

1	Using iNaturalist observations to detect disease in Red Mangroves (Rhizophora mangle)
2	Ryann E. Rossi
3	Department of Applied Ecology, North Carolina State University, Raleigh NC 27695
4	Correspondence: Ryann E. Rossi, <u>rerossi@ncsu.edu</u>
5	
6	
7	
8	
9	
10	
11	
12	
13	
14	
15	
16	
17	
18	
19	
20	
21	
22	
23	



•	14 4 .	
/	hstract:	,

24

25

26

27

28

29

30

31

32

33

34

35

36

37

38

39

Detection of disease over broad spatial scales is important to managing the spread of many diseases. One way to do this is to work with citizen scientists to collect data over broad spatial and temporal scales. Citizen science observations are becoming more widely available through web and app interfaces such as iNaturalist.org. iNaturalist.org provides passive sampling of organisms through photographs with a geolocation. These observations are often used to examine biodiversity and species monitoring, but, disease detection is also possible. Here, I demonstrate the utility of using iNaturlist.org observations of red mangrove to detect foliar disease symptoms such as lesions. I downloaded observations of red mangrove from iNaturalist.org, filtered them and examined images for foliar disease symptoms. Out of 153 filtered images, I found that 42% showed no signs of foliar disease while 58% did show foliar disease symptoms. I also found that observations of red mangrove were recorded from 15 countries in total, with 11 countries having at least one observation with foliar disease symptoms present. While small, this study demonstrates the utility of using resources such as iNaturalist.org to obtain preliminary disease observations which can be used to further focus in person disease surveys and sampling.

40

41

42

43

44

45

46



## **Introduction:**

Disease detection over broad spatial scales is often critical to managing the development and spread of many important agricultural and forest pathogens and pests. One way to increase disease detection is to involve citizens in the collection of this data. Recently, citizen scientists helped document and track the spread of Sudden Oak Death in California by learning how to identify and survey oak trees and alternative hosts for the pathogen (Meentemeyer et al. 2015; Rizzo & Garbelotto 2003). Citizen scientists that participated in the Sudden Oak Death project attended training sessions and then went into the field to detect and collect samples of the pathogen. Given the severity of this disease, this approach ensures citizen scientists collect the correct data. However, time and other resources are spent training and interacting with participants which may not be possible for disease monitoring of less severe diseases.

Another way to detect diseases with citizen scientists is through more passive sampling such as photographs with a geolocation. One such resource to use for this method is iNaturalist (https://www.inaturalist.org/). iNaturalist is a crowdsourced database of natural history observations. Photographs are typically included in observations so that observations can be validated (Freitag et al. 2016). Often, iNaturalist is used as a tool to record the biodiversity of species (Palumbo et al. 2012; Pimm et al. 2015). However, observations from iNaturalist could be used for much more such as species monitoring and detection of specific diseases and/or pests. For example, as of September 2017, there are 17 projects that ask contributors to focus images on detection of specific diseases and pests. Here, I demonstrate the utility of using general iNaturalist observations (i.e., not collected for a specific project) to detect foliar disease symptoms in mangrove forests.



70	Materials and Methods:
71	iNaturalist observations were downloaded in April 2017. "Red mangrove" was used as the search
72	term. Observations were individually examined to confirm <i>R.mangle</i> was correctly identified.
73	Images that were low resolution, taken at too far of a distance, or did not include leaves in the
74	image were excluded.
75	Following filtering of observations, images were examined for foliar disease symptoms. The
76	symptoms focused on were lesions (necrotic tissue on the leaf). Only images that had leaves
77	present with lesions were counted as having disease symptoms. We did not consider the number
78	of leaves in the images with lesions since we simply wanted to detect the presence or absence of
79	disease in the images.
80	
81	Results and Discussion:
82	I found that 16 out of the original 405 observations were not truly <i>R.mangle</i> . Of the remaining
83	389 observations that were <i>R.mangle</i> , 236 images were excluded leaving a total of 153
84	observations that were viable (i.e., could zoom to investigate leaves). The 153 observations were
85	from 15 countries (Table I). 42% (n=64) of observations showed no disease symptoms and 58%
86	(n=89) did show disease symptoms. Overall, disease symptoms were observed in 11 countries
87	(Table I). Mexico (n=53) and the United States (n=70) had the greatest number of observations
88	recorded with 49% and 70% of observations having disease, respectively.
89	
90	Overall, I found that iNaturalist observations of R.mangle can be used to detect foliar disease
91	symptoms. Although I focused on foliar disease detection, iNaturalist could also be used to
92	document canker disease on roots, scale insect outbreaks and herbivore damage (all of which



93	were identified in these observations). While these observations may not provide enough detail
94	to confirm a specific type of disease, this tool provides useful baseline information that can be
95	used to help direct more in-depth disease surveys.
96	To improve the use of iNaturalist observations for disease detection, specific projects could be
97	initiated that direct contributors to what they should focus their photos on. For example,
98	contributors could take multiple photos that capture the entire plant in addition to close-up
99	photos of whichever part of the plant exhibits disease symptoms.
100	
101	Acknowledgements:
102	I would like to thank the many citizen scientists and iNaturlaist.org for providing these data. This
103	work was funded by the National Science Foundation Graduate Research Fellowship DGE-
104	0946818.
105	
106	
107	
108	
109	
110	
111	
112	
113	
114	
115	



## 116 Table I:

Country	Disease	No . observatio	ns
Aruba	0	1	118
Bahamas	1	2	110
Belize	3	4	119
Cayman Islands	0	1	
Colombia	1	6	120
Costa Rica	1	1	
Cuba	2	3	121
Dominican Republic	0	1	122
Ecuador	1	4	122
Honduras	0	1	123
Mexico	26	53	
NA	2	2	124
Panama	1	1	425
Puerto Rico	1	2	125
Senegal	1	1	126
United States	49	70	120
	I		127



144	References:
145	Freitag A, Meyer R, and Whiteman L. 2016. Strategies employed by citizen science programs to
146	increase the credibility of their data. Citizen Science: Theory and Practice 1.
147	Meentemeyer RK, Dorning MA, Vogler JB, Schmidt D, and Garbelotto M. 2015. Citizen science
148	helps predict risk of emerging infectious disease. Frontiers in Ecology & the Environment
149	13:189-194. 10.1890/140299
150	Palumbo MJ, Johnson SA, Mundim FM, Lau A, Wolf AC, Arunachalam S, Gonzalez O, Ulrich JL,
151	Washuta A, and Bruna EM. 2012. Harnessing smartphones for ecological education,
152	research, and outreach. The Bulletin of the Ecological Society of America 93:390-393.
153	Pimm SL, Alibhai S, Bergl R, Dehgan A, Giri C, Jewell Z, Joppa L, Kays R, and Loarie S. 2015.
154	Emerging Technologies to Conserve Biodiversity. Trends in Ecology & Evolution 30:685-
155	696. https://doi.org/10.1016/j.tree.2015.08.008
156	Rizzo DM, and Garbelotto M. 2003. Sudden oak death: endangering California and Oregon
157	forest ecosystems. Frontiers in Ecology and the Environment 1:197-204.
158	