

Response to reviewers, comments in normal text, **responses in bold**. Line numbers refer to the final version of the manuscript.

Anonymous Referee #1 Received and published: 10 February 2020

5 The manuscript by Ganju et al. uses long-term high frequency measurements to quantify biogeochemical dynamics, coherence and metabolism in a shallow back-barrier estuary with focus on the role of submerged aquatic vegetation (SAV). The experimental design comprises four monitoring stations with water quality probes equipped with sensors for turbidity, chlorophyll-a, fDOM and oxygen. PAR sensors were included at two stations. It is concluded that SAV reduce sediment resuspension and thus Kd and that vegetated sites exhibited higher metabolic rates compared to un-vegetated sites.

10 Unfortunately, these conclusions are not fully supported by the experimental design and results as explained in detail below. This said, the obtained data seem to be of high quality and may be used to provide some insights into the spatial heterogeneity and metabolic characteristics of contrasting sites within an estuary.

Thank you for your positive comments. We have rephrased throughout to indicate that the turbidity-shear stress relationships indicate a seasonal pattern that may be related to bed stabilization by microphytobenthos and increased SAV aboveground biomass in the summer.

20 Specific comments: The experimental design is not well-chosen for testing the scientific hypothesis of the paper (i.e. that spatiotemporal variability/gradients in well-mixed estuaries is driven by the presence of SAV). Although there are two vegetated and two un-vegetated sites, these sites differ with respect to depth, substrate, eutrophication, etc making it impossible to attribute differences between vegetated and un-vegetated sites to SAV alone. The hypothesis/aim of the study should be revised.

We have revised the hypothesis on P1, 11-14 as follows:

25 **“We hypothesize that even in shallow, well-mixed estuaries there is strong spatiotemporal variability in ecosystem metabolism due to benthic and water column properties, and ensuing feedbacks to sediment resuspension, light attenuation, and primary production.”**

Also, the physical/hydro-morphological/biological characteristics incl. hydrodynamic properties of the monitoring sites as well as the rationale behind the choice of the sites should be carefully described.

30 **We have added details on the hydrodynamic climate (tide range and subtidal water level fluctuations), lithology, and organic matter content of the substrate; we have also added the rationale behind site selection, these changes are on P5, 16-27.**

It is not clear how the signal from the sensors were quality assured/rinsed for outliers, off sets, drifting etc. before use. Please describe any post processing procedure of the raw signals. The max values in table 3 don't indicate a “problem” with e.g. artificial spikes (which is somewhat surprising). However, the 0 values of turbidity, chl_a and fDOM seems a bit unrealistic.

35 **The USGS-required data report, cited in the manuscript, contains these details, and we have added relevant details from it on P7, 9-20:**

“Quality control and quality assurance checks were performed on all data and are described in detail in the data report (Suttles et al., 2017). These included removal of obvious spikes in

individual parameter time series using either a recursive filter or median filter technique, in which values that changed from one time point to the next by more than a set threshold were flagged and replaced with a fill value (e.g. -9999). Pre-deployment calibrations and post-deployment check were performed on all EXO2 sensors following YSI procedures outlined in the EXO User Manual (Revision F). Linear corrections were obtained from either post-deployment calibration checks or differences between in situ values from a fouled sensor and from a cleaned, calibrated sensor at the beginning of the next deployment. If data were obviously fouled and corrections were not possible, the data were replaced with fill values. WET Labs ECO-PARSB instruments were checked to verify counts were above their “dark” count (low-count) thresholds; therefore legitimate data collected during night were also discarded. All subsurface pressure data were corrected for changes in atmospheric pressure by using local barometric pressure data, from the meteorological station at CBWS when available, to give a more accurate representation of pressure caused by the overlying water.”

Zero values in certain parameters are common given the multi-point sensor calibrations over large ranges of variability. It is nearly impossible to accurately capture extremely high values and have sensitivity at the low range. There were ice-cover events during our campaign that nearly shut down advection, resuspension, and production. Values of many parameters were zero during this time.

Time series of chlorophyll, with high concentrations during resuspension events suggest domination of suspended microphytobenthos/dead microalgae.

Agreed, we failed to note that where we discussed peaks in chlorophyll during winter resuspension events in the Results section. We have added this on P11, 3-5.

Describe/discuss if microalgae primary production is dominated by pelagic and/or benthic microalgae.

We do not have direct estimates of microalgal production partitioned between sediments and the water column at any of our sites, and have added the following on P14, 17-21 to constrain this uncertainty:

“Though we do not have direct estimates of microalgal production partitioned between sediments and the water column at any of our sites, the sediment surface is nearly covered by SAV and macroalgae at the vegetated sites, therefore we expect that benthic microalgal production is low. At site CB11, K_d^{PAR} values indicate that very little light reaches the bottom, so we presume that water-column microalgae are dominant. For site CB06, we cannot constrain the partitioning between the two environments.”

It is concluded that the presence of SAV controls the shear stress-resuspension relation (P20, L 3). However, as the physical characteristics of especially the shallow sites suggest that microphytobenthos could be abundant, this would significantly influence the shear stress-resuspension relationship as well. Hence, SAV may not be the only/main explanation for the observed seasonal trend in shear stress-resuspension relation.

Agreed, the most general explanation is that depth controls the availability of light for both SAV and benthic algae; and the presence of both to some degree stabilize the bed in summer more than winter. At the deep sites, no such relationship exists, indicating no change in benthic characteristics. However, it is likely that in a vegetated sandy substrate, the seagrass would dominate bed stabilization over benthic microalgae given shading and coarser sediment. In fact, the data of Ellis et

al. (2015; cited now in the manuscript) indicate organic matter percentages of less than 2% on the sandy shoals. However, we have indicated that benthic algae may contribute to bed stabilization in the summer, and likely accounts for chlorophyll resuspension signals in winter.

5 Ellis, A.M., Marot, M.E., Wheaton, C.J., Bernier, J.C., and Smith, C.G., 2015, A seasonal comparison of surface sediment characteristics in Chincoteague Bay, Maryland and Virginia, USA: U.S. Geological Survey Open-File Report 2015-1219, <http://dx.doi.org/10.3133/ofr20151219>.

10 Results of significance test/SD would be highly appreciated, when presenting the results for the four sites (e.g. table 3 and text). Fx. it is states that mean (over what time period?) of chl_a concentration at station CB11 is twice as high as for the other stations (P10, L11-13). It would be nice to know that the observed difference is significant.

15 **We have tested significance for all parameters by comparing overlap in means and standard deviations. Based on overlap of the means and standard deviations, the only statistically significant difference between means is for mean production (P_g) at sites CB03 and CB06, and fDOM at site CB10 and CB11 (relative to site CB06). We have added these details in the text and table captions. However in a tidal system, differences between means are rarely significant compared to the tidally varying values which are substantially different. We have reworded where necessary to stress whether we are discussing peak, mean, or tidal-timescale values.**

20 More importantly, it is concluded that vegetated sites exhibited higher metabolic rates compared to un-vegetated sites (e.g. P20, L 6). This conclusion is not supported by the results in Table 1 where the apparent difference in P_g and R_f does not seem to be significant between site CB3, CB10 and CB11. Furthermore, P_n for all stations seems to be statistically similar.

25 **Again, in line with the comment above, the diurnal variability in P_g and R_f are higher at sites CB3, CB10 and CB11 – which is what we refer to with the term “metabolic rates” – though the means are not statistically different (except for between site CB03 and site CB06). We also state in the text that CB11 is similar to CB3 and CB10 because CB11 is a highly eutrophic site with high chlorophyll-a and water-column phytoplankton production. In terms of mean values, the only statistically significant difference is between sites CB03 and CB06, in terms of P_g . We have noted that in the caption and text.**

30 Direct PAR measurements (and derived K_d values) are only performed at the two shallow, vegetated sites. Although it is difficult to assess the quality/performance of the model from fig. 2 and 3, it seems reasonable that the model can be used to close the gaps in the time series at these stations. However, since light attenuation is dependent on e.g. particle size it seems unjustified to apply the model for stations where it is not calibrated/validated especially since the sediment at the different stations seems to differ in particle size and quality (mud vs sandy sediments and organics enrichment).

35 **The light model assessment for sites CB03 and CB10 is shown in supplementary Figure S2. Nonetheless, to test the sensitivity, we ran the light model with variation in the particle backscatter ratio (b_{bx}) which is particle size dependent, and added the results on P12, 14-22:**

40 **“Sediments at site CB06 and CB11 tend to be finer (Ellis et al., 2015) and may be more susceptible to flocculation, which would induce error in the light model. Therefore we ran the light model with variation in the maximum particle backscatter ratio (b_{bx}) which is particle size dependent. An increase in flocculated suspended-sediment would increase this ratio, thereby**

5 increasing the light attenuation and causing a larger difference in light attenuation between the sandy, vegetated sites and the siltier, unvegetated sites. Modifying the peak value as described above from 0.017 to 0.025 (47% increase; 0.025 value was used for a different mud-dominated estuary by Ganju et al., 2014) increased the median light attenuation at CB06 by 17% (from 1.35 to 1.58 m⁻¹) and at CB11 by 11% (from 1.67 to 1.86). Therefore the spatial variability in light attenuation is likely robust and not confounded by variability in particle size.”

10 Determining the optimal sampling frequency is a science in itself and the results regarding the influence of sampling frequency (table 3) is interesting, but as presented it seems as an unfinished story that is not properly treated in the result and discussion section. This part should be either up- or down scaled/skipped.

15 We believe that the concept that coarse temporal sampling masks spatial variability is an important one given the common daily sampling approach for dissolved oxygen in many impaired estuaries. We have expanded on this concept with examples from the literature in our revision, on P16, 19-23:

20 “Many estuary sampling programs, including “citizen science” efforts, often collect one daily sample of water quality parameters, including dissolved oxygen (Rheuban et al., 2016). Summers et al. (1997) demonstrated that month-long records of dissolved oxygen in a number of estuarine systems, resampled to replicate various non-continuous sampling programs, were not able to identify oxygen minima. This study expands on that finding by mimicking sampling programs within one large estuary, over an entire year.”

The present study do not examine gradients, so this term should be avoided (e.g. P1, L 13). Use “patchiness” or “spatial variability” instead.

25 **Agreed, we have revised throughout to “variability”.**

Technical correction: Table 3: State which time period (yearly, season, other?), min, max and mean is covering. Include SD for the mean values. Figure 6: add x-axis titles

We have revised to add time period to the caption, as well as standard deviation for the original data. We have added the x-axis (1/frequency) to Figure 6. Thank you for catching it.

Mario Hoppema (Referee) mario.hoppema@awi.de Received and published: 28 February 2020

30 The manuscript presents an impressive data set for four locations in an estuary, including several useful variables/parameters. The data interpretation reveals some interesting results and conclusions, which are definitely worth publishing in Ocean Science.

We appreciate the support and the willingness of the Editor to provide a comprehensive review.

However, in some cases there seems to be over-interpretation, for example the difference in metabolic rates between vegetated and unvegetated sites.

35 **In addition to our response to Reviewer #1 above, we would indicate that the daily rates are markedly higher between CB3 and CB10 compare with CB6 – which is the main distinction we discussed between vegetated and unvegetated. We indicated in the revision that for mean period values, only the gross production (P_g) values between site CB03 and CB06 are statistically different.**

Also the data processing should be explained in somewhat more detail.

See response above, we have added detail on the data processing on P7, 9-20.

5 For the net ecosystem metabolism, sometimes NEM is used, for example in Figure 11, and sometimes Pn, for example in Table 1. Please use only one single symbol or abbreviation throughout the manuscript.

We have corrected to “NEM” throughout.

P6, L3 please define RBR D|Wave

Revised to “RBR Virtuoso D|Wave pressure recorder”.

10 P6, L7 please define ADCP

Revised to “acoustic Doppler current profiler (ADCP)”.

P9, L2 “across sites across habitats” Modify piece of text?

Revised to “across sites and across habitats”.

15 P10, L15 I would recommend not to use NPQ as abbreviation, as it is only used twice; the reader possibly has to search for it here.

Revised to “non-photochemical quenching”.

P10, L19-20 “with the highest values consistently observed at CB11” I think one cannot state that, because the record at CB11 is far from complete; certainly the word “consistently” is misleading here.

Revised to “highest peak values over the overlapping period of record”.

20 P10, L20 “with lowest fDOM in the winter, possibly due to reduced biological activity” This does not sound like a good explanation. fDOM is not high during maxima of chlorophyll, which seems to indicate that it is not produced by biological activity. Moreover, the highest values seem to occur at site CB11, which receives most riverine freshwater.

25 **This was meant to suggest that reduced activity in the surrounding watershed and water bodies where DOM is produced, and then exported to the estuary with freshwater. We have clarified on P11, 11-13:**

“Concentrations were similar between the other sites, with lowest fDOM in the winter, possibly due to reduced dissolved organic material production in the surrounding watershed, and ultimately in freshwater runoff to the estuary.”

30 P11, L14-16 “Light attenuation at site CB11, with its proximity to freshwater and nutrient sources, was highest overall and more highly influenced by chlorophyll-a and fDOM than at other sites.” This is hard to believe when inspecting Figures 2-5. Moreover, there is much less data available for this site than for the others.

This is a fair comment, but the mean KdPAR is influenced strongly by the more frequent background conditions than the episodic values, which is why the mean ends up being larger at CB11, mostly caused by a higher background fDOM (Fig. 5). The partitioning of light attenuation between constituents is subject to error of the light model, but given that fDOM and chlorophyll-a are highest at that site (in terms of mean and peak values), it is expected that light attenuation would be highest there (despite lower turbidity than CB06). We have revised to indicate that this is only true during periods of overlap.

P12, L13-14 "between May and October" My view of the figures says that this should be "between May and September".

Revised.

P12, L14 "during November to April" I was say "during November to March"

Revised.

P12, L1920 "but instances of net autotrophy ($P_g > R_t$) occurred nearly 70% of the time at the vegetated sites" When I view Figure 11, neither autotrophy nor heterotrophy are statistically significant. This should be mentioned here.

It is a fair comment to suggest that we did not include statistical analyses to "prove" the existence of net autotrophy, but rather expressed the frequency of daily net autotrophic conditions. We have mentioned this on P13, L21-22:

"Gross primary production and respiration were largely balanced across all sites, but the frequency of daily net autotrophy ($P_g > R_t$) was higher at the vegetated sites."

P13, L10 at 1-7 day

Revised.

P13, L23-24 "The peak in spectral density was 30% higher at CB03 than CB10" This is really hard to see, if at all, in Fig. 6. Possibly the authors could add a comment to Figure 6 (Diss Oxygen) that the curve of CB3 lies under the one of CB10.

We have added a comment in the caption.

P17, L26 where the canonical C:N (I suggest to add canonical, since this is the well known Redfield ratio based on data from many locations)

Revised.

P19, L1-2 "In fact, modest net autotrophy prevailed during the summer season at vegetated sites but not at un-vegetated sites." This does not follow from the data in Figure 11, where NEM is around zero all through the year. See also comment above. Actually I am surprised by the uncertainty interval of NEM, which should have been formed from subtracting two larger terms. Because of this, I would expect it to be clearly larger and not smaller than the uncertainty of both of the terms.

If you compare NEM to P_g and R_t on an absolute basis, we agree that NEM is near zero. But our statement simply refers to the mean monthly value of NEM, which although much smaller than P_g and R_t , is still greater than zero at the vegetated sites during some seasons, which we define as

autotrophic. The error bars around NEM represent the standard deviation of the monthly mean estimates of NEM, which the reviewer is correct in that they are derived as the difference between P_g and R_r . But these difference calculations are performed on a daily basis, so the error bars represent the total variation of all ~30 daily P_g , R_r , and NEM rates and thus do not represent uncertainty carried over from each individual NEM calculation.

References

P22, L4 The usual abbreviation is: Limnol. Oceanogr.

P22, L7 What is this? Ref. No. [UMCES]CBL 04-105a? Report?

P22, L23 The usual abbreviation is: Estuar. Coast. Shelf Sci.,

10 P23, L16 change: Fremantle

P23, L22 The usual abbreviation is: Estuar. Coast. Shelf Sci.,

P24, L5 The usual abbreviation is: Estuar. Coast. Shelf Sci.,

P24, L16 If the journal consists of one word, the full name should be given: Biogeosciences

P25, L8 Modify journal name: Mar. Ecol. Prog. Ser.

15 P25, L16 Add pages and doi: S3-S16. doi:10.1890/05-0800.1

P25, L23 The usual abbreviation is: Limnol. Oceanogr.

P26, L7 The usual abbreviation is: Limnol. Oceanogr.

P26, L16 Add volume and pages: Mar. Geol., 404, 1-14

P26, L18 The usual abbreviation is: Limnol. Oceanogr.

20 P26, L22 Use full journal name: Biogeosciences

P27, L2 The usual abbreviation is: Estuar. Coast. Shelf Sci.,

P27, L12 Please give more info on this publication; report, website?

P27, L14 dito

P27, L16 The usual abbreviation is: Limnol. Oceanogr.

25 P27, L18 Ocean Mod.

P27, L20 Please add page numbers

We have corrected all reference errors.

Table 1 Please spell out standard deviation

Revised.

30 Figure 1 Please also define ADCP and CBWS .

Revised.

Figure 6 I think CB10 in red under the figure panels must be CB11 (unvegetated) Please indicate panels (a) – (d), and then also in the caption. Please explain the x-axis; in the main text the authors talk about “peak at 12.4 h”, but this is not reflected in the figure.

5 **Revised, the 12.4 peak, at ~ 0.5 days, is now noted in the caption.**

Figure 7 It is not clear to me how the oxygen saturation could be 130% in the middle of winter (e.g. at CB11).

10 **The periods where oxygen saturation exceed 130% do not truly occur in winter, which we would define as mid-December to mid- March. The exception to this is during a brief period in February and early March at CB 11, where chlorophyll-a exceeded 25 ug/L indicating that phytoplankton biomass was high and presumably, primary production rates.**

Figure 9 Please indicate panels (a) – (d), and then also in the caption. Please explain what the bottom structures in the panels mean, and why the data are shown despite these structures.

15 **Revised; the shaded areas are zones of the parameter space that are not statistically robust. We have clarified this in the caption.**

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Spatiotemporal variability of light attenuation and net ecosystem metabolism in a back-barrier estuary

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Abstract. Quantifying system-wide biogeochemical dynamics and ecosystem metabolism in estuaries is often attempted using a long-term continuous record at a single site, or short-term records at multiple sites due to sampling limitations that preclude long-term monitoring ~~at multiple sites~~. However, differences in the dominant primary producer at a given location (e.g., phytoplankton versus ~~benthic producers-submerged-aquatic-vegetation; SAV~~) control diel variations in dissolved oxygen and associated ecosystem metabolism, and may confound metabolic estimates that do not account for this variability. We hypothesize that even in shallow, well-mixed estuaries there ~~is~~ are strong spatiotemporal ~~variability gradients~~ in ecosystem metabolism due to ~~benthic and water column properties, the influence of submerged aquatic vegetation (SAV), and ensuing feedbacks to sediment resuspension,~~ light attenuation, and primary production. We tested this hypothesis by measuring hydrodynamic properties, biogeochemical variables (fluorescent dissolved organic matter (fDOM), turbidity, chlorophyll-a fluorescence, dissolved oxygen), and photosynthetically active radiation (PAR) over one year at 15 min intervals at paired channel (unvegetated) and shoal (vegetated ~~by eelgrass~~) sites in Chincoteague Bay, Maryland/Virginia, USA, a shallow back-barrier estuary. Light attenuation (K_{dPAR}) at all sites was dominated by turbidity from suspended sediment, with lower contributions from fDOM and chlorophyll-a. However, there was significant seasonal variability in the resuspension-shear stress relationship on the vegetated shoals, but not in adjacent unvegetated channels. This indicated that K_{dPAR} on the shoals was mediated by ~~submerged aquatic vegetation (SAV) SAV and possibly microphytobenthos~~ presence in the summer, which reduced resuspension and therefore K_{dPAR} . We also found that gross primary production (P_g) and K_{dPAR} were significantly negatively correlated on the shoals and uncorrelated in the channels, indicating that P_g over the vegetated shoals is controlled by a feedback loop between ~~benthic stabilization by SAV and/or SAV- microphytobenthos presence,~~ sediment resuspension, and light availability.

Metabolic estimates indicated substantial differences in net ecosystem metabolism between vegetated and unvegetated sites, with the former tending towards net autotrophy in the summer. Ongoing trends of SAV loss in this and other back-barrier estuaries suggests that these systems may also shift towards net heterotrophy, reducing their effectiveness as long-term carbon sinks. With regards to temporal variability, we found that varying sampling frequency between 15 min and 1 d resulted in comparable mean values of biogeochemical variables, but extreme values were missed by daily sampling. In fact, daily re-sampling minimized the variability between sites and falsely suggested spatial homogeneity in biogeochemistry, emphasizing the need for high-frequency sampling. This study confirms that properly quantifying ecosystem metabolism and associated biogeochemical variability requires characterization of the diverse estuarine environments, even in well-mixed systems, and demonstrates the deficiencies introduced by infrequent sampling on the interpretation of spatial [variability gradients](#).

1 Introduction

Back-barrier estuaries are biologically productive environments that provide numerous ecological, recreational, and economic benefits. Submerged aquatic vegetation (SAV) proliferates in these environments due to relatively shallow bathymetry and sufficient light availability, providing habitat for many fish and crustaceans (Heck and Orth 1980) as well as enhancing wave attenuation (Nowacki et al. 2017). Primary production in back-barrier estuaries and similar shallow marine ecosystems is relatively high given the shallow bathymetry, benthic light availability, and sometimes large SAV beds (e.g., Duarte and Chiscano 1999). Benthic communities within shallow ecosystems host other primary producers where SAV are absent, including microphytobenthos (Sundbäck et al. 2000) and various forms of macroalgae, and the relative contribution of these producers is altered by nutrient enrichment (e.g., McGlathery 2001, Valiela et al. 1997). In deeper, unvegetated habitats, phytoplankton may also contribute significantly to primary production, where the balance between water column and benthic primary production is dependent on depth, light availability, and nutrient levels, but it is unclear if total ecosystem primary production is affected by these factors (Borum and Sand-Jensen 1996).

A fundamental control on estuarine primary production is light availability, which is affected by bathymetry for benthic primary producers, but is also a function of other spatiotemporally dynamic variables (e.g., nutrient availability, sediment type). Models of light attenuation consider the role of suspended sediment, phytoplankton, and

colored dissolved organic matter (CDOM) concentrations in the water column, either through empirical formulations (Xu et al. 2005) or detailed models of scattering and absorption properties (Gallegos et al. 1990). Suspended-sediment concentrations are controlled by processes that vary on a variety of time scales (minutes, weeks, months), including bed composition, bed shear stress, resuspension, and advective inputs of sediment from external sources. In contrast, phytoplankton concentrations are a function of light and water column nutrients, while CDOM is driven by the input of terrestrial material through freshwater loading. The relative contributions of these constituents to light attenuation is dependent on local conditions, can vary spatially and seasonally based on external forcings (wind-induced resuspension, external inputs), and thus requires high-frequency measurements over space and time to measure all aspects of variability.

10 A wealth of literature describes the relationship between light and photosynthesis for marine photoautotrophs, including for SAV the role of self-shading, overall water-column conditions, and light attenuation (Sand-Jensen et al. 2006, Kemp et al. 2004, Duarte 1991). Light-photosynthesis interactions for SAV and phytoplankton can differ substantially, given that phytoplankton are vulnerable to water-column structure and mixing while SAV are rooted to sediments, ~~SAV are~~ vulnerable to epiphytic fouling under nutrient-enriched conditions (Neckles et al. 1993), and 15 ~~SAV~~ can engineer their own light environment through attenuation of flow velocity, wave energy, and therefore bed shear stress and resuspension (Hansen and Reidenbach 2013). Multiple studies have revealed self-reinforcing feedbacks within SAV beds, where SAV shoots and roots stabilize the sediment bed to reduce sediment resuspension, increasing the local net deposition of water-column particulates and improving local light conditions (e.g., Gurbisz et al. 2017). These feedbacks are complex, however, and depend on bed size, aboveground biomass, 20 and other factors (e.g., Adams et al. 2016). Consequently, healthy SAV beds are likely to generate high rates of primary production relative to adjacent unvegetated areas, but each habitat may respond differently to tidal, diurnal, and event-scale variations in physical forcing.

The spatiotemporal variability of light attenuation and primary production in back-barrier estuaries over seasonal timescales is not well-constrained due to large ~~variability~~ gradients in benthic habitats, nutrient and carbon 25 concentrations, and circulation. Nonetheless, numerous net ecosystem metabolism (NEM) estimates have been made with limited spatiotemporal information due to the inherent difficulty of continuous measurements at multiple locations. For example, Caffrey (2004) synthesized NEM across numerous estuaries, however some estuaries were

represented by 1-2 sampling locations within the system. Conversely, Howarth et al. (2014) quantified NEM at multiple locations in a small, eutrophic estuary, but sampling was limited to ~ 100 d over 7 years, thereby adding uncertainty to annual rates. Indeed, Staehr et al. (2012) suggest that undersampling and variability are both large sources of uncertainty in NEM estimates that are not well-understood.

5 We hypothesize that the type of benthic habitat and biophysical environment control spatial variations in primary production via their influence on light ~~availability via local~~ attenuation of light by resuspended particles and water-column phytoplankton biomass. Consequently, the quantification of spatial variability in ecosystem metabolism and subsequent ‘up-scaling’ to system-wide rates is influenced by sampling frequency. Therefore, the objective of this study is to quantify sub-hourly variations in light attenuation, water-column properties, and net ecosystem
10 metabolism and examine the relationship among these properties across habitats in a back-barrier estuary, Chincoteague Bay (Maryland/Virginia, USA) using a year-long deployment of high-frequency sensors. We first describe the observational campaign and analytical methods, followed by analysis of the light attenuation and associated forcing mechanisms. We then quantify gross primary production, respiration, and net ecosystem metabolism, which have not been studied comprehensively in this estuary, and relate it with the variability of light
15 attenuation across different habitats. Given the large spatial variability in bed sediment type, bathymetry, and dominant vegetation, we aim to quantify how temporal variations in light-attenuating substances, wave dynamics, dissolved oxygen, and metabolic rates are linked across these spatially distinct environments. Our conclusions highlight the importance of quantifying spatiotemporal variability in these processes, which indicate feedbacks between physical and ecological processes in estuarine environments that should be considered when evaluating
20 future ecosystem response.

2 Methods

2.1 Site description

25 Chincoteague Bay, a back-barrier estuary on the Maryland/Virginia Atlantic coast (Fig. 1), spans 60 km from Ocean City Inlet at the north to Chincoteague Inlet in the south. A relatively undeveloped barrier island separates the

estuary from the Atlantic Ocean. The mean depth is 1.6 m, with depths exceeding 5 m in the inlets. The central basin depths are approximately 3 m, and the eastern, back-barrier side of the bay is characterized by shallower vegetated shoals; the western side is deeper with no shoals (Fig. 1).

The coastal ocean tide range approaches 1 m, but is attenuated to less than 0.10 m in the center of the bay, where water levels are dominated by wind setup and remote offshore forcing (Pritchard, 1960); subtidal water level fluctuations throughout the bay approach 1 m (Nowacki and Ganju, 2018). River and watershed constituent inputs are minimal with the highest discharge and lowest salinities near Newport Bay in the northwest corner of the bay. Atmospheric forcing is characterized by episodic frontal passages in winter with strong northeast winds; summer and fall exhibit gentler southwest winds. Waves within the bay are predominantly locally generated, with substantial dependence on wind direction and fetch due to the alignment of the estuary along a southwest to northeast axis.

Wazniak et al. (2007) synthesized water quality metrics for Chincoteague Bay, indicating high spatial variability in chlorophyll-a, dissolved oxygen, and nutrient concentrations. Sites in the northern portion of the bay had generally poorer water quality, ostensibly due to higher nutrient loading and poorer flushing; Newport Bay had the lowest water quality index. Fertig et al. (2013) identified terrestrial sources of nutrients to central Chincoteague Bay, though the precise source (anthropogenic vs. naturally occurring) could not be determined.

2.2 Time-series of turbidity, chlorophyll-a, fDOM, dissolved oxygen, and light attenuation

Instrumentation was deployed from 10 August 2014 to 12 July 2015 (Suttles et al., 2017) at four locations in Chincoteague Bay (Fig. 1; Table 1). Instruments were recovered, downloaded, and serviced three times during that period (October 2014, January 2015, and April 2015). The sites were chosen to span a range of water depth, benthic type, and proximity to the coastal ocean; these represent a subset of the eight sites considered by Nowacki and Ganju (2018) in their sediment flux study. Here we include only the sites where light climate and/or water quality were measured comprehensively. Beginning in the southern portion of the estuary, site CB03 is within a seagrass meadow (primarily *Zostera marina*) on the eastern edge of the southern basin, at 1 m depth. Tide range is approximately 0.3 m (Nowacki and Ganju, 2018); the seabed is mainly composed of sand and silt, with organic content < 4% (Ellis et al., 2015). Moving northward, site CB06 is within the main channel north of the nominal boundary between the northern and southern basins, with a ~~silt~~mud-dominated bed, organic content of 6%, and no vegetation, ~~and~~a depth of 3 m, and a slightly diminished tide range of 0.2 m. Site CB10 is within a seagrass

meadow on the eastern side of the northern basin, with a sand-dominated bed and < 4% organic content, at 1 m depth, and tide range of 0.15 m. Lastly, site CB11 is in the central basin of Newport Bay within the northwest portion of the estuary, at a depth of 2.5 m, and tide range of 0.2 m. The seabed at this site is silt-dominated, with the highest observed organic content (>8%). A meteorological station was deployed at site CBWS, approximately 3.8 m above the water surface at Public Landing, Maryland on the central, western shore of Chincoteague Bay. Seagrass presence/absence was confirmed visually during instrument deployments; SAV bed area was obtained from the Virginia Institute of Marine Science Submersed Aquatic Vegetation Monitoring Program, which mapped SAV area from aerial imagery acquired annually in 2014 and 2015 (web.vims.edu/bio/sav/; Orth et al., 2016).

The shallow-water platform described by Ganju et al. (2014) was deployed at sites CB03 and CB10, within sandy patches of the seagrass meadows. The platform was designed to measure hydrodynamic, biogeochemical, and light parameters in the bottom half of a 1 m water column, and consists of an RBR Virtuoso D|Wave pressure recorder (+/- 0.05% accuracy), a pair of WetLabs ECO-PARSB self-wiping photosynthetically active radiation (PAR, 400-700 nm; +/-5% accuracy) sensors; a YSI EXO2 multi-parameter sonde measuring temperature, salinity, turbidity, dissolved oxygen, chlorophyll-a fluorescence, fluorescing dissolved organic matter (fDOM, a proxy for CDOM), pH, and depth (parameter-dependent accuracy available at <https://www.ysi.com/EXO2>); and a Nortek Aquadopp acoustic Doppler current profiler-ADCP measuring water velocity profiles (2 MHz standard at CB10 and 1 MHz high resolution at CB03; +/- 1% accuracy). All instruments except for the upper PAR sensor were mounted at 0.15 meters above the bed (mab) on a weighted fiberglass grate approximately 1 m x 0.5 m. The lower PAR sensor was recessed inside a PVC tube protruding from the bottom of the frame. The upper PAR sensor was mounted at 0.45 mab; the upper and lower sensors provide an estimate of light attenuation K_{dPAR} over the PAR spectrum (400-700 nm), calculated as:

$$K_{dPAR} = -\frac{1}{dz} \ln(PAR_{lower}/PAR_{upper}) \quad (1)$$

where dz is the distance between the two PAR sensors (0.3 m in this case). Light attenuation was calculated only between the hours of 1030 and 1530, when the angle of the sun relative to the deployment location was closest to 0 degrees. All sensors sampled at 15 min intervals, except for the wave recorders, which burst-sampled at 6 Hz every 3 min. Temperature, turbidity, and inner filter effects (IFE) have been shown to alter fDOM measurements (Baker, 2005; Downing et al., 2012). We corrected fDOM measurements to account for temperature, turbidity, and IFE,

according to Downing et al. (2012). Measurements of fDOM at turbidities > 50 NTU were removed due to interference with the fluorescence signal. Chlorophyll-a concentration was calculated from sensor-based fluorescence measurements by regressing fluorescence against discrete measurements of chlorophyll-a made on four dates (August and October 2014, January and April 2015) at all stations (Fig. S1). In short, water was collected in the field at the time of sensor sampling, and filtered through 0.7 µm GF/F filters, which were wrapped in foil, and frozen until laboratory analysis using standard methods (EPA Method 445.0). Non-photochemical quenching (NPQ) was accounted for by removing fluorescence measurements during periods of daylight (upper PAR sensor > 150 W/m², or ~40% of the record), and interpolating to fill gaps. Platforms at sites CB06 and CB11 were identical to platforms at CB03 and CB10 except for the omission of PAR sensors. Light attenuation at those sites was estimated using the model of Gallegos et al. (2011), discussed below; this enabled comparison between vegetated and unvegetated sites using the same fundamental variables to compute light attenuation.

Quality control and quality assurance checks were performed on all data and are described in detail in the data report (Suttles et al., 2017). These included removal of obvious spikes in individual parameter time series using either a recursive filter or median filter technique, in which values that changed from one time point to the next by more than a set threshold were flagged and replaced with a fill value (e.g. -9999). Pre-deployment calibrations and post-deployment check were performed on all EXO2 sensors following YSI procedures outlined in the EXO User Manual (Revision F). Linear corrections were obtained from either post-deployment calibration checks or differences between in situ values from a fouled sensor and from a cleaned, calibrated sensor at the beginning of the next deployment. If data were obviously fouled and corrections were not possible, the data were replaced with fill values. WET Labs ECO-PARSB instruments were checked to verify counts were above their “dark” count (low-count) thresholds; therefore legitimate data collected during night were also discarded. All subsurface pressure data were corrected for changes in atmospheric pressure by using local barometric pressure data, from the meteorological station at CBWS when available, to give a more accurate representation of pressure caused by the overlying water.

We investigated the frequency response of constituents using spectral density estimates (Welch's overlapped segment averaging estimator) and their uncertainties (Bendat and Piersol 1986) for dissolved oxygen, turbidity, chlorophyll-a, and fDOM. Uncertainty bounds on the spectra were used to test if differences in spectral density between sites were significant, where non-overlapping uncertainty envelopes suggest significant differences. We also

performed wavelet coherence analyses (Grinsted et al. 2004) between site CB03 (shoal) and CB06 (channel) for each constituent; wavelet coherence indicates co-variability of two signals at varying frequencies through time, and can identify mechanistic linkages between processes that have complex temporal coupling. The analysis was limited to these two sites due to the more complete data coverage. Statistics of the constituent time-series (mean, maxima, and minima) were computed with the original 15 min data, and with variable sampling intervals (1 h, 2 h, 24 h) to investigate the influence of sampling resolution on spatial [variabilitygradients](#) in biogeochemical variables.

Further details on collection protocols and access to the time-series data are reported by Suttles et al. (2017). The hydrodynamic results of this field campaign have been described in detail by prior studies (Ganju et al. 2016; Beudin et al. 2017; Nowacki and Ganju 2018); for the purposes of this paper we focus on the spatiotemporal variability of constituent concentrations and light attenuation.

2.3 Estimation of light attenuation contributions

We estimated light attenuation (for periods with missing PAR data or at sites with no PAR data) and relative contributions from turbidity, chlorophyll-a, and CDOM using the method of Gallegos et al. (2011). This formulation computes spectral attenuation in terms of suspended and dissolved constituents including the effects of water, CDOM, phytoplankton, and non-algal particulates (NAP, e.g., detritus, minerals, bacteria). We include absorption by four components: (1) absorption by water was assumed to follow the spectral characteristics of pure water; (2) CDOM absorption was taken proportional to fDOM concentration, with a negative spectral slope (Bricaud et al. 1981) set to $s_g=0.02 \text{ nm}^{-1}$ (Oestreich et al., 2016); (3) phytoplankton absorption was proportional to chlorophyll-a concentration and with the spectrum shape normalized by the absorption peak at 675 nm (initial value for peak absorption was taken as $a_{\lambda,675}=0.0235 \text{ m}^2 (\text{mg chl } a)^{-1}$, within the range provided by Bricaud et al. 1995); and (4) non-algal absorption was taken as proportional to the suspended-sediment concentration with a spectral shape (Bowers and Binding, 2006) that included a baseline of $c_{x,l}=0.0024 \text{ m}^2 \text{ g}^{-1}$ (Biber et al. 2008), an absorption cross-section of $c_{x,2}=0.04 \text{ m}^2 \text{ g}^{-1}$ (Bowers and Binding 2006), and a spectral slope of $s_x=0.009 \text{ nm}^{-1}$ (Boss et al. 2001). The backscattering ratio of water was set at 0.5, while CDOM is considered non-scattering (Mobley and Stramski 1997), and the particulate effective backscattering ratio b_{bx} was initially set at 0.017.

Given the high variability in turbidity and suspended-sediment concentrations due to wind-wave resuspension, we varied the value of b_{bx} as a function of turbidity to achieve the best agreement between the model and observations.

At turbidity below 50 NTU, b_{bx} is constant at 0.017, and then linearly declines to 0.0024 at turbidities above 220 NTU. This modification implies that backscattering by mineral particles dominates at low turbidities, while large organic aggregates dominate at the highest turbidities (Gallegos et al. 2011), and has the effect of enhancing attenuation by particulate-induced absorption at the highest turbidities. The lowest values of backscattering to scattering ratio b_{bx} in the literature range from 0.005 (Snyder et al. 2008) to about 0.002 (Chang et al. 2004). The chosen value of 0.0024 was obtained from Loisel et al. (2007). Morel and Bricaud (1981) described the b_{bx} ratio as decreasing with increasing absorption, which would be consistent with high turbidity situations. McKee et al. (2009) described a decrease of the b_{bx} ratio with increasing concentrations in a mineral-rich environment. The maximum sediment concentrations (15 mg/l) in that study were higher than previously mentioned studies, but smaller than concentrations observed in this environment (Nowacki and Ganju 2018). Determining the appropriate backscattering ratio at high turbidities is still poorly constrained and the minimum value used in this study might even be an overestimation (a sensitivity analysis is described in section 3.1). The calibration of the light model at shoal sites (CB03 and CB10) and subsequent application at channel sites assumes that the vertical variability in the bottom 0.3 m of the water column is similar among these sites.

2.4 Net ecosystem metabolism

The basic concept and method for computing community production and respiration (and ecosystem metabolism) was developed by Odum and Hoskin (1958) and, with numerous modifications, has been used since for estimating these rate processes in streams, rivers, lakes, estuaries and the open ocean (Caffrey 2004; Howarth et al. 2014). The technique is based on quantifying increases in oxygen concentrations during daylight hours and declines during nighttime hours as ecosystem rates of net primary production and respiration, respectively. The sum of these two processes over 24 h, after correcting for air-sea exchange, provides an estimate of net ecosystem metabolism. We utilized continuous oxygen concentration measurements at four locations in Chincoteague Bay (CB03, CB10, CB06, CB11) to compare differences in net ecosystem metabolism across sites and across habitats (phytoplankton versus SAV dominated, channel versus shoal). We computed daily estimates of gross primary production (P_g), ecosystem respiration (R_t), and net ecosystem metabolism ($NEM P_n = P_g - R_t$) using the approach of Beck et al. (2015), which utilizes weighted regression to remove tidal effects on dissolved oxygen time-series. The changes in dissolved oxygen concentrations used to compute metabolic rates were corrected for air-water gas exchange using the equation

$$D = K_a(C_s - C)$$

where D is the rate of air-water oxygen exchange ($\text{mg O}_2 \text{ L}^{-1} \text{ h}^{-1}$), K_a is the volumetric aeration coefficient (h^{-1}), and C_s and C are the oxygen saturation concentration and observed oxygen concentration ($\text{mg O}_2 \text{ L}^{-1}$), respectively. K_a was computed as a function of wind speed measured at a weather station installed at a dock (near CB10; Suttles et al. 2017). Details of the air-water gas calculation are incorporated into the R package WtRegDO (Beck et al. 2015) and described in detail in Thebault et al. (2008). The calculations utilized salinity, temperature, and dissolved oxygen times-series from the sensors at each platform, and atmospheric pressure and air temperature data from a nearby buoy (OCIM2 - 8570283 at the Ocean City Inlet, Maryland).

The oxygen data used to make metabolic computations were obtained from sensors deployed near-bottom in relatively shallow waters. Our metabolic computations assume that the water-column is well-mixed, which is necessary for the air-water flux correction to be valid and for the oxygen time-series to be representative of the combined water-column and sediments (e.g., Murrell et al. 2018). We tested the validity of this assumption at our two deeper sites (which were 2-3 m deep) using monthly vertical profile data for dissolved oxygen, temperature, and salinity from the Maryland Department of Natural Resources from 1999-2014. These data indicate instantaneous vertical dissolved oxygen differences of $>1 \text{ mg/L}$ on occasion over the 15 y record, generally during the productive summer months. The long-term mean vertical oxygen difference, however, was less than 0.5 mg/L during September to May, but between 0.5 and 0.9 mg/L during June-August, indicating that although we did not measure vertical gradients during our study, they do occasionally develop.

3 Results

3.1 Time-series of turbidity, chlorophyll-a, fDOM, dissolved oxygen, and light attenuation

Turbidity ranged from near zero to a maximum of over 400 NTU at site CB06 during a winter storm (December 2014) that induced waves exceeding 0.7 m (Figs. 2-5). The highest peak turbidities were observed at CB06, while sites CB03, CB10, and CB11 had similar statistical distributions of turbidity (Table 3), though the differences between mean values over the entire period were not statistically significant (Table 3). Turbidity at shoal sites was

highest in the winter, between November-April, while turbidity at site CB06 did not display a similar seasonal signal. In general, tidal resuspension appeared to be minimal although tidal advection after large wind-wave resuspension events was observed (Figs. 2-5). Spectrally, most of the energy in the turbidity signal was found at subtidal frequencies (i.e., > 1 day), with little correspondence between sites at tidal frequencies (Fig. 6).

5 Chlorophyll-a concentrations peaked just below 50 µg/L at site CB11, and below 30 µg/L at all other sites (Figs. 2-5). ~~Mean Chlorophyll-a concentrations were not statistically different comparable~~ across sites ~~CB03, CB10, and CB06~~, but ~~were nonetheless on average~~ twice as high at CB11 (Table 1). The largest concentrations were observed in winter during resuspension events, ~~likely representing resuspension of benthic microalgae~~, and during a broad spring bloom during March-April 2015. All sites showed a large decrease in chlorophyll-a concentration during a
10 period of ice formation in late February 2015, when advection and resuspension were largely halted in the estuary. Despite removing effects of ~~non-photochemical quenching NPQ~~, an attenuated diel signal remains in the spectra that may be a partial artifact or consistent with day-night variations in PAR (Fig. 6). The diurnal signal was comparable across all stations and was stronger than the tidal signal (Fig. 6).

Concentrations of fluorescing dissolved organic matter (fDOM) varied between 0 and 70 QSU, with the highest
15 ~~peak values consistently over the overlapping period of record~~ observed at CB11 in Newport Bay (Figs. 2-5). Concentrations were similar between the other sites, with lowest fDOM in the winter, possibly due to reduced ~~dissolved organic material production in the surrounding watershed, and ultimately in freshwater runoff to the estuary biological activity~~. Periodic large decreases in fDOM coincided with increases in turbidity due to wind-wave resuspension; despite correcting the fDOM time-series for turbidity interference, the fluorescence measurement was
20 likely attenuated beyond correction. Most of the energy in the fDOM signal was at subtidal frequencies (Fig. 6), there was a distinct peak at the M₂ tidal frequency (12.42 h ~~or ~0.5 days~~) corresponding to advection of fresher, fDOM-elevated water on ebb tides.

Dissolved oxygen percent saturation ranged from a minimum of 19% at site CB11 in the summer, to a maximum of over 200% at site CB10 (Fig. 7), with maximum diel variability in the summer. The channel sites (CB06 and CB11)
25 showed substantially attenuated diel variations as compared to the shoal sites (CB03 and CB10). Diel fluctuations in the winter were typically less than the subtidal changes. Spectral analysis clearly shows the dominance of 24 h diel fluctuations at all sites (Fig. 6), with higher energy at the vegetated shoal sites (CB03, CB10).

Direct measurements of K_{dPAR} at sites CB03 and CB10 were partially confounded by instrument fouling and malfunction of one or both sensors on each platform. Nonetheless, we successfully captured a wide range of conditions, with peak light attenuation of approximately 10 m^{-1} occurring at sites CB03 and CB10 in the winter during a sediment resuspension event (Figs. 2-5). Median K_{dPAR} was approximately 1 m^{-1} at both shoal sites, but event-driven magnitudes exceeded 2 m^{-1} multiple times during the deployment. The K_{dPAR} data were used to calibrate the light model (Fig. S2), which we implemented to reconstruct missing data at sites CB03 and CB10, and to estimate light attenuation at sites CB06 and CB11 where constituent concentrations were measured but no light data were collected. The full time-series of estimated light attenuation at the four sites demonstrates the strong seasonal variability of light attenuation at all sites (Figs. 2-5), with peak attenuation occurring during winter storms. Wave-induced sediment resuspension and advection were responsible for increased turbidity, which accounted for approximately 40% of the light attenuation at sites CB03, CB10, and CB11, and 61% at site CB06. Light attenuation at site CB11, with its proximity to freshwater and nutrient sources, was highest overall during the overlapping period of record and more highly influenced by chlorophyll-a and fDOM than at other sites. Median K_{dPAR} was highest at the two unvegetated channel sites, and lowest at the two vegetated shoal sites (Table 2), though the period-means are not statistically different. Turbidity, the strongest driver of K_{dPAR} , was generally lower during the warm season at the vegetated shoal sites and the turbidity generated for a given wind-wave induced bed shear stress was significantly reduced in summer at the vegetated sites, while there was no significant seasonal variation in the turbidity-shear stress relationship at unvegetated channel sites (Fig. 8).

Sediments at site CB06 and CB11 tend to be finer (Ellis et al., 2015) and may be more susceptible to flocculation, which would induce error in the light model. Therefore we ran the light model with variation in the maximum particle backscatter ratio (b_{ps}) which is particle size dependent. An increase in flocculated suspended-sediment would increase this ratio, thereby increasing the light attenuation and causing a larger difference in light attenuation between the sandy, vegetated sites and the siltier, unvegetated sites. Modifying the peak value as described above from 0.017 to 0.025 (47% increase; 0.025 value was used for a different mud-dominated estuary by Ganju et al., 2014) increased the median light attenuation at CB06 by 17% (from 1.35 to 1.58 m^{-1}) and at CB11 by 11% (from 1.67 to 1.86). Therefore the spatial variability in light attenuation is likely robust and not confounded by variability in particle size.

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Wavelet coherence of dissolved oxygen between site CB03 and site CB06 was maximized at diel timescales (Fig. 9), corresponding to co-varying oscillations due to daytime production and night-time respiration. Coherence was also high at lower frequencies, especially during winter when oscillations at both sites were small due to reduced production and respiration. Turbidity was coherent between channel and shoal sites primarily at subtidal timescales, corresponding to episodic multi-day resuspension events and generally high turbidity during most of the winter. Phase lag was minimal during times of high coherence for both of these parameters. Coherence for both chlorophyll-a and fDOM was minimal throughout the year, though the former demonstrated sporadic coherence at diel and multi-day timescales. The December-April period of generally high chlorophyll-a was similar between these sites and manifested as increased coherence at ~512-1024 h (21-42 d).

Statistics of the constituent time-series were computed with temporal sampling intervals of 15 min (original sampling interval), 1 h, 2 h, and 24 h (Figure 10; Table 3). While mean values were minimally altered, minima and maxima were significantly dampened for dissolved oxygen and turbidity. At site CB11, the original sampling interval captured a hypoxic event in September 2014 with a minimum value of 19% dissolved oxygen, while daily sampling yielded a minimum of 55%, above the hypoxic threshold. Similarly, at site CB10 supersaturation led to a maximum of over 200% dissolved oxygen, while daily sampling yielded a maximum of 152%. Maximum values of turbidity were similarly attenuated by daily sampling (Table 3). The effect of these sampling intervals is discussed later in the manuscript.

3.2 Net ecosystem metabolism

Estimates of ecosystem metabolism displayed strong seasonal variability, with elevated rates of P_g and R_t during warm months across all stations (Figs. 11, 12; Table S1). High rates of P_g and R_t persisted between May and ~~October~~September, when temperature and PAR peaked seasonally, and were consistently lower during November to ~~April~~March (Figs. 11, 12), and R_t was exponentially related to temperature across sites (although less so at CB06). ~~Daily~~ ~~Mm~~ metabolic rates were clearly higher at vegetated shoal sites (CB03, CB10), consistent with the strong diurnal signal in oxygen at these sites (Figs. 6, 7) and the presence of seagrass (Table 1). Temperature was strongly associated with rates of respiration across all sites, and respiration reached peak values under conditions of high temperatures and high rates of P_g (Fig. 12). Gross primary production and respiration were largely balanced across all sites, but ~~the frequency of instances of daily~~ net autotrophy ($P_g > R_t$) ~~was occurred higher~~ ~~nearly 70% of the time~~

at the vegetated sites, and were also persistent at CB06 (Fig. 13). P_g and R_t were more balanced at CB11, but we did not have enough data to capture the complete seasonal cycle at this site (Fig. 11).

The relationship between rates of P_g and light availability was site specific. At the vegetated sites, CB03 and CB10, P_g and K_{dPAR} were significantly negatively correlated, with high rates of P_g occurring during the periods of lowest K_{dPAR} and highest surface PAR (Fig. 14). At CB06 and CB11, variations in P_g and K_{dPAR} were not significantly correlated and K_{dPAR} was higher and P_g was lower at these sites than at CB03 and CB10 (Tables 1, 2; Figs. 13, 14).

Over the entire record, mean values were not statistically different except for mean P_g between sites CB03 and CB06 (Table 1). In summary, the highest daily metabolic rates we measured occurred during warm periods in vegetated shoals, where wave attenuation by seagrass reduced turbidity and K_{dPAR} .

4 Discussion

Net ecosystem metabolism and other water-quality parameters are typically measured over limited spatial (e.g. single point) and temporal (e.g. days-to-weeks) scales. An analysis of a comprehensive suite of high-frequency biological and physical measurements in Chincoteague Bay over an annual cycle revealed the primary drivers of light attenuation, the role of light attenuation in driving variations in gross primary production, the primary timescales of biogeochemical variability, and the effect of habitat type (i.e., vegetated versus un-vegetated; nutrient enrichment) on oxygen variability and net ecosystem metabolism. Turbidity dominated light attenuation variability and varied considerably at 1-7 day time scales, consistent with the frequency of storm passage. Storm-associated wind waves were the specific driver of resuspension and turbidity, and reduction of bed shear stress and turbidity in SAV-dominated shoal environments during summer increased light availability in these habitats. As a consequence, P_g was substantially higher in SAV-dominated shoals compared with adjacent plankton-dominated sites and P_g was negatively correlated with K_{dPAR} in the shoals, highlighting the role of short-term variability in light availability driving P_g . High rates of P_g and R_t in shoal environments led to much higher diurnal and seasonal-scale variability in dissolved oxygen in these habitats. Though we do not have direct estimates of microalgal production partitioned between sediments and the water column at any of our sites, the sediment surface is nearly covered by SAV and macroalgae at the vegetated sites, therefore we expect that benthic microalgal production is low. At site CB11, K_{dPAR}

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values indicate that very little light reaches the bottom, so we presume that water-column microalgae are dominant.

For site CB06, we cannot constrain the partitioning between the two environments.

4.1 Spatiotemporal variability of constituents

4.1.1 Interpretation of spectral signals

5 Spectral analysis of the constituent time-series elucidates significant differences between channel and shoal sites, as well as longitudinally within the estuary. With regards to dissolved oxygen, the peak in diurnal energy at shoal sites is markedly stronger relative to channel sites with no overlap in uncertainty bounds between channel and shoal sites, indicating higher local production/respiration due to SAV presence. Between sites CB03 and CB06, the diurnal peak in spectral density is reduced by 94%, and reduced by 83% between site CB10 and CB11. The peak in spectral
10 density was 30% higher at CB03 than CB10, perhaps due to higher SAV biomass at that site. This indicates the dominant role of seagrass in controlling oxygen concentrations locally, and the relatively weaker oxygen dynamics at the channelized, unvegetated sites. From a sampling perspective, this also suggests that interpreting oxygen dynamics with limited spatial resolution is confounded by heterogeneous benthic coverage.

The highest spectral densities for turbidity were observed at site CB06, across all frequencies. At the lowest
15 frequencies, corresponding to multi-day storms, there was no overlap in the uncertainty bounds. Nowacki and Ganju (2018) showed that sediment fluxes at this site were an order of magnitude larger than the other three sites, due to a strong response to wind events, which are typically multi-day events. The central location of the site suggests that local sediment transport would respond consistently to the seasonally variable winds, which tend to act along the axis of the estuary. Resuspension over shoals on either end of the estuary, and subsequent advection cause the
20 spatial integration of resuspension processes throughout the estuary at this main channel site.

Spectral density peaks of chlorophyll-a were similar between sites with significant overlap across frequencies, but site CB11 demonstrated the highest total spectral density, primarily at subtidal frequencies. The reduced exchange with other portions of the estuary, local nutrient inputs, and longer residence times increase the potential for phytoplankton growth at this station, which is known to be locally eutrophic (Fertig et al. 2009). This coincides with
25 the lowest oxygen minima across the four sites (Table 3), indicating eutrophication. Again, in regards to sampling

strategies, characterizing spatial variability gradients in eutrophication is confounded by the competing timescales of transport and biogeochemical processes.

While the tidal signal in fDOM was pronounced (peak at 12.42 h or 0.5 days), the bulk of the spectral density was in the lower frequency band at sites CB10 and CB11, in the northern half of the estuary. This likely represents a longitudinal gradient in freshwater input, with decreased salinity in the north where inputs are largest. The inverse correlation between salinity and fDOM in Chincoteague Bay was identified in prior work (Oestreich et al. 2016), and highlights the roles of tidal circulation, residence time, and freshwater input in biogeochemical and light attenuation patterns.

4.1.2 Channel-shoal coherence of water quality parameters

Wavelet coherence analyses demonstrate that though certain parameters are tightly coupled between channel and shoal, other parameters are not. The coherence of dissolved oxygen at diel and longer timescales, along with nearly no phase lag, is congruent with diel and seasonal patterns of production and respiration. The “breathing” of the estuary (Odum and Hoskin, 1958) appears as a near-simultaneous process regardless of location, and demonstrates the importance of local biogeochemical processes throughout the system. With regards to turbidity and sediment resuspension, the coherence at subtidal timescales indicates that channels integrate resuspension processes over the entire estuary; minimal phase lag suggests that a rapidly evolving local wave climate in response to episodic winds controls both resuspension and rapid advection through the channels (Nowacki and Ganju, 2018). Conversely, chlorophyll-a demonstrates coherence only over the seasonal timescale, and therefore advection and other tidal processes do not appear to control dynamic exchange of phytoplankton or benthic microalgae. Lastly, the lack of coherence in fDOM shows that freshwater input is relatively minor in the southern half of the estuary, and despite a significant peak in spectral density at the tidal timescale, the advection of freshwater (and fDOM) between channel and shoal is not significant. Despite a qualitatively coherent annual signal in fDOM, the analysis cannot detect coherence at this timescale with a single year of data. These complex relationships between parameters highlight both the intricate response to spatiotemporal variability in biophysical environments, and the necessity of high-frequency, spatially comprehensive measurements to understand estuarine ecosystem behavior.

4.1.3 Influence of temporal sampling resolution on spatial variability gradients

Many estuary sampling programs, including “citizen science” efforts, often collect one daily sample of water quality parameters, including dissolved oxygen (Rheuban et al., 2016). Summers et al. (1997) demonstrated that month-long records of dissolved oxygen in a number of estuarine systems, resampled to replicate various non-continuous sampling programs, were not able to identify oxygen minima. This study expands on that finding by mimicking sampling programs within one large estuary, over an entire year. Though mean biogeochemical values were relatively insensitive to sampling interval, maximum values were significantly modulated for all parameters, while minimum values for dissolved oxygen were also affected. With regards to dissolved oxygen specifically, we find that daily sampling dampens the spatial variability in maxima and minima between sites. For example, a daily sampling program would not detect hypoxic conditions at site CB11, and would only observe a 13% difference in dissolved oxygen as compared to nearby site CB10, while the actual difference was over 30%. This is relevant given the ubiquitous daily sampling programs in many estuaries, which cover numerous locations with infrequent sampling. This result suggests that characterizing differences in water-column conditions across space requires sampling at timescales finer than 1 day, especially in highly metabolic environments where diel oscillations in dissolved oxygen are large. Continuous monitoring at multiple locations in an estuary is difficult due to fouling and instrument limitations (e.g. power and memory), especially across large systems over an entire year. However, it may be more informative to conduct shorter term, continuous monitoring deployments at select locations rather than daily monitoring at many locations. In the case of Chincoteague Bay, characterizing the oxygen environment would be possible with seasonal, month-long deployments at four sites. This type of data would yield both an accurate assessment of minimum and maximum values, and allow for spectral analysis to capture tidal timescale variability.

4.2 Spatiotemporal variability of light attenuation

Light attenuation can appear to be controlled by uncoupled biological (i.e., nutrient loading and phytoplankton blooms) and physical (sediment resuspension) processes, but in reality the feedbacks between physics and biology are consistently present in estuaries. Wave-induced suspended-sediment resuspension is primarily responsible for light attenuation in shallow lagoons like Chincoteague Bay, and seagrass clearly modulates the magnitude and spatiotemporal variability in sediment resuspension through the attenuation of wave energy (e.g., Hansen and Reidenbach 2013). This wave-induced resuspension and turbidity occurs on the time scale of periodic wind events

observed in this system. During summer months, when vegetation densities are ostensibly the highest, the dependence (i.e. slope) of turbidity on bed shear stress decreases significantly at the vegetated shoal sites, while it is not significantly different at the unvegetated channel sites (Fig. 8). In the winter, when seagrass is largely absent, the slope increases at the shoal sites, indicating the influence of seagrass on bed stabilization in the summer. The role of seagrass in modulating their physical environment is demonstrated by this improvement in light availability when vegetation re-establishes in the warmer months. Furthermore, the dependence of P_g on K_{dPAR} at the vegetated shoal sites (Fig. 14) completes the positive feedback loop between benthic stabilizationseagrass, sediment resuspension, light attenuation, and primary production. It is important to note that the seasonal change in resuspension response at the shallow sites may also be partially attributed to increased microphytobenthos coverage in summer. However it is likely that in these vegetated sandy substrates, seagrass would dominate bed stabilization over benthic microalgae given shading and coarser sediment. In fact, the data of Ellis et al. (2015) indicate organic matter percentages of less than 2% on the sandy shoals.

Our measurements underscore the ability of comprehensive continuous measurements to capture multiple scales of variation in light attenuation. For example, subsampling our light attenuation measurements under fair weather conditions (wave height < median) leads to underestimation of median K_{dPAR} by as much as 27% (sites CB06 and CB10). Subsampling to periods with wave heights less than the 84th percentile leads to underestimation by 13% (site CB06). The effect of these changes can be large if sparse data are used to assess trends in biogeochemical variables, or to drive ecological models. In situations where continuous monitoring of light attenuation is difficult, light modeling using continuous measurements of turbidity, chlorophyll-a, and fDOM is a suitable proxy. With a reasonable number of discrete light attenuation samples, it is possible to calibrate a light model and estimate seasonal changes in contributions to light attenuation using the constituent measurements. This becomes useful when attempting to determine the relative influence of physical and biogeochemical processes on spatiotemporal variations in light attenuation, and the potential benefits of restoration activities.

Apart from Newport Bay in the northwest corner of the estuary, Chincoteague Bay is relatively unimpacted by external nutrient and organic material inputs, as compared to other lagoonal systems on the U.S. East Coast (e.g. Indian River Lagoon, Philips et al. 2002; Great South Bay, Kinney and Valiela 2011). For example, the relatively urbanized watershed surrounding Barnegat Bay, New Jersey, contributes larger nutrient loads to the northern portion

of the estuary (Kennish et al. 2007), however the highest light attenuation is observed in the southern portion, where sediment concentration are highest due to the availability of fine bed sediment and wind-wave resuspension (Ganju et al. 2014). In fact, in Barnegat Bay light attenuation is lowest in the more eutrophic, poorly flushed (Defne and Ganju 2015) northern portion due to coarser bed sediments and smaller wave heights (and less resuspension). In the more nutrient-enriched regions of the Maryland coastal bays north of Chincoteague Bay, chlorophyll-a is also much higher and likely makes a larger contribution to light attenuation (Boynton et al. 1996). In contrast, spatial variability gradients in the light field within Chincoteague Bay ~~is~~ driven by geomorphology (depth), the presence of submerged aquatic vegetation (CB03 and CB10), and the higher contribution of chlorophyll-a at a nutrient-enriched site (CB11; Boynton et al. 1996). Thus, generalizing the light attenuation in back-barrier estuaries is hampered by these subtle, habitat-specific controls on attenuation, but improvements in continuous monitoring will lead to an increased understanding of these controls.

4.3 Spatiotemporal variability in metabolism and relationship with benthic ecosystem and light attenuation

Clear differences in the magnitude and timescales of oxygen dynamics were present across habitats in Chincoteague Bay. Spectral density at the diurnal time-scale for dissolved oxygen was significantly higher at the two vegetated sites, where high rates of primary production during the day and associated respiration at night led to large changes in dissolved oxygen. Although this pattern is unsurprising given the high metabolic rates in these dense seagrass beds, this study is one of few studies that have clearly documented this pattern inside and outside of SAV beds within an estuary over annual timescales. Elevated metabolic rates in macrophyte dominated habitats relative to phytoplankton dominated habitats have been documented in other systems (D'Avanzo et al. 1996), where the canonical C:N molar ratio of phytoplankton (C:N = 6.6) is much less than that measured for Chincoteague Bay *Z. marina* (C:N = 25.8±7.2), indicating higher carbon and oxygen metabolism for a given amount of nitrogen (e.g., Atkinson and Smith 1983). Elevated magnitudes of daily oxygen change have also been associated with elevated phytoplankton biomass in Chincoteague Bay (e.g., Boynton et al. 1996) and other systems (e.g., D'Avanzo et al. 1996). This is consistent with the fact that the site at Newport Bay (CB11), which had the highest plankton chlorophyll-a concentrations across all sites, had a larger spectral density for dissolved oxygen at the diel frequency than at CB06. Newport Bay is known to have elevated nutrient inputs and concentrations associated with land-use in

its watershed (Boynton et al. 1996), and spectral density for chlorophyll-a was highest at this location (Fig. 6), suggesting that this site is responding to eutrophication.

Seasonal variations in metabolic rate estimates were typical of temperate regions, but were variable across space. At all stations, rates of P_g and R_t were highest in warmer months when incident PAR and temperature are highest in this region (e.g., Fisher et al. 2003), but rates were especially high in the later summer (August and September) at the vegetated sites. This late-summer period is typically when SAV biomass is highest and where the self-reinforcing effect of wave attenuation, reduced resuspension, and reduced K_{dPAR} allow for high rates of primary production (Figs. 8, 14). Indeed, the relationship between K_{dPAR} and P_g was strongest at these vegetated sites (Fig. 14), displaying the sensitivity to benthic primary production to light attenuation, which was dominated by turbidity. Elevated temperatures during this period also stimulate respiration (e.g., Marsh et al. 1986), which has been documented at the ecosystem scale in other SAV-dominated coastal systems (Howarth et al. 2014, Boynton et al. 2014). Seasonal variations in metabolic rates were comparatively lower at the unvegetated, non-eutrophic site (CB06), where low phytoplankton biomass and the absence of macrophytes leads to limited primary production and respiration rates.

The metabolic balance (ratio) between primary production and respiration is a metric of interest in coastal aquatic ecosystems as it provides an indication of whether the system is a relative source or sink of carbon with respect to the atmosphere (Stæhr et al. 2006). Many river-dominated estuaries and estuaries flanked by extensive tidal marshes are net heterotrophic, where respiration exceeds primary production (e.g., Kemp and Testa 2011). P_g and R_t are often balanced or indicate modest autotrophy in shallow lagoons where SAV is present (e.g., Ferguson and Eyre 2010, D'Avanzo et al. 1996). Autotrophy tends to persist in these habitats as submerged macrophytes like SAV generate high biomass where light availability is relatively high given low nutrient concentrations, low phytoplankton biomass, and sediment trapping within SAV beds. In fact, modest net autotrophy (on a monthly basis) prevailed during the summer season at vegetated sites but not at un-vegetated sites. The rates of P_g and R_t we estimated, which were typically between 200 and 400 mmol O₂ m⁻² d⁻¹, are comparable to those measured in other temperate ecosystems dominated by *Zostera marina* (Howarth et al. 2014) and net autotrophy in these low-nutrient environments is consistent with other coastal lagoons in the region (Giordano et al. 2012, Stutes et al. 2007). While the overall metabolic balance (P_g/R_t) of Chincoteague Bay is difficult to assess, given large differences in metabolic

balance across ~~a range of gradients of~~ depth and nutrient enrichment, we made a simple calculation where mean rates of net ecosystem metabolism at CB10 were multiplied by the area of SAV in 2015 (30 km²) and rates from CB06 are multiplied by the remaining area without SAV (276 km²). The calculation assumes that rates measured at these stations are representative of similar habitats across Chincoteague Bay, and does not account for the potential of benthic photosynthesis to occur in shallow regions not occupied by SAV. This approach generates an estimate of net metabolism for Chincoteague Bay of 317±461 Mmol O₂ yr⁻¹, which indicates autotrophy basin-wide, but the uncertainty is high. Given ongoing loss of SAV in this and other back-barrier systems, it is likely that the metabolic balance may shift towards heterotrophy in the future.

P_g and R_t were tightly coupled in the Chincoteague Bay stations during 2014-2015, suggesting that primary producers were the dominant sources of respiration in the ecosystem. At the vegetated stations, low plankton chlorophyll-a levels indicate that SAV were the dominant contributors to ecosystem metabolism, though macroalgae and epiphytic algae were also present. Alternatively, high chlorophyll-a levels at CB11 (Newport Bay) suggest that phytoplankton were dominant contributors to metabolism, as this site is known to be nutrient enriched. Alternate sources of primary production and respiration that could drive metabolism include benthic micro- and macro-algae and heterotrophic bacteria. Previously measured sediment oxygen uptake (SOU) rates during summer in sediment incubations without light suggest a mean SOU of 56.2 (±24.7) mmol O₂ m⁻² d⁻¹ (Bailey et al. 2005) which is ~50% of the mean summer respiration rates at CB06, but only 15% of rates measured at the phytoplankton-dominated CB11. Rates of benthic photosynthesis were not available at this site, but with a mean K_{dPAR} of 1.35 at CB06 and a water-column depth of 3 m, only a small fraction of surface PAR would be expected to reach the sediments. Thus, respiration at CB06 might be evenly split between the water-column and sediments, which fits well with previous cross-system comparisons of coastal marine ecosystems (Kemp et al. 1992, Boynton et al. 2018).

5 Conclusions

We tested the hypothesis that spatiotemporal differences in benthic habitat would exert a strong control on biogeochemical variables and primary production in an otherwise well-mixed estuary. We found a clear linkage between SAV presence/absence, dependence of turbidity on shear stress, the ensuing light attenuation, and gross primary production, reinforcing the positive feedback loop between seagrass presence and primary production.

Vegetated shoal sites exhibited higher metabolic rates, reduced sediment resuspension, and reduced light attenuation as compared to unvegetated channel sites. Light attenuation was dominated by wind-wave induced sediment resuspension, which was minimized during summer months in vegetated shoal sites when SAV aboveground biomass was at its peak. Furthermore, we found a clear causal linkage between SAV presence, the reduction of turbidity at a given shear stress, lowered light attenuation, and elevated gross primary production, reinforcing the positive feedback loop between seagrass presence and primary production. As a consequence, despite relatively balanced ecosystem production and respiration across all sites, high rates of primary production led to modest net autotrophy at sites vegetated by SAV. Future shifts towards phytoplankton-dominant habitat, given ongoing SAV loss, will likely shift these systems towards net heterotrophy and reduce their effectiveness as carbon sinks.

5

10 Interpreting spatial differences in biogeochemistry was strongly influenced by temporal sampling resolution, where daily sampling reduced spatial variability between sites and attenuated peak values significantly. These results demonstrate the value of high-temporal resolution measurements at multiple locations within a given estuary, due to the strong interplay between geomorphology, light attenuation, SAV dynamics, and ecosystem metabolism. The mechanisms identified here demonstrate a need to consider feedbacks between biological and physical processes in

15 estuaries, especially when constructing deterministic models or evaluating future ecosystem responses to eutrophication and climate change.

Code availability

Freely available MATLAB toolboxes, as cited in the text, were used for analyses.

20 Data availability

The time-series data are available at <https://stellwagen.er.usgs.gov/chincoteague.html>

Sample availability

N/A

Appendices

25 N/A

Team list

N/A

Author contribution

N.K.G., S.E.S, and J.M.T. designed the study, A.L.A. implemented the light model, all authors analyzed time-series
5 data and drafted the manuscript.

Competing interests

The authors declare that they have no conflict of interest.

Disclaimer

Use of brand names is for identification purposes only and does not constitute endorsement by the U.S. Government.

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Figure captions

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Figure 2. Time-series of wave height, upper and lower PAR, KdPAR, turbidity, chlorophyll-a, and fDOM at vegetated shoal site CB03.

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Figure 3. Time-series of wave height, upper and lower PAR, KdPAR, turbidity, chlorophyll-a, and fDOM at vegetated shoal site CB10.

Figure 4. Time-series of wave height, modeled KdPAR, turbidity, chlorophyll-a, and fDOM at unvegetated channel site CB06.

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10 **Tables**

Table 1. Water-column properties at the four study sites, SAV presence/absence, and metabolic rate estimates.

Temperature data are minimum and maximum values (min, max), salinity is the annual mean, chlorophyll-a is the annual mean, and metabolic rates estimates are means (± standard deviation*SD*) for the August-September (peak

SAV growth) period. Standard deviations of all parameters are sufficiently large such that differences between

15 means are not statistically significant except for P_e at site CB06 relative to site CB03.

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	CB03	CB06	CB10	CB11
<i>Depth (m⁻¹)</i>	1	3	1	2.5
<i>Temperature (degC)</i>	-1.6 – 30.8	-1.6 – 29.1	-1.7 – 30.1	-1.5 – 29.2
<i>Salinity</i>	26.8	26.7	26.2	24.5
<i>Chlorophyll-a (µg/L)</i>	5.2	5.6	5.9	11
<i>SAV presence</i>	Y	N	Y	N
<i>P_g (mmol O₂ m⁻² d⁻¹)</i>	310.7 (±162.2)	93.2 (±53.7)	281.9 (±198.5)	239.9 (±133.9)
<i>R_t (mmol O₂ m⁻² d⁻¹)</i>	304.1 (±179.7)	92.9 (±55.3)	267.0 (±198.2)	252.2 (±133.4)
<i><u>NEMP_n</u> (mmol O₂ m⁻² d⁻¹)</i>	6.7 (±36.8)	0.3 (±27.1)	14.9 (±46.6)	-12.3 (±57.8)

Table 2. Median light attenuation and relative contributions from turbidity, chlorophyll-a, and fDOM. Remainder of light attenuation contribution is from water (not shown), and equivalent between sites.

	CB03	CB06	CB10	CB11
K_{dPAR} (m ⁻¹)	1.00	1.35	1.19	1.67
K_{dPAR} (turbidity)	0.44 (44%)	0.82 (61%)	0.53 (45%)	0.66 (40%)
K_{dPAR} (chl-a)	0.11 (11%)	0.10 (7%)	0.11 (10%)	0.37 (22%)
K_{dPAR} (fDOM)	0.17 (17%)	0.15 (11%)	0.24 (20%)	0.31 (19%)

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Table 3. Mean, minima, and maxima for four water-quality parameters at four sites with variable temporal sampling resolution, over the entire period of record. Standard deviations of all parameters, at original temporal resolution, are sufficiently large such that differences between means are not statistically significant except for fDOM at sites CB10 and CB11 relative to site CB06.

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Site	Resolution	Dissolved oxygen (%)			Turbidity (NTU)			Chlorophyll-a (ug/L)			fDOM (Q)	
		Mean (std. dev.)	min	max	mean	min	max	mean	min	max	mean	min
CB03	15 min	104.4 (14.9)	40.9	182.8	14.9 (26.4)	0.0	427.4	5.2 (4.2)	0.2	29.9	12.6 (4.0)	0.0
	1 h	104.4	41.2	182.3	14.9	0.0	396.9	5.2	0.2	29.7	12.6	0.0
	2 h	104.4	44.4	180.1	14.9	0.0	396.9	5.2	0.2	29.0	12.6	3.4
	24 h	106.5	59.1	149.2	16.7	0.0	242.7	5.5	0.2	27.6	12.5	3.7
CB06	15 min	103.0 (5.69)	70.2	149.4	27.1 (47.2)	0.0	546.8	5.6 (5.0)	0.0	30.0	6.7 (3.3)	0.0
	1 h	103.0	75.0	149.0	27.0	0.0	546.8	5.6	0.0	29.7	6.7	0.0
	2 h	103.0	77.9	149.0	27.2	0.0	525.7	5.6	0.1	29.1	6.7	0.0
	24 h	104.2	92.8	141.5	28.7	0.0	356.7	5.2	0.2	25.4	6.7	0.0
CB10	15 min	108.1 (15.9)	53.4	214.5	13.6 (22.0)	0.0	375.0	5.96 (4.3)	0.2	39.5	16.7 (4.4)	3.4
	1 h	108.1	55.3	212.0	13.6	0.0	308.8	5.96	0.3	35.9	16.7	3.4
	2 h	108.1	56.3	212.0	13.6	0.0	308.8	5.86	0.3	35.8	16.7	3.4
	24 h	105.1	68.7	151.9	15.7	0.0	274.5	4.85-0	0.3	35.8	16.5	4.2

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CB11	15 min	99.3 (14.3)	19.2	160.3	12.4 (23.1)	0.0	275.2	11.0 (8.3)	0.5	49.6	26.0 (10.4)	5.4	Formatted: Font: 9 pt
	1 h	99.3	19.4	160.3	12.4	0.0	255.5	11.1	0.5	49.0	26.0	5.5	Formatted: Font: 9 pt
	2 h	99.3	20.9	160.3	12.4	0.0	255.5	11.1	0.5	40.9	26.1	5.5	Formatted: Font: 9 pt
	24 h	104.7	55.3	143.9	14.2	0.0	174.3	10.8	0.7	36.8	25.6	8.1	Formatted: Font: 9 pt
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Figures

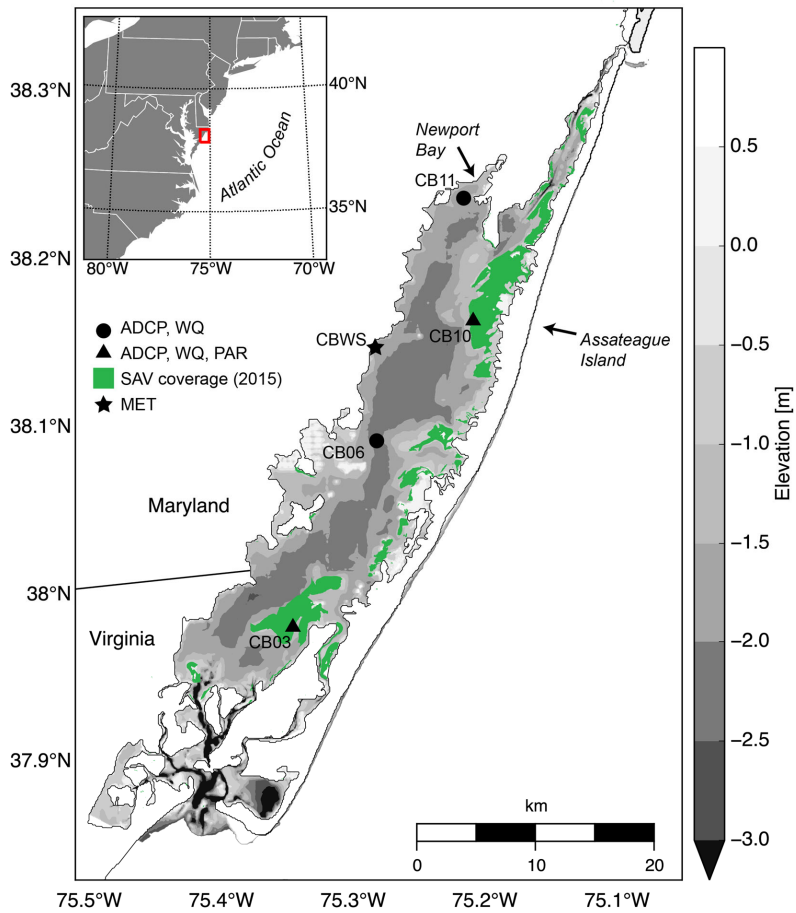


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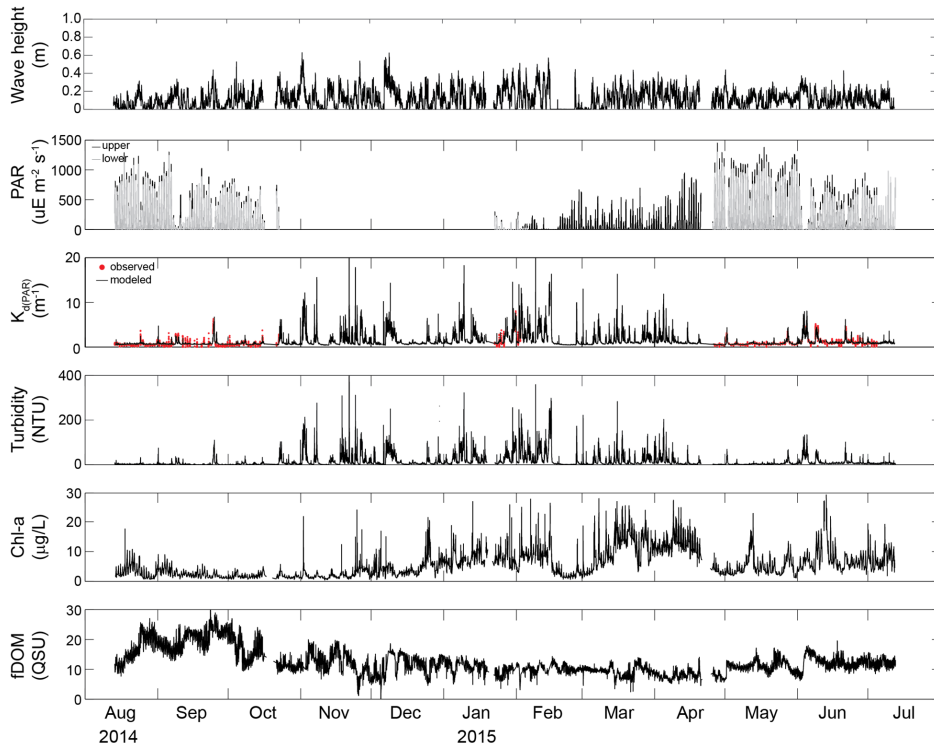


Figure 2. Time-series of wave height, upper and lower PAR, K_{dPAR} , turbidity, chlorophyll-a, and fDOM at vegetated shoal site CB03.

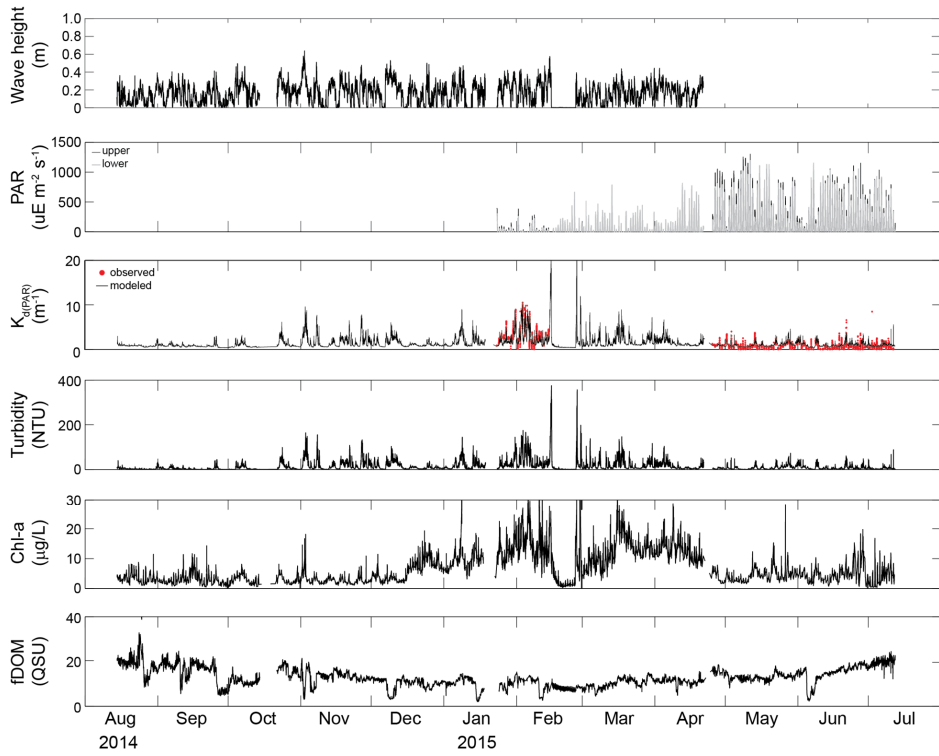


Figure 3. Time-series of wave height, upper and lower PAR, $K_{d(PAR)}$, turbidity, chlorophyll-a, and fDOM at vegetated shoal site CB10.

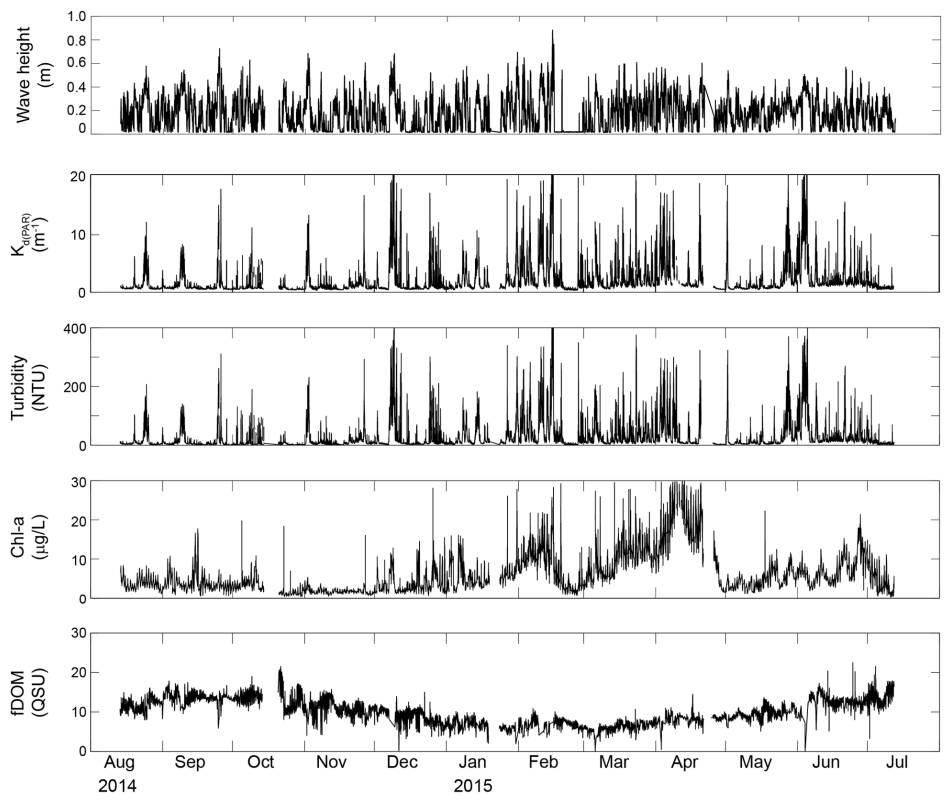


Figure 4. Time-series of wave height, modeled K_{dPAR} , turbidity, chlorophyll-a, and fDOM at unvegetated channel site CB06.

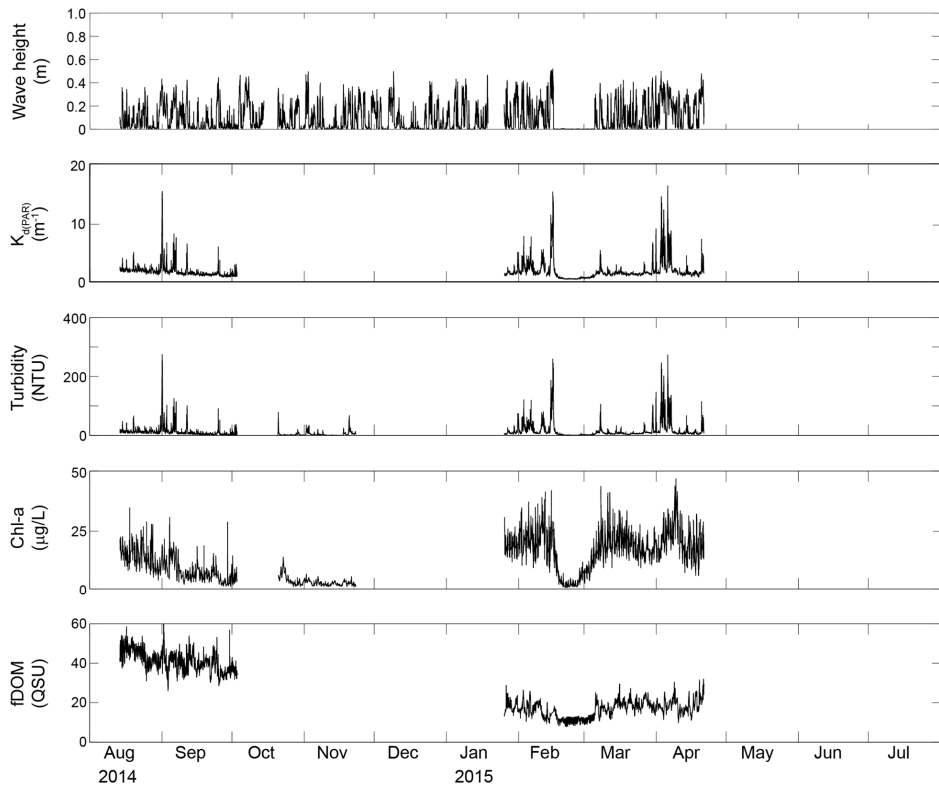
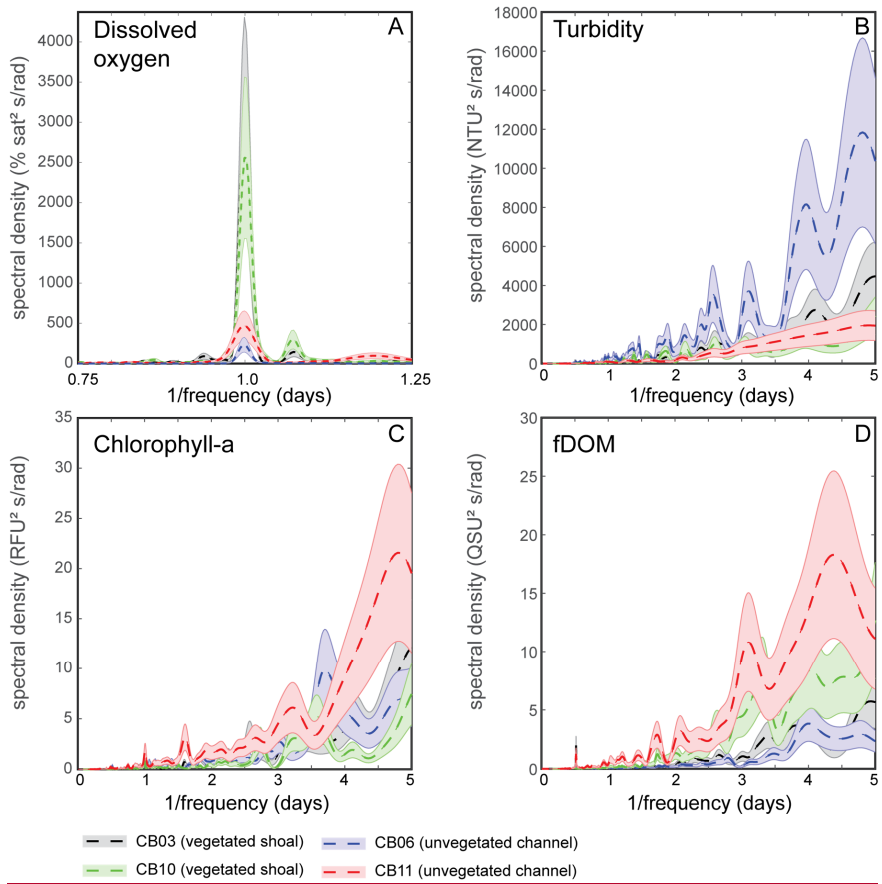


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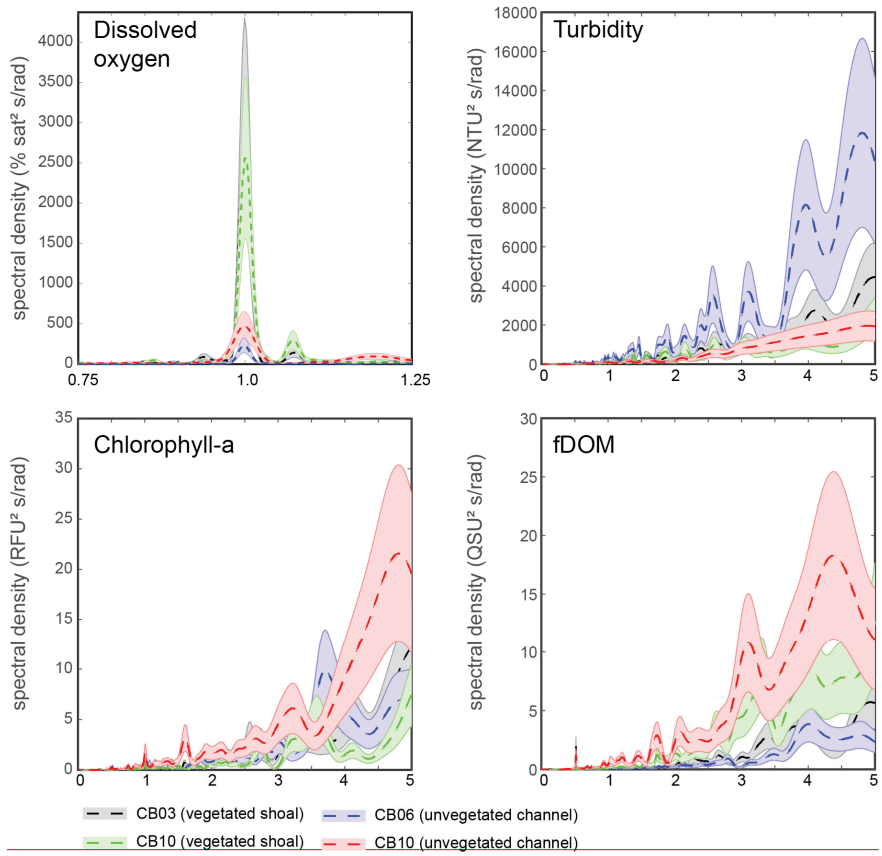


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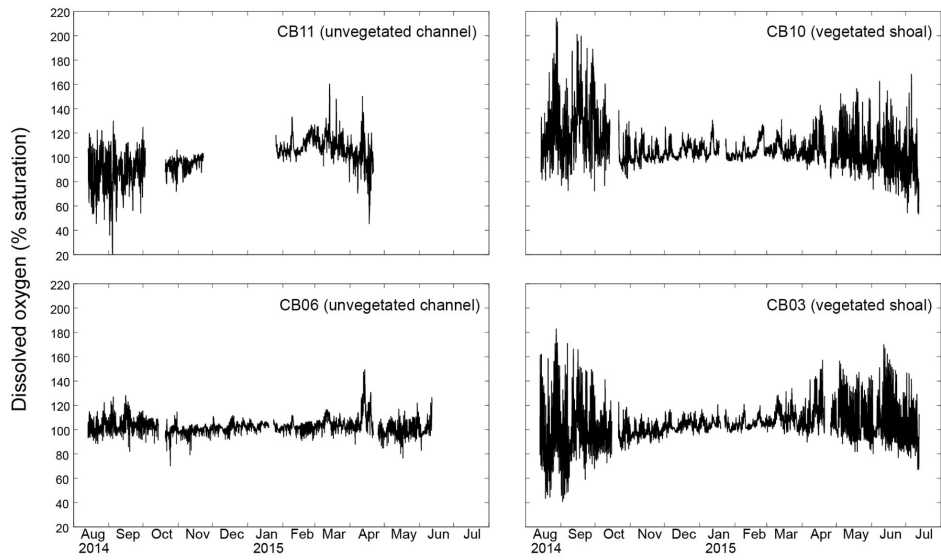


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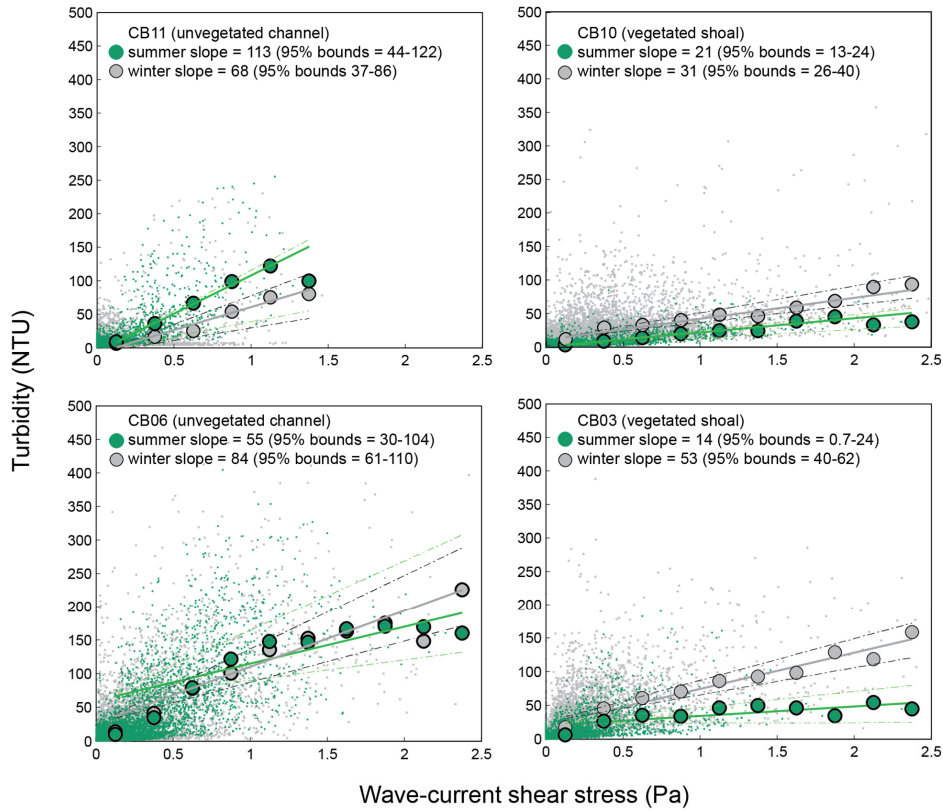
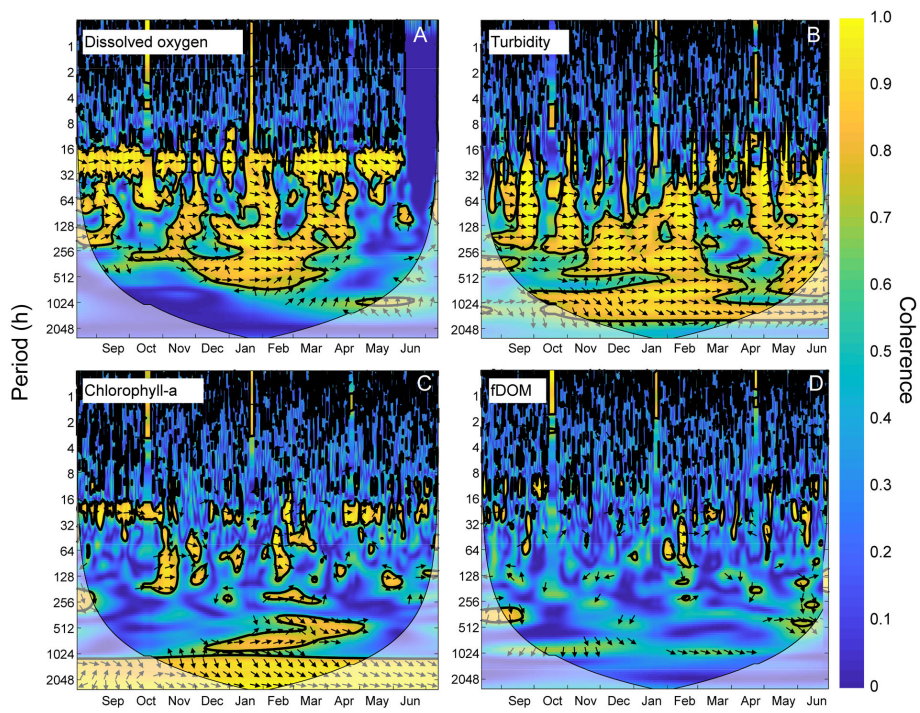


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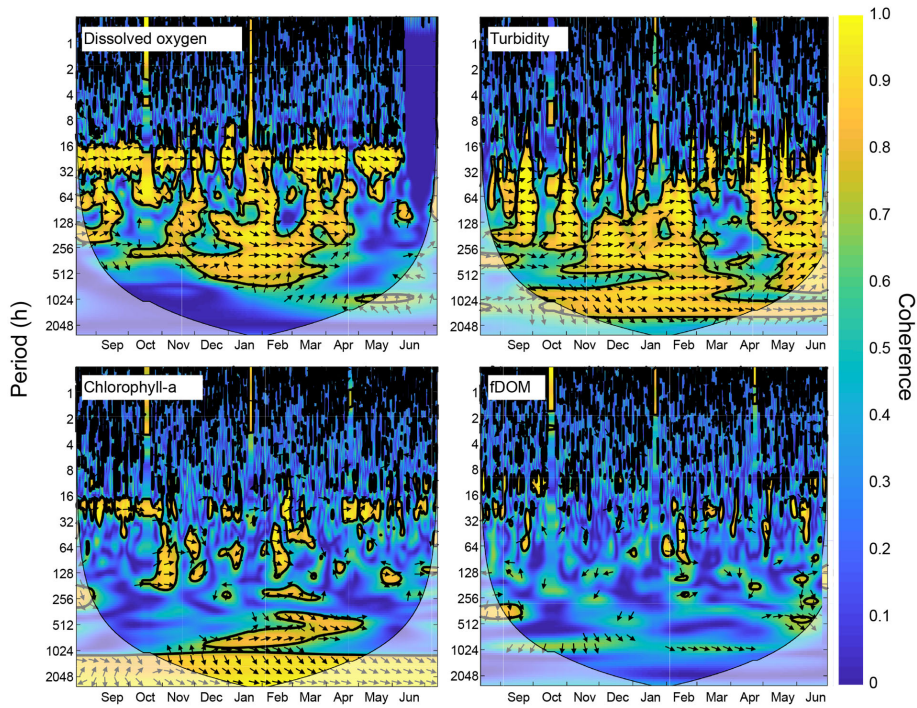


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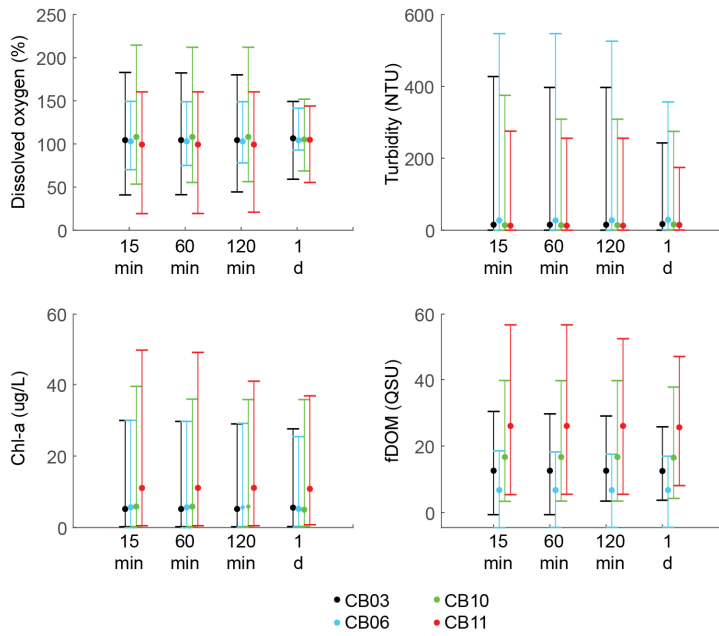


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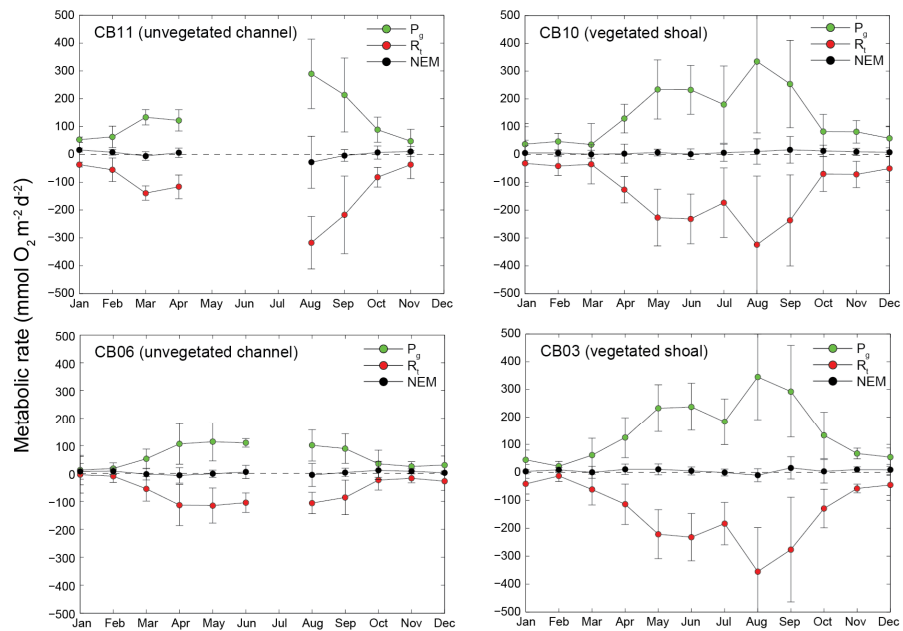


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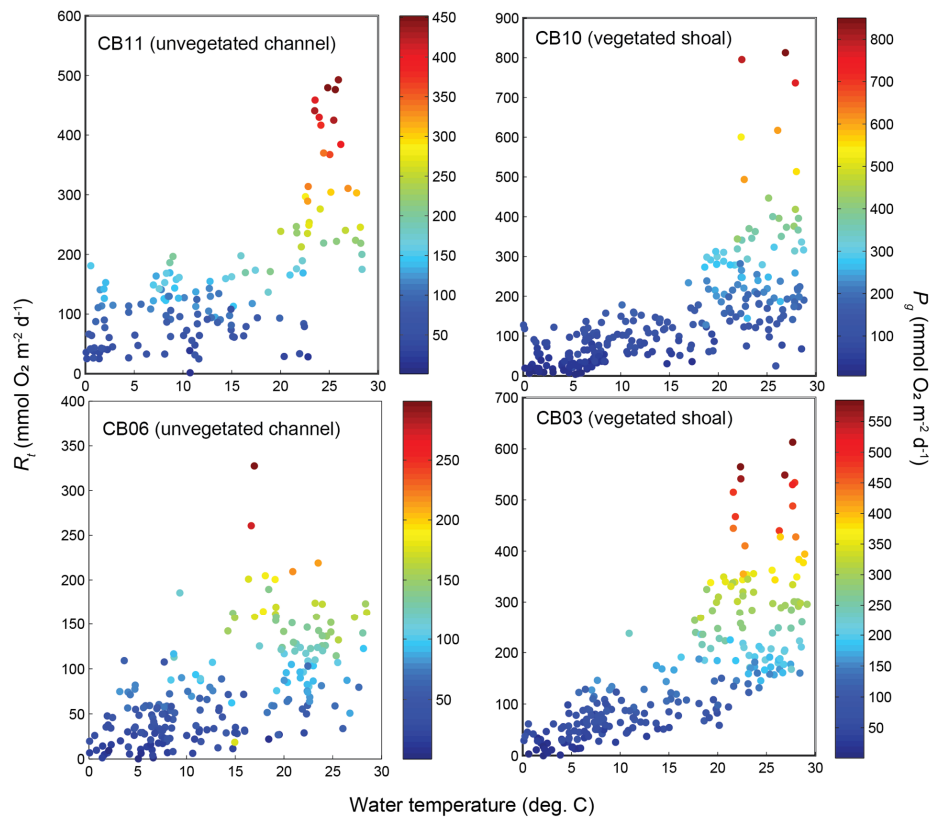


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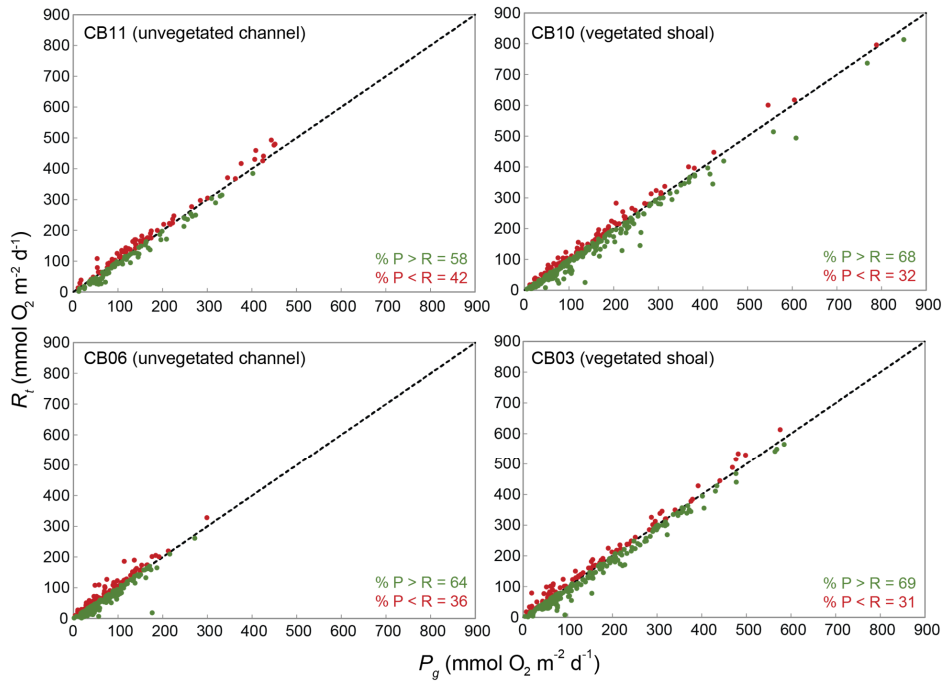


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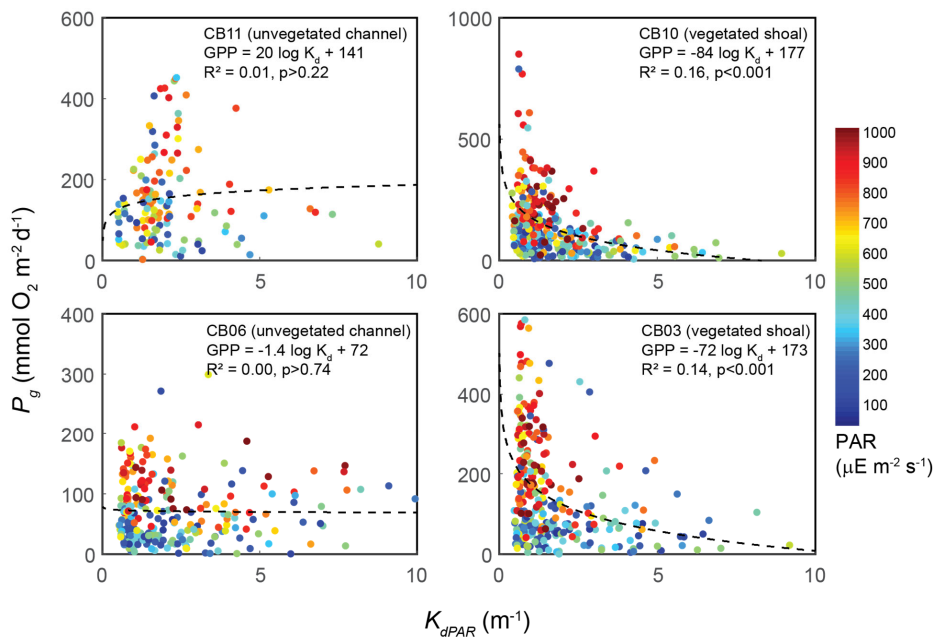


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