (first condition) higher than post-vocal
1033 dPTEs (second condition). Dashed lines indicate the opposite. *(Bottom)* Net information outflow
1034 (DI_{net}) from FAF (blue bars) and AC (orange bars), in the two conditions considered (pre-vocal vs.
1035 post-vocal). Significant differences across conditions are marked with stars (FDR-corrected
1036 Wilcoxon signed-rank tests; * $p_{\text{corr}} < 0.05$, ** $p_{\text{corr}} < 0.01$, *** $p_{\text{corr}} < 0.001$, n.s.: not significant; n =
1037 500). Grey numbers in the panels indicate effect sizes (d ; not shown for non-significant

1038 differences). Values were considered independently of whether there was previous significant
1039 directionality in any of the two conditions. Data shown as mean \pm sem. **(b)** Same as in **a**, but
1040 illustrating comparisons of directionality between pre-vocal vs. post-vocal conditions related to the
1041 vocalization of communication calls.

1042 **Fig. 4. Electrical stimulation of the FAF increases response strength in AC.** **(a)** Schematic
1043 representation of the paradigm for electrical and acoustic stimulation. The timestamps for acoustic
1044 stimulation (coloured according to delay; see also panels **e-g**) represent the delay of sound onset
1045 relative to the end of electrical stimulation train. **(b)** Oscillograms of the natural distress syllable
1046 and echolocation pulse used for acoustic stimulation. On the right, the normalized power spectra of
1047 both calls are shown (orange, distress; blue, echolocation). **(c)** Broadband (0 – 10kHz), raw data
1048 recorded simultaneously from FAF and AC (at representative depths of 50, 250 and 600 μ m)
1049 illustrating a single trial of electrical stimulation. Note that no electrical artefacts are visible in AC.
1050 **(d)** Auditory cortical LFPs (left column), and time-course of their energy (right), in response to
1051 either the distress syllable (top) or the echolocation pulse (bottom). Responses corresponding to the
1052 no-Estim conditions shown in blue; responses related to the Estim condition, in red. **(e)** Strength of
1053 auditory cortical ERPs in response to the distress syllable, across all recorded columns ($n = 20$) and
1054 depths. In blue, responses associated to the no-Estim condition; in red, those associated to the Estim
1055 condition (data as mean \pm s.e.m). **(f)** Corrected p-values obtained after statistical comparisons
1056 between response strengths related to Estim and no-Estim conditions, across all channels and
1057 delays (paired, FDR-corrected Wilcoxon signed rank tests, $\alpha = 0.05$). **(g-h)** Same as in **e-f**, but
1058 dealing with responses to the echolocation pulse. **(i)** Proportion iterations (out of 500) in which
1059 responses associated to the Estim condition were significantly larger than those associated to the
1060 no-Estim condition (same test as above). Data are presented across all channels and delays
1061 analysed, for responses to the distress and echolocation sounds. **(j)** Median effect size (Cliff's delta)
1062 for the same comparisons summarized in **i**.

1063 **Fig. 5. The FAF-AC network during vocal production.** **(a)** Oscillations in frontal and auditory
1064 cortices provide a neural correlate of vocal production, allowing the prediction of ensuing call type.
1065 Prediction is possible in complementary frequency bands in each region, and with opposite effects.
1066 **(b)** Schematic representation of causal interactions (within a TE framework) in the FAF-AC
1067 network. Strong top-down control, mostly in δ and γ frequencies, occurs during spontaneous
1068 activity (no-voc) and prior to vocal utterance. In the δ -band, information flows top-down in the
1069 circuit (FAF \rightarrow AC) during pre-vocal periods, but changes to bottom-up (AC \rightarrow FAF) information
1070 transfer during post-vocal periods. The directionality patterns and the strength of preferential causal
1071 interactions depend on the type of call produced, and on the timing relative to vocal onset. **(c)**
1072 Electrical stimulation results provide strong support to the notion that FAF alters the manner in
1073 which AC processes acoustic information, preferentially when animals listen to echolocation
1074 sounds. In comparison to listening of distress sounds, auditory cortical responses to echolocation
1075 sounds were more reliably enhanced after electrically stimulating the frontal cortex.

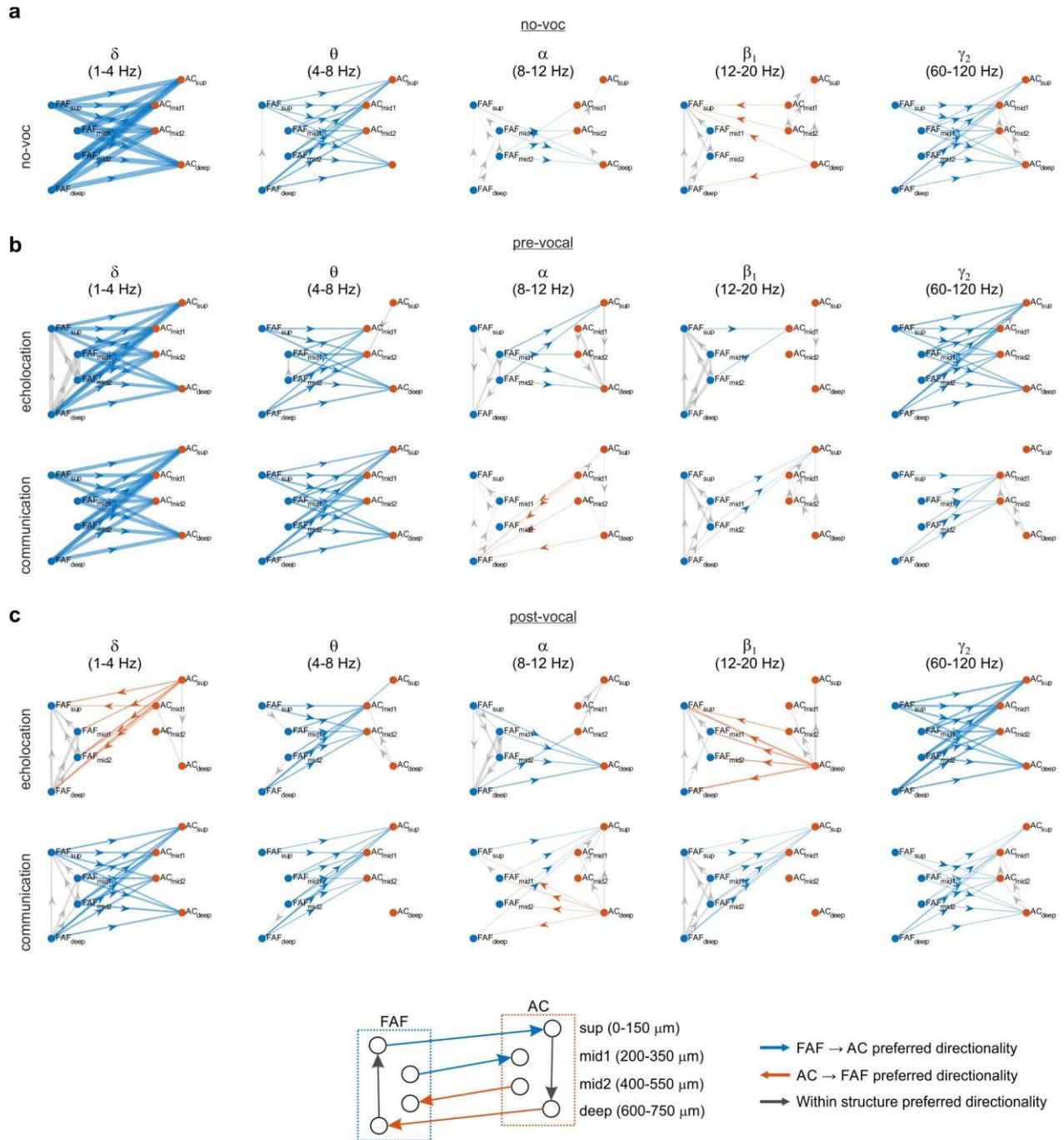
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1078 **Fig. 1. Pre-vocal oscillations in frontal and auditory cortices predict ensuing vocal output.**

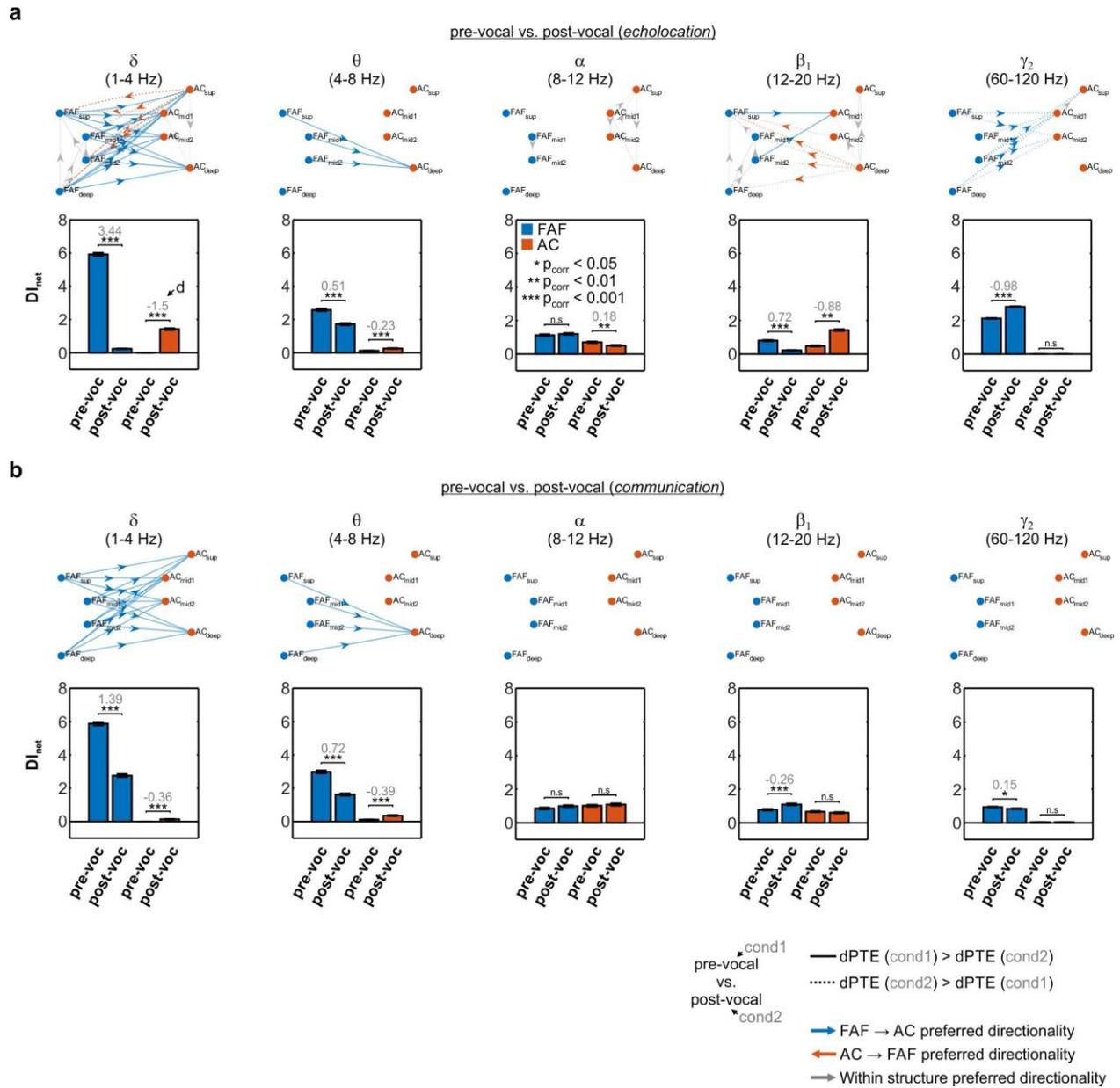
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1081 **Fig. 2. Directed connectivity patterns in the FAF-AC network.**

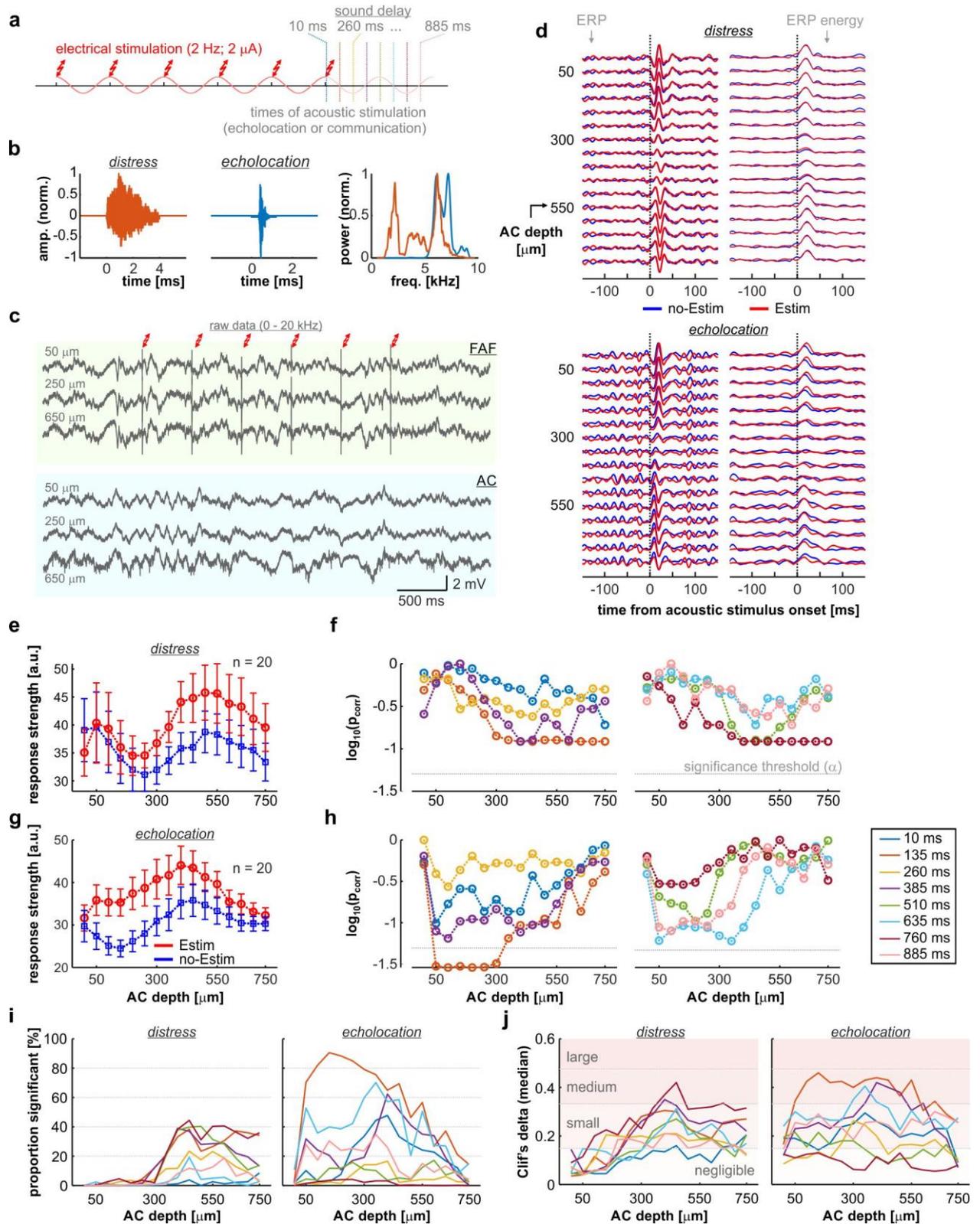
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1084 **Fig. 3. Pre-vocal and post-vocal directionality differences in the FAF-AC network.**

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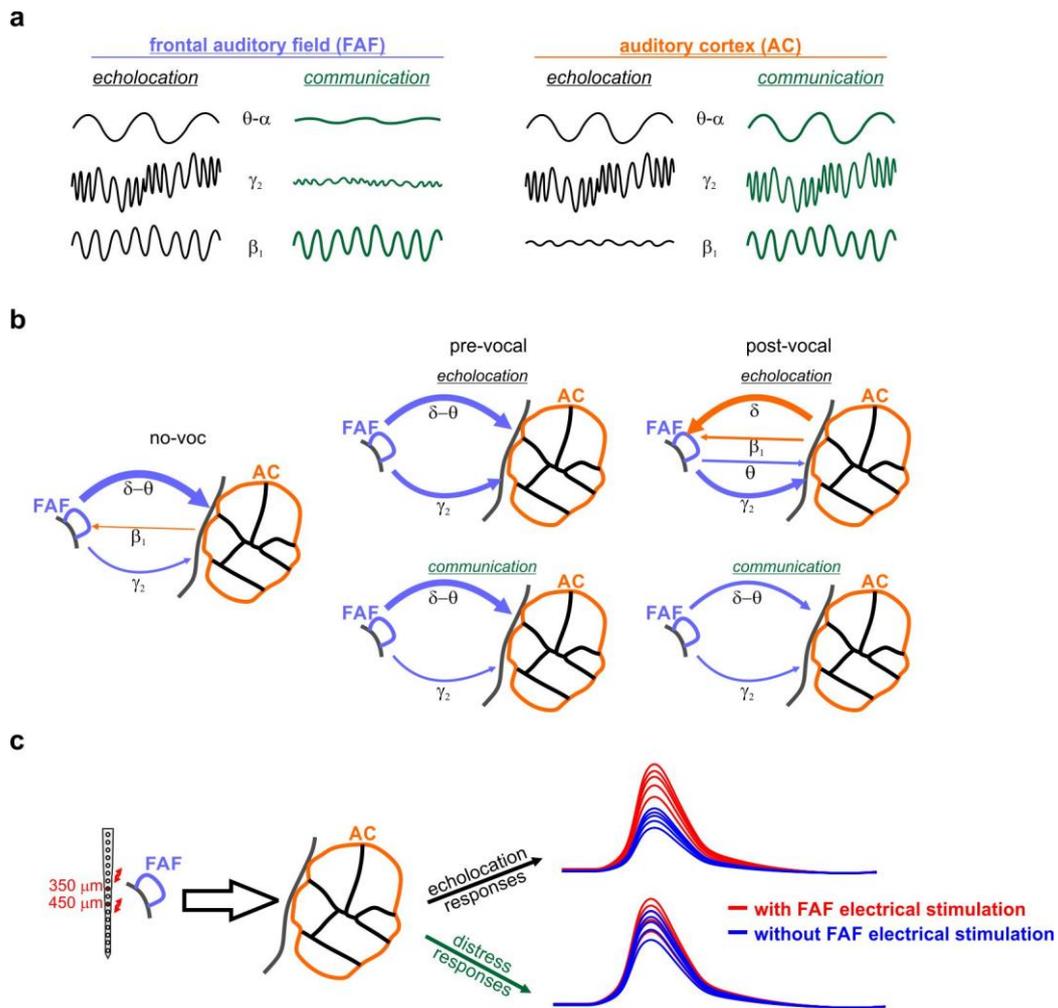


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Fig. 4. Electrical stimulation of the FAF increases response strength in AC.

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1090 **Fig. 5. The FAF-AC network during vocal production.**

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