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**Objective classification of North Atlantic right whale (*Eubalaena glacialis*) vocalizations to improve passive acoustic detection**

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25           **ABSTRACT**

26           Passive acoustic monitoring is playing an increasing role in the detection of endangered  
27 North Atlantic right whales (NARW). Previous acoustic monitoring has relied on a single  
28 stereotyped vocalization, the upcall. Here the entire repertoire produced by NARW during the  
29 winter and early spring in Cape Cod Bay, Massachusetts is described. An objective sound  
30 classification scheme and automatic classification algorithm were developed. Nine days of  
31 acoustic recordings were used for the data analysis and a total of 9,611 right whale sounds were  
32 identified. The objective classification scheme of right whale sounds allowed for rapid  
33 identification of a diversity of right whale sounds. These sounds were assigned to 6 classes of  
34 narrowband upcalls, downsweep, complex and high frequency calls, wideband gunshot sounds  
35 and complex sounds. Results indicate that the prevalence of upcalls varied from 28% of detected  
36 calls in January to 80% in April. Other classes of signals were also well represented in the  
37 repertoire including the narrowband complex (10-36%) and high frequency calls (1-26%),  
38 wideband gunshot sounds (4-25%) and wideband complex sounds (0 – 25%). The prevalence of  
39 non-upcall signals suggests that including more signals classes may improve rates of detection  
40 for right whales in the Cape Cod Bay habitat.

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44 **INTRODUCTION**

45 Research into the vocal behavior of the critically endangered North Atlantic right whale  
46 (*Eubalaena glacialis*) (NARW) has been conducted over the past decade (Matthews *et al.*, 2001;  
47 Mellinger *et al.*, 2007; Parks and Tyack, 2005; Parks *et al.*, 2011; Urazghildiiev *et al.*, 2009;  
48 Vanderlaan *et al.*, 2003). These studies are important to provide both baseline behavioral data  
49 and to inform conservation measures for this species. Long-term passive acoustic monitoring of  
50 NARW habitats has been used to successfully determine seasonal presence of individuals to  
51 contribute to near-real-time implementation of conservation measures (e.g., Van Parijs *et al.*,  
52 2009; Clark *et al.*, 2010).

53 One of three known areas which receive federal designation as a critical habitat area of  
54 NARW (Federal Register 59 FR 28793) is the Cape Cod Bay ecosystem. This area is used  
55 primarily for feeding, socializing, and as nursery area for cows and their calves (Clark *et al.*,  
56 2010). To acoustically detect the presence of NARW in this area, the upcalls, or contact calls, are  
57 typically used because this type of calls is the most stereotyped right whale vocalization that is  
58 known to be produced by both males and females (Parks and Tyack, 2005; Parks *et al.*, 2011).  
59 However, different call types classified as “tonals” and “gunshot sounds” have also been  
60 detected in Cape Cod Bay in March and April 2004 (Van Parijs *et al.*, 2009). Other call types that  
61 have been described from digital acoustic tag (DTAG) based recordings from North Atlantic  
62 right whales include “moans” (Matthews *et al.*, 2001), as well as “screams,” (corresponding to  
63 “high,” “hybrid,” and “pulsive” calls produced by Southern right whales (Clark, 1983)), “blows,”  
64 “warbles,” and “downcalls” (Parks *et al.*, 2011). The relative proportion of different right whale  
65 vocalization types produced varies seasonal and between habitats (Van Parijs *et al.* 2009).  
66 Therefore, ignoring call types other than upcalls may result in missing essential information

67 about the presence or behavior of NARW in a particular area. Therefore, a quantitative  
68 evaluation of the usage of the vocal repertoire of NARW and more objective methods for  
69 classification of their other call types in different habitats and times of year are necessary to  
70 improve passive acoustic detection for this species.

71 This study focused on characterizing the multiple call types recorded from NARW during  
72 the late winter and spring in Cape Cod Bay using data collected by a sparse array of bottom-  
73 mounted synchronized hydrophones. Due to the high variability in the parameters of detected  
74 NARW vocalizations, their classification is not a trivial problem. The choice of a practical  
75 classification scheme depends on the goals of classification as well as on the presence of well  
76 separated signal parameters. In this work, a classification scheme based on measurable signal  
77 parameters was developed to allow for the design of an automatic classifier. The goal of this  
78 work was to design a practical sound classification scheme, to develop an automatic classifier,  
79 and to obtain empirical distributions of signal classes to better understand right whale vocal  
80 behavior and to improve passive acoustic monitoring applications.

81

## 82 **METHOD**

### 83 **Acoustic recordings**

84 Data were collected using a passive acoustic monitoring system consisting of 11  
85 synchronized marine autonomous recording units (MARUs) located at known positions. The  
86 MARUs were approximately 8 km apart (Fig. 1). Each MARU consisted of an HTI-94-SSQ  
87 hydrophone (High Tech, Inc., Long Beach, Mississippi, USA) with a sensitivity of -168 dB re 1  
88 V/ $\mu$ Pa, an amplifier with a gain of 23.5 dB and A/D converter with a sensitivity of 103 Bit/V  
89 (Parks *et al.*, 2009). The units had a flat ( $\pm 1.0$  dB) frequency response between 10 – 2500 Hz.

90 All MARUs recorded continuously at a sampling rate of 5 kHz. Acoustic recordings were made  
91 from February 8 to May 3, 2012.

## 92 **Signal detection**

93 Signal detection was performed in three stages. Upcalls are considered to be the most  
94 stereotyped distinctive NARW vocalization; therefore an upcall detector (Urazghildiiev *et al.*,  
95 2009) was applied in the first stage to identify periods of right whale acoustic activity. The  
96 detector thresholds were selected in a way to detect >95% of upcalls clearly visible on the  
97 spectrograms in three or more channels. All signals detected in this stage were checked by the  
98 human operator and false detections occurring due to noise transients were removed from the  
99 testing data set. To detect other types of NARW vocalizations, an energy detector was applied in  
100 the second stage. The NARW vocalizations were identified from all automatically detected  
101 sounds by visual inspection of the signal spectrograms. Signals with SNR > 10 dB were used in  
102 this study. Signals were identified as NARW calls if they fit any of the call types described in the  
103 literature, such as moans, screams, downcalls, high calls, gunshot sounds, and other calls  
104 observable within the frequency range below 2.5 kHz (see, e.g., Matthews *et al.*, 2001; Parks *et*  
105 *al.*, 2011; Parks and Tyack, 2005; Van Parijs *et al.*, 2009; Clark, 1982; Clark *et al.*, 2010).

106 To avoid misclassification of NARW calls with similar calls produced by humpback  
107 whales, data collected after April 8, 2012, when the first song produced by humpback whales  
108 was detected in the North part of Cape Cod Bay by recorders 1 and 8, have not been used for the  
109 analysis. All selected calls were also checked to make sure that other humpback whale  
110 vocalizations, such as patterned repetitive vocalizations or humpback whale stereotyped calls  
111 described in Stimpert *et al.*, (2011) were not included in the testing data. In the third stage, visual

112 inspection of the spectrogram by a human operator was performed and all NARW vocalizations  
 113 not detected in in the previous stages were added to the testing data set.

## 114 **Signal classification**

115 Impulsive signals produced by NARW were classified based on types of distribution of  
 116 signal energy in time and frequency domain. One of the basic features for classification is the  
 117 local bandwidth of signals. Using this parameter, two main classes of NARW vocalizations were  
 118 introduced. The first class comprised locally narrowband frequency-modulated (FM) signals  
 119 whose energy is distributed in a vicinity of an instantaneous frequency and its harmonics. The  
 120 FM signals can be represented as

$$121 \quad s(t) = \sum_i A_i(t) \cos(2\pi i f(t)), \quad (1)$$

122 where function  $f(t)$  specifies frequency modulation of signals, and  $A_i(t)$  is an amplitude  
 123 modulation of the  $i$ th harmonic.

124 The second main class combined wideband signals having wider distribution of energy in  
 125 frequency domain, such as gunshot sounds, exhalations, and slaps (Matthews *et al.*, 2001; Parks  
 126 and Tyack, 2005; Clark, 1982).

127 The next feature used for classification of NARW vocalizations was the frequency band  
 128 occupied by signals. The major part of the first harmonics of locally narrowband NARW  
 129 vocalizations is distributed within two frequency bands situated lower and higher than 200 Hz.  
 130 Correspondingly, NARW tonals were classified as low frequency (upcalls, downcalls, moans, <  
 131 200 Hz) and high frequency (high calls, > 200 Hz) locally narrowband signals.

132 Classification of locally narrowband vocalizations was also made based on behavior of  
 133 the FM function  $f(t)$ . This function can either be monotonic or have a finite number of  
 134 inflection points, i.e., the points  $t_k, k = 1 \dots K$ , where  $df(t)/dt = 0$ . The more complicated a

135 locally narrowband vocalization is, the higher number  $K$  of inflection points its FM function has.  
136 By using this parameter, locally narrowband vocalizations were classified as simple ( $K \leq 1$ ) and  
137 complex ( $K \geq 2$ ) signals. Finally, simple FM signals were classified as upsweep, constant  
138 sweep, and downsweep signals depending on the number of points  $t_k$  where  $df(t)/dt$  is larger,  
139 equal or lower than zero, respectively.

140 The energy distribution of wideband NARW vocalizations in the time-frequency plane is  
141 less structured. The most distinctive and important type of wideband calls is the gunshot sound,  
142 which typically has shorter duration than other types of wideband vocalizations. Therefore two  
143 subclasses of wideband sounds were considered: gunshot sounds (and other slaps), and complex  
144 wideband vocalizations.

145

146

## 147 RESULTS

### 148 Classification scheme

149 Using the features of NARW vocalizations described in section II.B, we introduced the  
150 following classes of signals:

151 Class NU: locally narrowband low frequency upsweep signals. The FM function  $f(t)$  of  
152 NU signals has one or less local minimum or maximum and increases in time such that the  
153 number of points where  $df(t)/dt \geq 0$  is larger than 50% of total points associated with signal.  
154 The peak frequency of NU signals  $F_p < 200$  Hz and instantaneous bandwidth of the first  
155 harmonic  $B < 100$  Hz.

156 Class ND: locally narrowband low frequency constant sweep and downsweep signals.

157 Signals associated with this class were similar to NU calls except for the instantaneous frequency

158 of ND signals,  $f(t)$ , is a non-ascending function, i.e., the number of points where  $df(t)/dt \leq 0$   
159 is larger than 50% of total points associated with signal. Constant sweep and downsweep signals  
160 were combined into one class since the proportion of this type of vocalizations is relatively small  
161 as compared to other types of calls.

162 Class NC: locally narrowband low frequency complex signals. Signals of this class were  
163 similar to NU calls except for the number of local minimums or maximums is greater than 1.

164 Class NH: locally narrowband high frequency signals. This class comprised all locally  
165 narrowband calls having peak frequency  $F_p \geq 200$  Hz and any FM function  $f(t)$ .

166 Class WG: wideband gunshot sounds. The instantaneous bandwidth of WG signals  
167  $B \geq 100$  Hz and duration  $D < 1.5$  s. Separate frequency components of WG signals were  
168 typically indistinguishable on the spectrogram.

169 Class WC: wideband complex signals. The instantaneous bandwidth of WC signals  
170  $B \geq 100$  Hz and duration  $D \geq 1.5$  s. Frequency components of some WC signals were  
171 distinguishable on the spectrogram, but they were typically separated by less than 50 Hz.

172 Example spectrograms of signals from each class are shown in Figs. 2 – 7. The  
173 spectrograms were obtained using 1024 point FFT (frequency resolution is 4.88 Hz) with Hann  
174 window and 75% overlap. Note that the wideband signal displayed in Fig. 6, right, does not look  
175 like typical gunshot (see Fig. 6, left and middle). However, this signal was attributed to WG class  
176 due to its parameters' fit to this class.

177 Note that classification of locally narrowband signals by the complexity of the FM  
178 function is applicable to both low frequency and high frequency sounds. However, this kind of  
179 classification was applied to low frequency sounds only since upcalls traditionally used for  
180 passive acoustic detection of NARW belong to this class of locally narrowband signals. The need



181 for more detailed classification of low frequency sounds is explained by the fact that the  
182 detection performance of the automatic upcall detectors (Gillespie 2004, Urazghildiiev *et al.*,  
183 2009) degrades as the complexity of upcalls increases, so considering simple and complex low  
184 frequency tonal calls as separate classes can be useful for improving the efficiency of the  
185 automatic NARW detectors. For the sake of simplicity, all high frequency locally narrowband  
186 calls were associated with the same class.

### 187 **Automatic classifier**

188 Measurable signal parameters used for the design of the classification scheme described  
189 above allowed developing an automatic classifier. The block-diagram of the automatic classifier  
190 is represented in Fig. 7. The classifier was implemented as a custom-build Matlab program. It  
191 was used to aid the human operator to classify NARW vocalizations. The human operator  
192 checked classes computed by the automatic algorithm for all detected signals and changed  
193 classes of signals incorrectly classified automatically. Testing the performance of the automatic  
194 classifier was outside the scope of this work.

### 195 **Empirical distribution of classes**

196 A total of 9 chunks of acoustic recordings, 24 hours each, collected on February 9, 16, 23  
197 and 27; March 4, 12, 17, 23; and April 1 and 7, 2012, were used for the data analysis. These days  
198 were selected because the results of visual survey indicate the presence of NARW in the vicinity  
199 of the array and the days were temporally separated to allow for sampling of different individual  
200 whales.

201 The results of classification of all detected NARW vocalizations are represented in Table  
202 I. Daily distribution of detected NARW calls and their classes are displayed in Fig. 9. Fig. 10  
203 displays empirical distribution of classes of all detected NARW vocalizations.

204

## 205 **DISCUSSION**

206 The vocalization repertoire of NARW during their spring migration in Cape Cod Bay is  
207 highly variable. Using the distribution of signal energy in the time and frequency domains of  
208 detected right whale signals has led to a new scheme for the classification of NARW  
209 vocalizations. An important advantage of the proposed classification scheme is that it is based on  
210 a finite set of measurable signal parameters: instantaneous bandwidth, peak frequency, duration,  
211 and the number of inflection points in the FM function. Using these parameters, an automatic  
212 classifier was designed and implemented in this study

213 Statistical analysis and classification of the diversity of signals produced by right whales  
214 in this habitat yielded six robust, quantifiable classes to provide additional signals to aid in  
215 passive acoustic monitoring of this species. These classes included upcalls (NU), three additional  
216 tonal categories (downsweep (ND), narrowband complex (NC), high (NH)), and two broadband  
217 categories (gunshot/slaps (WG), and wideband complex (WC)). All previously reported call  
218 types (see, e.g., Matthews *et al.*, 2001; Parks *et al.*, 2011; Parks and Tyack, 2005; Van Parijs *et*  
219 *al.*, 2009; Clark 1982; Clark *et al.*, 2010) can be classified using the proposed classification  
220 scheme and an automatic classifier. Using distribution of signal energy in time and frequency  
221 domain, an automatic classifier was designed and implemented in this study.

222

223 The sound production activity levels of NARW changed significantly over the course of  
224 the observation period. Sound production rates were highest in February and decreased in March.  
225 The relative usage of the components of the vocalization repertoire also changed. The relatively  
226 simple upcalls (NU class) typically used for passive acoustic monitoring of NARW comprise  
227 48.7% of all detected NARW vocalizations, and daily occurrence of upcall changes from 28 % to  
228 80.2%. This finding alone is intriguing, as it indicates that while upcalls were always detected  
229 (by default based on our methods for detection of other right whale signals), their prevalence  
230 within the repertoire of signals produced on a given day varied widely. This variation is likely  
231 due to a combination of different individual whales and variability in the behavioral context of  
232 signal production.

233 Signal types that are not typically considered in passive acoustic monitoring or detection  
234 or right whales comprised a significant portion of the detected signals over the study period.  
235 Signals with more complex frequency modulation (NC) and high calls (NH) comprise 9.9 –  
236 35.8% and 0.5 – 25.8% of total detected calls, respectively. The percentage of high calls (NH  
237 class) was greater in February and March. The production rate of wideband gunshot sounds  
238 (WG) varied from 4.1 to 24.6%. These values are comparable to those reported in Van Parijs *et*  
239 *al.*, 2009, which found that in the Cape Cod Bay habitat, upcalls were the predominant call type,  
240 with fewer more complex tonal signals and low rates of gunshot sound production during spring  
241 months.

242 These types of analysis may also provide insight into the behavioral activity of right  
243 whales in different habitat areas or in different times of years (Van Parijs *et al.*, 2009). Previous  
244 studies have indicated different behaviors associated with the production of these different sound  
245 types (Parks *et al.*, 2005; Parks and Tyack 2005; Parks *et al.*, 2011). Analysis of seasonal

246 variation in the production of a single call-type, the upcall, in the Gulf of Maine showed both  
247 seasonal and diel trends in sound production (Mussoline *et al.*, 2012). Future studies could  
248 expand this type of analysis to include additional call types produced by right whales.

249 This study indicates that applying NARW upcall detectors (e.g., Dugan *et al.*, 2010;  
250 Gillespie, 2004; Urazghildiiev and Clark, 2006; Urazghildiiev *et al.*, 2009) may result in missing  
251 a large number of non-upcall vocalizations and potentially, in the worst case scenario, result in  
252 missing the detection of vocalizing NARW or groups of whales if their vocalization session does  
253 not contain upcalls. Therefore, we suggest that the design and testing of more complicated  
254 detection techniques sensitive to non-upcall vocalizations should be an area of future research  
255 and will potentially allow for more detailed behavioral information to be obtained from these  
256 passive acoustic monitoring systems.

257 This study should be considered preliminary, as it is based on the acoustic behavior of  
258 right whales from 9 days in a single year. Further studies are necessary to assess the added  
259 benefit of adding detection of these additional signals for right whale passive acoustic  
260 monitoring. Further studies should also involve processing more data, evaluation the individual  
261 statistical properties of vocalization sessions produced by each vocalizing NARW or groups of  
262 closely spaced NARW and the design of the tools for automatic detection and classification of all  
263 NARW vocalizations.

264

265 **ACKNOWLEDGMENTS**

266 The authors wish to thank Dr. Christopher W. Clark for help with lab support and useful  
267 comments. Thanks also to Christopher Tessaglia-Hymes, Christopher Tremblay and Jason John  
268 Michalec for deployment and recovery of the pop-up recording units and Janelle Morano for  
269 useful discussions.

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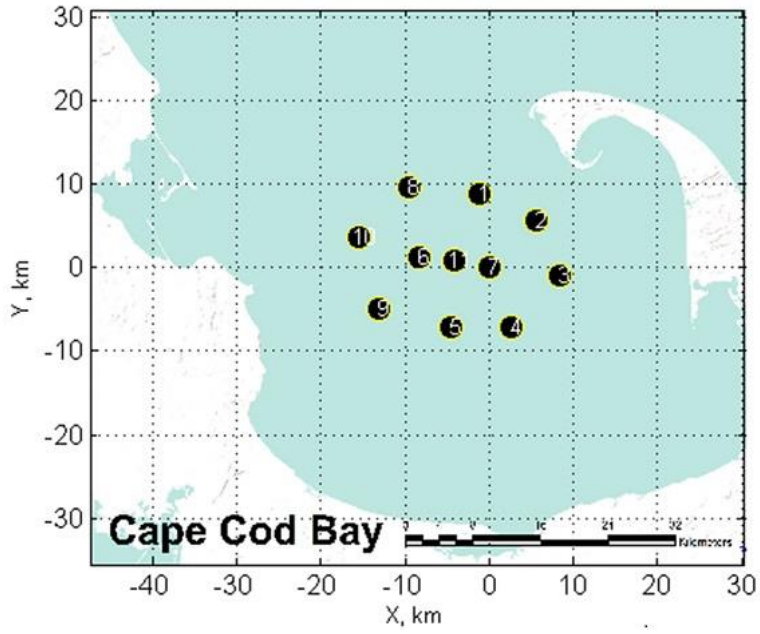
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TABLE I: The number of detected signals from each class.

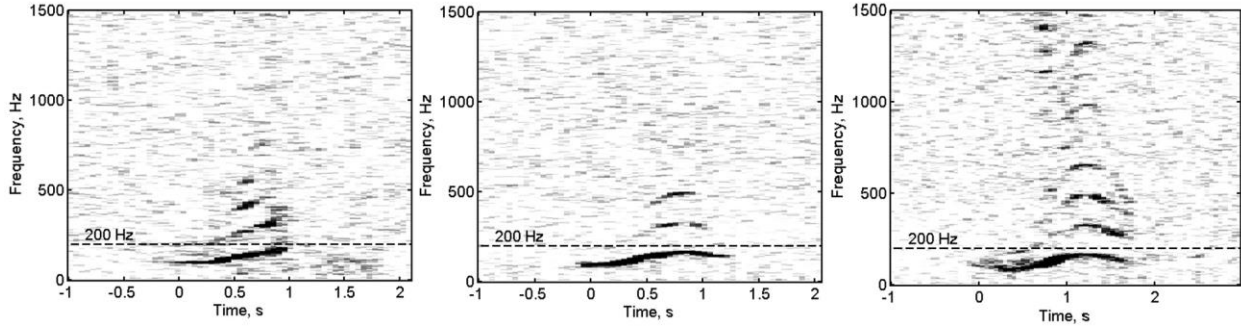
Date	Class						Nsig
	Locally Narrowband				Wideband		
	Low frequency			High frequency (NH)	Gunshot (WG)	Complex (WC)	
	Simple		Complex (NC)				
	Upsweep (NU)	Down sweep (ND)					
9-Feb	532 (28.1%)	10 (0.5%)	254 (13.4%)	490 (25.8%)	146 (7.7%)	464 (24.5%)	1896
16-Feb	535 (49.7%)	12 (1.1%)	252 (23.4%)	196 (18.2%)	55 (5.1%)	26 (2.4%)	1076
23-Feb	1002 (60.3%)	36 (2.2%)	317 (19.1%)	176 (10.6%)	99 (6.0%)	33 (2.0%)	1663
4-Mar	1091 (54.5%)	30 (1.5%)	350 (17.5%)	271 (13.6%)	151 (7.5%)	107 (5.3%)	2000
12-Mar	507 (77.5%)	3 (0.5%)	75 (11.5%)	40 (6.1%)	27 (4.1%)	2 (0.3%)	654
17-Mar	474 (32.4%)	22 (1.5%)	266 (18.2%)	299 (20.5%)	360 (24.6%)	40 (2.7%)	1461
23-Mar	286 (61.2%)	14 (3.0%)	92 (19.7%)	16 (3.4%)	57 (12.2%)	2 (0.4%)	467
1-Apr	111 (52.4%)	1 (0.5%)	76 (35.8%)	4 (1.9%)	18 (8.5%)	2 (0.9%)	212
7-Apr	146 (80.2%)	3 (1.6%)	18 (9.9%)	1 (0.5%)	14 (7.7%)	0 (0.0%)	182
Total	4684 (48.7%)	131 (1.4%)	1700 (17.7%)	1493 (15.5%)	927 (9.6%)	676 (7.0%)	9611

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329  
330 FIG. 1. Map of the sensor array geometry. MARU locations are shown as black,  
331 numbered circles.  
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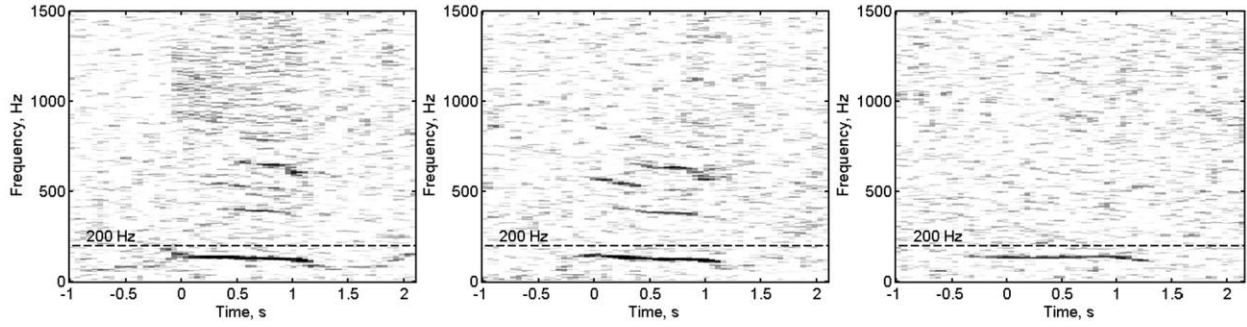
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FIG. 2. Spectrograms of narrowband up-sweep FM signals (NU). The spectrograms were

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obtained using 1024 point FFT with Hann window, 75% overlap, and sampling rate of 5 kHz.

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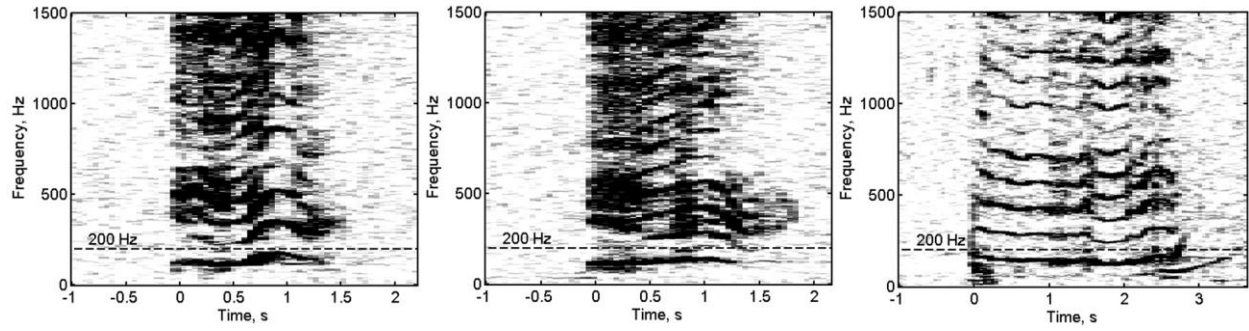


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FIG. 3. Spectrograms of narrowband downswEEP FM signals (ND).

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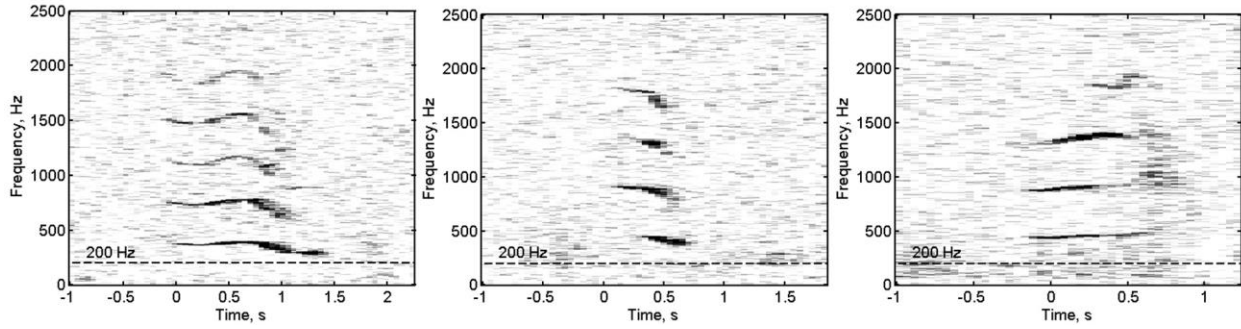


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FIG. 4. Spectrograms of narrowband complex FM signals (NC).

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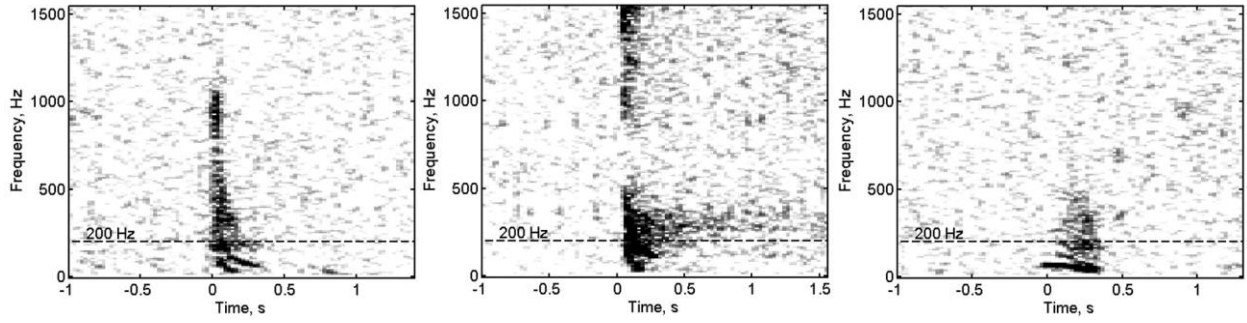


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FIG. 5. Spectrograms of narrowband high calls (NH).

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FIG. 6. Spectrograms of wideband gunshot sounds (WG).

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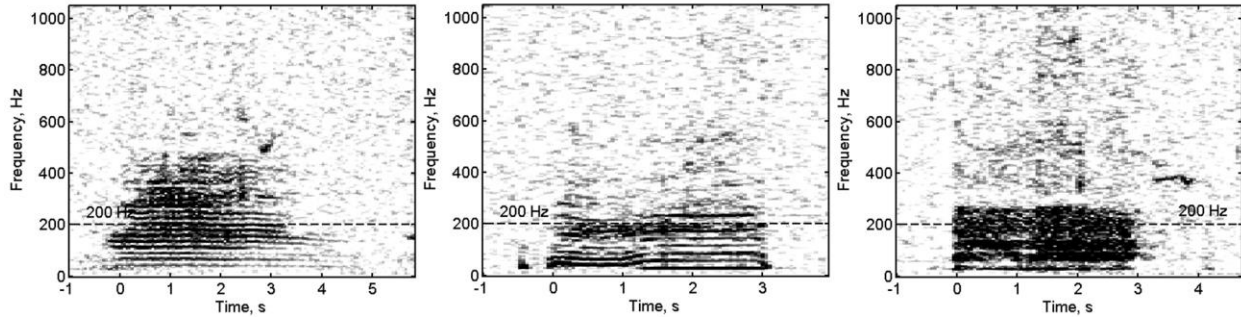
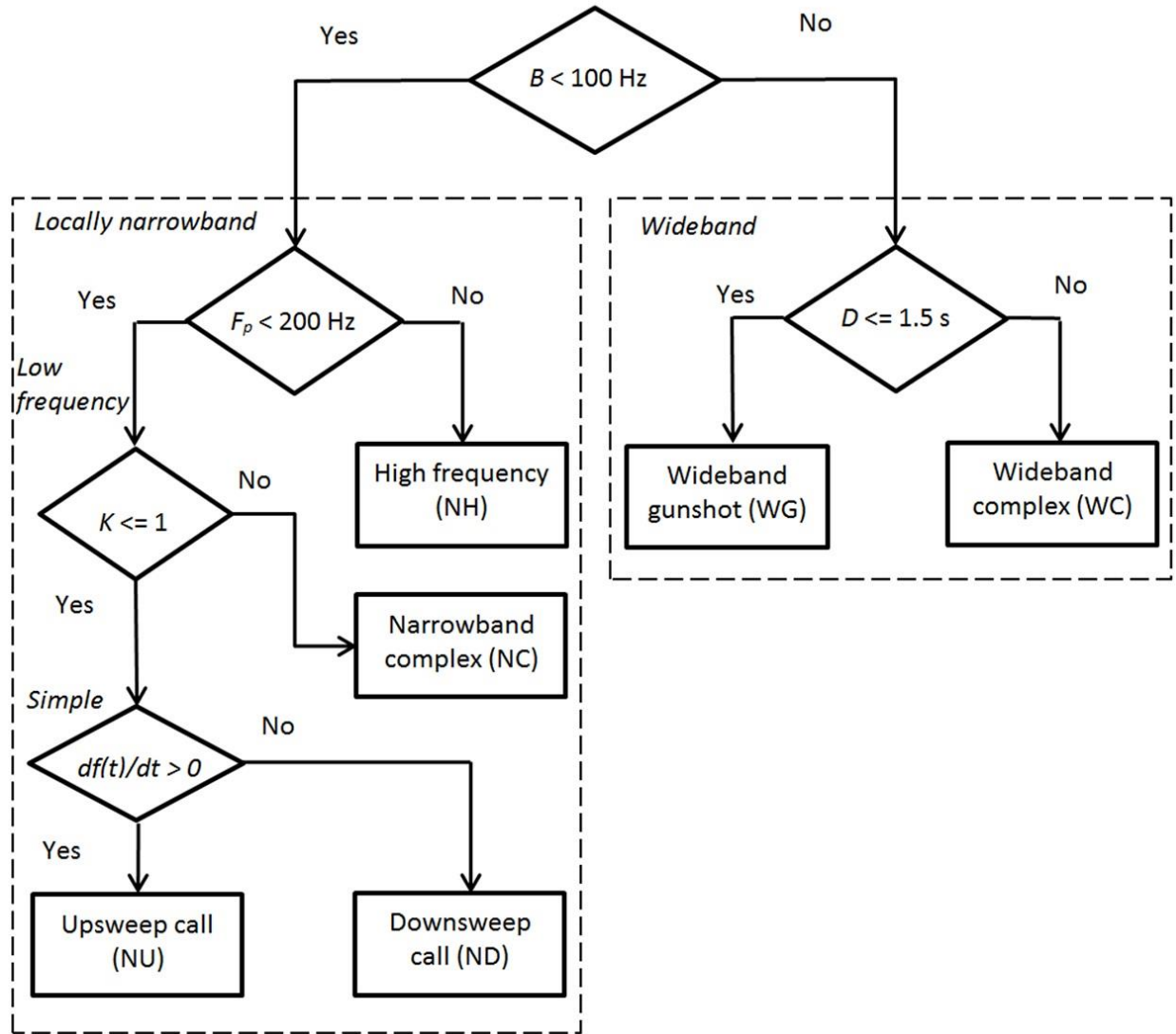


FIG. 7. Spectrograms of complex wideband signals (WC).





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FIG. 8. Block-diagram of the automatic classifier.

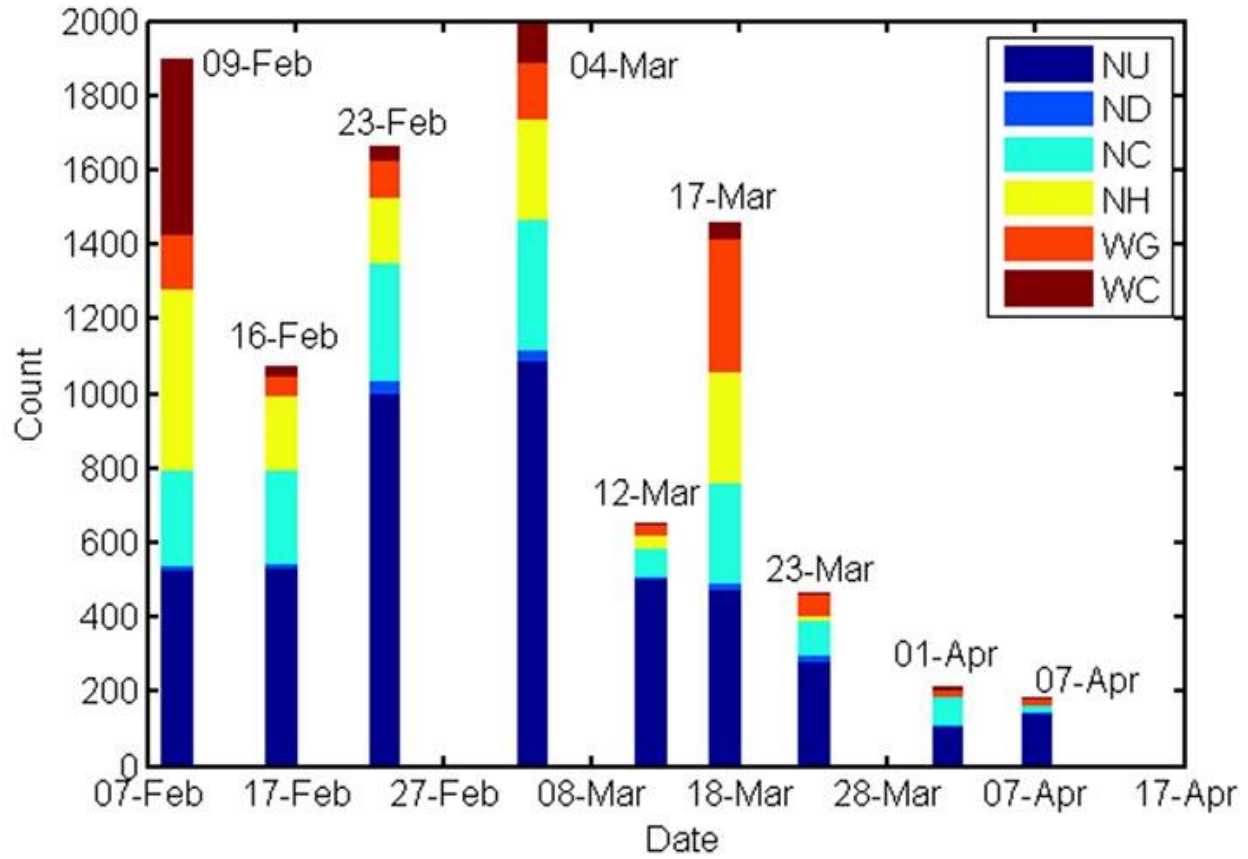


FIG. 9. Daily distribution of detected NARW calls and their classes.

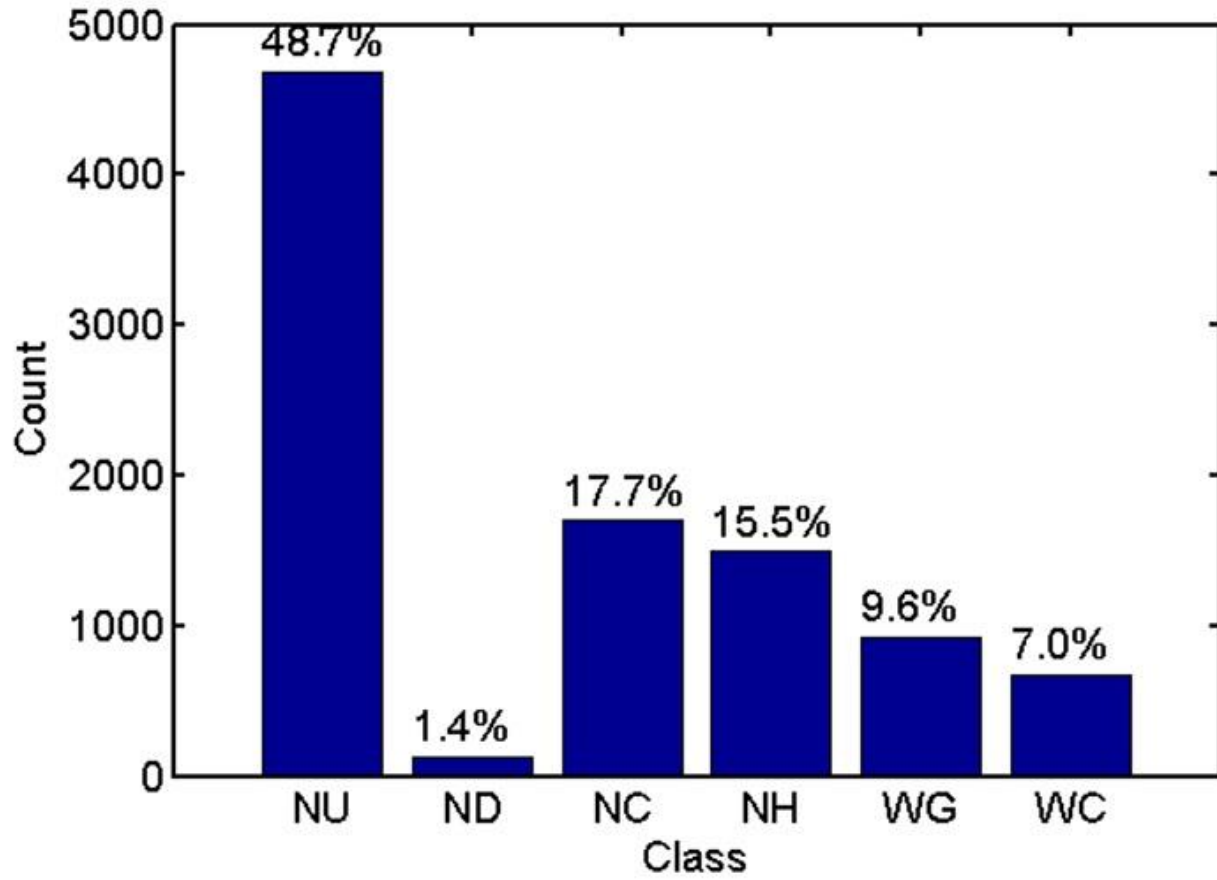


FIG. 10. Empirical distribution of classes of all detected NARW vocalizations.

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