

1 **Human versus automated detection of killer whale calls¹**

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9 (Dated: 7/14/12)

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Abbreviated title: Human vs automated detection of killer whales

1 This paper is intended for the special issue on Marine Mammals and Passive Acoustics.

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20 **Abstract:**

21 When endangered southern resident killer whales enter the urban estuary known as the Salish
22 Sea, their characteristic sound signals (calls, whistles, and echolocation clicks) blend with the
23 noise from commercial ships bound for the ports of Vancouver, Seattle, and Tacoma. In such
24 environments where signal-to-noise ratios are low or noise mimics signals, human listeners
25 provide validation of automatic detections and increase overall detection rates. By giving both
26 citizen scientists and computers real-time access to underwater sounds, the Salish Sea
27 Hydrophone Network allows comparison of human and automated detection rates. In a case
28 study where killer whales passed a proposed tidal turbine site 22 times, they were detected
29 acoustically by the automated algorithm 14 times (64% detection rate), by human listeners 10
30 times (45%), and by either system 17 times (77%). During local daylight hours, human detection
31 rates rose to nearly match automated levels and had zero false positives.

32

33 ASA-PACS numbers:

34 4330Sf Acoustical detection of marine life; passive and active

35 4380Ev Acoustical measurement methods in biological systems and media

36 I. Crowdsourcing real-time data streams

37 Over the last decade the proliferation of computers and mobile devices has fueled a rapid
38 expansion of scientific data collection by citizen scientists (Delaney et al. 2007; Honicky et al.
39 2008; Cochran et al. 2009; Sullivan et al. 2009; Dickinson et al. 2010). Some projects
40 crowdsource analysis of archived data sets (Orchive; WhaleFM; CamClikr), while others
41 develop games in which humans are challenged to solve difficult computational problems
42 (Cooper et al. 2010; EteRNA) or annotate large multimedia databases (Barrington et al., 2012).

43

44 Analysis of data streams by citizen scientists *in real-time* has many potential applications and
45 benefits, but is comparatively rare. Furthermore the most popular live data sources rarely
46 provide opportunities for citizen science. For example, wildlife web cameras like Cornell
47 University's BirdCams provide live feeds that are very popular, but only offer observers a chat
48 room. Similarly, few of the growing number of live hydrophone streams (Table I) have offered
49 listeners ways to become citizen scientists. One exception is OrcaLab which offers an in-
50 browser chat room to discuss killer whale sounds heard via real-time audio streams (Orca Live)
51 and an on-line archive that citizen scientists are indexing (Orchive).

52

53 In marine bioacoustics and passive marine mammal monitoring in particular, autonomous or
54 cabled recording systems often accumulate large volumes of data. Software and humans can
55 process such archives in search of patterns, but depending on the nature of the signal and the
56 noise, it can be computationally-intensive and time-consuming (Mellinger et al., 2007; Munger
57 et al., 2005). Furthermore, the detection procedure is often so specific that unexpected signals
58 are missed.

59

60 We describe an alternative approach in which live audio streams from cabled hydrophones are
61 provided to both computers and human listeners for real-time monitoring and signal detection.

62 Because continuous recordings are not archived, this technique greatly reduces demand for data
63 storage and post-processing resources.

64

65 The main benefit of live streaming and real-time analysis is that presence or absence information
66 can immediately guide mitigation, research, or stewardship. Live streaming also enlivens
67 education and enhances marine environmental awareness by enabling students and stewards to
68 monitor current ocean conditions. It motivates citizen scientists to monitor the data stream
69 because they seem to enjoy being the first to discover something and knowing they are
70 witnessing a natural phenomena as it happens.

71 **II. Real-time detection of endangered killer whales by humans and computers**

72 Both human and automated detection of real-time ocean sounds are useful in the noisy inland
73 waters of Washington State and British Columbia known as the Salish Sea. Haro Strait between
74 Vancouver Island and the San Juan Islands is the core summertime habitat of the southern
75 resident killer whales (SRKWs) -- an iconic population that is listed as endangered on both sides
76 of the U.S.-Canada border. The SRKWs should be easy to detect acoustically because they emit
77 a wide range of distinct vocalizations, including pulsed calls, whistles, and echolocation clicks
78 (Ford, 1989) and are acoustically active except when resting. Passive acoustic monitoring for
79 SRKWs in the Salish Sea is complicated, however, because about 20 commercial ships transit
80 Haro Strait daily (V. Veirs et al., in prep.). Many of these ships emit intense tones (Mckenna,
81 2011) or clicks that are similar to SRKW signals. Within this soundscape it is difficult to
82 automatically detect killer whale vocalizations without ship noises generating an overwhelming
83 number of false positives.

84

85 Yet, reliable real-time detection of SRKWs is needed to enable rapid responses by researchers
86 who seek field data from SRKWs. To meet these needs and surmount the automated detection

87 challenge, we have provided real-time hydrophone audio streams to the public and asked citizen
88 scientists to listen for whales (and other interesting sounds).

89

90 Real-time acoustic detections by a combination of humans and computers supplement location
91 reports from sighting networks, particularly at night and during inclement weather. These new
92 detections also help define SRKW habitat use and nighttime acoustic behavior.

93

94 Real-time locations of SRKWs are also useful for mitigating potential impacts on the population.

95 In the event of an oil spill, the Northwest Area Contingency Plan will need up-to-date locations

96 to guide decisions. Tidal power turbines and pile driving projects could reduce the potential

97 risks they pose to SRKWs with knowledge of their real-time distribution. If ship noise proves to

98 be a significant risk to SRKWs, then whale locations could be used to reroute vessels or adjust

99 their speed – just as is done for North Atlantic right whales in Massachusetts Bay (Clark et al.,

100 2010).

101

102 Real-time notification of live listening opportunities enables novel educational and outreach

103 activities designed to facilitate the recovery and conservation of the SRKW population. The

104 established popularity of orcas can combine with the inherent excitement of listening live to the

105 ocean in exhibits at marine educational centers around the region.

106 **III. The Salish Sea Hydrophone Network**

107 The Salish Sea Hydrophone Network is a cooperative effort of scientists, educators, managers,

108 and citizens in the Pacific Northwest (Fig.1). The Network is an outgrowth of the SeaSound

109 project – an effort initiated by The Whale Museum in the 1970s to monitor underwater sound at

110 Lime Kiln [aka Whale Watch] State Park on the west side of San Juan Island. In the early 2000s

111 Val Veirs developed the Orcasound laboratory 5 km to the north and then led The Whale

112 Museum to revamp the Lime Kiln node. Since 2006 the Network has grown and been

113 maintained through funding from the National Oceanic and Atmospheric Administration and
114 additional financial support from Colorado College, Beam Reach, the Washington Department of
115 Fish and Wildlife, and individual Network members.

116

117 Today the network consists of cabled hydrophone systems deployed across the critical habitat of
118 SRKWs (Fig. 1) and a web site – orcasound.net – that offers public access to real-time and pre-
119 recorded audio streams. In addition to the original partners (The Whale Museum at Lime Kiln,
120 and Orcasound laboratory), regional collaborators that host hydrophone nodes now also include
121 the Port Townsend Marine Science Center, the Seattle Aquarium, and the Makah Tribe in Neah
122 Bay. This expansion has helped to highlight differences in anthropogenic noise levels around the
123 Salish Sea, patterns of SRKW critical habitat use, as well as furthering our abilities to
124 acoustically detect SRKWs for citizen science, research, and stewardship projects.

125

126 With broader access to the Internet the Network has become a global experiment in sharing and
127 processing real-time underwater sound. Development of outreach activities by the Network
128 partners, as well as educational and advocacy organizations like Orca Network and Killer Whale
129 Tales, has led to better detection, classification, and notification software systems, as well as
130 enhanced networking of listeners, researchers, activists, managers, and marine end users.

131

132 Taken together, this Network of hydrophones, computers, and humans has evolved into an
133 automated and crowdsourced acoustic detection system. Hydrophones sense underwater sound
134 and shoreside computers digitize the signal. Software and citizen scientists detect signals,
135 classify sounds, and notify Network members.

136 **A. Network hardware**

137 A typical Network node has at least two hydrophones installed on site for redundancy and to
138 enable stereo listening. We have tried many makes and models over the years but are currently

139 using a combination of hydrophones manufactured by International Transducer Corporation,
140 Cetacean Research Technology, Reson, and LabCore. Most of the nodes now use LabCore
141 hydrophones because of their relatively low cost (\$300), durability, ease of use, and
142 minimization of cable and flow noise. The LabCore40 hydrophones however have a narrower
143 and less-flat frequency response than other hydrophones we have used. In general Network
144 hydrophones have been chosen to be sensitive from at least 500 Hz to 15 kHz. All hydrophones
145 are located within 100 m of the shoreline and are deployed at depths of 5-15 m for ease of
146 maintenance.

147

148 At Lime Kiln and Orcasound where wave and tidal power are high, we mount hydrophones on
149 weighted stands and use intertidal cable protectors. At nodes where a dock structure is available
150 (the Seattle Aquarium, Port Townsend, and Neah Bay), the most successful deployment method
151 is simply dangling the hydrophone vertically into the sea between pilings. The analog signal is
152 digitized (48000 samples/sec, 16 bit) using a sound board in a Desktop with an uninterrupted
153 power supply, or more recently a netbook with batteries capable of enduring brief electrical
154 outages. An Internet connection with an upload bandwidth of at least 100 kbps supports the
155 audio stream and our other data transmission needs.

156 **B. Detection by computers**

157 WHO_Listener (Washington Hydrophone Observatory Listener) is a computer program that
158 monitors a real-time audio feed and triggers on ephemeral sounds. The program is written in C+
159 + using Microsoft's Visual Studio (2008) and consists of the following independently-running
160 threads (Fig. 2): data acquisition, spectral analysis, amplitude calculation, noise monitoring,
161 event triggering, event reporting, and Internet uploading.

162

163 In the data acquisition (DAQ) thread, mono or stereo audio streams from a analog-to-digital
164 sound card are sent to a ring buffer containing the most recent 10 seconds of audio data. The

165 purpose of the ring buffer is to maintain access to earlier data needed by the triggering and
166 reporting threads. Each time that a new datum is added to the ring, the oldest datum is
167 overwritten and the event is posted to other threads.

168 ***1. Spectral analysis***

169 The spectral analysis thread takes data from the ring buffer and computes power spectral
170 densities (Fourier transform size 2048, Hanning window, 10% overlap). From these spectra, the
171 thread calculates short- and long-term average spectra. The short-term average power spectrum
172 is calculated by adding 10% of the new value of each frequency component to 90% of the
173 previous average. For the long term average, the weightings are 1% and 99%, respectively.
174 These choices result in e-folding times of about 0.5 and 5 seconds respectively.

175 ***2. Amplitude calculation***

176 The amplitude thread also takes data from the ring buffer and calculates short- and long-term
177 averages of the RMS amplitude using the same weighting scheme as in the spectral analysis.
178 Calibration of the amplitude signals is applied here.

179 ***3. Noise monitoring***

180 Every 5 seconds the power spectrum is saved to maintain a running average of power statistics
181 (means and standard deviations). Every ½ hour, these statistics are written to an archive text file
182 on the node computer.

183 ***4. Event triggering***

184 WHO_Listener has two detectors designed to trigger on SRKW signals. The peak-tracking
185 detector (PKT) is an energy sum detector. It monitors the evolution of peaks in the power
186 spectrum and triggers when harmonic structure and duration are consistent with typical SRKW
187 calls. The power-excursion detector (PWR) is an amplitude detector. It monitors signal power
188 and triggers when the signal power briefly exceeds a threshold typical of nearby SRKW calls,
189 whistles, or clicks (or other ephemeral sounds).

190

191 The peak-tracking detection is accomplished by computing the scalar product of sequential third
192 Fourier transforms. The averaged spectral densities from the spectral thread are themselves
193 Fourier transformed yielding a function similar to the cepstrum without the cepstrum's
194 intermediate log transformation. Then a final Fourier transform is applied yielding a function of
195 frequency that is sensitive to killer whale calls. A scalar parameter for tracking such tonal sounds
196 is constructed by taking the scalar product of the lower-frequency portion this third Fourier
197 transform of the original time-series data with the immediately previous such transform (delayed
198 0.1 seconds). This parameter (PKT) rises and falls smoothly when receiving signals that have
199 continuous and relatively simple frequency structure similar to killer whale calls.

200

201 The parameter used for the power-excursion detector (PWR) is the logarithm of the ratio of the
202 short-term (signal) and long-term (background) integrated power spectra that are computed in the
203 spectral analysis thread.

204

205 Both triggers have amplitude and timing criteria; the trigger must remain between specified
206 amplitude thresholds continuously within a specified time interval before a detection is logged.
207 When either the spectral signature (PKT) or the power amplitude (PWR) rises above a starting
208 threshold, a flag is raised reporting that a trigger may be starting. From that moment on, a timer
209 assesses incoming spectra or average signal power. If the trigger signal remains above the
210 starting and ending Schmidt trigger thresholds within the specified time interval, then an event
211 has been detected. After tuning during real-time operation, we initiate the start of a trigger when
212 a trigger parameter rises above 120% of a specified threshold and terminate when the signal drops
213 below 80% of that threshold. The thresholds that we use are 0.7 for PKT and 10 for PWR and
214 the minimum and maximum duration of a trigger must be between 0.5 and 5 seconds. With these
215 parameter settings, killer whale calls are rarely missed (not detected). This triggering process is

216 thus not based on an absolute amplitude threshold and can detect quiet calls when background
217 noise levels are low.

218 ***5. Event reporting***

219 Whenever a trigger occurs, a detection is written to disk. This involves going back in time in the
220 ring buffer by a specified number of seconds prior to the start of the trigger (usually 2 seconds)
221 and then copying data from the ring buffer until some time (usually 5 seconds) after the event
222 trigger has terminated. A WAV file is generated with metadata encoded in the WAV file.
223 Metadata including node location, sampling rate, word size, number of channels, etc. are stored
224 in the header of a (16-bit, 44.1 kHz sampling rate) WAV file and date/time and dB levels are
225 stored in a constructed filename. These WAV files are stored in the monitoring computer at the
226 location of the hydrophone node where the trigger occurred.

227 ***6. Internet uploading***

228 If WAV files have been created recently (within the last 5 minutes), then the ~1 MB WAV files
229 are converted to ~100 kB MP3 files and a ~10 kB thumbnail image of the spectrogram of the
230 trigger is created. These compressed files are uploaded to orcasound.net where every 5 minutes
231 server-side scripts archive the detections, enter metadata for each in a MySQL database, and send
232 analysts an HTML email with the thumbnails embedded and linked to the archived detections.

233 **C. Detection by humans**

234 From each Network node, live hydrophone signals are broadcast as 48 kbps stereo MP3 streams
235 by the WinAmp Shoutcast plugin to a commercial streaming server. The audio stream is served
236 to as many as 20 simultaneous listeners per node who play the stream through free software like
237 iTunes and WinAmp. Demand for the Lime Kiln audio was high enough in the summer of 2011
238 to warrant increasing the maximum number of listeners to 30.

239

240 The streams are accessed either through links on the orcasound.net web site or through a playlist
241 maintained by the listener's MP3 stream player software. Web site traffic (measured with Google
242 Analytics) during the period of April 2007 through April 2012 averaged 30 unique visitors per
243 day, but has a strong seasonal pattern. Visitors per day rises from a mid-winter low of about 10
244 to a summertime average of ~100, with occasional spikes to 200-350. Listeners are
245 predominantly from the U.S. (75%), Canada (13%), Europe (8%), and Australia (0.5%).

246

247 The web site introduces citizen scientists to the Network by stating scientific and educational
248 goals and describing each node. It also offers links to a “sound tutor” page and ways to
249 contribute to the observational effort.

250

251 The sound tutor page presents recordings of common Salish Sea sounds, including the SRKW
252 calls most often used by each pod, and links to online libraries of SRKW calls. By comparing
253 the common calls and call libraries to “greatest hits” of the Network, listeners quickly learn to
254 ignore ships while monitoring the live data stream for signals of interest. In the first 6 months of
255 2012, there have been 487 page views of the sound tutor page. Further learning occurs through
256 occasional formal trainings or through iterative email exchanges between citizen scientists and
257 Network administrators.

258

259 Other links on the website enable citizen scientists to report detections. Beginning listeners can
260 do this via email (detection@orcasound.net) to Network administrators. Intermediate listeners
261 log their listening sessions along with others in a collaborative Google spreadsheet (that
262 automatically notifies administrators when new edits occur). Advanced listeners are invited to
263 join an unmoderated email distribution list (locate@orcasphe.net) through which they can
264 communicate their detections and often preliminary classifications with other citizen scientists.

265

266 Some listeners have installed (often free) stream "ripping" software that allows them to record
267 the live stream. The software typically creates a WAV or MP3 file on their personal computer.
268 In some cases, listeners archive or publish these recordings on their own web sites and/or blogs.
269 In others, they email them to the Network analysts who verify, convert if necessary, and then
270 upload them to the orcasound.net detection archive.

271

272 With advanced listeners, it is not uncommon for metadata to be reported along with detections,
273 including: classifications of recorded sounds as SRKW calls, clicks or whistles; acoustic
274 inference of pod identity; or deduction of direction of travel from multi-node detection
275 sequences. The end product of the human detection system is the same as that yielded by the
276 automated system: a SRKW signal is detected in near-real-time.

277 **D. Classification**

278 Whether detections are received from humans or node computers, the next step is for an analyst
279 to classify the detection(s) and notify the Network of SRKW presence or absence. The analyst
280 can review the detection in two ways: open the email to examine the most recent detections
281 (usually 1-2 per email); or review many detections by querying the database based on a
282 particular node of interest, time period, trigger type, and/or dB range. With the current database
283 query and classification web forms, triggered sounds can be visually and aurally reviewed.

284

285 Trained users can classify the detections into categories such as orca calls, ship noise, speedboat
286 noise, and other noises. Queries can be constructed by selecting from various menus to search
287 the database for sounds based on node, date, time, dB level, orca classification(s), ship
288 classification(s), boat classification(s), and other classification(s).

289

290 A database front end allows users to query the database after selecting the various checkboxes
291 and drop-down menu items. Advanced listeners can classify sounds by annotating database
292 sounds with predetermined classification categories or tags.

293

294 The analyst can scroll through the table of spectrograms, click the check boxes of sounds to be
295 classified, and then specify and submit the classification via the form at the top of the page. To
296 listen to the sound associated with a particular spectrogram, the analyst simply clicks on one of
297 the spectrogram thumbnails; this activates a pop-up Flash player that displays a spectrogram
298 which scrolls by in real-time as the audio plays.

299

300 Whether the review occurs via near-real-time email or a subsequent database query, the detection
301 of killer whales is ultimately classified by humans (listeners and analysts), not by software. With
302 the database of spectrograms of detections, the analyst can quickly assess a whole sequence of
303 recent detections or review all sounds received in a day in just a few minutes. The system also
304 allows the analyst to create an HTML summary of the detections and spectrograms.

305

306 In the future, we could improve the confidence of the detection by verifying the classification
307 with other analysts or computer-based classification software. This has not been implemented
308 yet, but an obvious next step is to initiate an automatic notification whenever a qualified analyst
309 uses the web-based form to classify one of the triggered recordings as a SRKW call.

310 **E. Notification**

311 The population of listeners and web site visitors, along with a growing cadre of researchers,
312 stewards, and educators have been helping test key elements of the real-time notification system.
313 Notification from the automated detection system comes in up to three forms, depending on the
314 urgency and complexity of the notice: a text message via a private Twitter feed
315 (<http://twitter.com/killerwhales>); an email via the `locate@` list; and/or a blog post.

316

317 The most detailed level of analysis is a blog post by an analyst on the orcasound.net/wp web site
318 that includes the detections and spectrograms. A link to the post or an abbreviated classification
319 is circulated to the locate email list, and if a real-time listening opportunity still exists, the
320 detection is also reported via the Twitter feed.

321

322 As of June, 2012, the [locate@ listserv](mailto:locate@listserv) has 107 members who receive locations in real-time
323 emails or daily summaries. List membership is doubling every two years. List members include
324 Navy Region Northwest (via a special email address that is monitored by the Region Watch
325 Commander, ROC Bangor to improve Navy awareness of killer whale locations) and SRKW
326 recovery managers at the Northwest Regional Office of NOAA/NMFS.

327

328 As of June, 2012, the Twitter feed (currently private) has 60 followers, a number that is also
329 doubling every 2 years. The feed has provided 446 updates since its inception on 5/27/07, 61 of
330 them in the 12 months prior to June, 2012).

331 **IV. Results**

332 Admiralty Inlet is the gateway from the Pacific Ocean into Puget Sound (Fig. 1). Its strong and
333 consistent tidal currents have made it of interest as a potential tidal power site by both the
334 Snohomish County Power Utility District (SnoPUD) and the U.S. Navy. In the fall, SRKWs
335 commonly begin transiting Admiralty Inlet in pursuit of chum salmon returning to Puget Sound
336 rivers. Transits continue a few times per month through the spring.

337

338 The potential impacts of tidal turbines on SRKWs during these transits motivated SnoPUD to
339 fund a field study of how SRKWs use of the Inlet. A study area off Admiralty Head was
340 monitored from land and sea from October, 2009, through April, 2010. The intensive sighting
341 and listening effort at the study area and photo identification of SRKWs confirming when they

342 were outside of the study area enabled us to quantify both human and automated detection rates
343 during this time period.

344

345 Admiralty Inlet is a challenging site to conduct acoustic detections due to current velocities and
346 high shipping traffic. All shipping headed to and from the ports of Seattle, Tacoma and Everett
347 pass through this body of water. Mean sound pressure levels (0.156 - 30 kHz) in this location
348 have been reported at 117 dB re 1 Pa, with levels exceeding 100 dB re 1 Pa 99% of the time
349 (Bassett, Thomson, & Polagye, 2010).

350

351 Of the 22 times that SRKW transited the study area they were detected acoustically via the
352 PTMSC hydrophone 14 times (64%) by the automated algorithm and 10 times (45%) by human
353 listeners and overall, combining both approaches SRKW were detected 17 times (77%).

354

355 Of the total transits, we know that 14 occurred during the local daylight hours when the majority
356 of listeners were likely to be awake. The rates of detection of daylight transits were higher: 79%
357 by the automated system, 71% by human listeners, and 93% combined.

358

359 The Admiralty Inlet field study included a component that required researchers to collect data on
360 SRKW transits from land and from a research vessel. The acoustic detections were an integral
361 part of notifying these crews in enough time that they could collect data. Of the 14 daytime
362 transits, rapid response crews observed whales during 10 transits (71%) thus demonstrating the
363 value of this system to fulfilling the boat-based research objectives.

364 **V. Discussion**

365 Given that we know killer whales can remain silent while resting for hours at a time and that
366 noisy ships can mask distant vocalizations, we consider our acoustic detection of about two
367 thirds of the SRKW transits through Admiralty Inlet a success. Although use of passive acoustic

368 monitoring (PAM) for SRKW detection will always have some limitations, we believe the
369 automated and human detection rates can still be improved through simple means: iterative
370 testing and redesign of the automated triggering algorithms and global growth and education of
371 the human listening network.

372

373 Passive acoustic monitoring with the cabled hydrophone system at the Port Townsend Marine
374 Science Center enabled detection of many transits which would otherwise have gone undetected.
375 PAM was critical in learning that SRKWs do move through Admiralty Inlet at night. If we
376 assume our detection rate was constant throughout the study, then our in-situ results suggest that
377 SRKW vocalize at levels that are detectable from the shoreline during about half of their
378 nighttime transits of Admiralty Inlet. The automated detectors caught half of the 8 nighttime
379 transits, while humans only caught one. This proves that acoustic monitoring is particularly
380 helpful when sighting networks are inoperable, even in a noisy urban estuary. It also indicates
381 that globalizing the human listening network (across a wider range of time zones) could
382 dramatically improve detection rates when it is nighttime for the SRKWs.

383

384 While a friendly competition between human and automated detection systems has promoted the
385 advancement of both approaches, the best detection results were obtained through a combination
386 of the systems. The automated system provides continuity when the human system is small or
387 distracted. The human system offers discovery of novel sounds, detection of signals in noise,
388 and nuanced classification – all of which are currently challenging computational problems.

389

390 Overall, human listeners have proven to be excellent detectors of any novel signal in our
391 predominantly monotonous marine soundscape. This may be due in part because human brains
392 are capable of pre-attentive auditory change perception (Risto Näätänen, 2000). This has been
393 demonstrated by hydrophone listeners who can play a live stream in the background for hours as
394 they work on other tasks (or even sleep!) and still notice the first faint calls of approaching killer

395 whales. To evoke higher-level auditory processing, an auditory stimulus must violate a rule (or
396 perceived pattern) established by preceding stimuli (R Nääätänen et al., 2007). We speculate that
397 by training listeners which marine sounds are commonplace – through online tutoring and
398 listening experience – their minds establish that shipping noise is the “rule.” Thereafter, signal
399 detection can be elicited by a killer whale call or other unusual stimulus in humans who aren’t
400 consciously paying attention.

401

402 An unexpected benefit of real-time detection by humans has been the role the public have played
403 in not only monitoring but also mitigating underwater noise pollution. Citizen scientists have
404 proven highly effective at detecting rare or novel signals in the Salish Sea soundscape. In some
405 cases, the signals have been due to biological sources (like male harbor seals), but in others
406 listeners detected potentially harmful anthropogenic sounds, including: mid-frequency active
407 sonar, ship-to-submarine communications, underwater detonations, pile driving, and
408 exceptionally loud ships. Especially in response to hearing the military noises, citizens were
409 surprisingly efficient at seeking out responsible authorities, asking that the noise abate, and
410 documenting the noise incidents publicly.

411

412 **A. The classification challenge**

413 The online database and web-based analysis tools have made it easier to review detections and
414 classify orca calls. However, classifying the numerous false positives generated by
415 WHO_Listener triggering on ship noise across the Network is still tedious. To enable the
416 automated detection system to provide reliable real-time notifications of SRKW
417 presence/absence we need to automate the classification of the detections.

418

419 Prior to automation, however, we intend to compare results from a genetic-tree classification
420 scheme (Veirs, 2009) with classifications generated by humans. One approach is to invite

421 untrained listeners to judge similarity and thereby build a classification tree (Rehn et al., 2010).
422 Another is to set up a direct competition between automated methods and human classification,
423 scored with a reference library of calls.

424

425 We have already begun to invite our most-skilled listeners to classify detections in the online
426 database. A product of this effort will be a population of false detections that can be used to test,
427 refine, and supplement the WHO_Listener triggers.

428

429 **B. Real-time citizen science within social networks**

430 In the last couple years we have witnessed a rapid rise in public discussion of Network
431 hydrophone signals on Facebook and other social networks. This growth has exceeded the use of
432 the communication tools we created for our listeners (the detection@ email address, the Google
433 spreadsheet, the locate@ email list, and the Twitter feed). While this may be due to
434 shortcomings of our tools, we believe it is primarily because of the increased membership of
435 social networks, in particular those that focus on topics related to killer whales. The Orca
436 Network Facebook page, for example, often has the richest available observational data
437 contained in wall posts made by the 13,000 users (as of June, 2012) who like the page.

438

439 Unfortunately, the data they generate are easily lost to science because they occur in de-
440 centralized threaded conversations that are time-consuming to monitor. It is also difficult to
441 maintain data standards within many social networks (e.g. time zone) and to archive or organize
442 the observations in ways that enable further analysis.

443

444 A new challenge we face is how to utilize such enthusiastic citizen scientists and the
445 observations they voluntarily make. Our citizen scientists need training to use of web-based
446 tools like the email list, Twitter feed, and forms for reporting or querying data. In contrast,
447 many potential citizen scientists are already familiar with how to post information to social

448 networks. This leaves us pondering whether to create bioacoustic research opportunities within
449 existing social networks (e.g. by developing apps) or to enable citizen scientists to use their
450 social network identity to login to our web site where they can use improved tools to participate
451 in killer whale research.

452

453 **ACKNOWLEDGEMENTS**

454 From 2007-2012, the Network was supported by NOAA Northwest Fisheries Science Center and
455 Northwest Regional Office and for this we thank B. Hanson and L. Barre. Additional financial
456 support has come from the Washington Department of Fish and Wildlife, Network node hosts,
457 and individual hydrophone network listeners. At Lime Kiln, we are grateful to the Washington
458 State Parks and U.S. Coast Guard for use of the lighthouse. The Admiralty Inlet study was
459 funded by the Snohomish Public Utility District and would not have been possible without the
460 efforts of our collaborators: Dom Tollit of SMRU, Ltd., S. Berta and H. Garrett of Orca Network,
461 C. McLean of the Port Townsend Marine Science Center, E. Peirson, and R. Slade. The Network
462 keeps running thanks to members of the “A(coustics)-Team” around the region: D. Howitt, D.
463 King, R. Kodner, and S. Bräger at Lime Kiln; C. McLean and E. McRae of the Port Townsend
464 Marine Science Center; B. Nelson, D. Larson, S. Harvey at the Seattle Aquarium; J. Scordino
465 and A. Akmajian in Neah Bay; our past students from Colorado College and Beam Reach; and L.
466 Veirs at Orcasound. Finally, we are grateful to all who listen enthusiastically for the whales,
467 especially J. Hyde and L. Brockelhurst. Figures were created using the Generic Mapping Tools
468 (*Wessel & Smith, 1998*) and Google Drawing.

469

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523 **TABLES**

524 **Table I caption:**

525 Live or near-real-time public hydrophone audio streams.

526 **Table I:**

Salish Sea Hydrophone Network	http://orcasound.net
Orca Live	http://orca-live.net
Victoria Experimental Network Under the Sea (VENUS)	http://venus.uvic.ca/data/live-audio/
Neptune Canada	http://www.neptunecanada.ca/news/sound-gallery/
Pacific Wild	http://www.pacificwild.org
Hawaiian humpback	http://jupiterfoundation.org
Whalesong humpback	http://www.whalesong.net/
PALAOA Antarctica	http://icecast.awi.de:8000/PALAOA.MP3
Listening to the Deep Ocean Environment	http://listentothedeep.com

527

528

529 **FIGURE CAPTIONS**

530 **Figure 1:**

531 Distribution of nodes (black dots) in the Salish Sea Hydrophone network. The Admiralty Inlet
532 study site is located adjacent to Port Townsend to the east-northeast.

533

534 **Figure 2:**

535 Data flow in the WHO Listener PAM program. Hydrophones signals from digital acquisition
536 (DAQ) are stored temporarily in a ring buffer. After averages are computed over two time
537 intervals, transformations are made to construct triggers for frequency dependent events (PKT)
538 and for amplitude dependent events (PWR). Detections are archived locally and uploaded to an
539 on-line database.

540