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4 Diotic and dichotic frequency discrimination thresholds in musicians and non-musicians: relationships
5 between perception, musical ability and self-evaluated competence.

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18 **Abstract**

19 Pitch perception provides important information for musical and vocal communication. Numerous studies
20 have shown that musical training and expertise are associated with better pitch processing, however, it is
21 unclear what types of pitch percepts are plastic with music training. The current study addresses this issue
22 by measuring discrimination thresholds of Musicians (n=20) and Non-musicians (n=18) to diotic (same
23 sound to both ears) and dichotic (different sounds to each ear) stimuli created from four types of acoustic
24 computations: 1) pure sinusoidal tones, PT; 2) four-harmonic complex tones, CT; 3) iterated rippled
25 noise, IRN; and 4) interaurally correlated broadband noise, called “Huggins” or “dichotic” pitch sounds,
26 DP. Frequency Difference Limens (DLF) in each condition were obtained via a 3-alternative-forced-
27 choice adaptive task requiring selection of the interval with the highest pitch, yielding the smallest
28 perceptible fundamental frequency (F0) distance (in Hz) between two sounds. Music skill was measured
29 by an online test of musical Pitch, Melody and Timing (International Laboratory for Brain Music and
30 Sound Research, <https://www.brams.org/en/onlinetest/>). Musicianship, length of music experience and
31 self-evaluation of musical skill were assessed by questionnaire. Results showed musicians had smaller
32 DLFs in all four conditions and that thresholds were related to subjective and objective musical ability. In
33 addition, self-report of musical ability was shown to be a significant variable in group classification,
34 suggesting that the neurobehavioral profile of musicians includes self-evaluation of musical competence.

35

36 **Introduction**

37 Musical training is associated with better pitch encoding and perception (for review see (1)). In music,
38 pitch is the quality that most strongly defines the perception of the melodic contour. Each note in a
39 melody has a pitch that is related to the lowest frequency of an instrumental sound, called the fundamental
40 frequency (F0). The perception of pitch, however, is not only evoked by musical sound, but can emerge
41 from a wide variety of acoustic components (for review see (2)). A classic way to determine someone's
42 pitch perception ability is by measuring the smallest perceptible pitch change from a center frequency,
43 called a *difference limen for frequency* (DLF) (3, 4). In general, normal-hearing listeners can perceive a
44 change in as little as 2-3 Hz from a center frequency under optimal listening conditions (5). Musicians can
45 detect even smaller pitch changes, sometimes so minute that the change is undetectable by otherwise
46 normal-hearing non-musicians (2, 6, 7). Not surprisingly, increased acuity in musicians is not limited to
47 musical sounds, but extends to perception and processing of speech (8), non-speech ((6) for review (9))
48 and non-native language sounds (10).

49 A prevalent hypothesis is that musical training improves auditory encoding mechanisms that give rise to
50 pitch perception. However, the auditory system utilizes several mechanisms to encode pitch-related
51 acoustics and it is unclear which ones are most improved with music training. One way the auditory
52 system works is by representing the “temporal code” of a stimulus in which auditory neurons *phase-lock*,
53 firing at a rate that matches the period, or frequency inverse, of a sound (for review see (11)). During
54 temporal encoding, sounds trigger networks of neurons to compute and extract temporal patterns that can
55 give rise to pitch perception (12, 13). Music practice and performance could activate and strengthen the
56 temporal synchrony of these networks, thereby improving representation and higher-order computation
57 acuity. Another mechanism to encode pitch, called “place code”, functions such that different frequencies
58 activate discrete regions of the inner ear and subsequent nuclei, producing a tonotopic map of frequencies
59 at each processing station (for review, see (14)). For example, the perception of pitch rises as the region
60 of maximal activation on the basilar membrane moves closer to the base of the cochlea. Music training

61 could generate more precise and definite tonotopic maps due to top-down modulation induced by the
62 increased prevalence and relevance of sounds in the musician's environment (15-17). Finally, a pitch
63 perception can be generated by presenting different sound components to each ear, creating a *dichotic*
64 (binaural) or combined estimation of the sound's pitch (18, 19). Although the music-related mechanistic
65 hypotheses are less prevalent for dichotic plasticity, it is reasonable to suggest that music training could
66 increase the accuracy of communication between the left and right ears, particularly during azimuth
67 (horizontal) localization tasks such as identification of instruments in an orchestra (20). To encode sound,
68 the auditory system will use or integrate information gathered from each encoding strategy presented
69 above, depending on what acoustic features are present in the stimulus.

70 The working hypothesis that motivated this study was that music training engenders plasticity in specific
71 auditory encoding mechanisms. Particularly, we posited that sounds reliant on temporal encoding would
72 be impacted the most because playing music requires considerable focus on sound timing. To test this, we
73 measured DLFs in musicians and non-musicians using four different types of sounds with different pitch-
74 related acoustics (Fig. 1). Creation of the sounds was inspired by work describing how different degrees
75 of temporal, place and dichotic encoding mechanisms can be elicited in the auditory system (21). In order
76 to assess musical skill, all participants took an online musical test for pitch, melody and timing
77 measurements and filled out a questionnaire that probed duration of musical training and subjective self-
78 reports of musical skill and listening habits.

79 **Materials and Methods**

80 **PARTICIPANTS:** 38 individuals with audiometric thresholds within normal limits (<25dB HL for 0.25,
81 0.5, 1, 2, 3, 4, 6 and 8 kHz) and no history of neurological disorders participated in the study. Previous
82 research has shown that music-related brain plasticity is most effective when people begin playing music
83 early, continue, and are currently practicing (22-24). Therefore, subject inclusion criteria in the Musician
84 (MU) group included 1) self-identification as a musician via questionnaire and reported current
85 involvement in musical activities, 2) self-report of music training initiation before high school (e.g. before

86 grade 9, age 14-15) and 3) a total of at least 5 years in formal music education. 20 subjects fulfilled the
 87 criteria for MU group inclusion, with the remainder 18 subjects grouped into Non-musicians (NM).
 88 Group characteristics of age, music education, self-ratings and objective measures of musical skill (i.e.
 89 online aptitude test, for description see below) are presented in Table 1.

Group	Metric	Self-reported Music Education (yrs.)			Self-reported Musical Skill and Listening (Scale 1-9)		Objective Musical Skill Scores (MBEA) (%)			
		Age at Test	Age Began Music	Total Music Education	SR Musical Skill	SR Music Listening Frequency	Melody	Timing	Pitch	Total/Avg.
Non-musicians (n=18)	Mean	25.11	7.63	5.13	2.22	7.00	84.28	86.89	88.06	86.22
	Std. Dev.	1.906	2.875	2.416	1.629	1.680	7.466	6.296	11.254	5.342
Musicians (n=20)	Mean	23.75	7.60	12.20	7.05	7.20	90.90	93.10	95.75	92.9
	Std. Dev.	5.543	3.202	4.099	.759	1.609	6.782	4.909	4.541	3.782

Self-reported music education, music skill and listening frequency measures obtained via questionnaire and are reported in years. Only 8 Non-musicians had previous music education. Self-reported music skill was rated on a scale from 1-9, with 1 being “novice” and 9 denoting “professional”. Music listening frequency was rated on a scale from 1-9 with 1 being “never” and 9 “all the time”. Melody, Timing, Pitch and Average/Total musical skill scores obtained via online aptitude test (www.brams.org) and are reported in percent correct.

90

91 STIMULI. Sounds were 300 ms in duration, with two 60-ms raised cosine ramps for onset and offset.
 92 Figure 1 shows time waveforms (left panels) and frequency spectra (right panels) for the 440 Hz
 93 (standard) stimuli used in the study. 440 Hz was chosen because it is a familiar musical note (A4) that
 94 elicits strong phase-locking.

95 In order to test binaural mechanisms, we created a dichotic pitch (DP) stimulus, often called “Huggins’
 96 pitch,” which consists of dissimilar right and left inputs to make a dichotic estimation of a sound’s pitch
 97 (18, 19). DP stimuli were created with the Binaural Auditory Processing Toolbox for MATLAB8 using a
 98 transition width of 16%. DP sounds were made of white noise, diotic at all frequencies except for a
 99 narrow band at the F0 (440 Hz), over which the interaural phase transitioned progressively through 360°.
 100 Individuals were familiarized with DP perception through five online Demonstrations
 101 (<https://web.stanford.edu/~bobd/cgi-bin/research/dpDemos/>).

102

-Figure 1-

103 Figure 1. Study stimuli. Each row shows the 440 Hz stimulus waveform (right panel) and spectrum on a
104 logarithmic frequency scale (left panel). A. Dichotic pitch with an interaural phase shift of 440 Hz, B.
105 Pure Tone, C. Iterated Rippled Noise with a 64 iteration of delay and add at 1/440 s. D. Complex tone
106 with three overtone harmonics

107 In contrast, a pure tone (PT), shown in panel B of Figure 1 is the product of a sinusoidal function.
108 Sinusoids are thought to be encoded by place code mechanisms because they elicit narrow bands of
109 maximal activation at specific places in the tonotopic map of the cochlea. At lower frequencies ($< \sim 2$ kHz)
110 elicit additional phase-locked temporal codes at the frequency's period. Pure tones consisted of sinusoids
111 at a fundamental frequency (F0) of 440 Hz, chosen.

112 We also tested an iterated noise (IRN) stimulus which evokes a pitch perception that is primarily reliant
113 on temporal information (25-27). IRN stimuli, shown in Figure 1C, were created from Gaussian
114 broadband noise filtered from 80-3000 Hz with 64 iterations of delay and add durations at the inverse of
115 the F0 (440 Hz). The temporal regularity imposed on broadband noise gives rise to the perception of pitch
116 despite low spectral content.

117 Finally, we used a complex tone with three harmonic overtones (CT), which most closely resembles the
118 sound a musical instrument makes and relies on a combination of place and temporal codes. Complex
119 tones (Fig. 1D) consisted of a four-harmonic complex (h1-h4) with equal amplitude and the same F0 (440
120 Hz).

121 MUSICAL APTITUDE TEST: Individuals completed an online test through the International Laboratory
122 for Brain, Music, and Sound Research (BRAMS, www.brams.org/en/onlinetest/) that allows for the
123 assessment of the functioning of each musical component: 1) Melody, 2) Timing, and 3) Pitch ability. The
124 online test battery is based on the Montreal Battery for Evaluation of Amusia (MBEA) and consists of
125 musical phrases that vary along the melody, timing or pitch dimension (28). During the MBEA, listeners

126 perform a two-alternative forced choice task to determine whether two presented musical phrases are the
127 same or different. In the melody test, for example, the two choices may consist of an original melodic
128 contour and a scale- or contour-violated alternate. The output of the online test is a percentage correct for
129 each task, in addition to the average of all three categories. These percentages were recorded and utilized
130 in this study. Group means and standard deviations are shown in Table 1.

131 QUESTIONNAIRE: Participants' musical history was collected through a questionnaire probing a range
132 of information regarding subjective aptitude and measures of musicianship. We used the following
133 details and scale-based ratings to correlate with objective performance on psychoacoustic measures: 1)
134 Musician self-identification (e.g. "Are you a musicians?"), 2) Self-Report of Music Listening Frequency
135 on a scale of 1-9 3) Self-Report of Musical Skill on a scale of 1-9 4) Age of Music Start and 5) Years of
136 Consistent Practice. Group means and standard deviations are shown in Table 1.

137 DATA ANALYSIS: Tests of normality were computed on all variables. Results of these tests showed that
138 the pairs of MU and NM distributions were not significantly different from normal according to or
139 Shapiro-Wilk tests ($p < 0.01$), except for SR Musical Skill and BRAMS Pitch score (See Supplementary
140 Table 1). Examination of the detrended SR Musical Skill scores showed that one NM rated themselves
141 >1 Standard Deviation from normal and one MU rated themselves >-1 Standard Deviation from normal.
142 Examination of the detrended BRAMS pitch scores showed that one individual from each group scored $>-$
143 1 Standard Deviation from normal. Given that a skew in distribution was observed for two measures, we
144 provide observed power for each test and only conducted tests that were robust to the assumption of
145 normality (29, 30).

146 The group difference hypothesis was tested using a set of mixed repeated-measures ANOVAs
147 (RMANOVA) for group (Musicians vs. Non-musicians, 2 between-subjects factors) and the following
148 within-subjects factors: 1) sound type threshold (CT, PT, IRN, DP, 4 within-subjects factors), 2) sound
149 type standard deviation (CT, PT, IRN, DP, 4 within-subjects factors), 3) self-reported measures of music
150 skill and 4) listening frequency from the questionnaire (SR Musical Skill and SR Frequency of Music

151 Listening, 2 within-subject factors). F-statistics, p-values and the observed statistical power, ranging from
152 0 to 1, where the fraction represents the chance of failing to detect an effect, are reported with each
153 significant ANOVA result. Post-hoc t-tests were conducted when significant interaction effects with p-
154 val<0.05 were observed.

155 To examine the question of a relationship between DLFs, musicality and self-assessment, Pearson's r
156 correlations were computed. Pearson's r-values and p-values of the significance test are reported. In order
157 to discover the degree to which our dependent variables discriminate between NM and MU, a
158 discriminant function analysis with predictive classification of cases was conducted. The discriminant
159 analysis included all four DLF thresholds, BRAMS total score and self-reported measures for a total of
160 seven continuous, numeric variables and one categorical variable with two levels (NM, MU).

161 **Results**

162 *Pitch discrimination thresholds*

163 The 2X4 mixed RMANOVA on group (MU vs. NM, between-subjects factor) and threshold (CT, PT,
164 IRN and DP, within-subjects factors) showed a within-subjects main effect of sound type;
165 $F(3,108)=38.137$, $p<0.001$, Observed Power=1.0, an interaction effect $F(3,108)=11.754$, $p<0.001$,
166 Observed Power=0.999, and a between-subjects main effect $F(1,36)=29.205$, $p<0.001$, Observed
167 Power=1.0. Post-hoc t-tests were significant for all conditions ($p<0.016$).

168 Group means show that musicians had lower thresholds for each sound type category (Fig. 2, Supp. Table
169 1). Bar graphs in Figure 2 illustrates the group mean values for each sound type category, showing lower
170 means for the musician group in all categories, relative to Non-musicians. Taken together, the data show
171 that musicians can hear smaller pitch differences than Non-musicians in all four pitch-evoking sound type
172 categories, with the greatest difference in the DP condition and the smallest difference in the CT
173 condition.

174 -Figure 2-

175 Figure 2. Bar graph shows mean DLF thresholds (+/- 1 SE). Musicians have smaller (better) pitch
176 discrimination thresholds in all conditions, relative to non-musicians (* $p < 0.05$; ** $p < 0.01$).

177

178 To examine group differences in threshold variance a 2X4 mixed RMANOVA on group (MU vs. NM,
179 between-subjects factor) and threshold variance per sound type (DP, PT, IRN and CT, within-subject
180 factors). For this analysis, standard deviation was computed from the four recorded threshold
181 measurements per sound type. Results showed a within-subjects main effect of sound type;
182 $F(3,108)=18.265$, $p < 0.001$, Observed Power=1.0, an interaction effect $F(3,108)=3.866$, $p=0.011$,
183 Observed Power=0.811, and a between-subjects main effect $F(1,36)=11.702$, $p=0.002$, Observed
184 Power=0.914. Post-hoc t-tests were significant for DP and PT ($p < 0.007$), but not IRN and CT ($p > 0.058$).
185 Examination of group means showed that threshold variance is smaller in MU than NM in the DP and PT
186 condition (Supp. Table 2).

187 *Relationships between pitch discrimination thresholds, self-reports and musical aptitude measures*

188 Pearson's correlations show that better discrimination thresholds are associated with a higher self-report
189 of musical skill and better scores on all tests of BRAMs musical aptitude. Correlations are reported in
190 Supplementary Table 2. Figure 3 shows individual data for the representative correlations between DLFs,
191 self-report and BRAMS total score. Figure 3 (left column) illustrates that lower (better) DLFs are
192 associated with higher self-reports of musical skill. The spread of the data in Figure 3 also illustrates
193 greater variance in self-report among NM compared to MU, reflecting a wider range of self-assessed
194 musical experience in the NM group. Figure 2 (right column) shows the relationships between DLFs and
195 BRAMS total score. Consistent negative correlations suggest that smaller DLFs are associated with
196 higher musical aptitude.

197

-Figure 3-

198 Figure 3. Scatterplots of individual data for Musicians (red) and Non-musicians (black) with regression
 199 lines. Left column shows relationships between pitch discrimination thresholds self-reported (subjective)
 200 musical skill (scaled between 1-9, with 1 being novice, 9 professional). Higher self-report is associated
 201 with smaller (better) thresholds. Right column shows relationships between pitch discrimination
 202 thresholds behavioral scores obtained from the BRAMS musical skills test (objective). Higher score is
 203 associated with smaller (better) thresholds.

204 *Discriminant analysis*

205 A discriminant analysis was conducted to determine which of our variables contributed most to group
 206 separation and to test whether an individual's group category could be correctly identified based on our
 207 continuous numeric experimental measures. Continuous variables were DP, PT, IRN and CT DLFs as
 208 well as SR Musical Skill and BRAMS Avg./Total score. Table 2 shows significant mean differences were
 209 observed for all variables ($p < 0.021$) except for SR Music Listening Frequency ($p = 0.774$).

Table 2. Discriminant analysis results including tests of equality of group means and variable loadings.						
Metric	Wilks' Lambda	F	df1	df2	Sig.	Structure Matrix (Loadings)
Dich. Pitch DLF	18.502	18.502	1	36	<0.001	-.295
Pure Tone DLF	25.668	25.668	1	36	<0.001	-.348*
Itr. Rip. Noise DLF	6.333	6.333	1	36	.016	-.173
Comp. Tone DLF	16.646	16.646	1	36	<0.001	-.280
SR Musical Skill	141.793	141.793	1	36	<0.001	.817*
BRAMS Avg./Total	19.747	19.747	1	36	<0.001	.305*
Structure Matrix (Loadings) shows pooled within-groups correlations between discriminating variables and standardized canonical discriminant functions, *denotes important correlations >0.3						

211 The canonical discriminant function showed a significant association between groups and variables;
212 Wilks' Lambda=0.145, Chi-square=60.800, $p < 0.001$, accounting for 85.5% of the between-group
213 variability. Examination of the discriminant loadings (Table 4) showed three significant predictors (i.e.
214 > 0.3), namely SR Musical Skill (.811) and PT DLF (-.343), and BRAMS Avg./Total score (.301). The
215 weakest predictor was IRN DLF (-.169). Cross-validated classification showed that overall, 89.2% of the
216 subjects were correctly classified into MU and NM groups. It should be noted that log determinants of
217 this analysis showed large differences and Box's M was significant, suggesting that the assumption of
218 equality of covariance matrices was violated. However, this problem is somewhat allayed given that
219 normality is not a critical assumption for discriminant analysis.

220 **Discussion**

221 We have answered two main questions in this study: 1) Are musicians better at perceiving specific pitch-
222 related acoustics? and 2) Are psychoacoustic thresholds related to objective and subjective measures of
223 music ability? To answer these questions, we recorded psychoacoustic DLFs to four types of pitch-
224 eliciting sounds in 18 Non-musicians and 20 Musicians and obtained objective and subjective measures
225 of musicianship from all participants. Sound types for DLF measurement included Dichotic (Huggins')
226 pitch, Pure Tones, Iterated Rippled Noise and Complex Tones. Objective measures of musicianship were
227 obtained from the output of the BRAMS online music aptitude test and subjective assessment of music
228 ability and listening were obtained through the questionnaire.

229 To answer the first question, DLF data were subjected to a RMANOVA with four within-subject factors
230 of sound type and two between-subject factors of group. Results showed group differences across all
231 sound types, with the greatest differences for dichotic and pure tone stimuli. These data refute our initial
232 hypothesis that pitch-related temporal encoding mechanisms would be most impacted by musicianship;
233 instead suggesting that music-related plasticity is not restricted to types of pitches. The greatest difference
234 between Musician and Non-musician discrimination thresholds in the dichotic condition suggests that

235 higher-order mechanisms, such as those requiring a combination of sound across the ears, are greatly
236 impacted by musical training.

237 Several hypotheses could reasonably explain our findings. One hypothesis is that mechanisms of music-
238 related brain plasticity are not restricted to place or temporal code encoding mechanisms in peripheral or
239 brainstem nuclei (11), but may also occur cortically (12), or at least beyond the superior olive where
240 dichotic sounds first combine. Unfortunately, our current data do not permit further elucidation on the
241 veracity of this postulate because we do not have encoding data to test brainstem and cortical plasticity
242 specifically. An alternative hypothesis is that playing music sharpens one's ability to extract pitch
243 percepts in conditions where the pitch strength is less salient, such as the dichotic and iterated rippled
244 noise conditions. If this hypothesis were true, we might expect that the largest differences between the
245 two groups would be in the least salient conditions. Whereas the largest threshold difference is in the
246 dichotic condition (less salient pitch), the second largest threshold difference is observed in the pure tone
247 condition, which has the most salient pitch strength. Although our data do not directly address the issue of
248 pitch strength, the fact that the largest differences are observed with both strong and weak pitch percepts
249 diminishes this hypothesis' likelihood. A third, big picture, hypothesis is that Musicians possess a greater
250 aptitude to learn the task than Non-musicians. If this were true, we would expect Musicians to learn the
251 task faster than Non-musicians. A post-hoc examination of the within-session change in threshold showed
252 that Non-musicians did have more variability, measured by standard deviation (Supp. Table 4). However,
253 mean magnitudes of the within-session change in threshold over the four runs, computed by subtracting
254 the threshold obtained in the first run from the threshold obtained in the last run, did not appear to differ
255 between groups (Supp. Table 4). To verify our observations, we performed two RMANOVAs for sound
256 type and group on the standard deviation and within-session change data. Results showed that Musicians
257 had lower standard deviation in thresholds to dichotic and pure tone stimuli, compared to Non-musicians,
258 but only in the pure and complex tone conditions. No significant differences were observed for within- or
259 between-subject comparisons of the within-session threshold change magnitude. Taken together, these

260 data suggest that acclimatization or learning trajectories from task beginning to end is similar in
261 Musicians and Non-musicians and that musicianship positively influences dichotic and pure tone pitch
262 discrimination, in part by stabilizing threshold reliability.

263 In answering the second question, we showed evidence for a relationship between psychoacoustic pitch
264 discrimination and measures of subjective and objective music ability. The correlation data show that
265 discrimination thresholds across all four pitch types were negatively correlated with a higher subjective
266 rating of musicianship, such that individuals who rated themselves with musical ability closer to
267 “professional” on a subjective scale, could hear smaller pitch differences in all four sound conditions.
268 Conversely, individuals who rated themselves with a lower musical ability (i.e. closer to “novice” on the
269 same scale) had greater (poorer) DLFs. To the authors’ knowledge this is the first time that such a
270 relationship has been reported. The results imply that a person’s self-assessment can be a good predictor
271 of their psychoacoustic threshold. It should be noted however, that while the correlations between self-
272 reported music ability and DLF are significant, more than half of the r-values portray a moderately strong
273 relationship (i.e. <0.5); suggesting that other, untested, variables account for additional variance in the
274 relationship. Therefore, while the connection between basic sensory ability and self-assessment of music
275 ability is suggested here, it is only partially accounted for.

276 Regarding objective measures of music ability, it appears that the online music aptitude tests of Melody,
277 Pitch, and the average of all music scores showed a consistent relationship with our psychoacoustic test
278 results. In general, higher scores were related to smaller DLFs, suggesting that those who scored well on
279 the BRAMS tests could discriminate sounds with smaller pitch differences. It is interesting to note that
280 although Melody and Pitch scores correlated with psychoacoustic discrimination thresholds, Timing
281 scores did not. Relatedly, Melody and Pitch were correlated to each other; $r=0.419$, $p=0.010$, but neither
282 correlated with Timing; $p>0.222$. Timing scores did correlate, however, with SR Musical Skill; $r=0.419$,
283 $p=0.010$, and BRAMS Avg./Total score; $r=0.505$, $p=0.001$, suggesting that rhythmic ability is related to
284 musical aptitude and self-assessment of musical skill, but may be independent of pitch perception. Taken

285 together, the correlation data show that the ability to discriminate small pitch differences can be reflected
286 in global musical abilities and an individual's evaluation of their own musical aptitude. This implies that
287 sensory thresholds for pitch discrimination underlie, at least in part, one's musical ability and self-
288 appraisal of that ability. Furthermore, relationships between sensory threshold for pitch and more broad
289 measures of musicianship are not restricted to a specific mechanism of pitch processing.

290 The Discriminant Analysis allowed us to detect the degree to which our variables discriminate between
291 Musicians and Non-musicians. The variables that contributed most to the predictions of group
292 membership were 1) Self-report of musical ability on a scale of 1-9, 2) Pure Tone DLFs and 3) BRAMS
293 Avg./Total score. While the relationship between pure tone perception, musical aptitude and musicianship
294 is well established, the contribution of a self-report variable is novel as far as the authors' knowledge.
295 Here, we show that self-evaluation of musical competence can be meaningfully applied to classify groups
296 and is related to objective measures of music and perceptual ability. Self-evaluation of competence, or
297 self-competence is defined as the sense of one's capacity. (31) Previous data on this topic show that
298 general self-competence is as associated with measures of cognitive ability such as IQ and academic
299 achievement measured by GPA. (32) Our data support the argument that self-evaluation of competence is
300 a meaningful measure of ability and outcomes (33) and extend into musicianship.

301 In addition to the finding of self-report as a meaningful measure, the discriminant analysis showed
302 common characteristics of musicians include psychoacoustic, musical and self-evaluated abilities. This
303 gives rise to the notion that all three areas may interact to define a person who is talented or skilled in
304 music. It is interesting to note that the self-reported music listening scale did not distinguish between
305 groups. This supports several lines of research showing that active music-making, rather than listening
306 alone, is a catalyst for brain plasticity and internalized perceptual change (23, 34, 35).

307 In conclusion, this study sheds light on several aspects of musicianship. First, we show that the influence
308 of musicianship is not limited to pitch judgements involving monotic/diotic mechanisms but also includes
309 those that rely on dichotic integration. Second, our data show that basic perceptual thresholds are related

310 to measures of both subjective and objective musical ability. And third, the data suggest that self-
311 evaluation of musical ability is a meaningful part of musicianship such that high evaluation of
312 competence are characteristic of musician group members. Taken together, the data update the
313 neurobehavioral profile of musicians and extend creative ability measurements into new arenas.

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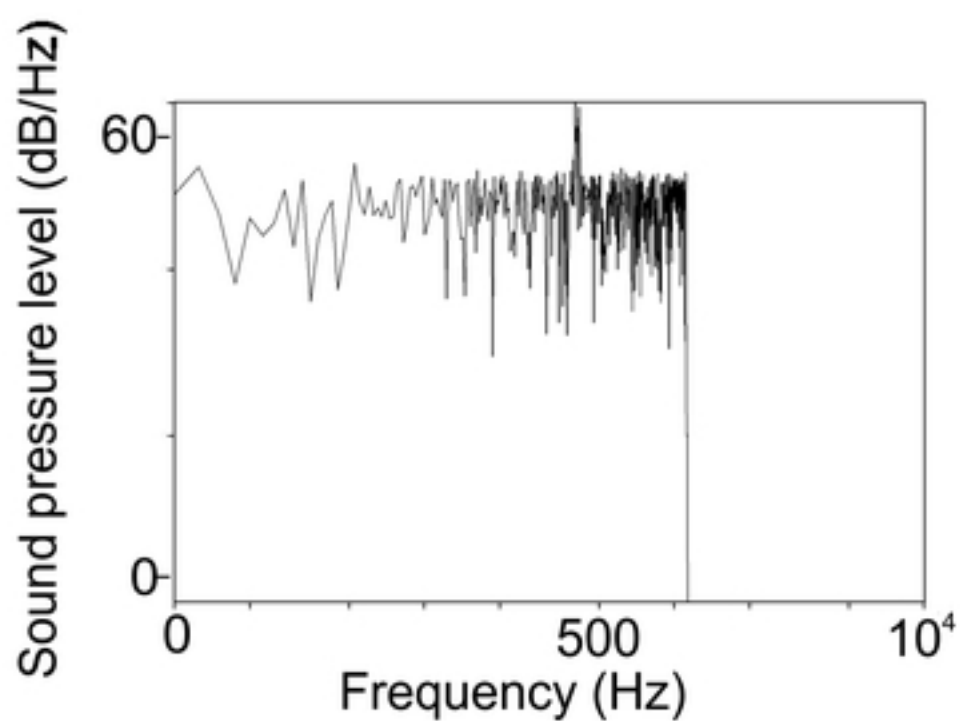
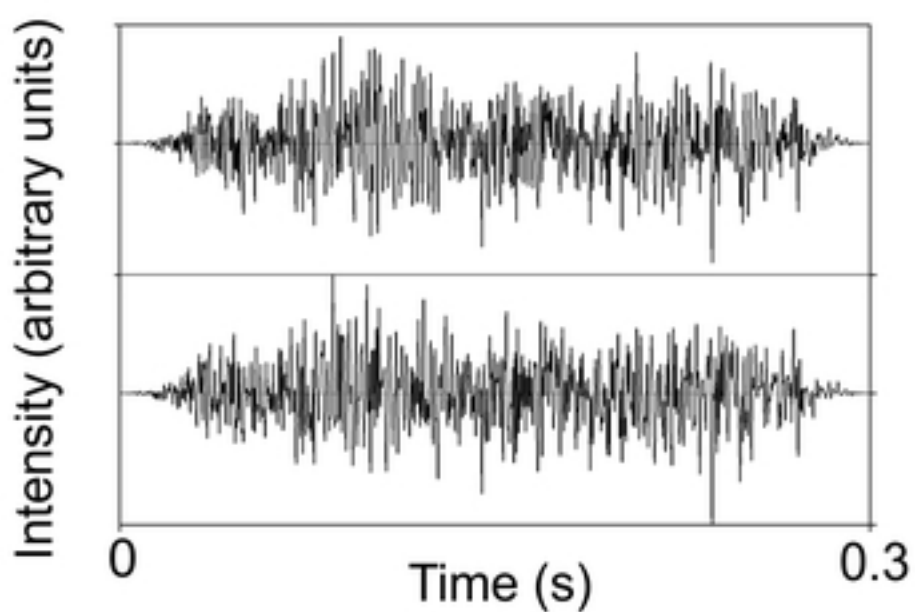
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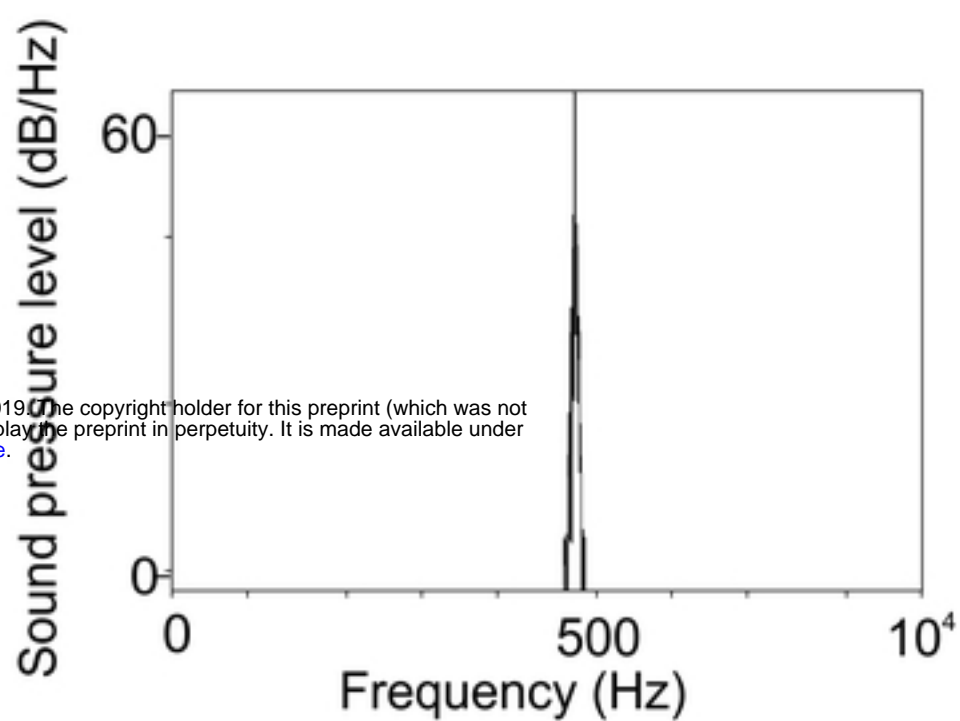
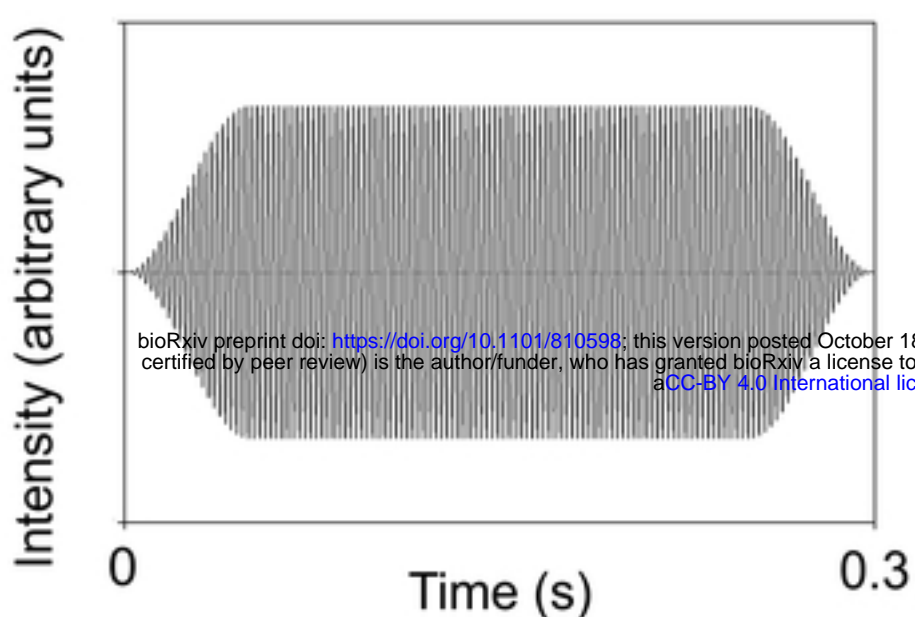
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Figure 1. Stimuli

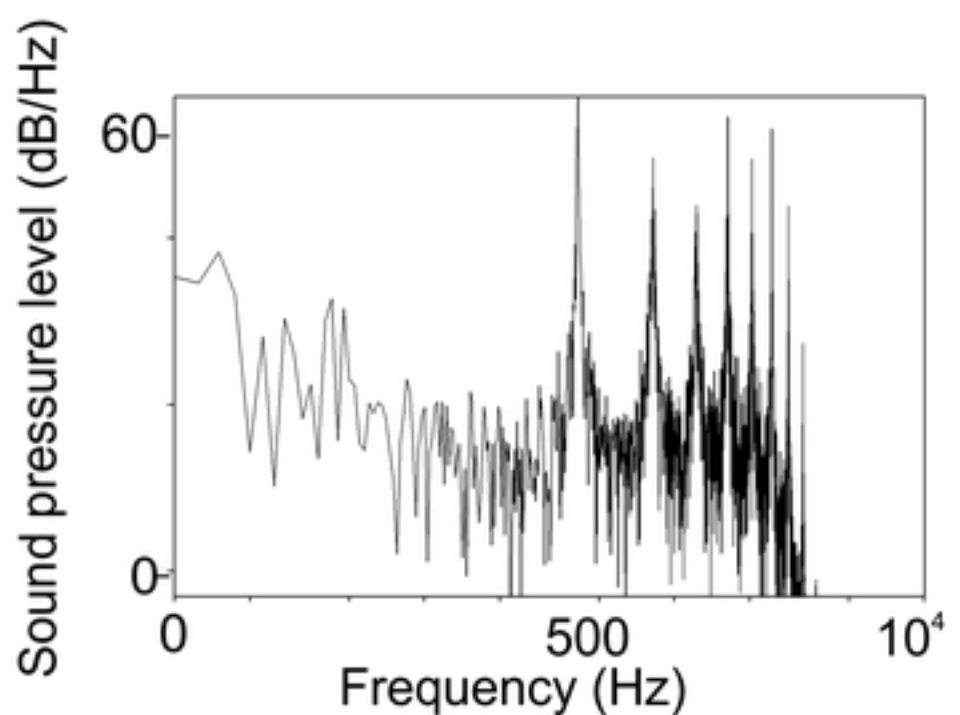
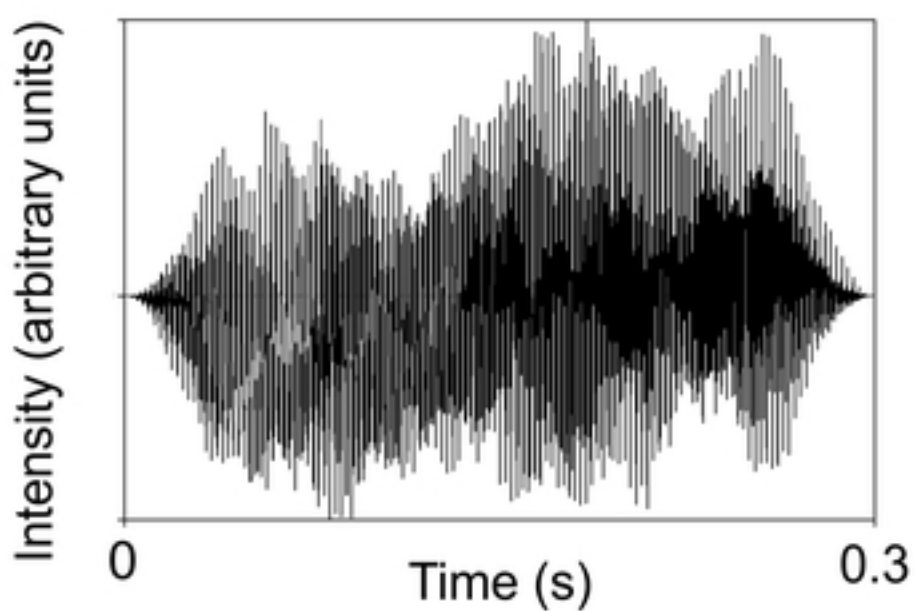
A. Dichotic Pitch



B. Pure Tone



C. Iterated Rippled Noise



D. Complex Tone

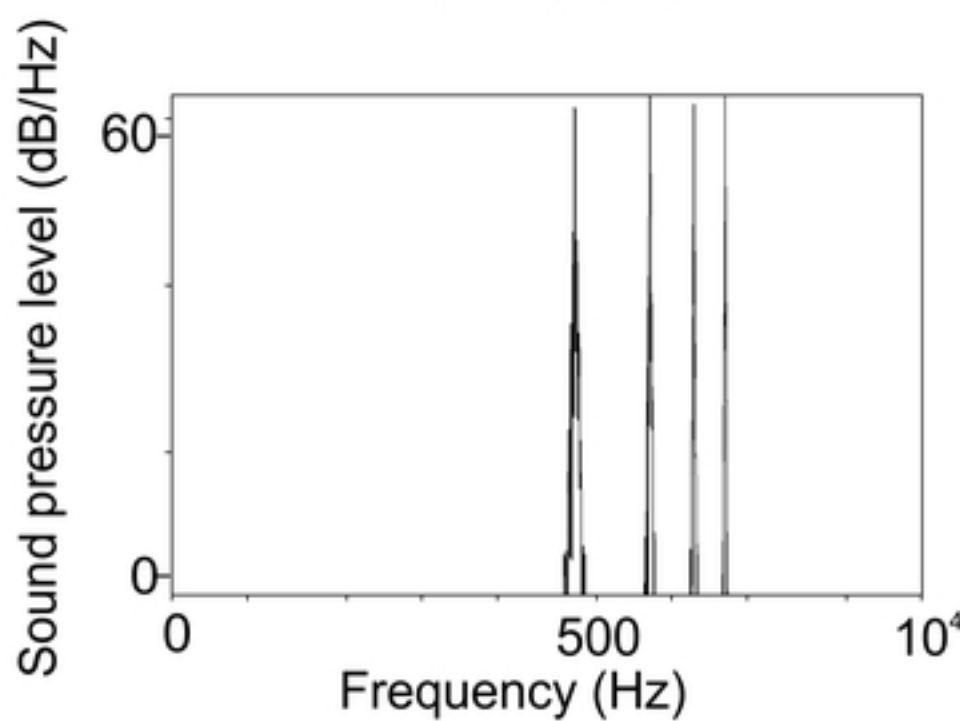
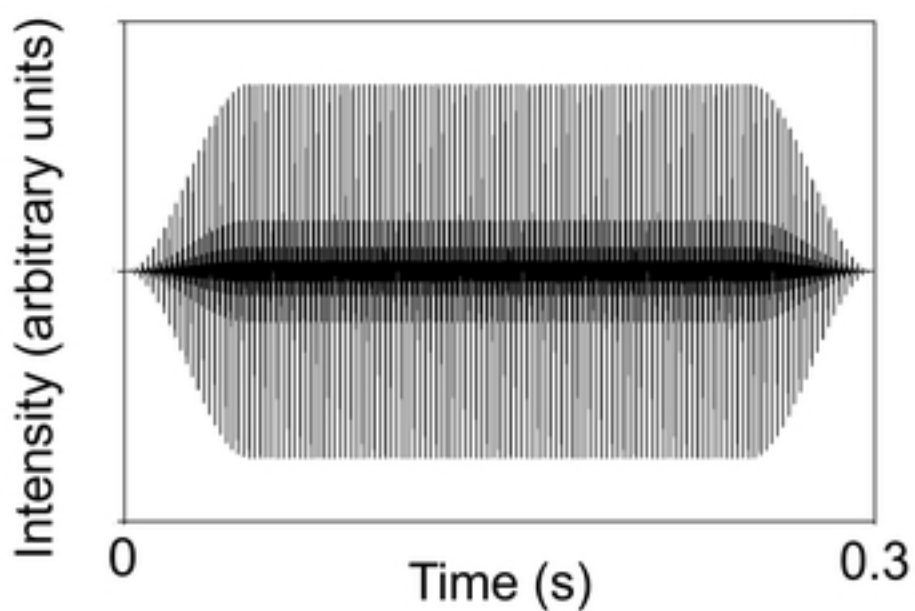


Figure 1

Figure 2.

Mean Pitch discrimination thresholds in Musicians and Non-musicians

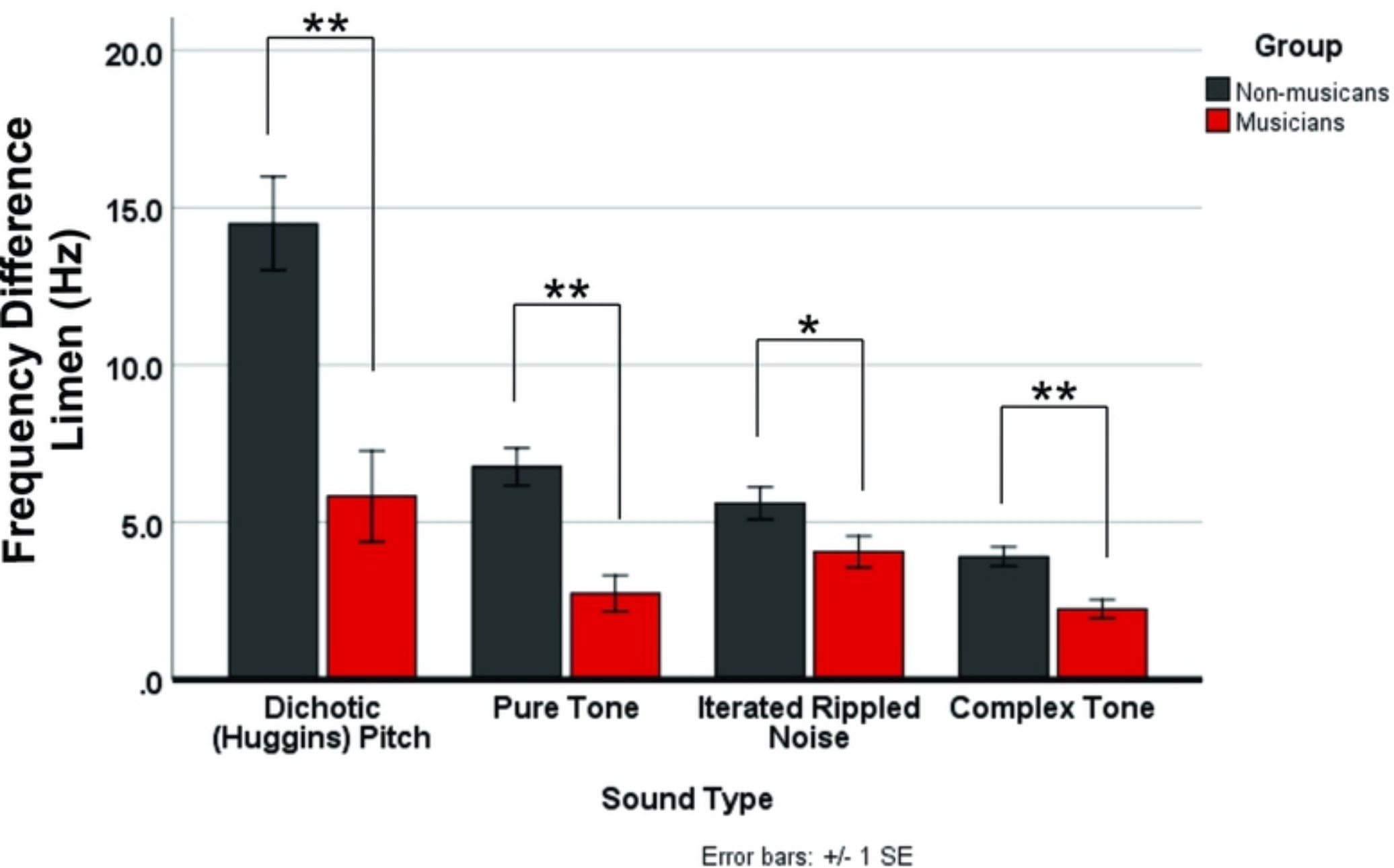


Figure 2

Figure 3.

Relationships between pitch discrimination thresholds, and measures of objective and subjective music ability

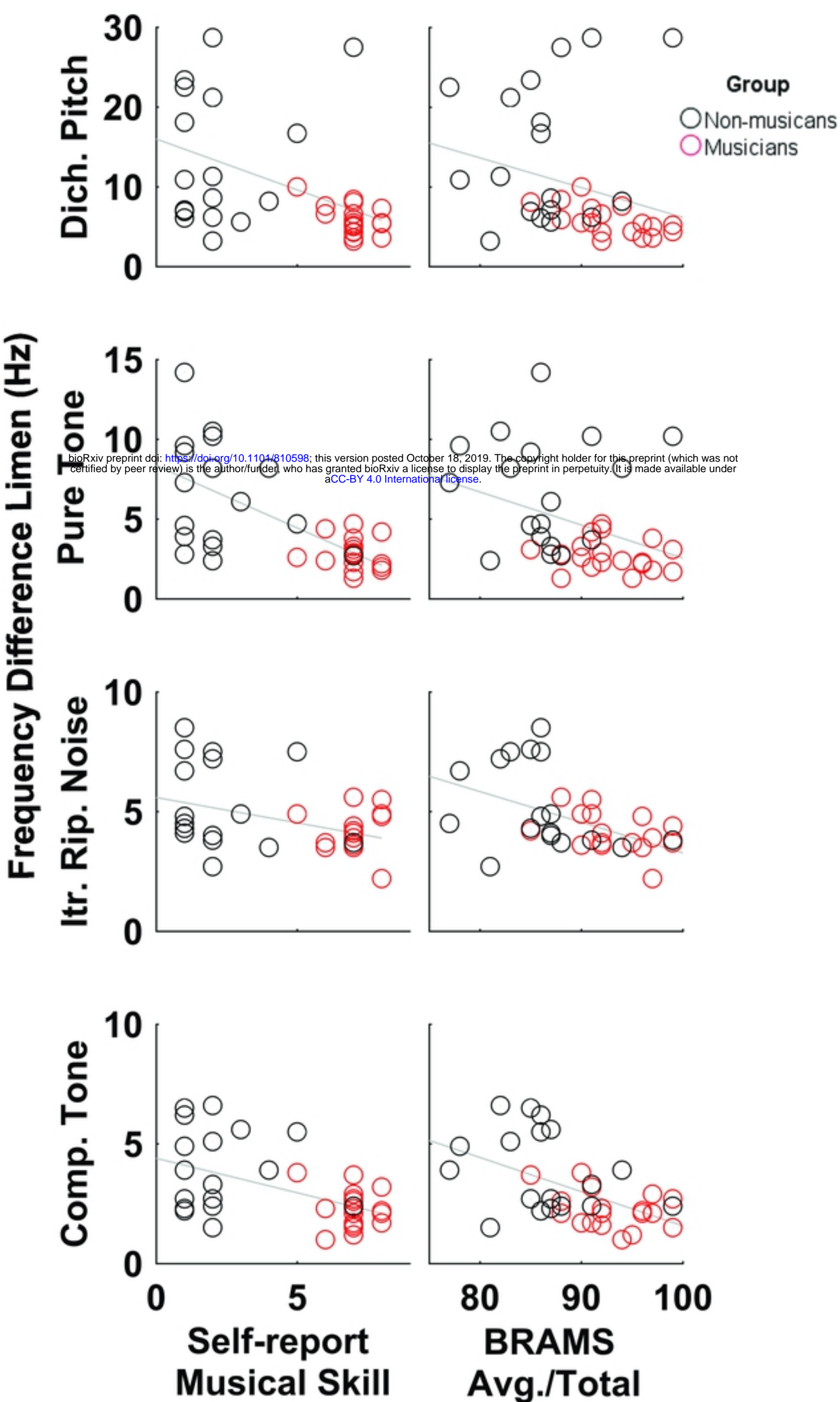


Figure 3