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## **Biomass-cover relationship for eelgrass meadows**

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14 *Zostera marina*

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18 **Abstract**

19 Eelgrass meadows play key roles in coastal ecosystems and the extent of the standing biomass is focal to  
20 address ecosystem functioning. Eelgrass cover is commonly assessed in marine monitoring programs while  
21 biomass sampling is destructive and expensive. Therefore, we have proposed a functional relationship that  
22 translates eelgrass cover into aboveground biomass using site-specific information on Secchi depth or light  
23 attenuation. The relationship was estimated by non-linear regression on 791 combined observations of  
24 eelgrass cover and biomass from eight different coastal sites in Denmark. Eelgrass biomass initially  
25 increased with cover and flattened out as cover exceeded 40-50% due to increased self-shading. Decreasing  
26 light energy with depth reduced the eelgrass biomass potential (assessed at 100% cover), and this reduction  
27 was stronger for coastal sites with lower water transparency. Moreover, the biomass potential varied  
28 seasonally from around 110-140 g DW m<sup>-2</sup> in spring months to a peak of 241 g DW m<sup>-2</sup> in August, consistent  
29 with other seasonal studies. The model explained 56% of the variation in log-transformed biomasses, but  
30 significant variation between coastal sites still remained, deviating between -23% and 39% from the mean  
31 relationship. These site-specific deviations could be due to differences in losses related to grazing, drifting  
32 algae and epiphytes, better light capture by dense canopies, as well as differences in how well light  
33 conditions within eelgrass meadows are represented by actual measurements of Secchi depth and light  
34 attenuation. The relationship can be employed to estimate eelgrass biomass of entire coastal ecosystems from  
35 observations of eelgrass cover and depth.

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## 37 **INTRODUCTION**

38 Eelgrass meadows play key functional roles in coastal ecosystems because eelgrass is an engineering species  
39 capable of modifying the benthic habitat structurally and metabolically (Gutierrez et al. 2011; Hemminga  
40 and Duarte 2000). The meadows increase the structural complexity of the seafloor and provide habitat for a  
41 variety of species, thereby stimulating biodiversity (Plummer et al. 2013). They are also highly productive  
42 and hence support secondary production and have a major effect on nutrient and carbon cycling in the coastal  
43 zone. In some areas eelgrass constitutes an important food source for birds (Clausen et al. 2012), but overall  
44 few species graze directly on eelgrass, and most of the biomass enters the detritivore food web or is buried  
45 (Cebrián et al. 1997). Moreover, eelgrass meadows dissipate wave energy and stabilize the sediments within  
46 and surrounding the meadows, which help protect the coast from erosion. The reduced wave energy further  
47 promotes particles trapping, and thereby contributes to increased water clarity as well as carbon sequestration  
48 in eelgrass sediments (van der Heide et al. 2011; McGlathery et al. 2012; Duarte et al. 2013). Seagrass  
49 sediments have indeed been identified as globally important carbon stocks (Fourqurean et al. 2012). These  
50 eelgrass-mediated ecosystem services depend on the standing biomass and the area cover of the meadows,  
51 which are, therefore, key variables to address in monitoring and management of coastal ecosystems.

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53 Mosaics of eelgrass patches and meadows occur on soft/sandy bottom of relatively protected waters from the  
54 shore and as deep as light levels allow, with the meadows confined to shallow depth ranges in turbid waters  
55 and extending deeper in clear waters (Duarte et al. 2007). The abundance of eelgrass typically declines  
56 exponentially with depth paralleling the extinction of light (Duarte 1991; Krause-Jensen et al. 2000).  
57 Physical exposure may reduce the abundance in shallow water, resulting in a bell-shaped distribution with  
58 depth (Krause-Jensen et al. 2003), and poor sediment quality or reduced oxygen levels may also lead to  
59 reduced eelgrass abundance (Koch 2001; Krause-Jensen et al. 2011). As the meadows respond to changing  
60 water and sediment quality, their distribution and abundance are often used as indicators of ecological status  
61 (Marbà et al. 2013).

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63 Biomass expressed as dry weight of carbon per m<sup>2</sup> seafloor is a relevant unit for quantifying eelgrass  
64 abundance and estimating structural and functional roles of the plant. Carbon biomass can also be quantified  
65 for other ecosystem components, which potentially allows addressing carbon flow through the ecosystem via  
66 coupling to process rates. But direct determination of biomass is destructive and resource-demanding as it  
67 requires harvesting the plants by divers and many biomass samples would be required to determine the large-  
68 scale eelgrass abundance. Eelgrass cover, assessed by divers, underwater video or remote sensing, is an  
69 alternative, non-destructive variable that is less costly compared to measuring biomass and suitable for  
70 assessment of eelgrass distribution and abundance at larger spatial scale. However, assessments of eelgrass  
71 cover do not couple as directly to ecosystem functions as biomass observations do. The combined benefit of  
72 non-destructive, large-scale and low-cost cover assessments and detailed biomass information relating more  
73 directly to ecological functions, could be obtained if robust relationships between coverage observations and  
74 biomass could be established to predict biomass distribution from coverage. For instance, observations of  
75 eelgrass cover along depth gradients from the shore and to the deepest extension of the meadows could be  
76 converted to biomass on the basis of such biomass-cover relationships. This would allow scaling cover to  
77 biomass over larger areas and potentially assessing eelgrass functions at an ecosystem scale.

78

79 Here we establish and test a generic relationship between eelgrass biomass and cover, taking into account  
80 factors such as depth, water clarity and time of the year. The relationship is developed based on monitoring  
81 data from the Danish coastal waters with combined information on eelgrass cover and biomass along depth  
82 gradients. The relationship allows the conversion of estimates of eelgrass cover to biomass along depth  
83 gradients. We thereby provide the basis for obtaining estimates of eelgrass biomass based on large-scale and  
84 long-term data sets on eelgrass cover. This opportunity is of great value e.g. in Denmark where the majority  
85 of monitoring data on eelgrass distribution and abundance is available solely as cover estimates.

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## 88 **MATERIALS AND METHODS**

89 Eelgrass cover has been monitored routinely since 1989 in ~50 different estuaries and coastal embayments  
90 (referred to as coastal sites in the following) within the Danish National Aquatic Monitoring and Assessment  
91 Program (DNAMAP). In addition to the regular eelgrass monitoring, data on the aboveground biomass of  
92 eelgrass were available from specific surveys in eight coastal sites which form the study areas of the current  
93 study (Table 1; Fig. S1). These data were extracted from the national marine monitoring database or from  
94 reports, in cases when data had not been submitted to the database. Eelgrass biomass was sampled between  
95 1990 and 2009.

96

97 Eelgrass biomass and cover were sampled in the growth season (March to October) by regional monitoring  
98 authorities with support from consultants. Sampling was carried out according to the same general protocol  
99 by experienced divers that regularly participate in intercalibration exercises as part of the monitoring  
100 program. Biomass samples were obtained by harvesting the aboveground biomass within a frame placed  
101 randomly within the eelgrass meadows where these covered the seafloor. The frame size varied between  
102 coastal sites from 0.09 to 0.25 m<sup>2</sup> and the number of samples per depth transect ranged between 1 and 24.  
103 The samples were dried at 105 °C (in few cases at 85 °C) for 24 h to constant weight and the biomass  
104 reported in g dry weight (DW) m<sup>-2</sup>. Before harvesting the biomass, the diver estimated the eelgrass cover  
105 within the frame in percent of the soft/sandy seafloor and recorded sampling depth. We quality-controlled  
106 the data by contacting the regional monitoring team and consulting monitoring reports to check that biomass  
107 estimates were correctly adjusted to varying frame sizes and represented aboveground biomass per m<sup>2</sup>. Data  
108 not conforming to the quality check were discarded. The resulting data set consisted of 852 combined  
109 biomass-cover observations distributed across eight coastal sites over the period 1990-2009 (Table 1).

110

111 The depth distribution of eelgrass biomass depends on the prevalent light conditions, and therefore seasonal  
112 means (March to September) of Secchi depth ( $Z_{SD}$ ) for the different coastal sites and years with eelgrass data

113 were calculated. Secchi depths were measured in all eight coastal sites as part of the DNAMAP; although not  
114 within the eelgrass meadows but at stations in the deeper part of the coastal site. In these shallow coastal  
115 sites the Secchi disk was occasionally visible at the bottom (censored data) and therefore censored data  
116 regression was employed (Carstensen 2010). Secchi depth means (March-September) were estimated for  
117 each year in all coastal sites and combined with the eelgrass biomass data, except for Kertinge Nor and  
118 Helnæs Bugt where Secchi depth observations were too few in the year (1996) with eelgrass biomass data  
119 and a Secchi depth mean over multiple surrounding years was calculated instead. Additionally, the light  
120 attenuation coefficient ( $K_d$ ) has been estimated from underwater PAR (photosynthetically active radiation)  
121 profiles as part of DNAMAP in more recent years (Pedersen et al. 2014), which only partially overlap the  
122 biomass samples in time as opposed to Secchi depth, which has been monitored regularly as part of  
123 DNAMAP. Mean values of the product between  $K_d$  and  $Z_{SD}$  for the eight different sites were calculated for  
124 comparison with the biomass model described below.

125

### 126 **Eelgrass biomass model**

127 The area-specific eelgrass biomass ( $B(C,Z)$ ) essentially depends on the density and size of eelgrass shoots,  
128 which is reflected in eelgrass cover ( $C(Z)$ ). Eelgrass growth and, hence to a large extent, biomass and cover  
129 mainly depends on the light energy reaching the eelgrass (Sand-Jensen and Borum 1991; Krause-Jensen et  
130 al. 2000), and this is a function of the depth of the sample ( $Z$ ) as well as the attenuation of light in the water  
131 column, expressed by the light attenuation coefficient ( $K_d$ ) which can be approximated from the Secchi depth  
132 ( $Z_{SD}$ ) (see below). Eelgrass biomass varies dynamically as a function of growth and loss processes, but we  
133 assumed that the balance between eelgrass growth and respiration can be described by a seasonal model with  
134 a depth component accounting for the reduced growth with lower light, whereas other loss processes than  
135 respiration, such as grazing and physical destruction, and shading by drifting macroalgae are unrelated to  
136 light and assumed to be reflected directly by the biomass and cover estimates. Hence, the model describes  
137 steady-state conditions for eelgrass biomass in each month (March-October). The relationship between  
138 eelgrass biomass and eelgrass cover, depth, and Secchi depth is derived step-by-step in the following.

139

140 We assumed that eelgrass biomass is related to eelgrass cover through a saturation-type of response,  
 141 displaying almost proportionality at low eelgrass coverage (no competition for light) but levelling off at  
 142 increasing cover due to increased competition for light. This can be formulated as:

$$143 \quad B(C, Z) = B_{max}(Z) \cdot \left(1 - \exp\left(-\frac{C(Z)}{k_c}\right)\right) \quad \text{Eq. (1)}$$

144 where  $B_{max}(Z)$  is the maximum attainable biomass at a given depth, and  $k_c$  is a parameter describing how fast  
 145 the relationship between biomass and cover levels off. Eelgrass biomass will approach  $B_{max}(Z)$  as  $C(Z)$   
 146 increases towards 100%.

147

148 The maximum biomass as a function of depth,  $B_{max}(Z)$ , depends on the light-regulated reduction of biomass  
 149 with depth, assuming that a certain light level can sustain a certain biomass (steady-state assumption). The  
 150 effect of light-limited growth can be modeled using a simple hyperbolic tangent function (Platt and Jassby  
 151 1976)

$$152 \quad B_{max}(Z) = B_{max} \cdot \tanh\left(\frac{I(Z)}{I_{sat}}\right) \quad \text{Eq. (2)}$$

153 where  $B_{max}$  is the maximum attainable biomass when there is no light limitation,  $I(Z)$  is the irradiance at  
 154 depth  $Z$ , and  $I_{sat}$  is a parameter equal to the irradiance level yielding 76 % of  $B_{max}$  (i.e.  $\tanh(1)=0.76$ ). Using  
 155 Lambert-Beer's law with  $K_d$  describing the light attenuation with depth the expression becomes

$$156 \quad B_{max}(Z) = B_{max} \cdot \tanh\left(\frac{I_0}{I_{sat}} \cdot \exp(-K_d \cdot Z)\right) \quad \text{Eq. (3)}$$

157

158 Assuming that the Secchi depth ( $Z_{SD}$ ) represents 20 % (see discussion) of the surface irradiance (i.e.

$$159 \quad K_d = \frac{-\log(0.2)}{Z_{SD}} \text{) the maximum eelgrass biomass becomes a function of } Z_{SD}$$

$$160 \quad B_{max}(Z) = B_{max} \cdot \tanh\left(\frac{I_0}{I_{sat}} \cdot \exp\left(\log(0.2) \cdot \frac{Z}{Z_{SD}}\right)\right) \quad \text{Eq. (4)}$$

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162 Thus, combining Equations (1) and (4) the eelgrass biomass can be formulated as function of coverage,  
 163 Secchi depth and depth as:

164  $B(C, Z) = B_{max} \cdot \tanh\left(\frac{I_0}{I_{sat}} \cdot \exp\left(\log(0.2) \cdot \frac{Z}{Z_{SD}}\right)\right) \cdot \left(1 - \exp\left(-\frac{C(Z)}{k_C}\right)\right)$  Eq. (5)

165 Measurements of eelgrass biomass typically have a right-skewed distribution with variation between  
 166 replicate samples increasing with the mean. For analyzing measured biomasses it is therefore more relevant  
 167 to consider the log-transform of the biomass

168  $\log(B(C, Z)) = \log(B_{max}) + \log\left(\tanh\left(\frac{I_0}{I_{sat}} \cdot \exp\left(\log(0.2) \cdot \frac{Z}{Z_{SD}}\right)\right)\right) + \log\left(1 - \exp\left(-\frac{C(Z)}{k_C}\right)\right)$  Eq. (6)

169 Eelgrass biomass accumulates during months when production exceeds respiration, which results in a  
 170 seasonal effect on biomass in addition to the direct effect of light attenuation in the water column. This  
 171 seasonal variation describing the balance between growth and respiration was modeled by estimating the  
 172 parameter  $B_{max}$  specific to each month with biomass observations.

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174 This non-linear model was fitted using the combined data set of eelgrass biomass, cover and Secchi depth by  
 175 means of non-linear maximum likelihood regression (PROC MODEL in SAS 9.3; SAS Institute, Cary, NC).

176 Model parameters were  $k_C$  and  $I_0/I_{sat}$  (describing the relative amount of surface radiation where light reduces  
 177 growth by 24%) as well as eight month-specific parameters for  $B_{max}$ . The non-linear estimation routine  
 178 iteratively found the optimal parameter estimates by ordinary least squares estimation. The eelgrass model  
 179 was tested by examining the distribution of the residuals, plotting them versus depth and cover and analyzing  
 180 their differences among coastal sites. The nature of the depth and cover relationships was assessed by  
 181 plotting the marginal relationships of eelgrass biomass versus the two predictors (cover and depth), i.e.  
 182 calculating eelgrass biomass adjusted for the predicted effect of cover and depth as well as interannual  
 183 variations in Secchi depth. Finally, the applicability of the model was tested by applying the estimated  
 184 relationship to four different transects, where cover and depth had been recorded.

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## 190 RESULTS

191 The combined data set (791 observations) represented a broad span of depths (0.7-7.8 m), Secchi depths  
192 (2.2-8.7 m), eelgrass cover (1-100%), and eelgrass biomass (0.2-573 g DW m<sup>-2</sup>) (Table 1). Although there  
193 were differences in sampling efforts across the 8 coastal sites, the data set appeared reasonably balanced and  
194 not biased towards a single coastal site. All of the eight different months (March-October) used to describe  
195 the seasonal variation in  $B_{max}$  were sampled at least at two coastal sites.

196 Eelgrass biomass varied over three orders of magnitude with an overall tendency to decline at depths >2 m  
197 (Fig. 1a). For observations representing full (100%) eelgrass cover, the biomass varied from 26 to 546 g DW  
198 m<sup>-2</sup>, while the highest biomass observation was actually measured for a cover of 85%. Eelgrass cover  
199 spanned broadly across the entire depth range (Fig. 1b), which allowed for estimating the eelgrass biomass  
200 dependency on depth and cover with small risk of correlated parameter estimates, particularly  $k_c$  and  $I_0/I_{sat}$ .  
201 Eelgrass biomass increased with cover across different depth strata in a similar manner, showing an initial  
202 increase in biomass at low eelgrass cover before flattening when the cover exceeded 40-50% (Fig. 2).

203 Examining the data and the residuals generated from Eq. (6), two observations were identified as outliers  
204 (Fig. 2); both having eelgrass biomass above 100 g DW m<sup>-2</sup> at a low cover of 1% and 10%. These  
205 observations were subsequently excluded from the model estimation.

206 The eelgrass biomass modelled from Eq. 6 explained 56% ( $R^2=0.56$ ) of the total variation in the log-  
207 transformed biomass observations without any systematic departures over the prediction range (Fig. 3). The  
208 residual variation was considerable (Root MSE=0.6469 on the log-scale), corresponding to about  $\pm 90\%$   
209 variation on individual observations. All parameter estimates were strongly significant (Table 2) and  
210 importantly, the correlation between the parameter estimates of  $k_c$  and  $I_0/I_{sat}$  was small ( $r=0.1243$ ). This  
211 implied that the depth and cover terms of Eq. (6) were determined almost independently of each other. The  
212 parameter estimate of  $k_c$  described that the eelgrass biomass reached a “saturation point” for eelgrass cover  
213 around 54%. Similarly, the parameter estimate of  $I_0/I_{sat}$  suggested that light limitation became important at

214 depths where the surface irradiance was reduced to less than 30%. The monthly parameter estimates for  $B_{max}$   
215 displayed a significant (Wald test statistic=57.75;  $p<0.0001$ ) and expected seasonal pattern increasing from  
216 around 110-140 g DW m<sup>-2</sup> in the spring months to a peak of 241 g DW m<sup>-2</sup> in August and then declined  
217 gradually in September and October to a level similar to that of June and July (~170 g DW m<sup>-2</sup>; Fig. 4).

218 The residuals of biomass estimates from Eq. (6) followed the normal distribution closely and did not show  
219 any systematic departures over the ranges of depth and cover (data not shown). However, the residuals  
220 varied significantly among coastal sites ( $F_{8,781}=14.15$ ;  $p<0.0001$ ). Accounting for site-specific differences  
221 only reduced the remaining residual variation slightly (Root MSE=0.6038 on the log-scale), corresponding to  
222  $\pm 83\%$  variation on individual observations. Thus, eelgrass biomass observations were quite variable with a  
223 considerable amount of variation unaccounted for. Mean differences among coastal sites were between -0.26  
224 and 0.33 on the log-scale (Table 3), corresponding to -23% and 39% deviation from the biomass-cover  
225 relationship estimated over the entire data set. So in addition to the estimated relationship representing the  
226 average across all coastal sites, there were site-specific characteristics yielding overall higher or lower  
227 eelgrass biomass.

228 The marginal relationships between eelgrass biomass and depth (accounting for variations in cover, Secchi  
229 depth, and month of sampling through the model) showed different decreases with depth among the coastal  
230 sites, i.e. different “biomass attenuation” with depth (Fig. 5), which were caused by differences in light  
231 attenuation among sites. In Køge Bugt and The Sound that had the highest water transparency (Table 1),  
232 eelgrass biomass only decreased slightly between 0.7 and 7.8 m depth. Roskilde Fjord and Odense Fjord had  
233 less clear waters and eelgrass biomass decreased already at depths >2 m (Fig. 5), although for Odense Fjord  
234 this was only clear from the estimated relationship as eelgrass biomass was not sampled deeper than 3.2 m.  
235 Differences between the estimated marginal relationships and observations, adjusted for variations in  
236 eelgrass cover, interannual variation in Secchi depth and month of sampling, were large for Køge Bugt  
237 (residuals 35% above the average) and The Sound (residual 21% below the average) (Fig. 5, Table 3). This  
238 site-specific bias was smaller for Roskilde Fjord (-2%) and Odense Fjord (-11%).

239 Similarly, the marginal relationships showed a steep proportional increase in eelgrass biomass with eelgrass  
240 cover in the range 0-20%, followed by a more gradual increase that almost flattened out when eelgrass cover  
241 exceeded 40-50% (Fig. 6). The relationships for the different sites were quite similar, since site-specific  
242 differences were based on the ratio between eelgrass sampling depth and Secchi depth ( $\frac{z}{z_{SD}}$  in Eq. 6), that  
243 exhibited small variations among sites (Table 1). As above, the relationship for Køge Bugt underestimated  
244 eelgrass biomass observations, whereas the relationships for Roskilde Fjord, Odense Fjord and The Sound  
245 overestimated biomass observations (cf. Table 3).

246 We calculated eelgrass biomass along four different transects where depth and eelgrass cover was monitored  
247 as part of the national monitoring program (Fig. 7). All transects started at shallow depths and extended  
248 beyond the eelgrass depth limits; however, depth did not increase continuously due to bottom topography.  
249 Eelgrass biomass largely followed variations in eelgrass cover, displaying shifts between dense meadows  
250 and bare sediments, but with relatively smaller biomass at deeper depths, which was most clearly seen in  
251 Køge Bugt and Roskilde Fjord (Fig. 7a,c). Eelgrass biomass was predicted at 100-200 g DW m<sup>-2</sup> in the dense  
252 meadows, whereas the less dense patches had lower biomass.

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## 256 **DISCUSSION**

257 We developed a general model that describes eelgrass biomass based on information on eelgrass cover for a  
258 given depth and season and with associated information on water transparency of the coastal site. Hence, the  
259 model allows a general conversion of eelgrass cover data to biomass. This model may constitute a useful tool  
260 as information on eelgrass biomass is highly valuable for addressing functional aspects of eelgrass meadows  
261 but sampling of biomass is destructive and costly while eelgrass cover is much easier and less costly to  
262 assess on the large scale. As eelgrass biomass and cover at given depths are highly dependent on light  
263 attenuation, the inclusion of a light attenuation term in the model enables a realistic fit to local light  
264 conditions and makes the model generally applicable for areas showing seasonality of eelgrass biomass  
265 similar to that of mid-latitude Danish coastal waters. The model also allows for converting aerial surveys of  
266 eelgrass cover into biomass, provided that the bathymetry and Secchi depth of the area are known.

### 267 *Light regulation of eelgrass biomass*

268 The model provided estimates of the light level ( $I_0/I_{sat}$ ) needed to support maximum eelgrass biomass, based  
269 on the assumption that Secchi depths represent 20% of the surface irradiance (PAR). This assumption  
270 corresponds to  $K_d \cdot Z_{SD} = -\log(0.2) = 1.61$ . Although this value corresponds to values reported for open  
271 seawater (~1.5-1.7),  $K_d \cdot Z_{SD}$  is generally higher in estuaries and coastal waters (~1.9-3.9) influenced by  
272 dissolved organic matter from land (Koenings and Edmundson 1991). The more recent monitoring  
273 observations of the light attenuation coefficient suggest that  $K_d \cdot Z_{SD}$  ranges from 1.7 in Odense Fjord and  
274 South Funen Archipelago to 2.1 in Roskilde Fjord (data not shown), corresponding to 12-18% of surface  
275 irradiance at the Secchi depth. These values are higher than the value employed in the eelgrass biomass  
276 model, but  $K_d$  and  $Z_{SD}$  are measured at deeper monitoring stations centrally located in the study sites,  
277 whereas eelgrass biomass was sampled in shallower nearshore environments, where sediment resuspension is  
278 more pronounced. Increased scattering from resuspended particles in the shallow environments reduces  
279  $K_d \cdot Z_{SD}$  (Gallegos et al. 2011), justifying the lower value applied in the model. Furthermore, eelgrass

280 biomass was sampled over a period (1990-2009) when nutrient inputs from Denmark were significantly  
281 reduced (Carstensen et al. 2006), which also led to a decrease in the ratio between scattering and absorbance  
282 (Pedersen et al. 2014). This suggests that  $K_d \cdot z_{SD}$  has increased over time and therefore was lower during the  
283 period of eelgrass biomass sampling, consistent with Pedersen et al. (2014) reporting an increase in  $K_d \cdot z_{SD}$   
284 in Roskilde Fjord from 1.8 (1985) to 2.2 (2008-2009). Unfortunately, light measurements within the eelgrass  
285 meadows were not available, but  $K_d \cdot z_{SD} = 1.61$  is not unrealistic given the arguments raised above.

286 The effect of light attenuation on eelgrass biomass was described as a biomass attenuation (Duarte 1991)  
287 with depth, and Eq. (2) and the  $I_0/I_{sat}$ -value (Table 2) suggest that 73% of  $B_{max}$  can be obtained at 30% of the  
288 surface irradiation, 55% of  $B_{max}$  can be obtained at 20% surface irradiation, and 30% of  $B_{max}$  can be obtained  
289 at 10% surface irradiation. The nature of the light-dependency for eelgrass biomass is poorly documented in  
290 the literature but our  $I_0/I_{sat}$ -value is possibly larger than the light level needed to support the depth limit of  
291 eelgrass, for which there is considerable documentation. Based on laboratory studies Olesen (1996) found  
292 that 11% of surface irradiance was needed to support eelgrass growth on an annual basis. Field studies have  
293 reported somewhat higher light levels at the depth limit probably because loss of biomass due to other factors  
294 than respiration contributes to defining the depth limit. Assuming 10% of the surface light at the Secchi  
295 depth, Nielsen et al. (2002) showed that 18% surface irradiance was available at the depth limit of Danish  
296 eelgrass meadows (Secchi depth ~4 m), while Krause-Jensen et al. (2011) found that 28% of surface  
297 irradiance was available at the average depth limit of eelgrass in Danish coastal waters (Secchi depth  
298 between 2.5 and 8 m). Combining these studies with the model results suggests that the eelgrass biomass at  
299 the depth limit represents 50-70% of  $B_{max}$ . Obviously, the biomass attenuation component is not useful for  
300 predicting depth limits (the biomass model is essentially unbounded towards deeper depths, cf. Eq. 2) and  
301 depth limits are described through disappearance of eelgrass cover, as input to the model.

### 302 *Seasonal variation in biomass*

303 The increase in eelgrass biomass from May to August followed by a decline in September and October fits  
304 well with the results from other studies at similar latitudes, which also show a biomass peak in

305 August/September (Sand-Jensen 1975; Olesen and Sand-Jensen 1994; Pedersen and Borum 1993; Clausen et  
306 al. 2014). The increase in eelgrass biomass from May to August reflects the main growth season for eelgrass  
307 with good light conditions and minimal physical exposure while the decline in biomass during autumn is a  
308 combined effect from decreasing light levels and losses of leaves and shoots during autumn storms. The  
309 timing of the biomass peak depends on latitude with earlier timing in the southern end of the distribution  
310 range and later timing towards the Arctic (Clausen et al. 2014). Hence, the model would need adjustment of  
311 the seasonal pattern if applied to eelgrass cover data from higher or lower latitudes.

312 The  $B_{max}$  estimate for August (Table 2) suggests a mean eelgrass biomass potential of 241 g DW m<sup>-2</sup> for  
313 100% eelgrass cover and no light limitation. This is consistent with Olesen and Sand-Jensen (1994), who  
314 investigated a broad selection of 40 temperate eelgrass meadows and found an average aboveground biomass  
315 of 245 g DW m<sup>-2</sup> (10-90% percentile range: 111-391 g DW m<sup>-2</sup>). Olesen and Sand-Jensen (1994) concluded  
316 that the maximum attainable biomass of eelgrass meadows during midsummer was relatively uniform among  
317 populations because self-shading within the stands sets an upper limit for biomass development. Probably for  
318 the same reason maximum eelgrass biomass shows no significant change with latitude (Clausen et al. 2014).

319 The average increase in eelgrass biomass from May to August was 129 g DW m<sup>-2</sup> and represented about a  
320 doubling. Such marked seasonality from spring to summer is characteristic for eelgrass meadows (Olesen  
321 and Sand-Jensen 1994; Duarte and Chiscano 1999; Clausen et al. 2014). For example, Sand-Jensen (1975)  
322 reported a quadrupling of the aboveground eelgrass biomass and a doubling of the belowground biomass  
323 from March to August in a shallow Danish embayment, paralleling a total production of about 1100 g DW  
324 m<sup>-2</sup> from April to October.

### 325 *Spatial variation and applicability of the model*

326 The estimated model translates eelgrass cover and depth into eelgrass biomass, provided that the Secchi  
327 depth is also known. Analysis of the residuals suggests that the model could introduce a substantial bias (-  
328 23%-39%, Table 3) in such biomass estimates, but this bias may also result from other factors (input data)  
329 than model bias. As discussed above, Secchi depth was monitored at deeper stations for all sites and this may

330 introduce a bias, because these data represent light conditions in the eelgrass meadows with varying degree  
331 of bias. Differences among areas in  $K_d \cdot z_{SD}$  for converting Secchi depths to light attenuation may also  
332 introduce a bias, and it is therefore better to use Eq. (3) for the light attenuation, if  $K_d$  is measured.

333 Variation among sites in the biomass-cover relationship was observed (Table 3). It cannot be excluded that  
334 differences between observers and slight differences in methods explain part of this variation as many divers  
335 contributed to the survey. If differences between observers could be ignored, site-specific variation in the  
336 biomass for a given cover likely reflects differences between sites in loss factors and/or growth conditions  
337 unrelated to light-attenuation in the water column (which the model accounts for). One such loss factor could  
338 be grazing, e.g. by water fowl, which would affect biomass without necessarily affecting cover and therefore  
339 would result in the model overestimating the actual biomass at such sites. Drifting algae, which tend to  
340 accumulate in eelgrass beds (Rasmussen et al. 2013, 2015) might also reduce the aboveground biomass  
341 without affecting cover and might be part of the explanation why the model overestimates the eelgrass  
342 biomass at sites such as Køge Bugt and Kertinge Nor, which have been known for large occurrences of  
343 drifting filamentous brown algae during the study period (Riisgård et al. 1995). By contrast, well-established  
344 and dense eelgrass meadows tend to facilitate their own growth and resilience through positive feed-backs,  
345 which may involve more efficient light utilization in the closed canopy, increased sedimentation and  
346 improved recycling of nutrients as well as increased top-down control of epiphytic algae on leaf surfaces  
347 (Gutierrez et al. 2011; Sand-Jensen et al. 2007; van der Heide et al. 2011) and may thereby maintain a larger  
348 biomass at a given cover. Such positive feed-backs could explain why the model underestimated the biomass  
349 at the South Funen Archipelago and The Sound, known for their well-developed eelgrass meadows (Krause-  
350 Jensen et al. 2000).

351 The perspective is to employ the model to scale-up eelgrass cover estimates for an entire coastal ecosystem  
352 to calculate nutrient and carbon budgets for eelgrass, and to compare these to similar budgets for the water  
353 column and other biological components (Neckles et al. 2012). This will allow to quantitatively assess the  
354 role of eelgrass in the biogeochemical cycling of elements in coastal ecosystems.

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449

**Table 1** Overview of data sets used for estimating the eelgrass aboveground biomass-cover relationship. For each coastal site is listed the number of observations, distributed over number of years, number of transects and the specific months sampled (by month number, i.e. 3=March, 4=April, etc.). Means and ranges for the variables used in the relationship are shown. Annual means (March-September) of Secchi depth were used and no range is given for sites with a single sampling year.

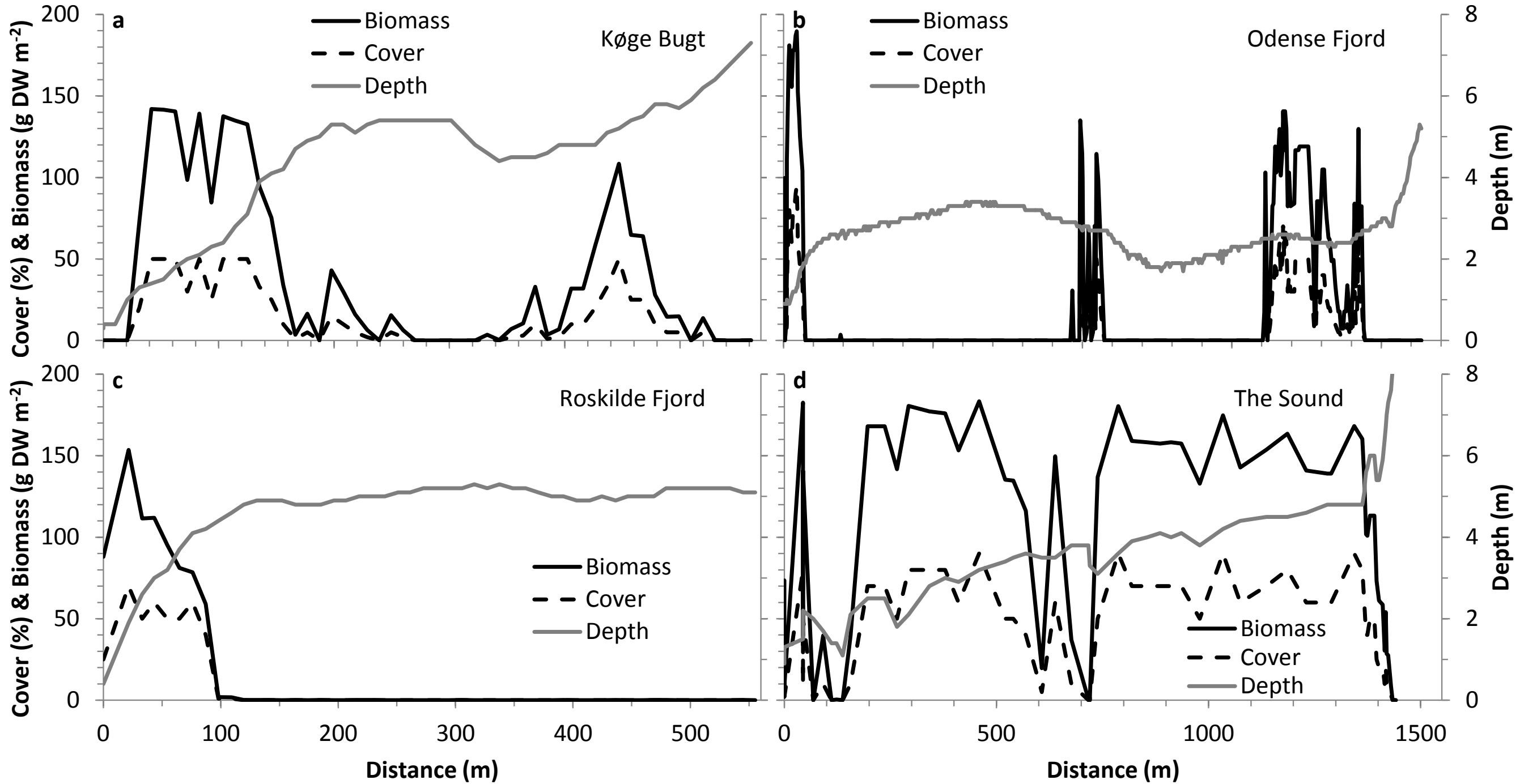
Coastal site	# of years	Months sampled	# of transect	# of obs.	Secchi depth (m)		Transect depth (m)		Cover (%)		Biomass (g DW m <sup>-2</sup> )	
					Mean	Range	Mean	Range	Mean	Range	Mean	Range
<b>Flensburg Fjord</b>	7	3-6, 8-9	1	83	5.0	4.4-5.4	2.8	1.0-5.0	66	1-100	148	2.8-546
<b>Helnæs Bugt</b>	1	9	7	38	5.6		1.5	0.7-2.5	72	50-95	176	64-312
<b>Kertinge Nor</b>	1	9	4	12	2.2		2.0	0.7-2.5	100	100	122	31-240
<b>Køge Bugt</b>	6	4-10	4	156	7.1	6.6-7.5	4.3	0.7-7.8	45	1-100	113	0.2-573
<b>Odense Fjord</b>	3	5-6, 8-10	3	120	3.2	2.8-3.5	1.6	0.8-3.2	81	20-100	128	5.0-388
<b>Roskilde Fjord</b>	6	3-8	10	105	3.9	2.7-4.5	2.4	1.0-6.0	46	2-100	113	2.9-399
<b>South Funen</b>	1	9	2	24	6.5		3.7	2.6-5.2	93	75-100	91	34-178
<b>Archipelago</b>												
<b>The Sound</b>	10	5-9	11	253	7.8	7.4-8.7	3.3	1.0-6.3	86	10-100	155	3.9-511

**Table 2** Parameter estimates obtained from Eq. (6) using 791 eelgrass biomass observations (log-transformed). SE=standard error of the parameter estimate.

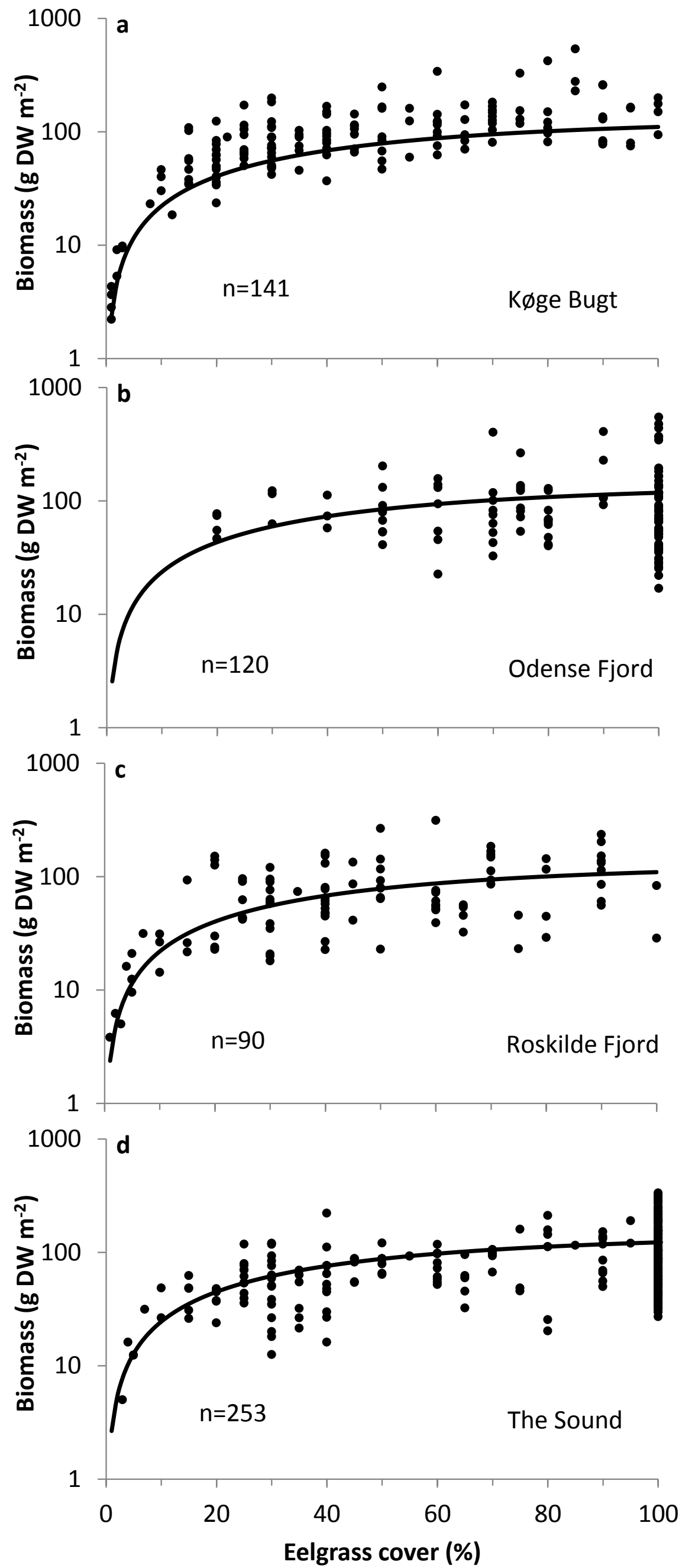
<b>Parameter</b>	<b>Estimate</b>	<b>SE</b>	<b><i>t</i>-test</b>	<b>Probability</b>
$k_c$	54.26	7.82	6.94	<.0001
$I_0/I_{sat}$	3.12	0.38	8.28	<.0001
$\log(B_{max})$ (Mar)	4.79	0.15	32.04	<.0001
$\log(B_{max})$ (Apr)	4.93	0.16	30.33	<.0001
$\log(B_{max})$ (May)	4.72	0.13	35.06	<.0001
$\log(B_{max})$ (Jun)	5.16	0.12	44.50	<.0001
$\log(B_{max})$ (Jul)	5.13	0.12	44.01	<.0001
$\log(B_{max})$ (Aug)	5.48	0.08	65.91	<.0001
$\log(B_{max})$ (Sep)	5.12	0.10	52.79	<.0001
$\log(B_{max})$ (Oct)	5.20	0.12	42.67	<.0001

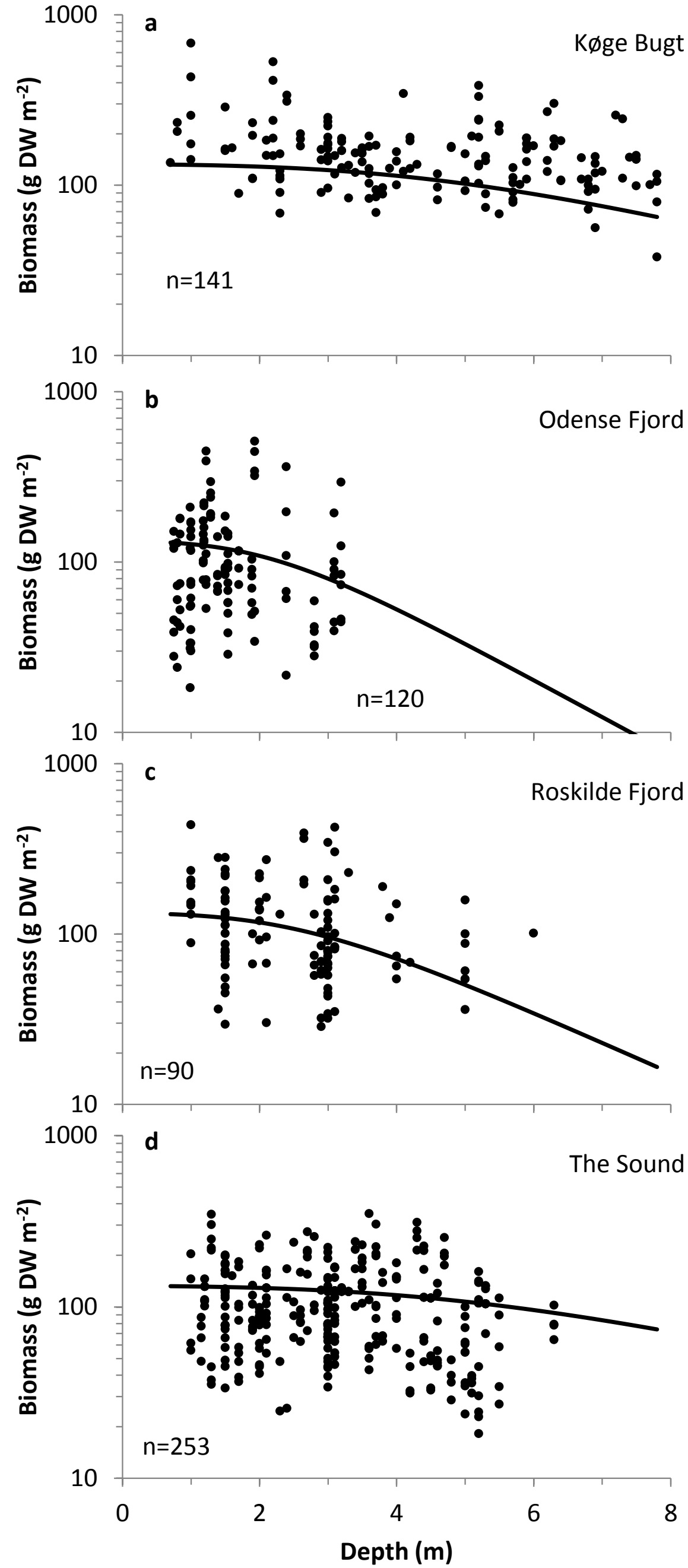
**Table 3** Residual variation from estimating Eq. (6) among coastal sites. For each coastal site is listed the mean of the residuals, the standard error of the mean (SE), the  $t$ -statistic for testing if the mean equals zero and its associated probability.

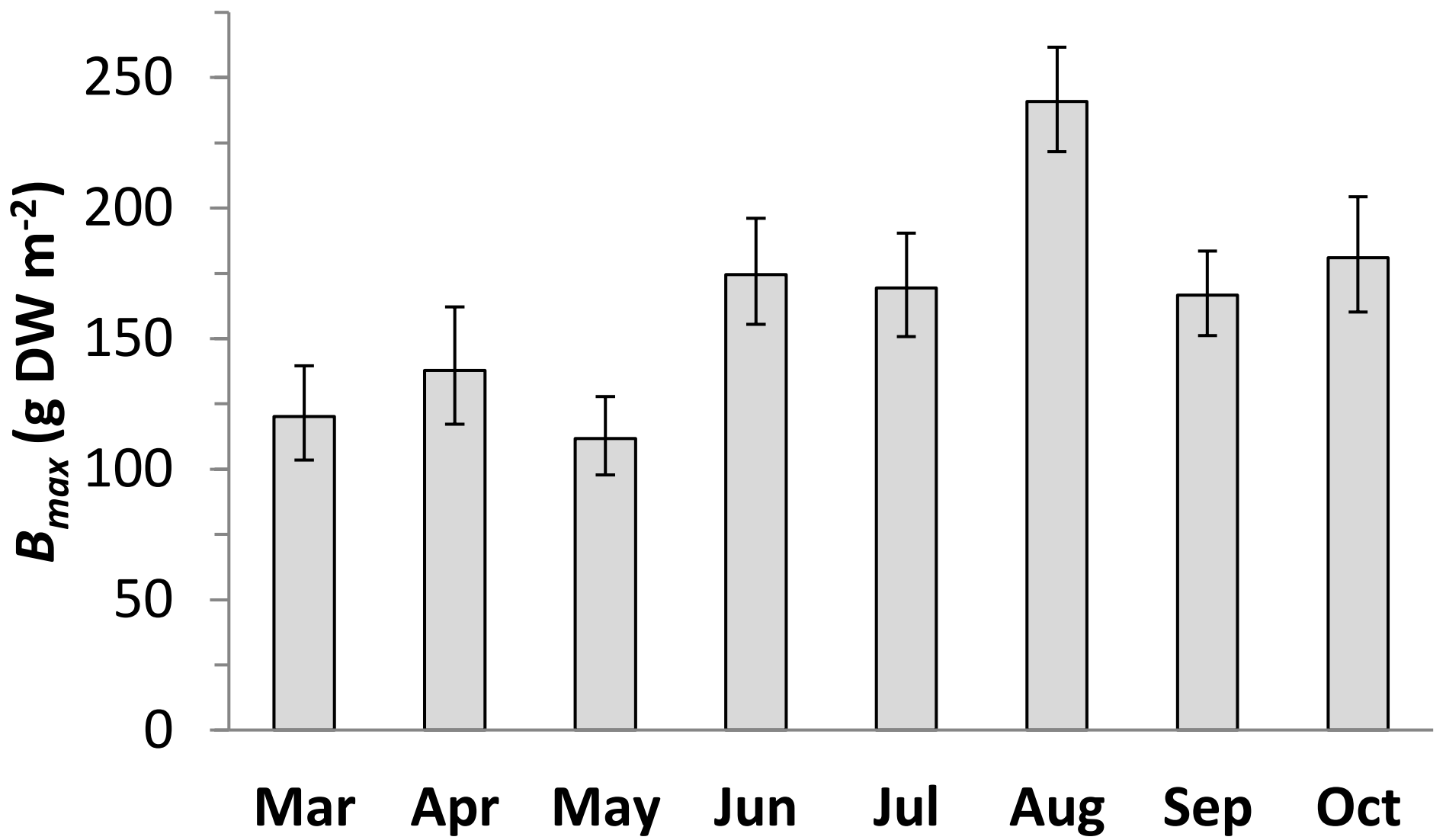
Coastal site	Mean	SE	$t$	$p$
Flensborg Fjord	0.256	0.067	3.82	0.0001
Helnæs Bugt	0.332	0.098	3.38	0.0007
Kertinge Nor	0.190	0.174	1.09	0.2770
Køge Bugt	0.298	0.048	6.16	<0.0001
Odense Fjord	-0.117	0.055	-2.13	0.0335
Roskilde Fjord	-0.016	0.059	-0.28	0.7823
South Funen Archipelago	-0.258	0.123	-2.10	0.0363
The Sound	-0.237	0.038	-6.25	<0.0001

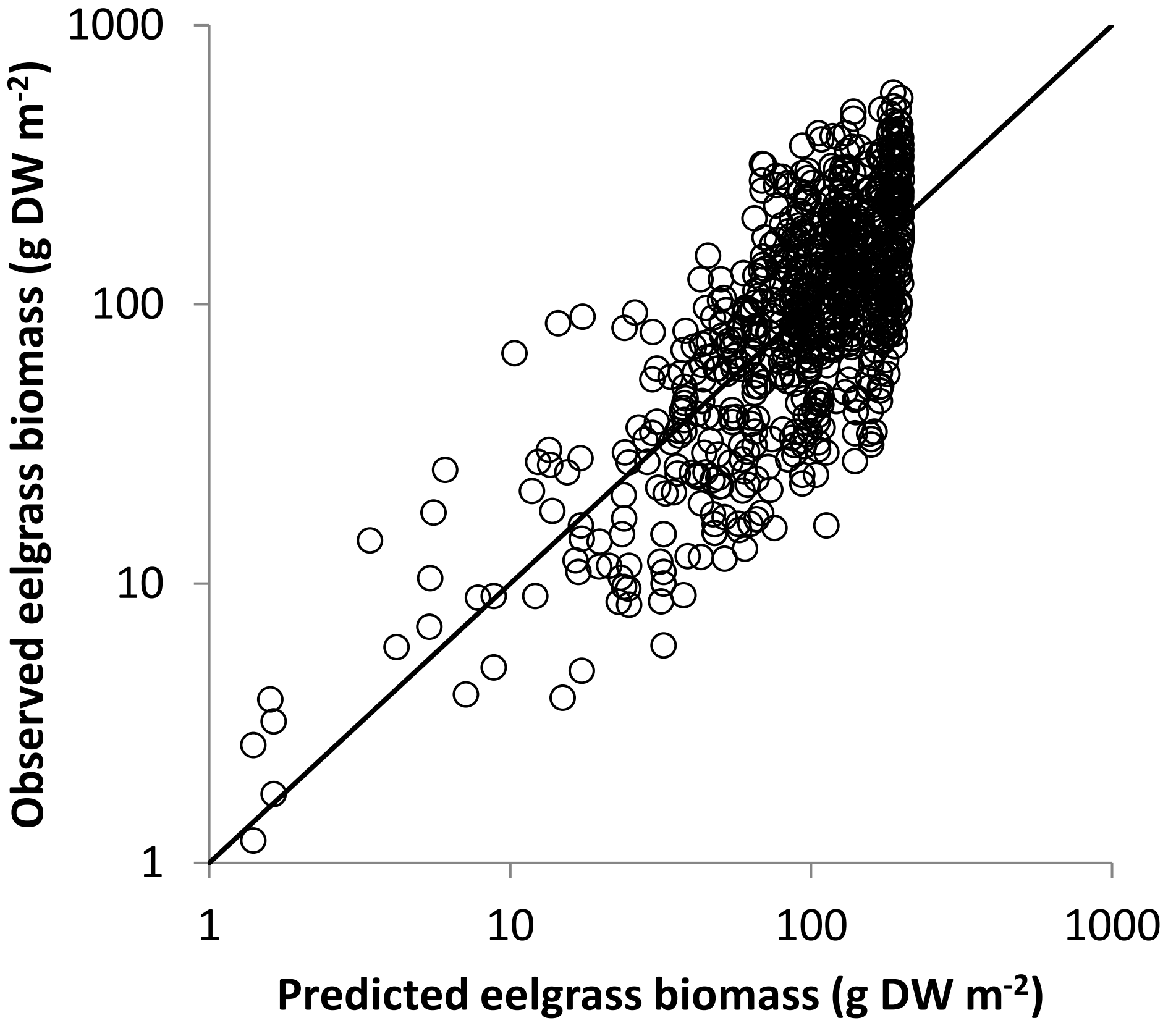


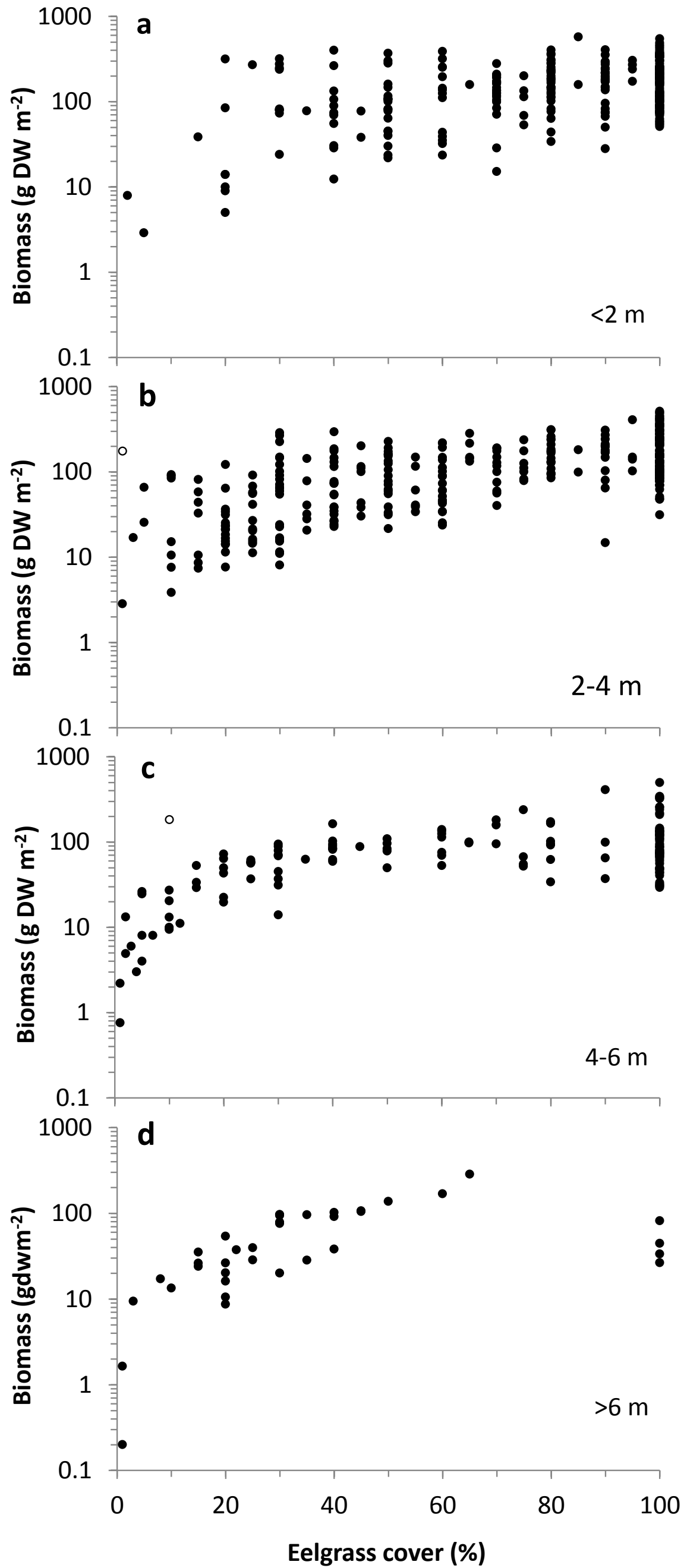


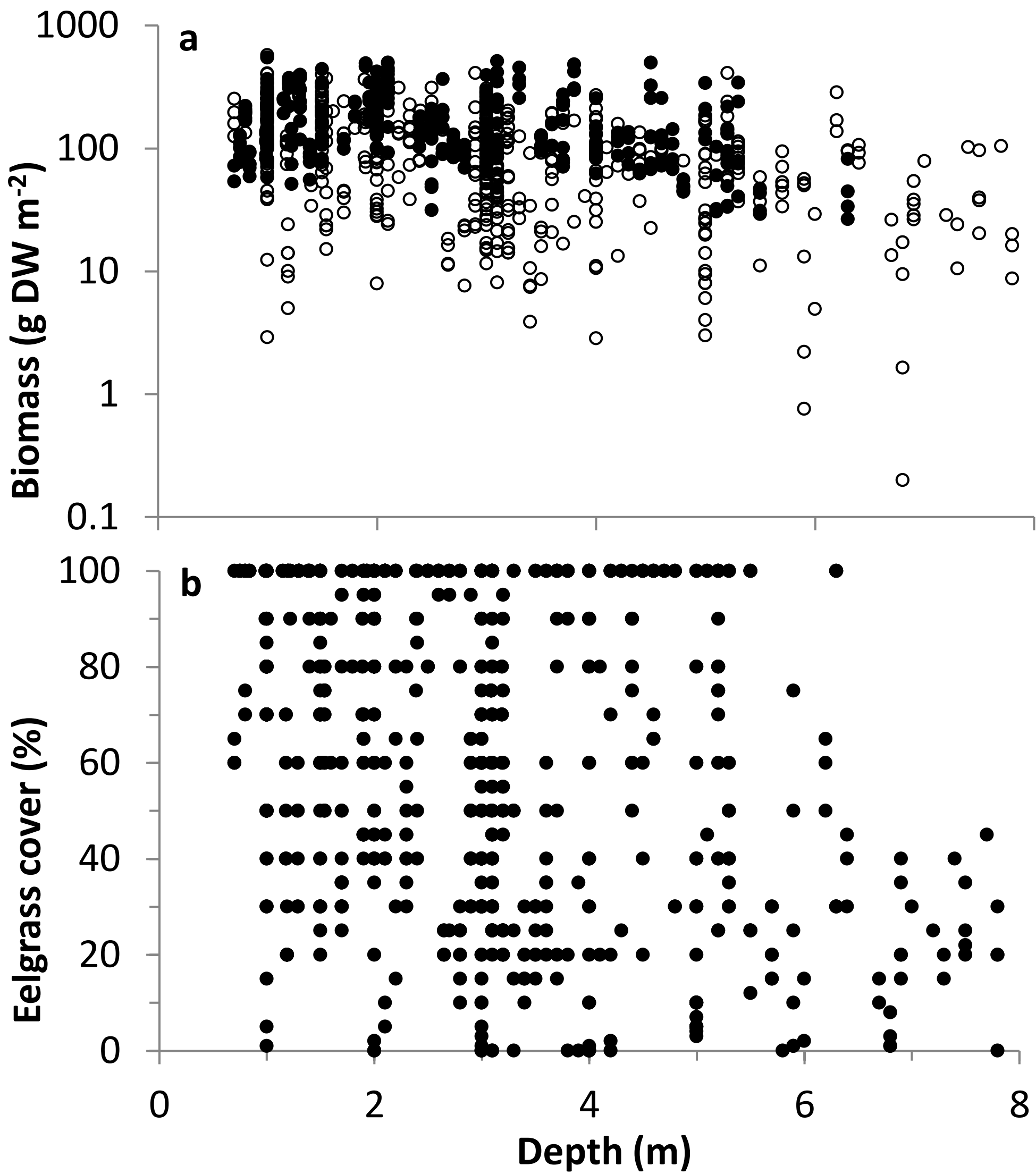












## FIGURE LEGENDS

**Fig. 1** Eelgrass aboveground biomass (**a**) and eelgrass cover (**b**) versus depth shown for all 791 samples. a) closed symbols are observations with 100% cover and open symbols have less than 100% eelgrass cover

**Fig. 2** Eelgrass aboveground biomass versus cover shown for different depth. **a:** 0.5-2 m, **b:** 2-4 m, **c:** 4-6 m, **d:** 6-8 m. Two outliers are shown with open symbols

**Fig. 3** Predicted versus observed eelgrass aboveground biomass.

**Fig. 4** Monthly estimates of maximum aboveground eelgrass biomass at 100% cover ( $B_{max}$ ). Estimates were obtained from back-transforming the estimates in Table 2. Error bars show the standard errors of the monthly estimates

**Fig. 5** Marginal relationships between aboveground eelgrass biomass and depth for four selected sites. Variations in eelgrass cover, interannual variations in Secchi depth and month of sampling were accounted for by adjusting observations (dots) and the modeled relationships (solid line) to a common eelgrass cover of 100% and an average over all months from March to October using the estimated relationship (Eq. 6) and annual means of Secchi depth. The four selected sites had the most biomass observations and a broad span in Secchi depths and eelgrass depth ranges (Table 1)

**Fig. 6** Marginal relationships between aboveground eelgrass biomass and cover for four selected sites. Variations in sampling depth, interannual variations in Secchi depth and month were accounted for by adjusting observations (dots) and the modeled relationships (solid line) to a mean sampling depth (Table 1) and an average over all months from March to October using the estimated relationship (Eq. 6) and annual means of Secchi depth. The four selected sites had the most biomass observations and a broad span in Secchi depths and eelgrass depth ranges (Table 1)

**Fig. 7** Application of the estimated eelgrass biomass vs. cover relationship to four transects from selected sites monitored in 2013. Depth and eelgrass cover are measured along transects of variable length and

converted to eelgrass biomass using Eq. (6). Secchi depths used in the equations were 7.2 m (Køge Bugt), 4.1 m (Odense Fjord), 3.7 m (Roskilde Fjord), and 8.9 m (The Sound)



## Supplementary information

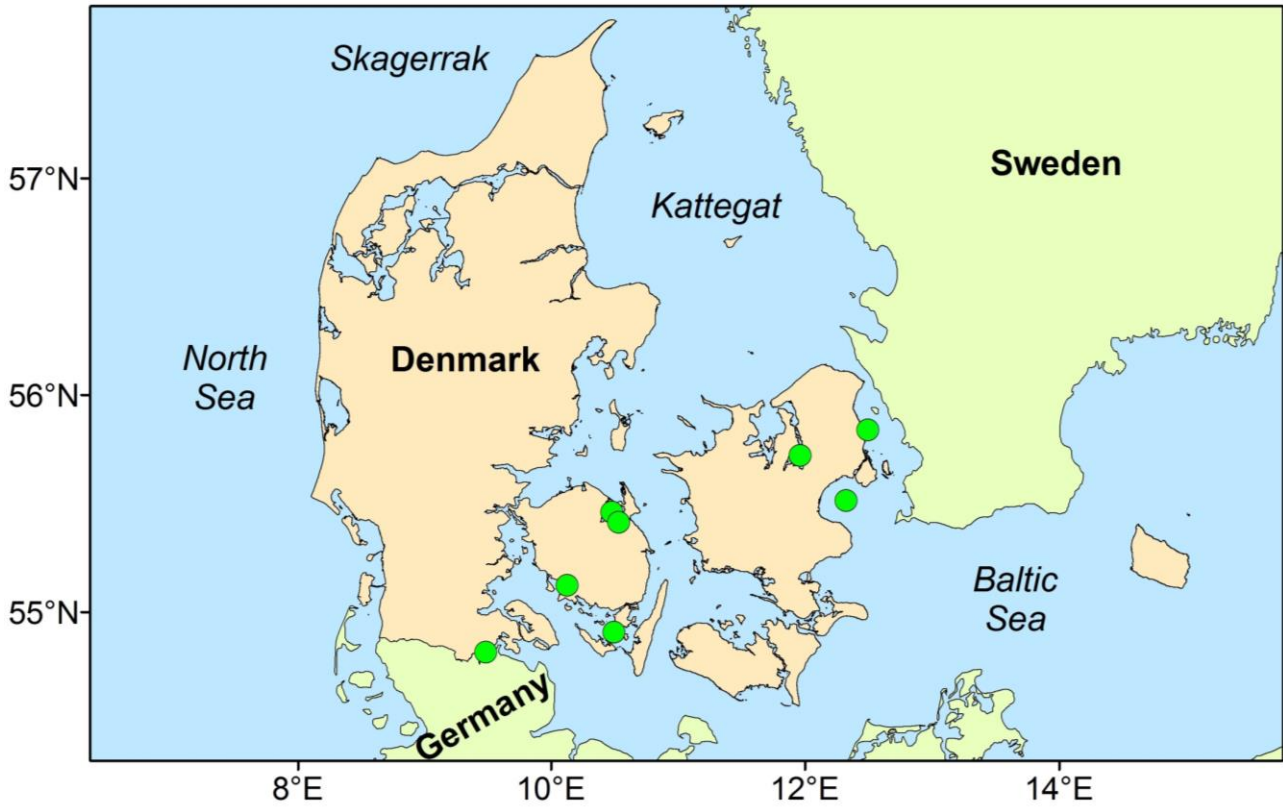


Fig. S1: Location of coastal sites where eelgrass biomass was sampled.