1	Seagrass collapse due to synergistic stressors is not
2	anticipated by phenological changes
3	Giulia Ceccherelli ^{1*} , Silvia Oliva ¹ , Stefania Pinna ¹ , Luigi Piazzi ¹ , Gabriele Procaccini ² ,
4 5	Lazaro Marin-Guirao ² , Emanuela Dattolo ² , Roberto Gallia ² , Gabriella La Manna ^{1,3} , Paola Gennaro ⁴ , Monya M. Costa ⁵ , Isabel Barrote ⁵ , João Silva ⁵ , Fabio Bulleri ⁶
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7	
8 9	¹ Dipartimento di Scienze della Natura e del Territorio, Polo Bionaturalistico, University of Sassari, via Piandanna 4, 07100 Sassari, Italy
10	² Stazione Zoologica Anton Dohrn, Villa Comunale, Napoli, Italy
11 12	³ MareTerra – Environmental Research and Conservation, Regione Sa Londra 9, Alghero (SS), Italy
13 14	⁴ Italian National Institute for Environmental Protection and Research (ISPRA), via di Castel Romano 100, Roma, Italy
15 16	⁵ CCMAR - Centre of Marine Sciences, University of Algarve, Faro, Portugal
17	⁶ Department of Biology, University of Pisa, via Derna 1, Pisa
18	
19	*corresponding author
20	GIULIA CECCHERELLI
21	Dipartimento di Scienze della Natura e del Territorio,
22	Polo Bionaturalistico,
23	University of Sassari,
24	via Piandanna 4, 07100 Sassari, Italy
25	phone: +39-079-228643
26	e-mail address: cecche@uniss.it
27	
28 29 30 31	Author contributions: GC, SO, and LP conceived and designed the experiments. GC, SO, SP and LP performed the experiments. GP, LMG, ED, RG, PG, MMC, IB, JS analysed the sampled material, FB analysed the data. GC and FB led the writing of the ms. All authors contributed critically to the drafts and gave final approval for publication.

running title: Synergistic stressors cause seagrass collapse

35 Abstract

34

36 Seagrasses are globally declining and often their loss is due to synergies among 37 stressors. We investigated the interactive effects of eutrophication and burial on the 38 Mediterranean seagrass, Posidonia oceanica. A field experiment was conducted to 39 estimate whether shoot survival depends on the interactive effects of three levels of 40 intensity of both stressors and to identify early changes in plants (i.e. morphological, physiological and biochemical, and expression of stress-related genes) that may 41 42 serve to detect signals of imminent shoot density collapse. Sediment burial and nutrient enrichment produced interactive effects on P. oceanica shoot survival, as 43 high nutrient levels had the potential to accelerate the regression of the seagrass 44 45 exposed to high burial (HB). After 11 weeks, HB in combination with either high or medium nutrient enrichment, caused a shoot loss of about 60%. Changes in 46 47 morphology were poor predictors of the seagrass decline. Likewise, few biochemical variables were associated with P. oceanica survival (the phenolics, 48 49 ORAC and leaf δ^{34} S). By contrast, the expression of target genes had the highest 50 correlation with plant survival: photosynthetic genes (ATPa, psbD and psbA) were 51 upregulated in response to high burial, while carbon metabolism genes (CA-chl, 52 PGK and GADPH) were down-regulated. Therefore, die-offs due to high 53 sedimentation rate in eutrophic areas can only be anticipated by altered expression of stress-related genes that may warn the imminent seagrass collapse. 54 55 Management of local stressors, such as nutrient pollution, may enhance seagrass 56 resilience in the face of the intensification of extreme climate events, such as floods.

57 Key words: burial, early warnings, eutrophication, multiple-stressors, Posidonia oceanica.

58

59 1. Introduction

60 Transitions between natural systems with radically different properties can occur 61 abruptly. Important examples can be found in ecology (e.g., lake eutrophication and coral 62 reef collapses), where regime shifts have consequences that are often irreversible (Bellwood et al. 2004, Carpenter 2011, Perry and Morgan 2017). Predicting and 63 64 anticipating catastrophic regime shifts represents a timely objective to improve our ability 65 to preserve biodiversity and ecosystem functioning in the face of escalating anthropogenic 66 pressures (Boettiger and Hastings 2013). Among all systems, the coastal marine are experiencing a wide range of human-induced alterations, the magnitude of which 67 68 increases with the local density of human populations. Generally, the various 69 environmental stressors do not act in isolation (Crain et al. 2008); rather, the effects of 70 individual stressors interact to generate cumulative impacts that can be greater or smaller 71 than the sum of their individual impacts (i.e. synergistic or antagonistic effects). 72 Nevertheless, predicting the effects of combinations of stressors is particularly challenging 73 because mechanisms underpinning impact are rarely elucidated and either threshold or 74 nonlinear responses to stressors remain unknown (Griffen et al. 2016). 75 The effects of multiple stressors on slow-growing, habitat-forming species, such as 76 certain seagrasses, are particularly threatening. Seagrass meadows are among the most 77 important and productive coastal systems (Costanza et al. 1997). They provide key 78 ecological services, including nursery grounds, habitat (for a review see Heck et al., 2003), 79 organic carbon production and export, nutrient cycling, sediment stabilization, trophic transfer to adjacent habitats (Hemminga and Duarte 2000, Larkum et al. 2006) and coastal 80 81 protection from erosion (Fonseca and Cahalan 1992, Fonseca and Koehl 2006). 82 Nonetheless, they are threatened by the rapid environmental changes caused by the 83 expansion of coastal human populations. Rapid, large-scale seagrass loss over relatively

84 short temporal scales has been reported throughout the world (Bulthuis 1983, Orth and 85 Moore 1983, Fourqurean and Robblee 1999, Marbà et al. 2005, Walker et al. 2006). 86 Stressors, such as sediments and nutrients inputs from terrestrial runoff, physical 87 disturbance (e.g. trawling, anchoring), invasive species, disease, aquaculture, overgrazing, 88 algal blooms and global warming, have been shown to cause seagrass declines at scales 89 ranging from square meters to hundreds of square kilometres (e.g. Munkes 2005, Orth et 90 al. 2006, Williams, 2007, Waycott et al. 2009, Bockelmann et al. 2011, Giakoumi et al. 91 2015). Overall, enhanced nutrients loading and sedimentation rates are likely the most 92 common and significant causes of seagrass decline (Unsworth et al. 2015). Indeed, the 93 current expansion of fish farming and other aquaculture practices (e.g., shellfish culture) 94 can have serious consequences on local populations of seagrasses through increased 95 deposition of organic matter and nutrients (Marbà et al. 2006). While eutrophication is 96 considered as the main cause of seagrass loss at a regional scale, burial of plants due to 97 anthropogenic-increased sedimentation or natural extreme events like storms or floods, 98 has been identified as an important cause for local die-offs (Short and Wyllie-Echeverria 99 1996, Erftemeijer and Lewis 2006, Orth et al. 2006, Cabaço et al. 2008, Cabaço and 100 Santos 2014).

101 Posidonia oceanica (L.) Delile is a slow-growing seagrass, endemic in the 102 Mediterranean and experiencing a widespread decline throughout the basin (Telesca et al. 103 2015). The regression of P. oceanica beds and the consequent expansion of alternative 104 habitats (e.g. algal turfs or dead seagrass rhizomes, generally referred to as "dead matte") 105 is particularly common in the proximity of urban areas (Montefalcone et al. 2009, 106 Tamburello et al. 2012). In addition to enhanced nutrient loading, P. oceanica meadows 107 are exposed to increased sedimentation rates as a consequence of beach nourishment, 108 dredging of waterways, shoreline armouring and severe climatic events (i.e. storms and 109 floods). Correlative and experimental studies have assessed the effects of both

110	eutrophication (Alcoverro et al. 1997, Delgado et al. 1999, Ruiz et al. 2001, Holmer et al.
111	2008) and burial on <i>P. oceanica</i> (Manzanera <i>et al.</i> 2011, Gera <i>et al.</i> 2014), but how
112	organic load levels change the effects of burial is yet to be explored. Within this context,
113	the identification of early warning signals for drastic declines would be a valuable tool for
114	the management of seagrass meadows (McMahon et al. 2013, Macreadie et al. 2014,
115	Roca et al. 2016). This goal has been pursued for pressing disturbances through
116	correlative approaches (e.g., van Katwijk et al. 2011) that do not, however, allow
117	estimating signs of imminent collapses.
118	Here, we investigate the interactive effects of eutrophication and burial on <i>P</i> .
119	oceanica. A field experiment was conducted 1) to estimate whether shoot survival
120	depends on the interactive effects of three levels of intensity of both stressors, using a full
121	factorial design, and 2) to identify early changes in plant attributes (<i>i.e.</i>
122	morphological/growth, physiological/biochemical, and expression of stress-related genes)
123	that may serve as signals of imminent shoot density collapse (early warnings of
124	degradation). In general, response time and sensitivity to stressors vary with the type of
125	variable examined; thus, understanding how sensitivity to stressors may change according
126	to the level of biological organization is essential to rationalise the choice of indicators of
127	impending seagrass decline and to design monitoring programmes (Roca et al. 2016).
128	Indicator specificity is expected to increase when moving towards lower levels of biological
129	organisation (sensu, Whitham et al. 2006), from the structural metrics to specific
130	physiological and molecular indicators (Adams and Greeley, 2000). Whether this general
131	rule holds for <i>P. oceanica</i> remains utterly unexplored.
132	2. Materials and Methods

133 2.1. Study site

134	The study was carried out in a shallow (5-8 m deep) continuous <i>P. oceanica</i>
135	meadow, on the north-west coast of Sardinia (40°34.1 N, 09°8.5 E). At this site, P.
136	$oceanica$ canopy structure in the inner meadow is well preserved (shoot density mean \pm SE
137	= 699.4 \pm 38.6 m ⁻² , <i>n</i> =28; canopy height mean \pm SE = 63.92 \pm 1.94 cm, <i>n</i> =35). At the site, the
138	grazing sea urchin, Paracentrotus lividus, is present at low densities, whilst juveniles of the
139	herbivore fish, Sarpa salpa, are common (GC, personal observations).

140 **2.2. Experimental design and set up**

141 The experiment started on the 29th of April 2015 and lasted until P. oceanica shoot 142 mortality exceeded 60% in some treatments (i.e. 11 weeks, see Results). The hypotheses 143 were tested by running a fully-factorial experiment of sediment burial (high, medium and 144 control) and nutrient enhancement (high, medium and ambient) treatments. Twenty-seven 145 circular patches (hereafter referred to as experimental units) were randomly selected 146 across the seagrass meadow, at least 3 m apart one from another. Within each unit, a 147 PVC cylinder (40 cm in height and diameter) was inserted about 10 cm deep into the 148 bottom, thus leaving about 30 cm of the cylinder above the sediment (Fig. S1). Each 149 cylinder enclosed between 81 and 109 P. oceanica shoots. 150 High, medium and control burial (HB, MB and CB) were obtained by adding 12.5 L, 151 5.0 L and 0 L of sediment (corresponding to a layer 10 cm, 4 cm and 0 cm tick), 152 respectively accordingly with Manzanera et al. (2011). We used washed sand for 153 playgrounds in our experiment, as this sand is similar in grain size (coarse sand, 0-1 mm) 154 and mineralogy (carbonate) to sediment at the study site. This ensured that all 155 experimental units were treated with the sand characterized by the same granulometry,

- 156 devoid of fauna and low in organic content. Depending on the treatment, P. oceanica
- 157 plants were buried with 4 cm of sediment over the ligula in HB, at the ligula height in MB,

and unburied in CB. When scuba divers filled experimental units with sand, care was taken
not to damage the leaves and to keep the plants upright during the process.

160 Also, high, medium and ambient nutrients (HN, MN and AN) were obtained by adding 161 80 g, 40 g and 0 g of homogenized fish fodder (protein 56.66%, fat 24.88%, cellulose 162 5.53%, ash 9.4%, phosphorous 1.35%, calcium 1.73% and sodium 0.46%) to the sediment 163 in each unit. Treatment levels were decided on the basis of nutrient release from offshore 164 fish farms (Garcia-Sanz et al. 2010). Fish fodder was used to simulate effects of 165 eutrophication related to land and coastal use change (from deforestation to aquaculture), 166 which often increases the organic matter and nutrients input into coastal sediments. The 167 use of fish fodder not only increases ammonium levels through mineralization, but also 168 fuels the production of sulfide (Burkholder et al. 2007). Sulfide is a strongly phytotoxic 169 compound, as it blocks the activity of cytochrome oxidase and other metal containing 170 enzymes, which may lead to massive seagrass die-offs (Govers et al. 2014).

171 Each unit was randomly assigned to one of the six combinations of treatments (n=3). 172 Three extra cylinders with large holes and not exposed to either sediment or fish fodder 173 addition were used as procedural controls. The outer edge of all units was not parted off 174 and a trans-cylinder clonal integration between shoots was not impeded with the aim not 175 to impose further stress on shoots. The effects of the experimental conditions on shoot 176 survival were evaluated every few weeks (see section 2.3.2) with the aim of catching the shoot mortality of about 20% and 60% in the harshest treatments that had corresponded 177 178 to week 3 and 11. Then, to identify indicators of mortality, shoot survival at week 11 was 179 related with morphological, physiological/biochemical variables and expression of stress-180 related genes, estimated at week 3.

181 **2.3. Data collection and analyses**

182 **2.3.1. Assessment of trophic enrichment**

183	In order to estimate concentrations of inorganic nutrients, samples were taken from
184	the water column at two dates chosen at random during the experiment (week 7 and 11).
185	At each date, two replicate water samples were taken in each unit using a 125-mL sterile
186	bottle about 5 cm above P. oceanica rhizomes. Samples were shaken and then filtered
187	(0.45 μm mesh size) as soon as they were brought to the surface. Samples were frozen in
188	liquid nitrogen for transportation to the laboratory, where concentrations of ammonia,
189	nitrate, nitrite and ortophosphate were determined using a continuous-flow AA3
190	AutoAnalyzer (Bran-Luebbe) and expressed in μ mol I ⁻¹ . Concentrations of dissolved
191	inorganic nitrogen (DIN, ammonia+nitrate+nitrite) and phosphorus (DIP as $P-PO_{4^{3-}}$) were
192	analysed by means of two one-way analyses of variance (ANOVA) testing the effect of
193	Nutrient enrichment (3 levels, HN, MN and AN) on the pooled data taken in each unit
194	(n=18). Cochran's C-test was used before each analysis to check for homogeneity of
195	variance, and data were transformed when necessary. SNK test was used for a posteriori
196	means comparisons (Underwood 1997).
197	2.3.2. <i>P. oceanica</i> survival

P. oceanica shoot density was counted at week 0, 3, 7, 9 and 11 to estimate shoot
survival (assessed as the percentage of shoots with leaves) through time. Shoot survival
(%) after 3 and 11 weeks was analysed by means of two 2-way analyses of variance
(ANOVA), including the factors Burial (3 levels, HB, MB and CB, fixed) and Nutrient
addition (3 levels, HN, MN and AN) both fixed and orthogonal (*n*=3). Cochran's C-test was
used before each analysis to check for homogeneity of variance and data were
transformed when necessary (Underwood 1997).

To evaluate the effects of procedural controls (PC), *P. oceanica* shoot survival at week 3 and 11 was also analysed by two one-way ANOVAs (PC vs. CBAN, *n*=3). Cochran's C-test and SNK test were run as explained above.

208 2.3.3. P. oceanica morphological/growth and physiological/biochemical variables 209 Six morphological variables (epiphyte load, % of leaves with necrosis, maximum leaf 210 length, mean leaf length, number of leaves, shoot biomass), leaf growth rate, and eleven 211 physiological/biochemical variables (leaf N, C and S content and corresponding isotopic 212 signature δ^{15} N, δ^{13} C, δ^{34} S, the antioxidant capacity through the oxygen radical absorbance 213 capacity, ORAC, Trolox equivalent antioxidant capacity, TEAC and phenolics) (Table 1) 214 were estimated after 3 weeks, when shoot mortality was still inconspicuous (see Results). Selection of these variables was based on the review made by Roca et al. (2016). 215

216 Among the morphological variables, epiphyte load was estimated as the weight of the 217 epiphytes obtained from scraping the shoots with a razor blade (referred to the total weight 218 of scraped leaves). The incidence of leaf necrosis was estimated as the percentage of 219 leaves showing marks of necrosis (black areas or spots). Leaf growth rate was assessed 220 using a modified leaf punching technique. At the beginning of the experiment, three shoots 221 per unit were marked by punching a hole just above the leaf base-leaf blade junction of the 222 outermost leaf with a hypodermic needle, and tagged with a plastic cable tie. After 20 223 days, the punched shoots were collected and the length of the newly produced tissue in 224 each shoot measured.

Among the physiological/biochemical variables, ORAC, TEAC and phenolics were extracted in frozen leaf tissue samples as described in Costa et al. (2005) and estimated following Huang et al (2002), Re et al. (1999), and Folin-Ciocalteu method (Booker and Miller 1998, Migliore *et al.* 2007), respectively. Leaf N, C, and S (%) were determined in samples of ca. 3.5 mg of dried, finely ground and homogenized material from each unit.
Leaf N, C and S isotopic signature, was analysed through elemental analyser combustion
for continuous flow isotope ratio mass spectroscopy.

Effects of Burial (3 levels, HB, MB and CB, fixed) and Nutrients (3 levels, HN, MN and AN) on morphological/growth and physiological/biochemical variables (*n*=3) at week 3 were analysed by means of two 2-way analyses of variance (ANOVA). Cochran's C-test was used before each analysis to check for homogeneity of variance (Underwood 1997).

236 2.3.4. P. oceanica expression of stress-related genes

237 The expression of ten stress-related genes (GADPH, SHSP, HSP90, PGK, CAB-151, 238 psbD, psbA, CA-chl, rbcl, and ATPa), (Table 1) were estimated after 3 weeks. At this aim 239 three shoots for each experimental unit were collected after 3 weeks of treatment. A 4 cm 240 leaf segment from the youngest fully mature leaves of each shoot (usually the second-rank 241 leaf) was collected and rapidly cleaned from epiphytes with a razor blade, towel-dried and 242 immediately stored in RNAlater® tissue collection solution (Ambion, Life Technologies). 243 Samples were then transported to the laboratory, preserved one night at 4 °C and stored 244 at -20 °C until RNA extraction.

245 After total RNA extraction (Mazzuca et al. 2013), RNA quantity and purity were 246 assessed by Nanodrop (ND-1000 UV-Vis spectrophotometer; NanoDrop Technologies) 247 and 1% agarose gel electrophoresis. Average Abs260/280 nm and Abs260/230 nm ratios 248 (2.0 and 1.8, respectively) have indicated the absence of protein and solvent 249 contaminations, while gel electrophoresis has showed intact RNA, with sharp ribosomal 250 bands. Total RNA (500 ng) was reverse-transcribed in complementary DNA (cDNA) with 251 the iScript™ cDNA Synthesis Kit (Bio-Rad) using the GeneAmp PCR System 9700 (Perkin 252 Elmer).

253 The target genes were selected on the basis of the two key events that mandatorily 254 occur in the light and dark reactions of photosynthesis and can be hampered by biotic and 255 abiotic stressful conditions, reducing plant productivity and survival (Ashraf and Harris, 256 2013). In light reactions, light energy is harvested by antenna pigments, channelled to 257 photosystem II (PSII) to produce photochemistry and converted into chemical energy (i.e. 258 ATP) and reducing power (*i.e.* NAPDH) by the flow of electrons along the electron 259 transport chain. Particularly, we have selected and analyzed genes coding for a putative 260 fundamental protein of the photosynthetic antenna complex (CAB-151): the two PSII core 261 proteins D1 and D2 (psbA and psbD) and a putative chloroplastic ATP synthase subunit 262 alpha (ATPa). In dark reactions, CO₂ is fixed into carbohydrates in the Calvin-Benson 263 cycle by using ATP and NADPH produced in the light reactions. In relation to this cycle, we 264 have selected several putative genes encoding key enzymes of the plant carbon 265 metabolism: the large subunit of the RuBisCO (rbcl), directly involved in CO₂ fixation, and 266 a chloroplast carbonic anhydrase (CA-chl) that catalyzes the conversion of HCO-3 into 267 CO2. We also selected two major enzymes involved in glycolysis, glyceraldehyde-3-268 phosphate dehydrogenase (GAPDH) and phosphoglycerate kinase (PGK), for obtaining 269 energy and carbon molecules from glucose. Finally, two general stress genes were also 270 included in the analysis as they are ubiquitous heat stress proteins in environmental 271 stress: the molecular chaperone (HSP90) and one putative small heat shock protein 272 (SHSP). All selected genes were already investigated in other studies (Table S1). Primers 273 sequences and GenBank Accession Numbers are reported in the reference in which the 274 genes have been selected for the first time. The expression stability of a set of three 275 putative reference genes already tested in P. oceanica (i.e. EF1A, 18S and L23; Serra et 276 al. 2012) was evaluated by running GeNorm (Vandesompele et al. 2002) and Normfinder 277 (Andersen et al. 2004). According to stability analysis, two genes, namely 18S and L23 278 (Fig. S2), were used to normalize gene expression data.

After normalizing by each primer efficiency (all >0.92% and R²>0.96), relative treatment gene expression values were calculated as the negative differences in cycles to cross the threshold value (- Δ CT) between the reference genes and the respective target genes (- Δ CT = CT_{reference} - CT_{target}). Subsequently, fold expression changes were calculated for graphical purposes according to the equation: fold expression change = ± 2([(Δ CTtreatment)-(- Δ CTcontrol)]).

Effects of Burial (3 levels, HB, MB and CB, fixed) and Nutrients (3 levels, HN, MN and AN) on the relative expression of target genes (-ΔCT values; n=3) were analysed by means of 2-way analyses of variance (ANOVA). Cochran's C-test was used before each analysis to check for homogeneity of variance (Underwood 1997).

289 2.3.5. Assessment of the predictive variables

290 In order to identify predictors of shoot survival of P. oceanica under different 291 combinations of burial and nutrients enhancement levels, separate multiple regressions 292 were run between each of the four groups of response variables included in the study 293 (morphological/growth, antioxidant, isotopes, and gene expression) and shoot survival. 294 More specifically, values of explanatory variables measured at week 3 were used as 295 predictors of shoot survival at week 11. Collinearity among covariates was assessed by 296 means of Variance Inflation Factor (VIF) procedures. Covariates with highest VIF values, 297 calculated using the R car package, were sequentially dropped from the model, until all 298 VIF values were smaller than 3, as recommended by Zuur et al. (2010). Linearity and 299 homogeneity of variances was visually checked by means of residual plots. Log 300 transformation of data was effective in enhancing homogeneity of variances. For each 301 group of variables, the best-fit model was selected by means of a stepwise procedure 302 (both directions) using the stepAIC function from the R MASS package (Venables and 303 Ripley 2002). This function selects the best model based on the Akaike Information

304	Criteria. The relationship between plant survival and predictive variables retained by the
305	different best-fit models were visualized by means of partial regression plots, using the
306	function avPlots from the car package. The relative importance (as percentage) and
307	bootstrap confidence intervals of the explanatory variables retained in the best fit models
308	were assessed by means of the Lindemann-Merenda-Gold (Img) method for calculating
309	sequentially weighted partial R^2 (Lindeman <i>et al.</i> 1980), using the R "relaimpo" package
310	(Gromping 2006). This method calculates an average coefficient of partial determination
311	for each model permutation using the individual contribution of each explanatory variable.

312 3. Results

313 **3.1. Inorganic nutrients and** *P. oceanica* survival

The addition of fish fodder to sediments resulted in a significant increase of DIN concentration in the water column above rhizomes ($F_{2,51}$ =10.93 p=0.0001; Fig. 1); as shown by the SNK tests, the increment in DIN was proportional to the amount of fodder added (HN>MN>AN). By contrast, there were no significant differences in DIP among treatments ($F_{2,51}$ =3.08 p=0.0547).

319 The comparison between CBAN and PC shows that there was no artefact of PVC 320 cylinders on shoot mortality at both week 3 (F1,4=0.25 p=0.6430) and week 11 (F1,4=0.04 321 p=0.8554) (Fig. 2). P. oceanica shoot survival was significantly affected by sand burial, as 322 about the 25% of shoots died by week 3 in the HB units (Fig. 2 and Table 2, HB<MB=CB). 323 Burial was the only stressor affecting mortality of the seagrass in the short-term (3 weeks). 324 After that, shoot survival decreased through time and, after 11 weeks, it was regulated by 325 the interactive effects of nutrient addition and burial (Table 2; Fig. 2). In units exposed to 326 high burial, shoot survival under high and intermediate nutrients addition was lower than at ambient nutrient levels (HBHN=HBMN<HBAN). In contrast, shoot survival was not affected 327

by nutrients levels in units exposed to either medium or ambient burial (SNK tests in Table
2). At high nutrients levels, shoot survival was lower in units exposed to high than medium
or ambient burial levels (HBHN<MBHN=CBHN). At medium nutrients levels, the survival
decreased with increasing severity of burial (HBMN<MBMN<CBMN), while at ambient
nutrient levels, there was no difference among burial treatments (Fig. 2 and Table 2,
CBAN=MBAN=HBAN).

334 Synergistic effects of both high and medium nutrients addition and high burial were 335 further assessed by comparing the average response of P. oceanica shoot survival (%), 336 both at each stressor level and at each combination of stressor levels, with those obtained 337 for the theoretical additive response of stressors levels in combination (Fig. 3). In fact, for 338 the average response given at high nutrient addition (HN), medium nutrient addition (MN), 339 and high burial (HB) when applied individually, the cumulative effects under additive 340 conditions (HN+HB and MN+HB) on shoot survival would have been about two fold higher 341 than that observed (Fig. 3). The difference in shoot survival between the average 342 experimental evidence and the theoretical additive estimates provides evidence for the synergism between nutrient (high and medium) and burial (high) stressors. 343

344 3.2. *P. oceanica* predictive variables

After three weeks since treatments, nutrient addition changed epiphyte load, necrosis and leaf growth rate, among morphological/growth variables, phenolics and the antioxidant response, ORAC and TEAC, among the physiological variables. However, there was no significant effect of nutrient enrichment on isotopes or gene expression, except for the stress gene HSP90, which experienced a slight inactivation (Fig. 4, 5, 6 and Table 3). In contrast, burial significantly increased epiphyte load and affected gene expression of GADPH, PGK, psbA, CA-chl and ATPa (Table 3). The only variables that responded to the interactive effects of nutrient addition and burial were the percentage of leaf necrosis and phenolics content, being one the mirror pattern of the other (Table 3): the first was significantly greater at high nutrient levels when plants were exposed to high burial (HN>MN=AN) and at high burial levels if plants were at ambient nutrients levels (HB>MB=CB), while the latter was smaller at high nutrients levels condition only when plants were exposed to HB (HN<MN=AN), and in high buried plants only at HN (HB<MB=CB).

359 The multiple regressions conducted with genetic, isotope, antioxidant and 360 morphological covariates explained the 60.6%, 33.2%, 36.4% and 27.8% (Adjusted R² 361 values) of the variance in shoot survival, respectively (Table 4). After simplification through 362 the step-wise procedure, the best fit model at the level of gene expression included two 363 explanatory variables, CA-chl and ATPa, although only the effects of the former was 364 significant, contributing nearly for 84% of the variation in shoot mortality explained by the 365 model (Table 5; Fig. 6). The positive estimate of CA-chl suggests that overexpression of 366 this gene at week 3 is positively correlated with shoot survival at week 11 (Fig. S3). CA-chl 367 is directly related with carbon metabolism (correlated to PGK and GADPH) and was, in 368 fact, down-regulated in treatments, such as HBHN and HBMN (SNK test, Table 6), where 369 P. oceanica had the lowest survival. ATPa is a gene involved in the photosynthesis 370 (correlated to psbA and psbD) and was up-regulated in high burial treatments (Fig. 6 and 371 Table 6).

Four explanatory variables were retained in the best fit isotope model and only one of these, namely leaf δ^{34} S, was significantly related with shoot survival at 11 weeks and accounted for about 50% of the total variation in shoot survival explained by the model (Table 4). However, although leaf δ^{34} S was positively (Fig. S4) related with shoot survival, the ANOVA did not identify any significant effect neither of nutrient addition nor of burial on leaf δ^{34} S content (Table 3). The best fit antioxidant model retained two variables, ORAC and phenolic content, both of which were significantly correlated with shoot survival. Phenolic contents accounted for a greater proportion (~63%) than ORAC (~37%) of the total variability in shoot mortality explained by the model (Table 4). Coefficient estimates were negative for ORAC and positive for phenolics (Fig. S5).

The morphological/growth best fit model retained 3 variables, epiphyte load, shoot biomass and growth (Table 4). The former variable accounted for most of the total variation in shoot survival explained by the model (~77%, Table 5). The negative estimate of the relationship coefficients (Table 4) indicates that increasing levels of epiphytic loading at week 3 were associated with lower shoot survival at week 11 (Fig. S6). Plants that had grown more in the first three weeks had greater chances of be alive at week 11.

388 DISCUSSION

389 4.1. Synergistic effects of nutrients and burial on *P. oceanica* mortality

390 *P. oceanica* shoot survival was affected by synergistic effects of burial and nutrient 391 addition over a short time. High sediment load (HB), corresponding to the complete burial 392 of meristems, increased seagrass mortality independently of the organic load in just three 393 weeks. Apparently, HB was the only level strong enough to cause rapid seagrass 394 mortality.

Over a longer term, burial and nutrient enrichment had interactive, negative effects on *P. oceanica* survival. In particular, high nutrients levels (HN) had the potential to accelerate the regression of the seagrass subjected to high burial. The use of a gradient of both stressors, manipulated in a crossed design, has also allowed detecting the rapid threshold responses to their combinations. Thus, by week 11, HB in combination with either HN or MN, caused a shoot loss of about 60%. In addition, the slow decrease in shoot survival under high burial without organic loading (HBAN), suggests that mortality at
HB was determined by the level of nutrient load, from the fastest at HN to the slowest at
AN.

404 Total shoot mortality did not occur within 11 weeks since the start of the experiment, 405 even under the most stressful conditions and it is unknown whether it would have occurred 406 on a longer term. However, several mechanisms may underpin the survival of some 407 shoots within experimental plots. First, the clonal integration with neighbouring shoots 408 immediately outside the unit edge: continuity between shoots was not prevented to avoid 409 further stress that would have biased the response of the plants. Thus, transfer of 410 metabolites (e.g. carbohydrates) from non-treated plants may have sustained the survival 411 of plants experimentally exposed to stressful conditions. Second, burial effect might have 412 not been homogeneous among shoots within each experimental unit, as rhizome height 413 varied among them and samples were only three-replicated. Third, individual response to 414 stress can differ. Differences in resilience can be related to individual characteristics (e.g. 415 age) of single ramets or to genetic peculiarities of single genets. More resilient genotypes 416 could respond better to the synergistic action of the applied stressors (Hughes and 417 Randall 2004).

418 4.2. Predictive variables of *P. oceanica* collapse

The protraction of the experiment for 11 weeks allowed detecting widespread mortality in experimental units exposed to HBHN and HBMN. Thus, besides the identification of lethal effects of burial and eutrophication on *P. oceanica*, our study enabled the identification of the plant attributes (morphological/growth, physiological/biochemical and transcriptomic) at week 3, prior to mortality, related to survival at week 11.

425 4.2.1. Morphological/growth variables

426 Except for the increase in epiphyte load, changes in morphological variables and 427 growth rate were little effective in predicting seagrass loss (Table 6). An increase in 428 epiphyte biomass, especially macroalgae, is a common response to eutrophic conditions 429 (Piazzi et al. 2016) and the ratio between epiphyte and leaf biomass in P. oceanica is one 430 of the descriptors used to assess the ecological quality of Mediterranean water bodies 431 under the European Water Framework Directive (Gobert et al. 2009, Oliva et al. 2012). 432 Our study shows that the epiphyte load can be used also as an indicator of burial stress, 433 as probably a consequence of lower phenolics content in the buried plants (Costa et al. 434 2015). Furthermore, signals of imminent mortality can also be gained by the percent of 435 leaves with necrosis; leaf necrosis has been previously suggested to increase with 436 eutrophication (Roca et al. 2016). Here, a higher proportion of leaves exposed to both high 437 burial and both high and medium nutrient addition had necrotic spots which were not 438 uniformly distributed along the blades, being mostly concentrated on the basal part of the 439 leaf. Finally, leaf growth was accelerated by nutrient addition, but was not related to burial, 440 suggesting that this variable is inadequate for predicting imminent seagrass degradation 441 due to the combination of stressors.

442 4.2.2. Physiological/biochemical variables

Based on the multiple regressions, the biochemical/physiological variables associated with the *P. oceanica* survival, were the phenolic content, ORAC and δ^{34} S. Among these, only phenolics are likely to be a useful predictor (positive coefficient), since differences in their concentration were also related to the interactive effect of nutrient addition and burial (Table 6). Phenolics can have multiple biological functions mainly related to the reproductive strategy, adaptation and survival to environmental disturbances, antimicrobial and antifouling properties. Their deposition as lignin in cell walls increases their mechanical strength and improves plant response against pathogens
and wounding. Variations in phenolics content have been observed in *P. oceanica* as a
response to changes in water quality and when competing with invasive species (Pergent *et al.* 2008, Migliore *et al.* 2007, Rotini *et al.* 2013).

454 Furthermore, ORAC, which reflects the ability of the plant metabolism to scavenge 455 oxygen reactive species (ROS, Mittler, 2002) through hydrogen atom donation (both 456 enzymatically and non-enzymatically), was negatively associated with P. oceanica 457 survival, and shoots with higher ORAC will have higher probability of mortality for the high 458 concentrations of ROS. However, it only responded to nutrient addition (Table 6). 459 Furthermore, the δ^{34} S content (positive coefficient) would suggest that the presence of 460 sulphide intrusion in leaf tissue, lower leaf δ^{34} S signature, could predict greater shoot 461 survival (Table S4). However, this isotopic signal did not respond uniformly to any treatment, probably due to the sediment type (i.e. coarse-carbonate sediments, Oliva 462 463 2012): indeed, the relationship between sediment sulfide and the 834S of P. oceanica tissues is known to be rather complex being controlled both by the sediment sulfide 464 465 concentrations and the oxygen status of the plants (Borum et al. 2005). Because of the 466 high variability, the use of δ^{34} S as an indicator for fish farm effects on *P. oceanica* has 467 already not been supported (Frederiksen et al. 2007). Therefore, among the biochemical/physiological variables only results in leaf phenolics content are promising 468 469 enough to promote it as an indicator of nutrient and burial stressors of P. oceanica (Table 470 6).

471 4.2.3. Expression of selected genes

Gene expression of target genes had the highest correlation with shoot survival.
Photosynthetic genes (ATPa, psbD and psbA) were up-regulated in response to high
burial suggesting that high sediment load tends to increase the number of PSII and ATP

475 synthase, probably to compensate for high energy consumption. By contrast, carbon 476 metabolism genes (CA-chl, PGK and GADPH) were down-regulated, being associated 477 with plant survival. In particular, the experimental treatments significantly affected the 478 ability of plants to fix C and recycle ATP through glucolysis or gluconeogenesis, as 479 suggested by the significant down-regulation of PGK. This response is likely to result in a 480 long-term starvation of the plants, likely contributing to the high mortality observed in these 481 treatments. In other words, burial produced a strong effect on the carbon metabolism of 482 plants, reducing carbon fixation and the production of energy: buried plants invested 483 energy to increase the amount of PSII and ATP synthase to sustain the thylacoid proton 484 gradient necessary to produce more ATP. Further, because HSP90 is an ATP-dependent 485 molecular chaperone it is possible that a strong ATP reduction experienced by plants from 486 the most intense treatments resulted in lower expression levels of this gene. It is also 487 known that inhibition of HSP90 induced cell death in plants (Nishizawa-Yokoi et al. 2010; 488 Moshe et al. 2016) and, likely, the observed down-regulation is promoting and anticipating 489 plant mortality two months in advance.

490 4.3. Conclusions

491 This study has estimated that P. oceanica shoot survival strongly depends on the 492 interactive effects of levels of intensity of nutrients and burial and it has identified some 493 early changes in plant attributes. Within the whole set of possible predictors of seagrass 494 collapse, only a few were associated with shoot survival (with different contribution to the 495 total variability) and these did not necessarily correspond to those affected by the 496 treatments (Table 6). Furthermore, they were attributed to a different reliability level based 497 on the association to shoot survival. Overall, the evaluation of the candidate indicators of 498 P. oceanica regression for the consequences of high organic load and sediment burial has 499 produced some very promising evidence for the gene CA-chl, and secondly of ATPa, while 500 phenolics and epiphyte load are likely to be less reliable predictors. Thus, the sensitivity of 501 P. oceanica attributes to these stressors has increased as the level of biological 502 organization decreases (according to expectations, Roca et al. 2016), underlining the 503 importance of implementing monitoring protocols with biochemical and molecular 504 indicators. However, high-replicated studies that include different field conditions (e.g. 505 sediment type, wave exposure) are needed so that consistency of variables' response 506 would be estimated and solid predictors incorporation promoted in any kind of 507 management program (i.e. assessment of ecosystem status, environmental quality, 508 impacts or the results of mitigation actions). 509 Nevertheless, no morphological change of shoot or leaf size should be expected after 510 a severe storm or flood that accumulates high loads of sediments, rich in organic matter, 511 over a P. oceanica bed; these attributes are likely to be reliable warnings only of pressing

disturbances over a longer temporal scale (Roca *et al.* 2016). Consequently, if other levels of biological organization besides the morphological are not considered, the seagrass collapse without early warning signals (Pace *et al.* 2015). Our results indicate that the interaction between two of the main regional anthropogenic stressors in temperate coastal ecosystems, eutrophication and high sediment loads (the deposition of organic detritus likely brings with it increased sediments, Terrados *et al.* 1999), can trigger the fast collapse of seagrass meadows without any major phenological change.

In the growing interest of identifying the type and role of interactions among local anthropogenic stressors in driving habitat shifts in marine ecosystems (Russell and Connell 2012), this controlled factorial experiment provides evidence that eutrophication and burial have non-additive consequences on seagrass beds. The input of excess nutrients (primarily nitrate and phosphate) to the marine environment is a global problem associated with a range of human activities, but coastal eutrophication through organic 525 matter dispersal represents a further source of disturbance. In fact, the addition of a 526 detrital layer to a seagrass bed will not just result in increased dissolved inorganic nutrients 527 to the sediments, it will have a whole stimulatory impact upon the microbial, fungal and 528 detrital feeding community (Danovaro et al. 1994). A likely additional effect of increased 529 organic detritus is that of increasing the levels of sulphide stress within the sediments with 530 follow-up negative effects upon the seagrass (Marbà et al. 2006). Indirectly, our results 531 provide strong evidence suggesting that development of coastal areas and their 532 associated human activities will have major impacts on the seagrass meadows and such 533 information should be used to identify appropriate local management actions to halt the 534 global loss of seagrasses in favour of alternative habitats composed by either macroalgae 535 or anoxic mud (Unsworth et al. 2015).

536 Finally, local anthropogenic stressors are thought to negatively interact with global 537 climatic stressors resulting in decline of many habitats, such as coral reefs, macroalgal 538 forests, mangroves and seagrasses. These findings suggest that shifts from seagrass 539 systems to dead rhizomes (*i.e.* dead matte) in areas characterized by poor water quality, 540 may become more common under future scenarios of climate change (Garcia *et al.* 2013), 541 as frequency and intensity of storms and floods are expected to increase.

542 Because our study has demonstrated that the effects of nutrients and sediment burial 543 on seagrasses are synergistic, strategies for reducing nutrient levels, a widely advocated 544 strategy for seagrass conservation, should be pursued bearing in mind the need to control 545 also sedimentation rates. Local management of nutrient loading can represent a valid tool 546 for mitigating the impacts of global stressors on marine macrophytes (Falkenberg et al. 547 2010). Our study shows that reducing nutrient loading may be not sufficient to enhance the 548 resistance of seagrasses to global stressors, such as seawater warming (NOAA 2016). 549 Given the cumulative nature of human impacts and the large small-scale variation in life-

550	traits occurring in coastal environments, one size fits all strategies are unlikely be
551	successful for sustaining the functioning of marine ecosystems in the face of climate
552	changes. In addition, our study makes a first step towards the identification of response
553	variables that may function as early warning signals of imminent seagrass collapse. It also
554	provides evidence for increased indicator specificity at lower levels of biological
555	organisation, promoting the need of implementing monitoring with molecular analyses. As
556	a final note of caution, the robustness of the response variables identified as most
557	promising must be assessed against variations in stressor intensity and background
558	abiotic and biotic conditions before their implementation in monitoring programs.
559	Acknowledgements
560	Analyses were performed at the Servizos de Apoio á Investigación-Universidade da

and leaf δ³⁴S). We are grateful to Capo Caccia-Isola Piana MPA administration for
facilitating the samplings within their domains. This research was supported by MIUR, the
Italian Ministry of Education and Research, (TETRIS grant 2011-2015) and by Fondazione
di Sardegna (EIMS grant 2016-2019). PG acknowledges the support of S. Porrello and E.
Persia at ISPRA laboratories for water nutrient analysis.

Coruña (Leaf N, Leaf C, and corresponding isotopic signatures), Iso-Analytical Ltd (Leaf S

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568 LITERATURE CITED

Adams SM, Greeley MS (2000) Ecotoxicological Indicators of Water Quality: Using Multi response Indicators to Assess the Health of Aquatic Ecosystems. Water, Air, and Soil
 Pollution 123:103-115.

572	Alcoverro T, Romero J, Duarte CM, López NI (1997) Spatial and temporal variations in
573	nutrient limitation of seagrass Posidonia oceanica growth in the NW Mediterranean.
574	Marine Ecology Progress Series 146:155-161.
575	Andersen CL, Jensen JL, Orntoft TF (2004) Normalization of realtime quantitative reverse
576	transcription-PCR data: a model-based variance estimation approach to identify genes
577	suited for normalization, applied to bladder and colon cancer data sets. Cancer
578	Research 64:5245–5250.
579	Ashraf M, Harris PJC (2013) Photosynthesis under stressful environments: An overview.
580	Photosynthetica 51 (2):163-190.
581	Bellwood DR, Hughes TP, Folke C, Nyström M (2004) Confronting the coral reef crisis.
582	Nature 429:827–833.
583	Boettiger C, Hastings A (2013) From patterns to predictions. Nature 493:157-158.
584	Bockelmann AC, Beining K, Reusch TBH (2011) Widespread occurrence of endophytic
585	Labyrinthula spp. in northern European eelgrass Zostera marina beds. Marine Ecology
586	Progress Series 445:109-116.
587	Booker FL, Miller JE (1998) Phenylpropanoid metabolism and phenolic composition of
588	soybean [Glycine max (L.) Merr.] leaves following exposure to ozone. Journal of
589	Experimental Botany 49:1191-1202.

Borum J, Pedersen O, Greve TM, Frankovich TA, Zieman JC, Fourqurean JW, et al.
(2005) The potential role of plant oxygen and sulphide dynamics in die-off events of
the tropical seagrass, *Thalassia testudinum*. Journal of Ecology 93:148–158.

- Bulthuis DA (1983) Effects of in situ light reduction on density and growth of the seagrass
 Heterozostera tasmanica (Martens ex Aschers.) den Hartog in Western Port, Victoria,
- 595 Australia. Journal of Experimental Marine Biology and Ecology 67:91–103.
- Burkholder JM, Tomasko DA, Touchette BW (2007) Seagrasses and eutrophication.
 Journal of Experimental Marine Biology and Ecology 350:46–72.
- Cabaço S, Santos R (2014) Human-induced changes of the seagrass *Cymodocea nodosa*in Ria Formosa lagoon (Southern Portugal) after a decade. Cahier de Biologie Marine
 55:101-108.
- Cabaço S, Santos R, Duarte CM (2008) The impact of sediment burial and erosion on
 seagrasses: a review. Estuarine, Coastal and Shelf Science 79 (3):354-366.
- 603 Carpenter J (2011) May the best analyst win. Science 331:698–699.
- Costa M, Barrote I, Silva J, Olivé I, Costa MM (2015) Epiphytes modulate *Posidonia oceanica* photosynthetic production, energetic balance, antioxidant mechanisms and
 oxidative damage. Frontiers in Marine Science doi.org/10.3389/fmars.2015.00111.
- 607 Costanza R, d'Arge R, de Groot R, Farber S, Grasso M, Hannon B, Limburg K, Naeem S,
- 608 O'Neill RV, Paruelo J, Raskin RG, Sutton P, Van den Belt M (1997) The value of the 609 world's ecosystem services and natural capital. Nature 387:253-260.
- Crain CM, Kroeker K, Halpern BS (2008) Interactive and cumulative effects of multiple
 human stressors in marine systems. Ecology letters 11:1304-1315.
- Danovaro R, Fabiano M, Boyer M (1994) Seasonal changes of benthic bacteria in a
 seagrass bed (*Posidonia oceanica*) of the Ligurian Sea in relation to origin,
 composition and fate of the sediment organic matter. Marine Biology 119:489-500.

Dattolo E, Ruocco M, Brunet C, Lorenti M, Lauritano C, Sanges R, De Luca P, Procaccini
G (2014) Response of the seagrass *Posidonia oceanica* to different light
environments: insight from a combined molecular and photo-physiological study.
Marine Environmental Research 101:225-236.

- Dattolo E, Marín-Guirao L, Ruiz J, Procaccini G (2017) Long-term acclimation to reciprocal
 light conditions reveals depth-related selection in the marine foundation species
 Posidonia oceanica. Ecology and Evolution DOI:10.1002/ece3.2731
- Delgado O, Ruiz J, Perez M, Romero R, Ballesteros E (1999) Effects of fish farming on
 seagrass (*Posidonia oceanica*) in a Mediterranean bay: seagrass decline after organic
 loading cessation. Oceanologica Acta 22:109–117.
- Erftemeijer PLA, Lewis RRR (2006) Environmental impacts of dredging on seagrasses: a
 review. Marine Pollution Bulletin 52:1553–1572.
- Falkenberg LJ, Connell BD, Russel SD (2010) Disrupting the effects of synergies between
 stressors: improved water quality dampens the effects of future CO₂ on a marine
 habitat. Journal of Applied Ecology 50:51-58.
- Fonseca MS, Cahalan JA (1992) A preliminary evaluation of wave attenuation by four
 species of seagrass. Estuarine, Coast Shelf Science 35:565-576.
- Fonseca MS, Koehl MAR (2006) Flow in Seagrass canopies: the influence of patch width.
 Estuarine, Coastal and Shelf Science 67:1-9.
- Fourqurean JW, Robblee MB (1999) Florida Bay: A recent history of ecological changes.
 Estuaries 22:345-357.

636	Frederiksen MS, Holmer M, Diaz-Almela E, Marba N, Duarte CM (2007) Sulfide invasion in
637	the seagrass Posidonia oceanica at Mediterranean fish farms: assessment using
638	stable sulphur isotopes. Marine Ecology Progress Series 345:93-104.

- Garcia-Sanz T, Ruiz-Fernandez JM, Ruiz M, Garcia R, Gonzalez MN, Perez M (2010) An
 evaluation of a macroalgal bioassay tool for assessing the spatial extent of nutrient
 release from offshore fish farms. Marine Environmental Research 70:189-200.
- Garcia R, Holmer M, Duarte CM, Marbà N (2013) Global warming enhances sulphide
 stress in a key seagrass species (NW Mediterranean). Global Change Biology
 19:3629–3639.
- Gera A, Pagès JF, Arthur R, Farina S, Roca G, Romero J, Alcoverro T (2014) The effect of
 a centenary storm on the long-lived seagrass *Posidonia oceanica*. Limnology and
 Oceanography 59:1910-1918.
- Giakoumi S, Halpern BS, Michel LN, Gobert S, Sini M, Boudouresque CF, Gambi MC,
 Katsanevakis S, Lejeune P, Montefalcone M, Pergent G, Pergent-Martini C, SanchezJerez P, Velimirov B, Vizzini S, Abadie A, Coll M, Guidetti P, Micheli F, Possingham
 HP (2015) Towards a framework for assessment and management of cumulative
 human impacts on marine food webs. Conservation Biology 29(4):1228-1234.
- Gobert S, Sartoretto S, Rico-Raimondino V, Andral B, Chery A, Lejeune P, Boissery P
 (2009) Assessment of the ecological status of Mediterranean French coastal waters as
 required by the Water Framework Directive using the *Posidonia oceanica* Rapid Easy
 Index, PREI. Marine Pollution Bulletin 58:1727–1733.
- Govers LL, de Brouwer JHF, Suykerbuyk W, Bouma TJ, Lamersc LPM, Smolders AJP,
 van Katwijk MM (2014) Toxic effects of increased sediment nutrient and organic matter
 loading on the seagrass *Zostera noltii*. Aquatic toxicology 155:253-260.

- Griffen BD, Belgrad BA, Cannizzo ZJ, Knotts ER, Hancock ER (2016) Rethinking our
 approach to multiple stressor studies in marine environments. Marine Ecology
 Progress Series 543:273–281.
- Gromping U (2006) Relative importance for linear regression in R: The package relaimpo,
 Journal of Statistical Software 17(1):1–27.
- Heck KL, Hays C, Orth RJ (2003) A critical evaluation of the nursery role hypothesis for
 seagrass meadows. Marine Ecology Progress Series 253:123–136.
- 667 Hemminga MA, Duarte CM (2000) Seagrass ecology. Cambridge University Press
- 668 Holmer M, Argyrou M, Dalsgaard T, Danovaro R, Diaz-Almela E, Duarte CM, Frederiksen
- 669 M, Grau A, Karakassis I, Marbà N, Mirto S, Pérez M, Pusceddu A, Tsapakis M (2008)
- Effects of fish farm waste on *Posidonia oceanica* meadows: synthesis and provision of
 monitoring and management tools. Marine Pollution Bulletin 56:1618-1629.
- Huang D, Ou B, Hampsch-Woodill M, Flanagan JA, Prior RL (2002) High-throughput
 assay of oxygen radical absorbance capacity (ORAC) using a multichannel liquid
 handling system coupled with a microplate fluorescence reader in 96-well format.
 Journal of Agricultural and Food Chemistry 50:4437-4444.
- Hughes AR, Stachowicz JJ (2004) Genetic diversity enhances the resistance of a
 seagrass ecosystem to disturbance. Proceedings of the National Academy of
 Sciences 101:8998–9002.
- 679 Larkum AWD, Orth RJ, Duarte CM, (2006) Seagrasses: Biology, Ecology and
 680 Conservation. Dordrecht (The Netherlands). Springer.

681	Lauritano C, Ruocco M, Dattolo E, Buia MC, Silva J, Santos R, Olivé I, Costa MM,
682	Procaccini G (2015) Response of key stress-related genes of the seagrass Posidonia
683	oceanica in the vicinity of submarine volcanic vents. Biogeosciences 12:4185-4194.

- Lindeman RH, Merenda PF, Gold RZ (1980) Introduction to bivariate and multivariate
 analysis, Scott, Foresman, Glenview, Ill.
- Macreadie PI, Baird ME, Trevathan-Tackett SM, Larkum AWD, Ralph PJ (2014)
 Quantifying and modelling the carbon sequestration capacity of seagrass meadows –
 a critical assessment. Marine Pollution Bulletin 83:430-439.
- Manzanera M, Alcoverro T, Tomas F, Romero J (2011) Response of *Posidonia oceanica* to burial dynamics. Marine Ecology Progress Series 423:47-56.
- Marbà N, Duarte CM, Díaz-Almela E, Terrados J, Álvarez E, Martínez R, Santiago R,
 Gacia E, Grau AM (2005) Direct evidence of imbalanced seagrass (*Posidonia oceanica*) shoot population dynamics along the Spanish Mediterranean. Estuaries
 28:53-62.
- Marbá N, Santiago R, Díaz-Almela E, Álverez E, Duarte CM (2006) Seagrass (*Posidonia oceanica*) vertical growth as an early indicator of fish farm-derived stress. Estuarine,
 Coastal and Shelf Science 67:475-483.
- Marín-Guirao L, Ruiz JM, Dattolo E, Garcia-Munoz R, Procaccini G (2016) Physiological
 and molecular evidence of differential short-term heat tolerance in Mediterranean
 seagrasses. Scientific Reports 6:28615.
- Mazzuca S, Bjork M, Beer S, Felisberto P, Gobert S, Procaccini G, Runcie J, Silva J,
 Borges AV, Brunet C, Buapet P, Champenois W, Costa MM, D'Esposito D, Gullstrom
 M, Lejeune P, Lepoint G, Olive I, Rasmusson M, Richir J, Ruocco M, Serra IA,

704	Spadafora A, Santos R (2013) Establishing research strategies, methodologies and
705	technologies to link genomics and proteomics to seagrass productivity, community
706	metabolism, and ecosystem carbon fluxes. Frontiers in Plant Science 4:38.

- McMahon K, Collier C, Lavery PS (2013) Identifying robust bioindicators of light stress in
 seagrasses: A meta-analysis. Ecological Indicators 30:7-15.
- 709 Migliore L, Rotini A, Randazzo D, Albanese NN, Giallongo A (2007) Phenols content and
- 2-D electrophoresis protein pattern: a promising tool to monitor *Posidonia meadows*health state. BMC Ecology 7:6.
- Mittler R (2002) Oxidative stress, antioxidants and stress tolerance. Trends in Plant
 Science 7:405-410.
- Montefalcone M (2009) Ecosystem health assessment using the Mediterranean sea- grass
 Posidonia oceanica: a review. Ecological Indicators 9:595-604.
- Moshe A, Gorovits R, Liu Y, Czosnek H (2016) Tomato plant cell death induced by
 inhibition of HSP90 is alleviated by *Tomato yellow leaf curl virus* infection. Molecular
 Plant Pathology 17:247-260.
- Munkes B (2005) Eutrophication, phase shift, the delay and the potential return in the
 Greifswalder Bodden, Baltic Sea. Aquatic Sciences 67:372–381.
- Nishizawa-Yokoi A, Tainaka H, Yoshida E, Tamoi M, Yabuta Y, Shigeoka S (2010) The
 26S Proteasome function and Hsp90 activity involved in the regulation of HsfA2
 expression in Response to oxidative stress. Plant and Cell Physiology 51:486-496.
- NOAA (National Oceanic and Atmospheric Administration) (2016) Extended reconstructed
 sea surface temperature (ERSST.v4). National Centers for Environmental Information.
 Accessed March 2016. www.ncdc.noaa.gov/data-access/marineocean-data/extended reconstructed-sea-surface-temperature-ersst.

- Oliva S (2012) Seagrass systems under human pressure: responses and assessment
 tools. PhD Thesis Universitat de Barcelona (Spain).
- 730 Oliva S, Mascaró O, Llagostera I, Pérez M, Romero J (2012) Selection of metrics based
- on the seagrass *Cymodocea nodosa* and development of a biotic index (CYMOX) for
 assessing ecological status of coastal and transitional waters. Estuarine, Coastal and
 Shelf Science 114:7-17.
- Orth RJ, Moore KA (1983) Chesapeake Bay: an unprecedented decline in submerged
 aquatic vegetation. Science 222:51-53.
- 736 Orth RJ, Carruthers TJB, Dennison WC, Duarte CM, Fourqurean JW, Heck KLjr, Hughes
- AR, Kendrick GA, Kenworthy WJ, Olyarnik S, Short FT, Waycott M, Williams SL
 (2006) A Global Crisis for Seagrass Ecosystems. Bioscience 56:987-996.
- 739 Pace ML, Carpenter SR, Cole JJ (2015) With and without warning: managing ecosystems
- in a changing world. Frontiers in Ecology and the Environment 13:460-467.
- Pergent G, Boudouresque CF, Dumay O, Pergent-Martini C, Wyllie-Echeverria S (2008)
 Competition between the invasive macrophyte *Caulerpa taxifolia* and the seagrass
 Posidonia oceanica: contrasting strategies. BMC Ecology 8:20.
- Perry CT, Morgan KM (2017) Bleaching drives collapse in reef carbonate budgets and reef
 growth potential on southern Maldives reefs. Scientific Reports 7:40581 DOI:
 10.1038/srep40581
- Piazzi L, Balata D, Ceccherelli G (2016) Epiphyte assemblages of the Mediterranean
 seagrass *Posidonia oceanica*: an overview. Marine Ecology 37:3-41.
- Prado P, Alcoverro T, Romero J (2008) Seasonal response of *Posidonia oceanica*epiphyte assemblages to nutrient increase. Marine Ecology Progress Series 359:8998.

752	Re R, Pellegrini N, Proteggente A, Pannala A, Yang M, Rice-Evans C (1999) Antioxidant
753	activity applying an improved ABTS radical cation decolorization assay. Free Radical
754	Biology and Medicine 26:1231-1237.

- Roca G, Alcoverro T, Krause-Jensen D, Balsby TJS, van Katwijk MM, Marbà N, Santos R,
 Arthur R, Mascaró O, Fernández-Torquemada Y, Pérez M, Duarte CM, Romero J
 (2016) Response of seagrass indicators to shifts in environmental stressors: a global
 review and management synthesis. Ecological Indicators 63:310-323.
- Rotini A, Belmonte A, Barrote I, Micheli C, Peirano A, Santos RO, Silva J, Migliore L
 (2013) Effectiveness and consistency of a suite of descriptors to assess the ecological
 status of seagrass meadows (*Posidonia oceanica* L. Delile). Estuarine, Coastal and
 Shelf Science 130:252-259.
- Ruíz JM, Pérez M, Romero J (2001) Effects of fish farm loadings on seagrass (*Posidonia oceanica*) distribution, growth and photosynthesis. Marine Pollution Bulletin 42(9):749 760.
- Russell BD, Connell SD (2012) Origins and consequences of global and local stressors:
 incorporating climatic and non-climatic phenomena that buffer or accelerate ecological
 change. Marine Biology 159(11):2633-2639.
- 769 Serra IA, Lauritano C, Dattolo E, Puoti A, Nicastro S, Innocenti AM, Procaccini G (2012)
- 770 Reference genes assessment for the seagrass *Posidonia oceanica* in different salinity,
 771 pH and light conditions. Marine Biology 159:1269-1282.
- Short FT, Wyllie-Echeverria S (1996) Natural and human-induced disturbance of
 seagrasses. Environmental Conservation 23:17-27.

774	Tamburello L, Benedetti-Cecchi L, Ghedini G, Bulleri F (2012) Variation in the structure of
775	subtidal landscapes in the NW Mediterranean Sea. Marine Ecology Progress Series
776	457:29-41.

- Telesca L, Belluscio A, Criscoli A, Ardizzone G, Apostolaki ET, Fraschetti S, Gristina M,
 Knittweis L, Martin CS, Pergent G, Alagna A, Badalamenti F, Garofalo G, Gerakaris V,
 Pace ML, Pergent-Martini C, Salomidi M (2015) Seagrass meadows (*Posidonia oceanica*) distribution and trajectories of change. Scientific Reports 5:12505.
- Terrados J, Duarte CM, Kamp-Nielsen L, Agawin NSR, Gacia E, Lacap D, Fortes MD,
 Borum J, Lubanski M, Greve T (1999) Are seagrass growth and survival affected by
 reducing conditions in the sediment? Aquatic Botany 65:175-197.
- Underwood AJ (1997) Experiments in ecology: their logical design and interpretation using
 analysis of variance. Cambridge University Press
- 786 Unsworth RKF, Collier CJ, Waycott M, McKenzie LJ, Cullen-Unsworth LC (2015) A
 787 framework for the resilience of seagrass ecosystems. Marine Pollution Bulletin 100:34788 46.
- Unsworth RKF, Cullen LC, Pretty JN, Smith DJ, Bell JJ (2010) Economic and subsistence
 values of the standing stocks of sea- grass fisheries: potential benefits of no-fishing
 marine protected area management. Ocean and Coastal Management 53:218–224.
- Unsworth RKF, Collier CJ, Waycott M, Mckenzie LJ, Cullen-Unsworth LC (2015) A
 framework for the resilience of seagrass ecosystems. Marine Pollution Bulletin 100:34 46.

795	Vandesompele J, De Preter K, Pattyn F, Poppe B, Van Roy N, De Paepe A, Speleman F
796	(2002) Accurate normalization of real-time quantitative RT-PCR data by geometric
797	averaging of multiple internal control genes. Genome Biology 3:0034.1.

- Van Katwijk MM, Van der Welle MEW, Lucassen ECHET, Vonk JA, Christianen MJA,
 Kiswara W, Inayat al Hakim I, Arifin A, Bouma TJ, Roelofs JGM, Lamerset LPM (2011)
 Early warning indicators for river nutrient and sediment loads in tropical seagrass
 beds: a benchmark from a near-pristine archipelago in Indonesia. Marine Pollution
 Bulletin 62:1512-1520.
- 803 Venables WN, Ripley BD (2002) Modern Applied Statistics with S. Fourth Edition.
 804 Springer, New York.
- Walker DI, Kendrick GA, McComb AJ (2006) Decline and recovery of seagrass
 ecosystems-the dynamics of change. Pages 551–565 in Larkum, A.W.D., Orth, R.J.
 and Duarte, C.M., eds. Seagrasses: Biology, Ecology and Conservation. Dordrecht
 (The Netherlands): Springer.
- Waycott M, Duarte CM, Carruthers TJB, Orth RJ, Dennison WC, Olyarnik S, Calladine A,
 Fourqurean JW, Heck KL, Hughes AR, Kendrick GA, Kenworthy WJ, Short FT,
 Williams SL (2009) Accelerating loss of seagrasses across the globe threatens coastal
 ecosystems. Proceedings of the National Academy of Sciences 106:12377–12381.
- Whitham TG, Bailey JK, Schweitzer JA, Shuster SM, Bangert RK, LeRoy CJ, Lonsdorf EV,
 Allan GJ, DiFazio SP, Potts BM, Fischer DG, Gehring CA, Lindroth RL, Marks JC, Hart
 SC, Wimp GM, Wooley SC (2006) A framework for community and ecosystem
 genetics: from genes to ecosystems. Nature Reviews Genetics 7:510–523.
- Williams SL (2007) Introduced species in seagrass ecosystems: status and concerns.
 Journal of Experimental Marine Biology and Ecology 350:89–110.

- 819 Zuur AF, Ieno EN, Elphick CS (2010) A protocol for data exploration to avoid common
- 820 statistical problems. Methods in Ecology and Evolution 1:3–14.

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Table 1. The *P. oceanica* response variables classified in the three levels of biological organization.

Morphological/growth		Physiological and biochemical	Gene expression
Epiphyte load (mg/sh) Necrosis (% leaf) Max leaf length (cm) Mean leaf length (cm) Num of leaves/shoot Shoot biomass (g DW) Leaf growth rate (cm/sh·day)	isotopes antioxidant	ORAC (µmol EQT/g DW) TEAC (µmol/g DW) Phenolics (mg/g DW) Leaf N (%) Leaf δ^{15} N (%) Leaf C (%) Leaf δ^{13} C (%) Leaf S (%) Leaf δ^{34} S (%)	GADPH SHSP HSP90 PGK CAB-151 psbD psbA CA-chl rbcl ATPa

		shoot survival			shoot survival		
		3	weeks			11 weeks	5
Source	df	MS	F	Р	MS	F	Р
Nutrients = (N)	2	168.8	0.90	0.4240	631.1	7.6	0.0041
Burial = (Bu)	2	1181.9	6.30	0.0084	4430.8	53.3	0.0000
N×Bu	4	278.1	1.48	0.2488	836.5	10.1	0.0002
Residual	18	187.5			83.1		
					C = 0.3214 (ns)		
					Tra	ansform: None	
SNK test			SE=4.5 S <mb=c< td=""><td></td><td colspan="3">N×Bu, SE=5.263 HB: HN=MN<an< td=""></an<></td></mb=c<>		N×Bu, SE=5.263 HB: HN=MN <an< td=""></an<>		
					MB:	HN=MN=	=AN
					CB:	HN=MN=	=AN
					HN:	HB <mb=< td=""><td>=CB</td></mb=<>	=CB
					MN	: HB <mb< td=""><td><cb< td=""></cb<></td></mb<>	<cb< td=""></cb<>
					AN:	HB=MB=	=CB

Table 2. ANOVAs on the effects of Nutrients (three levels: high, medium, and ambient; HN, MN,AN), and Burial (three levels: high, medium, and none; HB, MB, CB) on shoot survival after 3 and11 weeks after the start of the experiment. n=3. SNK tests results are reported below.

Table 3. ANOVAs on the effects of Nutrient addition (three levels: high, medium, and ambient;
HN, MN, AN) and Burial (three levels: high, medium, and none; HB, MB, CB) on
morphological/growth, physiological/biochemical and gene expression response variables.
Significant F values are reported in bold; * = p<0.05.

Variable	Nutrients F _{2,18}	Burial <i>F</i> _{2,18}	Nutrients x Burial <i>F</i> _{4,18}
Morphological/growth			
Epiphyte load	6.23 *	4.10 *	2.76
Necrosis	5.83 *	2.91	3.06 *
Max leaf length	1.48	2.60	1.33
Mean leaf length	1.58	2.41	1.45
Num of leaves/shoot	1.06	0.72	0.22
Shoot biomass	0.48	0.89	2.07
Leaf growth rate	3.90 *	0.39	2.14
Physiological and biochemical			
ORAC	5.01 *	3.11	2.84
TEAC	5.11 *	0.26	2.11
Phenolics	0.34	3.29	4.07 *
	0.50	0.05	4.70
Leaf N	0.53	0.05	1.70
Leaf δ ¹⁵ N	0.46	1.68	2.28
Leaf C	0.74	0.86	1.31
Leaf δ ¹³ C	2.64 3.44	0.32 0.16	0.63
Leaf S	3.44 2.91	2.66	0.99 0.20
Leaf 534S	2.91	2.00	0.20
Gene expression			
GADPH	1.04	4.21 *	1.17
SHSP	3.29	0.89	2.47
HSP90	4.06*	0.26	0.75
PGK	2.39	8.00 *	2.19
CAB-151	0.45	1.60	1.31
psbD	0.19	2.52	2.87
psbA	0.37	4.58 *	2.44
CA-chl	2.25	9.25 *	2.03
rbcl	0.10	0.67	1.45
ATPa	0.95	5.75 *	2.88

Table 4. Multiple regressions of *P. oceanica* shoot survival against different groups of response variables. Coefficient estimates (Estimate), standard errors (SE), t-values, and significance level (*P*-value) for variables retained in the best-fit model are reported.

Morphological/growth						
effect	Estimate	SE	t-value	P-value		
Epiphyte load	-0.00365	0.00104	-3.50	0.00193		
Shoot biomass	-0.17576	0.10531	-1.66	0.10869		
Leaf growth	0.06651	0.03766	1.766	0.09066		
Adjusted R ² =0.2779	F _{3,2}	F _{3,23} =4.33 P-value 0.0146				
Physiological and biochem (antioxidant)	ical					
effect	Estimate	SE	t-value	P-value		
ORAC	-0.0004	0.00013	-3.045	0.00557		
Phenolics	0.0007	0.00021	3.697	0.00113		
Adjusted R ² =0.3644	Fa	₂₄=8.45 P-val	ue 0.0016			
	• 2,2	24 01.10 1 10.1				
Physiological and biochem (isotopes)		<u>24 01.01 10.</u>	<u></u>			
Physiological and biochem		SE	t-value	<i>P</i> -value		
Physiological and biochem (isotopes)	ical			<i>P</i> -value		
Physiological and biochem (isotopes) effect	ical Estimate	SE	t-value			
Physiological and biochem (isotopes) effect Leaf N	ical Estimate -0.4591	SE 0.2735	t-value -1.678	0.1075		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N	ical Estimate -0.4591 -0.1782	SE 0.2735 0.1195	t-value -1.678 -1.491	0.1501		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N Leaf S	ical <u>Estimate</u> -0.4591 -0.1782 -0.5854 0.0627	SE 0.2735 0.1195 0.4482	t-value -1.678 -1.491 -1.306 2.356	0.1075 0.1501 0.2050		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N Leaf S Leaf δ^{34} S	ical <u>Estimate</u> -0.4591 -0.1782 -0.5854 0.0627 F _{2:}	SE 0.2735 0.1195 0.4482 0.0266 222=4.23 P-val	t-value -1.678 -1.491 -1.306 2.356	0.1075 0.1501 0.2050		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N Leaf S Leaf δ^{34} S Adjusted R ² =0.3321 Gene expression effect	ical <u>Estimate</u> -0.4591 -0.1782 -0.5854 0.0627	SE 0.2735 0.1195 0.4482 0.0266	t-value -1.678 -1.491 -1.306 2.356 ue 0.0108 t-value	0.1075 0.1501 0.2050 0.0278		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N Leaf S Leaf δ^{34} S Adjusted R ² =0.3321 Gene expression	ical <u>Estimate</u> -0.4591 -0.1782 -0.5854 0.0627 F _{2:}	SE 0.2735 0.1195 0.4482 0.0266 222=4.23 P-val	t-value -1.678 -1.491 -1.306 2.356 ue 0.0108	0.1075 0.1501 0.2050		
Physiological and biochem (isotopes) effect Leaf N Leaf δ^{15} N Leaf S Leaf δ^{34} S Adjusted R ² =0.3321 Gene expression effect	ical <u>Estimate</u> -0.4591 -0.1782 -0.5854 0.0627 F _{2:} Estimate	SE 0.2735 0.1195 0.4482 0.0266 22=4.23 P-val SE	t-value -1.678 -1.491 -1.306 2.356 ue 0.0108 t-value	0.1075 0.1501 0.2050 0.0278 <i>P</i> -value		

	% contribution				
Morphological/growth	Epiphyte Ioad	Leaf growth	Shoot biomass		
	76.7	12.6	10.7		
Physiological and biochemical (antioxidant)	Phenolics 63.0	ORAC 37.0			
Physiological and biochemical (isotopes)	δ ³⁴ S 49.5	Leaf S 19.2	δ ¹⁵ N 16.9	Leaf N 14.4	
Gene expression	CA-chl 83.8	ATPa 16.1			

Table 5. Rank of variables contributing most to the *P. oceanica* shoot survival (% contribution to R^2).

Table 6. List of variables evaluated as possible indicators of imminent collapse of *P. oceanica* due to Nutrient (= N) and Burial (= B) stressors. For each of the variables included in the study are reported the direction of stressor effects (\uparrow =increase and Ψ =decrease effect and 'no Ha' = no alternative hypothesis detected by the SNK tests), the sign and strength of their association with *P. oceanica* shoot survival on the basis of their relative contribution to the total variability explained by the best-fit model (arbitrary scale: 0-25%: low; 25-50%: medium; 50-75%: high; 75-100%: very high) and reliability based on the total variability explained by the regression model the variables belongs to (arbitrary scale: poor R²<40, good 60>R²>40, very good R²>60).

Variable	Stressor effect	Association to survival	indicator reliability
Epiphyte load Necrosis Leaf growth rate	↑N ↑B ↑N ↑B ↑N	very high - low	poor
ORAC TEAC Phenolics	no Ha ♠N ♥N ♥B	medium - high	роог
Leaf $\delta^{34}S$	no effect	medium	роог
GADPH	no Ha	-	
HSP90	no Ha	-	
PGK	₩в	- very high	very
psbA	no Ha	low	very gooc
CA-chl	↓ в		Ċ.
ATPa	↑ B		

Fig. 1 Water nutrients. Change (respect to controls) in mean water DIN (left axis) and $P-PO_4^{3-}$ (right axis) due to experimental nutrient addition. Dots are averages of two sampling times (n=6). Black, grey, and light grey colours indicate high, medium, and control burial (HB, MB, CB, respectively). Nutrient addition levels are separated by dashed lines

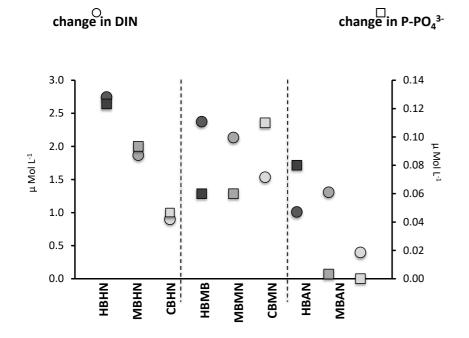
Fig. 2 *P. oceanica* shoot survival. Effect of experimental treatments (nutrient addition and burial) after 3, 7, 9 and 11 weeks (mean±SE, n=3). Black, grey, and light grey colours denote high, medium, and control burial treatments, respectively (HB, MB, CB, respectively). PC corresponds to Procedural controls

Fig. 3 *P. oceanica* survival. Average (n=3) shoot change in controls (CBAN), in each level of each single stressor (MB, HB, MN and HN), and in each combination of levels for multiple stressors interactions (MBMN, MBHN, HBMN, and HBHN). Grey and striped bars represent the experimental data and the theoretical additive response (calculated by summing the response to the single stressor levels), respectively

Fig. 4 Morphological/growth variables (mean+SE, n=9) change due to nutrient addition (H=high, M=medium and A=ambient) and burial (high, medium, and control)

Fig. 5 Physiological/biochemical variables (mean+SE, n=9) change due to nutrient addition (H=high, M=medium and A=ambient) and burial (high, medium, and control)

Fig. 6 Stress-related gene expression (mean+SE, n=9) change due to nutrient addition (H=high, M=medium and A=ambient) and burial (high, medium, and control)





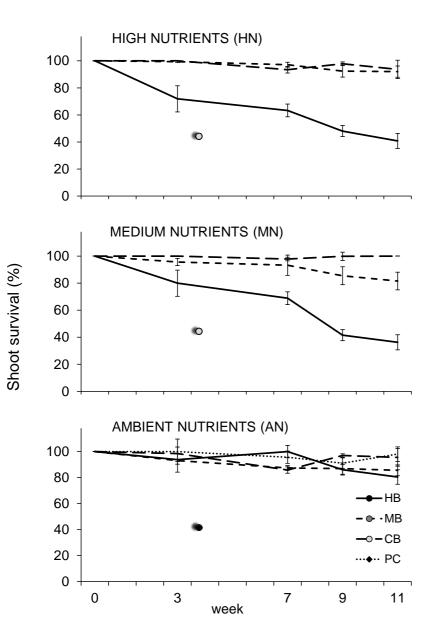


Fig 2

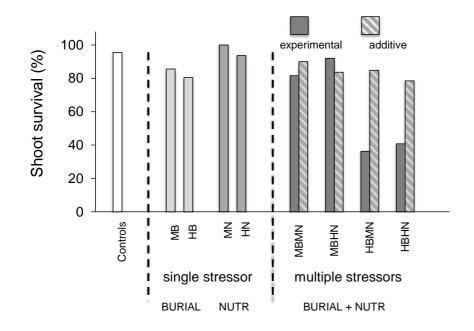
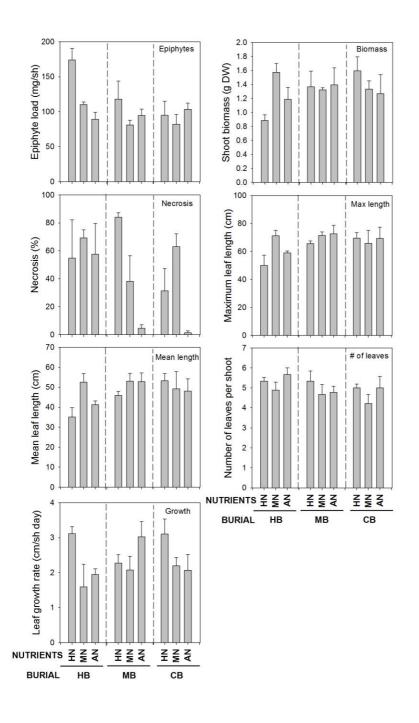


Fig 3





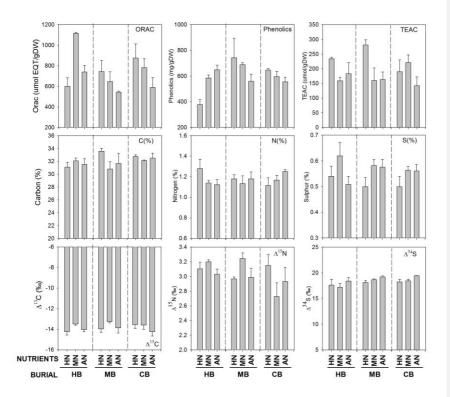
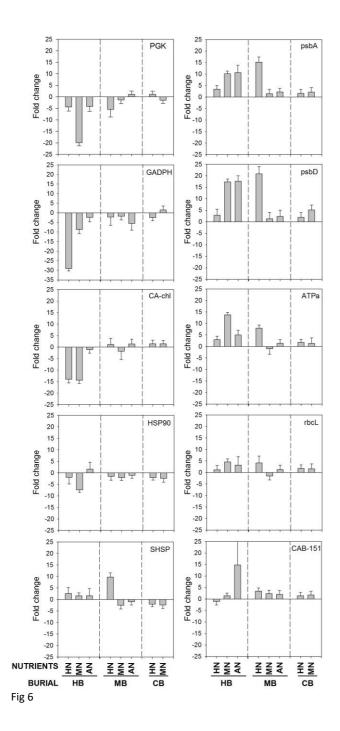


Fig 5



Seagrass collapse due to synergistic stressors is not anticipated by phenological changes

Giulia Ceccherelli, Silvia Oliva, Stefania Pinna, Luigi Piazzi, Gabriele Procaccini, Lazaro Marin-Guirao, Emanuela Dattolo, Roberto Gallia, Gabriella La Manna, Paola Gennaro, Monya M. Costa, Isabel Barrote, João Silva, Fabio Bulleri

Supplementary material

Table S1. List of genes analysed. Sequence of primers and stability curves are reported in the reference in which the genes have been selected for the first time. T: target genes; R: reference genes.

Gene name	Symbol	References	Туре
Photosystem II protein D1	PSbA	Dattolo <i>et al</i> . 2014	Т
Photosystem II protein D2	PSbD	Dattolo <i>et al</i> . 2014	Т
ATP synthase subunit alpha	ATPa	Marín-Guirao <i>et al.</i> 2016	Т
Chlorophyll a-b binding protein 151	CAB-151	Dattolo <i>et al</i> . 2014	Т
Phosphoglycerate kinase	PGK	Dattolo <i>et al</i> . 2017	Т
Carbonic anhydrase, chloroplastic	CA-chl	Dattolo <i>et al</i> . 2017	Т
Glyceraldehyde-3-phosphate dehydrogenase	GADPH	Serra <i>et al.</i> 2012	Т
Rubisco large subunit	rbcL	Marín-Guirao <i>et al</i> . 2016	Т
Heat Shock Protein 90	HSP90	Lauritano <i>et al</i> . 2015	Т
Small Heat Shock Protein	SHSP	Lauritano <i>et al.</i> 2015	Т
18S ribosomal RNA	18S	Serra <i>et al</i> . 2012	R
Elongation factor 1 alpha	EF1A	Serra <i>et al.</i> 2012	R
Ribosomal protein L23	L23	Serra <i>et al</i> . 2012	R



Fig. S1 PVC cylinder bordering an HBHN experimental unit. The coloured cable ties indicate shoots punched for leaf growth rate estimates.

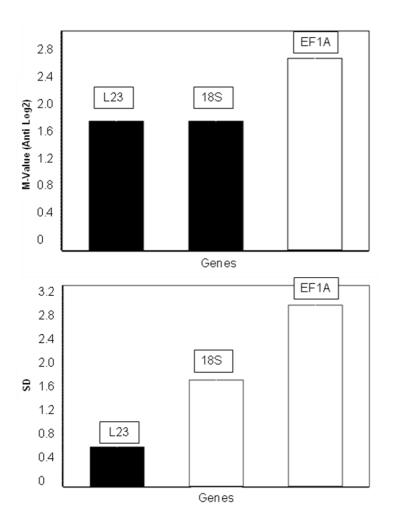


Fig. S2 Reference genes. Assessment of the stability of three putative reference genes. More stable genes are the ones in black, according to the software geNorm (top panel) and NormFinder (bottom panel). Both L23 and 18S have been utilized as reference genes in the analysis.

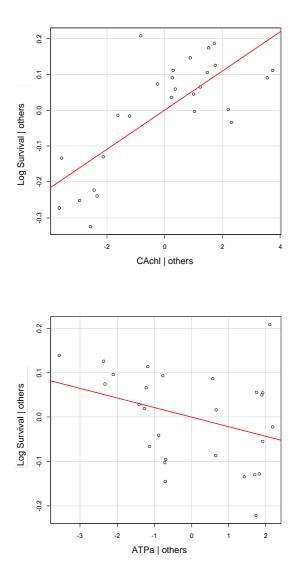


Fig. S3 Partial regression plots for each of the variables retained by the best-fit multiple regression model at the level of gene expression.

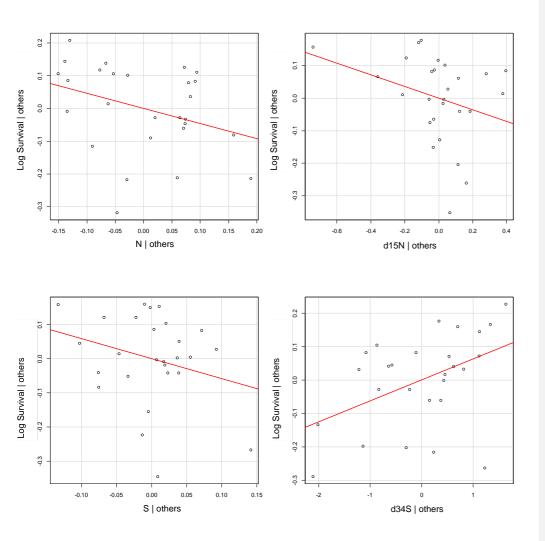


Fig. S4 Partial regression plots for each of the variables retained by the best-fit multiple regression model at the level of isotopes.

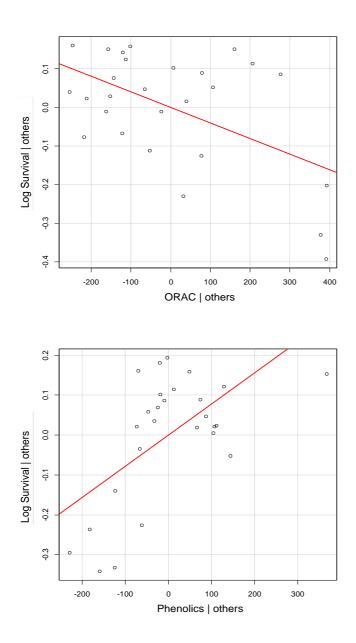


Fig. S5 Partial regression plots for each of the variables retained by the best-fit multiple regression model at the level of antioxidant.

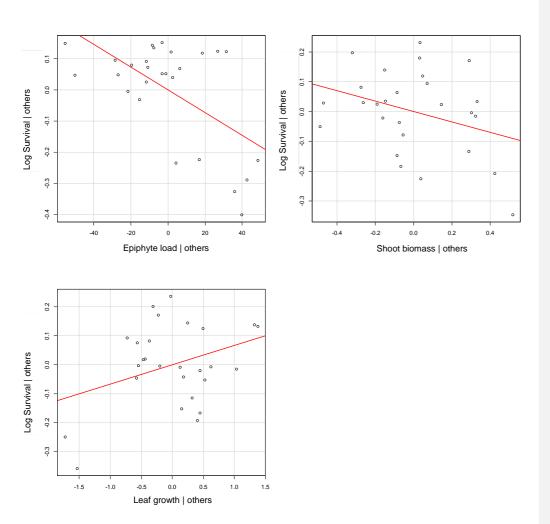


Fig. S6 Partial regression plots for each of the variables retained by the best-fit multiple regression model at the level of plant morphology/growth.