1	The effect of aggregation on visibility in open water
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7	Abstract
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9	Aggregation is a common life-history trait in open-water taxa. Qualitative
10	understanding of how aggregation by prey influences their encounter rates with predators is
11	critical for understanding pelagic predator-prey interactions and trophic webs. We extend a
12	recently-developed theory on underwater visibility to predict the consequences of grouping in
13	open water species in terms of increased visual detection of groups by predators. Our model
14	suggests that enhanced visibility will be relatively modest, with maximum detection distance
15	typically only doubling for a 100-fold increase in the number of prey in a group. This result
16	suggests that although larger groups are more easily detected, this cost to aggregation will in
17	many cases be dominated by benefits, especially through risk dilution in situations where
18	predators cannot consume all members of a discovered group. This in turn helps to explain the
19	ubiquity of grouping across a great variety of open-water taxa.
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23	Keywords: grouping, anti-predator defence, shoal, school, dilution of risk, attack abatement
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### Introduction

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28 Aggregation is a dominant feature of the life-histories of many organisms (e.g. flocks of 29 birds, shoals of fish, herds of ungulates, clusters of insect eggs). Of the many selective benefits 30 that group living can confer, those related to reducing predation risk (e.g. through collective 31 vigilance, collective defence, predator confusion, and risk dilution) seem the most ubiquitous (see [1] and [2] for reviews). However, these benefits to group living will be moderated or 32 33 even nullified if predators (and other antagonists – such as parasitoids) can detect groups at 34 greater distances than they can detect single individuals. There is a paucity of current theory 35 and empirical exploration related to the ability of predators to detect groups of prey, and how 36 that might be affected by different traits of that group. This is a significant handicap to 37 understanding the ecological consequences stemming from aggregation as a widespread anti-38 predatory defence. Firstly, aggregation plays a facilitating role in human harvesting of natural 39 populations; with many species targeted only because their tendency to aggregate makes 40 harvesting economically viable [3]. Secondly, as species introductions, range-changes and 41 extinctions alter ecosystems, our ability to predict consequences will critically depend on our 42 understanding of interspecific interactions (especially predation – given its ubiquity). Finally, 43 the detectability of aggregations is also important in ecological contexts other than predator-44 prey or host-parasite interactions. For example pollinators are attracted to larger aggregations 45 of flowers from a greater distance away, and this has been linked to increased detectability (e.g. [4 & 5]). A similar effect has also been reported for attraction of seed-dispersers to 46 47 aggregations of fruits (e.g. [6]). An improved quantitative understanding of how aggregation 48 affects visual detection is vital in these contexts too.

49

### 50 Existing theory on visual detection of groups of targets

51 The most commonly cited theory on the issue of group visibility remains that of Vine 52 [7,8]. Vine [7] considered a tightly packed chain of n individuals, each of width l and height h, 53 which form a straight line of length *nl* and height *h* when viewed from the side. He argued that 54 in terms of visual acuity the important dimension is the minimum one (*h* in this case), and thus the length of the line formed by the *n* individuals was irrelevant to its ease of detection, and 55 56 detection rate would be independent of group size. In Vine [8], he admitted that a body of 57 evidence existed suggesting that humans could more easily detect horizontal lines than points 58 of the same height. He thus revised his argument to suggest that the maximum distance (r) at 59 which a chain of individuals could be detected would take the form

- 60
- $61 r = A \left(\frac{nl}{h}\right)^B$
- 62

63 where A and B were empirically-determined constants. He suggested on the basis of 64 experiments with human observers that *B* seemed to be of the order 0.40-0.45. A consequence 65 of these values is that r initially rises steeply (r doubling as n increases from 1 to 5), but this 66 effect quickly wanes in strength at larger group sizes (with highly reduced differences in r67 above n = 50). Vine pointed out that the consequences of having less packed individuals so that 68 there are gaps in the viewed aggregation were unknown, and this remains true. To this we 69 would add that the contrast of individuals with the background will strongly influence detectability and is unexplored in this theory. Further, the ecological applicability of this work 70 71 is limited, since only a small fraction of natural aggregations involve long chains of 72 individuals.

Turner & Pitcher [9] provided the most influential theoretical work for the overall
effect of aggregation on not just detection but prey capture rates. Their theory explored two
simple alternative assumptions for the effect of group size on detection rates: assuming that the

76 probability per unit time of a group of prey being detected by a nearby predator is either 77 independent of group size (n), or increases linearly with n. The authors argued that these two 78 alternatives likely bracket reality for most natural systems. However, the distribution of real-79 world cases within this very wide bracket remains unclear. Ioannou et al. [10] argued that 80 detectability of a group should increase with the visual angle subtended by the group. 81 However, they did not speculate on how increasing visual angle will translate theoretically into 82 ecologically-relevant measures such as rate of detection. Treisman [11], on the basis of 83 unpublished experiments with humans, argued that the probability of target detection increases 84 linearly with increasing angular area of the target until a critical area is reached, after which 85 further increasing area brings no further improvement.

86

### 87 *Motivation for our work*

88 Modelling vision is considerably more tractable in pelagic environments than other 89 habitats; because the background against which objects are viewed is simple and predictable, 90 and detection range is strongly influenced by well-characterised patterns of absorption and 91 scattering of light. Hence, there have been a number of theoretical predictions of pelagic visual 92 detection (e.g. [12-14] and references therein). However, no previous study has explicitly 93 explored the detection of a group of individuals. Recently Nilsson et al. [15,16] have offered a 94 general theory for the visual detection of objects in this environment. Here we build on that 95 framework, extending it to the situation where an approximately spherical group of targets 96 (such as a tightly packed school of fish called a bait ball) is detected visually by something 97 with a camera eye (e.g. a cetacean or predatory fish) viewing it horizontally.

An understanding of the anti-predatory effectiveness of grouping is particularly
important in the pelagic realm, the largest habitat on the planet. Grouping is a common lifehistory trait in this environment. With no physical structures to offer protection, prey

101	aggregation is an important and common anti-predatory strategy. The consequences of
102	aggregation for rates of discovery by predators are critical for understanding pelagic predator-
103	prey interactions, and trophic webs. In the next section we develop a new theory for the effects
104	of group size and various ecological factors on maximum detection distance in an open-water
105	environment. A key part of this theory is the attenuation of light as it passes through water,
106	which is substantially greater than in air. Thus the application of our theory is currently limited
107	to aquatic systems, although it could be adapted to terrestrial and aerial situations where the
108	background against which prey are viewed is relatively simple (e.g. snowfields, the sky,
109	mudflats).
110	
111	Methods
112	General theory
113	We model detection of a compact spherical school of fish (henceforth called a bait
114	ball). We begin with the following definitions:
115	
116	$N_t$ number of photons collected by the retina in one integration time from the target bait
117	ball at a negligible viewing distance. We assume that the eye employs spatial
118	summation to collect all photons from the ball in one big 'pixel'. This is known as
119	optimal summation [16], which maximizes detection range, and thus provides an upper
120	bound for the effect of aggregation on visibility.
121	
122	$N_b$ number of photons from the background water (over a pixel the same angular size as
123	the bait ball pixel)
124	

- 125  $N_p$  number of photons scattered into the path between the viewer and the bait ball as 126 viewing distance increases. This is typically referred to as 'pathlight'.
- 127

128  $C_0$  inherent Weber contrast of the bait ball against the background water. Given by  $N_t/N_b$  – 129 1. For simplicity, we assume that the bait ball consists of enough individuals that it 130 appears as a solid wall of opaque fish. Thus the contrast of the ball equals the contrast 131 of the individual fish. This contrast attenuates with distance *r* following 132  $C = C_0 e^{(K\cos\theta - c)r}$ , where *c* and *K* are the beam and diffuse attenuation coefficients of the 133 water, and  $\theta$  is the viewing angle of the predator (0° for looking directly upwards, 180° 134 for looking directly downwards) [17].

135

The pelagic light field is approximately monochromatic at viewing angles greater than 48° from vertical (i.e. outside Snell's window) even at relatively shallow depths below the surface, and at all viewing angles at depths greater than approximately 100 m [18]. In these situations, the beam and diffuse attenuation coefficients of the water (c and K) can be considered to be approximately constant and equal to the values at the wavelength of peak penetration (480 nm in this study). In this case, the four terms defined above are related as:

143 
$$N_t = N_b (C_0 + 1) e^{(K\cos\theta - c)r}$$
, and  $N_p = N_b (1 - e^{(K\cos\theta - c)r})$  (1) and (2)

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The first part of (1) is obtained from solving the Weber contrast equation; the second
(exponential) part is obtained from the contrast attenuation equation given above. Equation (2)
is from the pathlight equation for horizontal viewing [17,19]. Now, from [16]:

149 
$$|N_t + N_p - N_b| = R_{\sqrt{N_t + N_p + N_b}}$$
 (3)

151at the maximum sighting distance, where *R* is the reliability coefficient. The photoreceptor152noise term introduced by Nilsson et al. [16] is negligible at the euphotic depths examined in153this study (<200 m) and thus excluded. Substituting (1) and (2) into (3) gives:</td>154155 $N_b | (C_0 + 1) e^{(K\cos\theta - c)r} + (1 - e^{(K\cos\theta - c)r}) - 1 | = R \sqrt{N_b [(C_0 + 1) e^{(K\cos\theta - c)r} + (1 - e^{(K\cos\theta - c)r}) + 1]}$  (4)156157Combining terms gives:158

159 
$$\sqrt{N_b} |C_0| e^{(K\cos\theta - c)r} = R\sqrt{C_0 e^{(K\cos\theta - c)r} + 2}$$
 (5)

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161 As mentioned above,  $C_0 e^{(K\cos\theta-c)r}$  is the apparent contrast of the bait ball at viewing distance r162 and thus is much less than two at the sighting distance unless the light levels are extremely 163 low, so (5) is well approximated as:

164

165 
$$\sqrt{N_b} |C_0| e^{(K\cos\theta - c)r} \cong R\sqrt{2}$$
(6)

166

167 From [16]:

168

169 
$$N_{b} = \frac{\pi}{4} \left(\frac{T}{r}\right)^{2} \left(\frac{\pi A^{2}}{4}\right) q \tau \Delta t \int_{\lambda_{1}}^{\lambda_{2}} \left(1 - e^{-kR(\lambda)t}\right) L_{b}(\lambda) d\lambda$$
(7)

171 where *T* is diameter of the bait ball. For the viewing organism *A*, *q*,  $\tau$ , and  $\Delta t$  are the diameter 172 of the pupil, the quantum efficiency of the photoreceptors, the ocular transmittance, and the 173 integration time of the photoreceptors respectively. The parameters *k* and *l* are the absorption 174 coefficient and the length of the photoreceptors respectively.  $L_b(\lambda)$  is the spectral radiance of 175 the background light and  $R(\lambda)$  is the normalized absorbance spectrum of the photoreceptors. 176 We define:

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178 
$$N_{0} = \left(\frac{\pi A^{2}}{4}\right) q \tau \Delta t \int_{\lambda_{1}}^{\lambda_{2}} \left(1 - e^{-kR(\lambda)l}\right) L_{b}(\lambda) d\lambda$$
(8)

179

180 which is the number of photons absorbed by a pixel that views a region one steradian in 181 angular area. This can be thought of as the product of the sensitivity of the eye and the amount 182 of light available for vision. Since the terms can not be separated, due to the weighted integral, 183 they are considered as one. Substituting equation (8) into equation (5) gives:

184

185 
$$\sqrt{\frac{\pi}{4}} \frac{T}{r} \sqrt{N_0} |C_0| e^{(K\cos\theta - c)r} \cong R\sqrt{2}$$
(9)

186

187 Now *T*, which is the diameter of the spherical bait ball, is related to the number of fish in the 188 target group n via:

189

$$190 T = \sqrt[3]{\frac{6nV_0}{\pi}} (10)$$

191

192 where  $V_0$  is the volume each fish occupies in the bait ball (including the fish and the

193 surrounding water). This volume varies by species and swimming speed, but is approximately

the cube of the body length of the fish for a school larger than 50 individuals [20]. Substituting(10) into (9) and rearranging gives:

197 
$$re^{(c-K\cos\theta)r} \cong \frac{|C_0|}{R} \sqrt{\frac{\pi}{8}} \sqrt{N_0} \sqrt[3]{\frac{6nV_0}{\pi}}$$
(10)

198

199 Gathering the constants and setting the reliability coefficient R to 1.96 (the value for 95%)

200 confidence of detection), we get:

201

202 
$$re^{(c-K\cos\theta)r} \cong 0.4 |C_0| \sqrt{N_0} \sqrt[3]{nV_0}$$
 (11a)

203

which can be solved for *r* as:

205

206 
$$r \cong \frac{W \left[ 0.4 \left( c - K \cos \theta \right) \left| C_0 \right| \sqrt{N_0^3 \sqrt[3]{nV_0}} \right]}{c - K \cos \theta} \text{ or }$$
(11b)

207

208 
$$r \cong \frac{1}{c} W \left( 0.4c \left| C_0 \right| \sqrt{N_0} \sqrt[3]{nV_0} \right)$$
 (11c)

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for horizontal viewing. W(*x*) is the LambertW function (the inverse of  $y = xe^x$ ), which can be calculated using Matlab, Maple, Mathmatica and other computational packages.

212

## 213 Specific example parameter values

214 For the visual system of an Atlantic blue marlin predator (*Makaira nigricans*),

- 215 representative values are: pupil diameter A = 0.019 m, integration time  $\Delta t = 0.017$  s, ocular
- transmittance  $\tau = 0.8$ , and quantum efficiency q = 0.34. The photoreceptors of the marlin have

a peak absorbance at 480 nm, an absorption coefficient of 0.035  $\mu$ m<sup>-1</sup>, and a length of 57  $\mu$ m 218 [21].

219

The background radiance spectra  $(L_b(\lambda))$  were modeled using measured profiles of inherent optical properties and commercial radiative transfer software (HydroLight 5.1, Sequoia Scientific). The ability of radiative transfer theory to accurately model oceanic radiance distributions has been validated by *in situ* measurements of selected radiances and irradiances in multiple studies (e.g. [22][23]). The agreement between modeled and measured spectral radiances is particularly good in oceanic waters, which are easily characterized (reviewed by [18]).

227 Depth profiles of inherent optical properties and chlorophyll-a concentration from 228 tropical oceanic water (approximately Jerlov oceanic type I; [24]) needed for the radiative 229 transfer software were obtained from Drs. Andrew Barnard, Scott Pegau and Ronald Zaneveld 230 (College of Oceanic and Atmospheric Sciences, Oregon State University, Corvallis, Oregon), 231 who collected them using a dual path, multiband absorption and attenuation meter (ac-9, 232 WETLabs) and fluorometer in the Equatorial Pacific (0°0'N 177°21'W). Absorption and 233 beam attenuation coefficients (at 412, 440, 488, 510, 532, 555, 650, and 676 nm) were 234 measured to a depth of 199 m and chlorophyll-a concentration was measured at 1 m intervals 235 to a depth of 110 m (Figure 1a).

Underwater radiance distributions were calculated from 400 to 700 nm at 10 nm intervals and the surface to 200 m depth at 10 m intervals. The sky was assumed to be cloudless, the sea to be calm, and the sun at the zenith. The sky irradiance was calculated using the Radtran model ([25]), and the sky radiance angular distribution was calculated using the semi-empirical model given in [26]. Both models account for atmospheric effects, such as the reddening of the sun as it approaches the horizon, and are well established. Pure water

242	absorption was taken from [27], and the particle scattering phase function was an average-
243	particle phase function based on measurements by Petzold [28]; tabulated values are given by
244	Mobley [18, table 3.10]. Chlorophyll fluorescence was calculated from the measured
245	chlorophyll-a concentration using a modeled phytoplankton absorption spectrum taken from
246	[29] and a fluorescence efficiency of 0.02 that was independent of excitation wavelength.
247	Raman scattering by the water molecules was also included [30]. These values were used to
248	calculate estimates of the number of photons captured per steradian per integration time $(N_o)$ .
249	Figure 1b shows the values for three oceanic predators as a function of depth.
250	
251	
252	Results
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254	In figure 2 we plot maximum horizontal sighting distance $(r)$ as a function of the
255	number of fish in the group $(n, varying over three orders of magnitude from 10 to 10,000)$ and
255 256	number of fish in the group ( $n$ , varying over three orders of magnitude from 10 to 10,000) and depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast
256	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast
256 257	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would
256 257 258	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would be found in a relatively cryptic silvery fish, such as a herring or sardine, a value of 0.5 would
256 257 258 259	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would be found in a relatively cryptic silvery fish, such as a herring or sardine, a value of 0.5 would be for a typical reef fish, and a value of one would be for a black fish. We solve equation (11)
256 257 258 259 260	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would be found in a relatively cryptic silvery fish, such as a herring or sardine, a value of 0.5 would be for a typical reef fish, and a value of one would be for a black fish. We solve equation (11) for the maximum distance ( $r$ ) at which the group of individuals of approximate individual
256 257 258 259 260 261	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would be found in a relatively cryptic silvery fish, such as a herring or sardine, a value of 0.5 would be for a typical reef fish, and a value of one would be for a black fish. We solve equation (11) for the maximum distance ( $r$ ) at which the group of individuals of approximate individual lengths of 10 cm can be detected. Our key results are however qualitatively unchanged for
256 257 258 259 260 261 262	depth (from 0 at the surface to 200 m depth) for three values of the inherent Weber contrast between the fish and the background (0.25, 0.5 and 1). For reference, a contrast of 0.25 would be found in a relatively cryptic silvery fish, such as a herring or sardine, a value of 0.5 would be for a typical reef fish, and a value of one would be for a black fish. We solve equation (11) for the maximum distance ( $r$ ) at which the group of individuals of approximate individual lengths of 10 cm can be detected. Our key results are however qualitatively unchanged for different sized fish. For example, by inspection of equations (9) and (11), we can see

266 Our first key prediction concerns the inherent visual contrast of the prey against the 267 background. It is unsurprising that r increases with increasing inherent contrast ( $C_0$ ) of the 268 prey. What is less obvious is that this effect is non-linear: having a four times greater contrast 269 does not increase sighting distance four-fold. This is because sighting distance is related to a 270 function of the natural logarithm of the contrast (the effect of change in contrast can be seen in 271 greatest detail in Figure 3). Also less obvious is that there is no strong effect of contrast on the 272 shape of the *r*-*n* relationship, and so we would not expect the visibility costs of grouping to be 273 inherently different for prey of different contrasts against the background. By inspection of 274 equation 11, all three terms on the right will be the same in this regard: inherent contrast, the 275 square root of photons (the weighted product of the intensity of illumination and sensitivity of 276 the viewer), and the cube root of the bait ball volume (and so the length of the individual prey 277 fish) all affect the relative sighting distance in a similar way. As a rule of thumb, the number 278 of photons drops by a factor of ten every 70 m in clear oceanic water [18], so the square root 279 drops by a factor of about three. Thus, as an example, cutting contrast to a third of what it was 280 has the same effect as moving the bait ball 70 m deeper or cutting the number of individuals by 281 a factor of 27. This line of arguments may explain why schooling pelagic fish nearly always 282 invest in mirrored scales that reflect much of the incident light to drop their inherent contrast 283 considerably [31].

Our next key result is the effect of depth, with sighting distances being maximised at around 100m, above that the dominant factor is higher attenuation of horizontally travelling light (high *c*) caused by suspended particles (e.g. phytoplankton), and below that the dominant factor is low incident light levels (leading to low  $N_0$ ) caused by attenuation of sunlight as it passes through the surface waters above.

Our primary interest has been in predicting the relationship between maximum sighting
distance (*r*) and group number (*n*). Unsurprisingly, *r* increases with *n* under all circumstances.

291 It is also perhaps unsurprising that the *r*-*n* relationship flattens as *n* increases, but what is less 292 obvious is that (even over the broad range of *n* considered) there is no saturation of the curve. 293 That is, we can still see appreciable increase in r as n changes from 1,000 to 10,000 for 294 example. This is due to optimal summation, which allows the fish to make its visual pixel the 295 same size as the bait ball. Thus, at least over the situations we model, there is no ceiling effect 296 whereby after a group reaches a certain size, further increases in size do not increase the ease 297 of detection of the group. However of most interest is the relatively modest effect of 298 increasing n: under all the circumstances that we explored, increasing the group size by two 299 orders of magnitude (i.e. multiplying *n* by a factor of 100) causes *r* to rise by less than a factor 300 of two. Even this is likely an over-estimate, since not all animals employ optimal summation. 301 By assuming optimal summation we are finding the longest possible sighting range. The 302 relatively modest costs of grouping in terms of increased visual detection may be relatively 303 easily outweighed by benefits through risk dilution, collective vigilance and/or confusion 304 effects, all of which have been demonstrated to increase rapidly with increasing group size (see 305 Discussion and [1]). If maximum sighting distance doubles, then this would suggest that the 306 volume of space over which the prey can be detected increases by a factor of eight. Thus, our 307 model predicts that (as would be expected) a group of 5,000 pelagic prey is more obvious to 308 predators than a group of 50, and this should increase the rate at which the larger group is 309 discovered by predators, but only by a factor of eight or less, the exact number depending on 310 the details of the predator's foraging strategy.

In Figure 3, we show a greater range of values of contrast for three levels of  $N_0$ , corresponding (for our Blue Marlin viewer) to depths of about 50, 100 and 200 m (beam attenuation coefficient *c* is considered to be at a constant value of 0.1 for all three situations). This emphasizes that the inevitable rise in sighting distance with increasing group size can be counteracted by a decrease in the inherent contrast, leading to our prediction that the larger the

characteristic shoal size of fish the stronger the selection pressure should be for morphologicaladaptations (most obviously mirrored scales) that reduce contrast.

318 It is important to note that our visual model allows the minimum contrast threshold of 319 the viewer to go well below levels that have actually been measured. It may very well be that 320 natural predators do achieve these low levels, but so far this has not been demonstrated in large 321 pelagic predators. The lowest threshold measured for fish (and indeed for any animal) is 0.005 322 [32]. Figure 4 shows the sighting distance versus depth and group size using the same 323 procedures as used to generate Figure 2 but with the additional constraint that the contrast 324 threshold of the viewer cannot go below 0.005. As can be seen by comparing Figures 2 and 4, 325 this added constraint does not change any of our qualitative conclusions. Interestingly though, 326 group size does not affect sighting distance at all at shallower depths under this constraint. For 327 the viewer to get any advantage when viewing larger groups at these depths, its contrast 328 threshold would have to be very low indeed. One thing that is obvious from both figures 2 and 329 4 is that water clarity has the biggest influence on sighting distance, because it is the only 330 variable outside the (very slowly increasing) LambertW function. This is why the schools can 331 be seen at greater distance at depth despite it being darker, so long as the water is clearer, an 332 effect commonly experienced by scuba divers as they drop below the murky surface layer to 333 the darker but clearer depths.

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# 336

#### Discussion

The main prediction of our model is that in general a 100-fold increase in the number of individuals in a group will only lead to at most a doubling in the range at which prey are visible to a predator and so (in a worst case where individuals could be detected from all angles) the

larger group might be detected eight times as frequently as the smaller. We now consider theanti-predatory benefits of grouping for comparative purposes.

342 The benefits of risk dilution can sometimes be substantial. If the predator is relatively 343 small in comparison to the prey and not as fleet as the prey, then it may only be able to capture 344 a single individual from a group. In this case, the dilution benefits of being in a group a 345 hundred times larger (and having the risk of being the selected individual reduced by a factor 346 of 100) will far outweigh the eight-fold increase in frequency of encounter of the group with a 347 predator. However at the opposite extreme where the predator is large (or hunts in packs) and 348 fleet compared to the prey, then all of a discovered group may be consumed and there is no 349 dilution benefit to grouping. In general, available empirical evidence (reviewed in [2]) suggests 350 both these extreme situations are commonplace, and ecologically and taxonomically 351 widespread. We can conclude that in cases where a single attack captures only a single 352 individual or small fraction of the prey group, and a predator cannot repeatedly attack a 353 discovered group, then dilution benefits will exceed the visibility costs estimated here.

354 Although predator confusion resulting in a reduced ability to capture prey when faced 355 with larger moving groups has often been demonstrated (see Beauchamp [2] for a review) the 356 effect of group size has rarely been quantified, and current theory does not allow strength of 357 confusion and prey survival to be quantitatively linked [33]. However the confusion effect can 358 be strong, In the most thorough study of the effects of group size, Landeau & Terborgh [34] 359 demonstrated that predatory bass were always successful in quickly capturing a single minnow 360 when both were in an experimental arena together. In contrast, this success rate (for capturing every single minnow) dropped to 11% despite an extended time for interaction when the prey 361 362 group size was increased to 15. Our model suggests that such an effect could again dominate the cost of increased ease of detection of larger groups. 363

364 If we turn to increased vigilance as another anti-predatory benefit of grouping, the most 365 relevant data from the recent extensive review of Beauchamp [2] is that of Kenward [35] on 366 the characteristic distance at which flocks of woodpigeons reacted to apparent attacks by a 367 trained goshawk. Single pigeons reacted on average when the goshawk was only 4m away, this 368 distance increased four-fold for flocks of between two and ten birds and ten-fold for flocks of 369 more than fifty. Interpretation of such data is complicated because there may be a lag between 370 detection of the approaching predator and flight response, but this is likely to be low in this 371 system where predators are much more successful if they can pin prey to the ground, and in 372 any case such a lag is likely to be bigger for large flocks where risk dilution will be substantial. 373 However, as with confusion, it is difficult to quantify the relationship between early predator 374 detection and prey survival. Clearly there is a dearth of data quantifying how vigilance benefits 375 of aggregation change with aggregation size, but given the modest costs of increased detection 376 estimated here there is at the very least no reason to reject the possibility of vigilance benefits 377 outstripping these costs. However, vigilance for predators is particularly beneficial in situations 378 where forewarned prey can flee to a place of safety, and this option is generally not available in 379 pelagic environments.

380 Finally, another factor that mitigates the cost of larger groups being detectable at a 381 greater distance is that, for finite prey populations, increases in group size correspond with a 382 decrease in the total number of groups in the environment. This reduction in the density of 383 groups means at any one time a predator will be a greater distance on average from the nearest 384 group [9,10]. However, evaluation of the consequences of this for predators and prey would require consideration of how such aggregation changed not just average distance from prey but 385 386 also predator activity budgets and search strategies (i.e. in terms of speed and direction of 387 travel during searching). This is an open but tractable problem theoretically, which (in common

with all work on predation) should benefit from a step change in our ability to collect data on
free-living animals through miniaturisation of on-board data-loggers [36]
Finally, we also note that the benefits of remaining in a group often appear to hold across a
broad range of group sizes, and after attack on the group has begun. Observation of bait balls
suggests that the tenancy to aggregate remains even as the ball of fish is whittled away by a
group of predators [37].

394

395 In summary, we have been able to offer an estimate of the likely consequences of 396 grouping in open water species in terms of increased visibility of groups to predators. Our 397 model suggests that such enhanced visibility will be relatively modest, with maximum 398 detection distance only doubling for a 100-fold increase in the number of individuals in the 399 group. This suggests that although larger groups will likely be detected and attacked more 400 often by predators, cost to grouping will in many cases be dominated by benefits through 401 (some or all of) risk dilution, predator confusion, and enhanced collective detection of 402 approaching predators. This helps to explain the ubiquity of grouping across a great variety of 403 open-water taxa – the greatest predation cost to this behaviour is likely to be dominated by 404 expected benefits.

405

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410

411	References
412	[1] Krause J, Ruxton GD 2002 Living in groups. Oxford: Oxford University Press.
413	[2] Beauchamp G 2014 Social predation: how group living benefits predators and prey. New
414	York: Academic Press.
415	[3] Brierley AS, Cox MJ 2015 Fewer but not smaller schools in declining fish and krill
416	populations. Curr, Biol. 25, 75-79
417	[4] Grindeland, JM, Sletvold N, Ims RA 2005 Effects of floral display size and plant density on
418	pollinator visitation rate in a natural population of Digitalis purpurea. Funct. Ecol., 19,
419	383-390.
420	[5] Makino TT, Ohashi K, Sakai. S 2007 How do floral display size and the density of
421	surrounding flowers influence the likelihood of bumble bee revisitation to a plant?
422	Funct. Ecol. 21, 87-95
423	[6] Sargent S 1990 Neighborhood effects on fruit removal by birds: a field experiment with
424	Viburnum dentatum (Caprifoliaceae). Ecology 71, 1289-1298.
425	[7] Vine I 1971 Risk of visual detection and pursuit by a predator and the selective advantage
426	of flocking behaviour. J. theor. Biol. 30, 405-422
427	[8] Vine I 1973. Detection of prey flocks by predators. J. theor. Biol. 40, 207-210.
428	[9] Turner GF, Pitcher TJ 1986 Attack abatement: a model for group protection by combined
429	avoidance and dilution. Am. Nat., 128, 228-240.
430	[10] Ioannou CC, Bartumeus F, Krause J, Ruxton GD 2011 Unified effects of aggregation
431	reveal larger prey groups take longer to find. Proc. R. Soc. Lond. B, 278, 2985-2990.
432	[11] Treisman M 1975. Predation and the evolution of gregariousness. I. Models for
433	concealment and evasion. Anim. Behav. 23, 779-800.
434	[12] Asknes DL, Utne, ACW 1997 A revised model of visual range in fish. Sarsia 82: 137–147

- 435 [13] Johnsen S 2002. Cryptic and conspicuous coloration in the pelagic environment. *Proc. R.*436 *Soc. Lond. B* 269, 243-256.
- 437 [14] Johnsen S, Sosik HM 2003 Cryptic coloration and mirrored sides as camouflage strategies
- 438 in near-surface pelagic habitats: implications for foraging and predator avoidance.
- 439 *Limnol. Ocean.* **48**, 1277-1288.
- [15] Nilsson DE, Warrant EJ, Johnsen S, Hanlon R, Shashar N 2012 A unique advantage for
  giant eyes in giant squid. *Curr. Biol.* 22, 683-688.
- [16] Nilsson DE, Warrant EJ, Johnsen S 2014 Computational visual ecology in the pelagic
  realm. *Phil. Trans. R. Soc. Lond. B* 369, 20140038)
- 444 **[17]** Duntley SQ. 1966 *Principles of underwater lighting*. In Underwater Photo Optics I 1966
- 445 Jun 1 (pp. 1-7). International Society for Optics and Photonics.
- 446 [18] Mobley, C.D., 1994. *Light and water: radiative transfer in natural waters*. Academic
  447 press.
- 448 **[19]** Mertens, L. E. (1970). *In-water photography: theory and practice*. John Wiley & Sons.
- 449 [20] Pitcher TJ, Partridge BL. 1979 Fish school density and volume. *Mar. Biol.* 54, 383-94.
- 450 [21] Fritsches KA, Marshall NJ, Warrant EJ 2003 Retinal specializations in the blue marlin:
- 451 eyes designed for sensitivity to low light levels. *Mar & Freshwater Res* 54, 1-9.
- 452 [22] Mobley CD, Gentili B, Gordon HR, Jin Z, Kattawar GW, Morel A, Reinersman P,
- 453 Stamnes K, Stavn RH. 1993 Comparison of numerical models for computing
  454 underwater light fields. *Applied Optics.* 32, 7484-504.
- 455 [23] Stramska M, Stramski D, Mitchell BG, Mobley CD. 2000 Estimation of the absorption
- 456 and backscattering coefficients from in-water radiometric measurements. *Limnol.* &
- 457 *Ocean.* **45**, 628-41.
- 458 [24] Jerlov NG. 1976 Marine optics. Elsevier

- 459 [25] Gregg WW, Carder KL. 1990 A simple spectral solar irradiance model for cloudless
  460 maritime atmospheres. *Limnol & Ocean.* 35,1657-75.
- [26] Harrison AW, Coombes CA. 1988 An opaque cloud cover model of sky short wavelength
  radiance. *Solar Energy.* 41, 387-92.
- [27] Pope RM, Fry ES. 1997 Absorption spectrum (380–700 nm) of pure water. II. Integrating
  cavity measurements. *Applied optics* 36, 8710-23.
- 465 [28] Petzold H, editor. 1977 *Die neuen Körpertherapien*. Paderborn: Junfermann.
- 466 [29] Prieur L, Sathyendranath S. 1981 An optical classification of coastal and oceanic waters
- 467 based on the specific spectral absorption curves of phytoplankton pigments, dissolved
- 468 organic matter, and other particulate materials. *Limnol. & Ocean.* **26**, 71-89.
- 469 [30] Gordon HR. 1999 Contribution of Raman scattering to water-leaving radiance: a
- 470 reexamination. *Applied Optics*. 38, 3166-74.
- 471 **[31]** Denton EJ 1970. On the organization of reflecting structures in some marine animals.
- 472 Phil. Trans. R. Soc. B 258, 285-313
- 473 [32] Cronin TW, Johnsen S, Marshall NJ, Warrant EJ 2014. *Visual Ecology*. Princeton
- 474 University Press.
- 475 [33] Ward A, Webster M 2016 Sociality: the behaviour of group-living animals. Berlin:
- 476 Springer
- 477 [34] Landeau L, Terborgh J 1986 Oddity and the 'confusion effect' in predation. *Animal*478 *Behav.* 34, 1372-1380.
- 479 [35] Kenward RE 1978 Hawks and doves: factors affecting success and selection in goshawk
  480 attacks on woodpigeons. *J. Anim. Ecol.* 47, 449-460.
- 481 [36] Kays R, Crofoot,MC, Jetz W, Wikelski M. 2015. Terrestrial animal tracking as an eye on life and
  482 planet. *Science*, *348*(6240), p.aaa2478.

- 483 [37] Vaughn-Hirshorn RL, Muzi E, Richardson JL, Fox GJ, Hansen LN, Salley AM, Dudzinski KM,
- 484 Würsig B. 2013 Dolphin underwater bait-balling behaviors in relation to group and prey ball
- 485 sizes. Behavioural processes. **98**,1-8.

### 487 **Figure Captions**

488

489 Figure 1: The optical parameters in the water studied. ) A) The beam attenuation coefficient 490 (at 480 nm) as a function of depth. B) The number of photons (on a log scale) absorbed by the 491 eye of an Atlantic Blue Marlin (*Makaira nigricans*: pupil diameter = 0.019 m) in one 492 integration time if looking horizontally and viewing a full steradian in a sample of clear 493 oceanic water ( $N_0$ ). To show the effect of pupil diameter on  $N_0$ , values for two other pelagic 494 predators – the Atlantic Mackerel (Scomber Scombrus: pupil diameter = 0.0096 m) and the 495 Bluefin Tuna (*Thunnus thynnus*: pupil diameter = 0.036 m) – are also given. All visual 496 parameters other than pupil diameter remain that of the marlin. 497 Figure 2: Sighting distance (in meters) of a spherical bait ball of fish (each being 0.1 m in 498 length) as a function of depth and number of fish in the ball (on a log scale). The left, middle, 499 and right panels correspond to fish with inherent contrasts of 0.25, 0.5 and 1.0 respectively. 500 The complex effect of depth on sighting distance is due to the fact that deeper water is both 501 darker and clearer, which affect sighting distance in opposite ways. 502 Figure 3: Sighting distance (in meters) of a spherical bait ball of fish (each being 0.1 m in 503 length) as a function of both the inherent contrast and number of fish in the ball (on a log 504 scale). The left, middle, and right panels correspond to the number of photons absorbed by a 505 marlin eye ( $N_0$ , see text) at daytime depths in clear oceanic water of approximately 200, 100, 506 and 50 meters respectively (the beam attenuation coefficient  $c = 0.1 \text{ m}^{-1}$  throughout). 507 508 Figure 4: Sighting distance (in meters) of a spherical bait ball of fish (each being 0.1 m in

509 length) as a function of depth and number of fish in the ball (on a log scale). The left, middle,

510 and right panels correspond to fish with inherent contrasts of 0.25, 0.5 and 1.0 respectively. In

- 511 this case (as opposed to the results shown in Figure 2), the minimum contrast threshold of the
- 512 viewer is not allowed to drop below 0.005, which is the lowest value measured in any animal.