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Validating Citizen Science Monitoring of Ambient Water Quality for the United Nations Sustainable Development Goals

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Abstract

Citizen science (CS) may be described as research carried out by members of the public with the aim of gathering scientific information for the purpose of aiding in scientific projects. It has many potential advantages, including data collection at a scale not possible by professional scientists alone. The United Nations (UN) has recently recognized citizen science as a potential source of data that may contribute to the UN Sustainable Development Goals (SDGs). The availability of relatively inexpensive water quality monitoring field equipment suitable for CS suggests great potential for increased spatial coverage far beyond that of traditional, laboratory-based monitoring networks for water quality. In support of work towards the achievement of Sustainable Development Goal 6: “Clean Water and Sanitation”, this study tested the use of such field equipment by citizen scientists for SDG Indicator 6.3.2: “Proportion of bodies of water with good ambient water quality”. Data generated by 26 citizen scientists were compared with the results produced by an accredited laboratory. The results compared well for most parameters, suggesting that citizen science may be able to contribute towards monitoring ambient water quality for the Sustainable Development Goals.

Keywords: SDG 6; capacity development; volunteer monitoring; United Nations; community science

1. Introduction

34 SDG Indicator 6.3.2 is defined as the “proportion of bodies of water with good ambient water
35 quality” (UNEP, 2018). Together with SDG Indicator 6.3.1 on the “proportion of wastewater
36 safely treated”, these indicators provide a means of monitoring progress towards achieving
37 SDG Target 6.3 with the aim of improving global water quality. Due to the issues facing many
38 Member States regarding the collection of sufficient data on ambient water quality, the United
39 Nations has expressed significant interest in the potential for citizen science to contribute to
40 supporting progress towards achieving the ambient water quality SDG Indicator 6.3.2 (UNEP,
41 2018). The indicator methodology currently makes use of a water quality index that
42 summarizes data gathered through the analysis of basic core water quality parameter groups,
43 namely oxygen, salinity, nitrogen, phosphorus and acidification (UN Water, 2018). All
44 Member States are asked to monitor to this level and are required to report a national indicator
45 score designed to reflect overall water quality in that region (UNEP, 2018). As part of the
46 United Nation’s 2017 baseline data drive, submissions were received from 52 of the 193
47 Member States, comprising data of varying levels of coverage and completeness (UNEP,
48 2018). The data drive highlighted that some Member States were prevented from reporting on
49 the ambient water quality indicator for SDG 6 due to insufficient monitoring activities, and that
50 other States with limited resources focused on monitoring a few key water bodies (UNEP,
51 2018).

52 Citizen science refers to the participation of citizens in scientific projects with the
53 objective of gathering scientific information (Bonney *et al.*, 2014; Silvertown, 2009). The
54 practice employs the joint efforts of both professional scientists and members of the public,
55 who need not hold any preliminary knowledge or training on the subject matter, but who
56 volunteer to collaborate with professionals to conduct scientific research (Cappa *et al.*, 2018;
57 Dickinson & Bonney, 2012). Although citizen science traces its roots back to the beginnings
58 of modern science (Cohn, 2008), scientific research involving volunteers has seen a surge in
59 popularity in recent years (McKinley *et al.*, 2017). The United Nations has recognized citizen
60 science as potentially being a necessary source of support for the monitoring of ambient water
61 quality for SDG 6 (UNEP, 2018). Greater effort is therefore needed in order to encourage the
62 use of this cost-effective and abundant resource. The five core water quality parameter groups
63 of the ambient water quality SDG Indicator 6.3.2 (oxygen, salinity, nitrogen, phosphorus and
64 acidification) may be measured using a range of simple and inexpensive field techniques that
65 are accessible to citizen science networks (UNEP, 2018). Thus, where the proper resources are
66 put in place to ensure responsible data collection and submission, citizen science networks

67 could prove a vital source of additional data on ambient water quality by providing greater
68 spatial and temporal coverage of data than is currently possible through the sole use of
69 traditional, laboratory-based monitoring networks (UNEP, 2018).

70 A number of challenges remain before citizen science can be seen as a viable method
71 of scientific research producing reliable data that can be used to support scientific and decision-
72 making processes across a diversity of fields, including those relating to the monitoring of
73 ambient water quality for the Sustainable Development Goals. The most significant barrier to
74 the widespread use of citizen science is the perception of scientists who question the quality
75 and reliability of data produced by non-professionals (Burgess *et al.*, 2017; Fore *et al.*, 2001;
76 Penrose & Call, 1995; Riesch & Potter, 2013). Data quality issues are not isolated to citizen
77 science monitoring programmes – experienced researchers also make errors. However, the
78 perception that volunteer-generated data would not be well received by the scientific
79 community contributes to a prejudice against its use (Crall *et al.*, 2011; Dickinson *et al.*, 2010;
80 Foster-Smith & Evans, 2003; Riesch & Potter, 2013). In contrast, numerous studies have shown
81 that volunteers are capable of collecting data of equal quality to that of professional scientists,
82 provided they are given the proper training and resources, and provided the study design
83 matches the collectors’ abilities, and many validation studies to date have reported the high
84 standard of water quality data collected by citizen scientists (Dyer *et al.* 2014; Herman-Mercer
85 *et al.*, 2018; Levesque *et al.*, 2017; Loiselle *et al.*, 2016; Loperfido *et al.*, 2010; McGoff *et al.*,
86 2017; Muenich *et al.*, 2016; Safford & Peters, 2017; Scott & Frost, 2017; Shelton, 2013;
87 Thornhill *et al.*, 2017; Thornhill *et al.*, 2018; Wilderman & Monismith, 2016). Water quality
88 and water resource management within EU Member States is governed by the Water
89 Framework Directive (WFD), a piece of European Commission legislation, that requires the
90 incorporation of public participation in its implementation, mainly through public consultation
91 and information supply (Hadj-Hammou *et al.*, 2017; Van der Heijden & Ten Heuvelhof, 2013).
92 As with the methodology for the ambient water quality indicator for SDG 6, Member States
93 within the EU have the freedom to develop their own strategies for the monitoring and
94 assessment of waterbodies (Van der Heijden & Ten Heuvelhof, 2013). While public input has
95 been encouraged with regard to both the WFD and ambient water quality SDG Indicator 6.3.2
96 (UNEP, 2018; Van der Heijden & Ten Heuvelhof, 2013), the specific role of citizen science in
97 monitoring and assessing water quality is limited, and no study to date has explored the
98 potential for citizen science to support ambient water quality monitoring as part of the SDGs
99 specifically.

100 This study explored whether a group of citizen scientists based in Killarney, Co. Kerry,
101 Ireland, were capable of collecting high-quality data on a number of the core and alternative
102 ambient water quality parameters associated with SDG Indicator 6.3.2. The citizen scientists
103 conducted analyses on water samples using simple citizen science field kits provided by
104 FreshWater Watch (<https://freshwaterwatch.thewaterhub.org/>), the freshwater initiative of the
105 global NGO, Earthwatch (<https://earthwatch.org/>). The overall accuracy of the citizen science
106 field kits was evaluated by comparison with an ISO/IEC 17025:2017 accredited laboratory in
107 Co. Kerry, Ireland. The feasibility of citizen science to support monitoring of ambient water
108 quality parameters for the SDGs was assessed. The challenges and opportunities encountered
109 with applying this scientific approach to monitoring for the ambient water quality SDG
110 Indicator 6.3.2 are discussed here.

111

112 **2. Methods**

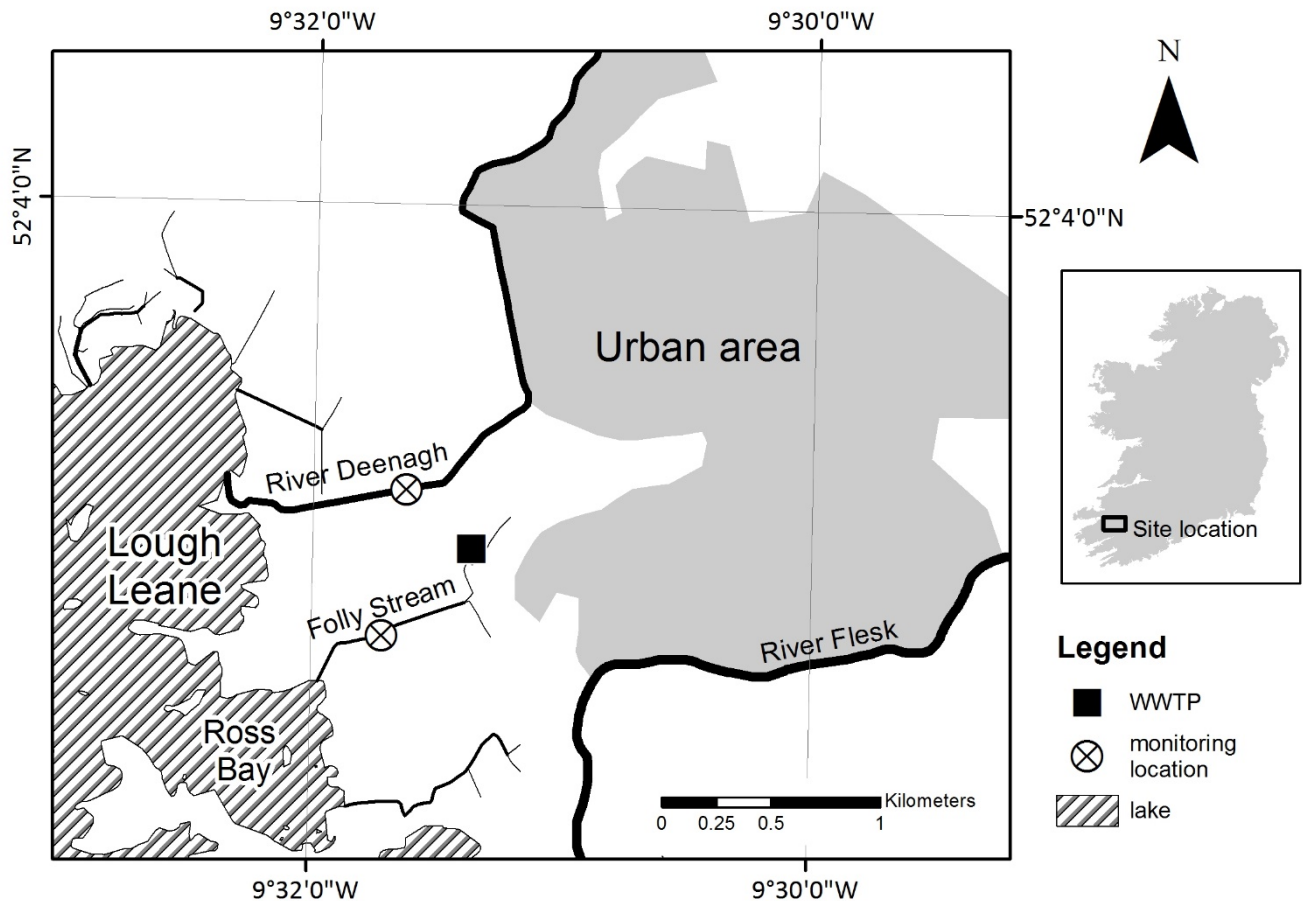
113 *2.1 Participant Recruitment*

114 Participants were recruited from St. Brendan's College, Killarney, Co. Kerry, Ireland, from a
115 class of 74 male students, between the ages of 16 and 17. Each student was given a screening
116 survey to assess their interest in science, environmental issues and working outdoors. A total
117 of 34 students were identified as potential participants for the project, based on the level of
118 interest shown by their responses to the screening survey. They then took part in a briefing
119 session and underwent training. The level of training among citizen scientists can influence the
120 accuracy of monitoring data (Fore *et al.*, 2001), therefore training was provided to all potential
121 participants. During the training session, students were taught about water quality issues within
122 freshwater ecosystems and the background to the research project, namely the UN Sustainable
123 Development Goals and the potential for citizen science to contribute to supporting SDG 6.
124 FreshWater Watch training materials provided the baseline for training of all participants, and
125 this was supplemented with a demonstration of the analysis techniques using water samples
126 provided for the purpose of training. Having been split into small groups, the students were
127 allowed time to practice using the analytical kits within the classroom under the supervision of
128 the trainer, who was able to provide feedback and answer questions. Following this practical
129 training session, all students were required to complete a training quiz, to confirm that the
130 participants were sufficiently trained and that their results could be trusted for uploading to the
131 FreshWater Watch global database (<https://freshwaterwatch.thewaterhub.org/content/data->

132 [map](#)). Based on the results of the training quiz, 28 students were selected to participate in the
133 research study.

134

135 2.2 Site Description



136

137 Figure 1. Locations of the monitoring sites within the River Deenagh and Folly Stream catchments in southwest
138 Ireland.

139 Lough Leane is a freshwater lake located within Killarney National Park, draining a catchment
140 of 553 km² near the town of Killarney, County Kerry in southwest Ireland. The rivers Flesk,
141 Deenagh and Long Range are the main sources of input to Lough Leane, which flows to the
142 Atlantic Ocean via the River Laune (Jennings *et al.*, 2013). The Folly stream is a minor stream
143 of approximately 1.5 km in length that drains a small area of roughly 0.9 km² and enters Lough
144 Leane near Ross Bay. The main wastewater treatment plant for the town of Killarney is located
145 1km upstream of Ross Bay. Two Storm Water Overflows (SWOs) carrying untreated

146 wastewater enter the Folly stream during times when the WWTP is under stress from high-
147 inputs (Irish Water, 2018).

148 The River Deenagh and Folly stream were identified as suitable for inclusion in this
149 study due to the evident differences in water quality between the two waterbodies. Monitoring
150 at the Folly stream has indicated that good status surface water standards for ammonia and
151 biochemical oxygen demand (BOD) are exceeded both upstream and downstream of the
152 wastewater treatment plant. Good status standard for orthophosphate is also exceeded
153 downstream of the plant (Environmental Protection Agency, 2012). It was acknowledged in
154 the last waste water discharge license application that the Folly stream was unable to
155 accommodate the discharge from the WWTP, despite the fact that it operated well within its
156 design parameters and capacity (Environmental Protection Agency, 2012). The Folly stream
157 has appeared as a cause of local concern in recent years due to the deteriorating water quality,
158 though it is currently not monitored by the EPA and is not assigned a status under the Water
159 Framework Directive (Environmental Protection Agency, 2012). Conversely, a number of EPA
160 monitoring stations are located along the length of the River Deenagh, with the most recent
161 assessment determining that the two lower stations located near Killarney town achieved
162 “Good” ecological status (Environmental Protection Agency, 2019). The differences in water
163 quality between the two waterbodies allowed for an examination of the effectiveness of the
164 FreshWater Watch equipment in more and less polluted environments.

165 A preliminary survey was carried out on 24th February 2019 and two sampling sites
166 were carefully selected based on accessibility and safety, one located on the River Deenagh
167 (52° 3’ 17” N, -9° 31’ 38” W) and another along the Folly stream (52° 2’ 56” N, -9° 31’ 44”
168 W) (Figure 1). On the day of sampling conditions at both sites were calm with a steady water
169 flow and average water levels. The sampling site at the River Deenagh was located upstream
170 of a bridge and featured clear water and a rocky bottom with bank vegetation on one side of
171 the river and a small pedestrian path on the other. The surrounding and overhead vegetation
172 consisted of deciduous forest. The sampling site along the Folly stream featured murky water
173 and a muddy bottom, with thick bank vegetation and a surrounding deciduous woodland.

174

175 *2.3 SDG Indicator 6.3.2 Parameters*

176 The five core water quality parameter groups for the ambient water quality SDG Indicator 6.3.2
177 are outlined in Table 1. Some parameters are included in the methodology in order to

178 characterize the water quality in a particular waterbody, while others provide a direct measure
 179 of water quality for ecosystem or human health (UN Water, 2018). Deviation from normal
 180 ranges (such as with salinity and acidification) and comparison of measured values with target
 181 values (in the case of phosphorus, nitrogen and oxygen) allow for the detection of instances
 182 where the waterbody may be experiencing harmful impacts. This enables the classification of
 183 water quality as either “good” or “not good” in relation to these target values for each
 184 monitoring location. The classifications are aggregated by catchment, and then nationally, to
 185 generate the indicator percentage (UN Water, 2018).

186 The water quality data which feed into the indicator are derived from in-situ
 187 measurements and analysis of water samples. The citizen science field kits provided by
 188 FreshWater Watch (FWW) were capable of measuring four of the recommended ambient water
 189 quality parameters: Orthophosphate, Nitrate, Electrical Conductivity and pH. The field kits did
 190 not include tests for the other recommended parameter, dissolved oxygen (DO), so Chemical
 191 oxygen demand (COD) was included here.

192

193 Table 1. Recommended monitoring parameters (in bold) required for the water quality index used for SDG
 194 Indicator 6.3.2 for three water body types. Alternative parameters (in italics) may be substituted for the
 195 recommended parameters, depending on data availability and applicability for specific water body types (UN
 196 Water, 2018).

Parameter group	Parameter	River	Lake	Groundwater
Oxygen	Dissolved oxygen <i>Biological oxygen demand, Chemical oxygen demand</i>	x	x	
Salinity	Electrical conductivity <i>Salinity, Total dissolved solids</i>	x	x	x
Nitrogen*	Total oxidised nitrogen <i>Total nitrogen, Nitrite, Ammoniacal nitrogen</i>	x	x	
	Nitrate**			x
Phosphorus	Orthophosphate <i>Total phosphorus</i>	x	x	
Acidification	pH	x	x	x
* Countries should include the fractions of N and P which are most relevant in the national context				
** Nitrate is suggested for groundwater due to associated human health risks				

197

198

199 **2.4 Citizen Analyses**

200 Sampling took place on 22nd March 2019 as part of an activity for World Water Day. At each
201 sampling site a large plastic bucket was first rinsed three times in the water from the sampling
202 site. Taking care not to disturb the sediment, the bucket was then filled from the centre of the
203 waterbody and placed in a secure location on the bank, where the sample water was mixed well
204 with a clean plastic spatula. All sampling by citizen scientists was conducted using the sample
205 water contained in the bucket, therefore minimizing any spatial and temporal differences
206 between results. The samples taken for analysis at an accredited laboratory were also taken
207 from the same sample of water in the same bucket. The citizen scientists wore gloves while
208 sampling and a large sheet of plastic tarp was placed on the ground where volunteers could
209 place equipment in order to avoid contamination of the water sample and materials used.

210 Nitrate (NO₃-N), phosphate (PO₄-P) and chemical oxygen demand (COD) Kyoritsu
211 PackTest (Kyoritsu Chemical-Check Lab, Corp., Tokyo, Japan) water chemistry kits were
212 obtained from FreshWater Watch (Earthwatch Institute, Oxford, United Kingdom). All
213 parameters were measured in transparent plastic tubes which are designed to mix a small water
214 sample with reagents that produce increasing colour values with increasing concentration
215 (Scott & Frost, 2017). The PO₄-P method using 4-aminoantipyrine with phosphatase enzyme
216 (Berti *et al.*, 1988), and nitrate NO₃-N method using zinc and subsequently following the Greiss
217 method (Nelson *et al.*, 1954), provided nutrient concentrations that fell into one of seven
218 categories ranging from <0.02 - >1.0 mg/L P and <0.2 - >10 mg/L N (Table 2) (Scott & Frost,
219 2017). Chemical oxygen demand was determined by an oxidation reaction with potassium
220 permanganate in an alkaline medium, which provided concentrations ranging across seven
221 categories from 0-5 to >100 mg/L O₂ (Table 2) (Kyoritsu, n.d.). pH was determined with
222 Simplex Health (Simplex Health, Wollaston, United Kingdom) pH test strips which were held
223 in the sample water for 3 seconds and subsequently matched to a colour chart. Electrical
224 conductivity was measured using hand-held Lohand Biological (Hangzhou Lohand Biological
225 Co., Ltd, China) conductivity meters dipped into the sample water for approximately 15
226 seconds until the reading in µS/cm stabilized (Table 2). Each participant received a copy of the
227 instructions on how to conduct each test and recorded all their data on their own individual
228 datasheet, covering both sites. Replicate samples were taken by citizens at each site – fourteen
229 students sampled each parameter twice in Site 1 and three times in Site 2, while the other half
230 of the participants did the opposite, thus taking a total of five measurements for each parameter
231 across the two sites.

232 A total of 27 datasheets were received following sampling and one was rejected because
 233 it was incorrectly completed. Data analysis was conducted on the results collected by 26
 234 participants in the study, resulting in a total of 66 measurements for most parameters at Site 1
 235 and 64 measurements for each parameter at Site 2 (Table 5).

236

237 Table 2. Ranges of measurement of the equipment used by citizen scientists to analyse various water quality
 238 parameters at the River Deenagh and Folly stream.

Parameter	Units	FWW Equipment Range								
Orthophosphate	mg/L P	<0.02	0.02-0.05	0.05-0.1	0.1-0.2	0.2-0.5	0.5-1.0	>1.0		
Nitrate	mg/L N	<0.2	0.2-0.5	0.5-1.0	1.0-2.0	2.0-5.0	5.0-10.0	>10.0		
Chemical Oxygen Demand	mg/L O ₂	0.0-5.0	5.0-10.0	10.0-13.0	13.0-20.0	20.0-50.0	50.0-100.0	>100.0		
pH	pH Unit	< 4.5	4.5 – 5	5 – 5.5	5.5 – 5.75	Increments of 0.25 up to 7.5	7.5 - 8	8 – 8.5	8.5 - 9	> 9
Electrical Conductivity	µS/cm	10 - 1990 +/- 10 µS/cm precision								

239

240

241 2.5 Laboratory Analyses

242 At each site three samples were taken from the bucket of sample water and transported to the
 243 Southern Scientific Services laboratory at Farranfore, Co. Kerry within 20 minutes of
 244 collection for preservation and analysis. The laboratory holds ISO/IEC 17025:2017
 245 accreditation for general requirements for the competence of testing and calibration
 246 laboratories (Southern Scientific Services, 2019). All methods used for the analysis of the
 247 various parameters are listed in Table 3. Orthophosphate and Nitrate were determined by
 248 spectrophotometry; pH and electrical conductivity were analysed using Rohasys MINILAB
 249 Multi Parameter robot (ROHASYS BV, Rijen, Netherlands); chemical oxygen demand was
 250 determined using a closed-reflux, colorimetric method (Table 3).

251

252 Table 3. Laboratory methods from Standard Methods for the Examination of Water and Wastewater 23rd Edition
 253 (Baird *et al.*, 2017) used in the analysis of water samples as part of this study by the accredited laboratory.

Parameter	Standard Reference/SOP	Range of Measurement	Accuracy of Measurement	Equipment/Technique
Orthophosphate	APHA, 4500P-E, 23Ed., (2017) / SPC 027c	0.01-12 mg/L P	+/- 0.001	Spectrophotometry by Aquakem 250 Autoanalyser
Nitrate	APHA, 4500NO3-E, 23Ed., (2017) /SPC 027g	0.25-45 mg/L N	+/- 0.001	
Chemical Oxygen Demand	APHA, 5520D, 23Ed., (2017) / SPC 016	10-30,000 mg/L	+/- 0	HACH/Colorimetric
pH	APHA, 4500B-H+, 23Ed., (2017) / SPC 052	4 - 10 pH Units	+/- 0.01	Rohasys Minilab
Electrical Conductivity	APHA, 2510B, 23Ed., (2017) / SCP 052	14.7 -111,900 μ S/cm @ 20°C	+/- 0.1	

254

255

256 *2.6 Data Analyses and Considerations*

257 The test kits provided by FreshWater Watch produced a categorical classification for the
258 concentration of various water quality parameters within a sample of water. The categories for
259 each parameter are outlined in Table 2. The outcomes of citizen scientist sampling are
260 displayed in a frequency distribution table – the most frequently chosen concentration range,
261 as well as the range containing the “true” laboratory value, are shown (Table 5). As the data is
262 categorical, the concentration range containing the laboratory value could be considered the
263 “correct” result, while results in all other categories could be considered incorrect. However
264 due to the nature of the testing kits and the colorimetric method by which a value is determined,
265 difficulty can arise for users when deciding between concentration ranges, as there is no
266 distinctive colour difference between one concentration range and the next. When the “true”
267 laboratory value falls close to the border of one of the concentration ranges it is understandable
268 for citizen scientists to struggle with choosing the correct result. For this reason, results
269 recorded one concentration range outside the “correct” concentration range are included in the
270 discussion on percentage agreement and the accuracy of citizen science monitoring of ambient
271 water quality. Opinion is also divided on an adequate level of percentage agreement in research.
272 To one researcher 70% agreement is adequate, whereas another would not consider 70%
273 agreement a sufficient level to answer their research questions (Aceves-Bueno *et al.*, 2017). A
274 general rule of thumb describes an agreement level of 75% as a minimum acceptable level of
275 agreement (Graham *et al.*, 2012; Hartmann, 1977; Stemler, 2004). This was the acceptance
276 level adopted by this investigation.

277

278 **3. Results**

279 3.2 Water Quality Testing

280 Table 4 shows the results of water quality analyses conducted by an accredited laboratory in
281 Kerry on samples taken from the River Deenagh (Site 1) and Folly stream (Site 2). Results of
282 analyses of the same water quality parameters by citizen scientists are displayed in Table 5,
283 and the percentage of their results in agreement with those obtained by the laboratory are
284 highlighted in bold (Table 5). Of the five ambient water quality parameters analysed, citizen
285 scientists demonstrated good agreement in their measurements of three – Orthophosphate,
286 Nitrate and Electrical Conductivity. The other two parameters, pH and Chemical Oxygen
287 Demand, showed less agreement with the laboratory results (Table 5).

288 Across both sites the majority of volunteer results for Orthophosphate were either in
289 agreement with the laboratory value or else fell into a concentration range just above or below
290 this (Table 5a). A similar result can be seen for Nitrate where between 81.3-84.8% of results
291 across both sites fell within or just outside the concentration range corresponding to the
292 laboratory value for Nitrate (Table 5b). However, greater variation can be seen in the
293 distribution of results outside this concentration range (Table 5b). The results of electrical
294 conductivity tests by citizen scientists at the River Deenagh were also positive, with 77.4% of
295 results falling within or just outside the laboratory value of 180 $\mu\text{S}/\text{cm}$. At the Folly stream the
296 results showed less agreement, with many citizen scientists overestimating the conductivity
297 value at that site (Table 5e).

298 The results of Chemical Oxygen Demand tests were less compatible with the laboratory
299 results; citizen scientists showed poor agreement of COD values in both the River Deenagh
300 (0.0%) and Folly stream (2.6%) (Table 5c). The percentage of citizen scientist results recorded
301 within or just outside the laboratory result was lower at 28.8% and 11.0% for sites 1 and 2
302 respectively. Citizen scientists were unable to measure pH accurately to within or just outside
303 the concentration range agreeable with the laboratory result in either the River Deenagh (0.0%)
304 or Folly stream (21.9%) (Table 5d).

305 The contrasting nature of the River Deenagh and Folly Stream is reflected in the results
306 obtained by both citizen scientists and the accredited laboratory. Though Nitrate and pH levels
307 did not appear to differ much between the two sites, Orthophosphate, Chemical Oxygen
308 Demand and Electrical Conductivity levels were noticeably higher at the Folly Stream than in
309 the River Deenagh (Tables 4 and 5). Irrespective of the levels of agreement between citizen
310 and laboratory results, the volunteers and FWW testing kits were capable of revealing a

311 difference in water quality between the two sites that supports current conclusions on the nature
 312 of these waterbodies.

313

314 Table 4. Results of analyses of water samples taken from the River Deenagh (Site 1) and Folly stream (Site 2) by
 315 an ISO/IEC 17025:2017 accredited laboratory. The means of the three laboratory analyses was calculated for each
 316 parameter and used for comparison with results gathered by citizen scientists.

Parameter	Units	Site 1				Site 2			
		Sample 1	Sample 2	Sample 3	Mean	Sample 1	Sample 2	Sample 3	Mean
Orthophosphate	<i>mg/L P</i>	0.02	0.01	0.02	0.02	0.10	0.10	0.10	0.10
Nitrate	<i>mg/L NO3-N</i>	2.4	2.5	2.6	2.5	2.5	2.4	2.4	2.4
Chemical Oxygen Demand	<i>mg/L O2</i>	<10	11	10	11	15	14	17	15
pH	<i>pH Unit</i>	7.5	7.5	7.5	7.5	7.2	7.1	7.1	7.1
Electrical Conductivity	<i>µS/cm @ 20°C</i>	180	179	180	180	427	434	432	431

317

318

319 Table 5. Results of citizen scientist water quality sampling at the River Deenagh (Site 1) and Folly stream (Site
 320 2) using the FreshWater Watch water quality testing kits. The number and percentage of results obtained by citizen
 321 scientists within each concentration range are shown. The citizen scientist results in agreement with the results
 322 obtained for each parameter by an accredited laboratory are highlighted in bold.

a) Orthophosphate		b) Nitrate		c) Chemical Oxygen Demand	
Range (mg/L P)	Site 1 Results	Site 2 Results	Range (mg/L N)	Site 1 Results	Site 2 Results
<0.02	29 (43.9%)	0 (0.0%)	<0.2	0 (0.0%)	0 (0.0%)
0.02-0.05	35 (53.0%)	13 (20.3%)	0.2-0.5	1 (1.5%)	6 (9.4%)
0.05-0.1	2 (3.0%)	27 (42.2%)	0.5-1.0	9 (13.6%)	5 (7.8%)
0.1-0.2	0 (0.0%)	23 (35.9%)	1.0-2.0	12 (18.2%)	4 (6.3%)
0.2-0.5	0 (0.0%)	1 (1.6%)	2.0-5.0	42 (63.6%)	31 (48.4%)
0.5-1.0	0 (0.0%)	0 (0.0%)	0.5-10.0	2 (3.0%)	17 (26.6%)
>1.0	0 (0.0%)	0 (0.0%)	>10.0	0 (0.0%)	1 (1.6%)
Total	66 (100.0%)	64 (100.0%)	Total	66 (100.0%)	64 (100.0%)
				Range (mg/L O2)	
				0.0-5.0	46 (69.7%)
				5.0-10.0	19 (28.8%)
				10.0-13.0	0 (0.0%)
				13.0-20.0	0 (0.0%)
				20.0-50.0	1 (1.5%)
				50.0-100.0	0 (0.0%)
				>100.0	0 (0.0%)
				Total	66 (100.0%)
					22 (34.4%)
					35 (54.7%)
					4 (6.3%)
					2 (3.1%)
					1 (1.6%)
					0 (0.0%)
					0 (0.0%)
					64 (100.0%)

d) pH		Site 1 Results	Site 2 Results
<4.5	0 (0.0%)	0 (0.0%)	0 (0.0%)
4.5 - 5	0 (0.0%)	0 (0.0%)	0 (0.0%)
5 - 5.5	10 (15.2%)	0 (0.0%)	0 (0.0%)
5.5 - 5.75	45 (68.2%)	2 (3.1%)	2 (3.1%)
5.75 - 6	8 (12.1%)	14 (21.9%)	14 (21.9%)
6 - 6.25	2 (3.0%)	26 (40.6%)	26 (40.6%)
6.25 - 6.5	1 (1.5%)	4 (6.3%)	4 (6.3%)
6.5 - 6.75	0 (0.0%)	2 (3.1%)	2 (3.1%)
6.75 - 7	0 (0.0%)	1 (1.6%)	1 (1.6%)
7 - 7.25	0 (0.0%)	0 (0.0%)	6 (9.4%)
7.25 - 7.5	0 (0.0%)	7 (10.9%)	7 (10.9%)
7.5 - 8	0 (0.0%)	2 (3.1%)	2 (3.1%)
8 - 8.5	0 (0.0%)	0 (0.0%)	0 (0.0%)
8.5 - 9	0 (0.0%)	0 (0.0%)	0 (0.0%)
>9	0 (0.0%)	0 (0.0%)	0 (0.0%)
Total	66 (100.0%)	64 (100.0%)	64 (100.0%)

e) Electrical Conductivity		Results Site 1	Results Site 2
Range (µS/cm)	Range (µS/cm)	Results Site 1	Results Site 2
110	410	1 (1.6%)	1 (1.6%)
130	420	4 (6.5%)	8 (12.9%)
150	430	3 (4.8%)	10 (16.1%)
160	440	6 (9.7%)	11 (17.7%)
170	450	15 (24.2%)	20 (32.3%)
180	460	30 (48.4%)	9 (14.5%)
190	470	3 (4.8%)	1 (1.6%)
Total	480	62 (100.0%)	2 (3.2%)
	Total		62 (100.0%)

325 **4. Discussion**

326 *4.1 Can citizen science help support monitoring for SDG Indicator 6.3.2?*

327
328 Overall the results of the water quality analyses indicated that citizen scientists were able to
329 measure water quality parameters to within or just outside the laboratory value for between
330 79.7% and 99.9% of measurements for Orthophosphate and Nitrate, establishing them as two
331 of the parameters most compatible with the laboratory results (Table 5a-b). Electrical
332 conductivity measurements were a little more variable, with between 46.7% and 82.3% of
333 results falling within or just outside the laboratory value (Table 5e). Chemical oxygen demand
334 and pH were the parameters showing the least agreement with the laboratory results (Table 5c-
335 d). Concentration ranges just outside the concentration range containing the laboratory result
336 were taken into account when discussing percentage agreement and the overall accuracy of
337 results. While this was deemed necessary to account for the difficulty volunteers experienced
338 in choosing between concentration ranges due to the colorimetric nature of the testing kit, it
339 must be recognized that this method likely overestimates the percentage agreement due to the
340 inclusion of results at the extreme, opposite ends of the outer concentration ranges which were
341 not in any way misinterpreted.

342 The five water quality parameters chosen for inclusion in this research study form the
343 basis of the most basic monitoring level for ambient water quality under SDG Indicator 6.3.2,
344 the ambient water quality indicator for SDG 6 (UNEP, 2018). Results of citizen testing of
345 Orthophosphate, Nitrate and Electrical Conductivity proved reasonably accurate based on the
346 percentages of results in agreement with laboratory analyses for these parameters (Table 5a-b
347 & 5e). This was partly expected for both nutrient tests given the positive conclusions drawn by
348 other researchers who have used the Kyoritsu PackTest water chemistry kits provided through
349 FreshWater Watch to allow citizen scientists to measure Orthophosphate and Nitrate (Levesque
350 *et al.*, 2017; Loiselle *et al.*, 2016; McGoff *et al.*, 2017; Scott & Frost, 2017; Shupe, 2017;
351 Thornhill *et al.*, 2017; Thornhill *et al.*, 2018; Xu *et al.*, 2017). Two of these studies (Levesque
352 *et al.*, 2017; Thornhill *et al.*, 2017) noted that between 65.8% and 81% of results obtained by
353 citizen scientists for both parameters were in agreement with laboratory results, a slightly
354 higher level of agreement than was noted in this investigation. Interest level has been identified
355 as an important motivational variable in a student's academic performance and an influencing
356 factor in how much attention is paid to a particular activity (Hidi & Harackiewicz, 2000;
357 Schiefele, 1991, 1996). It is therefore possible that the slightly lower level of agreement with

358 laboratory results witnessed in this study compared to others involving FreshWater Watch
359 volunteers could be attributed to lower interest levels on the parts of the students, compared to
360 those of volunteers giving time out of their everyday schedule. An investigation into whether
361 differences in interest levels influence the accuracy of results obtained using the kits may prove
362 beneficial for recruitment purposes for future citizen science projects. Other published research
363 studies focusing on testing water quality using citizen scientists have opted for the use of total
364 reactive phosphorus (Hach Aquacheck Cat. 27571-50) and nitrate field test strips (HACH,
365 2745425; Hach Aquacheck Cat. 27454-25) (Loperfido *et al.*, 2010; Muenich *et al.*, 2016) and
366 observed mixed results. No other published studies could be found on citizen science water
367 quality testing involving the use of the Lohand Biological meters for conductivity. The
368 performance of the meters in the field and their agreement with the laboratory results was very
369 good at the River Deenagh (Table 5e), though they did not perform as well at Folly stream,
370 potentially indicating that they are less reliable in more polluted environments. Other published
371 studies have made use of YSI Professional Plus multi-probes (Shelton, 2013), EuTech
372 ECTestr™ 11 probes (Storey *et al.*, 2016), Oakton PCtestr meters (Shupe, 2017), and the
373 LaMotte PockeTester meter (Wilderman & Monismith, 2016) for measuring electrical
374 conductivity and have reached mostly positive conclusions on their use. However, while also
375 useful, these instruments are considerably more expensive than the Lohand Biological meters
376 provided through FreshWater Watch.

377 The test for Chemical Oxygen Demand followed an identical procedure to those used
378 for Orthophosphate and Nitrate, albeit with a slightly longer time for colour development
379 before reading the result, yet the accuracy of the results was vastly different (Table 5c). The
380 test procedure for pH was also extremely simple, involving dipping a Simplex Health test strip
381 into the water for 3 seconds and determining the result after 15 seconds, yet despite this
382 simplicity great variability can be seen within the results. As the participants were already
383 familiar with the testing procedure for Chemical Oxygen Demand due to its similarity to other
384 parameters, and the simplicity of the pH test left little opportunity for error, variability in the
385 results of both parameters would suggest that less accurate and precise measurements
386 potentially stemmed from a difficulty in interpreting the results rather than a difficulty in
387 correctly carrying out the tests themselves to avoid contamination and reduce error (Table 5c-
388 d). Further investigations using these tests may prove beneficial in determining their accuracy,
389 and the ease with which results can be interpreted, before they could be applied to routine
390 monitoring of ambient water quality for the Sustainable Development Goals. Other published

391 studies have investigated pH using pH field test strips (Sigma-Aldrich, P-4411; Aquaspex™
392 pH-Fix 4.5-10.0) (Muenich *et al.*, 2016; Storey *et al.*, 2016) and Oakton PCtestr meters (Shupe,
393 2017) with mixed reviews. Citizen science studies to date measuring dissolved oxygen have
394 made use of the YSI Professional Plus multi-probes (Shelton, 2013) and LaMotte Direct
395 Reading Titrator kits (Storey *et al.*, 2016) with mixed results. This study measured Chemical
396 Oxygen Demand as an alternative to dissolved oxygen, yet also recorded mixed results on the
397 test's accuracy, possibly suggesting that the technology behind citizen science tests has not yet
398 advanced to the stage where accurate measurements of oxygen or oxygen demand can be taken
399 (Table 5c). However, given the multitude of published studies revealing positive results for
400 orthophosphate, nitrate and electrical conductivity with the use of various citizen science
401 equipment, finding affordable and reliable testing equipment for these parameters especially
402 should not be too great a challenge. This may allow for the initial establishment of citizen
403 science as a core source of support for ambient water quality monitoring as part of the SDGs.

404 As noted above, the percentage agreement between citizen scientist and laboratory
405 results was slightly lower in this investigation than in others involving FreshWater Watch
406 volunteers using identical testing equipment (Levesque *et al.*, 2017; Thornhill *et al.*, 2017).
407 While the lower interest levels of the students may have had an effect on the accuracy of the
408 results, neither study carried out by Levesque *et al.*, (2017) or Thornhill *et al.*, (2017) revealed
409 a 100% agreement rate between volunteer and laboratory results. This may suggest that while
410 interest and training levels do hold some influence over operator error and the accuracy of
411 results (Fore *et al.*, 2001), technology is the main limiting factor when it comes to the accuracy
412 and success of citizen science. Though technology has been a huge contributor to the
413 advancement of citizen science in recent decades (Silvertown, 2009) it also remains as a barrier
414 in certain circumstances where it is considered unreliable or unaffordable. Other published
415 studies have opted for the use of more accurate equipment with positive results (Shelton, 2013),
416 though this is unrealistic for most citizen science programmes due to the substantial associated
417 cost. Though extremely affordable, a limitation of the equipment provided by FreshWater
418 Watch for the purpose of monitoring for the ambient water quality indicator is the colorimetric
419 method by which the range of values is determined. This rather subjective process provides
420 difficulty for the user when determining whether the result lies within one range or another
421 when the true result may in fact lie on the border of the kit ranges. This happened at both sites
422 in this study when analyzing Orthophosphate, for example (Tables 2 & 4).

423

424 Other studies using the same equipment provided by FWW have also cited difficulties
425 in determining results where the existence of low nutrient concentrations means results falling
426 into the two lowest concentration categories limit finer scale analysis of nutrient patterns
427 (Levesque *et al.*, 2017; Scott & Frost, 2017). A review by Newman *et al.*, (2012) into the future
428 of citizen science using emerging technologies concluded that future citizen science
429 programmes will need to “choose appropriate technology” for the project participants. Based
430 on these observations, it is clear that further advancements in technology, whether to produce
431 a more precise and accurate result that cannot be misinterpreted, or to allow for easier
432 interpretation of a more ambiguous result, are still necessary before citizen monitoring may be
433 accepted as reliable enough to support data collection on ambient water quality as part of SDG
434 6: “Clean Water and Sanitation”.

435 On the other hand, adjustments to the assessment methods themselves may further
436 increase the ease with which citizen and professional data may be integrated for the purpose of
437 ambient water quality monitoring. During the global roll-out of the ambient water quality SDG
438 Indicator 6.3.2 a number of challenges regarding the methodology were identified, namely
439 issues surrounding the establishment of target values to determine whether a waterbody has
440 good ambient water quality or not. The current method of determining an absolute measure of
441 water quality through the comparison of measured values with target values is greatly
442 influenced by the target values selected, and thus could result in misleading interpretations of
443 water quality depending on whether the target values selected are lenient or strict (UNEP,
444 2018). As this study has revealed, while citizen science cannot provide numerical measures of
445 the parameters for the ambient water quality indicator that are as accurate as those obtained by
446 an accredited laboratory, it can indicate a concentration range for each parameter (Table 5a-b
447 & 5e). Citizen science may therefore be more applicable to a monitoring methodology in which
448 the focus shifts from target values to target ranges, allowing for the easier integration of citizen
449 science data with that of professionals. A less specific assessment method, in which the results
450 of water quality tests may encompass a range of values rather than conforming to a black-or-
451 white target value may therefore prove more approachable and applicable for citizen science
452 monitoring networks hoping to aid in the determination of ambient water quality. Assessing
453 the appropriateness of potential methods for applying citizen science monitoring to target
454 ranges in support of the ambient water quality SDG Indicator 6.3.2 should prove an important
455 focus of future studies. Another factor which must be considered is the comparability of citizen
456 science data worldwide. Differences in study design and data validation procedures have

457 oftentimes resulted in difficulty when determining the accuracy of citizen science (Storey *et*
458 *al.*, 2016). This study therefore chose to assess the quality of citizen data through comparisons
459 made with professionally-generated laboratory data, a validation procedure common in citizen
460 science water quality monitoring programmes (Muenich *et al.*, 2016; Levesque *et al.*, 2017;
461 Loiselle *et al.*, 2016; Scott & Frost, 2017; Thornhill *et al.*, 2017; Thornhill *et al.*, 2018). When
462 it comes to applying citizen science monitoring programmes to the collection of data on
463 ambient water quality for SDG Indicator 6.3.2, guidelines and protocols will have to be clearly
464 established in order to allow for the generation of comparable data, as is the case with
465 laboratory results worldwide through the use of Standard Operating Procedures (SOPs). At the
466 time of writing FreshWater Watch had collected 22,092 datasets on water quality throughout
467 the world, over 10,000 in Europe alone. While this database is a wonderful resource for
468 comparing water quality worldwide through the use of FreshWater Watch testing equipment,
469 comparisons and the integration of data with other citizen science programmes will prove
470 complicated should the advantages offered by the collection of vast amounts of data be
471 overcome by the unavoidable biases introduced via the use of different testing kits and
472 procedures. Careful consideration must therefore be given to how citizen science may be used
473 to effectively support the monitoring of ambient water quality for the Sustainable Development
474 Goals when there currently exists so many options for testing equipment, as evidenced above.
475 While greater leniency is called for through the use of target ranges for monitoring under the
476 ambient water quality indicator, stricter regulations will need to be put in place in order to
477 establish the guidelines and protocols necessary to ensure that high-quality and
478 intercomparable volunteer data is generated on ambient water quality. These considerations
479 would allow for the production of more comparable data in both developed and developing
480 nations with well-established citizen science communities. Applying citizen science in an
481 approach as such should also allow for the more effective integration of volunteer monitoring
482 programmes with current professional activities in developing nations where a lack of capacity
483 to collect and analyse water quality data required for SDG Indicator 6.3.2 hinders their ability
484 to report on ambient water quality (United Nations, 2018).

485

486 **5. Conclusions**

487 This study assessed the applicability and feasibility for citizen science to contribute towards
488 monitoring activities supporting SDG Indicator 6.3.2 on the “Proportion of bodies of water

489 with good ambient water quality”. It showed that citizen scientists can produce data on
490 Electrical Conductivity and on Orthophosphate and Nitrate concentrations, in two Irish
491 waterbodies that agreed with the analysis of these parameters at an accredited laboratory.
492 However, the precision and accuracy of the tests used for Chemical Oxygen Demand and pH
493 need further development. Through the positive conclusions drawn for three of the five water
494 quality parameters analysed, this study has demonstrated the potential of citizen science to
495 contribute to water quality monitoring for **the Sustainable Development Goals**. The limitations
496 in accuracy of the field kits used here may present challenges for how the data can be integrated
497 into existing monitoring activities.

498

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509

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