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1	Validating Citizen Science Monitoring of Ambient Water Quality for the
2	United Nations Sustainable Development Goals
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13 Abstract

14 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 15 16 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 17 source of data that may contribute to the UN Sustainable Development Goals (SDGs). The 18 availability of relatively inexpensive water quality monitoring field equipment suitable for CS 19 suggests great potential for increased spatial coverage far beyond that of traditional, laboratory-20 based monitoring networks for water quality. In support of work towards the achievement of 21 22 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 23 water with good ambient water quality". Data generated by 26 citizen scientists were compared 24 25 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 26 27 ambient water quality for the Sustainable Development Goals.

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

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- 32 33
- 1. Introduction

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

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

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

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

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

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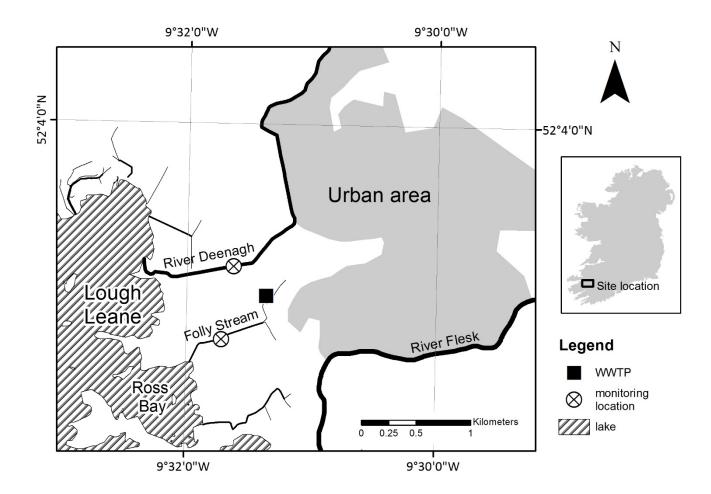
112 **2.** Methods

113 2.1 Participant Recruitment

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

- map). Based on the results of the training quiz, 28 students were selected to participate in the
 research study.
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135 *2.2 Site Description*



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Figure 1. Locations of the monitoring sites within the River Deenagh and Folly Stream catchments in southwestIreland.

Lough Leane is a freshwater lake located within Killarney National Park, draining a catchment of 553 km² near the town of Killarney, County Kerry in southwest Ireland. The rivers Flesk, Deenagh and Long Range are the main sources of input to Lough Leane, which flows to the Atlantic Ocean via the River Laune (Jennings *et al.*, 2013). The Folly stream is a minor stream of approximately 1.5 km in length that drains a small area of roughly 0.9 km² and enters Lough 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 wastewater enter the Folly stream during times when the WWTP is under stress from high-inputs (Irish Water, 2018).

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

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

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175 2.3 SDG Indicator 6.3.2 Parameters

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

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

The water quality data which feed into the indicator are derived from in-situ measurements and analysis of water samples. The citizen science field kits provided by FreshWater Watch (FWW) were capable of measuring four of the recommended ambient water quality parameters: Orthophosphate, Nitrate, Electrical Conductivity and pH. The field kits did not include tests for the other recommended parameter, dissolved oxygen (DO), so Chemical 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 Biological oxygen demand, Chemical oxygen demand	X	х	
Salinity	Electrical conductivity	Х	Х	Х
	Salinity, Total dissolved solids			
Nitrogen*	Total oxidised nitrogen	Х	х	
	Total nitrogen, Nitrite, Ammoniacal nitrogen			
	Nitrate**			Х
Phosphorus	Orthophosphate	Х	х	
	Total phosphorus			
Acidification	рН	Х	х	Х
* Countries sho	ould include the fractions of N and P which are more	st relevant in the	national context	
** Nitrate is su	ggested for groundwater due to associated human l	health risks		

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199 2.4 Citizen Analyses

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

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

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

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Table 2. Ranges of measurement of the equipment used by citizen scientists to analyse various water qualityparameters 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	0.5-10.0	>10.0					
Chemical Oxygen Demand	mg/L O2	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					
рН	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										

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241 2.5 Laboratory Analyses

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

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

Parame te r	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
Nitrate	APHA, 4500NO3-E, 23Ed., (2017) /SPC 027g	0.25-45 mg/L N	+/- 0.001	Aquakem 250 Autoanalyser
Chemical Oxygen Demand	APHA, 5520D, 23Ed., (2017) / SPC 016	10-30,000 mg/L	+/- 0	HACH/Colorimetric
рН	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 μS/cm @ 20°C	+/- 0.1	

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256 2.6 Data Analyses and Considerations

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

277

278 **3. Results**

279 3.2 Water Quality Testing

Table 4 shows the results of water quality analyses conducted by an accredited laboratory in 280 Kerry on samples taken from the River Deenagh (Site 1) and Folly stream (Site 2). Results of 281 analyses of the same water quality parameters by citizen scientists are displayed in Table 5, 282 and the percentage of their results in agreement with those obtained by the laboratory are 283 highlighted in bold (Table 5). Of the five ambient water quality parameters analysed, citizen 284 scientists demonstrated good agreement in their measurements of three - Orthophosphate, 285 286 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 agreement with the laboratory value or else fell into a concentration range just above or below 289 290 this (Table 5a). A similar result can be seen for Nitrate where between 81.3-84.8% of results across both sites fell within or just outside the concentration range corresponding to the 291 laboratory value for Nitrate (Table 5b). However, greater variation can be seen in the 292 distribution of results outside this concentration range (Table 5b). The results of electrical 293 conductivity tests by citizen scientists at the River Deenagh were also positive, with 77.4% of 294 results falling within or just outside the laboratory value of 180 µS/cm. At the Folly stream the 295 results showed less agreement, with many citizen scientists overestimating the conductivity 296 value at that site (Table 5e). 297

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

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

- difference in water quality between the two sites that supports current conclusions on the nature
- 312 of these waterbodies.
- 313
- Table 4. Results of analyses of water samples taken from the River Deenagh (Site 1) and Folly stream (Site 2) by
- an ISO/IEC 17025:2017 accredited laboratory. The means of the three laboratory analyses was calculated for each
- and parameter and used for comparison with results gathered by citizen scientists.

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

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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.

nand	Site 2 Results	22 (34.4%)	35 (54.7%)	4 (6.3%)	2 (3.1%)	1(1.6%)	0 (0.0%)	0 (0.0%)	64 (100.0%)																		
c) Chemical Oxygen Demand	Site 1 Results	46 (69.7%)	19 (28.8%)	0 (0.0%)	0 (0.0%)	1 (1.5%)	0 (0.0%)	0 (0.0%)	66 (100.0%)		Results Site 2	1 (1.6%)	8 (12.9%)	10 (16.1%)	11 (17.7%)	20 (32.3%)	9 (14.5%)	1 (1.6%)	2 (3.2%)	62 (100.0%)							
c) Ch	Range (mg/L O2)	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	Total	e) Electrical Conductivity	Range (µS/cm)	410	420	430	440	450	460	470	480	Total							
	Site 2 Results	0(0.0%)	6 (9.4%)	5 (7.8%)	4 (6.3%)	31 (48.4%)	17 (26.6%)	1 (1.6%)	64 (100.0%)	e) Electrical	Results Site 1	1 (1.6%)	4 (6.5%)	3 (4.8%)	6 (9.7%)	15 (24.2%)	30 (48.4%)	3 (4.8%)	60 (100 000)	070.001 20							
b) Nitrate	Site 1 Results	0 (0:0%)	1(1.5%)	9 (13.6%)	12 (18.2%)	42 (63.6%)	2 (3.0%)	0 (0.0%)	66 (100.0%)		Range (µS/cm)	110	130	150	160	170	180	190	Toto1	1 0141							
	Range (mg/L N)	<0.2	0.2-0.5	0.5 - 1.0	1.0-2.0	2.0-5.0	0.5 - 10.0	>10.0	Total			•															
	Site 2 Results	0 (0:0%)	13 (20.3%)	27 (42.2%)	23 (35.9%)	1(1.6%)	0 (0.0%)	0 (0.0%)	64~(100.0%)		Site 2 Results	0 (0.0%)	0 (0.0%)	0 (0.0%)	2 (3.1%)	14 (21.9%)	26 (40.6%)	4 (6.3%)	2 (3.1%)	1(1.6%)	6 (9.4%)	7 (10.9%)	2 (3.1%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	64~(100.0%)
a) Orthophosphate	Site 1 Results	29 (43.9%)	35 (53.0%)	2 (3.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	66 (100.0%)	d) pH	Site 1 Results	0 (0.0%)	0 (0.0%)	10 (15.2%)	45 (68.2%)	8 (12.1%)	2 (3.0%)	1 (1.5%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	66 (100.0%)
3	Range (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	Total		Range (pH Units)	< 4.5	4.5 - 5	5 - 5.5	5.5 - 5.75	5.75 - 6	6 - 6.25	6.25 - 6.5	6.5 - 6.75	6.75 - 7	7 - 7.25	7.25 – 7.5	7.5 - 8	8 - 8.5	8.5 - 9	> 9	Total

325 **4. Discussion**

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

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

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

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

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

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

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

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

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

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

485

486 **5.** Conclusions

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

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

498

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