1	CHALLENGES TO THE ASSESSMENT OF BENTHIC POPULATIONS
2	AND BIODIVERSITY AS A RESULT OF RHYTHMIC BEHAVIOUR: VIDEO
3	SOLUTIONS FROM CABLED OBSERVATORIES
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58 All marine species studied thus far show rhythmic temporal patterns in Abstract 59 their behavioural, physiological, and molecular functions, which are collectively known 60 as biological rhythms. Biological rhythms are generated by biological clocks that time 61 biological functions and are synchronised by geophysical cycles such as the solar light-62 dark cycle and tidal cycle. On continental margins, behavioural rhythms can be detected 63 by diel (i.e., 24-h based) or seasonal periodical trawling as a consequence of massive 64 inward and outward displacements of populations to and from the sample areas. As a 65 result, significant errors in population/stock and biodiversity assessments performed by 66 trawling may occur if timing of sampling is not taken into account. The increasing 67 number of cabled and permanent multiparametric seafloor observatories now allows 68 direct, continuous, and long-lasting monitoring of benthic ecosystems and analysis in 69 relation to several habitat cycles. This review describes the adaptation of this 70 technology to investigations of rhythmic behaviour by focusing on automated video-71 imaging. Diel fluctuations in the number of video-observed individuals can be used as a 72 measure of average population rhythmic behaviour. The potential implementation of 73 automated video image analysis in relation to animal tracking and classification 74 procedures based on the combined use of morphometric tools and multivariate statistics 75 is detailed in relation to populational and community studies. Based on video cameras 76 mounted at multiparametric cabled observatories, an integrated time-series analysis 77 protocol using chrono-biomedical procedures is proposed to place video-recorded bio-78 information in an oceanographic context.

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Keywords: Behavioural rhythms, marine chronobiology, cabled multiparametric
underwater observatories and networks, automated video-imaging, tracking, species
classification, Elliptic Fourier Analysis (EFA), Fourier Descriptors (FD), Scale-

Invariant Feature Transform (SIFT), Red-Green-Blue (RGB) coordinates, bio- and
habitat data integrated time-series analysis, internal tides, diel captures, nektobenthic
movements

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The missing chronobiology framework in marine ecological studies

Chronobiology in deep water areas

89 The evolution of life occurs within a framework of deterministic (predictable) habitat 90 fluctuations (i.e., geophysical cycles; Aschoff 1981), such as light intensity or tidal 91 cycles. As a result, biological processes are governed by internal oscillatory 92 mechanisms, i.e., biological clocks, that have evolved as mechanisms to regulate the 93 period and phase of rhythmic behaviour and physiology (Daan & Aschoff 2001, Carr et 94 al. 2003). Because biological clocks are 'entrained' (or synchronised) to specific 95 environmental cycles, the biological rhythms they generate are in tune with the cyclic 96 temporal habitat of each species. Biological rhythms are collectively defined as 97 fluctuations in biological functions at any level of complexity, from molecular, 98 physiological, and behavioural processes in individuals, up to oscillations at the 99 population and community level (Aguzzi & Company 2010). Chronobiology is a rapidly developing field that studies the generation and entrainment of biological 100 101 rhythms using approaches that include reductionist strategies such as molecular biology 102 up to holistic level ecology (Naylor 2010).

103 The definition of rhythmic activity is usually the first step in biological rhythm 104 analysis, being the result of all underlying physiological fluctuations. Presently, 105 rhythmic behaviour is mostly studied in terrestrial animals by conducting laboratory 106 tests in which geophysical cycles are simulated or eliminated (i.e., constant conditions) 107 to reveal intrinsic (endogenous) mechanisms underlying the functioning of biological 108 clocks (Johnson et al. 2003). This laboratory-oriented research allowed the study of

biological timing systems in controlled conditions, conditions that cannot be easily 109 110 satisfied in field studies Laboratory research responds to the needs of biomedicine (e.g., 111 chronopharmacology), which studies behaviour as a phenotypic expression of molecular 112 clock mechanisms (Tosini & Aguzzi 2005, Hu et al. 2007, Bertolucci et al. 2008). For 113 example, the neurophysiologic basis of circadian (24-h endogenously based) clocks and 114 their light entrainment pathways are currently being studied in several mammalian and 115 insect species (e.g., Golombek & Rosenstein 2010) to establish the complex 116 arrangement of neural oscillator cells in the nervous system and their relationships with 117 peripheral tissues (Welsh et al. 2010).

118 In marine animals, rhythmic behaviour occurs in the form of predictable 119 variations in the rates of swimming, walking, and crawling, which are difficult to study 120 in the laboratory given the three-dimensional character of the marine environment (i.e., 121 the water column plus the seabed), where these activities take place (Aguzzi & 122 Company 2010). Laboratory and field research on this topic has chiefly been developed 123 using shallow or inland water fishes and decapod crustaceans of commercial interest 124 (e.g., related to fisheries and aquaculture) or other invertebrates (e.g., the gastropods 125 Aplysia or the American horseshoe crab; Lickey et al. 1976, Watson et al. 2008) based 126 on the combination of pragmatism in sampling (accessibility of animals for collection) 127 and high rates of survival in controlled laboratory conditions (Reebs 2002, Naylor 128 2005). If from one side, terrestrial chronobiology is well studied and control processes 129 deeply characterised, marine chronobiology has only been developed to a similar degree 130 for the intertidal zone (Naylor 2010). Data on rhythmic behaviour for species inhabiting 131 the continental margin areas of shelves and slopes, including the deep sea are currently 132 scant (Aguzzi et al. 2004b). The paucity of data on rhythmic behaviour becomes 133 dramatically evident as the depth range inhabited by species increases. Laboratory 134 studies using deep-water continental margin species (i.e., those of the middle and lower shelves or slopes) are less common, with major animals studied being decapod
crustacea such as the Norway and the American clawed lobsters (*Nephrops norvegicus*and *Homarus americanus*) (reviewed by Aguzzi & Sardà 2008; Aguzzi et al. 2008).

138 In a context where the behavioural habit is unknown for the vast majority of 139 marine species, rhythmic displacement of individuals can fundamentally influence the 140 perception of sampled population sizes and therefore communities' compositions 141 according to the time of the day (Navlor 2005). Accordingly, biodiversity observations 142 should be somehow linked down to individual behavioural performances in relation to 143 changing habit conditions. This fact has been poorly considered to date, as day-night 144 scheduled field sampling procedures are not often employed due to a lack of suitable 145 technology or sufficient resources for repetition. The aim of this review is to describe 146 the application of the newly developed video methods associated with permanent cabled 147 observatory stations to the study of the behavioural rhythms of benthic species. This 148 technology has the potential to provide important data for evaluation of the reliability of 149 population, stock and biodiversity assessments in depth contexts where sampling 150 repeatability and duration are strongly constrained.

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Ecological implications of behavioural rhythms

A number of biological traits, including rhythmic behaviour, are still unknown for the majority of benthic species, including many of commercial interest (Company & Sardà 1998, 2000, Company et al. 2003). Rhythmic movements related to the sea bed environment occur in a complex habitat scenario characterised by marked depthdependent variations in key oceanographic parameters.

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Behavioural rhythms as unknown displacement typologies in the complex environmentalcontext of continental margins

Light plays a central role in the regulation of the rhythmic behaviour of animals 161 162 inhabiting bottom areas within the twilight zone range (Chiesa et al. 2010). Because 163 light intensity decreases and spectral quality narrows with increasing depth (Herring 164 2002), day-night driven rhythms should be also weaker in slope species in comparison 165 with shallower-dwelling shelf species (Aguzzi et al. 2009a). In the oligotrophic western 166 Mediterranean Sea, the twilight zone ends on the middle slope (Margalef 1986), and 167 light can potentially regulate the circadian system of benthic species down to depths of 168 1000 m (Tobar & Sardà 1992, Sardà et al. 2003). As light progressively fades with 169 depth, the day-night control of behaviour decreases, and therefore, other geophysical 170 variables and cycles are expected to control the behavioural rhythms of benthos.

171 It is known that internal tides influence the rhythmic behaviour of fishes (e.g., zoarcids) inhabiting the aphotic deep sea (Aguzzi et al. 2010b). Wagner et al. (2007) 172 173 show that deep-sea fishes living at continental rise and abyssal depths, beyond the reach 174 of solar light, have an endogenous clock mediated through release of melatonin (Priede 175 et al. 1999). Water flow variations based on a 12.4-h tidal cycle affect the rhythmic 176 production of this hormone, the function of which is related to variations in metabolic 177 activity in association with animal behavioural activation (Wagner et al. 2007). In areas 178 where the tidal pull is negligible (e.g., the Mediterranean Sea), other cycles, such as 179 weaker atmosphere-driven inertial currents, appear to be an effective means of 180 geophysical control over the rhythmic behaviour and physiology of benthos (Aguzzi et 181 al. 2009e). Anyway, the rhythmic behaviour of deep water species in relation to depth-182 variable light intensity regimes and geographically variable tidally or inertially driven 183 current flows remains very poorly understood.

A traditional sampling methodology for the study of behavioural rhythms in marine species is the temporally scheduled use of pelagic and bottom trawling. Rhythmic catch patterns associated with these temporally modulated sampling methods are considered as a proxy for rhythmic displacements of populations into and out ofseabed sampling areas (Aguzzi et al. 2009d).

189 The behavioural rhythms of many pelagic species are characterised by diel (i.e., 190 24-h based) vertical migrations (DVMs) that occur within the water column in response 191 to defined light intensity variations (Naylor 2006, Gaten et al. 2008) typified by ascent 192 towards the surface at night time and descent at dawn. Repetition of pelagic sampling at 193 different depth strata at similar times has allowed the characterisation of migration 194 extension ranges for different demographic components of micro- and macronekton 195 (i.e., gender-related and ontogenetic modulations of behaviour) (Foxton 1970a,b, Franqueville 1971, Foxton & Roe 1974, Herring & Roe 1988, Frank & Widder 2002). 196 197 Within the DVM displacement category, a diel cycle of occurrence of benthopelagic 198 species is reported in seabed collections taken at different times (Aguzzi et al. 2006a). 199 Animal catches increase during daytime when benthopelagic populations descend in the 200 water column, hiding on the sediment surface from pelagic visual predators (Aguzzi et 201 al. 2006a).

202 Similarly detailed knowledge regarding the behaviour of benthos is still a long 203 way off. Several species apparently perform long-range migrations within the benthic 204 boundary layer, following bathymetric gradients across continental shelves and slopes 205 (Benoit-Bird & Au 2006). These movements can be broadly defined as nektobenthic 206 (Cartes et al. 1993). Catches of populations of these species vary rhythmically at a 207 certain depth stratum as the result of arrival and departure of schools (Aguzzi & 208 Company 2010). The occurrence of this behaviour has been inferred in most of cases by 209 combining direct observations, such as trawl catch patterns, with indirect ones, such as 210 comparisons of morphological and metabolic adaptations (Company & Sardà 1998, 211 Pulcini et al. 2008, Aguzzi et al. 2009b, Antonucci et al. 2009).

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213 The effects of behavioural rhythms on population/stock and biodiversity assessments

214 Deep oceans represent one of the last frontiers for human exploration on Earth, with 215 deep-sea marine biology being a relatively recent scientific discipline (Glover et al. 216 2010a). Ocean zones with water depths below 200 m cover 75% of the planet's surface, 217 but very little is known about the species inhabiting these zones and their life histories 218 (Waterman 2001). Deep-water ocean areas potentially host some of the ecosystems with 219 the highest biodiversity levels on the planet (Ramirez-Llodra et al. 2010). Most of these 220 deep-sea areas along continental margins are threatened by increasing anthropogenic 221 pressures (Sheppard 2000) mainly related to searching for minerals, oil, and gas, with 222 commercial fisheries apparently exerting the greatest impact (Benn et al. 2010). For 223 example, a deep water fishery in the NE Atlantic Ocean trawling down to a maximum 224 depth of 1500 m has been shown to deplete the demersal fish assemblage, affecting up 225 to 77 species of fish in addition to the 3 main species targeted by the fishing vessels 226 (Bailey et al. 2009, Priede et al. 2011). Furthermore, in these deep-sea areas 227 anthropogenic changes in species composition should be distinguished from more long-228 term natural ones that can be associated with climate (Bailey et al. 2006, Billet et al. 229 2009).

230 Studies on biodiversity in deep-water areas should be linked to the concept of 231 community functioning in the face of habitat changes through time (Smith et al. 2009). 232 The management of new fisheries grounds, or any other area where human activity may 233 represent a potential source of impacts, requires precise data on population demography 234 and distribution as well as the evaluation of biodiversity (Pauly et al. 2003, Morato et al. 235 2006). Commercially targeted species represent the focus of management policies, and 236 the preservation of their biomass for sustainable exploitation also requires investigation 237 of all other surrounding species, from which data must be derived and integrated into 238 increasingly holistic models (Roberts 2002). This requirement is one of the main justifications for biodiversity studies in deep-water areas and attempts to link
community functioning with habitat changes through space and time (Danovaro et al.
2010).

242 One of the great limitations to improving the reliability of estimating species 243 abundance estimates using trawl surveys is the overall variability of data collected at sea 244 (Godø 1998). In trawl surveys, the time of the day is usually preserved as constant, or it may represent a random character due to technical factors (Raffaelli et al. 2003). 245 246 Random sampling with respect to time produces unpredictable variability in the species 247 composition of catches (Bahamon et al. 2009). Planning temporally scheduled sampling 248 to standardise field data has been generally accepted in terrestrial ecology studies for 249 quite some time (Park 1941) but this approach is still inconsistently applied in marine 250 research (Naylor 2005).

251 Trawl timing is intuitively perceived as a biasing factor when sampling is not 252 repetitively performed at a given depth. Changes in species abundances and 253 composition are related to the rhythmic behaviour of individuals (Aguzzi & Company 254 2010). Therefore, sampling at a given depth should be repeated at different times of day 255 and in different seasons to evaluate the biasing effects of time on the reported trawl data 256 (Aguzzi & Bahamon 2009). Unfortunately, these factors are not often taken into 257 account. There is a conflict between allocation of effort to sampling and space or time 258 not only in deep-water sampling but also for coastal-water sampling (Azzurro et al. 259 2011). An increase in the surveyed area, as required for the assessment of population distributions, usually implies a concomitant reduction in the frequency of repetitive 260 261 sampling within a certain location (Benoit & Swain 2003). Only temporally scheduled 262 sampling within a fixed area can provide the data required for the characterisation of 263 behaviour rhythms.

264 Within this framework, specifically planned day-night trawl surveys show the 265 occurrence of broad diel variations in communities from different continental margin 266 areas exposed to fluctuations in solar light intensity (Petrakis et al. 2001, Benoit & 267 Swain 2003, Carpentieri et al. 2005, Benoit-Bird & Au 2006, Benoit-Bird et al. 2008, 268 Aguzzi & Bahamon 2009). For example, in the western Mediterranean Sea, diel 269 differences in catch composition are the product of the rhythmic appearance and 270 disappearance of populations from trawl-hauling windows (Figure 1). This variation can 271 also follow a seasonal pattern. Several decapod and fish species appear to modify their 272 depth ranges from autumn-winter to spring-summer (Figure 2; Aguzzi et al. 2009a). In 273 benthic species, the number of individuals collected at a given depth and location may 274 vary over the year, not only because of the reproductive cycle (i.e., the recruitment of 275 pelagic larvae into benthic adult cohorts; e.g., Company et al. 2003), but also due to the 276 modulation of behavioural rhythms in response to changing photoperiod lengths. Some 277 populations appear to move into darker, deep waters when the duration of the 278 photophase increases, while others move to shallower depths, where they become 279 exposed to brighter daytime environmental illumination levels (Aguzzi et al. 2009a, 280 Aguzzi & Company 2010). Such seasonal bathymetric changes have also been observed 281 in other continental margin and abyssal areas. For example, seasonal changes in the 282 population size-frequency distribution of grenadier fishes (Coryphaenoides spp.) have 283 been detected in both the Central North Pacific Ocean at a depth of 5800 m (Priede et al. 284 2003) and on the Porcupine abyssal plain of the North Atlantic Ocean at a depth of 4800 285 m (Smith et al. 1997). It was proposed that seasonal changes in size-frequency indicate 286 the occurrence of basin-scale seasonal migrations of these fishes across the seafloor into 287 and out of the study area.

288 Studies employing day-night scheduled trawling between 700 m and 1000 m 289 depths have also reported diel variations in the composition of communities (Cartes et al. 1993, Sardà et al. 2003), indicating the effect exerted by the few remaining photons
on the extremely sensitive visual organs of mid-slope species. Despite these
observations, population, biomass and biodiversity studies making extensive use of
oceanographic data very rarely consider solar light measurements as a potentially
helpful tool for interpreting data (e.g., Hart et al. 2010).

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296 The needs of a new observational technology

297 Changes in communities over a 24-h or a seasonal period within a benthic sampling 298 area represent the sum of all species rhythmic displacements that result from synchronic 299 movement of all individuals of a population within different depths of the water column 300 and the continental margin. In this sense, investigations of behavioural rhythms in 301 populations of different species in an area can be placed within the broad framework of 302 studies on ecosystem functioning (Kronfeld-Schor & Dayan 2003).

303 In this context, sampling methodologies and their applicability represent a major 304 bottleneck in understanding the biology of species inhabiting deep-water continental 305 margins and the deep sea. Trawling is still one of the most effective and more 306 economically feasible methods of sampling (Raffaelli et al. 2003, Sardà et al. 2004). 307 Trawling studies are broadly conducted over large seabed areas for the assessment of 308 the distribution and demography of populations (e.g., stock assessment), as well as for 309 overall biodiversity evaluations (Coll et al. 2010). Pluriannual surveys are common on 310 several of the world's continental margins for these purposes, for example, in the 311 Mediterranean (Relini & Piccinetti 1994, Bertrand et al. 2002), parts of the North 312 Atlantic, such as the Gulf of St. Lawrence (Bailey & Elner 1989, Hurlbut & Clay 1990), 313 the Bering Sea (Bakkala 1993), and the Pacific Ocean (Knuckey & Dichmon 2009).

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Automated video-imaging from cabled observatories for the study of

behavioural rhythms

317 The lack of tools for obtaining direct ethological observations leads to major knowledge 318 gaps regarding the behavioural mechanism underlying trawl sampling variability. 319 Technological limitations in direct observation capabilities are at the base of the 320 presently scarce modelling capacity regarding population/stock and biodiversity 321 assessments as well as ecosystem functioning in continental margin areas. Improvement 322 in this field requires a new observational technology to monitor community changes 323 produced by the synergic sum of behavioural rhythms of all constituting species. This 324 technology must be i) continuous and long-lasting with respect to the frequency and 325 duration of data collection; ii) automated in relation to the real-time processing of 326 biological data; and iii) multiparametric in terms of the integrated acquisition and 327 processing of biological and habitat data (see Section on Multiparametric measurements 328 and chronobiological analytic protocols for the characterisation of ecosystem 329 functioning).

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monitoring

Cabled seafloor observatories for multiparametric video and oceanographic

Behavioural rhythms of species within a community represent the core of ecosystem temporal functioning having still poorly known implications on results of assessment surveys in continental margin areas and the deep-sea. The study of rhythmic behaviour at diel and seasonal scales requires a technological development mostly directed towards coupled acquisition and analysis of diversified and enlarged datasets over periods of time larger than one year (Matabos et al. 2011).

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340 The state of art in video sampling technology and cabled observatories as innovative

341 technology

342 In the last 30 years, the development of video-imaging in association with multisensor 343 measurements for habitat characterisation has been applied to understanding patterns of 344 biomass and biodiversity variations at different spatial and temporal scales (reviewed by 345 Glover et al. 2010b). Technological effort has been applied to development of observational tools for remote exploration and monitoring of the sea. Remotely 346 347 Operated Vehicles (ROVs), Autonomous Underwater Vehicles (AUVs), and non-348 permanent camera stations have provided glimpses of the continental margin and, more 349 generally, of deep-sea ecosystem functioning over a range of temporal windows (Allen 350 et al. 1978, Stoner et al. 2008). However, none of these technologies can currently be 351 used to produce continuous time series of integrated biological, geological, and 352 oceanographic data over consecutive years, as is required for reliable monitoring of 353 benthic ecosystems in relation to present challenges of management and conservation.

354 Cabled multiparametric seafloor observatories connected to the shore for power 355 and real-time data transmission represent a substantial innovation in this respect. In 356 principle, the data provided by such platforms satisfies the above needs. Cabled 357 observatories use a vast array of instruments on submarine platforms, including 358 seismometers, acoustic current profilers, hydrophones, sonar, echo-sounders, 359 conductivity and temperature sensors, and pressure sensors (Aguzzi et al. 2011b). In 360 addition to imaging cameras, the biological tools available for these experiments 361 include plankton samplers, sediment traps capturing falling particulates to the deep-sea 362 floor, turbidity sensors, and benthic flow simulation chambers. Chemical and gas 363 sensors for measuring carbon dioxide, oxygen, methane, and nitrates are also available 364 (Matabos et al. 2011).

Sensors deployed over the seabed are only a part of the relatively complex 365 366 instrument arrays that can be set up in cabled observatories. Profiler instrument 367 packages installed at scientific nodes can provide multi-parametric measurements of 368 ecological processes within the water column above. An example of that is represented 369 by the Regional Scale Nodes associated with the US Ocean Observatory Initiative 370 (OOI) (see Section on Networking of seafloor cabled observatories as a tool for future 371 predictive and informative studies); these are sensors attached to profilers that vertically 372 traverse the ocean from just above the seafloor to the subsurface. CTDs, dissolved 373 oxygen sensors, current meters, and fluorometers are expected to be installed in these 374 profiler instrument packages to provide temporal and spatial measurements over almost 375 the entire oceanic depth range.

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377 A complex management scenario for data acquisition

Data acquisition still represents a critical step depending on sensor typologies and specifications (Majumder et al. 2002). In the three-dimensional context of the seabed including the overlying water column, multiparametric coordinated data collection requires establishment of acquisition and management protocols to optimise these processes (Bahamon et al. 2011). For example, even a single habitat parameter, such as the temperature from CTDs, yields highly complex information when measured repeatedly over time at different depths, including the seabed (Figure. 3).

Interest in database technology is growing as new tools become available, such as Structured Query Language (SQL) applications that are capable of storing and managing large volumes of spatial and time series of data, though they fail to support multi-dimensional arrays as a core data type (Aguzzi et al. 2011b). Relational mapping methods (Barry & Stanienda 1998) are gaining growing interest as an alternative database technology to develop an object-relational approach to effectively create a new interface for an existing database management system, thus allowing object-orienteddata and relational data to be combined in a single framework.

393 Data fusion is a fairly new informatics discipline, as it represents the capacity to 394 integrate data from different instruments (Klein 2004, Mitchell 2007). Technologies 395 should be adopted if they can be constantly calibrated and in which inter-instrumental 396 transfer of data is possible. A good example of this type of technology, though from a 397 different field, is provided by meteorological stations that can be coupled with networks 398 used to forecast weather conditions in relation to animal health and production 399 (Menesatti et al. 2009b). Thus, local dairy cow housing environments can be monitored 400 by sensors of different types integrated into the same managing system. Within these 401 networks, only a few stations for which calibrated instruments are available can be used 402 for forecasting purposes.

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404 *Video-imaging as novel bio-sensor approach*

Geophysical and oceanographic sensors are able to directly measure parameters driving processes of interest. In contrast, one of the difficulties associated with marine biology research is the lack of sensors capable of life functions directly at the level of individuals, species and communities. Most bio-sensors are indirect, since they measure products of life processes in terms of chemical derivates (e.g., dissolved oxygen, chlorophyll or nitrates; Aguzzi et al. 2011b).

411 Cabled observatories often have video cameras (Favali & Beranzoli 2006, Favali 412 et al. 2010). In relation to benthic ecosystems, these cameras potentially represent the 413 primary effective bio-sensor at individual animal, population, and species levels, 414 provided that automated image analysis is sufficiently developed to classify animals 415 based on differences in shape, texture and colour. Video cameras may thus be used to 416 efficiently detect fluctuations in the numbers of individuals of different species over 417 time as a proxy of population behavioural rhythms (Aguzzi et al. 2009c, 2010b). When 418 image time series are coupled with other physical parameters collected synchronously, 419 information on rhythmic activities can be related to environmental fluctuations within 420 the habitat under investigation (Matabos et al. 2011). This could be helpful in searching 421 for the geophysical cycles and processes ultimately controlling the behaviour of 422 individuals and populations of different species as one of the major driving forces of 423 ecosystem functioning (see Section on Multiparametric measurements and 424 chronobiological analytic protocols for the characterisation of ecosystem functioning).

425 Previous time-lapse imagery studies carried out with semi-autonomous 426 observatories (e.g., landers) have indicated the effectiveness of cameras for the 427 characterisation of ecosystem functioning. Kemp et al. (2008) described the DOBO 428 system (Deep Ocean Benthic Observatory), which is capable of remaining on the 429 seafloor for up to 12 months releasing bait at regular intervals. Time series observations 430 were obtained for squat lobsters (*Munidopsis* spp.) and macrourid and zoarcid fishes at 431 a depth of 3664 m in the Atlantic Ocean. This type of system is capable of providing 432 information during winter months when sampling using surface vessels is not feasible 433 due to adverse weather conditions. DOBO is also equipped with multiple sensors for 434 conductivity, temperature and pressure and an Acoustic Doppler Current profiler.

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video imaging

Investigation of biological rhythms of marine populations through time lapse

Video cameras on cabled seafloor observatories are often installed as an accessory to the primary purpose of acquisition of geological, chemical, and physical data. Hence these video sources are rarely used for continuous, long-term ecosystem surveying. One limitation on their use is the lack of an acceptable level of automation of image processing. (Costa et al. 2011). Human inspection of videos is time-consuming and represents an important bottleneck for the extraction of quantitative information from
video footage (Walther et al. 2004, Cline et al. 2007, Glover et al. 2010b).

During the past two decades automated video-imaging techniques have been increasingly used to extract quantitative biological information for a wide variety of land and marine species (Table 1). This information includes not only the shape and the colour of organisms for the purpose of achieving suitable levels of automatic discrimination of species (i.e., classification) once identified in frames, but also the ability to follow and quantify animal behaviour (i.e., tracking).

451 Implementing procedures for animal tracking and classification is crucial for 452 automated video-image analysis related to investigations of the behavioural rhythms of 453 marine populations. Tracking is the process of identifying the same animal as it 454 displaces within a set of temporally consecutive frames (Voss & Zeil 1995, Lipton et al. 455 1998, Chraskova et al. 1999, Edgington et al. 2004, Walther et al. 2004). Classification 456 is the grouping of each newly identified animal within a pre-established category, which 457 is usually the species (Strachan & Nesvadba 1990, Lipton et al. 1998, Dah-Jye et al. 458 2008). While tracking is critical for characterisation of behavioural rhythms (i.e., the 459 counting of individuals over time), classification is necessary for the characterisation of 460 communities at a local scale.

461 Tracking procedures should be considered at different temporal scales in relation 462 to two major aspects of animal behaviour (Aguzzi et al. 2009c,d, Menesatti et al. 463 2009a): i) the average temporal presence of an organism in the Field Of View (FOV), 464 according to its speed of displacement; and ii), the average time interval required by an 465 organism to re-occur in the FOV, according to the periodicity of its behaviour rhythm. 466 In relation to the first aspect, frame collection should be set at frequencies 4-5 times 467 higher than the average probability of spotting a targeted organism in the FOV 468 (Nyquist-Shannon frequency; Shannon 1949). Regarding the second aspect, video

469 acquisition has to be carried out over a temporal window encompassing several cycles 470 (equivalent to the periodicity) in the fluctuation of a certain habitat driver to allow 471 consistent and reliable time-series analysis (see Section on Multiparametric 472 measurements and chronobiological analytic protocols for the characterisation of 473 ecosystem functioning). This requirement is particularly strong in field studies in which 474 the measured rhythms are noisy due to the variable level of synchronism in the activities 475 of all individuals constituting the population (Aguzzi et al. 2010a,b, Azzurro et al. 476 2011).

477 Depending on frame acquisition frequency, video-imaging techniques can be 478 broadly subdivided in two major groups: 'single frame' (i.e., a collection of temporally 479 isolated and still images, as in the case of 'time-lapse photography' (Sherman & Smith 480 2009) and 'motion vision', the continuous acquisition of frames at ≥ 24 fps 481 conventionally known as "video footage" (Aguzzi et al. 2010b). In the first mode, the 482 frequency of acquisition tends to be lower than the movement rate of animals, which 483 implies that tracking cannot be performed, and only classification is achievable. In the 484 motion vision mode, the frequency of image acquisition is usually higher than 485 displacement velocity, and tracking of animals through consecutive frames can be 486 carried out together with the computation of their trajectories of displacement. Frame 487 interval should be chosen to match the expected movement rate of the species.

Estimation of local population abundances by video-imaging is hindered by the problem of repeat counting of individuals as they move around in the FOV. (Trenkel et al. 2004). One way of avoiding this is by estimating displacement trajectories and subsequently eliminating all initially counted individuals within the same frame set (Smith et al. 1993). Video studies on behavioural rhythms of populations may produce biases in the estimation of counted individuals when trajectories are not considered. However, because those biases are constant throughout the sampling period, they do not 495 mask inherent fluctuations in the individuals counted as a result of behaviour rhythms496 (Aguzzi et al. 2010b).

497 Having identified displacing animals in single frames, the problem reduces to 498 tracking their displacement across a set of temporally consecutive frames. Assuming 499 that an animal is already tracked in a temporally continuous group of frames, certain 500 algorithms can be used to match the identified animal in a new frame, hence connecting 501 it to the precedent trajectory. This assignation can be done by comparing an object's 502 position with its expected position for each event, extrapolated from its positions in the 503 past with each object being assigned to the event that it matches best (Edgington et al. 504 2004). Computation of trajectories (spatial tracking) can be implemented using Kalman 505 Filters (Cline et al. 2007) as a group of mathematical equations providing an efficient 506 and recursive solution to the method of minimum squares (Lauritzen 1981). This 507 solution allows computing an optimum statistical spatial estimation of an object at any 508 time point. For this purpose, the available information at time t - 1 is updated with the 509 information during the following t moment. Another algorithmic approach includes 510 recursive Bayesian filters (Betke et al. 2007). This approach solves the problem of 511 tracking a single object in an image sequence recursively by estimating the state (s) of 512 the object in the current frame (t), defined as s(t), based on its state in the previous 513 frame, s(t - 1), and by filtering measurement x(t) in the current frame.

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Methods for the automated classification of tracked animals

516 Identification of individuals to species or classification is fundamental to the study of 517 behavioural rhythms using automated video image analysis. The automated counting of 518 individuals can be efficiently carried out only after a suitable level of efficiency in the 519 pattern recognition of targeted species is achieved. Classification can be carried out by 520 considering the outlines of animals (i.e., the profile given by the silhouette) or their521 colour and physical properties.

522 The 'Pattern Recognition problem' consists of determining a procedure that can 523 assign each entity to its proper class on the basis of its attributes. Automated systems for 524 pattern recognition and classification are based on a supervised approach and, for this 525 reason, should be tuned in their efficiency by a training process. This programming procedure consists of a 'visual census' performed by trained operators on a subset of 526 527 data for the entities considered, with this subset being referred to as the training set. 528 Thus, a training set consists of entities characterised by a set of attributes that belong to known classes (Nieddu & Patrizi 2000). In the case of imaging products from cabled 529 530 observatories, training sets are represented by a collection of selected images depicting 531 the targeted species viewed from different angles. (Aguzzi et al. 2011b).

532

533 Statistical tools for classification

534 Multivariate classification analyses can be generally divided into two main categories: 535 unsupervised and supervised (Costa et al. 2011). For unsupervised techniques, grouping 536 or clustering methods for multivariate elements (x-block) are based on functional 537 relationships among these elements (i.e. distances or variances). They do not require a 538 priori knowledge of the class categories. By contrast, in supervised techniques, class 539 attribution is provided by single or multiple variables (y-block). Thus, multivariate 540 methods are forced to cluster into *a priori* established classes. Unsupervised methods 541 are mainly applied in an exploratory sense, when the aim is to analyse or visualise non-542 forced aggregating relationships (unsupervised) among elements.

543 Concerning supervised techniques, it is possible to distinguish two main 544 analytical approaches: classification and modelling. Classification analysis requires a 545 decision rule, referred to as the 'classification criterion', to sort objects into classes on

546 the basis of selected quantitative features (Jayas et al. 2000). By contrast, in modelling, 547 it is possible to attribute objects not only into one or more classes but also to none (i.e., 548 in this case, the object is an outlier). Modelling techniques calculate the 'prediction 549 probability' using a classification threshold for each modelled class (Costa et al. 2011). 550 The modelling efficiency is indicated by statistical parameters, such as 'sensitivity' and 551 'specificity'. Sensitivity represents the percentage of the objects of a category accepted by the modelled class. Specificity is the percentage of objects different from the 552 553 modelled classes that were rejected by the classification criterion. However, for 554 classification, a matrix of correct classifications can be used (Costa et al. 2008b).

555 The choice of suitable statistics to treat the different categories of data (i.e., 556 morphometric or colorimetric descriptors) represents an important step in the 557 development of automated video image analysis. The many different algorithms that can 558 be used for classification can be summarised into the following categories:

- 559 1. Probabilistic-based:
- 560 Parametric Linear Discriminant Analysis (Li & Yuan 2004) and
 561 Quadratic Discriminant Analysis (Wu et al. 1996).
- 562 Non parametric Potential Function (Pei & Xie 1999).
- 563
 2. Distance-based: k-nearest neighbour (Findik et al. 2010), Soft
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 Fartial Least Squared Discriminant Analysis (Costa et al. 2008b).
- 5663. Experience-based: Support Vector Machine (Amit & Boaz 2005) and567Artificial Neural Networks (Costa et al. 2009c).

As stated by Mitchie et al. (1994) "...Machine Learning aims to generate classifying expressions simple enough to be understood easily by humans. They must mimic human reasoning sufficiently well to provide insight into the decision process...". Machine-learning approaches can be implemented by combining video572 acquired morphological descriptors and multivariate statistics. Existing methods are 573 based on random forest classifiers (i.e., an ensemble of decision trees, where each tree is 574 a computer-generated structure analogous to a dichotomous key) taken directly from 575 descriptors (reviewed by Lytle et al. 2010). In a random forest, each tree is constructed 576 using a process that incorporates a random component to introduce variation amongst 577 the trees.

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579 The fuzzy logic approach in animals' classification

580 In extreme environments such as the deep sea, where the wide range of animal diversity 581 is often currently unexplained, the use of a fuzzy logic approach within object 582 classification could return an interesting way of classifying new (not yet classified) 583 objects in a trained guild. A fuzzy set is a set in which the elements have degrees of 584 membership. An element of a fuzzy set can be full member, or a partial membership 585 value can be assigned to an element that is no longer restricted to only two values; 0 or 586 1 but also any value in between (Bahamon et al. 2009). Fuzzy systems have the 587 capability of representing classification decisions explicitly in the form of fuzzy 'if-then' 588 rules. Fuzzy sets allow the assignment of partial- and multiple-value memberships. 589 Fuzzy systems use vague, imprecise or uncertain information to generate simpler, more 590 suitable models that are easier to handle and more familiar to human thinking. 591 Concerning quantitative features for the automated classification of marine organisms, 592 these classifications can be based on shape, colour, texture or a Scale-Invariant Feature 593 Transform (SIFT) algorithm (see below).

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595 Morphometric and colorimetric descriptors for automated animals' classification

596 For quantitative image analysis (i.e., chromatic or hyperspectral derived) different 597 elaborative steps finalised altogether to enhance the chromatic properties (contrast, equalisation, amplification, denoising, smoothing, and etc.) and to extract the sensible part of the image can be used to differentiate objects from the background (segmentation and thresholding). This procedure can be used to measure sensible objects by morphometric, optical densing or texture mapping (Bharati 2004, Li 2008, Lillholm 2003, Huang 2003). Among the different features to be extracted and processed with image analysis tools, shape and colour are some of the most important and easy to be extracted.

The shape of a given animal can be automatically classified and assigned to a species by many different methods, including univariate techniques, such as fractal indices or ratios, or multivariate techniques, such as Fourier methods (Rohlf & Archie 1984). These tools allow the recognition of an organism through the fitting of its profile with a set of harmonic functions.

610 In Elliptic Fourier Analysis (EFA), the animals' shape can be studied by profile 611 digitisation (Figure 4; Costa et al. 2009a, 2010, 2011). A set of points is automatically 612 positioned along the outline in an equiangular or equidistant fashion. The contour is 613 transformed into an incremental harmonic function when a set of x, y coordinates is 614 computed for each point. That function is then fitted with a set of different ellipses to 615 approximate its variation with the highest precision. At the end of the fitting process, 616 each biological sample is represented by a set of ellipses, each of which has four 617 coefficients. As a result, a matrix comprising all individuals with their respective ellipse 618 coefficients can be obtained. This matrix is the input required for multivariate statistical 619 analysis (Costa et al. 2011).

Fourier Descriptors (FD) are also employed for the automated recognition of tracked animals during the study of behavioural rhythms (Toth & Aach 2003, Veeraraghavan et al. 2005). FD can be utilised to describe the shape outline of a biological form in terms of its frequency variability (Aguzzi et al. 2009c). This 628 SIFT is an algorithm employed in computer imaging used to detect and describe 629 local features in images (Lowe 1999). Interesting points of a particular organism can be 630 extracted to provide a 'feature description' of the object. This description, extracted 631 from a training image, can then be used to identify the object in a test image containing 632 many other objects (Figure 5; Aguzzi et al. 2011a). A significant requirement of this 633 process is that the set of features extracted from the training image must be robust 634 enough to allow efficient identification, despite changes in image scale, noise, 635 illumination, and local geometric distortion. This method robustly identifies objects 636 even under conditions of clutter and partial occlusion because the SIFT feature 637 descriptor is invariant to scale, orientation, and affine distortion and is partially 638 invariant to illumination changes (Lytle et al. 2010).

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population behavioural rhythms

Flowcharts of image processing and animal classification for the study of

642 Different automated video-imaging trials have been elaborated in the past few years 643 when working with imaging products from cabled observatories in different continental 644 margin areas. The procedure of image treatment at each video station necessarily requires the implementation of specific algorithms, due to different light levels in 645 646 relation to artificial lighting conditions, as well as turbidity, fouling and substrate 647 texture/heterogeneity. Based on tracking and classification processes, general image 648 treatment flowcharts can be proposed for the automated processing of time series of 649 images collected by cabled observatories (e.g., the Canadian VENUS in Saanich Inlet or the Japanese Sagami Bay observatories at 103 and 1100 m depth, respectively; see on
Networking of seafloor cabled observatories as a tool for future predictive and
informative studies), to study population behavioural rhythms.

653 Colour features, numerically defined by coordinates in n-dimensional colour 654 spaces, can be important descriptors for animal classification (Bruno & Svoronos 2005). 655 For example, in combination with shape analysis, the RGB (Red-Green-Blue) content of 656 organisms can be added to increase recognition efficiency. Morphological classification 657 can be then based on EFA, FD, SIFT, and FD + RGB semi-automated procedures 658 (Aguzzi et al. 2009c, 2010b, 2011a). Each tracked animal can be recognised on the basis 659 of its set of FDs and its average colour content coordinates (Figure 5; Aguzzi et al. 660 2011a). A library of manually supervised and classified images is required for each 661 target species (Storbeck & Daan 2001, Culverhouse et al. 2003, Edgington et al. 2006). 662 A subset of frames is chosen, and manual classification of animals is performed by a 663 human observer through interactive scripts. Images of animals from different angles can 664 be saved in a binary format to allow later morphometric comparisons and recognition. 665 Moving animals are identified and assigned to species as pre-established categories by 666 using multivariate supervised statistics, which allow constructing colorimetric and 667 morphological models of reference (see the next section).

668 The acquired frames are processed for grey-scale transformation and 669 segmentation (i.e., binarisation of the coloured image in black and white) (Figure 6A, 670 B). Subsequent subtraction of consecutive frames is then applied to detect all moving 671 objects, and areal thresholding permits selecting object sizes of interest that correspond 672 to the general body size of animals (Figure 6C). The profile of identified objects can 673 then be superimposed on the original image according to an area that corresponds to the 674 animal's body with its peculiar RGB coordinates (Figure 6D). Both the animals' 675 profiles and their RGB contents can be subsequently classified by multivariate

676	supervised statistics (i.e., K-Nearest Neighbour, KNN or Partial Least Square
677	Discriminant Analysis, PLSDA; reviewed by Costa et al. 2011), according to
678	morphological models of reference extracted from libraries of manually classified
679	images (i.e., the training set).

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Acoustic tracking as a complementary method to automated video-imaging in investigations of the behaviour of marine species

A concern related to use of video-imaging is that the lights necessary to acquire the images may influence the behaviour of the organisms being observed. Widder et al. (2005) used red light to illuminate a scene and demonstrated differences in the fauna observed. The use of acoustic imaging may solve problems of photic contamination (i.e., lighting at moments of frame/footage acquisition) at depths below the twilight zone.

689 Active sonar sensors oriented vertically upwards from a sea floor observatory 690 seem to present a scope similar to video cameras in relation to population behavioural 691 rhythms. These instruments allow depiction of zooplankton and pelagic fish 692 displacements within the pelagic environment (reviewed by Lemon et al. 2007, 693 Kaartvedt et al. 2009, Borstad et al. 2010, Cox & Borchers 2011). Horne et al. (2010) 694 deployed the Deepwater Echo Integrating Marine Observatory System (DEIMOS) with 695 Simrad 38 kHz upward-looking sonar at a depth of 890 m in Monterey Bay off the coast 696 of California and recorded seasonal variability in diel migrations. Borstad et al. (2010) 697 recorded a 2-year time series using a 200 kHz system deployed in Saanich Inlet, British 698 Columbia and made major progress in the analysis and presentation of this type of large 699 dataset, which was derived from a 0.3 Hz ping rate. The Deep-sea Long-term 700 Observatory System (DELOS) at 1500 m depth off Angola is not presently cabled, 701 being retrieved yearly for data downloading (see Section on Networking of seafloor

cabled observatories as a tool for future predictive and informative studies). This
multiparametric platform is also equipped with Kongsberg high resolution 675 kHz
scanning sonar oriented horizontally outward to a range of 150 m. This positioning
enables tracking movements of fishes and other targets around the observatory.
Appropriately configured arrays can track whales in three dimensions, identifying
different types of behaviours and feeding events (Nosal & Fraser 2007).

708 Acoustic technology in general, does not allow ultimate species classification, 709 although multiple emissions frequency and resulting echo-trace analysis (using 710 frequency responses) can distinguish the sizes of moving animals, thus allowing some distinctions to be made (Petitgas et al. 2003, Jech & Michaels 2006, Reier Knudsen & 711 712 Larsson 2009). Acoustic video cameras could be used in the benthic environment to 713 create high-resolution images of moving animals, similarly to what can be done using 714 video-imaging systems (Mueller et al. 2006). However, the problem of species 715 classification on acoustic-image products still remains. Acoustic imaging could 716 efficiently replace video cameras only in cases when targeted taxa show very different 717 morphologies (i.e., without the requirement of distinction based on colour and its 718 patterning) (Boswell et al. 2008). Recent methodological improvements reached an 719 acceptable level of acoustic classification with a reduced group of small fishes under 720 laboratory controlled conditions (Roberts et al. 2011).

An alternative to optical methods for observing the behaviour of organisms is acoustic passive and active tracking, together with the use of tags or transponders (Sarriá et al. 2001). As an example of this, the DELOS platform is equipped with passive listening devices (Bagley et al. 2011). These sensors can detect sounds produced by crustaceans, fishes, and cetaceans. Individual fishes can be tracked using ingestible transponders that are tracked by sonar mounted on an observatory platform (Priede & Bagley 2000). Using ingestible acoustic transmitters, Priede et al. (1994)

728 detected seasonal changes in the swimming speeds of abyssal grenadier fishes in the NE 729 Pacific Ocean. The global Ocean Tracking Network (OTN; O'Dor et al. 2010) uses a 730 chain of acoustic observatories around the world to detect fishes with surgically-731 implanted acoustic identity tags (O'Dor & Stokesbury 2009). Using multiple 732 observatories, OTN will be able to track mobile species throughout their annual or 733 multi-annual migration patterns. The data and power capabilities of cabled observatories 734 are enabling great advances to be made in the application of acoustic methods to long-735 term time series research.

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the characterisation of ecosystem functioning

Multiparametric measurements and chronobiological analytic protocols for

739 The temporal functioning of continental margin and deep sea communities according to 740 habitat cycles at diel and seasonal scales is essentially unknown at present (Barans et al. 741 2005, Wagner et al. 2007). Reports on species presence and abundance from the deep 742 ocean are too limited in both space and time mainly because of observational 743 technological limitations (see Section on The missing chronobiology framework in 744 marine ecological studies). Marine species display rhythmic behaviours in response to 745 geophysical cycles, and these responses affect observable populations during field 746 sampling. This phenomenon influences population/stock and biodiversity assessments.

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From descriptive to more cause-effect-based deep water ecology

The integrated processing of biological and habitat information is a pending issue in continental margin ecology and represents one of the most promising ways to move beyond the current, chiefly descriptive, approach. The transition from descriptive to a more cause-effect-based marine research could be achieved by implementing the conceptual framework of integrated time-series analysis (Dickey & Bidigare 2005, 757 The temporal regulation of physiological processes within a whole organism 758 involves the clock-coordinated activities of oscillators in the central nervous system and 759 in several peripheral tissues (Dibner et al. 2010, Harrington 2010). The centralised 760 control of biological clocks over all biological rhythms in an animal's body can be 761 studied by characterising the direction of flow of regulating information (Yamazaki et 762 al. 2000). This characterisation can be methodologically approached by obtaining 763 synchronic time series for different physiological or molecular indicators in different 764 tissues and organs and by detecting the peaks of rhythms as markers of the activity of 765 the oscillator (Fukuhara et al. 2005). This type of comparison seeks anticipations and 766 delays in the peaks of each oscillator in relation to each other, which could yield 767 information on the direction of the flow of controlling information (Tosini and Aguzzi 768 2005). Similarly, in the case of multiparametric data acquisition from seafloor stations, 769 time series of several habitat parameters should be compared and cross-checked to 770 portray which incremental combination of such parameters causes a subsequent 771 biological response (i.e., peaks in the counted individuals of a given population).

772 The statistical tools required for such analyses are already available in 773 commercial chronobiological software applications which are used to statistically 774 describe rhythms at molecular, physiological and behavioural level (e.g., Aguzzi & 775 Chiesa 2005, Aguzzi et al. 2006b; Tosini et al. 2007). However, further development is 776 needed for these tools to become more user-friendly, to fully satisfy the needs of 777 biological and habitat data multiparametric analyses in field study of populations, i.e., in 778 terms of applying automated sequential steps for data pre-processing (i.e., filtering; see 779 below) and consequent time series analysis.

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Integrated waveform analysis as a protocol for multiparametric data treatment A general protocol for the automatic and integrated treatment of bio- and habitat data can be proposed (Figure 7) according to the principle of anticipation and delay as a marker of cause-effect relationships between habitat fluctuations (i.e., the cause) and

784 785 consequent population rhythms (i.e., the effect). Firstly, time series of biological 786 observations (i.e., the number of video-counted individuals for each selected species) 787 should be acquired with a frequency that corresponds to the timing of the recording of 788 concomitant habitat parameters (Figure 7A). Multiparametric data banks should then be 789 implemented to store such diversified information in columns (i.e., for all species and 790 selected habitat parameters), all starting with the same reference time. Secondly, time 791 series should be pre-processed for data filtering to eliminate excess noise (e.g., moving 792 average and binning or outlier removal; see review in Levine et al. 2002) (Figure 7B). 793 Thirdly, pre-processed time series should then be screened for inherent periodicity with 794 periodogram functions (Figure 7C).

795 Waveform computing represents the fourth step in the proposed data analysis 796 protocol, which, when successfully carried out in an integrated fashion, could lead to 797 the discovery of interesting hints in order to discuss any potential cause-effect 798 relationships between habitat fluctuations and the behavioural response of species 799 (Aguzzi et al. 2004a,c). Although phase-relationship associations in visual counting of 800 species (i.e., by video-imaging) with the light-dark or tidal cycles do not demonstrate 801 per se a causal relationship, integrated waveform analysis provide a temporal linkage 802 between the circadian system of species and several potential fluctuating environmental 803 cues (i.e., zeitgebers). In the laboratory, the phase relationship of behavioural rhythms 804 associated with the light-dark cycle accounts for the diurnal or nocturnal niche of 805 species, demonstrating under controlled conditions a causal relationship between behaviour, underlying physiology, and light intensity variation (Aguzzi et al. 2005,
Chiesa et al. 2010). The same correlation between behavioural rhythms and physical
variables does not necessarily imply entrainment in the field. However, this represents a
good attempt at description of ecosystem dynamism based on behavioural changes of
individuals constituting the community, within the framework of surrounding variable
habitat conditions.

812 Integrated waveform analysis can be performed as follows. Time series of both 813 behavioural and environmental variables can be subdivided into sub-sets with a length 814 corresponding to the periodicity identified in periodogram analysis (for example, of 24-815 h length for the day-night cycle or of 12.4-h length for the tidal cycle; Figure 7D). 816 Values from all subsets of bio- and habitat data should then be averaged for 817 corresponding data bins, thus obtaining a consensus average curve (Figure 7E). 818 Significant increments (i.e., onset and offset as phase markers of rhythm) can then be 819 identified from values above the Midline Estimating Statistic of Rhythm (MESOR; 820 Aguzzi et al. 2003a,b). The MESOR is conceptually equivalent to a daily mean as 821 threshold used for significance: all average estimates of the waveform are re-averaged, 822 and the resulting value is plotted onto the consensus curve as a threshold horizontal line. 823 Above-MESOR values represent the peak (i.e. as significant increment in the biological 824 fluctuation), the temporal amplitude of which can be represented as a continuous line in 825 time. The temporal limits of the activity/rest intervals associated with behavioural 826 rhythms in relation to measured habitat parameters can be assessed, and the indications 827 that they provide about peaks in temporal relationships as markers of a cause-effect 828 relationship are of particular interest (Aguzzi et al. 2005, 2006a,b, Costa et al. 2008a). 829 Onset of peaks of different biological and habitat waveforms can then be compared over 830 a standard 24-h cycle. Thus, the existence over consecutive cycles of significant 831 increases in behavioural activity prior to the increase in the state of fluctuation of other

habitat variables could represent the phenomenon of anticipation, as a marker of rhythm
entrainment (Figure 7F). Additionally, the waveform area calculated at a given time
interval can be used to estimate the amount of activity related to any endogenous or
exogenous component regulating behavioural rhythms (Chiesa et al. 2005, AnglèsPujolrès et al. 2006).

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and informative studies

Networking of seafloor cabled observatories as a tool for future predictive

840 Seafloor cabled observatory video studies on the behavioural rhythms of animal 841 populations provide an important source of data for modelling variability in sampling 842 related to these rhythms responding to changeable habitat conditions (Aguzzi et al. 843 2010a). Regardless of their quantification efficiency, local direct population and 844 biodiversity estimations using video methods should be undertaken very cautiously. 845 Derived animal counts should not be extrapolated to population demography given the 846 overly geographically localised character of the observations. Additionally, a lack of 847 records for a given species does not imply its absence within a geographical area 848 (Barans et al. 2005), as such records are challenged by remote video observations using 849 food (i.e., bait) attractors (e.g., Cappo et al. 2004). For example, abyssal demersal fishes 850 can be observed using free-fall lander technology, as they can be actively attracted to 851 bait placed within view of seafloor cameras (Priede & Bagley 2000).

Accordingly, data from a permanent video cabled observatory can only assist in understanding how temporal biases in population/stock and biodiversity assessments may occur if the timing of sampling is not considered. In the future, increasing knowledge about the behavioural rhythms of deep-water species and their modulation based on geophysical cycles will be of great value to researchers in modelling such biases in other sampling methods, such as trawling (Barans et al. 2005). 858

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Spatial coverage of existing or planned cabled observatory networks

860 Presently, insufficient spatial coverage greatly biases any potential estimate of 861 population demography and local biodiversity in continental margin areas. (Coll et al. 862 2010). The easiest solution to this problem would be increasing the number of 863 observational sites and the length of monitoring periods in a given area. An interesting 864 opportunity for extending the currently limited underwater monitoring capability in 865 terms of both space and time is presented by the creation of networks of permanent 866 multisensor cabled seafloor observatories, deployed from coastal areas to abyssal plains. 867 Although at the present state of technological development, this objective is not realistic 868 in the short term, the increase in the number of these platforms represents a real future 869 opportunity to perform prolonged observations in a wide range of habitat conditions 870 (Beranzoli et al. 2002, Favali & Beranzoli 2006, Favali et al. 2006a,b, Frugoni et al. 871 2006, Manuel et al. 2010).

872 A sustained stream of high quality interdisciplinary data on the marine 873 environment associated with appropriate sampling is being acquired to study 874 biological, geological and oceanographic interlinked phenomena (Lampitt et al. 2010). 875 Network configurations are particularly suitable for disciplines that require real-time 876 integration of different submarine nodes deployed over large areas. Several countries 877 and consortia have started long-term multiparametric monitoring initiatives by installing 878 individual or networks of permanent and integrated cabled observatories (Table 2 lists 879 marine multiparametric cabled seafloor observatories bearing video cameras) for a 880 variety of purposes over large geographic areas (Favali et al. 2006a, Favali & Beranzoli 881 2009, Favali et al. 2010). Although ecological discovery and ecosystem bio-monitoring 882 are not prime targets of these observatories, they often include associate science 883 components that could quite easily host the type of research under consideration here.

884 Countries currently driving large international programmes aimed at creating
885 underwater permanent cabled installations networks are Japan, Europe, the United
886 States, and Canada.

Japan is presently working on the development of alert systems for earthquakes and associated tsunamis, which is currently one of the major requirements in seismically active areas (reviewed by Priede et al. 2004, Favali & Beranzoli 2006, Favali et al. 2006a, Mikada et al. 2010). The Dense Oceanfloor Network system for Earthquakes and Tsunamis (DONET 2011) consists of a network of seismic observational nodes along several locations in the Japan Trench (i.e., off-Tokachi-Kushiro, Suruga, Boso, Izu Peninsula, Sanriku, and in Sagami Bay).

894 The European Seafloor Observatory Network (ESONET) EC project indicated in 895 its scientific objectives a number of observatory locations planned for different 896 continental margin depths and latitudes including the Arctic (Svalbard), the Norwegian 897 margin, the Nordic Seas (Faroese), Irish waters (Porcupine/Celtic; CELTNET), the 898 Azores, the Iberian area of the Gulf of Cadiz, the Ligurian Sea, the Western Ionian Sea 899 (Eastern Sicily), the Hellenic area, the Black Sea, the Marmara Sea, and the Canary 900 Islands (PLOCAN) (Favali et al. 2010). The large-scale research infrastructure 901 European Multidisciplinary Seafloor Observatory (EMSO 2011), forms part of the 902 European Strategy Forum on Research Infrastructures roadmap (ESFRI 2011) and 903 intends to develop permanent nodes of the underwater network at these sites. Within the 904 Central Mediterranean, aims are related to geophysical and oceanographic applications 905 and to the capture of neutrinos crossing the Earth by means of photomultiplier tubes 906 (Amram et al. 2000, Priede et al. 2008, Ageron et al. 2009, Aguilar et al. 2010, Favali et 907 al. 2011). In particular, the Western Ionian Sea cabled node of EMSO manages a variety 908 of different instruments covering the scientific fields of geophysics, physical 909 oceanography and bio-acoustics (Favali et al. 2006a, 2011) acquiring the data in real910 time. In the future it is planned to host video cameras at these locations.

911 Off the coast of the United States of America video cabled platforms are fully 912 operative in the deep-water Californian margin (e.g. the Eye-in-the-Sea ,EITS, 913 http://www.mbari.org/earth/mar_tech/EITS/eits_live.html from the Monterey 914 Accelerated Research System - MARS; see Table 2). Other platforms that could also be 915 easily equipped with cameras are the shallow-water Martha's Vineyard Coastal 916 Observatory (MVCO 2011) of Massachusetts' Katama Air Park and the Long-term 917 Environmental Observatory (LEO 2011) of New Jersey. The USA is also promoting the 918 complex Ocean Observatories Initiative (OOI 2011) based initially on the deployment 919 of permanent oceanographic buoys in the years from 2010 to 2014, which could be used 920 in future for the deployment of cabled observatories.

921 NEPTUNE Canada (NC; see Table 2) is an operating cabled regional network 922 across the Juan de Fuca plate in the Northeast Pacific off Vancouver Island, British 923 Columbia (Barnes et al. 2008), which is part of the Ocean Networks Canada 924 Observatory. The system now provides the continuous power and bandwidth to collect 925 integrated data on physical, chemical, geological, and biological gradients at temporal 926 resolutions relevant to the dynamics of the earth-ocean system. The location of this 927 system over an entire tectonic plate provides a unique setting in which to study the 928 influence of geological, physical, and tectonic processes on species dynamics and 929 biodiversity in a comprehensive manner at temporal scales ranging from minutes to 930 decades. Furthermore, that network encompasses a high diversity of ecosystems, 931 including a coastal embayment (with both hardground and soft-bottom environments), 932 the continental slope, a marine canyon, methane hydrate fields, and an abyssal plain 933 with hydrothermal vents. The scientific outputs of the network will be easily applicable 934 to other locations in the World Ocean. The NEPTUNE Canada regional network is
935 currently equipped with six video camera systems located at depths ranging from 20-936 980 m, and two additional systems will be deployed in 2011 at hydrothermal vents on 937 the Endeavour segment of the Juan de Fuca Ridge, at a depth of ca. 2190 m. These 938 cameras are operated through a web interface available anywhere on the planet, and 939 videos are freely accessible and available for download through the NC website 940 (NEPTUNE 2011).

941 Although, in several cases, the primary target of observatory deployment is not 942 image acquisition, presently installed nodes potentially represent a powerful laboratory 943 for investigating the modulation of behavioural rhythms of local animal populations 944 correlated with geophysical cycles (and events), such as those involving annual 945 formations of water masses, light intensity fluctuations, and internal tides or inertial 946 currents. Because of differences in geographic location, detected rhythms ultimately 947 respond to local latitude, oceanographic setting, water depth, seafloor morphology, and 948 the likely effects of climate change associated with global warming (Canals et al. 2006, 949 Somot et al. 2006).

950 The acquisition of biological data by the use of video-imaging systems will 951 likely become more widespread with the development of seafloor observatory 952 technology and networking. Multidisciplinary data acquisition is not currently being 953 conducted using standardised protocols of reference; this lack of standardisation inhibits 954 interdisciplinary analysis, as will be required, for example, to improve geo-hazard 955 warning and to evaluate the influences of climate change and anthropogenic impacts 956 (Ruhl et al. 2011). Consequently the scientific community is in urgent need of standard 957 protocols to acquire biological and habitat data with these platforms in different areas of 958 the planet, to provide a global understanding of the dynamics of marine ecosystems and 959 predict their evolution in the context of changing oceans.

960

Future issues to address in the use of cable observatory technology

With the increase in, the number of cabled video observatories, the efficiency of automated image processing (i.e., biodata), and the integration of biological and habitat data analysis, it will be possible to undertake a series of studies of general interest within the context of novel ocean observing framework topics (IFSOO 2011).

966

967 Ecosystem dynamic functioning in areas with geophysical cycles of conflicting 968 periodicity

In the near future, the use of video observation networks should allow for the study of the oceanographic modulation of behavioural rhythms in species with a wide distribution range, which are, thus, subjected to different combinations of day-night, internal tidal or inertial current cycles. Within the twilight zone, the light intensity cycle varies as a function of depth and latitude, the latter acting through the poleward decrease in angle of elevation of the sun (Fraenkel 1986).

975 Day-night shifts in seabed-living or demersal communities are found at 976 continental margin depths within the twilight zone stratum at all latitudes from the 977 tropical to polar regions such as off Antarctica (Kaufmann et al. 2003). Studies based on 978 temporally scheduled trawling along continental slopes, rises, and abyssal plains, show 979 that day-night variations in catches resulting from species behavioural rhythms are 980 widespread, showing great potential for installation of video observatories (Figure 8A), 981 At the same time, complex propagation patterns of internal tidal waves may occur due 982 to the effect of seafloor topography (Figure 8B) (Garret 2003, Garret & Kunze 2007), 983 which would lead to the occurrence of semidiurnal or inertial behavioural rhythms in 984 populations (Wagner et al. 2007, Aguzzi et al. 2010b).

985 From a global perspective, the deployment of an integrated network of 986 permanent video observatories is fully justified for the purpose of understanding 987 geographic differences in community responses under mixed day/night, internal tidal, 988 and inertial regimes. Observatories located in shelf and slope areas, where productivity 989 is generally low and transparency high, are particularly well suited for the study of diel 990 and seasonal behavioural rhythms. Other observatories in the almost tideless, 991 oligotrophic Mediterranean Sea represent excellent platforms for the study of day-night-92 based community rhythms (reviewed by Aguzzi et al. 2009e).

993 By contrast, remote monitoring by observatory sites within coastal turbid areas 994 (including estuaries) may not be as useful because of inherent difficulties in tracking 995 animals and classifying them in poor visibility or when fouling occurs (Aguzzi et al. 996 2009c). Other observatories located in ocean areas associated with strong tidal motion 997 could be used to identify community responses to water speed increases, irrespective of 998 the level of sunlight exposure. Furthermore, deeper observatories in the Mediterranean 999 Basin may facilitate the investigation of behavioural rhythms correlated with 1000 geophysical cycles involving water displacement that are much weaker than those 1001 produced by internal tides, such as atmospherically driven deep water inertial currents 1002 with an 18-h periodicity (reviewed by Aguzzi et al. 2009d).

1003

1004 The role of large diel nektobenthic and benthopelagic migrations in the indirect day1005 night regulation of deep-sea ecosystems

The implementation of integrated permanent video observation networks should be useful for studying how the behavioural rhythms of continental margin species act as regulators of ecosystem functioning in different depth strata. Species receiving daynight-based light intensity entrainment in superficial depth strata may influence the functioning of deeper ecosystems in aphotic continental margin areas. Nektobenthic migrants likely perform large (i.e., few kilometres) diel horizontal migrations (DHMs) by analogy with diel vertical migrations (DVMs; see Section The missing 1013 chronobiology framework in marine ecological studies), the extension of which is still 1014 poorly described. Moreover, the occurrence and nektobenthic typology of rhythmic 1015 behaviour in the majority of continental margin species is currently poorly described. In 1016 contrast to the pelagic environment, where vertical population displacements have been 1017 sampled by by temporally coordinated net tows at different depths, such a sampling 1018 design has not often been used in shelf and slope trawl surveys (Aguzzi & Company 1019 2010).

1020 The relevance of DHMs is currently largely underestimated by marine ecology, 1021 despite their potential to involve a number of different fish and decapod species 1022 (Suetsugu & Otha 2005, Benoit-Bird & Au 2006). The ecological and adaptive 1023 implications of DHM movements could be similar to those already proposed for DVMs 1024 in the pelagic environment (Zaret & Suffern 1976, Aksens & Giske 1993, De Robertis 1025 2002, Naylor 2006). The activation/suppression induced by DHMs with respect to the 1026 behaviour of seabed deep-sea species might follow a pattern similar to the one proposed 1027 in the scattering layer-interception hypothesis (i.e., diel changes in behavioural 1028 activation are the product of predator/ prey interactions; reviewed by Fock et al. 2002).

1029 Video-image acquisition from bathymetrically coupled video stations, 1030 functioning continuously for 24-h in a synchronized manner (Figure 9), could be used to 1031 identify which species are DHM movers and which adopt a benthopelagic behaviour 1032 (i.e., as a sub- category within DVM; reviewed by Aguzzi and Company 2010). 1033 Unveiling temporal shifts in the number of video-counted individuals on continental 1034 shelves and slopes for a given species at equivalent time points could be considered as 1035 markers of DHM displacements. Conversely, the presence of fluctuations in the number 1036 of video-counted individuals only at a single station, usually the deeper, may indicate 1037 the occurrence of benthopelagic DVMs (Aguzzi et al. 2006c). It is likely that slope-1038 positioned, cabled observatories will be able to detect DVMs in shrimps shifting 1039 between the seabed and the overlying benthic boundary layer (i.e. benthopelagic 1040 movements). These diel displacements may rhythmically bring animals within the 1041 camera FOV during certain parts of the day-night cycle (usually the night), similar to 1042 results from temporized bottom trawling (Aguzzi et al. 2006c). This phenomenon is 1043 likely to be more evident on slopes than shelf regions because of the greater depth range 1044 available for vertical movements. In addition to intrinsic behaviour, predator diel migrations are also influenced by movements of macro zooplankton or micronekton 1045 1046 prey species in relation to environmental light and its effect on visual predation (Pusc et 1047 al. 2002).

Both DVM and DHM movements affect the presence of predators and prey over 24-h periods in deep water areas below the twilight zone. These movements may represent an indirect signal for the behavioural synchronisation to the day-night cycle of species residing in deep-sea areas. The NEPTUNE platforms located in Barkley canyon at 400 m and 920 m (see Table 2) offer the opportunity of studying these nektobenthic movements in a coordinated fashion and their effect on ecosystem functioning at different depths on the continental margin (Huchet et al. 2010).

1055

1056 Indirect seasonal (i.e., photoperiodic) regulation of deep-sea ecosystems:
1057 characterisation of endogenous and exogenous signals for growth and reproduction

The existence of a mechanism controlling behaviour on both diel and seasonal bases through detection of photophase or scotophase duration (i.e., photoperiodism) is currently under discussion for several vertebrate and invertebrate animal models (reviewed in Refinetti 2006). Such a common mechanism could explain reported rhythms related to the growth and reproduction of a number of vertebrate and invertebrate species. Photoperiodic responses in marine animals may involve a circadian clock that can determine the temporal difference between long and short days (i.e., photoperiodic time measurement), for example, measuring the reciprocal temporal
distance at dusk and dawn behavioural peaks (reviewed by Aguzzi & Company 2010).

1067 Deep continental margin species from the Western Mediterranean Sea appear to 1068 carry out bathymetric displacements according to seasonal variations in the length of the 1069 photoperiod, though these presently remain non-quantified (Aguzzi et al. 2009a). 1070 Seasonal behaviour modulation in species performing nektobenthic displacements may 1071 be the result of changes in interspecific prey-predator relationships in varying light 1072 conditions (Kornfeld-Schor & Dayan 2003). In this context, networks of 1073 bathymetrically linked video observation nodes may contribute to the characterisation of 1074 these poorly understood phenomena.

1075 The bathymetric ranges of the distributions of animal species may also change 1076 on a seasonal basis according to synchronisation of moulting and reproduction cycles 1077 (Aguzzi & Company 2010). Modulation of the biological rhythms in benthic species 1078 may follow a photoperiodic-like pattern, even at depth ranges where solar light is 1079 completely absent and where the only available light source is bioluminescence. 1080 However, in the dark deep sea, although most species, particularly at abyssal depths 1081 seem to exhibit continuous or semi-continuous reproductive patterns (reviewed in 1082 Young 2003); studies performed in the last decade on deep continental margin areas 1083 have demonstrated the occurrence of seasonal reproductive cycles at aphotic depths 1084 (Company & Sardà 1998 2000, Company et al. 2003). This finding suggests that there is 1085 photoperiodic regulation of the physiology of animals in these deep strata (Aguzzi et al. 1086 2007).

1087 The nature of the environmental clues employed by deep-water continental-1088 margin species to time their seasonal rhythms to the photoperiod are presently 1089 unknown. Seasonal phytodetritus concentration peaks within the upper water column 1090 may act as a temporal feeding synchroniser for benthos reproduction (Ecklebarger & Walting 1995, Lisin et al. 1997). This synchronisation may allow these species to obtain a suitable food supply in their larval phase, thus increasing the survival rate of the dispersing offspring (Gage & Tyler 1992, Tyler et al. 1994). Again, continuous video monitoring of benthic deep-sea communities coupled with concurrent multiparametric collection of habitat data may prove critical for understanding the cause-effect relationships between species' behavioural rhythms and habitat fluctuations.

1097

1098 Faunal changes in a globally warming world

1099 The current and predicted impacts of global climate change on living organisms are a 1100 cause of major concern (Walther et al. 2002). The importance of monitoring such 1101 transformations in marine communities is absolutely clear within the marine science 1102 community (Occhipinti-Ambrogi 2007, Lepetz et al. 2009, Jennings & Brander 2010). 1103 Climate-driven variation affects aquatic organisms, which show complex behavioural 1104 responses in displacement and habitat utilisation correlated with temperature and 1105 salinity changes in continental margin areas (Hawkins et al. 2008). In coastal areas, 1106 poleward shifts of species distributions (Perry et al. 2005, Lenoir et al. 2011) and the increasing success of thermophilic taxa versus the retreat of cold water species (Bianchi 1107 1108 & Morri 2003, Azzurro 2008) can be considered among the most obvious consequences 1109 of this phenomenon in marine communities. Other detectable effects are related to shifts 1110 in phenology, i.e., in the timing of life-history events that occur at specific times each 1111 year (Edwards & Richardson 2004.). To follow and understand these phenomena, 1112 generalising analyses from a local to a global scale should be conducted. These analyses 1113 will obviously require sampling to be greatly extended in both time and space, which 1114 usually implies that an extensive socio-economical effort will be required to sustain the 1115 required technology (Giske et al. 2001).

1116 New approaches, new technologies and, most importantly, new ways to collect 1117 and share data are thus critical elements in exploring patterns that act at the largest 1118 geographical scale (Decker & O'Dor 2003). In this context, the role of modern ocean 1119 observation methods, such as marine cabled seafloor observatories, can be considered as 1120 promising and cost-effective, not only in deep water but also in coastal areas (e.g., 1121 OBSEA; see Table 2). These devices, which collect samples remotely, can handle large 1122 volumes of data; however, they have not yet been employed to carry out long-term 1123 studies on faunal changes in association with modifications of selected habitat 1124 characteristics, such as temperature or light intensity/turbidity conditions in relation to 1125 eutrophication. Nevertheless, researchers around the world are working diligently 1126 towards finding new solutions to monitor marine communities over large space and 1127 time scales (Decker & O'Dor 2003, Lepetz et al. 2009). Cabled marine observatories are 1128 perfectly suited for this purpose, and research could immediately benefit from these 1129 powerful innovations, building the capacity for world-wide marine research. These 1130 remote video recorders can support continuous, efficient semi-quantitative sampling in a 1131 variety of different environmental conditions, from near shore to the deep sea. By using 1132 these devices, it will become possible for researchers to plan long-term studies for the 1133 real-time identification of changes in the geographic distribution of target species based 1134 on to an early detection system, as intended within the frame of the Ocean Tracking 1135 Network (OTN 2011).

One of the main advantages of the suggested approach is the potential to overcome the variability generated by many other climate-independent forcing factors that can influence species distributions. The coordination of as many observatories as possible throughout the seas of the world would allow the realisation of complex experimental designs suited to achieve a general picture of global responses to climatic changes on a continuous basis. This coordination would imply the need for an effort related to the development of standardised methodologies and coordination with a 'network' of different observations representing the sole feasible way of disentangling global signals from other sources of variability that act at regional and local scales (Briand 2008, Lepetz et al. 2009).

1146 To limit and to better allocate effort, the acquisition of remotely video-collected 1147 data could be restricted to a small set of indicator species (i.e., key species) (Briand 1148 2008), and these indicators should be chosen from sensitive organisms that are easy to 1149 track and identify. Among other potential biotic indicators, fishes exhibit excellent 1150 qualities for both ecosystem and climate monitoring, such as high dispersal potential, 1151 ecological differentiation, general non-resilience, and sensitivity to temperature 1152 (Stephens et al. 1988, Roessig et al. 2004). Moreover, their large size, ease of 1153 identification and the simplicity of their automated classification based on 1154 morphometric tools (Aguzzi et al. 2009c, 2011b) make them ideal candidates for 1155 sensing studies using remote devices, such as cabled observatories.

1156

1157 The connectivity of benthic populations as a result of rhythmic behaviour in cyclically1158 changing seascapes

In the marine environment, the rhythmic displacement of species occurs within seabed 1159 1160 or water column corridors associated with recurrent oceanographic features (Roff & 1161 Evans 2002). Roff & Taylor (2000) characterised different strata of the water column 1162 and seabed as patches using parameters such as water temperature, depth/light, 1163 stratification/mixing regimes, substratum type, and exposure/slope. They termed the 1164 marine components identified by this classification as 'seascapes'. Marine seascapes are 1165 different from landscapes, as they are more dynamic, intermittent, and present a more 1166 complex geography (Longhurst 2007). The activities of animals occur within certain 1167 seascapes associated with entraining geophysical cycles (i.e., being constituted by a set of particular and recurrent habitat conditions fitting with species ecological niche requirements). Seascapes hence rhythmically modulate the rhythmic behaviour of animals, and seascape characteristics can be measured in a multiparametric fashion by seafloor cabled observatories.

Farina & Belgrano (2004) defined as the 'ecofield' the portion of the surrounding environmental space directly perceived by animals during their behavioural activities of energy collection, preservation, and transformation. Internal tidal currents, light intensity cycles, and variations in the length of photoperiods, represent signals that cyclically affect the ecofield of animals (reviewed by Aguzzi & Company 2010), modifying the predisposition of their populations to undergo rhythmic displacements within seascapes as corridors.

1179 Rhythmic behavioural activation may contribute to the potential dispersion of 1180 benthos among geographically separated groups. In this sense, behavioural rhythms 1181 may be used to explain dispersals by adults (and not only of larvae) among distant 1182 populations. Characterisation of how animals displace in their associated medium when 1183 reacting to temporal changes in its condition is of importance for characterising these potential corridors (reviewed in Kritzer & Sale 2006). Cabled platforms allow the 1184 1185 multiparametric measurement of marine processes through time by concurrently 1186 acquiring several different classes of oceanographic, geologic, and chemical data. Video 1187 observations can be efficiently used to portray the presence of animals within a multi-1188 dimensionally measured habitat framework, which is also represented by 1189 multiparametric assets at a corresponding time. This multiparametric monitoring in 1190 association with bio-video observations could provide a coherent operative and 1191 experimentally testable view of species ecological niches: the presence of animals can 1192 be measured at any moment in relation to (and possibly as a result of) habitat condition and the multidimensional characterisation of this habitat condition depends on theavailable sensors.

1195 Under this scenario, multiparametric video observatories within a network of 1196 distant nodes could provide important data on why certain deep-sea species exhibit 1197 certain geographic ranges of distribution (Levin et al. 2001), clarifying their level of 1198 connectivity based on the enhancement or suppression of the activities of adults at the 1199 onset of certain habitat conditions (i.e., seascape turnover in a certain observational 1200 area). For example, tidal pull may generate deep-sea currents of consistent speed. 1201 Benthic fishes may suppress their swimming activity upon water flow speed increases 1202 to maintain a certain location, or alternatively, they may exhibit increased activity 1203 leading to their entering the dispersing flow (Aguzzi et al. 2010b).

Cabled multiparametric video observatories endowed with moored arrays may be used to portray the behavioural reaction of benthic animals with different levels of motility (e.g., crawlers, walkers or swimmers) to internal tidal-driven flow variations. This research methodology may provide important data on behaviour and dispersion ranges in relation to stepping stone models that has been proposed for dispersal through favourable habitats such as the whale falls (reviewed by Smith & Baco 2003).

1210

1211 Community-based responses to seismic events: towards a new biologically based alert
1212 system

Grant & Halliday (2010) reviewed a number of reports on the putative anticipated perception of seismic events by animal communities. Some species may disappear from a certain area a few hours prior to these events, without resuming their presence and behavioural activities for several days. In this context, the study of responses of single species and communities to earthquakes could be used to predict these events if automated methods of animal monitoring could be developed with sufficient precision.

1219 Cabled multiparametric seafloor stations could be used to implement early 1220 warning systems based on sudden drops in video-counted individuals for a certain 1221 abundant species. With the increased automation efficiency of video image analysis, the 1222 number of individuals of key species in certain geologically active areas could be used 1223 to predict an incoming seismic event based on anticipated population disappearances 1224 from the FOV of cameras. In the case of Japan, this technological possibility could be explored using existing functioning platforms, such as that deployed at a depth of 1100 1225 1226 m in Sagami Bay (Central Japan; see Table 2). Long-lasting automated visual counting 1227 of zoarcid eelpout fishes has already been successfully carried out using this platform 1228 (Aguzzi et al. 2010b). A decrease in the population of zoarcid fishes occurred in 1996 1229 after a major seismic event in the area (Figure 10). Unfortunately the frame sampling 1230 rate was too low (one picture each 12-h) to demonstrate whether this decrease truly 1231 anticipated the event. Drops in individual counts prior to a seismic event could be 1232 resolved using a higher rate of frame acquisition (e.g., a frame per 10 min). The 1233 disappearance of animals below a certain threshold level (defined by an average 1234 computed from observational counts of previous days) may be indicative of the anticipatory perception and reaction of zoarcids related to the incoming seismic event. 1235

1236 If this dynamic could be proven, then it could be incorporated into an automated 1237 alert system. However, the Sagami population of zoarcids show rhythmic changes in 1238 local counts correlated with internal tidal motions (Aguzzi et al. 2010b), which greatly 1239 complicates such an operation. Other non-swimming and more benthic species could be 1240 then studied in relation to this application because their behaviour may be less 1241 dependent upon tidal flow changes.

1242

1243 Time series responses to stochastic pulses of nutrients in deep-sea ecosystems

1244 A potent method for understanding the dynamics of complex systems in engineering 1245 (e.g., electronics, mechanical structures, and acoustics) and economics is to test the 1246 impulse response of the system. A single impulse typically results in a deflection and a 1247 series of oscillations of decreasing amplitude, depending on the damping and resonant 1248 frequency of the system (e.g., ringing of a bell struck by a hammer). Similar principles 1249 can be employed to study the reaction of deep-sea ecosystems to stochastic pulses of 1250 nutrients. Deep-sea areas host food-depleted environments (reviewed by Childress 1251 1995). The enrichment of the deep seafloor with an artificial pulse of particulate organic 1252 matter can be studied by video monitoring of the response of fauna and micro-bacteria 1253 (Witte et al. 2003a,b, Gillibrand et al. 2006). Deployment of cetacean carcasses mimics 1254 natural carrion falls exploited by a succession of organisms (Smith & Baco 2003). 1255 Kemp et al. (2008) used DOBO (see Section on Automated video-imaging from cabled 1256 observatories for the study of behavioural rhythms) to perform time-series observations 1257 of porpoise (*Phocoena phocoena*) carcasses for up to 6 months at bathyal depths and 1258 recorded rhythmic alternations between the numbers of squat lobsters (Munidopsis 1259 crassa) and octopus arrivals (Benthoctopus spp.). Glover et al. (2010b) used real-time 1260 video from a cabled observatory to record the time series of events at whale bones 1261 deployed on the seafloor and showed that such systems have great potential for 1262 elucidating the response characteristics of marine ecosystems.

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Conclusions

1265 The use of cabled observatories for the assessment of population/stock size, 1266 biodiversity, species dynamics, ecosystem functioning, and the responses of 1267 communities to deterministic or stochastic changes in environmental conditions requires 1268 an international effort regarding methods standardisation for comparative studies. 1269 Future actions should promote a coordinated strategy for the use of permanent

1270 observatories, and standardised imaging census protocols should be adopted to permit 1271 large-scale analyses. These simple actions will supply a 'missing tool' in the exploration 1272 of global environmental changes and allow targeting which are the greatest challenges 1273 at different spatial and temporal scales. The methods here proposed are straightforward, 1274 and the development of automation in video and habitat data recording, processing and 1275 analysis could certainly capitalise on existing cabled stations. To realise these goals, the 1276 cooperation of a global network of interested scientists, companies and governments is 1277 certainly desirable.

1278 The study of behavioural rhythms of marine animals by means of cabled 1279 permanent video observatories at all depths of continental margin areas could be 1280 successfully performed by adherence to of prerequisites related to image acquisition, 1281 processing, handling and interpretation of the results. Suggestions for the efficient and 1282 reliable implementation of video studies of rhythmic behaviour using permanent 1283 observatories are listed below:

1284 1. Using video-imaging at permanent cabled video stations to portray local 1285 population sizes and biodiversity should be addressed with caution. Spatial limitations 1286 of these observational points may impede species detection in a given and larger area. It 1287 should be also noted that animal tracking without implementation of trajectory 1288 computing does not allow previously counted individuals to be discarded in consecutive 1289 frames. Video observations can only be used to disclose the biasing effect of 1290 behavioural rhythms on population/stock and biodiversity assessments obtained from trawling when the timing of sampling is not considered in relation to local day-night 1291 1292 (depth-dependent) or internal tidal and inertial (geographic-dependent) entraining 1293 cycles. Space biases could be reduced, hopefully in the near future, by expanding video 1294 acquisition locations as a part of observatory networks.

1295 2. Organism classification procedures must be implemented to facilitate the 1296 extraction of biological information from video sources following the combined use of 1297 morphometric tools and multivariate statistics (e.g., KNN and PLSDA). Supervised 1298 classification of organisms within frame subsamples including the creation of training 1299 sets (libraries of images), increases classification efficiency (i.e., a machine-learning 1300 approach). That process must be differentially tuned for each deployed cabled 1301 observatory, being the local environmental context, video and light equipment and 1302 residing species often different.

1303 3. Animal counts obtained from the automated processing of videos must be 1304 placed within the larger context of synchronously collected habitat data (i.e., gathered at 1305 the same time frequency). This requires implementation of multiparameter data banks in 1306 which biological, geological, and physical oceanographic data can be processed together 1307 for comparison. For such processing, the following steps already in use in the field of 1308 chronobiology are proposed in order to allow i) identification of the periodicity and 1309 phase of the fluctuations; and ii) establishment of temporal linkage in their phases (i.e., 1310 maxima timings), thus studying potential cause-effect relationships between 1311 geophysical fluctuations and population behavioural responses.

4. Among the diverse sensors available, light measuring devices are surprisingly still missing from platforms located within benthic areas of the twilight zone range. The diurnality and nocturnality of species represent fundamental characteristics of their ecological niche. Diel changes in light act cyclically at a very basal level of central nervous system functioning in animals (i.e., on biological clocks), hence potentially setting the timing of consequent behavioural activities.

1318 5. The implementation of automated video image analysis for species tracking
1319 and classification, as well as the development of efficient and user-friendly
1320 multiparametric information treatment tools should result in the creation of a new

generation of embedded technology coupling software for image processing andstatistical data treatment.

1323 6. Multiparametric cabled observatories can overcome present limitations in 1324 gathering biological (i.e., video-imaging), geological, chemical, and finally physical 1325 oceanographic information, which in turn favour the understanding of ecosystem 1326 functioning. In this context, there is for the first time the opportunity to place any biological observation (i.e., video-counted animals) acquired at a high constant 1327 1328 frequency, within a multidimensional space of habitat parameters also measured at a 1329 corresponding moments. A species niche is a hyperspace made by vectors representing 1330 the variation of habitat variables constraining its survival (i.e., the Hutchinson 1331 definition; Whittaker et al. 1973). Multiparametric cabled observatories within a 1332 network measuring biologic, geological, chemical, and oceanographic parameters 1333 represent the first operative tool to study species niches over space and time.

1334 7. Multiparametric monitoring represents the necessary basis for determination 1335 of presence or absence of a given species and, hence definition of population rhythmic 1336 behaviour within certain configurations of habitat variables. Although the behaviour of 1337 species and communities can also be depicted over time using acoustic technologies, 1338 only video-imaging can ultimately discriminate individuals of given species without the 1339 need for verification by other means e.g., pelagic/benthic trawling). Therefore, video 1340 cameras can be placed at the highest rank of a multisensor scale of complexity in which 1341 oceanographic (physical) and chemical (geologically- or life-derived) variables are at 1342 the base (Figure 11). This ranking of sensors is related to the character of the obtained 1343 information, with individual, population or species detection being the result of 1344 surrounding physical, chemical, and geologic conditioning (i.e., the ecological niche 1345 measured in a multiparametric fashion).

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References

- Adlerstein, S. & Ehrich, S. 2003. Patterns in diel variation of cod catches in North Sea
 bottom trawl surveys. *Fisheries Research* 63, 169-178.
- Aglen, A., Engas, A., Huse, I., Michalsen, K. & Stensholt, B.K. 1999. How vertical fish
 distribution may affect survey results. *ICES Journal of Marine Science* 56, 345360.

- Ageron, M., Aguilar, J.A., Albert, A., Ameli, F., Anghinolfi, M., Anton, G., Anvar, S.,
 Ardid, M., Aubert, J.J., Aublin, J., et al. 2009. Performance of the first
 ANTARES detector line. *Astroparticle Physics* **31**, 277-283.
- 1376 Aguilar, J.A., Samarai, A.L., Albert, A., Anghinolfi, M., Anton, G., Anvar, S., Ardid,
- 1377 M., Assis Jesus, A.C., Astraatmadja, T., Aubert, J.-J., et al. 2010. Measurement 1378 of the atmospheric muon flux with a 4 GeV threshold in the ANTARES neutrino 1379 telescope. *Astroparticle Physics* **33**, 86-90.
- Aguzzi, J., Abelló, P. & Depledge, M. 2004a. Endogenous cardiac activity rhythm of
 the Northwestern Mediterranean *Nephrops norvegicus* (L.) inhabiting the upper
 slope. *Marine Freshwater Behaviour Physiology* 37, 55-64.
- Aguzzi, J. & Bahamon, N. 2009. Modeled day-night biases in decapod assessment by
 bottom trawling survey. *Fisheries Research* 100, 274-280.
- Aguzzi, J., Bahamon, N. & Marotta, L. 2009a. Modulation of activity rhythms in
 continental margin decapods by light availability and predatory behaviour of *Nephrops norvegicus. Marine Ecology* **30**, 366-375.
- 1388 Aguzzi, J., Bullock, N.M. & Tosini, G. 2006a. Spontaneous internal desynchronization
- 1389 of locomotor activity and body temperature rhythms from plasma melatonin 1390 rhythm in rats exposed to constant dim light. *Journal of Circadian Rhythms* **4**, 6.
- Aguzzi, J. & Chiesa, J.J. 2005. The cardiac activity of *Nephrops norvegicus* (Decapoda:
 Nephropidae): the relationship between ultradian and circadian rhythms. *Journal of Crustacean Biology* 25, 577-584.
- Aguzzi, J., Chiesa, J., Abelló, P. & Díez-Noguera, A. 2005. Temporal modifications in
 cardiac rhythmicity of *Nephrops norvegicus* (Crustacea: Decapoda) in relation to
 trawl capture stress. *Scientia Marina* 69, 369-374.
- 1397 Aguzzi, J., Chiesa, J.J., Caprioli, R., Cascione, D., Magnifico, G., Rimatori, V. & Costa,
- 1398 C. 2006b. Preliminary evidences of circadian fan activity rhythm in Sabella

- *spallanzanii* (Gmelin, 1791) (Polychaeta: Sabellidae). *Scientia Marina* 70, 727724.
- Aguzzi, J. & Company, J.B. 2010. Chronobiology of deep water continental margin
 decapods. *Advances in Marine Biology, an Annual Review* 58, 155-225.
- 1403 Aguzzi, J., Company, J.B., Abelló, P. & García, J.A. 2006c. Ontogenetic changes in the
- 1404 vertical migratory rhythms of benthopelagic shrimps *Pasiphaea multidentata* and *P*.
- 1405 *sivado* (Crustacea: Caridea). *Marine Ecology Progress Series* **335**, 167-174.
- 1406 Aguzzi, J., Company, J.B., Costa, C., Menesatti, P., Bahamon, N. & Sardà, F. 2010a.
- 1407 Activity rhythms in the deep-sea crustacean: Chronobiological challenges and 1408 potential technological scenarios. *Frontiers in Bioscience* **16**, 131-150.
- 1409 Aguzzi, J., Company, J.B. & Sardà F. 2004b. The feeding activity rhythm of Nephrops
- *norvegicus* (L.) of the Western Mediterranean shelf and slope grounds. *Marine Biology* 144, 463-472.
- Aguzzi, J., Company, J.B., Sardà, F. & Abelló, P. 2003a. Circadian oxygen
 consumption patterns in continental slope *Nephrops norvegicus* (Decapoda:
 Nephropidae) in the western Mediterranean. *Journal of Crustacean Biology* 23, 749757.
- Aguzzi, J., Company, J.B., Sardà, F., Sánchez, J., García, J.A. & Rotllant, G. 2004c. Is
 the glucose concentration in the hemolymph a suitable indicator of circadian
 rhythmicity in *Nephrops norvegicus* (Decapoda: Nephropidae)? *Crustaceana* 77,
 213-229.
- 1420 Aguzzi, J., Costa, C., Antonucci, F., Company, J.B., Menesatti, P. & Sardà, F. 2009b.
- 1421 Influence of rhythmic behaviour in the morphology of decapod Natantia. *Biological*
- 1422 *Journal of the Linnaean Society* **96**, 517-532.

- 1423 Aguzzi, J., Costa, C., Furushima, Y., Chiesa, J.J., Company, J.B., Menesatti, P., Iwase
- R. & Fujiwara, Y. 2010b. Behavioural rhythms of hydrocarbon seep fauna in
 relation to internal tides. *Marine Ecology Progress Series* 418, 47-56.
- 1426 Aguzzi, J., Costa, C., Matabos, M., Antonucci, F., Juniper, K. & Menesatti, P. 2011a.
- 1427 Automated video-imaging applications for the detection of benthic crustaceans and
- 1428 bacterial mat coverage at VENUS cabled observatory. *Sensors* SUBMITTED.
- 1429 Aguzzi, J., Costa, C., Menesatti, P., Fujwara, Y., Iwase, R. & Ramirez-Llorda, E.
- 1430 2009c. A novel morphometry-based protocol of automated video-image analysis for
 1431 species recognition and activity rhythms monitoring in deep-sea fauna. *Sensors* 9,
 1432 8438-8455.
- Aguzzi, J., Costa, C., Menesatti, P., García, J.A. & Sardà, F. 2009d. Monochromatic
 blue light entrains diel activity cycles in the Norway lobster, *Nephrops norvegicus*(L.) as measured by automated video-image analysis. *Scientia Marina* 73, 773-783.
- Aguzzi, J., Cuesta, J.A., Librero, M. & Toja, J. 2005. Daily and seasonal feeding
 rhythmicity of *Palaemonetes varians* (Leach, 1814) from Southwestern Europe. *Marine Biology* 148, 141-147.
- 1439 Aguzzi, J., Manuèl, A., Condal, F., Nogueras, M., del Rio, J., Costa, C., Menesatti, P.,
- 1440 Sardà, F., Toma, D., Puig, P., Guillen, J. & Palanques, A. 2011b. The new Seafloor
- 1441 Observatory (OBSEA) for remote and long-term coastal ecosystem monitoring.
 1442 Sensors 11, 5850-5872.
- Aguzzi, J., Puig, P. & Company, J.B. 2009e. Hydrodynamic, non-photic modulation of
 biorhythms in the Norway lobster, *Nephrops norvegicus* (L.). *Deep-Sea Research I*
- 1445 **56**, 366-373.
- 1446 Aguzzi, J., Ramirez-Llodra, E., Telesnicki, G. & Camps, M. 2007. Day-night activity
- 1447 rhythm of the cold seep shrimp Alvinocaris stactophila (Caridea:

- Alvinocarididae) from the Gulf of Mexico. *Journal of the Marine Biological Association of the United Kingdom* 87, 1175-1180.
- Aguzzi, J. & Sardà, F. 2008. A history of recent advancements on *Nephrops norvegicus*behavioural and physiological rhythms. *Reviews in Fish Biology and Fisheries*1452 18, 235-248.
- Aguzzi, J., Sardà, F., Abelló P., Company, J.B. & Rotllant, G. 2003b. Diel patterns of *Nephrops norvegicus* catchability (Decapoda: Nephropidae) in the western
 Mediterranean: a perspective through depth and season. *Marine Ecology Progress Series* 258, 201-211.
- 1457 Aguzzi, J., Sbragaglia, V., Sarriá, D., García, J.A., Costa, C., Del Río, J., Mànuel, A.,
- Menesatti, P. & Sardà, F. 2011c. A new RFID technology for the laboratory
 monitoring of behaviour in marine organisms. *Sensors* 11, 9532-9548.
- Aguzzi, J., Sarriá, D., García, J.A., Del Rio, J., Sardà, F. & Manuel, A. 2008. A new
 tracking system for the measurement of diel locomotor rhythms in the Norway
 lobster, *Nephrops norvegicus* (L.). *Journal of Neuroscience Methods* 173, 215-224.
- 1463 Aksens, D.L. & Giske, J. 1993. A theoretical model for aquatic visual feeding.
 1464 *Ecological Modelling* 67, 233-250.
- Allen, P.Z., Thorndike, E.M., Sullivan, B.C.H. & Gerard, R.D. 1978. Observations of
 the deep-sea floor from 202 days of time-lapse photography. *Nature* 272, 812-814.
- Amit, D. & Boaz, L. 2005. Support vector machine-based image classification for
 genetic syndrome diagnosis. *Pattern Recognition Letters* 26, 1029-1038.
- 1469 Amram, P., Anvar, S., Aslanides, E., Aubert, J.J., Azoulay, R., Basa, S., Benhammou,
- 1470 Y., Bernard, F., Bertin, V., Billault, M., et al. 2000. Background light in potential
- 1471 sites for the ANTARES undersea neutrino telescope. *Astroparticle Physics* 13, 127-
- 1472 136.

- 1473 Anglès-Pujolràs, M., Chiesa, J.J., Díez-Noguera, A. & Cambras, T. 2006. Motor activity
- 1474 rhythms of forced desynchronized rats subjected to restricted feeding. *Physiological*1475 *Behaviour* 88, 30-38.
- 1476 Antonucci, F., Costa, C., Aguzzi, J. & Cataudella, S. 2009. Ecomorphology of morpho-
- 1477 functional relationships in the family of Sparidae: A quantitative statistic approach.
- 1478 *Journal of Morphology* **270**, 843-855.
- Aschoff, J. 1981. Free-running and entrained circadian rhythms. In *Handbook of Behavioral Neurobiology Vol. 4*, J. Aschoff. (ed). New York: Plenum Press, 81-93.
- Atkinson, D.B. 1989. Diel movements of beaked redfish and implications of these for
 stratified random bottom trawl estimates of biomass and abundance. *North American Journal of Fisheries Management* 9, 163-170.
- 1484 Azzurro, E. 2008. The advance of thermophilic fishes in the Mediterranean Sea:
- 1485 overview and methodological questions. In *Climate Warming and Related Changes* 1486 *in Mediterranean Marine Biota*, F. Briand (ed.). Monaco: CIESM Workshop
- 1487 Monographs **35**, 39-46.
- 1488 Azzurro, E., Aguzzi, J., Maynou, F., Chiesa, J.J. & Savini D. 2011. Diel rhythms in
- shallow Mediterranean rocky-reef fishes: a chronobiological approach with thehelp of trained volunteers. *Marine and Freshwater Research* ACCEPTED.
- Bahamon, N., Aguzzi, J., Bernardello, R., Ahumada, M.-A., Puigdefabregas, J., Muñoz,
 E., Velásquez, Z., Cruzado, A. 2011. The new real-time atmospheric and pelagic
 Operational Observatory of the Catalan Sea (OOCS) in the submarine Blanes
 canyon (NW Mediterranean Sea). *Sensors* SUBMITTED.
- Bahamon, N., Sardà, F. & Aguzzi, J. 2009. Fuzzy diel pattern in commercial
 catchability of deep-water continental margin species. *ICES Journal of Marine Science* 66, 2211-2218.

- Bailey, D.M., Collins, M.A., Gordon, J.D.M., Zuur, A.F. & Priede, I.G. 2009. Longterm changes in deep-water fish populations in the North East Atlantic: deeperreaching effect of fisheries? *Proceedings of the Royal Society of London B* 275
 1965-1969.
- Bailey, R.F. & Elner, R.W. 1989. Northwester Atlantic Snow crab fisheries: lessons in
 research and management. In *Marine Invertebrate Fisheries. Their Assessment and Management*, J.F. Caddy (ed.). New York: John Wiley and Sons, 261-280.
- Bagley, P.M., Smith, K.L., Bett, B., Priede, I.G., Rowe, G., Clarke, J., Walls, A., Ruhl,
 H.A. & Bailey, D.M. 2011. The DELOS project: Long-term observatories in an
 oil field environment in the Tropical Atlantic Ocean. In *Seafloor Observatories:*
- A New Version of the Earth from the Abyss, P. Favali et al. (eds). Italy: ElsevierPraxis.
- Bailey, D.M., Ruhl, H.A. & Smith, K.L. 2006. Long-term change in benthopelagic fish
 abundance in the abyssal North-East Pacific Ocean. *Ecology* 87, 549-555.
- 1512 Bakkala, R.G. 1993. Structure and Historic Changes in the Ground Fish Complex of the
- 1513 *Eastern Bearing Sea.* Seattle: NOAA/National Marine Fisheries Service (NMFS)
 1514 Technical Reports 114.
- Barange, M., Pillar, S.C., Huse, I. & Hutchings, L. 2005. Vertical migration,
 catchability and acoustic assessment of semi-pelagic Cape horse mackerel *Trachurus trachurus capensis* in the southern Benguel. *African Journal of Marine Science* 27, 459-469.
- Barans, C.A., Ardent, M.D., Moore, T. & Schmidt, D. 2005. Remote video revisited: A
 visual technique for conducting long-term monitoring of reef fishes on the
 continental shelf. *Marine Technology Society Journal* **39**, 80-88.
- Barnes, C.R., Best, M.M.R., Johnson, F., Phibbs, P. & Pirenne, B. 2008. Latest
 developments in building the world's first regional cabled observatory:

- 1524 NEPTUNE Canada. Building the world's first regional cabled observatory:
 1525 NEPTUNE Canada. *Journal of Ocean Technology* 3, 1-3.
- Barry, D. & Stanienda, T. 1998. Solving the Java object storage problem. *IEEE*-*Computer* 31, 33-40.
- 1528 Benn, A.R., Weaver, P.P., Billet, D.S.M., van den Hove, S., Murdock, A.P., Doneghan,
- G.B. & Le Bas, T. 2010. Human activities on the deep seafloor in the North East
 Atlantic: an assessment of spatial extent. *PLoS ONE* 5, e12730.
- Benoit-Bird, K.J. & Au, W.W. 2006. Extreme diel horizontal migrations by a tropical
 nearshore resident micronekton community. *Marine Ecology Progress Series* 319,
 1-14.
- Benoit-Bird, K.J., Zirbel, M.J. & McManus, M.A. 2008. Diel variation of zooplankton
 distributions in Hawaiian waters favours horizontal diel migration by midwater
 micronekton. *Marine Ecology Progress Series* 367, 109-123.
- Benoit, H.P. & Swain, D.P. 2003. Accounting for length- and depth-dependent diel
 variation in catchability of fish and invertebrates in an annual bottom-trawl
 survey. *ICES Journal of Marine Science* 60, 1298-1317.
- Beranzoli, L., Favali, P. & Smriglio, G. (eds) 2002. Science-Technology Synergy for *Research in the Marine Environment: Challenges for the XXI Century.*Amsterdam: Elsevier.
- Bertolucci, C., Colognesi, I., Caruso, P., Aguzzi, J., Chen, Z., Foà, A., Tosini, G.,
 Bernardi, F., McKnight, S.L. & Pinotti, M. 2008. Evidence for an overlapping
 role of CLOCK and NPAS2 transcription factors in liver circadian oscillators.
- 1546 *Molecular and Cellular Biology* **28**, 3070-3075.
- Bertrand, J., Gil de Sola, L., Papacostantinou, P., Relini, G. & Souplet, A. 2002. The
 general specifications of the MEDITS surveys. *Scientia Marina* 66, 9-17.

- Betke, M., Hirsh, D.E., Bagchi, A., Hristov, N.I., Makris, N.C. & Kunz, T.H. 2007.
 Tracking large variable numbers of objects in clutter. In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR'07).*Minneapolis MN: IEEE Press, 1-8.
- 1553 Bharati, M.H., Liu, J.J. & MacGregor, J.F. 2002. Image texture analysis: methods and 1554 comparisons. *Chemometrics and Intelligent Laboratory Systems* **72**, 57-71.
- Bianchi, N. & Morri, C. 2003. Global sea warming and 'tropicalization' of the
 Mediterranean Sea: Biogeographic and ecological aspects. *Biogeographia* 24,
 319-327.
- 1558 Billett, D.S.M., Bett, B.J., Reid, W.D.K., Boorman, B. & Priede, I.G. 2009. Long-term
- 1559 change in the abyssal NE Atlantic: The 'Amperima Event' revisited. *Deep-Sea*1560 *Research II* 57, 1406-1417.
- Blaber, S.J.M., Brewer, D.T. & Harris, A.N. 1994. Distribution, biomass and
 community structure of demersal fishes of the Gulf of Carpentaria, Australia. *Australian Journal of Marine and Freshwater Research* 45, 375-396.
- Borstad, G., Brown, L., Sato, M., Lemon, D., Kerr, R. & Willis, P. 2010. Long
 zooplankton time series with high temporal and spatial resolution. In *Proceedings of IEEE-OCEANS'10.* Seattle WA: IEEE Press, 1-9.
- Boswell, K.M., Wilson, M.P. & Cowan Jr., J.H. 2008. A semiautomated approach to
 estimating fish size, abundance, and behavior from Dual-Frequency
 Identification Sonar (DIDSON) data. *North American Journal of Fisheries Management* 28, 799-807.
- Briand, F. (ed.) 2008. *Climate Warming and Related Changes in Mediterranean Marine Biota.* Monaco: CIESM Workshop Monographs 35.
- 1573 Bruno, I., Costas, G., González, C. & Paz, X. 2000. Feeding chronology of Yellowtail
- 1574 flounder (Limanda ferruginea) and American plaice (Hippoglossoides

- 1575 platessoides) on Grand Bank (NAFO Division 3N). NAFO Scientific Council
 1576 Studies 33, 103-116.
- Bruno, T.J. & Svoronos, P.D.N. 2005. CRC Handbook of Fundamental Spectroscopic
 Correlation Charts. Boca Raton, London, New York: Francis & Taylor CRC
 Press.
- Budd, A.F., Johnson, K.G. & Potts, D.C. 1994. Recognizing morphospecies in colonial
 reef corals: I. Landmark-based methods. *Paleobiology* 20, 484-505.
- Canals, M., Puig, P., Heussner, S., Durrieu de Madron, X., Palanques, A. & Fabres, J.
 2006. Flushing submarine canyons. *Nature* 444, 354-357.
- Cappo, M., Speare, P. & De'ath, G. 2004. Comparison of baited remote underwater
 video stations (BRVUS) and prawn (shrimp) trawls for assessment of fish
 biodiversity in inter-reefal areas of the Great Barrier Reef marine park. *Journal of Experimental Marine Biology and Ecology* 302, 123-152.
- Carpentieri, P., Colloca, F. & Ardizzone, G.D. 2005. Day-night variations in the
 demersal nekton assemblage on the Mediterranean shelf-break. *Estuarine and Coastal Shelf Science* 63, 577-588.
- Carr, A.J., Johnston, J.D., Semikhodskii, A.G., Nolan, T., Cagampang, F.R.J.A.,
 Stirland, J.A. & Loudon, A.S. 2003. Photoperiod differentially regulates circadian
 oscillators in central and peripheral tissues of the Syrian hamster. *Current Biology*1594 13, 1543-1548.
- Cartes, J.E., Sardà, F., Company, J.B. & Llenorat, J. 1993. Day-night migrations by
 deep-sea decapod crustaceans in experimental samplings in the Western
 Mediterranean Sea. *Journal of Experimental Marine Biology and Ecology* 171, 6373.

- 1599 Chiesa, J.J., Aguzzi, J., García, J.A., Sardà, F. & de la Iglesia, H. 2010. Light intensity
- determines temporal niche switching of behavioural activity in deep water *Nephrops norvegicus* (Crustacea: Decapoda). *Journal of Biological Rhythms* 25, 277-287.
- 1602 Chiesa, J.J., Anglès-Pujolràs, M., Díez-Noguera, A. & Cambras, T. 2005. Activity
- 1603 rhythm of golden hamster (*Mesocricetus auratus*) can be entrained to a 19-h light-
- 1604 dark cycle. *American Journal of Physiology* **289**, 998-1005.
- 1605 Childress, J.J. 1995. Do animals present biochemical adaptations to depths? *Trends in*1606 *Ecology and Evolution* 10, 30-36.
- 1607 Chraskova, J., Kaminsky, Y. & Krekule, I. 1999. An automatic 3D tracking system with
 1608 a PC and a single TV camera. *Journal of Neuroscience Methods* 88, 195-200.
- 1609 Cline, D.E., Edgington, D.R. & Mariette, J. 2007. An automated visual event detection
- system for cabled observatory video. In *Proceedings of IEEE-OCEANS'07*.
 Vancouver BC: IEEE Press, 1-5.
- 1612 Coll, M., Piroddi, C., Steenbeek, J., Kaschner, K., Ben Rais Asram, F., Aguzzi, J.,
- 1613 Ballesteros, E., Bianchi, C.N., Corbera, J., Dailianis, T., Danovaro, R., Dimitris, V.,
- 1614 Estrada, M., Froglia, D., Galil, B.S., Gasol, J.M., Gertwagen, R., Gil, J., Gorsky, G.,
- 1615 Guilhaumon, F., Kesner-Reyes, K., Kitsos, M.-S., Koukouras, A., Lampadariou, N.,
- 1616 Laxamana, E., Lloris, D., López-Fé De La Cuadra, C.M., Lotze, H.K., Martin, D.,
- 1617 Moullot, D., Oro, D., Raichevich, S., Rius-Barile, J., Saiz-Salinas, J.I., San Vicente,
- 1618 C., Somot, S., Templado, J., Turon, X., Villanueva, R. & Voultsiadou, E. 2010.
- 1619 Biodiversity of the Mediterranean Sea: status, patterns, and threats. *PLoS I* **5**, e11842.
- 1621 Company, J.B. & Sardà, F. 1998. Metabolic rates and energy content of deep-sea
 1622 benthic decapod crustaceans in the western Mediterranean Sea. *Deep-Sea*1623 *Research I* 45, 1861-1880.

- 1624 Company, J.B. & Sardà, F. 2000. Growth parameters of deep-water decapod
 1625 crustaceans in the Northwester Mediterranean Sea: A comparative approach.
 1626 *Marine Biology* 136, 79-90.
- 1627 Company, J.B., Sardà, F., Puig, P., Cartes, J.E. & Palanques, A. 2003. Duration and
 1628 timing of reproduction in decapod crustaceans of the NW Mediterranean
 1629 continental margin: is there a general pattern? *Marine Ecology Progress Series*1630 261, 201-216.
- 1631 Costa, C., Aguzzi, J., Chiesa, J.J., Magnifico, G., Cascione, D. & Caprioli, R. 2008a.
 1632 Evidences on the transient disruption of *Sabella spallanzanii* (Polychaeta:
 1633 Sabellidae) endogenous fan activity rhythm as result of stress of capture and
 1634 acclimation to laboratory. *Italian Journal of Zoology* **75**, 337-344.
- 1635 Costa, C., Aguzzi, J., Menesatti, P., Antonucci, F., Rimatori, V. & Mattoccia, M. 2008b.
 1636 Shape analysis of different populations of clams in relation to their geographical
 1637 structure. *Journal of Zoology* 276, 71-80.
- 1638 Costa, C., Antonucci, F., Pallottino, F., Aguzzi, J., Sun D.-W. & Menesatti, P. 2011.
- 1639 Shape analysis of agricultural products: a review of recent research advances and
- 1640 potential application to computer vision. *Food and Bioprocess Technology* **4**, 673-
- 1641
 692.
- 1642 Costa, C., Loy, A., Cataudella, S., Davis, D. & Scardi, M. 2006. Extracting fish size
 1643 using dual underwater cameras. *Aquaculture Engineering* 35, 218-227.
- 1644 Costa, C., Menesatti, P., Aguzzi, J., D'Andrea, S., Antonucci, F., Rimatori, V.,
- 1645 Pallottino, F. & Mattoccia, M. 2010. External shape differences between sympatric
- 1646 populations of commercial clams *Tapes decussatus* and *T. philippinarum*. Food and
- 1647 *Bioprocessing Technology* **3**, 43-48.
- 1648 Costa, C., Menesatti, P., Paglia, G., Pallottino, F., Aguzzi, J., Rimatori, V., Russo, G.,
- 1649 Recupero, S. & Reforgiato, R.G. 2009a. Quantitative evaluation of Tarocco sweet

- 1650 orange fruit shape using opto-electronic elliptic Fourier based analysis. *Postharvest*1651 *Biology and Technology* 54, 38-47.
- 1652 Costa, C., Pallottino, F., Angelini, C., Proietti, M., Capoccioni, F., Aguzzi, J.,
 1653 Antonucci, F. & Menesatti, P. 2009b. Colour calibration for quantitative biological
 1654 analysis: A novel automated multivariate approach. *Instrumentation Viewpoint* 8,
 1655 70-71.
- 1656 Costa, C., Scardi, M., Vitalini, V. & Cataudella, S. 2009c. A dual camera system for
 1657 counting and sizing Northern Bluefin Tuna (*Thunnus thynnus*; Linnaeus, 1758)
 1658 stock, during transfer to aquaculture cages, with a semi automatic Artificial Neural
 1659 Network tool. *Aquaculture* 291, 161-167.
- 1660 Cox, M.J. & Borchers, D.L. 2011. Estimating the density of Antarctic krill (*Euphausia superba*) from multi-beam echo-sounder observations using distance sampling
 1662 methods. *Journal of the Royal Statistical Society C (Applied Statistics)* 60, 301-316.
- 1663 Culverhouse, P.F., Williams, R., Reguera, B., Herry, V. & González-Gil, S. 2003. Do
 1664 experts make mistakes? A comparison of human and machine identification of
 1665 dinoflagellates. *Marine Ecology Progress Series* 247, 17-25.
- Daan, S. & Aschoff, J. 2001. The entrainment of circadian systems. In *Handbook of Behavioral Neurobiology: Circadian Clocks*. Takahashi, J.S. & Turek, F. (eds).
 New York: Kluwer Academic/Plenum Publishers, 7-42.
- 1669 Dah-Jye, L., Archibald, J.K., Schoenberger, R.B., Dennis, A.W. & Shiozawa, D.K.
- 1670 2008. Contour matching for fish species recognition and migration monitoring.
 1671 *Studies in Computational Intelligence* 122, 183-207.
- 1672 Danovaro, R., Company, J.B., Corinaldesi, C., D'Onghia, G., Galil, B., Gambi, C.,
- 1673 Gooday, A.J., Lampadariou, N., Luna, G.M., Morigi, C., Olu, K., Polymenakou, P.,
- 1674 Ramirez-Llodra, E., Sabbatini, A., Sardà, F., Sibuet, M. & Tselepides, A. 2010.

- 1675 Deep-Sea biodiversity in the Mediterranean Sea: the known, the unknown, and the
- 1676 unknowable. *PLoS I* **5**, e11832.
- 1677 Decker, C.D. & O'Dor, R. 2003. Census of marine life: unknowable or just unknown?
 1678 *Oceanologica Acta* 25, 179-186.
- 1679 De Robertis, A. 2002. Size-dependent visual predation risk and the timing of vertical
- 1680 migration: An optimized model. *Limnological Oceanography* **47**, 925-933.
- 1681 Dibner, C., Schibler, U. & Albrecht, U. 2010. The mammalian circadian timing system:
- 1682 Organization and coordination of central and peripheral clocks. *Annual Review of*1683 *Physiology* 72, 517-549.
- 1684 Dickey, T.D. & Bidigare, R.R. 2005. Interdisciplinary oceanographic observations: The
 1685 wave of the future. *Scientia Marina* 69, 23-42.
- 1686 DONET 2011. http://www.jamstec.go.jp/jamstec-e/maritec/donet; Last access:
 1687 07/07/2011.
- 1688 Eckelbarger, K.J. & Walting, L. 1995. Role of phylogenetic constrains in determining
 1689 the reproductive patterns in deep-sea invertebrates. *Invertebrate Biology* 114, 2561690 269.
- Edgington, D.R., Cline, D.E., Davis, D., Kerkez, I. & Mariette, J. 2006. Detecting,
 tracking and classifying animals in underwater video. In *Proceedings of the MTS/IEEE-OCEANS'06*. Boston MA: IEEE Press, 1-5.
- 1694 Edgington, D.R., Cline, D.E., Kerkez, I., Mariette, J. 2007. Detecting, tracking, and 1695 classifying animals in underwater observatory video. In *Proceedings of the*
- 1696 Underwater Technology Symposium (UT07)/ Workshop on Scientific Use of
- 1697 *Submarine Cables & Related Technologies (SSC07).* Tokyo, Japan: IEEE Press.
- 1698 Edgington, D.R., Walther, D., Cline, D.E., Sherlock, R.E. & Koch, C. 2004. Detecting
- and tracking animals in underwater video. In Proceedings of the IEEE Computer

- 1700 Society Conference on Computer Vision and Pattern Recognition (CVPR'04).
- 1701 Washington D.C.: IEEE Press, 544-549.
- 1702 Edgington, D.R., Walther, D., Salamy, K.A., Risi, M., Sherlock, R.E. & Koch, C. 2003
- 1703 Automated event detection in underwater video. In Proceedings of the IEEE-
- 1704 *OCEANS'03*. San Diego CA: IEEE Press, 2749-2753.
- Edwards, M. & Richardson, A.J. 2004. Impact of climate change on marine pelagic
 phenology and trophic mismatch. *Nature* 430, 881-883.
- 1707 Embleton, K.V., Gibson, C.E. & Heaney, S.I. 2003. Automated counting of
- phytoplankton by pattern recognition: A comparison with a manual counting
 method. *Journal of Plankton Research* 25, 669-681.
- European Commission for Research Infrastructures (ESFRI) 2011.
 http://cordis.europa.eu/esfri/roadmap.htm; Last Access: 07/07/2011.
- European Multidisciplinary Seafloor Observatory (EMSO) 2011. http://www.emsoeu.org; Last access: 07/06/2011.
- 1714 Farina, A. & Belgrano, A. 2004. The eco-field: A new paradigm for landscape ecology.
- 1715 *Ecological Research* **19**, 107-110.
- Favali, P. & Beranzoli, L. 2006. Seafloor observatory science: A review. *Annals of Geophysics* 49, 515-567.
- Favali, P. & Beranzoli, L. 2009. EMSO: European Multidisciplinary Seafloor
 Observatory. *Nuclear Instruments and Methods in Physics Research A* 602, 21-27.
- 1720 Favali, P., Beranzoli, L., D'Anna, G., Gasparoni, F. & Gerber, H.W. 2006a. NEMO-
- 1721 SN1. The 1st 'Real-Time' Seafloor Observatory of ESONET. *Nuclear Instruments*
- and Methods in Physics Research A 567/2, 462-467.
- 1723 Favali, P., Beranzoli, L., D'Anna, G., Gasparoni, F., Marvaldi, J., Clauss, G., Gerber,
- 1724 H.W., Nicot, M., Marani, M.P., Gamberi, F., Millot, C. & Flueh, E.R. 2006b. A fleet

- of multiparameter observatories for geophysical and environmental monitoring at
 seafloor. *Annals of Geophysics* 49, 659-680.
- 1727 Favali, P., Beranzoli, L., Italiano, F., Migneco, E., Musumeci, M. & Papaleo, R. 2011.
- 1728 NEMO-SN1 observatory developments in view of the European Research
- 1729 Infrastructures EMSO and KM3NeT. Nuclear Instruments and Methods in Physics
- 1730 *Research Section A* **626-627**, 53-56.
- 1731 Favali, P., Person, R., Barnes, C.R., Kaneda Y., Delaney, J.R. & Hsu, S.K. 2010.
- 1732 Seafloor observatory science. In *Proceedings of the Observations and Information*
- 1733 for Society conference (OceanObs'09): Sustained Ocean, J. Hall et al. (eds). Venice,
- 1734 Italy: ESA Publication, 306.
- 1735 Findik, O., Babaoğlua, I. & Ülkera, E. 2010. A colour image watermarking scheme
- based on hybrid classification method: Particle swarm optimization and k-nearest
 neighbour algorithm. *Optics Communications* 283, 4916-4922.
- Fock, H., Uiblein, F., Koster, F. & von Westernhage, H. 2002. Diel and habitatdependent resource utilisation by deep-sea fishes at the Great Meteor seamount:
- 1740 Niche overlap and support for the sound scattering layer interception hypothesis.
- $1741 \qquad M \stackrel{!}{} E = 1 \qquad D \qquad C \stackrel{!}{} 244 \ 210 \ 222$
- 1741 *Marine Ecology Progress Series* **244**, 219-233.
- 1742Foxton, F. 1970a. The vertical distribution of pelagic decapods (Crustacea: Natantia)1743collected on the second cruise 1965. I. The Caridea. Journal of the Marine1744Example 1965. I. The Caridea. Journal of the Marine
- 1744 *Biological Association of the United Kingdom* **50**, 939-960.
- 1745 Foxton, F. 1970b. The vertical distribution of pelagic decapods (Crustacea: Natantia)
- 1746 collected on the second cruise 1965. II. The Penaeidea and general discussion.
- 1747 *Journal of the Marine Biological Association of the United Kingdom* **50**, 961-1000.
- 1748 Foxton, F. & Roe, H.S.J. 1974. Observations on the nocturnal feeding of some
- 1749 mesopelagic decapod crustacea. *Marine Biology* **28**, 37-49.

- Fraenkel, P.L. 1986. *Water lifting devices*. Rome: Food & Agriculture Organization of
 the United Nations (FAO), Rome: *FAO Irrigation and Drainage Paper* 43.
- 1752 Frank, T.M. & Widder, E.A. 2002. Effects of a decrease in downwelling irradiance on
- 1753 the daytime vertical distribution patterns of zooplankton and micronekton. *Marine*
- 1754 *Biology* **140**, 1181-1193.
- Franqueville, C. 1971. Macroplancton profond (Invertébrés) de la Méditerranée nord
 occidentale. *Téthys* 3, 11-56.
- 1757 Frugoni, F., Beranzoli, L., Favali, P. & Basili, A. (eds) 2006. Special issue dedicated to
- Giuseppe Smriglio. From land networks to seafloor observatories. Annals of *Geophysics* 49, 515-879.
- 1760 Fukuhara, C., Aguzzi, J., Bullock, N.M. & Tosini, G. 2005. Effect of long term-
- exposure to constant dim light on the circadian system of rats. *Neurosignals* 14, 117-125.
- 1763 Gage, J.D. & Tyler, P.A. 1992. Deep-sea Biology. A Natural History of Organisms at
- 1764 *the Deep-sea Floor*. Cambridge: Cambridge University Press.
- 1765 Garrett, C. 2003. Internal tides and ocean mixing. *Science* **301**, 1858-1859.
- Garret, C. & Kunze, E. 2007. Internal tide generation in the deep ocean. *Annual Review of Fluid Mechanic* **39**, 57-87.
- 1768 Gaten, E., Tarling, G., Dowse, H., Kyriacou, C. & Rosato, E. 2008. Is vertical migration
- 1769 in Antartic krill (*Euphausia superba*) influenced by an underlying circadian rhythm?
- 1770 *Journal of Genetic* **87**, 473-483.
- 1771 Gillibrand, E.J.V., Bagley, P.M., Jamieson, A., Herring, P.J., Partridge, J.C., Collins,
- 1772 M.A., Milne, R. & Priede, I.G. 2006. Deep-sea benthic bioluminescence at artificial
- 1773 food falls, 1000 to 4800 m depth, in the Porcupine Seabight and Abyssal Plain,
- 1774 North East Atlantic Ocean. *Marine Biology* **150**, 1053-1060.

- 1775 Giske, J., Huse, G. & Berntsen, J. 2001. Spatial modelling for marine resource 1776 management, with a focus on fish. *Sarsia* **86**, 405-410.
- 1777 Glover, A.G., Gooday, A.J., Bailey, D.M., Billet, D.S.M., Chevaldonné, P., Colaço, A.,
- 1778 Copley, J., Cuvelier, D., Desbruyères, D., Kalogeropoulou, V., Klages, M.,
- 1779 Lampadariuou, N., Lejeusne, C., Mestre, N.C., Paterson, G.L.J., Perez, T., Ruhl, H.,
- 1780 Sarrazin, J., Soltwedel, T., Soto, E.H., Thatje, S., Tselepides, A., Van Gaever, S. &
- 1781 Vanreusel, A. 2010a. Temporal changes in deep-sea benthic ecosystems: A review
- 1782 of the evidence from recent time-series studies. *Advances in Marine Biology An*
- 1783 *Annual Review* **58**, 1-95.
- 1784 Glover, A.G., Higgs, N.D., Bagley, P.M., Carlsson, R., Davies, A.J., Kemp, K.M., Last,
- 1785 K.S., Norling, K., Rosenberg, R., Wallin, K.-A., Källström, B. & Dahlgren, T.G.
- 1786 2010b. A live video observatory reveals temporal processes at a shelf-depth whale1787 fall. *Cahiers de Biologie Marine* **51**, 375-381.
- 1788 Godø, O.R. 1998. What can technology offer the future fisheries scientist possibilities
- 1789 for obtaining better estimates of stock abundance by direct observations. *Journal of*
- 1790 Northwest Atlantic Fishery Science 23, 105-131.
- Golombek, D.A. & Rosenstein, R.E. 2010. Physiology of circadian entrainment. *Physiological Review* 90, 1063-102.
- 1793 Gordoa, A. & Macpherson, E. 1991. Diurnal variation in the feeding activity and catch
- 1794 rate of cape hake (*Merluccius capensis* and *M. paradoxus*) off Namibia. *Fisheries*
- 1795 *Research* **12**, 299-305.
- Grant, R.A. & Halliday, T. 2010. Predicting the unpredictable; Evidence of pre-seismic
 anticipatory behaviour in the common toad. *Journal of Zoology* 281, 263-271.
- 1798 Hart, T.D., Clemons, J.R., Wakefield, W.W. & Heppel, S.S. 2010. Day and night
- abundance, distribution, and activity patterns of demersal fishes on Heceta Bank.
- 1800 *Oregon Fishery Bulletin* **108**, 466-477.

- 1801 Hawkins, S.J., Moore, P.J. Burrows, M.T., Poloczanska, E., Mieszkowska, N., Herbert,
- 1802 R.J.H., Jenkins, S.R.R., Thompson, C., Genner, M.J. & Southward, A.J. 2008.
- 1803 Complex interactions in a rapidly changing world: Responses of rocky shore 1804 communities to recent climate change. *Climate Research* **37**, 123-133.
- 1805 Herring, P. 2002. *The Biology of the Deep Ocean*. Oxford: Oxford University Press.
- 1806 Herring, P.J. & Roe, H.S.J. 1988. The photoecology of pelagic oceanic decapods.
 1807 Symposium of the Zoological Society of London 59, 263-290.
- Harrington, M. 2010. Location, location: Important for jet-lagged circadian
 loops. *Journal of Clinical Investigations* 120, 2265-2267.
- 1810 Hjellvik, V., Godø, O.R. & Tjøstheim, D. 2004. Diurnal variation in acoustic densities:
- 1811 why do we see less in the dark? *Canadian Journal of Fisheries and Aquatic*1812 Sciences 61, 2237-2254.
- Horne, J.K., Urmy, S.S. & Barbee, D.H. 2010. Using sonar to describe temporal
 patterns of oceanic organisms from the MARS observatory. In *Proceedings of IEEE-OCEANS'10.* Seattle WA: IEEE Press, 1-7.
- 1816 Hu, K., Scheer, F.A., Ivanov, P.Ch., Buijs, R.M. & Shea, S.A. 2007. The
- 1817 suprachiasmatic nucleus functions beyond circadian rhythm generation.
 1818 *Neuroscience* 149, 508-517.
- Huang, P.W. & Dai, S.K. 2003. Image retrieval by texture similarity. *Pattern Recognition* 36, 665-679.
- 1821 Huchet, G., Moore, K. & Tossell, D. 2010. NEPTUNE cabled ocean observatory:
- 1822 Scientific module deployment. In *Proceedings of the SubOptic Conference and*1823 *Convention*. Yokohama, Japan: SubOptic Press, 1-8.
- 1824 Hurlbut, T., & Clay, D. 1990. Protocols for research vessel cruises within the Gulf
- 1825 region (demersal fish) (1970-1987). *Canadian Manuscript Report of Fisheries and*
- 1826 *Aquatic Sciences* **2082**, 143.

- 1828oceanobserving.IFSOOConsultativeDraftversion71829(http://www.oceanobs09.net/).
- Janusz, J., Lipinski, M.R. & Draganik, B. 1998. Distribution and biology of the
 roundnose grenadier (*Coryphenoides rupestris*) from the Mid-Atlantic Ridge.
 Results of Polish investigations in the years 1986 1997 and 1998. Polish Reports
 from Research on the Mid-Atlantic Ridge.
- Jayas, D.S., Paliwal, J. & Visen, N.S. 2000. Multi-layer neural networks for image
 analysis of agricultural products. *Journal of Agricultural Engineering Research* 77,
- 1836 119-128.
- Jech, M.J. & Michaels, W.L. 2006. A multifrequency method to classify and evaluate
 fisheries acoustic data. *Canadian Journal of Fisheries and Aquatic Sciences* 63,
 2225-2235.
- Jennings, S. & Brander, K. 2010. Predicting the effects of climate change on marine
 communities and the consequences for fisheries. *Journal of Marine Systems* 79,
 418-426.
- Johnson, C.H., Elliott, J.A. & Foster, R. 2003. Entrainment of circadian programs. *Chronobiology International* 20, 741-774.
- 1845 Kaartvedt, S., Røstad, A., Klevjer, T.A. & Staby, A. 2009. Use of bottom-mounted
 1846 echo sounders in exploring behavior of mesopelagic fishes. *Marine Ecology*1847 *Progress Series* 395, 109-118.
- 1848 Kaufmanna, R.S., Fisherb, E.C., Gillb, W.H., Kingb, A.L., Laubacher, M. & Sullivan,
- 1849 B. 2003. Temporal patterns in the distribution, biomass and community structure of
- 1850 macrozooplankton and micronekton within Port Foster, Deception Island,
- 1851 Antarctica. *Deep-Sea Research II* **50**, 1765-1785.
- 1852 Kelly, N.E., Shea, E.K., Metaxas, A., Haedrich, R.L. & Auster, P.J. 2010. Biodiversity
- of the Deep-Sea continental margin bordering the Gulf of Maine (NW Atlantic):
 Relationships among sub-regions and to shelf systems. *PLoS I* 5, e13832.
- 1855 Kemp, K.M., Jamieson, A.J., Bagley, P.M., Collins, M.A. & Priede, I.G. 2008. A new
- 1856 technique for sequential periodic bait-release at a camera platform in the deep sea;
- trial at 3664m depth in the Charlie Gibbs Fracture Zone, Mid-Atlantic Ridge. *Deep Sea Research II* 55, 218-228.
- 1859 Kennelly, S.J., Graham, K.J., Montgomery, S.S., Andrew, N.L. & Brett, P.A. 1993.
- 1860 Variance and cost-benefit analyses to determine optimal duration of tows and levels
- 1861 of replication for sampling relative abundances of species using demersal trawling.
- 1862 *Fisheries Research* **16**, 51-67.
- 1863 Keskinen, E., Henrik Nygård, E.L., Røstad, A. & Thormar, J. 2004. New findings of
 1864 diel vertical migration in high Arctic ecosystems. Svalbard: University Centre
 1865 Publication Series (UNIS), Report AB320.
- 1866 Klein, L.A. 2004. Sensor and data fusion: A tool for information assessment and
 1867 decision making. International Society for Optical Engineering-SPIE Press
 1868 Monograph.
- 1869 Knuckey, I. & Dichmont, C. 2009. Review of Baywide annual Port Phillip Bay trawl
 1870 Program. Report to Fisheries Victoria. Victoria: Fishwell Consulting and CSIRO.
- 1871 Knuckey, I. & Hudson, R. 2007. Resource survey of the Great Australian Bight trawl
- 1872 fishery-2006 (2006 GABTF Resource Survey). AFMA Project R2005/1084.
 1873 Fishwell Consulting.
- 1874 Kritzer, M. & Sale, P.F. (eds) 2006. *Marine Metapopulations*. San Diego: Elsevier
 1875 Academic Press.
- 1876 Kronfeld-Schor, N. & Dayan, T. 2003. Partitioning of time as an ecological resource.
- 1877 *Annual Review of Ecology and Systematic* **34**, 153-181.

- 1878 Lampitt, R.S., Favali, P., Barnes, C.R., Church, M.J., Cronin, M.F., Hill, K.L., Kaneda,
- 1879 Y., Karl, D.M., Knap, A.H., McPhaden, M.J., Nittis, K.A., Priede, I.G., Rolin, J.-F.,
- 1880 Send, U., Teng, C.-C., Trull, T.W., Wallace, D.W.R. & Weller, R.A. 2010. In situ
- 1881 sustained Eulerian observatories. In Proceedings of the Observations and
- 1882 Information for Society conference (OceanObs'09): Sustained Ocean, J. Hall et al.
- 1883 (eds). Venice, Italy: ESA Publication, **306**.
- Lauritzen, S.T. 1981. Time-series analysis in 1880. A discussion of contributions made
 by T.N. Thiele. *International Statistical Review* 49, 319-333.
- 1886 Le Loeuff, P. & Intes, A. 1999. Macrobenthic communities on the continental- shelf of
- 1887 Côte-d'Ivoire. Seasonal and diel cycles in relation to hydroclimate. *Ocenologica*1888 Acta 22, 529-550.
- Lemon, D.D., Chave, R.A., Clarke, M.R., Dewey, R.K. & Macoun, P. 2007. Inverted
 echo sounder on a cabled observatory. In *Proceedings of the IEEE-OCEANS'07*.
 Vancouver BC: IEEE Press, 1-7.
- 1892 Lenoir, S., Beaugrand, G. & Lecuyer, E. 2011. Modelled spatial distribution of marine
- 1893 fish and projected modifications in the North Atlantic Ocean. *Global Change*1894 *Biology* 17, 115-129.
- 1895 Long-term Ecosystem Observatory (LEO) 2011. http://www.ccpo.odu.edu/TOO/; Last
 1896 access: 07/06/2011.
- 1897 Lepetz, V., Massot, M., Schmeller, D.S. & Clobert, J. 2009. Biodiversity monitoring:
- 1898 Some proposals to adequately study species' responses to climate change.
 1899 *Biodiversity and Conservation* 18, 3185-3203.
- 1900 Levin, L.A., Etter, R.J., Rex, M.A., Gooday, A.J., Smith, C.R., Pieda, J., Stuart, C.T.,
- 1901 Hessler, R.R. & Pawson, D. 2001. Environmental influences on regional deep-sea
- 1902 species diversity. *Annual Review of Ecology and Systematic* **32**, 51-93.

- Levine, J.D., Funes, P., Dowse, H.B. & Hall, J.C. 2002. Signal analysis of behavioural
 and molecular cycles. *BMC Neuroscience* 3, 1.
- Li, M. & Yuan, B. 2004. 2D-LDA: A statistical linear discriminant analysis for image
 matrix. *Pattern Recognition Letters* 26, 527-532.
- Lickey, M.E., Block, G.D., Hudson, D. & Smith, JT. 1976. Circadian oscillators and
 photoreceptors in the gastropod *Aplysia*. *Photochemistry and Photobiology* 23,
 253-273.
- 1910 Lillholm, M., Nielase, M. & Griffin, L.D. 2003. Feature-based image analysis.
 1911 *International Journal of Computer Vision* 52, 73-95.
- 1912 Lipton, A.J., Fujiyoshi, H. & Patil, R.S. 1998. Moving target classification and tracking
- 1913 from real-time video. In Proceedings of the IEEE Workshop on Applications of
 1914 Computer Vision (WACV'98). Princeton NJ: IEEE Press, 8-14.
- 1915 Lisin, S.E., Hannan, E.E., Kockevar, R.E., Harrold, C. & Barry, J.P. 1997. Temporal
 1916 variation in gametogenic cycles of vesicomyd clams. *Invertebrate Reproduction*1917 *and Development* **31**, 307-318.
- 1918 Liu, J.C., Hwang, W.L., Chen, M.S., Tsai, J.W. & Lin, C.H. 2003. Active contour
- model using wavelet modulus for object segmentation and tracking video
 sequences. International Journal of Wavelets, Multiresolution and Information
 Processing 1, 93-113.
- 1922 Longhurst, A. 2007. *Ecological Geography of the Sea*. Burlington MA: Academic1923 Press.
- Lorenzen, C.J. 1972. Extinction of light in the ocean by phytoplankton. *Journal of the Council for the International Exploitation of the Sea* 34, 262-267.
- 1926 Lowe, G. 1999. Object recognition from local scale-invariant features. In Proceedings
- 1927 of the International Conference on Computer Vision (ICCV'07). Kerkyra, Greece:
- 1928 IEEE Press, 1150-1157.

- 1929 Lytle, D.A., Martínes-Muñoz, G., Zhang, W., Larios, N., Shapiro, L., Paasch R.,
- 1930 Moldenke, A., Mortensen, E.N., Todorovic, S. & Dietterich, T.G. 2010. Automated
- 1931 processing and identification of benthic invertebrate samples. *Journal of Natural*
- 1932 *American Benthological Society* **29**, 867-874.
- Majumder, S., Scheding, S. & Durrant-Whyte, F.H. 2002. Multisensor data fusion for
 underwater navigation. *Robotics and Autonomous Systems* 35, 97-108.
- Mánuel, A., Nogueras, M. & Del Rio, J. 2010. OBSEA an Expandable Seafloor
 Observatory. *Sea Technology* 51, 7 only.
- 1937 Margalef, R. 1986. *Ecología*. Barcelona, Spain: Ediciones Omega.
- 1938 Matabos, M., Aguzzi, J., Robert, K., Costa, C., Menesatti, P., Company, J.B. & Juniper,
- 1939 S.K. 2011. Multi-parametric study of behavioural modulation in demersal decapods
- at the VENUS cabled observatory in Saanich Inlet, British Columbia, Canada. *Journal of Experimental Marine Biology and Ecology* 401, 89-96.
- 1942 Martha's Vineyard Coastal Observatory (MAVCO) 2011.
- 1943 http://www.whoi.edu/mvco/description/description2.html; Last access: 07/06/2011.
- 1944 May, J.L. & Blabe, S.J.M. 1989. Benthic and pelagic fish biomass of the upper 1945 continental slope off eastern Tasmania. *Marine Biology* **101**, 11-25.
- 1946 McQuinn, I.H., Simard, Y., Stroud, T.W.F., Beaulieu, J.-L. & Walsh, S.J. 2005. An
- adaptive, integrated "acoustic-trawl" survey design for Atlantic cod (*Gadus morhua*) with estimation of the acoustic and trawl dead zones. ICES Journal of
 Marine Science 62, 93-106.
- 1950 Menesatti, P., Aguzzi, J., Costa, C., García, J.A. & Sardà, F. 2009a. Video-image
- analysis for microcosm experiments on activity rhythms with multiple individuals of
- 1952 Norway lobster, *Nephrops norvegicus* (L.). *Journal of Neuroscience Methods* 184,
- 1953 161-168.

- 1954 Menesatti, P., D'andrea, S., Cavalieri, A., Baldi, M., Lanini, M., Vitali, A., Lacetera, N.,
- 1955 Bernabucci, U. & Nardone, A. 2009b. Microclimatic indexes to compare different
- 1956 dairy cows housing systems in different climatic conditions. In *Proceedings of the*
- 1957 60th European Association of Animal Production (EAAP). Barcelona (Spain):
- 1958 Wageningen Academic Publishers *Book of Abstract* **15**, 117.
- 1959 Mikada, H., Kasahara, J., Fujii, N. & Kumazawa, M. 2010. Active monitoring using
- 1960 submarine cables-leveraging offshore cabled observatory for passive monitoring.
- 1961 *Handbook of geophysical exploration: Seismic exploration* **40**, 473-491.
- 1962 Mitchell, H.B. 2007. *Multi-Sensor Data Fusion: An Introduction*. Berlin: Springer1963 Verlag.
- Mitchie, D., Spiegelhalter, D.J. & Taylor, C. 1994. *Machine Learning, Neural and Statistical Learning*. New York: Ellis Horwood.
- Morato, T., Watson, R., Pitcher, T.J. & Pauly, D. 2006. Fishing down the deep. *Fish and Fisheries* 7, 24-34.
- 1968 Mueller, R.P., Brown, R.S., Hop, H. & Moulton, L. 2006. Video and acoustic camera
- 1969 techniques for studying fish under ice: A review and comparison. *Reviews in Fish*
- 1970 *Biology and Fisheries* **2**, 213-226.
- 1971 Navia, A.F., Mejía-Falla, P.A. & Giraldo, A. 2007. Feeding ecology of elasmobranch
- 1972 fishes in coastal waters of the Colombian eastern tropical Pacific. *BMC Ecology* 7, 81973 only.
- 1974 Naylor, E. 2005. Chronobiology: implications for marine resources exploitation and
 1975 management. *Scientia Marina* 69, 157-167.
- 1976 Naylor, E. 2006. Orientation and navigation in coastal and estuarine zooplankton.
 1977 *Marine and Freshwater Behaviour and Physiology* **39**, 13-24.
- 1978 Naylor, E. 2010. Chronobiology of Marine Organisms. Cambridge: Cambridge
- 1979 University Press.

- 1980 Neat, F.C., Wright, P.J., Zuur, A.F., Gibb, I.M., Gibb, M.F., Tulett, D., Righton, D.A. &
- Turner, R.J. 2006. Residency and depth movements of a coastal group of Atlantic
 cod (*Gadus morhua* L.). *Marine Biology* 148, 643-654.
- 1983 Neilson, J.D., Clark, D., Melvin, G.D., Perley, P. & Stevens, C. 2003. The diel-vertical
- 1984 distribution and characteristics of pre-spawning aggregations of pollock (*Pollachius*
- 1985 *virens*) as inferred from hydroacoustic observations: The implications for survey
- 1986 design. *ICES Journal of Marine Science* **60**, 860-871.
- 1987 NEPTUNE 2011. http://www.neptunecanada.ca; Last access: 18/04/2011.
- 1988 Nieddu, L. & Patrizi, G. 2000. Formal methods in pattern recognition: A review.
 1989 *European Journal of Operational Research* 120, 459-495.
- Nosal, E.M. & Frazer, L.N. 2007. Sperm whale three-dimensional track, swim
 orientation, beam pattern, and click levels observed on bottom-mounted
 hydrophones. *Journal of the Acoustical Society of America* 122, 1969-1978.
- 1993 Nucci, M.E., Costa, C., Scardi, M. & Cataudella, S. 2010. Preliminary observations on
- Bluefin Tuna (*Thunnus thynnus*, Linnaeus 1758) behaviour in captivity. *Journal of*Applied Ichthyology 26, 95-98.
- Occhipinti-Ambrogi, A. 2007. Global change and marine communities: Alien species
 and climate change. *Marine Pollution Bulletin* 55, 342-352.
- 1998 O'Dor, R. & Stokesbury, M.J.W. 2009. The Ocean Tracking Network: Adding marine
- animal movements to the Global Ocean Observing System. In *Tagging and tracking*
- 2000 of marine animals with electronic devices; Reviews: Methods and technologies in
- *fish biology and fisheries* 9. J.L. Nielsen et al. (eds). Berlin: Springer, 91-100.
- 2002 O'Dor, R., Miloslavich, P. & Yarincik, K. 2010. Marine biodiversity and biogeography:
- 2003 Regional comparisons of global issues, an Introduction. *PLoS I* **5**, e11871.
- 2004 Ocean Observatory Initiative (OOI) 2011. http://www.oceanobservatories.org; Last
- 2005 access: 07/06/2011.

- 2006 Ocean Tracking Network (OTN) 2011. http://oceantrackingnetwork.org/index.html;
 2007 Last access: 18/04/2011.
- 2008 Park, O. 1941. Concerning the community symmetry. *Ecology* 22, 164-167.
- 2009 Pauly, D., Alder, J., Bennett, E., Christensen, V., Tyedmers, P. & Watson, R. 2003. The
- 2010 future for fisheries. *Science* **302**, 1359-1361.
- 2011 Pei, J.H. & Xie, W.X. 1999. Adaptive multi thresholds image segmentation based on
- 2012 potential function clustering. *Chinese Journal of Computers* **7**, 758-762.
- 2013 Perry, A.L., Low, P.J., Ellis, J.R. & Reynolds, J.D. 2005. Climate change and
 2014 distribution shifts in marine fishes. *Science* 308, 1912-1915.
- 2015 Petitgas, P., Massé, J., Beillois, P., Lebarbier, E. & Le Cann, A. 2003. Sampling
- 2016 variance of species identification in fisheries acoustic surveys based on automated
- 2017 procedures associating acoustic images and trawl hauls. *ICES Journal of Marine*2018 *Science* 60, 437-445.
- 2019 Petrakis, G., MacLennan, D.D. & Newton, A.W. 2001. Day-night and depth effects on
 2020 catch rates during trawl surveys in the North Sea. *ICES Journal of Marine Science*2021 58, 50-60.
- 2022 Pillar, S.C. & Barange, M. 1997. Diel variability in bottom trawl catches and feeding
 2023 activity of the Cape hakes of the west coast of South Africa. *ICES Journal of*2024 *Marine Science* 54, 485-499.
- Priede, I.G. & Bagley, P.M. 2000. In situ studies on deep-sea demersal fishes using
 autonomous unmanned lander platforms. *Oceanography and Marine Biology: An Annual Review* 38, 357-392.
- Priede, I.G., Bagley, P.M. & Smith Jr., K.L. 1994. Seasonal change in activity of
 abyssal demersal scavenging grenadiers *Coryphaenoides (Nematonurus) armatus* in
 the eastern North Pacific Ocean. *Limnology and Oceanography* **39**, 279-285.

- Priede, I.G., Deary, A.R., Bailey, D.M. & Smith, K.L. 2003. Low activity and seasonal
 change in population size structure of grenadiers in the oligotrophic abyssal central
 North Pacific Ocean. *Journal of Fish Biology* 63,187-196.
- 2034 Priede, I.G., Godbold, J.A., Niedzielski, T., Collins, M.A., Bailey, D.M., Gordon,
- 2035 J.D.M. & Zuur, A.F. 2011. A review of the spatial extent of fishery effects and
- 2036 species vulnerability of the deep-sea demersal fish assemblage of the Porcupine
- 2037 Seabight, Northeast Atlantic Ocean (ICES Subarea VII). *ICES Journal of Marine*2038 *Science* 68, 281-289.
- 2039 Priede, M., Solan, M., Mienert, J., Person, R., van Weering, T.C.E., Pfannkuche, O.,
- 2040 O'Neill, N., Tselepides, A., Thomsen, L., Favali, P., Gasparoni, F., Zitellini, N.,
- 2041 Millot, A., Gerber, H.W., DeMiranda, J.M.A., Klages, M. & Sigray, P. 2004.
- ESONET- European Sea Floor Observatory Network. In *Proceedings of the IEEE- OCEANS'04*. Kobe, Japan: IEEE Press, 2155-2163.
- Priede, I.G., Jamieson, A., Heger, A., Craig, J. & Zuur, A.F. 2008. The potential
 influence of bioluminescence from marine animals on a deep-sea underwater
 neutrino telescope array in the Mediterranean Sea. *Deep-Sea Research I* 55, 14741483.
- 2048 Priede, I.G., Williams, L.M., Wagner, H.-J., Thom, A., Brierley, I., Collins, M.A.,
- 2049 Collin, S.P., Merrett, N.R. & Yau, C. 1999. Implication of the visual system in
- 2050 regulation of activity cycles in the absence of solar light: 2-[125I]iodomelatonin
- 2051 binding sites and melatonin receptor gene expression in the brains of demersal deep-
- sea gadiform fishes. *Proceedings of the Royal Society Series B* **26**, 2295-2302.
- 2053 Pulcini, D., Costa, C., Aguzzi, J. & Cataudella, S. 2008. Light and shape: A contribution
- 2054 to demonstrate morphological differences in diurnal and nocturnal teleosts. *Journal*
- 2055 *of Morphology* **269**, 375-385.

- Pusc, C., Fock, H., Porteiro, F.M. & von Westernhagen, H. 2002. Interaction of
 mesopelagic fish and shallow topography in different latitudes of the NE Atlantic.
 ICES Theme session on oceanography and ecology of seamounts: Indications of
 unique ecosystems, CM 2002/M:10.
- 2060 Raffaelli, D., Bell, E., Weithoff, G., Matsumoto, A., Cruz-Motta, J.J., Kershaw, P.,
- 2061 Parker, R., Parry, D. & Jones, M. 2003. The ups and downs of benthic ecology:
- 2062 considerations of scale, heterogeneity and surveillance for benthic-pelagic coupling.
- 2063 *Journal of Experimental Marine Biology and Ecology* **285/286**, 191-203.
- 2064 Ramirez-Llodra, E., Brandt, A., Danovaro, R., De Mol, B., Escobar, E., German, C.R.,
- 2065 Levin, L.A., Martinez-Arbizu, P., Menot, L., Buhl-Mortensen, P., Narayanaswamy,
- 2066 B.E., Smith, C.R., Tittensor, D.P., Tyler, P.A., Vanreusel, A. & Vecchione, M.
- 2067 2010. Deep, diverse and definitely different: Unique attributes of the world's largest
 2068 ecosystem. *Biogeosciences* 7, 2851-2899.
- 2069 Reebs, S.G. 2002. Plasticity of diel and circadian activity rhythms in fishes. *Reviews of*
- 2070 Fish Biology and Fisheries 12, 349-371.
- 2071 Refinetti, R. 2006. Circadian Physiology. New York: Francis and Taylor.
- 2072 Reier Knudsen, F. & Larsson, P. 2009. Discriminating the diel vertical migration of fish
- and *Chaoborus flavicans* larvae in a lake using a dual-frequency echo sounder. *Aquatic Living Resources* 22, 273-280.
- 2075 Relini, G. & Piccinetti, C. 1994. Ten years of trawl surveys in Italian Seas (1985-1995).
- 2076 FAO Fishery Report **533**, 21-41.
- 2077 Roberts, C.M. 2002. Deep impact: The rising toll of fishing in the deep-sea. *Trends in*2078 *Ecology and Evolution* 17, 242-245.
- 2079 Roberts, P.D., Jaffe J.S. & Trivedi, M.M. 2011. Multiview, broadband acoustic
- 2080 classification of marine fish: A machine learning framework and comparative
- analysis. *IEEE Journal of Oceanic Engineering* **36**, 90-104.

- 2082 Roessig, J.M., Woodley, C.M., Cech Jr., J.J. & Hansen, L.J. 2004. Effects of global
 2083 climate change on marine and estuarine fishes and fisheries. *Reviews of Fish*2084 *Biology and Fisheries* 14, 251-275.
- 2085 Roff, J.C. & Evans, S.M.J. 2002. Frameworks for marine conservation: Non-
- 2086 hierarchical approaches and distinctive habitats. *Aquatic Conservation: Marine and*2087 *Freshwater Ecosystems* 12, 635-648.
- 2088 Roff, J.C. & Taylor, M.E. 2000. National frameworks for marine conservation a
 2089 hierarchical geophysical approach. *Aquatic Conservation: Marine and Freshwater*2090 *Ecosystems* 10, 209-223.
- 2091 Rohlf, F.J. & Archie, J.W. 1984. A comparison of Fourier Methods for the description
- of wing shape in mosquitoes (Diptera: Culicidae). *Systems Biology* **33**, 302-317.
- 2093 Ruhl, H.A., André, M., Beranzoli, L., Çağatay, N.M., Colaço, A., Cannat, M.,
- 2094 Dañobeitia, J.J., Favali, P., Géli, L., Gillooly, M., Greinert, J., Hall, P.O.J., Huber,
- 2095 R., Karstensenm, J., Lampitt, R.S., Larkin, K.E., Lykousis, V., Mienert, J., Miranda,
- 2096 J.M., Person, R., Priede, I.G., Puillat, I., Thomsen, L. & Waldmann, C. 2011. Open-
- 2097 ocean observatories in Europe. *Progress in Oceanography* IN PRESS.
- 2098 Ryer, C.H., Rose, C.S. & Iseri, P.J. 2010. Flatfish herding behaviour in response to
- trawl sweeps: A comparison of diel responses to conventional sweeps and elevated
 sweeps. *Fishery Bulletin* 108, 145-154.
- 2101 Sardà, F., Company, J.B. & Castellón, A. 2003. Intraspecific aggregation structure of a
- shoal of Western Mediterranean (Catalan coast) deep-sea shrimp, Aristeus
- 2103 antennatus (Risso, 1816), during the reproductive period. Journal of Shellfish
- 2104 *Research* **22**, 569-579.
- Sardà, F., D' Onghia, G., Politou, C.Y. & Tselepides, A. (eds) 2004. Mediterranean
 Deep-sea biology. *Scientia Marina* 68, 1-204.

- 2107 Sassa, C., Tsukamoto, Y., Yamamoto, K. & Tokimura, M. 2010. Spatio-temporal
- distribution and biomass of *Benthosema pterotum* (Pisces: Myctophidae) in the shelf
 region of the East China Sea. *Marine Ecology Progress Series* 407, 227-241.
- 2110 Sarriá, D., Molino, E., Rotllant, G., Mànuel, A., Aguzzi, J. & Sardà, F. 2011.
- 2111 Monitoring species in artificial reefs using acoustic communications.
 2112 *Instrumentation Viewpoint* 11, 48 only.
- Shannon, C.E. 1949. Communication in the presence of noise. *Proceedings of the Institute of Radio Engineers* 37, 10-21.
- 2115 Sheppard, D.C. 2000. Seas at the Millennium: an Environmental Evaluation. Global
 2116 Issues and Processes. Amsterdam and Oxford: Pergamon Press.
- 2117 Sherman, A.D. & Smith Jr., K.L. 2009. Deep-sea benthic boundary layer communities
- and food supply: A long-term monitoring strategy. *Deep-Sea Research II* 56, 17541762.
- 2120 Shotton, D.M., Rodriguez, A., Guil, N. & Trelles, O. 2000. Object tracking and event
- 2121 recognition in biological microscopy videos. In Proceedings of the 15th
- 2122 International Conference on Pattern Recognition (ICPR'2000). Barcelona, Spain:
- 2123 IEEE Press, 226-229.
- Simmonds, E.J., Armstrong, F. & Copland, P.J. 1996. Species identification using
 wideband backscatter with neural network and discriminant analysis. *ICES Journal of Marine Science* 53, 189-195.
- Smith, C.R. & Baco, A.R. 2003. Ecology of whale falls at the deep-sea floor. *Oceanography and Marine Biology: An Annual Review* 41, 311-354.
- 2129 Smith, K.L., Kaufmann, R.S. & Wakefield, W.W. 1993. Mobile megafaunal activity 2130 monitored with a time-lapse camera in the abyssal North Pacific. *Deep-Sea*
- 2131 *Research I* **411**, 2307-232.

- Smith, A., Priede, I.G., Bagley, P.M. & Addison, S.W. 1997. Interception and dispersal
 of artificial food falls by scavenging fishes in the abyssal NE Atlantic: Early season
 observations prior to annual deposition of phytodetritus. *Marine Biology* 128, 329336.
- 2136 Smith, K.L., Ruhl, H.A., Bett, B.J., Billett, D.S.M., Lampitt, R.S. & Kaufmann, R.S.
- 2137 2009. Climate, carbon cycling, and deep-ocean ecosystems. *Proceedings of the*2138 *National Academy of Sciences* 106 19211-19218.
- Somot, S., Sevault, F. & Déqué, M. 2006. Transient climate change scenario simulation
 of the Mediterranean Sea for the twenty-first century using a high-resolution ocean
- circulation model. *Climate Dynamics* **27**, 851-879.
- 2142 Spampinato, C., Chen-Burger, Y.-H., Nadarajan, G. & Fisher, R.B. 2008. Detecting,
- tracking and counting fish in low quality unconstrained underwater videos. In *Proceedings of the 3rd International Conference on Computer Vision Theory and*

2145 *Applications (VISAPP'2)*. Madeira, Portugal: VISAPP Press, 514-519.

- Stephens Jr., J.S., Hose, J.E. & Love, M.S. 1988. Fish assemblages as indicators of
 environmental change in nearshore environments. In *Marine organisms as*
- 2148 *indicators*, D.F. Soule & G.S. Keppel (eds). Berlin: Springer, 91-105.
- 2149 Stoner, A.W., Ryer, C.H., Parker, S.J., Auster, P.J. & Wakefield, W.W. 2008.
- 2150 Evaluating the role of fish behaviour in surveys conducted with underwater
- 2151 vehicles. *Canadian Journal of Fish and Aquatic Sciences* **65**, 1230-1243.
- Storbeck, F. & Daan, B. 2001. Fish species recognition using computer vision and a
 neural network. *Fisheries Research* 51, 11-15.
- Strachan, N.J.C. & Nesvadba, P. 1990. Fish species recognition by shape analysis of
 images. *Pattern Recognition* 23, 539-544.
- 2156 Suetsugu, K. & Otha, S. 2005. Day and nighttime changes in species composition of
- 2157 Deep-Sea demersal fishes. *Journal of Oceanography* **61**, 187-196.

- Tobar, R. & Sardà, F. 1992. Annual and diel light cycle as a predictive factor in the
 fisheries for the prawn (*Aristeus antennatus* Risso, 1816). *Fisheries Research* 15,
 169-179.
- Tosini, G. & Aguzzi, J. 2005. Effects of space flight on circadian rhythms. In *Experimentation with the animal model in space*, G. Sonnenfield (ed.). Elsevier:
 Amsterdam, 165-174.
- Tosini, G., Aguzzi, J., Bullock, N.M., Liu, C. & Kasamatsu, M. 2007. Effect of
 photoreceptor degeneration on circadian photoreception and free-running period in
 the Royal College of Surgeons Rat. *Brain Research* 1148, 76-82.
- 2167 Toth, D. & Aach, T. 2003. Detection and recognition of moving objects using statistical
- 2168 motion detection and Fourier Descriptors. In *Proceedings of the 12th International*
- 2169 *Conference on Image Analysis and Processing (ICIAP'03).* Los Alamitos CA: IEEE
 2170 Press, 430-435.
- Trenkel, V.M., Lorance, P. & Mahévas, S. 2004. Do visual transects provide true
 population density estimates for deepwater fish? *ICES Journal of Marine Science*61, 1050-1056.
- Tyler, P.A., Campos-Creasy, L.S. & Giles, L.A. 1994. Environmental control of quasicontinuous and seasonal reproduction in deep-sea benthic invertebrates. In *Reproduction, Larval Biology and Recruitment of the Deep-sea Benthos*, C.M.
- 2177 Young & K.J. Eckelbarger (eds). New York: Columbia University Press, 158-178.
- Veeraraghavan, A., Roy-Chowdhury, A.K. & Chellappa, R. 2005. Matching shape
 sequences in video with applications in human movement analysis. *IEEE Transactions on Pattern Analysis and Machine Intelligence* 27, 1896-1909.
- 2181 Vivekanandan, E., Mweyiappan, E.E., Mohanraj, G., Nammalwar, P., Thangaraj,
- 2182 Subramanian, V. & Sarvesan, R. 2002. Marine fishery resources of Chennai:
- 2183 Landings by day and night trawlers. In *Proceedings of the 5th Indian Fishery Forum*.

- S. Ayyappan et al. (eds). Orissa, India: Asian Fishery Society, Indian Branch
 (AFSIB), Mangalore, and AoA, 331-334.
- Voss, R. & Zeil, J. 1995. Automatic tracking of complex objects under natural
 conditions. *Biological Cybernetics* 73, 415-423.
- 2188 Wagner, H.J., Kemp, K., Mattheus, U. & Priede, I.G. 2007. Rhythms at the bottom of
- the deep-sea: Cyclic current flow changes and melatonin patterns in two species of
- demersal fish. *Deep-Sea Research I* **54**, 1944-1956.
- 2191 Walther, G.R., Post, E., Convey, P., Menzel, A., Parmesan, C., Beebee, T.J., Fromentin,
- J.M., Hoegh-Guldberg, O. & Bairlein, F. 2002. Ecological responses to recent
 climate change. *Nature* 416, 389-395.
- 2194 Walther, D., Edgington, D.R. & Koch, C. 2004. Detection and tracking of objects in
- 2195 underwater video. In Proceedings of the IEEE Computer Society Conference on
- 2196 Computer Vision and Pattern Recognition (CVPR'04). Washington D.C.: IEEE
 2197 Press, 544-549.
- Waterman, T. 2001. Evolutionary challenges of extreme environments (Part 2). *Journal of Experimental Zoology* 291, 130-168.
- 2200 Watson, W.H., Bedford, L. & Chabot, C.C. 2008. Rhythms of locomotion expressed by
- *Limulus polyphemus*, the American horseshoe crab: II. Relationship to circadian
 rhythms of visual sensitivity. *Biological Bulletin* 215, 46-56.
- Welsh, D.K., Takahashi, J.S. & Kay, S.A. 2010. Suprachiasmatic nucleus: Cell
 autonomy and network properties. *Annual Review of Physiology* 72, 551-577.
- 2205 Williams, R.N., Lambert, T.J., Kelsall, A.F. & Pauly, T. 2006. Detecting marine
- animals in underwater video: Let's start with salmon. In *Proceedings of the 12th*
- 2207 Americas Conference on Information Systems (AMCIS). Acapulco, Mexico:
- 2208 AMCIS Press, 1482-1490.

- Whittaker, R.H., Levin, S.A. & Root, R.B. 1973. Niche, habitat, and ecotope. *The American Naturalist* 107, 321-338.
- Widder, E.A., Robison, B.H., Reisenbichler, K.R. & Haddock, S.D.H. 2005. Using red
 light for in situ observations of deep-sea fishes. *Deep-Sea Research* 52, 2077-2085.
- 2213 Witte, U., Aberle, N., Sand, M. & Wenzhöfer, F. 2003a. Rapid response of a deep-sea
- benthic community to POM enrichment: An in situ experimental study. *Marine Ecology Progress Series* 251, 27-36.
- 2216 Witte, U., Wenzhöfer, F., Sommer, S., Boetius, A., Heinz, P., Aberle, N., Sand, M.,
- 2217 Cremer, A., Abraham, W.R., Jørgensen, B.B. & Pfannkuche, O. 2003b. In situ
- experimental evidence of the fate of a phytodetritus pulse at the abyssal sea floor.
- 2219 *Nature* **424**, 763-766.
- Wu, W., Mallet, Y, Walczak, B, Penninckx, W, Massart, D.L., Heuerding S. & Erni, F.
 1996. Comparison of regularized discriminant analysis linear discriminant analysis
 and quadratic discriminant analysis applied to NIR data. *Analytica Chimica Acta*329, 257-265.
- Yamada, Y. & Forger, D. 2010. Multiscale complexity in the mammalian circadian
 clock. *Current Opinion in Genetic and Development* 20, 626-633.
- 2226 Yamazaki, S., Numano, R., Abe, M., Hida, A., Takahashi, R., Ueda, M., Block, G.D.,
- Sakaki, Y., Menaker, M. & Tei, H. 2000. Resetting central and peripheral circadian
 oscillators in transgenic rats. *Science* 288, 682.
- Young, C.M. 2003. Reproduction, development and life-history traits. In *Ecosystems of the deep oceans*, Tyler, P.A. (ed). Elsevier: Amsterdam, 381-426.
- Yousif, A. & Sabrah, M.M. 2004. Catchability of the red mullet (*Upeneus* spp.) from
 North-West Red Sea during autumn. *Egyptian Journal of Aquatic Research* 30,
 2233 290-305.

Zaret, T.M. & Suffern, J.S. 1976. Vertical migration in zooplankton as a predator
avoidance mechanism. *Limnological Oceanography* 21, 804-813.

Table 1 Main studies on video image analysis of moving biological objects

Method	Theme	Citation
Coarse-coded channel for classification	Automated taxonomy of dinoflagellates for microscopy	Culverhouse et al. 2003
Segmentation and neural network-based classification	Automatic counter to identify and measure phytoplankton	Embleton et al. 2003
Wavelet-based method for identification based on profile	Cell motion detection	Liu et al. 2003
Movement tracking	Bacterial rotation filming	Shotton et al. 2000
Image subtraction and segmentation for identification		Edgington et al. 2003
Saliency-based system with Kalman Filters for tracking	Remotely Operated Vehicle (ROV) observations for pelagic community monitoring	Edgington et al. 2004
Saliency-based system with Kalman Filters for detection and tracking		Edgington et al. 2006
Bayesian classifier for detection and tracking		Edgington et al. 2007
Image subtraction for detection and tracking		Walther et al. 2004
Saliency-based algorithm for detection	Permanent cameras for benthic community monitoring	Cline et al. 2007, Aguzzi et al. 2009c, 2010b
Artificial neural networks and discriminant analysis for semi-automated sizing and counting	Dual (stereo) camera system for cultured fish	Simmonds et al. 1996, Williams et al. 2006, Costa et al. 2006, 2009c, Nucci et al. 2010
Image texture and colour analysis for classification	Reef monitoring	Spampinato et al. 2008

Method	Theme	Citation
Neural networks and discriminant analysis on shapes	Fish species classification by computer vision	Storbeck & Daan 2001
Discriminant analysis of shapes	Fish classification from photograph banks	Strachan & Nesvabda 1990
Frame subtraction and image filtering for tracking	Measurements of rhythmic movements in individual lobsters in laboratory conditions	Aguzzi et al. 2009d
Frame subtraction, image filtering, and shape matching for tag identification and tracking	Measurements of rhythmic movements in a group of lobsters in laboratory conditions	Menesatti et al. 2009a, Aguzzi et al. 2011c
Landmark based geometric morphometry	Morphologic discrimination of scleractinian reef corals	Budd et al. 1994
Multi-target initiation, clutter modelling, event analysis and multiple hypothesis filtering for tracking	Permanent infrared cameras for bat monitoring	Betke et al. 2007
Mirroring systems for three-dimensional trajectories tracking	Tracking of swimming-flying species	Chraskova et al. 1999
Frames subtraction, segmentation and pattern matching for tracking	Tracking of flying insects filmed under natural conditions	Voss & Zeil 1995

Table 2. Marine multiparametric cabled seafloor observatories bearing video cameras and equipped for the long-term and real-time monitoring of continental margin ecosystems. Information on geographic location and water depth (m) is provided along with the web page of reference (URL, as active or inactive at last access the 07/06/2011) and numeral code as referenced for Figure 8B.

Code	Name/Acronym	Location	Depth	URL
1	Monterey Accelerated Research System (MARS)	Monterrey (California, USA)	891	www.mbari.org/mars ACTIVE
2	Victoria Experimental Network Under the Sea (VENUS)	i) Saanich Inlet (British Columbia, Canada) ii) Bonne Bay Observatory (Strait of Georgia, New Foundland, Canada)	100 300	www.venus.uvic.ca ACTIVE http://bbo.physics.mun.ca/ background INACTIVE
2	North East Pacific Time Series Undersea Network Experiment (NEPTUNE)	i) Barkley Canyon (Juan de Fuca plate, Canada) ii) Hydrothermal vent site (Juan de Fuca plate, Canada)	400 920 3200	www.neptunecanada.ca ACTIVE
3	Deep-ocean Environmental Long-term Observatory System (DELOS)	Angola	1500	www.delos-project.org (not presently cabled being retrieved yearly for data downloading) ACTIVE
4	JAMSTEC Off- Hatsushima (DONET)	Sagami Bay (Japan)	1175	www.jamstec.go.jp/scdc/top_ e.html ACTIVE
5	JAMSTEC off-Cape Muroto (DONET)	Nankai Trough (Japan)	3572	www.jamstec.go.jp/scdc/top_ e.html ACTIVE
6	Life-Streamed Underwater Video Observatory (ESONET)	Koster Fjord (Sweden)	30	www.sams.ac.uk/sams- news/Media%20releases/wor ld-first-scientific-video- underwater-observatory ACTIVE
7	Expandable Seafloor Observatory (OBSEA- ESONET)	Western Mediterranean (Vilanova i la Geltru, Spain)	20	http://sites.upc.edu/~www- sarti/OBSEA/info/intro/intro _eng.html ACTIVE

2266 Figure 1 Variation in species composition as reported from 24-h trawl hauls 2267 conducted continuously over consecutive days at different depths (i.e., shelf: 100 m; 2268 slope: 400 m) in the western Mediterranean in autumn-winter (October 1998). The 2269 temporal variation in the species composition within different samples is shown by 2270 similarity index-based cluster analysis (A), which discriminates day from night hauls 2271 (bold black in grey boxes) (adapted by Bahamon et al. 2009). Variations in catches are 2272 the product of changing species abundances (B) according to day-night cycles (grey vertical bars indicate night), as can be seen for some ecologically and economically 2273 2274 important crustacean, fish and decapod species (all values are surface density estimates (no. ind. km⁻²) standardised as a percentage of the maximum for comparison). Fish 2275 2276 catch patterns were obtained from NERIT project unpublished data. Decapod catch 2277 plots were adapted from different studies, all reviewed in Aguzzi et al. (2009a).

2278

2279 Figure 2 Seasonal differences in population sampling on the shelf (100 m) and 2280 on the slope (400 m) for species in the example described in Figure 1B. Total sampled 2281 animals are reported for two different seasons for comparison: autumn-winter (October; 2282 32 and 34 catches on the shelf and the slope, respectively) and spring-summer (June; 32 2283 catches on the shelf and the slope). Differences in population catchability were 2284 considered as significant (*) when the total number of animals sampled in October and 2285 June showed variation equal to or greater than twofold. Data for fishes were obtained 2286 from NERIT project unpublished results. Decapod estimates were drawn from Aguzzi 2287 et al. (2009a).

2289 Figure 3 Simulated temperature sections drawn in the x, y, and z directions in a 2290 water parcel sampled at different periods (x axis, time 1 to time 4) in two locations (y2291 axis; locations m and n) and at three depths (z axis, surface, middle waters, and bottom). 2292 The volume array characterises the conditions for a water parcel, with temperature 2293 slightly decreasing with time but strongly decreasing with depth from 21 to 13°C. This 2294 simulation was performed according to the deployment configuration of the Operational 2295 Observatory of the Catalan Sea (OOCS; http://www.ceab.csic.es/~oceans/) maintained 2296 by the Centre for Advanced Studies in Blanes (CEAB-CSIC) in the NW Mediterranean 2297 Sea (41° 39.7' N; 2° 54.6' E; operating depths between 50 and 200 m). The simulation 2298 was performed considering the environmental conditions found in late summer in the 2299 area. OOCS employs an oceanographic buoy moored at the sampling station between 2300 the continental shelf and the slope at the head of the Blanes submarine canyon. Apart 2301 from the meteorological conditions provided by the buoy's instrumentation, the upper 2302 100 m are monitored with multi-parametric submarine instruments measuring the 2303 temperature, salinity, current velocity and direction, visible light, turbidity, dissolved 2304 oxygen and chlorophyll fluorescence conditions every 30 m (Bahamon et al. 2011).

2305

2306 Figure 4 Protocol for the discrimination of clam shells between two species, 2307 Tapes decussatus and T. philippinarum, as carried out by Elliptic Fourier Analysis 2308 (EFA) as an example of automated video image classification on a specimen profile 2309 (adapted from Costa et al. 2008b, 2010). Dorsal digital images (A) are acquired in 2310 colour and subsequently transformed to grey-scale (B). Image binarisation (C) is 2311 required to detect the shell outline. A series of points are placed equidistantly (each 2°; 2312 D) in a Euclidean space starting from an arbitrary origin; a set of x, y coordinates is then 2313 computed for each point. The points are joined, and an outline is obtained (E; profile 2314 acquisition). The resulting harmonic function is fitted with an increasing number of ellipses and a matrix (F) comprising all individuals (in columns) with all fitted ellipses (in rows, from A to Z, each with its own four coefficients: a, b, c, and d) in the rows is obtained. Supervised or non-supervised multivariate analyses can be performed to screen the sample for any clustering of shapes. A mean shape outline (G) can be then obtained for each clustering to represent the within cluster morphological variation.

2320

2321 Figure 5 Example of automated processing for identification of the galatheid 2322 squat lobster, Munida quadrispina, in Saanich Inlet using the VENUS platform video 2323 camera based on Red-Green-Blue (RGB) colour coordinates values, Fourier Descriptors 2324 (FDs) and Scale invariant Feature Transform (SFIT), as adapted from Aguzzi et al. (2011a). For the colour content analysis, the original RGB image (A) is filtered and 2325 2326 corrected for the background by means of a top-hat filter (B). The next step is segmentation using a threshold value corresponding to the 95th percentile of Euclidean 2327 2328 distances (C). Then, the Euclidean distances between the Red and Green channel are 2329 calculated for each pixel of the filtered TopHat image (D); the FDs of the animal shape 2330 are then classified by PLSDA modelling, where red circles show the correctly classified 2331 objects, and the green profile shows the object outline (E). For SIFT analysis, the Red 2332 channel of the original RGB image is used at filtering (F). Then, features extraction 2333 with the SIFT algorithm is performed, and all of the extracted features are shown in 2334 green, while all of the features recognised as belonging to Munida are shown in red (G).

2335

Figure 6 Flowchart describing the processing of animal tracking and classification in automated video image analysis of digital videos obtained from the permanent deep-sea station of Sagami bay (1100 m depth, Central Japan) (adapted from Aguzzi et al. 2009c). The identification of moving animals (within a circle: an eelpout; a red crab, *Paralomis multispina*; and finally a snail, *Buccinum soyomaruae*) occurs at frame subtraction (A-B) and after image binarisation and area filtering (C; the removal of fouling allows identification of the animals' shapes). At profile extraction (D), shapes and RGB content coordinates can be automatically processed for classification in relation to morphological models of reference (i.e. from a library of images as training set) by means of supervised multivariate statistics.

2346

2347 Figure 7 Proposed conceptual scheme for automated and integrated time-series 2348 analysis of different biological (black) and habitat (grey) datasets obtained from 2349 permanent marine multiparametric stations. Arrows indicate the sequence of different 2350 steps involved in the data processing (A, data acquisition; B, data filtering; C, 2351 periodogram screening; D, data sub-set partitioning; E, waveform computing, and active 2352 phase duration assessment by the establishment of a daily threshold mean, i.e., Midline 2353 Estimating Statistic of Rhythm-MESOR; and finally, F, integrated waveform analysis). 2354 In the ultimate stage (F), the continuous thick horizontal line depicts the temporal 2355 duration of significant increases (i.e., the peak) in the fluctuation of biological (black) 2356 and habitat (grey) parameters. The dashed vertical rectangle in this analysis indicates the sensible time lapse (ΔT) when several habitat indicators (grey) significantly increase 2357 2358 their state of fluctuations prior to the consequent increase in the biological response 2359 (black) as a marker of a cause-effect relationship.

2360

Figure 8 World distribution maps depicting oceanic zones with different sunlight exposure where field studies detected the occurrence of variations in sampled species based on behavioural diel rhythms (A), along with tidal forces (B). Computed global mean sea surface photosynthetically available radiation (PAR, measured as $E_im^ ^2dy^{-1}$) was obtained by Aqua MODIS satellite measurements for the period between 4 July 2002 and 28 February 2011 (digital data provided by NASA). The approximate 2367 locations of most recent trawl studies at depths equal to or greater than 40 m where diel 2368 changes in benthic or demersal fishes were reported in this PAR global scenario. A 2369 depth threshold of 40 m was chosen because this is the average depth of worldwide light 2370 extinction (Lorenzen 1972). The present authors selected only studies on diel catches 2371 for fishes because these are usually more prone to automated tracking and classification 2372 given their size compared with other species, such as crustaceans (Aguzzi et al. 2011b). For zones where multiple studies were conducted, only the most recent one has been 2373 2374 considered. The approximate locations of presently existing cabled seafloor 2375 observatories (see Table 2 for codes) is presented in relation in relation to internal tidal 2376 forces (amplitude of the principal lunar semidiurnal constituent of tides, M2 at a period of 12.42 hours) and propagation (adapted from: http://www.aviso.oceanobs.com, by 2377 2378 Legos and CLS-Space Oceanography Division, distributed by Aviso/CNES). Cotidal 2379 lines depict the phase every 30 degrees originating at amphidromic points, where the 2380 tidal range is zero.

2381

2382 Figure 9 Proposed bathymetric network of permanent multiparametric video 2383 stations (A) on the shelf and slope (indicated as 1 and 2, respectively) and modelled 2384 visual count patterns over 24-h (B) for diel nektobenthic horizontal migrations (DHMs) 2385 and diel vertical benthopelagic migrations (DVMs). A bathymetric integrated network 2386 of video cameras is required to portray the role of both types of migrations in the 2387 temporal functioning of deep water and deep-sea ecosystems. DHM and DVM can be 2388 distinguished by comparing the numbers of tracked animals at equivalent times at 2389 stations 1 and 2. Arrows depict the diurnal (white) and nocturnal (black) direction of 2390 population displacement.

2392 Figure 10 Time series (between 1 April and 20 May 2006) of visually counted 2393 eelpouts (zoarcid fishes) prior and after to two major seismic events (on 20-21 and 30 2394 April and 2 May) as reported by the imaging system of the Sagami Bay multiparametric 2395 cabled observatory (1100 m depth; Hatsushima, Central Japan). Digital frames acquired 2396 at midday and midnight (the rhomb series), depict a consistent drop in the number of 2397 counted individuals prior to and after the first of these seismic events. This pattern was 2398 detected by adding a two-step moving average (thick black line). Seismic data were 2399 collected from the **USGS-NEIC** Earthquake Database 2400 (http://earthquake.usgs.gov/earthquakes/eqarchives/epic/epic rect.php; Last access: 2401 19/04/2011).

2402

2403 Figure 11 Organisation scheme for measuring capability at different levels of 2404 ecosystem complexity using sensors installed in multiparametric cabled observatories. 2405 Sensors for physical measurements (i.e., chemical, geological, and oceanographic) 2406 portray ecosystem dynamism at a very deterministic baseline level, accounting 2407 indirectly for events affecting life performances. At intermediate levels of 2408 environmental complexity, there are sensors that measure biological performance only 2409 in an indirect manner (e.g., chlorophyll sensors). A maximum level of environmental 2410 complexity is obtained through direct biological process measurements, which are taken 2411 by video-imaging sensors that directly allow species recognition and the counting of 2412 individuals within local populations (by automated video image analysis).





Aguzzi et al., Fig. 2





Aguzzi et al., Fig. 3





В

С

D

Α

RGB/FD based



Aguzzi et al., Fig. 5



Aguzzi et al., Fig. 6





Aguzzi et al., Fig. 8A






