Size and transparency influence diel vertical migration patterns in copepods.

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5 1 Abstract

Diel vertical migration (DVM) is a widespread phenomenon in aquatic environments. The
primary hypothesis explaining DVM is the visual predator evasion hypothesis, which suggests
that zooplankton migrate to deeper waters to avoid detection during daylight. However, visual risk also depends on a copepod's morphology. In this study, we investigate hypotheses
related to morphology and DVM: (H1) size increases visual risk and will increase DVM depth
and (H2) copepod transparency will reduce visual risk and thus reduce DVM depth. In-situ
Copepod images were collected across several cruises in the Sargasso Sea using an Underwater
Vision Profiler 5. Copepod morphology was characterized from these images and a dimension
reduction approach. The results show a clear relationship in which larger copepods have a
larger DVM signal. Darker copepods also have a larger DVM signal, however only amongst
the largest group of copepods. This suggests multiple morphological traits influence copepod

2 Scientific Significance Statement

Diel Vertical Migration is a widespread phenomenon across marine and freshwater systems.

The predator evasion hypothesis suggests that DVM occurs as zooplankton attempt to escape visual predators. Yet, DVM itself is a costly and risky behavior. Thus, DVM should only occur when visual risk is high. Several studies have shown that copepod size influences the magnitude of DVM. However, an individual's visual risk may include traits beyond simply size. In this study, we utilize an in-situ imaging tool to reveal how copepod morphological traits influence DVM. Our findings show that both size and transparency influence DVM. This finding highlights that DVM is a complex behavior driven by copepod traits. Furthermore, this study exemplifies the ability of new technology to draw insights into plankton ecology.

28 3 Introduction

Diel vertical migration (DVM) is a wide spread phenomena with large consequences in ocean ecosystems. DVM is the process of pelagic organisms vertically moving in the water column on a daily basis, often travelling dozens to hundreds of meters (Bianchi and Mislan 2016). This large-scale event occurs across many taxa, from plankton to fish (Brierley 2014). However, DVM is particularly notable in zooplankton communities, whose migrations contribute substantially to biogeochemical cycles (Steinberg and Landry 2017; Archibald et al. 2019; Siegel et al. 2023). Zooplankton communities, largely dominated by copepods (Turner 2004), will feed in surface layers of the ocean at night then migrate into deeper waters during day-

- 37 time. Through this movement, copepods actively transport carbon to depth. Additionally,
- 38 Kelly et al. (2019) described zooplankton DVM to be a major component of mesopelagic food
- 39 webs. Thus to understand pelagic food webs and nutrient cycles, it is critically important to
- 40 understand the drivers of DVM.
- 41 Predominantly, zooplankton DVM is the movement from deep waters at daytime to shallower
- waters at night (Hays 2003; Bianchi and Mislan 2016). The leading explanation for this
- pattern is the predator-avoidance hypothesis (Bandara et al. 2021). This hypothesis posits
- 44 zooplankton evacuate the sunlit surface to evade visual predators then ascend at night to
- 45 feed. However, the massive migration undertaken by these copepods is energetically expensive
- 46 (Maas et al. 2018; Robison et al. 2020). Therefore, the visual predator evasion hypothesis
- 47 implies that DVM is a result of visual risk exceeding migration costs. However, a copepod's
- visual risk to a visual predator depends on morphological features (Aksnes and Utne 1997).
- Notably a copepod's size can increase visual detection. Several studies have documented that
- 50 copepod size influences DVM magnitude (Hays et al. 1994; Aarflot et al. 2019). Presumably, a
- 51 copepod's transparency will also influence DVM. Hays et al. (1994) reported that pigmentation
- 52 explained variation in DVM frequency. However, few other studies have investigated this at
- 53 length. One barrier to studying a relationship between copepod morphology and DVM is the
- 54 difficulty of accurately recording traits.
- 55 In-situ imaging tools offer great potential to better describe copepod DVM. By directly ob-
- serving copepods, new insights into their behavior and traits can be resolved (Ohman 2019).
- 57 For example, Whitmore and Ohman (2021) used an in-situ imaging device to describe a rela-

tionship between copepod abundance with a particulate field rather than chlorophyll-a. Such findings are facilitated by the fact imagery data records an individual's exact position. Additionally, a copepod's true appearance can be documented whereas net-collected organisms are often physically deformed or lacking color due to decomposition or preservation. Some studies have noted a copepod DVM with in-situ imagery data (Pan et al. 2018; Whitmore and Ohman 2021). However, direct tests of DVM-related hypotheses with such data have not been conducted.

In this study, we utilized in-situ imaging to evaluate how copepod morphological traits influence patterns. We specifically test the hypotheses that, (H1) size increases visual risk and will increase DVM magnitude and (H2) copepod transparency will reduce visual risk and thus reduce DVM. If these morphologically based hypotheses are true, then the larger and darker copepods will have the largest DVM signals.

70 4 Methods

4.1 CTD profiles and UVP imaging of copepods

Data were collected aboard the R/V Atlantic Explorer in collaboration with the Bermuda Atlantic Time-series Study (BATS) (Steinberg et al. 2001). In-situ images of plankton were acquired using an Underwater Vision Profiler (UVP5) (Picheral et al. 2010). The original sampling methodology and instrument specification followed details described in Barth and Stone (2022). The UVP was attached to the CTD rosette and deployed regularly on cruises

to the Sargasso Sea from June 2019 - December 2021. Typical monthly cruises included ~13 profiles with average descents to 1200m (Supplemental Figure 1). In this study, we investigated general trends in DVM by pooling together casts across multiple cruises. This approach is 79 necessitated by the small sampling volume of the UVP and low abundance of plankton which 80 requires aggregation of data to resolve trends (Barth and Stone 2022). While there was 81 some variation between cruises (Supplemental Figure 2), this oligotrophic system is relatively consistent across seasons. Additionally, every cruise had an approximately equal number of 83 day and night casts. Profiles were assigned to be day or night based on locally calculated nautical dawn and nautical dusk times using the R package suncalc 0.5.1. The UVP records images of large particles (>600 μ m ESD). However, living particles are not reliably identifiable below 0.9 mm (Barth and Stone 2022). All recorded images were processed using Zooprocess (Gorsky et al. 2010), which provides several metrics related to size, grey value, and shape complexity. These features were then used to automatically sort images using 89

using Zooprocess (Gorsky et al. 2010), which provides several metrics related to size, grey value, and shape complexity. These features were then used to automatically sort images using Ecotaxa (Picheral et al.). All images were manually verified by the same trained taxonomist. In total, 294,913 images were recorded. Of these, 85.2% were images of debris or artefacts. The smallest identified copepod was 0.940mm ESD and the largest was 5.904mm ESD. Across all casts, copepods were the most common organism, composing 58.7% of all identified, living particles. In total, there were 4151 individual copepods images.

5 4.2 Morphological Grouping

Zooprocess measures and collects several morphologically relevant parameters. To create relevant groups of copepods, a dimension reduction approach was used. Similar methods have 97 been successfully utilized to provide novel insights to marine snow (Trudnowska et al. 2021), copepod dynamics in the Arctic (Vilgrain et al. 2021), and temporal trends in phytoplankton communities (Sonnet et al. 2022). First, 18 morphologically relevant parameters were selected 100 to be included in a principal Components Analysis (PCA), following (Vilgrain et al. 2021). 101 Parameters can be described as relating to size (e.g. major axis, feret diameter, ESD), grey 102 intensity (e.g. mean grey value at 625nm wavelength light), shape (e.g. elongation, symmetry), 103 and shape complexity (e.g. fractal dimension). The PCA was weighted by the volume sampled 104 in a 1-m depth bin for each observation. This approach provides a correction for the UVP's 105 variable descent speed which can cause duplicate imaging of individuals. While this phenomena 106 has a minor impact on overall results (Barth and Stone 2022), we used the weighted approach 107 to assure that no individual features were overrepresented. All morphological descriptors were 108 scaled and centered prior to inclusion in the analysis. The model was constructed using the R 109 package FactoMineR 2.7. principal components were deemed to be significant if their eigen-110 values were greater than 1. This approach yielded 4 PCs which described 87.3% of the total 111 variation in morphological parameters, with 34.5\% and 26.5\% in the first two components 112 respectively. This four principal component space provides a "morphospace" to characterize 113 copepods. 114

To address our morphology-DVM hypotheses, we constructed discrete morphological groups

based on the first two principal components. Groups along each of the principal components
were defined as low (below 25th percentile), mid (25th-75th percentile) and high (greater than
75th percentile). To address the size-dependent hypothesis (H1), groups were assigned as low,
mid, or high along PC1. Then to assess if color/transparency was a secondary factor (H2),
within each PC1 group, PC2 groups were constructed as low, mid, or high. In total, this
created 9 groups (e.g. Low PC1-Low PC2, Low P1-mid PC2, etc).

4.3 Copepod vertical structure & DVM

4.3.1 Vertical distribution of copepods

Copepods in this system are well documented to undergo DVM (Steinberg et al. 2000; Schnetzer and Steinberg 2002; Maas et al. 2018). However, there have not been direct measurements of DVM with in-situ imaging data. First, to assess which portion of the water column copepods were utilizing for DVM, we visualized the average vertical structure. The concentrations of each morphological group (based on PC1 and PC2) were calculated in 20m depth bins for each UVP profile. These binned-profiles were then averaged together based on time of day.

30 4.3.2 Weighted mean depth variability

Weighted mean depth (WMD) is a common metric to describe vertical structure and DVM in zooplankton (Ohman et al. 2002; Ohman and Romagnan 2016; Aarflot et al. 2019). However, with in-situ imagery, this approach presents a few challenges. WMD cannot be calculated

individually for each profile then averaged because each profile had a different descent depth. Additionally, the small and uneven sampling volume of the UVP can make single casts too 135 variable to reliably resolve abundance. Yet, understanding variation around the WMD is nec-136 essary to compare DVM strength across groups. Here, we introduce a depth-bin constrained 137 bootstrap approach to define WMD with a 95\% confidence interval. To do this, the concentra-138 tion of each group, was calculated in 20m depth bins for each profile. Then all profiles from 139 the same time of day were 'pooled'. This provides a distribution of concentrations in each 140 depth-bin. Traditional bootstrapping randomly samples, with replacement, all observations. 141 With vertically structured data however, full random sampling would bias estimates towards 142 the surface. To avoid this, samples were "bin-constrained" such that for each iteration, a random observation was sampled within each depth bin, then replaced for the next iteration. 144 A maximum depth was set to 600m based on observations of vertical profiles which indicated 145 this to be the maximum point of day/night differences. This approach effectively created a 146 random profile by resampling a concentration, $conc^*$, from each depth bin, d. This profile then 147 was used to calculate a bootstrapped weighted mean depth, WMD^* . This was done for each 148 morphological group q, at each time of day t.

$$WMD_{g,t}^{*} = \sum_{i}^{N=60} \frac{d_{i}(conc_{i,g,t}^{*})}{\sum_{i}^{N=60} conc_{i,g,t}^{*}}$$

The distribution of $WMD_{g,t}^*$ then was used to calculate a bootstrapped mean and 95% confidence interval. The 95% CIs could be compared between times of day and morphological groups to assess the strength of DVM. Using PC1 to assess size, the WMD was compared between the three PC1-groups by percentile level. Then to assess the effect of transparency the WMD was compared between PC2-groups within each PC1-grouping. A larger signal of DVM would be evident by a clearly deeper (non-overlapping 95% CI) daytime WMD.

156 4.4 Data availability

All data and code are made available via (https://github.com/TheAlexBarth/DVM_Migration-

Morphology). All supplemental figures, tables, and analyses are available at (https://thealexbarth.github.io/DVI

5 Results

5.1 Morphological Groups

The PCA revealed four major axis of variability (Figure 1). The first axis (PC1, 34.23% of variability) was largely explained by increasing values related to size, such as perimeter (loading score = 0.927) and feret diameter (loading score = 0.910). The second axis (PC2, 27.24% of variability) can be interpreted as a gradient of transparent to dark individuals. PC2 was largely anticorrelated with mean grey value (higher values indicate a more transparent individual) (loading score = -0.920). As noted in the methods, PC3 and PC4 were both related to the orientation of the copepod and the appendage visibility respectively (Supplemental Figure 3).

 169 The morphological groupings were assigned along PC1 as low, mid and high. Then along PC2,

groups were assigned within each PC1-group (Figure 1). To confirm the morphospace grouping

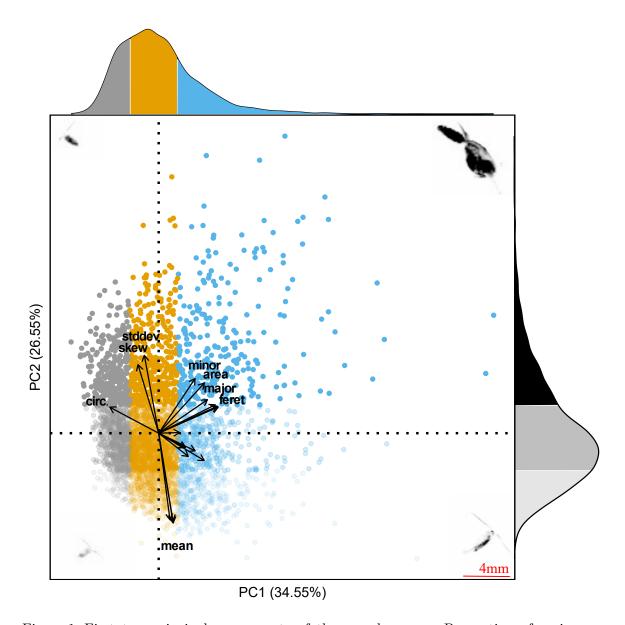


Figure 1: First two principal components of the morphospace. Proportion of variance explained by the two axis is . Each point represents an individual copepod. The color and transparency of each point corresponds to the morphological groups based on pecentile along each axis. Marginal distribution display the proportion of observations in each group. Representative vignettes of copepods are shown in the corners corresponding to their place in the morphospace. 4mm scale bar in the bottom right is shown for the vignettes.

resulted in ecologically relevant categories, the morphological groups were compared against known copepod metrics. Across all PC1-groups, there was a clear difference in feret diameter. The median feret diameter of the low group was 1.97mm. The median feret diameter of the mid and high groups were 2.84mm and 4.83mm, respectively (Figure 2A). All groups were significantly different from one another (Dunn Kruskall-Wallace test, p < 0.001). PC2 groups as a whole were also significantly different from one another (Dunn Krustall-wallace test, p < 0.001). However, within each PC2-group, there was a clear tendency for larger copepods (high PC1 group) to be more transparent (Figure 2B).

179 5.2 Vertical Profiles of Morphological Groups

For all groups, the 20m-binned profiles show a notable structure. While copepods were ob-180 served throughout the mesopelagic (Supplemental Figure 4), the majority of day/night differ-181 ences were observed above 600m (Figure 3). For all morphological groups, there was a peak in 182 nighttime concentration in the lower epipelagic (50m-200m). Similarly, there was a decrease 183 in average daytime concentration over the same region. This pattern is particularly apparent 184 for the groups which are mid and high on both PCs (Figure 3B, C, E, F). Across all groups, 185 both average daytime and nighttime concentration were low in the upper mesopelagic (200m-186 300m). Then, there was a peak in average daytime concentration in the depth bins in the 187 mid-mesopelagic (400m-600m).

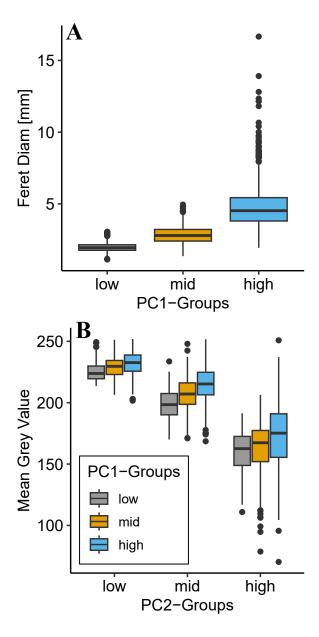


Figure 2: Comparison of morphological groups to relevant parameters. Groups were constructed along principal components with low as below 25th percentile, mid as 25th-50th percentile, and high as above 75th percentile. (A) PC1 groups are significantly different along feret diameter and display a clear trend for size. (B) PC2 groups are significantly different in terms of mean grey value. Note that a low mean grey value indicates a darker copepod.

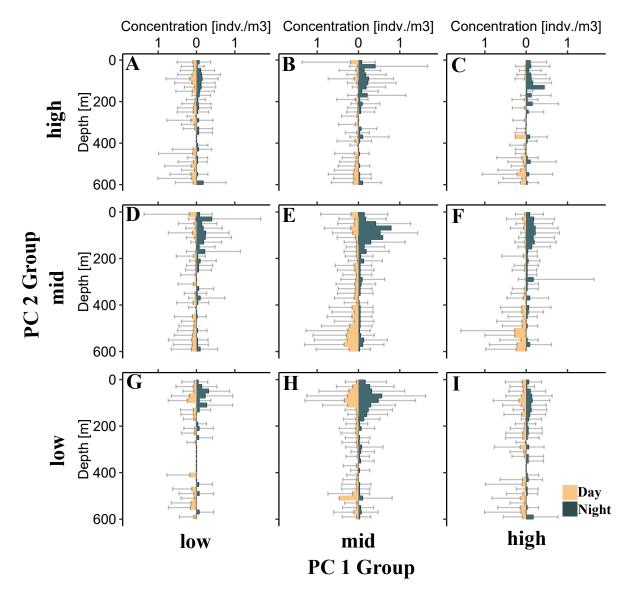


Figure 3: Average vertical profile of different copepod morphological groups. Bars display average concentration in a 20m depth bin. On each panel, left-side bars (tan) correspond to daytime while right-side (teal) bars correspond to nighttime. Standard deviation is shown for each 20m depth bin. Each panel corresponds to a morphological group along PC1 (size axis) and PC2 (transparency axis). (A) low PC1, high PC2; (B) mid PC1, high PC2; (C) high PC1, high PC2; (D) low PC1, mid PC2; (E) mid PC1, mid PC2; (F) high PC1, mid PC2; (G) low PC1, low PC2; (H) mid PC1, low PC2; (I) high PC1, low PC2

89 5.3 Weighted mean depth analysis

The bin-constrained bootstrap approach provided a direct method to compare DVM between 190 groups. Size (PC1) had a clear effect on DVM magnitude. First, for all PC1 groups, daytime 191 WMD 95% bootstrapped confidence intervals (95% CIs) were deeper and non-overlapping 192 with the nighttime 95% CIs (Figure 4). This indicates a clear DVM pattern. However, the 193 differences in day and night CIs varied between morphological groups. All PC1 groups had 194 a similar, overlapping nighttime 95% CI in the lower epipelagic (~145m - ~200m). However, 195 there was a clear difference in the depth of the daytime 95% CIs. The small (low PC1) group 196 had the shallowest 95% CI (235.2m-296.0m). The mid PC1 group's daytime 95% CI was 197 slightly deeper (309.0m-347.3m). The large (high PC1) group daytime 95\% CI was even lower 198 (352.3m-405.0m). 199

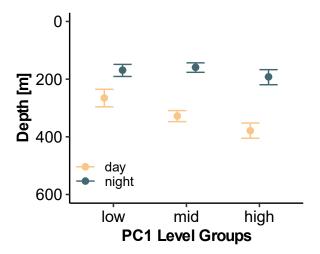


Figure 4: Mean bootstrapped weightd mean depth and 95% confidence intervals for copepods of different morphological groups. Low, mid, and high groups correspond to the different percentiles along PC1 from the morphospace. PC1 largely is explained by size metrics, with higher scores indicating a larger copepod.

When considering the influence of transparency (PC2) on DVM magnitude, we compared PC2 groups within their PC1 grouping. This approach was warranted because of the tendency 201 for size to have a slight effect on transparency (Figure 2). At this level of comparison, there 202 were several notable trends. For the smaller copepods (low PC1), once the data were split 203 into PC2 groups, the wider 95% CIs indicate little to no DVM signal. Generally, the daytime 204 95% CIs and nighttime 95% CIs are overlapping or near-overlapping (Figure 5A). With mid 205 sized copepods, there was a clear DVM signal. However, all PC2 groups appeared to have 206 a similar DVM magnitude with each group's daytime 95% CIs overlapping with each other 207 (Figure 5B). There was a difference in DVM magnitude across PC2 groups within the largest 208 copepods. The more transparent copepods (low PC2 group) showed no DVM signal, with a 209 shallow daytime WMD. However, the darker copepods (mid and high PC2 groups) had deeper 210 daytime WMDs (Figure 5c). 211

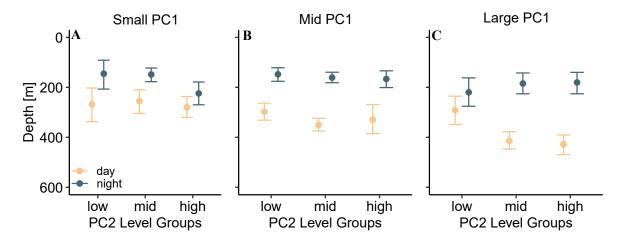


Figure 5: Mean bootstrapped weighted mean depth and 95% confidence intervals shown by copepod morphological groups along PC2 (transparency). Each panel represents a different size group of copepods (PC1 groups).

212 6 Discussion

13 6.1 Copepod morphospace

In this study, we built on methods for describing morphospaces from similar in-situ imaging 214 studies (Vilgrain et al. 2021; Trudnowska et al. 2021; Sonnet et al. 2022). The PCA-defined 215 morphospace with the present data aligns well with the prior applications. Interestingly, the proportion of variation explained by each axis in the morphospace defined on Arctic copepods 217 by Vilgrain et al. (2021) is extremely similar to the morphospace axes in this study. It 218 is possible that this is an artifact of the similarity of input data. Given the UVP has a 219 limited range of observable size classes (Picheral et al. 2010), only copepods above a certain 220 size were fed into both PCAs. Nonetheless, it is striking that the two morphospaces are 221 similar considering the vastly different community compositions between the Arctic ocean and 222 subtropical gyres (Soviadan et al. 2022). 223

224 6.2 Morphology and DVM

The pattern of DVM described in this study is consistent with the general nighttime ascent DVM pattern (Bianchi and Mislan 2016; Bandara et al. 2021). The average vertical profiles display a clear day/night difference (Figure 3). In each 20m depth bin, there was large variation, often exceeding the average concentration. This large variation however, was expected. There can be considerable variation between UVP estimates of zooplankton abundance (Barth and Stone 2022). Additionally, in this study we pooled casts across multiple seasons. Variability

in copepod DVM has been described across seasons (Whitmore and Ohman 2021). While 231 seasonal variability in DVM is in an interesting question in the Sargasso Sea, the nature of our 232 dataset did not lend itself to this investigation. However, despite the need to pool UVP casts 233 across cruises, the signal of DVM was still observable. Previous studies using in-situ imaging 234 have also noted a signal of DVM with copepods (Pan et al. 2018; Whitmore and Ohman 2021). 235 Yet due to small and uneven sampling, it can be a challenge to quantify DVM using in-situ 236 imaging. As presented in this paper, bin-constrained bootstrapping offers a robust method to 237 quantify WMD and investigate DVM hypotheses. 238 Copepod size had a clear effect in which larger copepods migrated further. This finding 239 is consistent with several studies which have documented a size-dependent relationship for 240 copepod DVM (Ohman and Romagnan 2016; Aarflot et al. 2019; Pinti et al. 2019). Ohman and Romagnan (2016) noted that moderate-size copepods had the largest migrations. While 242 this may seem contradictory to the present study, the difference between study systems needs 243 be taken into account. The copepods described in the large (high PC1) group had a mean feret 244 diameter of nearly 5mm. Conversely, in Ohman and Romagnan (2016)'s study the "moderate" 245 copepods ranged from 4mm-6mm. An effect of transparency on copepod DVM was only 246

copepods ranged from 4mm-6mm. An effect of transparency on copepod DVM was only observed in the large copepod group. The large but more transparent copepods (low PC2, high PC1) did not have a detectable DVM signal. Yet the darker copepods (mid and high PC2) had a large DVM signal. Hays (2003) described that copepod pigmentation could explain increased DVM with small (<1mm) copepods. The lack of a transparency effect for the mid- and low PC1 groups in our study is surprising. One possibility is that the small, transparent copepods

were not well sampled by the UVP (Figure 2). Alternatively, some copepods which do not 252 migrate may have pigmentation to avoid damage from UV radiation. The grey-value recorded 253 in UVP-imaged copepods can be indicative of many features beyond simply pigmentation, 254 notably egg-sacs and gut contents (Vilgrain et al. 2021). Such characteristics vary much more 255 between individuals and can have varied influences on DVM (PEARRE Jr. 2003). Thus the 256 relationship between color and DVM is the result of a delicate balance of minimizing multiple 257 ecological and biological risks (Hansson 2004; Hylander et al. 2014). While well documented, 258 predator avoidance may not always be the primary selective pressure on copepod traits. For 259 example, if the costs of migration are too large for some copepods, they will remain near the 260 surface. However, these copepods then are exposed to UV light and may increase pigmentation 261 to reduce damage. Overall, our results reveal a complex dynamic between copepod traits and 262 DVM behavior. Additionally, we show that in-situ imaging systems can be used to investigate 263 ecological hypotheses. 264

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