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1           **Introduction to the Special Issue: Marine and**  
2           **Coastal GIS for Geomorphology, Habitat Mapping,**  
3           **and Marine Reserves**

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8           **Keywords** Marine GIS, acoustic remote sensing, marine geomorphology, benthic  
9           habitat, marine reserve/sanctuary/protected area, marine ecology, seafloor/seabed  
10          mapping, satellite remote sensing

11          This sixth special issue on Marine and Coastal Geographic Information Systems (M&CGIS)  
12          is the first to be based on an organized series of presentations at a conference, the 2008  
13          Association of American Geographers (AAG) Annual Meeting in Boston, Massachusetts,  
14          USA. The papers were selected and peer reviewed for publication in this special issue  
15          under the theme “Marine Geomorphology as a Determinant for Essential Life Habitat:  
16          An Ecosystem Management Approach to Planning for Marine Reserve Networks” (see  
17          presentations and resources online at <http://marinecoastalgis.net/aag08>). The sessions were  
18          cosponsored by the Coastal and Marine, Geographic Information Science and Systems,  
19          and Biogeography specialty groups of the AAG. The unifying goal of these sessions was  
20          to examine critically the growing body of data suggesting that the underlying geology and  
21          geomorphology of marine environments dictate the location of critical life habitat for a  
22          variety marine species. For example, it is becoming clearer that spawning aggregations of  
23          many species of commercially important reef fishes commonly occur at the windward edge  
24          of reef promontories that jut into deep water (e.g., Heyman et al. 2007; Heyman et al. 2005).  
25          As another example, seamounts serve as attractors for pelagic fishes and as stepping stones  
26          for transoceanic species dispersal (e.g., de Forges et al. 2000; Stocks et al. 2004). The broad  
27          implications of these findings suggest that geomorphology might be used as a proxy for (or  
        at least help to identify) critical life habitat for marine species and thus serve to advance

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28 the application of ecosystem-based management (EBM)<sup>1</sup> to the design of marine reserve  
29 networks (e.g., Lubchenco et al. 2007; Halpin et al. 2007; Halpern et al. 2008).

30 With only 5–10% of the world's seafloor mapped with the resolution of similar studies  
31 on land (Sandwell et al. 2003; Wright 2003), marine geomorphology still represents a  
32 persistent gap in our knowledge. Recent advances in technology have increased the array  
33 of available tools and the accuracy and speed at which the physical aspects of marine  
34 and coastal areas can be mapped. Sea bottom geomorphology and habitat information can  
35 be gained through hydrographic surveys with single beam and multibeam eco-sounders  
36 and sidescan sonar mounted on boats, submersibles, or remotely operated vehicles  
37 (ROVs). Satellite-based remote sensors (e.g., Landsat, QuickBird, and IKONOS) and  
38 aircraft-mounted sensors (e.g., LiDAR) have also been successfully used for seafloor  
39 mapping. Water column properties (e.g., salinity, temperature, current speed and direction,  
40 chlorophyll content, turbidity, nutrients) can be measured directly with boat-based or  
41 *in-situ* instruments or remotely with satellite-based sensors (e.g., Aqua, Terra, Seawifs, and  
42 Modis). To go along with the physical information described above, ecological information  
43 (e.g., species composition, abundance) almost always needs to be evaluated by direct  
44 observations and/or with photography and video acquired by ROVs or submersibles. These  
45 data are highly variable in space and time so characterization requires multiple observations  
46 over various seasons and times. Reliable ecological characterizations therefore can be  
47 prohibitively expensive and time consuming and require re-measurement for monitoring. A  
48 major goal of our symposium and this focus issue is to illustrate state-of-the-art examples  
49 of how researchers have classified, integrated, and analyzed physical and ecological data  
50 sources using various algorithmic approaches in M&CGIS to reveal geomorphology as  
51 a proxy for habitat. Given the paucity of available data marine habitat data, and the need  
52 for rapid and large expansion in marine reserves networks coverage, geomorphological  
53 habitat proxies can assist managers in making timely recommendations for high-priority,  
54 critical habitats for inclusion within marine reserves.

55 Analyses of these data provide answers for three fundamental types of questions as  
56 follows:

- 57 1. What are the locations and shapes of benthic physical forms (e.g., platforms,  
58 seamounts, ledges, trenches, or abrupt changes in slope or geomorphic features),  
59 under what conditions (e.g., complexity of seafloor, levels of temperature, salinity,  
60 characteristics of bottom current regime), and what are the associated species and  
61 their uses of these habitats (e.g., feeding grounds, spawning aggregation sites,  
62 nursery habitats), as indicated by species composition or abundance over time?
- 63 2. What should be the habitat classification categories for a particular region,  
64 especially in relation to the adjacent coastal ecology? In reality, how accurate are  
65 the classifications derived by quantitative algorithms with regard to where certain  
66 species are colonizing? Should these quantitative approaches be standardized  
67 somehow?
- 68 3. Which habitats and locations are “biological hotspots” and/or areas of essential life  
69 habitat for multiple species (e.g., areas of high biodiversity, areas of high marine  
70 productivity such as upwelling areas, spawning aggregation sites, important feeding

<sup>1</sup>Defined by Feeley et al. (2008) as applying “current scientific understanding of ecosystem structure and processes to achieve more coordinated and effective management of society's multiple uses of and interests in the services provided by the ecosystem. EBM does not prescribe a particular outcome; instead, it acknowledge that changing the ecosystem can also change the services it provides.”

*Introduction to the Special Issue*

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71 grounds, nursery or juvenile habitat), and what should be the resulting decisions  
72 for monitoring, management, and ultimately conservation?

73 Articles in this special issue are divided into two main groups, the first focusing  
74 essentially on the geomorphology, habitat, and the necessary hydrographic surveying of  
75 study areas, while the second group is made up of methodological papers focusing on tools  
76 or techniques for classifying or merging data, while also interpreting the geomorphology.  
77 The first group of papers begins with a study by Kobara and Heyman that uses marine  
78 geomorphology as a predictor of the locations of spawning aggregation sites of Nassau  
79 grouper in the Cayman Islands. The paper illustrates that spawning aggregation sites are  
80 concentrated at the shelf edges and seaward-most tips of similarly shaped reef promontories  
81 that jut into deep water. Wedding and Friedlander use GIS analysis of bathymetric LiDAR  
82 (light detection and ranging) data to assess four marine protected areas (MPAs) in Hawaii  
83 to determine which geomorphic measures demonstrate important relationships with reef  
84 fish assemblage structure, and hence would ultimately serve as the best ecological criteria  
85 to guide future MPA design. In spite of variations in habitat complexity between sites and  
86 the important relationships between fish distribution and various LiDAR-derived habitat  
87 metrics, protection from fishing is found to be the best predictor of fish biomass at all sites.  
88 Kracker et al. moves the realm of benthic habitat mapping up into the water column by  
89 using hydroacoustic fisheries surveys and subsequent GIS analyses to assess patterns of  
90 fish biomass in relation to bottom habitat in the Gray's Reef National Marine Sanctuary on  
91 the inner continental shelf of Georgia. Their analysis illustrates that correlations between  
92 biota and habitat are better near the seafloor (e.g., proximity to ledges is a good predictor of  
93 high biomass in near-bottom regions) than in the water column. Yet overall, the techniques  
94 provide an efficient, nondestructive way to quantify fish biomass and associated habitats  
95 and will be applicable in other locations.

96 In the second group of papers largely on methodology, Su et al. move the emphasis  
97 into the realm of satellite remote sensing by presenting a method for deriving nearshore  
98 bathymetry from IKONOS multispectral satellite imagery using a nonlinear inversion model  
99 (the Levenberg-Marquardt algorithm) but with a new, automated method for calibrating the  
100 parameters in the model. Their analysis confirms that the derived bathymetry is slightly more  
101 accurate and stable for deeper benthic habitats than bathymetry derived from conventional  
102 log-linear models. Similarly, Hogrefe et al. present methods for deriving accurate nearshore  
103 bathymetry from IKONOS imagery but through a different approach of gauging the  
104 relative attenuation of blue and green spectral radiation (the Lyzenga method). They then  
105 combine that derived bathymetry with a 10-m terrestrial digital elevation model to create  
106 a seamless coastal terrain model of the topography and bathymetry of Tutuila, American  
107 Samoa, out to a surrounding depth of ~250 m. The results have positive implications  
108 for defining marine-terrestrial units (MTUs) that span the land-sea interface. This will in  
109 turn enable quantitative correlations between upland land use practices and the vitality  
110 of downstream reef communities, as measured by coral and fish species composition and  
111 diversity. Erdey uses high-resolution multibeam bathymetry from the Point Reyes National  
112 Seashore, California, as input to the bathymetric position index algorithm to create initial  
113 classifications of seafloor geomorphology. In concert with this, she analyzes backscatter  
114 intensity with multivariate statistical tools to delineate sediment textural classes. All methods  
115 are encapsulated into a new toolbox using the capabilities of ArcGIS ModelBuilder.  
116 Iampietro et al. also use high-resolution multibeam bathymetry and backscatter data, along  
117 with submersible and remotely-operated vehicle (ROV) video data at Cordell Bank National  
118 Marine Sanctuary (CBNMS) and the Del Monte shale beds of Monterey Bay, California, to

119 produce preliminary species-specific habitat suitability models for eight rockfish species.  
120 They use a generalized linear model (GLM) approach to produce the habitat classes, along  
121 with supervised texture classification from backscatter mosaics. They find that the GLM  
122 is reasonably portable from one location to another; that is, the model for *S. flavidus*  
123 (yellowtail rockfish) generated at CBNMS is at least as efficient at predicting yellowtail  
124 rockfish distributions at Del Monte. This could still fail in other circumstances where  
125 depth holds a strong inverse correlation with the probability of occurrence, or there is a  
126 failure to incorporate many other factors such as substrate type, temperature, currents, food  
127 availability, predation, and recruitment.

128 Studies such as Erdey and Iampietro et al. beg the question of whether researchers are  
129 ready to move toward a *standardization* of algorithmic seafloor classification approaches.  
130 It is useful here to distinguish between classifications that are visual, as opposed to those  
131 which are algorithmic. Visual classification relies on local expert knowledge to delineate  
132 distinct seafloor features and subsequent classifications of geomorphology by mere visual  
133 inspection of the data, by hand and/or with computerized drawing tools that work on  
134 an underlain image of a base map. This mode of classification therefore possesses high  
135 information content, but may also be subjective, laborious, expensive (if costs for human  
136 labor are a factor), and with resolution limited by time or patience. Algorithmic approaches  
137 are almost always quantitative, usually automatic or at least semi-automatic, and allow the  
138 user to refine the classification at certain stages in the process based on visual observation.  
139 This mode of classification, while subject to artifacts, is usually more repeatable, less  
140 expensive, and with resolution limited only by the source data.

141 In further examining various algorithmic approaches, GIS analyses involving  
142 quantitative assessment of the shape of the seafloor for habitat characterization have  
143 traditionally included slope and aspect of terrain, but also the more rigorous approach  
144 of topographic position index (TPI), which measures where a point is in the overall  
145 landscape/seascape in order to identify features such as ridges, canyons, slopes, midslopes,  
146 etc., and at whatever scale a topographic or bathymetric grid will support. This approach  
147 comes from the field of landscape ecology (see the review in Bridgewater 1993), based in  
148 part on the ecological land unit/landscape position algorithms of Fells (1995), Anderson  
149 et al. (1998), Guisan et al. (1999), Jones et al. (2000), and then Weiss (2001). Iampietro and  
150 Kvittek (2002) have championed TPI for the seafloor, Wright et al. (2005) and Lundblad  
151 et al. (2006) have extended it a bit further (calling it bathymetric position index or BPI) and  
152 codifying it as ArcGIS extension, while Lanier et al. (2007) have introduced an important  
153 variation on it (the surface interpretation method or SIM), taking further advantage of the  
154 latest 2.5-dimensional capabilities of ArcGIS.

155 Another important parameter that is calculated is seafloor roughness (i.e., the  
156 bumpiness of the seafloor, especially in terms of how convoluted and complex a surface is,  
157 and over cartographic map scales that are larger than TPI/BPI). Jenness (2003, 2004)  
158 developed a method for calculating a type of roughness called “rugosity,” which is  
159 essentially the ratio of study region’s surface area to planar surface area. Ardron (2002)  
160 has taken a slightly different approach where flow direction (the number of facets in a  
161 grid) and relief variability are combined to produce a “bottom complexity.” Sampson et al.  
162 (2008) have recently pointed to still another variation as developed by Sappington et al.  
163 (2007), which calculates “ruggedness” by measuring the dispersion of vectors orthogonal  
164 to a terrain surface. This method is much less correlated with (and hence distorted by) slope  
165 than the rugosity algorithm.

166 There is also a range of approaches for ecological habitat modeling involving biological  
167 data in concert with bathymetry, and extending from the seafloor (using depth, distance to

168 shelf break or shore) up into the water column. These include the aforementioned GLM,  
169 generalized additive modeling or GAM, the classification and regression tree or CART,  
170 environmental envelope models, canonical correspondence analysis, and Bayesian models,  
171 as summarized by Guisan and Zimmerman (2000), adapted for the marine environment by  
172 Redfern et al. (2006) and codified in GIS software by Best et al. (2006). It seems that it is  
173 always the biology that will provide the greatest challenge.

174 In assessing these many algorithmic approaches, standardization via scale (regional  
175 size of a hydrographic survey) as well as by resolution of the data (size of grid cells)  
176 will be key. While we are still far from a *standardization* of algorithmic approaches,  
177 our choices of algorithms should be governed by a detailed knowledge of the species  
178 of interest and the scale at which that organisms perceive the environment. This can be  
179 difficult to track when analyzing multiple species in a marine reserve. Further, while  
180 most researchers are using similar sonar systems for gathering multibeam bathymetry  
181 and backscatter (e.g., Reson, Kongsberg-Simrad, Acoustic Marine Systems, GeoSwath)  
182 the processing procedures for these data are not altogether standardized either (especially  
183 for backscatter). In addition, the level of detail in classification is going to depend on  
184 whether one also has access to satellite data and subsurface data, in addition to the standard  
185 acoustics and the groundtruthing visuals from ROV, submersible, or SCUBA, along with  
186 the associated uncertainties in mapping units (e.g., Halley and Jordan 2008). Differences  
187 in classifications for shallow versus deepwater regions will also be quite significant, even  
188 within the same study area (as pointed out by Lundblad et al. 2006 and Wilson et al.  
189 2007).

190 These issues were discussed at length during a panel session held as part of the 2008  
191 AAG presentations spawning the submissions to this special issue. Participants reported that  
192 efforts in Europe (e.g., Mapping European Seabed Habitats or MESH, a major European  
193 Union-funded initiative to harmonize mapping approaches and collate habitat maps in  
194 NW Europe, <http://www.searchmesh.net>) and Australia (e.g., Geoscience Australia; Heap  
195 2006) are moving towards standardizing a classification approach. In the U.S., the Coastal  
196 Marine Ecological Classification Standard (CMECS) managed by NOAA and NatureServe  
197 (Madden et al. 2005, Madden and Grossman 2008) will be important to consider when  
198 identifying marine ecoregions or when mapping from “ridge to reef” (i.e., the connectivity  
199 between upland watersheds, intertidal zones, and shallow coastal areas including reefs).  
200 This is where offshore classification categories must be integrated with those for wetland  
201 and intertidal regions (e.g., Heyman and Kjerfve 1999).

202 The papers in this special issue shed light on these issues and may lay the groundwork  
203 for the future development of a standard decision-tree or matrix of classification approaches,  
204 governed by map scale and species. At some point a standard classification dictionary for  
205 various settings (e.g., tropical coral reef substrate vs. continental shelf shale beds, etc., deep  
206 vs. shallow) and accompanying generic bathymetric, backscatter and biological datasets  
207 might be considered as tools for all to work with when testing these various approaches  
208 and the GIS extensions that encode them. Further dialogue will be welcomed as to what  
209 standard features should appear on a benthic habitat map, not just to aid scientists, but  
210 to communicate effectively to managers and policy-making stakeholders in the process of  
211 designing or monitoring a marine reserve.

212 We would like to point out parallel efforts that relate very nicely to the body of  
213 work presented in this special issue. The *Marine Geodesy* papers here of course focus  
214 on marine GIS and remote sensing aspects of benthic habitat mapping, but a special  
215 issue for *The Professional Geographer* (Heyman and Wright, submitted) draws upon the  
216 same organized sessions at AAG mentioned at the beginning of this article, with papers

217 not only on the physical and resource geography aspects of benthic habitat, but also on  
218 marine policy. A special issue of the journal *Geomatica* (Devillers and Gillespie 2008) is  
219 devoted to marine geomatics (“geomatics” being the equivalent term in Canada for GIS,  
220 remote sensing, geodesy, and photogrammetry), and featuring papers on the acquisition,  
221 processing, management, and dissemination of data from the seafloor, the subsurface,  
222 the water column (including pelagic biomass), and the sea surface. Mapping of the Arctic  
223 seafloor and the Canadian continental shelf, including benthic habitat, are additional themes,  
224 as well as the emergence of ocean sensor networks and ocean observatories. Interested  
225 readers should also take note of the annual GeoHab (marine **Geological** and biological  
226 **Habitat** mapping) conference (see <http://geohab.org>), and the recent monograph based  
227 on papers presented at this conference since its inception in 2001 (Todd and Greene  
228 2008).

229 To conclude, we would like to thank all of the contributors to this special issue of  
230 *Marine Geodesy* for their enthusiasm and skill in authoring these articles. We thank the  
231 many reviewers for their thoughtful insights and care in commenting on and improving  
232 all of the manuscripts. Finally, we thank Editor-in-Chief Dr. Rongxing (Ron) Li for his  
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234 assistance in publishing this one.

## 235 **References**

- 236 Anderson, M. G., M. D. Merrill, and F. B. Biasi. 1998. *Connecticut River watershed analysis: Eco-*  
237 *logical communities and neo-tropical migratory birds*. Boston, MA: The Nature Conservancy.  
238 Accessed June 4, 2008, from <http://conserveonline.org/docs/2001/03/ctrivsumm.doc>
- 239 Ardron, J. 2002. A GIS recipe for determining benthic complexity: An indicator of species richness.  
240 In *Marine geography: GIS for the oceans and seas*, J. Breman (ed.), p. 169–175. Redlands, CA:  
241 ESRI Press, CA.
- 242 Best, B. D., D. L. Urban, P. N. Halpin, and S. S. Qian, 2006. Linking multivariate habitat modeling into  
243 ArcGIS with the ArcRstats toolbox. *Proceedings of the Society for Conservation Biology, 20th*  
244 *Annual Meeting*, San Jose, CA (code accessible online from the Marine Geospatial Ecology Lab  
245 of P. Halpin. Accessed June 4, 2008, from <http://www.nicholas.duke.edu/geospatial/software> or  
246 <http://mgel.env.duke.edu/tools>).
- 247 Bridgewater, P. B. 1993. Landscape ecology, geographic information systems and nature conservation.  
248 In *Landscape ecology and geographic information systems*, R. Haines-Young, D. R. Green and  
249 S. Cousins (eds.), pp. 23–36. London: Taylor and Francis.
- 250 de Forges, B. R., J. A. Koslow, and G. C. B. Poore. 2000. Diversity and endemism of the benthic  
251 seamount fauna in the southwest Pacific. *Nature* 405:944–947.
- 252 Devillers, R. and R. Gillespie. In press. Geomatics for the blue planet. *Geomatica* 62(4).
- 253 Feeley, M. H., S. C. Pantoja, T. Agardy, J. C. Castilla, S. C. Farber, I. V. Hewawasam, J. Ibrahim,  
254 J. Lubchenko, B. J. McCay, N. Muthiga, S. B. Olsen, S. Sathyendranath, M. P. Sissenwine, D.  
255 O. Suman, and G. Tamayo. 2008. *Increasing capacity for stewardship of oceans and coasts: A*  
256 *priority for the 21st century*. Washington, DC: Ocean Studies Board, National Academies Press,  
257 p. 13.
- 258 Fels, J. and R. Zobel. 1995. *Landscape position and classified landtype mapping for the statewide*  
259 *DRASTIC mapping project*. Raleigh, North Carolina: North Carolina State University. Technical  
260 Report VEL.95.1 to the North Carolina Department of Environment, Health, and Natural  
261 Resources, Division of Environmental Management.
- 262 Guisan, A., S. B. Weiss, and A. D. Weiss. 1999. GLM versus CCA spatial modeling of plant species  
263 distribution. *Plant Ecology* 143:107–122.
- 264 Guisan, A. and N. E. Zimmerman. 2000. Predictive habitat distribution models in ecology. *Ecological*  
265 *Modelling* 135:147–186.

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7

- 266 Halley, V. and A. Jordan. 2008. Addressing spatial uncertainty in mapping southern Australian  
267 coastal seabed habitats. In *Mapping the seafloor for habitat characterization*, B. J. Todd and  
268 H. G. Greene (eds.), pp. 157–170. St. John's, Newfoundland, Canada: Geological Association  
269 of Canada, Geological Association of Canada Special Paper 47.
- 270 Halpin, P., P. Crist, and S. Carr. 2007. The coastal-marine manager's toolkit for ecosystem-based  
271 management. *Proceedings of the 27th Annual ESRI User Conference*, San Diego, CA.
- 272 Halpern, B. S., K. L. McLeod, A. A. Rosenberg, and L. B. Crowder. 2008. Managing for cumulative  
273 impacts in ecosystem-based management through ocean zoning. *Ocean & Coastal Management*  
274 51(3):203–211.
- 275 Heap, A. 2006. Classifying Australia's seascapes for marine conservation. *AusGeo News*, 84.  
276 Retrieved June 4, 2008, from <http://www.ga.gov.au/ausgeonews/ausgeonews200612/conservation.jsp>.
- 278 Heyman, W. and B. Kjerfve. 1999. Hydrological and oceanographic considerations for integrated  
279 coastal zone management in southern Belize. *Environmental Management* 24(2):229–245.
- 280 Heyman, W. D., B. Kjerfve, K. L. Rhodes, R. T. Graham, and L. Garbutt. 2005. Cubera snapper,  
281 *Lutjanus cyanopterus*, spawning aggregations on the Belize Barrier Reef over a six year period.  
282 *Journal of Fish Biology* 67(1):83–101.
- 283 Heyman, W. D., J.-L. B. Ecochard, and F. Biasi. 2007. Low-cost bathymetric mapping for tropical  
284 marine conservation—a focus on reef fish spawning aggregation sites. *Marine Geodesy* 30(1):  
285 37–50.
- 286 Heyman, W. D. and D. J. Wright. In press. Marine geomorphology in the design of marine reserve  
287 networks. *The Professional Geographer*.
- 288 Iampietro, P., and R. Kvitek. 2002. Quantitative seafloor habitat classification using GIS terrain  
289 analysis: Effects of data density, resolution, and scale. *Proceedings of the 22nd Annual ESRI*  
290 *User Conference*, San Diego, CA.
- 291 Jenness, J. S. 2003. *Grid surface areas: Surface area and ratios from elevation grids*. Jenness Enter-  
292 prises: ArcView Extensions. Retrieved June 4, 2008, from [http://www.jennessent.com/arcview/arcview\\_extensions.htm](http://www.jennessent.com/arcview/arcview_extensions.htm).
- 294 Jenness, J. S. 2004. Calculating landscape surface area from digital elevation models. *Wildlife Society*  
295 *Bulletin* 32(3):829–839.
- 296 Jones, K. B., D. T. Heggem, T. G. Wade, A. C. Neale, D. W. Ebert, M. S. Nash, M. H. Mehaffey, K.  
297 A. Hermann, A. R. Selle, S. Augustine, I. A. Goodman, J. Pedersen, D. Bolgrien, J. M. Viger, D.  
298 Chiang, C. J. Lin, Y. Zhong, J. Baker, and R. D. Van Remortel. 2000. Assessing landscape  
299 conditions relative to water resources in the western United States: A strategic approach.  
300 *Environmental Monitoring and Assessment* 64:227–245.
- 301 Lanier, A., C. Romsos, and C. Goldfinger. 2007. Seafloor habitat mapping on the Oregon continental  
302 margin: A spatially nested GIS approach to mapping scale, mapping methods, and accuracy  
303 quantification. *Marine Geodesy* 30(1/2):51–76.
- 304 Lubchenco, J., S. Gaines, K. Grorud-Colvert, S. Airame, S. Palumbi, R. Warner, and B. Simler  
305 Smith. 2007. *The science of marine reserves* (2nd Edition, U.S. version), 22 pp. Corvallis, OR:  
306 Partnership for the Interdisciplinary Studies of Coastal Oceans, <http://www.piscoweb.org>.
- 307 Lundblad, E., D. J. Wright, J. Miller, E. M. Larkin, R. Rinehart, T. Battista, S. M. Anderson, D. F.  
308 Naar, and B. T. Donahue. 2006. A benthic terrain classification scheme for American Samoa.  
309 *Marine Geodesy* 29(2):89–111.
- 310 Madden, C. J., D. H. Grossman, and K. L. Goodin (eds.). 2005. *Coastal and marine systems of*  
311 *North America: Framework for an ecological classification standard: Version II*. Arlington, VA:  
312 NatureServe, 60 pp.
- 313 Madden, C. J. and D. H. Grossman. 2008. A framework for a coastal/marine ecological classification  
314 standard (CMECS). In *Mapping the seafloor for habitat characterization*, B. J. Todd and  
315 H. G. Greene (eds.), pp. 185–210. St. John's, Newfoundland, Canada: Geological Association  
316 of Canada, Geological Association of Canada Special Paper 47.
- 317 Pennisi, E. 2002. Survey confirms coral reefs are in peril. *Science* 297:1622–1623.

- 318 Redfern, J. V., M. C. Ferguson, E. A. Becker, K. D. Hyrenbach, C. Good, J. Barlow, K. Kaschner,  
319 M. F. Baumgartner, K. A. Forney, L. T. Ballance, P. Fauchald, P. Halpin, T. Hamazaki, A. J.  
320 Pershing, S. S. Qian, A. Read, S. B. Reilly, L. Torres, and F. Werner. 2006. Techniques for  
321 cetacean-habitat modeling. *Marine Ecology Progress Series* 310:271–295.
- 322 Sampson, D. W., A. R. Wilbur, and S. D. Ackerman. 2008. Mapping seafloor surficial geologic habitat  
323 in Massachusetts state waters. *Abstracts of the Association of American Geographers Annual*  
324 *Meeting*, Boston, Massachusetts. Retrieved June 4, 2008, from <http://marinecoastalgis.net/aag08>
- 325 Sandwell, D., S. Gille, J. A. Orcutt, and W. Smith. 2003. Bathymetry from space is now possible.  
326 *Eos, Transactions of the American Geophysical Union* 84(5): 37, 44.
- 327 Sappington, J. M., K. M. Longshore, and D. B. Thompson. 2007. Quantifying landscape ruggedness  
328 for animal habitat analysis: A case study using bighorn sheep in the Mojave Desert. *Journal of*  
329 *Wildlife Management* 71(5):1419–1427.
- 330 Stocks, K. I., G. W. Boehlert, and J. F. Dower. 2004. Towards an international field program  
331 on seamounts within the Census of Marine Life. *Archive of Fishery and Marine Research*  
332 51(1–3):320–327.
- 333 Todd, B. J. and Greene, H. G. (eds.). 2008. *Mapping the seafloor for habitat characterization*. Geo-  
334 logical Association of Canada Special Paper 47, St. John's, Newfoundland, Canada:Geological  
335 Association of Canada, 519 pp.
- 336 Weiss, A. D. 2001. Topographic positions and landforms analysis (Map Gallery Poster). *Proceedings*  
337 *of the 21st Annual ESRI User Conference*, San Diego, CA.
- 338 Wilson, M. F. J., B. O'Connell, C. Brown, J. C. Guinan, and A. J. Grehan. 2007. Multiscale terrain  
339 analysis of multibeam bathymetry data for habitat mapping on the continental slope. *Marine*  
340 *Geodesy* 30:3–35.
- 341 Wright, D. J. 2003. Introduction. In *Undersea with GIS*, D. J. Wright (ed.), p. xiii–xvi. Redlands,  
342 CA: ESRI Press.
- 343 Wright, D. J., E. R. Lundblad, E. M. Larkin, R. W. Rinehart, J. Murphy, L. Cary-Kothera, and K.  
344 Draganov. 2005. *ArcGIS Benthic Terrain Modeler*, Corvallis, OR: Oregon State University,  
345 Davey Jones Locker Seafloor Mapping/Marine GIS Laboratory. Retrieved June 4, 2008, from  
346 <http://www.csc.noaa.gov/products/btm/>

#### 347 **Web Sites (last accessed July 28, 2008)**

- 348 AAG 2008 Special Sessions, *Marine Geomorphology as a Determinant for Essential Life Habitat:*  
349 *An Ecosystem Management Approach to Planning for Marine Reserve Networks* – [http://](http://marinecoastalgis.net/aag08)  
350 [marinecoastalgis.net/aag08](http://marinecoastalgis.net/aag08)
- 351 GeoHab Conference Series (marine **G**eological and biological **H**abitat mapping) – <http://geohab.org>
- 352 MESH, Mapping European Seabed Habitats – <http://search.mesh.net>