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Deeper habitats and cooler temperatures moderate a climate-driven seagrass disease

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Deeper habitats and cooler temperatures moderate a climate-driven disease in an essential marine habitat

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2. On revision: All data and R scripts used to generate the analyses presented here will be publicly available via the Cornell University eCommons Repository by the time of publication (<https://doi.org/10.7298/6ybh-w566>).

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Title

Deeper habitats and cooler temperatures moderate a climate-driven disease in an essential marine habitat

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Abstract

Eelgrass creates critical coastal habitats worldwide and fulfills essential ecosystem functions as a foundation seagrass. Climate warming and disease threaten eelgrass, causing mass mortalities and cascading ecological impacts. Subtidal meadows are deeper than intertidal and are valuable fish nursery grounds that may also provide refuge from the temperature-sensitive seagrass wasting disease. From cross-boundary surveys of 5,761 eelgrass leaves from Alaska to Washington and assisted with a machine-language algorithm, we measured outbreak conditions.

Across summers 2017 and 2018, predicted disease prevalence was nearly 40% lower for subtidal than intertidal leaves; in both tidal zones, disease risk was lower for plants in cooler conditions. Even in the environmentally more stable subtidal meadows, we observed high

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3 46 disease levels, with half of the sites exceeding 50% prevalence. Models predicted reduced
4 47 disease prevalence and severity under cooler conditions, confirming a strong interaction between
5 48 disease and temperature. At both tidal zones, prevalence was lower in more dense eelgrass
6 49 meadows, suggesting disease is suppressed in healthy, higher density meadows. These results
7 50 underscore the value of subtidal eelgrass and meadows in cooler locations as refugia, indicate
8 51 that cooling can suppress disease, and have implications for eelgrass conservation and
9 52 management under future climate change scenarios.
10 53

12 54 **Keywords**

13 55 Seagrass, eelgrass, marine disease, seagrass wasting disease, climate change, climate refugia
14 56

16 57 **Introduction**

17 58 The increasing incidence and severity of disease outbreaks [1–3]—fueled by acute and
18
19 59 prolonged warming ocean temperatures [1,4–9]—makes disease ecology on both land and sea a
20 60 priority in the portfolio of climate change research. Temperature-sensitive pathogens that target
21
22 61 marine foundation species like corals and eelgrass (*Zostera marina*), a temperate seagrass
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24 62 species, can be especially devastating, given their pivotal roles in driving marine ecosystem
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26 63 structure and function [7,9–11]. Eelgrass has the largest global distribution of any marine
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28 64 angiosperm, and grows in shallow, coastal areas throughout the northern hemisphere, spanning
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30 65 from Baja, Mexico to Alaska [12]. Seagrass wasting disease, caused by the protist *Labyrinthula*
31
32 66 *zosteriae*, is one of the current threats to the health and sustainability of global seagrass meadows
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34 67 [13,14]. The pathogen consumes plant chloroplasts [15], impairs photosynthesis [16], produces
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36 68 distinctive black lesions [17–19], and reduces eelgrass growth and belowground sugar stores in
37
38 69 natural meadows [20]. Historical disease outbreaks in the 1930s reduced some eelgrass meadows
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40 70 along the Atlantic coasts by 90% and dramatically altered their structure and function [21,22],
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42 71 reducing waterfowl and invertebrate populations [21,23–25], and altering the water quality in
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44 72 coastal regions [26]. Eelgrass disease outbreaks continue to persist in temperate seas worldwide
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46 73 [9,27–32], and can result not only in local extinctions, but also in the loss of the valuable
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3 74 ecosystem services eelgrass provides: carbon sequestration, sediment stabilization, water
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5 75 filtration, nutrient cycling, and habitat formation [33–35].
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8 76 Warming ocean temperatures and wasting disease can independently and synergistically
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10 77 interact and harm eelgrass. Rising temperature, including increased frequency and intensity of
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12 78 marine heat waves [36], is the most prominent global change factor impacting seagrass
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14 79 ecosystems [37,38], which are declining globally [39]. Warmer temperatures are associated with
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16 80 dramatic reductions in eelgrass growth [40,41], net primary production [42], density [8,43], and
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18 81 biomass [44]. Dramatic examples include widespread mortality of eelgrass in the Chesapeake
19
20 82 Bay, Virginia [42] and other seagrass in Western Australia [47] from marine heatwaves.
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22 83 Following recent marine heatwaves, shallower, warmer estuaries also had reduced eelgrass
23
24 84 biomass compared to deeper, cooler estuaries [45]. Further, warmer temperatures under climate
25
26 85 change projections are expected to substantially shift eelgrass ranges northward and increase
27
28 86 eelgrass susceptibility to anthropogenic and natural stressors like disease [46].
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33 87 Along with rising temperatures, seagrass wasting disease is among one of many multiple
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35 88 stressors threatening global seagrass meadows [14,48]. Climate change is predicted to increase
36
37 89 disease impacts on eelgrass health and meadow resistance [14]. Certain abiotic conditions—
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39 90 including warm temperatures—were implicated in historic wasting disease outbreaks [26,49,50].
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41 91 More recently, elevated temperatures [9,27,32] were associated with higher disease levels in
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43 92 natural meadows. Field surveys also suggest wasting disease and warmer temperatures facilitated
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45 93 seagrass declines in Sicily, Italy [27] and North America [9,13,14,32,51]. Lab experiments
46
47 94 demonstrate the causative agent, *L. zosterae*, grows faster at warmer temperatures up to 25° C
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49 95 [52,53], though the exact mechanisms underlying this relationship remain unknown [54]. Certain
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51 96 eelgrass biometrics are also associated with greater wasting disease. Field surveys detected
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3 97 significant, positive correlations between disease metrics and eelgrass leaf area and negative
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5 98 correlations between disease and shoot density [9,29,30]. Many other environmental parameters
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8 99 influence eelgrass health and survival (ex: exposure to waves and desiccation stress, salinity,
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10 100 sediment), though temperature, light, and nutrients are the most important for eelgrass health and
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12 101 productivity [40,55,56]. Despite the growing understanding of the role of climate and other
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15 102 environmental drivers on wasting disease, little is known about factors that lead to better
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17 103 outcomes for natural meadows, such as cooler, higher latitudes or deeper water.
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19 104 To capture a broad range of environmental conditions, better understand the synergistic
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21 105 effects of climate and disease on this foundation species, and determine the potential for cold,
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23 106 deep refugia, disease surveys spanning a wide latitude and depths in the northern range of
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26 107 eelgrass distribution are essential. Previous studies reported that disease was lower in deeper
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28 108 eelgrass meadows (-4 m mean low low water, -2 to -5 m, respectively) in the San Juan Islands,
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30 109 Washington, and Sweden [29,31]. This suggests the hypothesis that deeper, subtidal eelgrass
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33 110 meadows may provide plants with more favorable climatic conditions—and less favorable
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35 111 conditions for the pathogen—that allow them to persist [57,58]. Similar patterns were found
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38 112 among three species of algae, which had more severe infections in shallower regions compared
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40 113 to those at depth [59]. Refugia from climate change and disease pressure could potentially
41
42 114 mitigate local extinctions due to disturbances [58]. Already, deeper habitats serve as refugia from
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44 115 marine heatwaves for seaweeds [60], corals [61], temperate reefs [62], and eelgrass [45]. These
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47 116 examples highlight how deeper marine environments could reduce the impacts of climate change
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49 117 and pathogenic stressors, and exemplify the need to further understand host-pathogen
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51 118 interactions in these environments.
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3 119 We aimed to test the following hypotheses: (i) Disease is reduced in meadows at higher
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5 120 latitudes with cooler temperatures. (ii) Disease levels are lower in deeper, subtidal eelgrass
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8 121 compared to the more environmentally stressful conditions of shallower, intertidal eelgrass. (iii)
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10 122 Disease is higher in high-density eelgrass meadows, since the disease transmits via direct contact
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12 123 with infected leaves [15]. To address these, we surveyed seagrass wasting disease in eelgrass
13
14 124 meadows throughout their northern range from Puget Sound, Washington to Southeast Alaska in
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16
17 125 the Northeast Pacific to explore how disease varied across eight degrees latitude, tidal zones
18
19 126 (intertidal or subtidal), environments, and time. Altogether, we surveyed 5,761 eelgrass leaves
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21 127 from paired, adjacent intertidal and subtidal eelgrass meadows for leaf-specific measurements
22
23 128 (leaf area, disease prevalence, and severity) and site-specific biometrics (density and canopy
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25 129 height). Intertidal eelgrass meadows are exposed to more stressful, extremely variable
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28 130 environmental conditions at low tide, including higher temperatures, desiccation, UV stress, and
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31 131 at high latitudes, scouring by sea ice [63,64]. In contrast, deeper, subtidal meadows are
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33 132 constantly submerged and have more stable environmental conditions. Just as environmental
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35 133 conditions can vary dramatically with elevational gradients and influence disease dynamics on
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37 134 land [65]—so too can the environment and disease vary with depth in our oceans. Because
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40 135 intertidal eelgrass is exposed at low tide to greater environmental stressors, it could be more
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42 136 vulnerable to infection in a changing climate. Intertidal environments could also be more
43
44 137 conducive to pathogen growth. Given the ecological significance of eelgrass meadows—
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46 138 particularly as fish nursery and feeding grounds [66]—and the relatively little that is known
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48 139 about disease at depth [31], we made investigation of subtidal disease a key research priority in
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50 140 this project.
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142 **Methods**

143 **Field surveys.** We surveyed 19 intertidal and subtidal eelgrass meadows across four
144 geographic regions: Southeast Alaska (AK); British Columbia, Canada (BC); San Juan Islands,
145 Washington (SJ); and Puget Sound, Washington (PS) (Figure 1A, Figure S1, Table S1). Regions
146 spanned sea surface temperature gradients and ranged from urban environments with high human
147 impacts to remote environments with minimal to no development. For example, British
148 Columbia sites were in the Hakai Lúxvbálís Conservancy, the largest marine protected area
149 along coastal British Columbia (BC Parks), while Puget Sound sites in Washington were heavily
150 urbanized, with some adjacent to a wastewater treatment plant and railroads. Surveys occurred in
151 the summers of 2017 and 2018, when disease levels peak in temperate eelgrass [9,28,32,67]. Due
152 to logistical constraints, we had to stagger our sampling periods as such: We surveyed British
153 Columbia in late June, Puget Sound in early July, San Juan Islands in mid-late July, and Alaska
154 in early August. Within a given region, we surveyed all sites on the same low-tide series.

155 In each region, we surveyed 3-5 paired intertidal and subtidal eelgrass meadows, except
156 in British Columbia where three sites were strictly intertidal or subtidal. The San Juan Islands
157 have a history of wasting disease monitoring [9,29,30] and recent, significant meadow declines
158 [9,68]. For each field survey, we ran three, 20-m transects parallel to shore in the middle of both
159 intertidal and subtidal meadows. We sampled intertidal meadows at low tide and subtidal
160 meadows using SCUBA or snorkeling (Supplemental Video 1). During 2017, we recorded the
161 GPS coordinates at the ends of all intertidal transects for subsequent monitoring in 2018, so that
162 we could compare the same parts of the meadows between years. We tracked subtidal transect
163 locations using GPS coordinates from boats, dive compass headings, and in some cases,
164 anchored subtidal transect markers. At each site, we haphazardly collected 120 intertidal and 60

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3 165 subtidal leaves (n=40 leaves/transect, n=20 subtidal leaves/transect). Given the constraints of
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5 166 working underwater, the significantly larger size of subtidal eelgrass leaves compared to
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7 167 intertidal leaves, and the greater processing time required to process larger leaves, we collected
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10 168 fewer subtidal leaves. Intertidal meadows were at approximately +1 m and subtidal meadows
11
12 169 were at depths ranging from approximately -1.8 to -6 m mean low low water. Because disease
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14 170 susceptibility and levels can vary with the age of eelgrass leaves [29], we standardized our
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16
17 171 collections to the third-rank (third youngest) leaf from each shoot, following other published
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19 172 approaches [9]. For a subset of sites in British Columbia, San Juan Islands and Puget Sound, we
20
21 173 also measured shoot density and canopy height from quadrats at three points along each transect
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23 174 (0, 10, 20 m). Due to logistical constraints, we did not measure density in any subtidal Puget
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26 175 Sound meadows in 2018. We stored all leaves in bags with seawater on ice or in a refrigerator
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28 176 until processing for image analyses.
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33 178 **Disease quantification.** In lab, we gently scraped epiphytes from eelgrass leaves using
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35 179 soft, flexible rulers. We scanned eelgrass leaves between two transparency sheets with a Canon
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37 180 CanoScan LiDE 220 scanner at 600 dpi resolution within 24 hours of collection. This created
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39 181 digital images of eelgrass leaves for subsequent leaf area and disease measurements. Given that
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41 182 some subtidal leaves were nearly 3 m long, we scanned only diseased or potentially diseased
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43 183 portions of subtidal leaves for more efficient processing. Consequently, we measured the lengths
44
45 184 and widths of each subtidal leaf by hand prior to scanning, and used these to calculate subtidal leaf
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47 185 areas. We scanned entire intertidal leaves, which were smaller than subtidal leaves, and used leaf
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49 186 areas measured by a machine-learning algorithm.
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3 187 To precisely measure leaf-level disease prevalence and severity, we leveraged the
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5 188 Eelgrass Lesion Image Segmentation Analyzer (EeLISA), a robust algorithm that identified and
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7 189 measured healthy and diseased tissue on all images of scanned eelgrass leaves [9,32,69]. The
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10 190 algorithm calculated disease prevalence (presence/absence of disease) and lesion area for each
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12 191 leaf, along with leaf area estimates for intertidal leaves. Using leaf-level prevalence, we
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14 192 calculated transect- and site-level mean prevalence (proportion of infected individual plants); we
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16 193 calculated severity (proportion of infected leaf area) using lesion and leaf area measurements at
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18 194 leaf, transect, and site-levels. Importantly, this award-winning algorithm was instrumental in
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20 195 enabling us to efficiently and consistently survey disease across a broad, latitudinal scale, as
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22 196 previous methods of measuring disease lesions by hand would have severely limited the scope of
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24 197 our surveys; measuring diseased lesions by hand can take more than 30 minutes for one eelgrass
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26 198 leaf and can be a significant bottleneck for disease analyses [69].
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33 200 **Pathogen confirmation.** We confirmed that the black-edged, necrotic lesions we
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35 201 identified as wasting disease were caused by the pathogen *L. zosterae* and asymptomatic, healthy
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37 202 eelgrass did not contain *L. zosterae* using qPCR analyses (n=98 eelgrass leaves tested), following
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39 203 established protocols [9,28,32,70]. Subsequent qPCR analyses of diseased eelgrass from the San
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41 204 Juan Island, WA sites also confirmed the presence of *L. zosterae* [32].
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47 206 **Temperature & salinity data.** To determine the relationship between disease and sea
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49 207 surface temperatures, we assessed remote-sensed sea surface temperatures for all sites from
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51 208 January – June 2017 and 2018, following previously published methods [9,32]. Briefly, we
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53 209 extracted Group for High Resolution Sea Surface Temperatures (GHRSSST) Level 4, Multi-Scale
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3 210 Ultra-High Resolution (MUR) daily temperatures for each site from the Jet Propulsion
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5 211 Laboratory OPeNDAP portal [71]. For each site, temperatures were extracted from a 1 x 1 km
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7 212 area over the ocean; these sea surface temperatures were measured at the site-level and did not
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10 213 differentiate between subtidal and intertidal meadows, as our surveys did not extend beyond a 1
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12 214 x 1 km area at each site.

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14 215 To evaluate sea surface temperatures relative to each site, we calculated five different
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16 216 temperature anomaly metrics for each month (from January – June 2017 and 2018, respectively),
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18 217 consistent with previous work exploring impacts of temperature anomalies on marine
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20 218 environments [5,6,32]; we did not use absolute temperatures. All temperature metrics were
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22 219 calculated based on the daily, satellite-derived sea surface temperature for each site and the long-
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24 220 term, 17-year mean (2002 – 2018) monthly temperature for the site. The five temperature
25
26 221 anomaly metrics included: CDiffMean (cumulative difference between daily temperature and
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28 222 long-term mean), CDiffMeanHeat (cumulative positive difference between daily temperature and
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30 223 long-term mean), CDiffMeanCold (cumulative negative difference between daily temperature
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32 224 and long-term mean), CDiffT90Heat (cumulative positive difference between daily temperature
33
34 225 and long-term 90th percentile monthly temperature), CDiffT90Cold (cumulative negative
35
36 226 difference between daily temperature and long-term 90th percentile monthly temperature). These
37
38 227 temperature anomalies were cumulative temperature differences summed over a one-month
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40 228 period. We restricted temperatures from January – June of 2017 and 2018, since we began our
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42 229 disease surveys in late June of each year, and we did not want to include site temperatures after
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44 230 we had already collected eelgrass. We specifically did not include temperature anomalies for
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46 231 regions sampled after June (AK, SJ, PS) because we wanted to run temperature anomaly models
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48 232 that compared disease across all regions and sites simultaneously, rather than separate, region-
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3 233 specific models. All temperature metrics from January – June 2017 and January – June 2018
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5 234 were centered and scaled, then subset by month for subsequent models, described below.
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10 236 **Statistical analyses.** We performed all statistical analyses in R version 4.1.2 [72] and
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12 237 visualized data using the packages *ggplot*, *ggpubr*, and *RcolorBrewer* [73–75]. Data exploration
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14 238 and subsequent model fitting and validation were carried out following published protocols [76].
15
16 239 We incorporated remote-sensed sea surface temperatures into models to determine the effects of
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18 240 environment (temperature anomaly) and eelgrass biometrics (leaf area, density) on disease
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20 241 prevalence and severity. We used the *glmmTMB* function in the *glmmTMB* package to fit
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22 242 binomial generalized linear mixed models for prevalence [77], and the *lmer* function and *lme4*
23
24 243 package to fit linear mixed effects regression models for severity [78]. Fixed effects in all models
25
26 244 included tidal zone (subtidal vs intertidal), year, temperature anomaly, and leaf area, and
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28 245 interactions (detailed below); subsequent models also included eelgrass density. We centered and
29
30 246 scaled all numeric fixed effects—leaf area, density, and temperature anomaly—in order for the
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32 247 models to converge. To account for the hierarchical sampling design, we included the random
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34 248 nested effects of region, site, tidal zone, and transect in all models. Our nested design allowed for
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36 249 disease comparisons across broad environmental and spatiotemporal gradients.
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42 250 Given that some parameters were only measured at a subset of sites for both years, we
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44 251 ran several different models on our data. The most comprehensive prevalence and severity
45
46 252 models include data from all sites (n=5761 and n=3457 leaves, respectively; Table S2).
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48 253 Subsequent prevalence and severity models used a subset of the dataset, which included density
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50 254 (n=4090 and n=2549 leaves; Table S3). All data and R scripts used to generate the analyses
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3 255 presented here will be publicly available via the Cornell University eCommons Repository
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5 256 (<https://doi.org/10.7298/6ybh-w566>).
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10 258 *Developing leaf area, temperature, and disease models*
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12 259 To determine the best binomial generalized linear mixed model structure for leaf-level
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14 260 prevalence (Table S2), we ran models that included fixed effects of leaf area, tidal zone, year,
15
16 261 temperature anomaly, and interactions between some of these terms. We only tested interactions
17
18 262 that were biologically meaningful, such as leaf area and tidal zone interactions or leaf area and
19
20 263 year interactions, but not tidal zone and year interactions. Such interactions were considered
21
22 264 potentially biologically meaningful, since subtidal eelgrass leaves are considerably longer and
23
24 265 wider compared to those in intertidal zones [79]. Likewise, leaf area could interact with year, if
25
26 266 one year was warmer or cooler than another, since temperature strongly influences eelgrass
27
28 267 growth [40,41]. We subset temperature anomaly metrics to the month of March for this stage of
29
30 268 model development, as March included a range of temperatures above and below the long-term,
31
32 269 historical mean. The best-fit prevalence model structure had the lowest AICc (corrected Akaike
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34 270 Information Criterion) and included the following fixed effects and interactions: tidal zone, year,
35
36 271 leaf area, temperature anomaly, leaf area*tidal zone, leaf area*year. We then used this model
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38 272 structure to test subsequent monthly temperature anomaly models, switching out the five
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40 273 different temperature anomaly metrics described above (CDiffMean, CDiffMeanHeat,
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42 274 CDiffMeanCold, CDiffT90Heat, CDiffT90Cold), calculated on a monthly basis from January to
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44 275 June. This allowed us to determine which month's temperature metrics were the best fit for the
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46 276 prevalence model. We used AICc to select the best-fit, leaf-level prevalence model, which
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3 277 included a March cold temperature anomaly (CDiffMeanCold, n=5761 leaves, Table S2). We
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5 278 validated the model by assessing diagnostic plots made with the DHARMA package [80].
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8 279 We followed a similar process to develop the linear mixed effects regression model for
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10 280 leaf-level severity (Table S2). Because we used a hurdle model approach for analyzing disease
11
12 281 severity, we only included data for leaves with disease and excluded healthy individuals; we also
13
14 282 logit-transformed severity since the data were bound between 0 and 1, following established
15
16 283 protocols [81]. The best-fit, leaf-level severity model had the lowest AICc and included the
17
18 284 following fixed effects and interactions: tidal zone, year, leaf area, temperature anomaly, and leaf
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20 285 area*temperature anomaly. This model included a March cold temperature anomaly
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22 286 (CDiffT90Cold, n=3457 leaves, Table S2). To evaluate the model for normality and
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24 287 homogeneity of residuals, we visually checked diagnostic plots created with the *plot_model*
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26 288 function in the sjPlot package [82].
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33 290 *Developing leaf area, temperature, density, and disease models*

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35 291 We developed additional prevalence and severity models (Table S3) based on the subset
36
37 292 of sites for which we had eelgrass density—British Columbia, San Juan Islands, Puget Sound—
38
39 293 following the model development and selection process described above. The best-fit, binomial
40
41 294 generalized linear mixed model for leaf-level prevalence (Prev Mod 2) included the following
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43 295 fixed effects and interactions: tidal zone, year, leaf area, cold temperature anomaly
44
45 296 (CDiffMeanCold) for March, density, leaf area*CDiffMeanCold, CDiffMeanCold *mean
46
47 297 density, tidal zone*mean density (n=4090 leaves, Table S3). The best-fit, linear mixed effects
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49 298 regression hurdle model for leaf-level severity (Sev Mod 2) included the following fixed effects
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51 299 and interactions: tidal zone, year, leaf area, temperature anomaly (CDiffMean) for March,
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3 300 density, year*CDiffMean (n=2549 leaves, Table S3). For this model, we also used a “bobyqa”
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5 301 optimizer to support model convergence. As before, we used DHARMA diagnostic plots and qq-
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7 302 plots to assess respective models [80,82].
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12 305 **Results**

13 306 *Broad disease patterns*

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17 307 Disease was significantly higher in 2018 compared to 2017 (Table S2). Among the four
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19
20 308 regions, disease prevalence (proportion of infected individual plants) and severity (proportion of
21
22 309 tissue infected) increased in all regions in 2018 except for Puget Sound, which had reduced
23
24 310 disease (Figure 1B, Figure S2, Table S4). The most dramatic changes in disease between years
25
26 311 were in the intertidal. Intertidal prevalence in Alaska shifted from $22.05 \pm 2.61\%$ to $61.11 \pm$
27
28 312 3.08% the subsequent year, and regional severity changed from $1.62 \pm 0.34\%$ to $11.22 \pm 1.08\%$
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30 313 (mean \pm SE, Figure 1B, Figure S2). Spatially, leaf-level disease prevalence and severity were
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32 314 reduced at higher latitudes compared to lower latitude regions, though disease varied
33
34 315 considerably between sites (Figure 1B, Figure S2). This latitudinal gradient was more apparent
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36 316 in the higher-resolution severity data, with Alaska and British Columbia reporting lower disease
37
38 317 severity across both years and tidal zones compared to regions further south (Figure 1B).
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43
44 318 Prevalence and severity were significantly lower in subtidal meadows compared to the
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46 319 intertidal (glmm and lmer, $p < 0.001$, Table S2). When averaged across both years, the mean
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48 320 prevalence for intertidal eelgrass was $66.0 \pm 0.79\%$, compared to $50.4 \pm 1.06\%$ among subtidal
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50 321 plants (probability \pm SE). At the site-level, disease prevalence ranged from $7.93 \pm 3.43\%$ to
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52 322 100% among intertidal eelgrass and from $8.45 \pm 3.32\%$ to $95.23 \pm 2.7\%$ among subtidal eelgrass
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3 323 (mean \pm SE, Figure S2). Out of 70 total intertidal and subtidal sampling events across the two
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5 324 years, 41 had a mean prevalence greater than 50%, indicating widespread infection (Figure S2).
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7 325 Differences in severity between tidal zones were even more striking (Figure 1B, Table S2).
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9 326 When averaged across both years, severity for intertidal plants was $10.05 \pm 0.27\%$, compared to
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11 327 $3.12 \pm 0.17\%$ among subtidal plants (mean severity \pm SE). Site-level disease severity ranged
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13 328 from $0.14 \pm 0.096\%$ to $33 \pm 1.85\%$ among intertidal eelgrass, compared to $0.054 \pm 0.029\%$ to
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15 329 $16.3 \pm 2.78\%$ among subtidal eelgrass (mean \pm SE, Figure 1B). Of the 70 sampling events, 23
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17 330 had a mean severity greater than 10% (Figure 1B).
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24 332 *Leaf area, temperature, and disease models*

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26 333 We tested five temperature metrics calculated for each month (January – June) when
27
28 334 developing leaf-level prevalence and severity models. Of these, March temperature anomalies
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30 335 were in the best-fit models, based on the lowest AICc. Sea surface temperatures in March 2017
31
32 336 and 2018 varied regionally, with generally colder absolute temperatures in higher-latitude
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34 337 regions (Figure S3). All regions experienced warmer temperatures in March 2018 than March
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36 338 2017 except for Puget Sound, which was cooler that year (Figure 3A). This coincided with
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38 339 reduced disease prevalence and severity in Puget Sound relative to 2017 (Figure 1B, Figure S2).
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42 340 Leaf-level, summertime prevalence significantly decreased with cooler March
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44 341 temperatures, as predicted (glmm, $p < 0.001$, Table S2). Predicted prevalence decreased with
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46 342 cooler March temperature anomalies (CDiffMeanCold) for both intertidal and subtidal eelgrass
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48 343 (Figure 3B). Other significant predictors for leaf-level prevalence included: tidal zone, year, leaf
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50 344 area, leaf area*tidal zone, and leaf area*year (glmm, $p < 0.001$, Table S2). Across both tidal
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345 zones, transect-level disease prevalence was positively associated with cumulative March cold
346 temperature anomalies and leaf areas (Figure 2A, Figure S5).

347 Similarly, leaf-level severity significantly decreased with cooler March temperatures
348 (lmer, $p < 0.001$, Table S2). Among diseased leaves, predicted summertime severity decreased
349 with cumulative, 90th percentile cold March temperature anomalies in subtidal and intertidal
350 eelgrass (Figure S4). Compared to absolute cold temperature anomalies measured on a daily
351 basis, this cold temperature anomaly (CDiffT90Cold) is the accumulation of negative differences
352 between each site's daily temperatures and the long-term 90th percentile mean temperatures for
353 March 2017 and 2018. Other significant predictors of leaf-level severity include tidal zone, year,
354 and leaf area*CDiffT90Cold (lmer, $p < 0.001$, Table S2). For intertidal leaves, disease severity
355 was positively associated with cumulative, 90th percentile March cold temperature anomalies and
356 leaf areas, though these associations were not as apparent among subtidal leaves (Figure S5,
357 Figure S6).

359 *Leaf area, temperature, density, and disease models*

360 We measured densities in three of the four surveyed regions: British Columbia, San Juan
361 Islands, and Puget Sound. Short survey times in the remote sites in Alaska precluded density
362 measurements. Mean eelgrass densities varied among sites and tidal zones and between years for
363 several sites (Figure S7). Shoot densities were significantly higher in intertidal meadows
364 compared to subtidal in the San Juan Islands (t-test: $t(178) = 4.01$, $p < 0.001$) and Puget Sound
365 ($t(103) = 2.60$, $p = 0.01$), but not in British Columbia (Figure S4; $t(124) = -1.82$, $p = 0.07$). At the
366 transect level, low-density intertidal eelgrass had higher disease prevalence and severity
367 compared to eelgrass at higher densities (Figure S8). Changes in mean density in 2018 were not

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3 368 strongly associated with the prior year mean severity (*data not shown*), suggesting that other
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5 369 factors likely interact with disease to influence eelgrass persistence.
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8 370 Leaf-level prevalence was significantly, inversely associated with mean shoot density
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10 371 (glmm, $p < 0.001$, Table S3). High disease levels were associated with reduced eelgrass densities
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12 372 in both subtidal and intertidal meadows (Figure S8). The best-fit prevalence and density model
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14 373 included the following predictors, all of which were significant: tidal zone, leaf area, year, March
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16 374 cold temperature anomaly (CDiffMeanCold), density, leaf area* CDiffMeanCold,
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18 375 CDiffMeanCold*density, tidal zone*density. Interactions between temperature and density had
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20 376 the most pronounced effect on predicted prevalence at low densities. At low densities, lower
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22 377 predicted disease prevalence was associated with cooler temperatures, while higher predicted
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24 378 prevalence was associated with warmer temperatures (Figure S9). This association was
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26 379 consistent at mean densities, but did not persist at high eelgrass densities.
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31 380 Leaf-level severity was not significantly associated with mean shoot density (lmer,
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33 381 $p > 0.05$, Table S3). The best-fit, hurdle severity model included the following: tidal zone, leaf
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35 382 area, year, March temperature anomaly (CDiffMean), density, year* CDiffMean. There was not
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37 383 a consistent association between March temperature anomaly, eelgrass densities, and predicted
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39 384 severity in 2017 and 2018.
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45 386 *Eelgrass biometrics*
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47 387 Consistent with previous work [79], eelgrass leaves were smaller at shallower depths
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49 388 (Figure S7). Mean canopy height was $599.02 \pm 9.99\%$ in intertidal eelgrass and $1068.71 \text{ mm} \pm$
50
51 389 14.58% in subtidal eelgrass when averaged across years (mean \pm SE, Figure S7). Mean leaf area
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53 390 was also smaller among intertidal eelgrass compared to subtidal eelgrass. Across both years,
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3 391 mean leaf area was $1935.14 \text{ mm}^2 \pm 24.31\%$ in intertidal eelgrass and $5267.93 \text{ mm}^2 \pm 72.76\%$ in
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5 392 subtidal eelgrass (mean \pm SE, Figure S7). When modeled with temperature, leaf area was
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7 393 significantly, positively associated with leaf-level disease prevalence (glmm, $p < 0.001$, Figure
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9 394 3B, Table S2). Although subtidal eelgrass leaves were on average nearly three times larger than
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11 395 intertidal eelgrass, disease prevalence and severity were significantly lower in subtidal plants.
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17 397 *qPCR*

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19 398 We successfully confirmed the presence of *L. zosterae* in 19 out of 49 symptomatic,
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21 399 lesioned eelgrass from British Columbia and Puget Sound using qPCR. All asymptomatic
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23 400 eelgrass tested from these regions were qPCR negative for the pathogen ($n=49$). We isolated *L.*
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25 401 *zosterae* from diseased eelgrass in the San Juan Islands to confirm pathogen presence (*data not*
26
27 402 *shown*). Other studies also confirmed *L. zosterae* in diseased eelgrass in the San Juan Islands and
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29 403 Alaska [9,29,30,32,53,83]; these findings support that our visual identification of dark, necrotic
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31 404 lesions were caused by *L. zosterae*.
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38 406 **Discussion**

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41 407 The two study years, 2017 and 2018, captured outbreak conditions of relatively high
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43 408 disease levels across a wide latitude in the northern range of eelgrass, from Puget Sound to
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45 409 Alaska, including some relatively undisturbed, remote locations. Our observed disease
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47 410 prevalence and severity levels are comparable to those documented in other intertidal and
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49 411 subtidal eelgrass meadows in the Northeast Pacific [32], including the San Juan Islands [9],
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51 412 though severity levels are considerably higher than those observed in Sweden [31]. Previous
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53 413 work indicates that in natural meadows, diseased eelgrass can have reduced growth rates and
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3 414 belowground sugar reserves and lesions can rapidly outpace leaf growth [20]. Thus, the surveyed
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5 415 eelgrass meadows with high disease levels could have compromised growth and potentially
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7 416 survival. Against this backdrop of high disease levels, disease risk varied highly across both
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9 417 latitude and tidal zone. Disease prevalence and severity were reduced at cooler sites, the cooler
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11 418 year, and in higher latitudes. This confirms seagrass wasting disease is among the growing
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13 419 number of temperature-sensitive marine diseases [5,10,32].
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17 420 Of the temperature metrics tested in prevalence and severity models, March cold
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19 421 temperature anomalies were the best predictors for summertime disease levels. Regions with
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21 422 cooler temperatures that may either kill or slow the growth of *L. zosterae* could have lower
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23 423 summer disease levels. While most regions experienced cooler temperatures and reduced disease
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25 424 in 2017, the exception was a cooler Puget Sound in 2018, which stood out as reflecting a
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27 425 temperature-disease association. Disease prevalence and severity were markedly lower in Puget
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29 426 Sound that year, coinciding with cooler La Niña conditions—including increased upwelling—
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31 427 that provided more cool, saline water to the area in spring 2018 [84]. This local anomaly in
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33 428 cooler temperatures and lower disease further supports the notion that cooler temperatures
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35 429 suppress disease. In contrast, warmer spring temperatures could allow the pathogen to
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37 430 proliferate, causing disease outbreaks by the summer. Similar associations between June positive
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39 431 temperature anomalies and elevated disease were recently observed in intertidal eelgrass in the
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41 432 Northeast Pacific [9,32]. Based on these findings, spring temperatures could serve as an early
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43 433 indicator for summertime disease outbreaks.
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49 434 Sites spanned environmental and latitudinal gradients and allowed us to measure disease
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51 435 across a broad spatial scale. Our results indicate widespread disease prevalence across all sites,
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53 436 and suggest that sites with severe infections could be at-risk for future declines. Further, they
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3 437 indicate that even remote meadows with minimal human impacts, like Alaska and British
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5 438 Columbia, are at-risk for disease outbreaks. Since high-latitude meadows had lower disease
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7 439 compared to those at lower latitudes—and given that eelgrass ranges are expected to expand
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9 440 northward under climate change scenarios [46]—these northern meadows should be carefully
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11 441 monitored as potential refugia against disease and warm temperatures. A number of factors were
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13 442 confounded with geographic region, including timing of sampling, latitude, and human impacts
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15 443 (ex: coastal development, water quality). While our study design could not partition the variation
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17 444 associated with these factors, these may be important in influencing wasting disease dynamics.
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19 445 For example, coastal urbanization could compromise eelgrass health, since nutrient enrichment
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21 446 from runoff triggers algal blooms and suspended sediments limit light, stressors that caused
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23 447 seagrass loss in an urban Florida estuary [85]. Future studies could target analyses on multiple
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25 448 wasting disease stressors.

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31 449 Across regions and years, disease prevalence and severity were significantly lower in
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33 450 subtidal than intertidal meadows. When averaged across both years, disease severity was nearly
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35 451 three times lower in subtidal meadows, suggesting deeper habitats buffered the effects of
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37 452 environmental stressors and disease. Subtidal eelgrass may be more resilient and thus more
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39 453 resistant to wasting disease compared to intertidal eelgrass, and these deeper meadows could
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41 454 serve as refugia from future disease outbreaks and climate change conditions. This is consistent
42
43 455 with findings that 20 years after mass eelgrass die-offs in the Chausey Archipelago, France,
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45 456 recovery was mostly limited to subtidal meadows [86]. Similar to terrestrial plants in
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47 457 environmental extremes [87], intertidal eelgrass that is exposed to highly variable environmental
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49 458 conditions at low tide—high and low temperatures, salinity, desiccation, UV stress [64]—may be
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51 459 more physiologically stressed and at-risk to infection compared to subtidal meadows, which are
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3 460 not exposed at low tide and may be more disease-resilient. Similarly, deep temperate reefs act as
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5 461 refugia against marine heatwaves for habitat-forming corals, seaweeds, and eelgrass in Virginia
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7 462 and the Northeast Pacific, buffering against the harsh environmental conditions to which
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10 463 organisms at shallower depths are exposed [45,60–62,88].

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12 464 Lower disease occurred in more dense eelgrass meadows and at cooler temperatures,
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14 465 regardless of tidal zone, but this association was more pronounced in intertidal meadows.
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16 466 However, this pattern is contrary to our hypothesis and disease theory, which would predict
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18 467 higher disease levels in high-density meadows, given that one of the mechanisms of seagrass
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20 468 wasting disease transmission is via direct contact between infected and healthy leaves [15].
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22 469 Meadows with low eelgrass densities could have already experienced disease outbreaks or
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24 470 stressful conditions, leaving a reduced number of survivors with high disease prevalence and
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26 471 severity. Given that we observed strong interactions between temperature and density on disease
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28 472 prevalence, patchy meadows are likely more at-risk to synergies between thermal and disease
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30 473 stressors. Recent work corroborates similar findings on the resiliency of deeper eelgrass habitats,
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32 474 as deeper meadows had positive or neutral changes in eelgrass density following a marine
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34 475 heatwave, compared to significant declines in warmer, shallower meadows [45]. As such, high
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36 476 density eelgrass meadows under lower climate stress should be prioritized for conservation.
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42 477 Generally, the mean densities, canopy heights, and leaf areas we observed were
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44 478 comparable to those in other eelgrass meadows in the Northeast Pacific [9,30,32]. The higher
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46 479 densities and reduced canopy heights and leaf areas in the San Juan Islands and Puget Sound,
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48 480 WA meadows are consistent with established differences in eelgrass growth patterns between
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50 481 tidal zones [89]. Intertidal and subtidal densities varied considerably, with orders of magnitude
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52 482 higher densities occurring at some sites compared to others in the same tidal zone. Densities
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3 483 were more consistent in subtidal meadows year to year than intertidal meadows, further
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5 484 supporting our hypothesis that they are more environmentally stable and resilient against
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8 485 environmental disturbances; this is also reflected in lower disease in subtidal meadows. Our
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10 486 findings that leaf area and disease prevalence were significantly, positively associated also aligns
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12 487 with previous findings [9,29,30]. Based on leaf area alone and the usual association between
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14 488 disease and leaf size, subtidal meadows should have more disease, yet subtidal meadows
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16 489 consistently had reduced prevalence and severity. Again, this suggests greater resilience to
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19 490 disease of deeper, natural eelgrass meadows.
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22 491 We specifically designed surveys to determine the association between temperature and
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24 492 disease in natural eelgrass meadows spanning the high biodiversity Northeast Pacific.
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26 493 Temperature is an important driver of historic and current wasting disease outbreaks worldwide
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28 494 [9,13,14,27,32,51]. Our machine-learning algorithm, EeLISA, enabled us to prioritize precise
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30 495 disease measurements and scale up our surveys. Field surveys that span broad, spatiotemporal
31
32 496 scales are essential to tracking and predicting disease outbreaks in a rapidly changing ocean, and
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34
35 497 are needed to inform conservation and management decisions [90–92].
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38 498 Connecting across scales from individuals, tidal zones, sites, and geographic regions, this
39
40 499 large-scale field survey furthers our understanding of seagrass wasting disease dynamics in a
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42 500 changing ocean. Notably, it shows an association between reduced eelgrass disease, cooler
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44 501 temperatures, higher eelgrass densities, and deeper habitats. Our findings underscore a central
45
46 502 need in managing marine resources in a rapidly warming climate: mapping resilient refugia.
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48 503 Surveys also reveal the conservation value of subtidal meadows as climate refugia. Though
49
50 504 largely out of sight, expansive subtidal meadows cover more of Earth's surface area than
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53 505 intertidal meadows and create essential spawning habitat and nursery areas for innumerable fish
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3 506 and other organisms [93]. This new indication of an important refuge from disease significantly
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5 507 increases the value of subtidal meadows, many of which are declining within the Salish Sea
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7 508 [9,68] and globally [39]. While previous field surveys compared wasting disease in eelgrass at
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9 509 different intertidal [9,29] and subtidal zones [31], no prior studies have compared disease
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11 510 between tidal zones. A relatively understudied aspect of wasting disease in eelgrass, these deeper
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13 511 refugia provide important opportunities for future conservation efforts.
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17 512 This new information about lower wasting disease risk in cooler climates, cooler years,
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19 513 and deeper meadows can improve eelgrass management. First, to best inform conservation and
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21 514 preservation of these key habitats under mounting climate stress, continued monitoring of
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23 515 eelgrass meadows is essential, especially to monitor and track temperature-sensitive disease
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25 516 outbreaks. Intertidal meadows are most tractable for disease surveys, since they not only are
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27 517 easier to access from shore, but also have higher levels of disease, are more at risk, and may
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29 518 provide earlier warning of declines. Second, more protections should also be considered for both
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31 519 intertidal and subtidal meadows to buffer against future climate and disease-driven declines,
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33 520 especially in areas prone to more frequent, rapid warming with higher risk for disease outbreaks.
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35 521 Because subtidal meadows have the highest potential as safe havens against environmental and
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37 522 pathogenic stressors, eelgrass conservation activities should focus on protecting subtidal
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39 523 meadows. Given the increasing frequency and intensity of marine heatwaves [36,94] and other
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41 524 mounting environmental changes, understanding the synergistic effects of climate change and
42
43 525 marine diseases on foundation species is critical to the sustainability of our oceans and planet [7].
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49 526
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15 544 of interest to declare.
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20 546 **Data availability statement**

21 547 All data and R scripts used to generate the analyses presented here will be publicly
22 548 available via the Cornell University eCommons Repository by the time of publication
23 549 (<https://doi.org/10.7298/6ybh-w566>).
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796 **Figure Captions**

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798 Figure 1. (A) Locations for seagrass wasting disease surveys in Alaska, British Columbia, San
799 Juan Islands, and Puget Sound in summers 2017 and 2018. Surveys included paired subtidal and
800 intertidal eelgrass meadows. Map made in ArcGIS. (B) Site-level disease severity reflect lower
801 disease in subtidal meadows and generally higher disease in 2018; $n=5761$ blades (mean \pm SE).
802 Sites are arranged north to south, top to bottom within and by regions. Sites with missing bars
803 did not have eelgrass and do not represent that there was not any disease present (intertidal:
804 Triquet N, Choked; subtidal: Hakai).
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807 Figure 2. (A) Correlations between measured March cumulative negative temperature anomaly
808 and measured transect-level disease prevalence in intertidal and subtidal meadows. Bands
809 represent 95% CI. Temperature anomalies are centered and scaled. Also shown are
810 representative eelgrass in (B) intertidal and (C) subtidal meadows. Image E credit: A Hausner.

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812 Figure 3. (A) Cumulative March negative sea surface temperature anomalies in 2017 and 2018.
813 Cooler temperatures in 2018 in Puget Sound (PS) corresponded with lower disease levels that
814 year. (B) Predicted disease prevalence in 2017 and 2018 given observed cumulative March
815 negative temperature anomalies and mean, scaled leaf area. Predictions are based on the leaf-
816 level prevalence model in Table S3. Bands represent 95% confidence intervals. Temperature
817 anomalies are centered and scaled.

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For Review Only

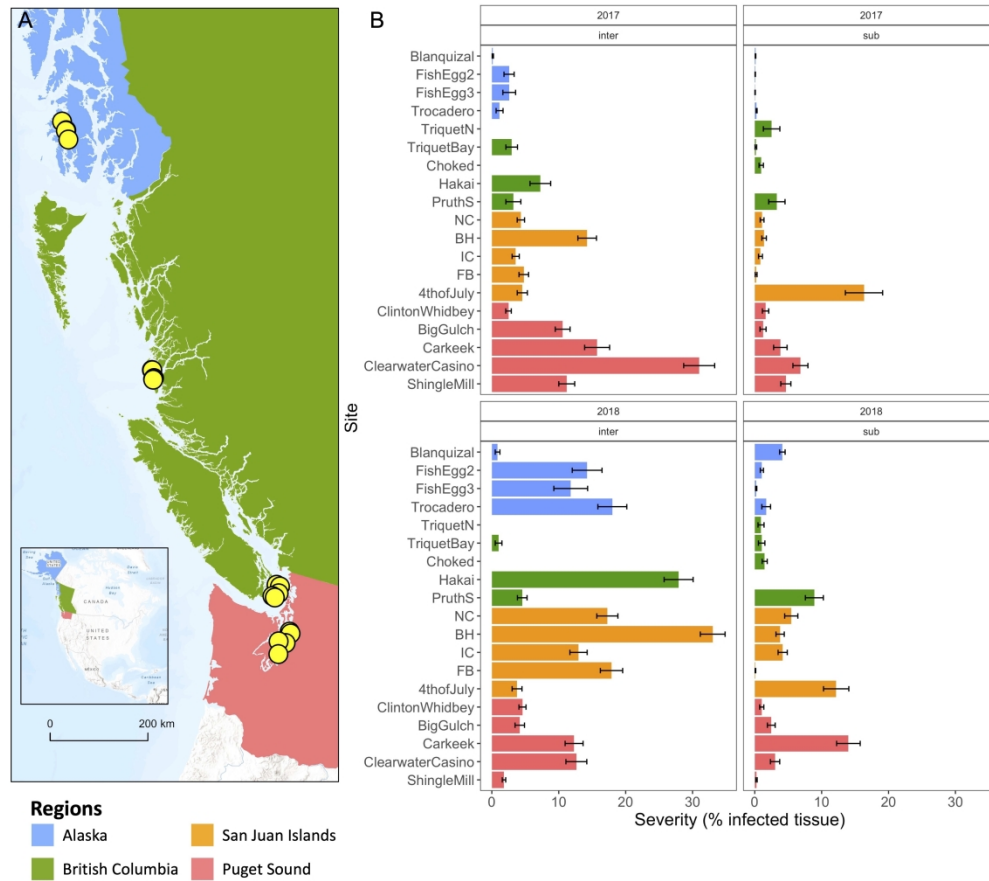


Figure 1. (A) Locations for seagrass wasting disease surveys in Alaska, British Columbia, San Juan Islands, and Puget Sound in summers 2017 and 2018. Surveys included paired subtidal and intertidal eelgrass meadows. Map made in ArcGIS. (B) Site-level disease severity reflect lower disease in subtidal meadows and generally higher disease in 2018; $n=5761$ blades (mean \pm SE). Sites are arranged north to south, top to bottom within and by regions. Sites with missing bars did not have eelgrass and do not represent that there was not any disease present (intertidal: Triquet N, Choked; subtidal: Hakai).

1587x1410mm (72 x 72 DPI)

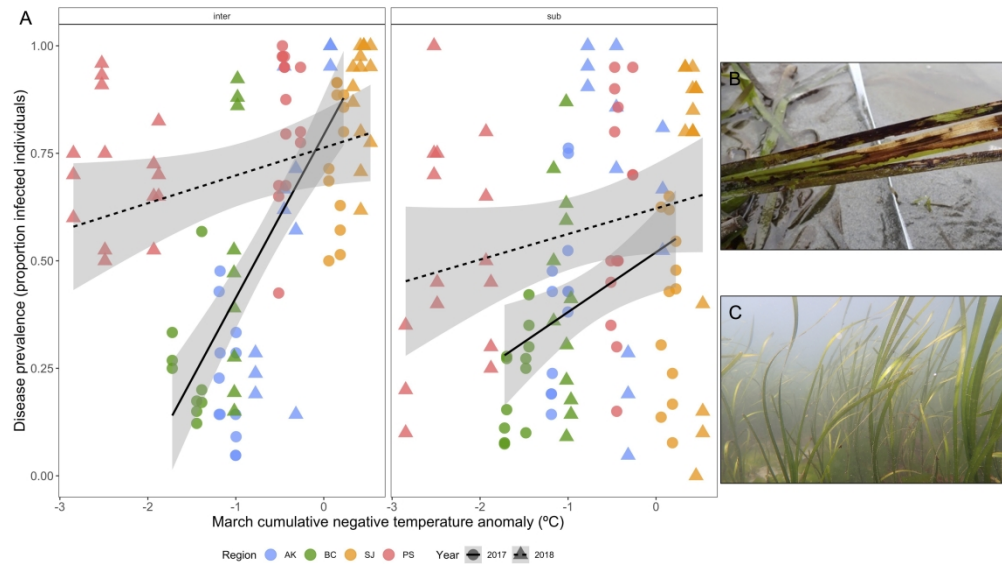


Figure 2. (A) Correlations between measured March cumulative negative temperature anomaly and measured transect-level disease prevalence in intertidal and subtidal meadows. Bands represent 95% CI. Temperature anomalies are centered and scaled. Also shown are representative eelgrass in (B) intertidal and (C) subtidal meadows. Image E credit: A Hausner.

2822x1587mm (72 x 72 DPI)

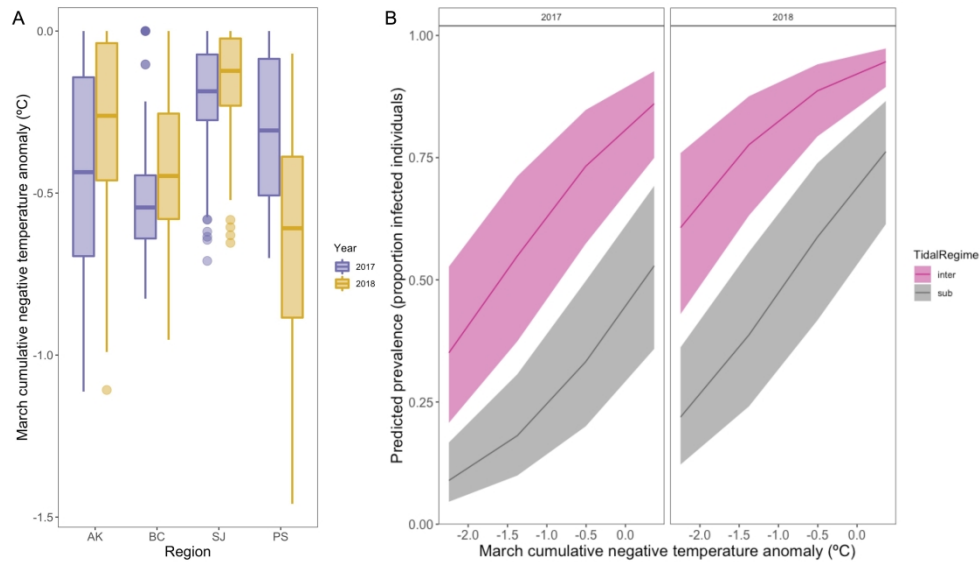


Figure 3. (A) Cumulative March negative sea surface temperature anomalies in 2017 and 2018. Cooler temperatures in 2018 in Puget Sound (PS) corresponded with lower disease levels that year. (B) Predicted disease prevalence in 2017 and 2018 given observed cumulative March negative temperature anomalies and mean, scaled leaf area. Predictions are based on the leaf-level prevalence model in Table S3. Bands represent 95% confidence intervals. Temperature anomalies are centered and scaled.

2822x1587mm (72 x 72 DPI)