Using revisitation data for predicting changes in ecosystem functioning across protected and non-protected coastal dunes

Greta La Bella[1], Alicia T.R. Acosta [1], Tommaso Jucker[2], Manuele Bazzichetto [3], Marco Andrello [4,5], Marta Gaia Sperandii [1], and Marta Carboni[1]

[1] Department of Science, Roma Tre University, Italy;
[2] School of Biological Sciences, University of Bristol, UK;
[3] Department of Biological, Geological and Environmental Sciences, University of Bologna, Italy;
[4] Institute of Anthropic impacts and Sustainability in the marine environment, National Research Council, CNR-IAS, via della vasca navale 79, 00146, Rome, Italy
[5] National Biodiversity Future Center, Palermo, Italy

CORRESPONDING: Greta La Bella, greta.labella@uniroma3.it

ARTICLE IMPACT STATEMENT

Hindcasting ecosystem functions from historical vegetation data enables monitoring protection efficacy in preserving ecosystem functionality

KEYWORDS: BEF; biodiversity; ecosystem functioning; Resurvey; Natura 2000; protected areas; temporal changes; trait-based approach

AUTHOR CONTRIBUTIONS

GLB, ATRA, and MC conceived and designed the study. GLB collected the data, led the statistical analysis, and wrote the first draft of the manuscript. All authors contributed to results interpretation, made significant contributions to the manuscript, and approved it for publication.

ACKNOWLEDGMENTS

This work was supported by the Grant of Excellence Departments 2018–2022 and the Grant of Excellence Departments 2023–2026, MIUR Italy. GLB, ATRA, MA and MC acknowledge the support of the project “National Biodiversity Future Center – NBFC" funded under the National Recovery and Resilience Plan (NRRP), Mission 4 Component 2 Investment 1.4 - Call for tender No. 3138 of 16 December 2021, funded by the European Union – NextGenerationEU. Project code CN_00000033, Concession Decree No. 1034 of 17 June 2022 adopted by the Italian Ministry of University and Research, CUP F83C22000730006.

Using revisitation data for predicting changes in ecosystem functioning across protected and non-protected coastal dunes

ABSTRACT

Protected areas are generally designed to conserve biodiversity. However, how well they also contribute to maintaining ecosystem functions that plant diversity supports has rarely been explicitly tested, often due to the lack of historical ecosystem function data. Here, we used a trait-based approach to reconstruct past ecosystem functioning and examine its change over the last 15 years in protected and unprotected coastal dune ecosystems, checking where functions remain stable over time. First, we resurveyed vegetation in quasi-permanent plots and measured in the present several ecosystem functions related to biomass production, carbon, water, nutrient cycling, erosion control, and invasion resistance across six coastal dune sites in Central Italy. Second, using these data, we quantified Biodiversity-Ecosystem Function (BEF) relationships and employed them to hindcast past ecosystem functions based on historical vegetation surveys. Finally, as a case study, we applied this method to assess temporal changes in ecosystem functioning under three protection regimes: national protected areas (i.e. strict protection), Natura 2000 sites (loose protection), and non-protected areas. Biomass production, carbon, and water regulation increased over time in non-protected areas, likely due to an expansion of ruderal and non-native species, that are usually more productive. Within Natura 2000 sites, communities showed a decrease in erosion control and invasion resistance, due to the loss of important dune-building species and the spread of non-natives. Only within national protected areas, ecosystem functions did not undergo significant temporal changes, and invasion resistance even increased. Our results suggest that ecosystem functioning remained stable over time only in areas under strict protection. More broadly, our study demonstrates the potential for using revisitation data in combination with locally estimated BEF relationships to hindcast past ecosystem functioning, providing a valuable tool for monitoring long-term functional changes in response to conservation measures.

INTRODUCTION

Monitoring ecosystem functions is critical for assessing the health and resilience of ecosystems and ensuring their ability to provide essential services to human societies. The post-2020 Global Biodiversity Framework (GBF) of the Convention on Biological Diversity (CBD) emphasizes not only the need to halt biodiversity loss but also to maintain and enhance ecosystem functions and services, as reflected in its goals and targets (CBD, 2022). Similarly,

the United Nations Sustainable Development Goals (SDGs) highlight the importance of sustaining ecosystem integrity to support human well-being (UN, 2015). Despite these global commitments, measuring ecosystem functions presents significant challenges due to their complex, multi-dimensional nature and the difficulty of obtaining long-term data. Many ecosystem processes operate at large spatial and temporal scales, requiring indirect methods or models to infer past dynamics. This limitation complicates the evaluation of conservation strategies, such as protected areas and restoration efforts, which should aim to preserve not only species diversity but also the ecological processes that sustain ecosystems. Without robust estimates of past ecosystem functions, it remains difficult to determine whether conservation actions have successfully maintained or restored functional integrity.

The relationship between Biodiversity and Ecosystem Functioning (BEF) has been widely documented across natural ecosystems, although its strength and direction can vary depending on the local environmental conditions and community composition (Hagan et al., 2021; Hanisch et al., 2020; La Bella et al., 2024; Ratcliffe et al., 2017; Valencia et al., 2015; van der Plas, 2019). Given this relationship, shifts in biodiversity can significantly alter ecosystem processes and functions. Nevertheless, while biodiversity changes are increasingly being reported, only few studies have quantified the extent and direction in which ecosystem functions are changing (Brauman et al., 2020). As ecosystem functions largely underpin vital services provided to people (Garland et al., 2021; MEA, 2005), monitoring changes in ecosystem functioning is of crucial importance for human well-being.

To prevent further biodiversity loss and safeguard the last remaining intact natural areas, European countries have implemented national or international conservation measures (Heslenfeld et al., 2008), including the establishment of protected areas (CBD, 2022; UN, 2015). By conserving biodiversity protected areas also offer the opportunity to preserve the variety of ecosystem functions and processes that biodiversity supports. However, since most protected areas are not explicitly managed to maintain ecosystem functionality, their ability to counteract changes in ecosystem functions remains uncertain.

Overall, studies assessing temporal changes in ecosystem functioning in real-world ecosystems are scarce for several reasons (DeCock et al., 2023; Waldén et al., 2023; Williams et al., 2005). First, the interest in quantifying ecosystem functions in natural systems has only surged in recent years (Garland et al., 2021; van der Plas, 2019). Although systematic data collection has been established in some long-term ecological monitoring networks (e.g. the Biodiversity Exploratories; Fischer et al., 2010), these remain relatively localised. As a result, comparison with past conditions are still limited (Kremen & Ostfeld, 2005; Meyer et al., 2015).

Second, unlike biodiversity metrics, measuring ecosystem functions is challenging and methods are still highly debated (Byrnes et al., 2014; Garland et al., 2021; Manning et al., 2018; Meyer et al., 2015). In fact, while some ecosystem functions can be quantified through direct, well-accepted indicators (e.g. biomass production), for other functions choosing the most appropriate indicator is less clear (e.g. water regulation) (Farnsworth et al., 2017; Garland et al., 2021; Jax, 2005). Moreover, measuring ecosystem functions is usually resource-intensive and time-consuming, requiring technical expertise and equipment which were until recently not widely accessible (Meyer et al., 2015). Consequently, reconstructing past data in order to track temporal changes in ecosystem functioning requires innovative solutions.

Plant functional traits are increasingly recognised as reliable predictors of several ecosystem functions and processes, such as biomass production, carbon sequestration, and nutrient cycling (de Bello et al., 2010; Díaz et al., 2007; Funk et al., 2017; Lavorel & Garnier, 2002). As a result, plant traits are now routinely used in BEF research together with taxonomic diversity for quantifying the effect of species and communities on ecosystems. For example, species-rich communities dominated by fast-growing acquisitive plant species have been frequently associated to fast biogeochemical cycles (Cornwell et al., 2008; Funk et al., 2017; Garnier et al., 2015; Lavorel & Garnier, 2002). Given this relationship, several trait-based approaches have been proposed to predict ecosystem functioning beyond the observed data, e.g. to larger datasets (Bodenhausen et al., 2023; Chen et al., 2023), or for predicting ecosystem functions under future scenarios of environmental change (Klumpp & Soussana, 2009; Madani et al., 2018; Quétier et al., 2007). BEF relationships can thus potentially also be used to hindcast past functioning levels using historical data, with even less uncertainty compared to e.g. forecasts based on many possible future scenarios. This potential application provides a chance to compare past and present ecosystem functions and, thus, to assess how ecosystems have responded to the last decades of global changes. As such, here we propose predictive modelling to reconstruct past ecosystem functions based on historical vegetation composition in order to track temporal changes in functioning, particularly in highly endangered ecosystems.

Coastal dunes are very peculiar ecosystems, rich in highly specialized flora with an outstanding conservation value (Acosta et al., 2009; Van der Maarel, 2003). They also provide multiple essential services to society, e.g. coastal protection against the sea (Liquete et al., 2013) and recreational activities (Drius et al., 2019a). Despite their valuable features, coastal dunes are among the most threatened ecosystems in Europe, especially in the Mediterranean basin (Janssen et al., 2016; Prisco et al., 2020). High anthropogenic impact related to urban expansion and tourism as well as the spread of invasive alien species, have led to extensive

degradation of coastal habitats and to biodiversity declines (Carboni et al., 2010; Defeo et al., 2009; Drius, Bongiorni, et al., 2019; Malavasi et al., 2016; Schlacher et al., 2007). Still, the consequences of habitat degradation for the functioning of coastal dune ecosystems have been rarely quantified.

In this study, we tackled two key objectives: (1) testing the predictive ability of vegetation resurveys for hindcasting ecosystem functioning and assessing possible gains and losses in functions; (2) exploring the performance of different types of protected areas in maintaining the level of ecosystem functioning stable over time. Specifically, we developed an approach that integrates vegetation resurvey data with biodiversity–ecosystem functioning (BEF) models to explore temporal trends in key ecosystem functions. We then applied this framework (Fig. 1) to coastal dune ecosystems, comparing temporal changes in ecosystem functioning across three protection regimes: nationally designated strictly protected areas, sites belonging to the European Natura 2000 network (protected areas that allow a sustainable use of natural resources), and non-protected areas.

METHODS

Study area and protection status

The research was conducted on coastal dune ecosystems situated along the Tyrrhenian coast of Central Italy, specifically in the region Lazio, consisting of approximately 250 km of coast. This geographical area falls within the Mediterranean climatic region and consists of Holocene sandy dunes forming a narrow stretch along the shoreline (Acosta et al., 2009).

For determining the protection status, we acquired spatial data related to the World Database on Protected Areas (WDPA) and World Database on Other Effective Area-based Conservation Measures (WD-OECM), from the "Protected Planet" data portal (UNEP-WCMC and IUCN, 2025). We considered three categories of protection: nationally designated protected areas, including both national parks and natural reserves (hereafter national protected areas); sites belonging to the Natura 2000 network, under Habitat Directive 92/43/EEC and Bird Directive 79/409/EC, which, compared to national protected areas allow a sustainable, albeit limited, use of nature (e.g., controlled dune stabilization or regulated tourism; Dudley, 2008; EC, 2025; Tsiafouli et al., 2013); and areas without any legal protection status. In cases of overlap between protection categories, such as Natura 2000 sites located within national protected areas, we assigned the highest protection level.

Historical and revisitation data

We used an Italian database of coastal dune vegetation resurveying ("RanVegDunes" dataset; Sperandii et al., 2017) consisting for the study area in 286 quasi-permanent random plots (4 m^2), i.e. georeferenced plots that can be reliably relocated over time, despite not being physically marked in the field (Kapfer et al., 2017).These plots were uniformly distributed across six study sites (Appendix S1) covering all herbaceous habitats of the coastal vegetation zonation along the sea-inland gradient in central Italy, i.e. the upper beach, embryo dunes, shifting dunes, and dune grasslands (see Acosta et al., 2007 for a description of the habitats). Overall, plot fall within two national protected areas, five Natura 2000 sites, and six non protected areas (Appendix S1 and S2). Plots were historically surveyed between 2005-2009 (historical data) and then revisited between 2017-2021 (revisitation data), thus after a period of 10-15 years. In each plot, sampling consisted in recording the number and percentage cover of all vascular plant species during the peak of the growing season (between April and May). Sampling method was consistent between historical survey and revisitation (see Sperandii et al., 2019 for details about the resurveying protocol). We classified species into natives and non-natives following (Galasso et al., 2024).

Plant traits

We compiled functional trait data from local databases incorporated into TRY (Kattge et al., 2020), aggregating plant traits measured within the study area of Central Italy. Specifically, we chose four aboveground and four belowground traits: plant height (H; measured in m), leaf area (LA; measured in cm2), specific leaf area (SLA; measured in mm2/mg), leaf dry matter content (LDMC, measured in mg/g), root diameter (RD; measured in mm), specific root length (SRL; measured in m/g), root tissue density (RTD; measured in g/cm3), and root dry matter content (RDMC; measured in g/g). These traits are associated with the leaf and roots economic spectra (Carmona et al., 2021; Díaz et al., 2016), and are known to be related to several ecosystem processes and functions (Bardgett et al., 2014; de Bello et al., 2010; Freschet et al., 2021; Lavorel & Garnier, 2002; Roumet et al., 2016), including on coastal dunes (La Bella et al., 2023, 2024). Trait values were averaged across individuals by species, then log-transformed, centred, and scaled (Májeková et al., 2016).

Environmental conditions

To account for climatic variability potentially affecting changes in vegetation between the two sampling periods, we obtained mean annual temperature (MAT), mean temperature in the coldest quarter (Bio11), mean annual precipitation (MAP), and precipitation in the warmest

quarter (Bio18) of each plot from CHELSA, a high-resolution (30 arcs, ~1 km) global database of climate data (Karger et al., 2021). However, due to high collinearity among these variables (see Appendix S3), we retained MAT as the sole climatic predictor in the models to avoid multicollinearity issues and improve model interpretability. In coastal dunes, the distance from the sea is commonly associated with a spatial gradient of environmental disturbance and, to a lesser extent, stress (Carboni et al., 2011). This means that communities located closer to the shoreline are typically exposed to higher levels of natural disturbances, such as tidal storms, salt spray, wind, and soil burial. Stress factors, such as water and nutrient limitation due to the high permeability of sandy soils, may also covary with sea distance, although not always linearly (McLachlan & Defeo, 2017). Despite such heterogeneity, sea distance remains a widely used proxy for capturing the composite sea-inland environmental gradient in coastal dune systems. Therefore, we included the shortest distance between each plot and the shoreline (hereafter sea distance) as a predictor in the models (Bazzichetto et al., 2016; La Bella et al., 2023). The shorelines of 2006 and 2017 were obtained through photointerpretation from satellite images in a QGIS environment (Version 3.28). We used the shoreline of 2006 for historical plots and the shoreline of 2017 for revisitation plots to quantify sea distance as the shortest Euclidean distance of each plot from the shoreline. As tidal variations along Mediterranean Sea are relatively small, ranging between 0.2-1 m (Fenu et al., 2013), the effect of tides on sea proximity measurement is negligible.

BEF subset

Given the logistical constraints of measuring ecosystem functions across the entire dataset, we extracted from the revisitation dataset a subset of 110 plots to collect data related to ecosystem functioning in addition to plant biodiversity (hereafter, BEF subset). We used a stratified random approach based on biodiversity data and environmental conditions in order to represent the full range of variation observed across both historical and revisitation datasets. For this subset, in spring-summer season of 2021, we repeated the floristic survey (as explained above) and concurrently collected ecosystem function variables related to supporting and regulating services (Garland et al., 2021; MEA, 2005). Supporting services were biomass production, soil water regulation, soil carbon stock, and soil nutrients cycling, while as regulating services we assessed erosion control and invasion resistance (Table 1). Aboveground plant living biomass was used as an indicator of biomass production potential across plots (Scurlock et al., 2002). Water regulation, carbon stock, and nutrient cycling were

quantified from soil samples collected in each plot. A full description of the sampling and analysis of these ecosystem functions can be found in Supporting Information (Appendix S4).

In coastal dunes, erosion control is mainly operated by rhizomatous geophyte grasses that, through a dense root network, entrap sandy sediments and stabilize the dune. For this function, rhizomatous geophyte grasses are commonly referred to as dune-building or engineering species (Durán & Moore, 2013; Levin et al., 2008; Stallins, 2005). Here we used the total cumulative cover of the dune-building grasses *Calamagrostis arenaria, Thinopyrum junceum,* and *Sporobolus virginicus*, as a measure of the erosion control function. Invasion resistance was assessed as 100% minus the cumulative percent cover of the most abundant non-native species in the study area (Cascone et al., 2021), i.e. *Carpobrotus edulis*, *Carpobrotus acinaciformis*, *Agave americana subsp. americana*, and *Yucca gloriosa* (Balvanera et al., 2006; Quijas et al., 2010). Thus, higher values indicate greater resistance to invasion due to the lower cover of non-natives, while lower values indicate lower resistance to invasion.

Table 1 Variables and methods used to measure ecosystem functions.

Category	Ecosystem functions	Variables	Measurement method
Supporting	Biomass production	Aboveground plant biomass	Weight of oven-dried plant biomass
	Water regulation	Soil water holding capacity (WHC)	Gravimetric method (Robertson et al., 1999)
		Soil electrical conductivity (EC)	Electrical conductivity meter
	Carbon stock	Soil total organic carbon (TOC) Soil total carbon (TC)	Thermal-gradient analysis
	Nutrient cycling	Soil total nitrogen (TN) CN ratio	Thermal-gradient analysis
		Soil total content of P, K, Mg, Zn, Ni, Cu, and Fe	Mehlich III method (Mehlich, 1984)
Regulating	Erosion control	Total cover of dune-building grasses: *Calamagrostis arenaria, Thinopyrum junceum,* and *Sporobolus virginicus* (Durán & Moore, 2013; Levin et al., 2008; Stallins, 2005)	

	Invasion resistance	100 minus the cumulative percent cover of non-native species (*Carpobrotus edulis, Carpobrotus acinaciformis, Agave americana subsp. americana,* and *Yucca gloriosa;* Balvanera et al., 2006; Quijas et al., 2010)

Biodiversity and ecosystem functioning metrics calculation

Our approach assumes that past trait-environment relationships have remained stable over time, but this may not always hold, particularly in ecosystems that have undergone significant anthropogenic or climatic changes. Consequently, although the original resurveying database included 286 plots (each visited twice), we only hindcasted ecosystem functions in plots where environmental conditions were expected to have remained relatively stable over time, minimizing the risk of violating this assumption. In particular, we excluded plots: (1) in which vegetation disappeared at the time of the resurvey due to anthropogenization or coastal erosion (n = 50); (2) belonging to the drift lines' annual vegetation (n = 21), due to their ephemeral nature (Maun, 2009); (3) where species with trait information did not account for at least 80 % of the total vegetation (n = 8) (Pakeman & Quested, 2007); (4) where environmental conditions and biodiversity measures were out of the prediction range to avoid extrapolation (n = 25). Although plots where vegetation disappeared may provide interesting insights over the consequences of habitat loss, we chose not to consider them in the analysis due to the unfeasibility of assigning reliable values of ecosystem functions. The final dataset consisted of 203 plots with two repeated measures each, located either inside national protected areas (n = 69) and Natura 2000 sites (n = 89), or in non-protected areas (n = 45).

Biodiversity metrics

In each plot, we calculated several diversity metrics for historical and revisitation data. In particular, we assessed both taxonomic and functional aspects of the community, including species richness, total vegetation cover, above- and below-ground functional diversity (FD_{above} and FD_{below}), and the Community-Weighted Mean (CWM) of each trait.

Species richness was quantified as the total count of recorded species, while total vegetation cover as the cumulative percentage of plant cover within the plot area.

For each trait, the community-weighted mean (CWM) was computed using the “functcomp” function (package FD; Laliberté et al., 2014). To characterize the functional diversity of plant communities, we calculated Rao’s quadratic entropy (RaoQ) as a measure of overall trait dissimilarity. In addition, to capture other key aspects of functional diversity, we also computed functional richness (FRic) and functional evenness (FEve), which represent the extent of trait

space occupied by the community and the regularity of species' trait distribution within that space, respectively (Villéger et al., 2008). Functional diversity indices were measured separately for above and belowground traits. We first generated a multi-trait dissimilarity matrix across all species ("gawdis" function; de Bello et al., 2021) and then used the "dbFD" function (package FD; Laliberté et al., 2014) to quantify above- and belowground functional dissimilarity (FD_{above} and FD_{below}), functional richness ($FRic_{above}$ and $FRic_{below}$), and functional evenness ($FEve_{above}$ and $FEve_{below}$). However, to avoid inflating model complexity and potential multicollinearity, we ran separate models for each functional diversity index. As the results obtained with FRic and FEve were qualitatively similar to those based on RaoQ functional dissimilarity (FD), we present only the models including RaoQ in the main text and report the others in the Supporting Information (Appendix S5 and S6).

Ecosystem functioning

Ecosystem functions assessed using more than one variable (see Table 1), i.e. water regulation, soil carbon stock, and nutrient cycling were quantified using the averaging approach (Byrnes et al., 2014; Maestre et al., 2012). This method calculates the mean of the Z-score values across the EF variables, providing a straightforward composite measure of an ecosystem function.

Statistical analysis

In order to reconstruct past ecosystem functions and assess how they changed over time, we used a stepwise approach (Fig. 1). We first assessed BEF relationships in a subset of vegetation plots and estimated the predictive ability of biodiversity for hindcasting ecosystem functioning. Then, we applied the estimated BEF relationships to predict ecosystem functions of a large resurveying vegetation dataset as well as past ecosystem functioning levels based on historical vegetation data. Finally, we applied this method to explore temporal trends in ecosystem functioning under different protection regimes.

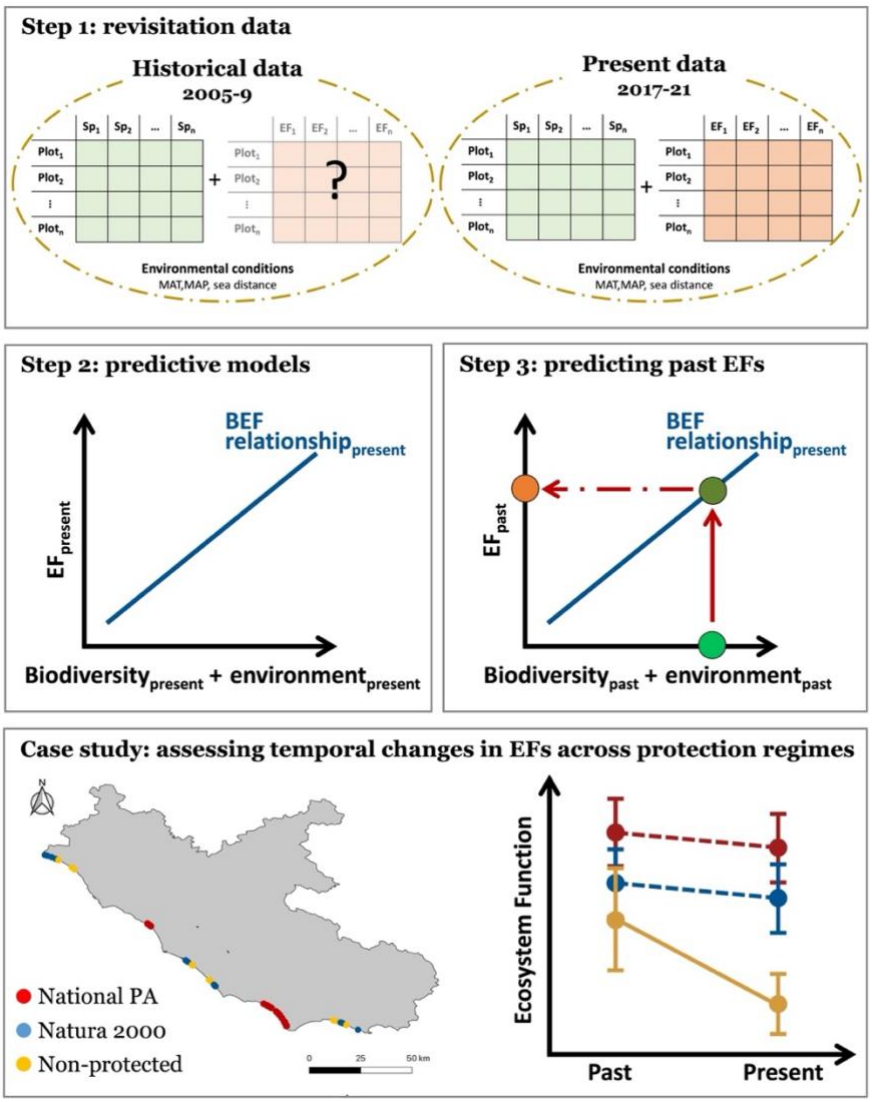

Figure 1 Workflow used to assess temporal changes in ecosystem functioning. **Step 1**: in 2021, we resurveyed vegetation in quasi-permanent plots and collected several ecosystem function variables. **Step 2**: using these data, we assessed BEF relationships. **Step 3**: we employed BEF relationships to hindcast past ecosystem functioning based on historical vegetation surveys. **Case study:** we used this approach to assess temporal changes in ecosystem functioning across different protection regimes, i.e. national protected areas (PA), Natura 2000 sites, and non-protected areas, along the coastal dunes of Lazio region (Italy).

Predictive models for ecosystem functions

Past erosion control and invasion resistance were quantified directly from historical vegetation records, as these functions are derived from vegetation data. In contrast, past information on biomass production, carbon stock, water regulation, and nutrient cycling was not available for plots outside the BEF subset or for the historical surveys, and therefore

required predictions. Thus, using both random forest (randomforest package; Breiman et al., 2018) and linear mixed-effect models ("lmer" function, lme4 package; Bates et al., 2014), we fit current BEF relationships based on the data of the BEF subset. Separate models were fit for each ecosystem function considered as response variable, i.e. biomass production, soil carbon stock, water regulation, and nutrient cycling. In each model, biodiversity metrics and environmental variables were used as predictors. Before conducting the analysis, we assessed multicollinearity among predictors and excluded those with a Variance Inflation Factor (VIF) exceeding 3 (Zuur et al., 2010; see also Appendix S3). The final predictors included SR, total vegetation cover, FD_{above}, FD_{below}, H_{CWM}, SLA_{CWM}, $LDMC_{CWM}$, $RDMC_{CWM}$, RTD_{CWM}, sea distance, and MAT. Additionally, in the mixed-effect models, we incorporated the statistical interaction between sea distance and the biodiversity variables to address potential impacts of stress and disturbance on the BEF relationships (La Bella et al., 2024), resulting in the initial estimation of 20 fixed-effects parameters before model selection. For each response variable, we initially fitted mixed-effects models including study sites as a random factor. The significance of random effects was assessed through likelihood ratio tests comparing models with and without the random term. If the random effect was not significant, it was removed, and the simpler linear model was retained to improve parsimony. We applied a model selection and averaging approach using the "dredge" function (MuMIn R package; Barton & Barton, 2023), which ranks all possible models based on their Akaike Information Criterion (AIC), accounting for both model fit (via log-likelihood) and complexity. We then averaged the parameters of all models with $\Delta AIC < 2$ (Burnham & Anderson, 2002; Richards, 2005). Model residuals underwent visual inspection to ensure homoscedasticity and normality (Zuur et al., 2010). Before the analysis, biomass production was log-transformed to improve model fit and reduce skewness in the response variable.

Cross-validation

We used leave-one-out cross-validation (LOOCV) to assess the predictive power of the selected models related to supporting ecosystem functions (Yates et al., 2023). It consisted in randomly leaving out from the model training set one plot per run, which was used as testing set to predict the ecosystem function variable starting from the plant community and environmental data. However, since within each site plots where moderately close to each other due to the narrow spatial conformation of Mediterranean coastal dunes, we applied the B-LOOCV proposed by Ploton et al. (2020) to account for spatial autocorrelation. This method leaves out from the training set the testing plot and all its neighbours included within a spatial

buffer. We tested several exclusion buffer sizes, with radius ranging between 0 m and 500 m with intervals of 50 m. This spatial extent has been previously advised for Mediterranean costal dunes (Malavasi et al., 2018). The procedure was repeated 1000 times for each ecosystem function model. For each iteration, the model predictive performance was quantified using several goodness-of-prediction statistics: Pearson correlation between observed and predicted values (Pearson *r*); root mean squared error (RMSE); and mean absolute error (MAE). Response variables were scaled and centred (mean = 0 and standard deviation = 1) before the cross validation in order to obtain standardized and comparable measures of the predictive performance (i.e. not influenced by the unit scale). We then calculated the mean and standard deviation of each statistical metric, grouped by buffer size to evaluate if the model predictive accuracy was affected by spatial autocorrelation. The model performance was consistent across sizes of exclusion buffer (Appendix S7), indicating low spatial dependence between training and test sets. Linear models and random forest performance was qualitatively similar (see Table 2 for linear models and Appendix S8 for random forest), thus we kept only mixed-effect models for predictions.

We finally selected BEF models that showed a satisfactory predictive ability (e.g. Pearson $r > 0.5$) and used them to predict the ecosystems functions of the whole resurveying dataset in both time periods, i.e. past and present.

Temporal changes in ecosystem functioning across different protection regimes

We evaluated differences in temporal changes in ecosystem functioning in communities under different protection regimes using linear mixed-effect models ("lmer" function, lme4 package; Bates et al., 2014). Ecosystem functions were analysed separately and used as separate response variables. Time, protection status, and their interaction were used as fixed effects. The "Time" variable consisted in two levels, i.e. past and present; while protection status had three levels, i.e. national protected areas, Natura 2000 sites, and non-protected areas. Study site and plot identity nested in study site were included as random intercepts to account for the temporal repeated measure in each plot and for the spatial dependence of plots within sites. We performed a Moran's test to verify for any remaining spatial autocorrelation of the residuals of each model ("moran.test" function, spdep package; Bivand & Wong, 2018), which revealed no spatial autocorrelation (Appendix S9). For each model, residuals were visually inspected to ensure assumptions of homoscedasticity and normality (Zuur et al., 2010). Finally, Tukey's post hoc tests were used to assess the significance of temporal trends in the three classes of protection (emmeans package; Lenth, 2022). We consider non-significant changes

as an effective conservation outcome of the protection regime. Since species related to invasion resistance or erosion control were not present in all plots, we analysed temporal trends in these two functions only in plots were non-native or dune-building species were recorded in at least one survey period. For invasion resistance and erosion control, which are proportion variables bounded between 0 and 1, we used beta regression models (glmmTMB package; Brooks et al., 2017). Prior to the analysis, we transformed the response variables to avoid extremes of 0 and 1 (response variable value/1× (number of observations − 1) + 0.01)/number of observations; Cribari-Neto & Zeileis, 2010). We then used diagnostic plots of the DHARMa package (Hartig, 2017) and tested the model for uniformity, dispersion, and homoscedasticity. To account for uncertainty in the predicted past ecosystem functions, we propagated uncertainty from fixed effects, random effects, and residual variance. We first obtained 1000 samples of the model's coefficients using the "sim" function (arm package; Gelman et al., 2015). For each set of simulated coefficients, we computed the predicted values of the modelled functions and added normally distributed residual noise with standard deviation equal to the model's residual standard deviation (function "sigma"), to account for unexplained variability. Then, to propagate the uncertainty we fit the above-mentioned mixed-effect models testing temporal change in ecosystem functions across different protection levels for each set of simulated predictions (n = 1000), generating a distribution of results that incorporates the three sources of uncertainty, i.e. from fixed effects, random effects, and residual variance. Finally, we extracted the mean values and assessed the significance of temporal changes for each protection level using confidence intervals (Appendix S11).

To evaluate the robustness of our findings on the effects of protection status on temporal changes in ecosystem functioning, while accounting for external confounding factors (Geldmann et al., 2018; Joppa & Pfaff, 2009), we performed additional analyses having first pre-processed data using propensity score matching (PSM). We performed pairwise comparisons between protection levels. For each comparison, we matched plots based on environmental similarity (MAT, distance from the sea, and latitude) using nearest-neighbour matching with replacement and a 1:1 ratio (function "matchit", R package MatchIt; Ho et al., 2007). Covariate balance was evaluated using standardised mean differences and Love plots. We then tested for differences in the magnitude of temporal change in ecosystem function ($\Delta EF = EF_{t0} - EF_{t1}$) between pairs of protection levels (i.e. "non-protected" vs "Natura 2000", " non-protected " vs "national protected areas", "Natura 2000" vs "national protected areas") using Welch's t-tests on matched samples, to assess whether the degree of protection influenced the

observed changes over time. The results of this analysis confirmed the general pattern observed with the mixed-effect models not accounting for matching (Appendix S10).

RESULTS

Predictive models for ecosystem functions

The models of biomass production, water regulation, carbon stock, and nutrient cycling showed a satisfactory predictive ability (Pearson $r > 0.5$; Table 2). Among these, the model for biomass production had the highest predictive ability (Pearson $r = 0.8$).

Table 2 Predictive models of ecosystem functions. The table shows the goodness-of-prediction of each model evaluated with several statistics, i.e. Pearson *r*, RMSE, and MAE. "Model" column indicates whether the final model after model selection consisted in a linear model ("lm") or a linear mixed-effect model ("lmer"). Multiple R^2 and conditional R^2 are provided for each model. Pearson *r* is the correlation between observed and predicted values. RMSE stand for root-mean-square error while MAE for mean absolute error.

EFs	Model	Multiple/ conditional R^2	Pearson *r*	RMSE	MAE
Biomass production	lm	0.760	0.730	0.605	0.489
Water regulation	lmer	0.583	0.581	0.706	0.599
Carbon stock	lmer	0.603	0.566	0.740	0.627
Nutrient cycling	lmer	0.531	0.524	0.832	0.691

Across the models, total vegetation cover and H_{CWM} were the most important predictors of ecosystem functioning (Fig. 2a), with a general positive effect across all functions, suggesting that in open, spatially heterogeneous habitats such as coastal dunes, structural dominance and space occupancy may be the most important drivers. Other biodiversity component also contributed to ecosystem functioning, but their effect was less consistent. Communities with low species richness and higher RTD improved biomass production. Communities dominated by species with high SLA and low functional diversity in the aboveground traits supported greater water regulation. Low functional diversity in the aboveground traits promoted soil carbon stock. Finally, we also tested the interaction between biodiversity metrics and sea

distance, and we found that some biodiversity effects vary along the sea-inland environmental gradient (Fig. 2b). For example, the effect of aboveground functional diversity on nutrient cycling decreased along the sea-inland gradient.

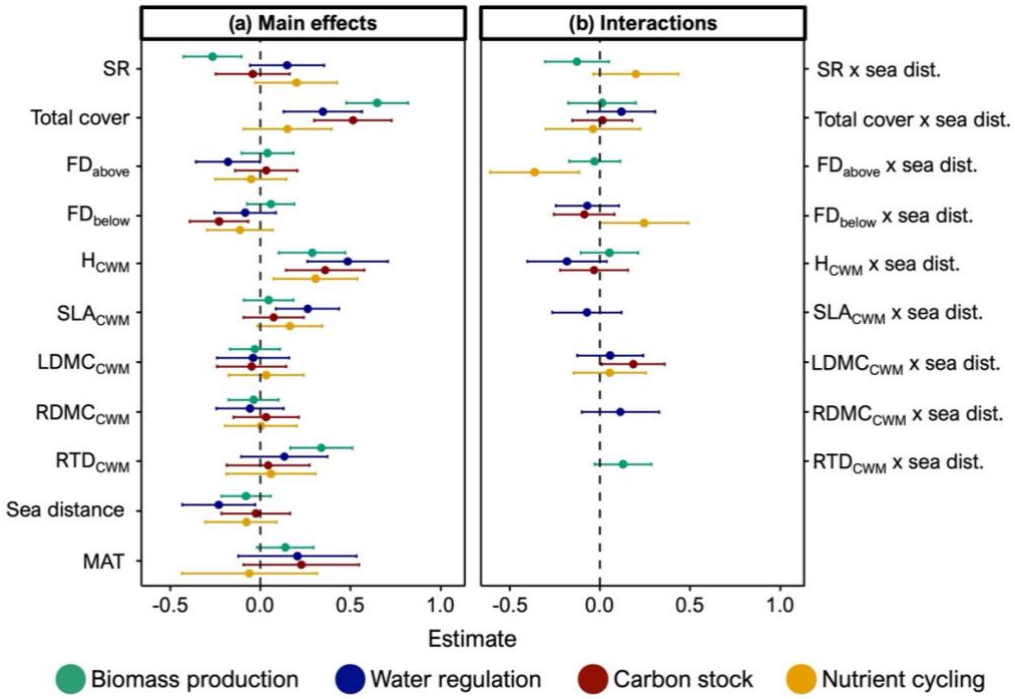

Figure 2 Results of the predictive models. Dots indicate the standardized regression coefficients while error bars the associated 95% confidence intervals. Panel (**a**) shows the direct effects of biodiversity and environmental conditions on ecosystem functioning. Panel (**b**) refers to the interaction coefficients which describe how the effect of the biodiversity metrics on ecosystem functions varies along the sea inland gradient. A negative effect size would indicate e.g. that the effect of SR on EF decreases as sea distance increases. SR, species richness; FD, functional diversity (RaoQ); H, height; SLA, specific leaf area; LDMC, leaf dry matter content; RDMC, root dry matter content; RTD, root tissue density; MAT, mean annual temperature; sea dist., sea distance.

Temporal changes in ecosystem functioning across different protection regimes

We identified temporal changes in all ecosystem functions. However, these changes were not consistent across different protection regimes. Specifically, in non-protected areas, biomass production, water regulation, and carbon stock increased over time (respectively: estimate = 0.491, $P = 0.003$; estimate = 0.767, $P = 0.001$; estimate = 0.516, $P < 0.001$; Fig. 3a-c) while invasion resistance decreased (estimate = -0.721, $P < 0.001$; Fig. 3f). Inside Natura 2000 sites, water regulation increased (estimate = 0.403, $P < 0.001$; Fig 3b) while erosion control and

invasion resistance decreased over time (respectively: estimate = - 0.838, $P < 0.001$; estimate = -0.444, $P < 0.001$; Fig 3e-f). In contrast, within national protected areas most functions remained stable over time and invasion resistance even increased (estimate = 0.722, $P < 0.001$; Fig. 3e).

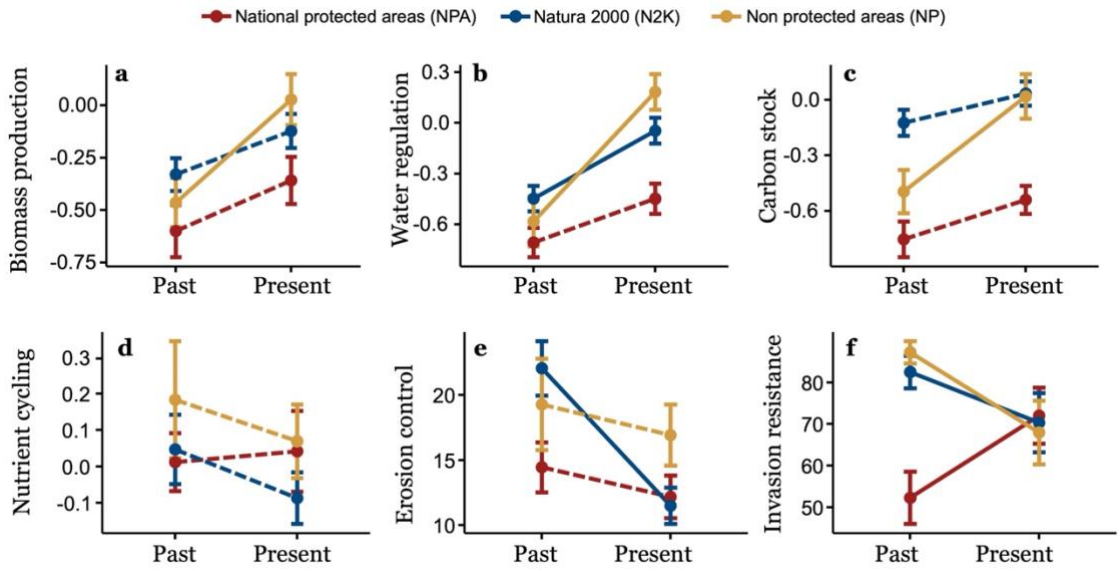

Figure 3 Temporal changes in ecosystem functions. Dots indicate the mean values and lines the associated standard errors. Continuous lines refer to significant changes, while dotted lines to non-significant.

DISCUSSION

In this study, we proposed a new application of revisitation studies in combination with locally estimated BEF relationships to trace back past ecosystem functions and assess how these have changed over the last 10-15 years under different regimes of protection. Specifically, in this section, we first discuss the ability of biodiversity to predict ecosystem functioning in coastal dunes as well as the effectiveness and limitations of using BEF relationships to hindcast past ecosystem functioning. Then, we highlight the relevance for conservation of our approach by proposing a possible implementation, i.e. monitoring how ecosystem functioning has changed in the context of ongoing global changes as well as the performance of different protection regimes in counteracting these temporal changes in the study area.

Biodiversity as predictor of past ecosystem functioning

In this study, we tested the ability of plant diversity and community composition to model several ecosystem functions in coastal dunes, such as biomass production, water regulation, carbon stocks, and nutrient cycling. Since BEF relationships are often highly context dependent

(Jucker et al., 2016; La Bella et al., 2024; Ratcliffe et al., 2017; Spohn et al., 2023), our models also include environmental conditions and how these interact with the effect of plant diversity and community composition on ecosystem functioning. Including the environmental context is necessary to properly model BEF relationships, and it improves the predictive power of biodiversity (Díaz et al., 2007; La Bella et al., 2023, 2024; Lavorel et al., 2011; Van Der Plas et al., 2020). Here, we found that plant diversity, community composition, and environmental conditions can effectively predict the ecosystem functions analysed. Among all functions, biomass production was the one that was best predicted. This result is in line with the strong plant diversity-biomass relationship found in a wide range of natural ecosystems (Duffy et al., 2017; Garnier et al., 2015; van der Plas, 2019). In contrast, the predictive ability of the models for carbon stock, water regulation, and nutrient cycling was lower compared to biomass production, but still sufficient to make meaningful predictions.

Interestingly, we found that some ecosystem functions tend to decrease along the sea-inland gradient, indicating that close to the sea ecosystem functions are generally higher. This pattern is in line with previous studies showing that in coastal dunes some ecosystem functions, such as pollination, can display strong spatial heterogeneity, with functional peaks often associated with pioneer communities near the seashore (e.g., drift lines and shifting dunes), due to the presence of keystone species (Fantinato et al., 2018).

Limitations of hindcasting ecosystem functions

While our study highlights the potential of using revisitation data and BEF relationships for hindcasting past ecosystem functions, some limitations should be acknowledged. First, the availability and quality of historical vegetation data pose a constraint, as they can vary in taxonomic resolution, plot re-location accuracy, and sampling methodologies, potentially introducing biases in reconstructing past ecosystem functioning. Second, our hindcasting approach assumes that relationships between biodiversity, environmental conditions, and ecosystem functions remain stable over time. This is a common assumption in predictive ecological models including forecasting (Dietze, 2017), species distribution models (Elith & Leathwick, 2009) and trait-based projections of community or ecosystem change (Laughlin & Messier, 2015). However, we acknowledge that BEF relationships may shift due to changes in environmental drivers, disturbance regimes, or species interactions. While we assessed spatial model accuracy via cross-validation, we could not evaluate temporal predictive performance due to a lack of independent historical data. To overcome these two issues, and to minimise uncertainty and avoid extrapolation errors, we restricted predictions to a subset of the original

revisitation dataset, excluding 30% of plots. Specifically, we excluded plots belonging to highly variable or ephemeral habitats (namely the drift lines' annual vegetation), plots lost due to anthropogenic changes or coastal erosion, and plots with environmental conditions or biodiversity values falling outside the models' prediction range (see Methods for details). Moreover, our simulation-based approach incorporates parameter and residual uncertainty, improving robustness.

Although our hindcasting framework has inherent limitations, it can provide valuable insights where long-term ecosystem function data are lacking. By reconstructing past ecosystem functions based on historical vegetation records, it offers a novel tool to assess long-term ecological changes and support conservation planning. While further testing across ecosystems is needed, this approach can be particularly relevant in light of the growing availability of resurvey data, such as ReSurveyEurope (Knollová et al., 2024), ReSurveyDunes (Acosta et al., 2025), BioTIME (Dornelas et al., 2025), LOTVS (Sperandii et al., 2022), or VESTA (Świerkosz & Reczyńska, 2022).

Temporal changes in ecosystem functioning across different protection regimes

One of the main arguments for biodiversity conservation relies on the ability of biodiversity to support ecosystem functioning (Balvanera et al., 2006). However, although many BEF studies have reported positive relationships between biodiversity and ecosystem functioning (Le Bagousse-Pinguet et al., 2019; Maestre et al., 2012), several others have highlighted that these relationships are not universally positive and can vary depending on context (Hagan et al., 2021; van der Plas, 2019). In line with this, our findings show that biodiversity can have positive, negative or no effects on ecosystem functioning. Conserving biodiversity may therefore not necessarily ensure the conservation of ecosystem functioning (O'Connor et al., 2021). To explore this further, we applied our hindcasting approach to compare temporal changes in ecosystem functions across three protection regimes, i.e. national protected areas, Natura 2000 sites, and non-protected areas. Our case study illustrates the utility of our method for conservation and long-term ecological monitoring by revealing how biodiversity changes affect ecosystem functioning.

In the study area, significant vegetation shifts have occurred over the past 10-15 years both inside and outside Natura 2000 sites (Sperandii et al., 2020). As a result, we detected also marked temporal changes in ecosystem functions. Specifically, in non-protected areas and Natura 2000 sites, we observed that most ecosystem functions underwent temporal changes with a general increase of biomass production, carbon stock, and water regulation and a decline

in invasion resistance and erosion control. These patterns likely reflect the spread of fast-growing ruderal and non-native species observed in Italian coastal dunes (Del Vecchio et al., 2015, 2016), which often replace native dune-building species responsible of erosion control (Sarmati et al., 2025). Additional trait-based analyses (Appendix S12) supported these trends also in our study area, showing a significant decrease in $LDMC_{CWM}$ in non-protected areas and an increase, although non statistically significant, in SLA_{CWM} in Natura 2000 sites, suggesting a potential shift towards communities with more acquisitive strategies. While ruderal and non-native species may temporarily enhance certain functions, such as biomass production and biogeochemical cycles (Liao et al., 2008; Marzialetti et al., 2025; Rout & Callaway, 2009; Simberloff, 2011; Wilsey et al., 2023), they may, in contrast, undermine long-term ecosystem stability and biodiversity (Del Vecchio et al., 2015; La Bella et al., 2023; Sperandii et al., 2021). It is important to mention that, in coastal dunes, key services include erosion control, invasion resistance, and habitat provisioning for biodiversity (Drius et al., 2019a, 2019b; Liquete et al., 2013), while supporting services such as biomass production, water regulation, and nutrient cycling are relatively low compared to other habitats (Van Der Biest et al., 2017). As such, in these ecosystems higher functioning may not necessarily translate into more valuable services to human societies, and only key functions should be prioritised by conservation actions. Our findings indicate that non-protected coastal dunes, which are not subject to formal management, are shifting toward more productive ecosystems, likely due to the spread of non-native species, with the risk of losing their key characteristics in the near future. On the contrary, Natura 2000 sites are not effectively protecting dune-building species responsible for dune stability, resulting thereby in the loss of erosion control potential, and not contrasting the expansion of non-native species. These findings question the efficacy of the existing set of Natura 2000 sites of our study area to counteract pressures such as tourism, urban expansion, or climate change-driven disturbances, as observed also by other studies (Bricca et al., 2024; Ricci et al., 2023; Tsiafouli et al., 2013) and to prevent the decline of ecosystem functioning or, as showed by Sperandii et al. (2020), of biodiversity.

In contrast, within national protected areas most ecosystem functions remained stable over time and invasion resistance even increased. Although we lack specific data on local management actions, the observed reduction in non-native species suggests the possible influence of control efforts. These results are encouraging, indicating that the national protected areas considered here have been effectively managed to maintain functioning ecosystems beyond biodiversity, even in vulnerable ecosystems like coastal dunes. Our findings support previous research suggesting that stricter protection can enhance ecosystem functioning

(Lecina-Diaz et al., 2019), although evidence remains mixed (Castro et al., 2015; Olmo et al., 2025).

The contrasting outcomes between national protected areas and Natura 2000 network likely reflect their distinct objectives and management practices. Natura 2000 network aims to maintain endangered species and habitats at "favourable conservation status" (Commission, 2009), while national protected areas often pursue broader goals, including the conservation of ecosystems and natural processes (CBD, 2022). Still, the most substantial differences lie in management: national protected areas typically benefit from dedicated funding and personnel, whereas Natura 2000 sites often do not (Hochkirch et al., 2013; Kati et al., 2015; Maes et al., 2013). In coastal dunes, effective conservation measures include fencing or wooden walkways to reduce trampling (Acosta et al., 2013; Prisco et al., 2021), invasive species control (Lazzaro et al., 2020), and limitation on mechanical beach cleaning (Battisti et al., 2023; Roig et al., 2009), therefore practices that require considerable financial resources for implementation and long-term maintenance..

Finally, it is important to remember that this work is not intended to assess the overall success of different protection regimes but rather to examine how revisitation data can be used to monitor changes in ecosystem functioning within an existing network of protected areas. Here, we provide insights into the dynamics of ecosystem functioning, highlighting the potential of the hindcasting approach proposed for monitoring and conservation purposes. However, as the number of sites included in this study is relatively limited, the findings from these specific locations should be generalized with caution.

Final remarks

Although most revisitation studies primarily focus on quantifying biodiversity changes in response to global changes, they also offer a unique opportunity to evaluate temporal trends in the ecosystem functions that biodiversity supports. A major limitation, however, is the lack of direct historical measurements of ecosystem processes. Here, to overcome this data gap, we proposed an innovative practical solution, namely hindcasting past ecosystem functions using present-day BEF relationships. This approach enables to jointly assess temporal changes in biodiversity and ecosystem functioning, providing a more comprehensive understanding of the impacts of global changes on natural ecosystems. In this light, we demonstrated a possible application and the relevance for conservation of this approach, showing how ecosystem functioning has changed over the last 15 years in protected and unprotected coastal dunes.

Many conservation efforts currently focus on preserving individual species or habitats, often neglecting crucial functions and services that ecosystems provide, despite their central role goals and targets of the post-2020 global biodiversity framework. Here, we showed that areas designed to conserve biodiversity can also help maintain ecosystem functions. Still, the mere designation of a protected area does not guarantee conservation success, as evidenced by the functioning decline observed in the Natura 2000 sites of this study. Preserving well-functioning coastal dune ecosystems may instead require rigorous, well-funded conservation measures, like those implemented in national protected areas. These findings highlight the need to monitor within protected areas not only biodiversity but also trends in ecosystems functioning to ensure long-term conservation success.

REFERENCES

Acosta, A., Carranza, M. L., & Izzi, C. F. (2009). Are there habitats that contribute best to plant species diversity in coastal dunes? *Biodiversity and Conservation*, *18*(4), 1087–1098. https://doi.org/10.1007/s10531-008-9454-9

Acosta, A., Ercole, S., Stanisci, A., Pillar, V. D. P., & Blasi, C. (2007). Coastal Vegetation Zonation and Dune Morphology in Some Mediterranean Ecosystems. *Journal of Coastal Research*, *2007*(236), 1518–1524. https://doi.org/10.2112/05-0589.1

Acosta, A. T. R., Di Biase, L., Sarmati, S., Allevato, E., Angiolini, C., Bagella, S., Bazan, G., Bertacchi, A., Brancaleoni, L., & Buffa, G. (2025). ReSurveyDunes—A data resource of resurveyed coastal dune vegetation plots in Italy. *Vegetation Ecology and Diversity*, *62*, 1–6.

Acosta, A. T. R., Jucker, T., Prisco, I., & Santoro, R. (2013). Passive Recovery of Mediterranean Coastal Dunes Following Limitations to Human Trampling. In M. L. Martínez, J. B. Gallego-Fernández, & P. A. Hesp (Eds.), *Restoration of Coastal Dunes* (pp. 187–198). Springer Berlin Heidelberg. https://doi.org/10.1007/978-3-642-33445-0_12

Balvanera, P., Pfisterer, A. B., Buchmann, N., He, J.-S., Nakashizuka, T., Raffaelli, D., & Schmid, B. (2006). Quantifying the evidence for biodiversity effects on ecosystem functioning and services. *Ecology Letters*, *9*(10), 1146–1156. https://doi.org/10.1111/j.1461-0248.2006.00963.x

Bardgett, R. D., Mommer, L., & De Vries, F. T. (2014). Going underground: Root traits as drivers of ecosystem processes. *Trends in Ecology & Evolution*, *29*(12), 692–699. https://doi.org/10.1016/j.tree.2014.10.006

Barton, K., & Barton, M. K. (2023). *Package 'MuMIn'*.

Bates, D., Mächler, M., Bolker, B., & Walker, S. (2014). *Fitting Linear Mixed-Effects Models using lme4*. https://doi.org/10.48550/ARXIV.1406.5823

Battisti, C., Perchinelli, M., Vanadia, S., Giovacchini, P., & Marsili, L. (2023). Monitoring Effectiveness of an Operational Project on Two Threatened Landbirds: Applying a Before–After Threat Analysis and Threat Reduction Assessment. *Land*, *12*(2), 464. https://doi.org/10.3390/land12020464

Bazzichetto, M., Malavasi, M., Acosta, A. T. R., & Carranza, M. L. (2016). How does dune morphology shape coastal EC habitats occurrence? A remote sensing approach using airborne LiDAR on the Mediterranean coast. *Ecological Indicators*, *71*, 618–626. https://doi.org/10.1016/j.ecolind.2016.07.044

Bivand, R. S., & Wong, D. W. S. (2018). Comparing implementations of global and local indicators of spatial association. *TEST*, *27*(3), 716–748. https://doi.org/10.1007/s11749-018-0599-x

Bodenhausen, N., Hess, J., Valzano, A., Deslandes-Hérold, G., Waelchli, J., Furrer, R., Van Der Heijden, M. G. A., & Schlaeppi, K. (2023). Predicting soil fungal communities

from chemical and physical properties. *Journal of Sustainable Agriculture and Environment*, *2*(3), 225–237. https://doi.org/10.1002/sae2.12055

Brauman, K. A., Garibaldi, L. A., Polasky, S., Aumeeruddy-Thomas, Y., Brancalion, P. H. S., DeClerck, F., Jacob, U., Mastrangelo, M. E., Nkongolo, N. V., Palang, H., Pérez-Méndez, N., Shannon, L. J., Shrestha, U. B., Strombom, E., & Verma, M. (2020). Global trends in nature's contributions to people. *Proceedings of the National Academy of Sciences*, *117*(51), 32799–32805. https://doi.org/10.1073/pnas.2010473117

Breiman, L., Cutler, A., Liaw, A., Wiener, M., & Liaw, M. A. (2018). Package 'randomforest'. *University of California, Berkeley: Berkeley, CA, USA*, *81*, 1–29.

Bricca, A., Deola, T., Zerbe, S., Bagella, S., Rivieccio, G., Wellstein, C., & Bonari, G. (2024). Higher levels of protection do not consistently improve habitat quality: Insights from Mediterranean and Alpine shrublands. *Biological Conservation*, *293*, 110571. https://doi.org/10.1016/j.biocon.2024.110571

Brooks, M., Bolker, B., Kristensen, K., Maechler, M., Magnusson, A., Skaug, H., Nielsen, A., Berg, C., & Van Bentham, K. (2017). *glmmTMB: Generalized Linear Mixed Models using Template Model Builder* (p. 1.1.10) [Dataset]. https://doi.org/10.32614/CRAN.package.glmmTMB

Burnham, K. P., & Anderson, D. R. (2002). Model selection and multimodel inference: A practical information-theoretic approach. *2nd Ed. New York, Springer-Verlag*.

Byrnes, J. E. K., Gamfeldt, L., Isbell, F., Lefcheck, J. S., Griffin, J. N., Hector, A., Cardinale, B. J., Hooper, D. U., Dee, L. E., & Emmett Duffy, J. (2014). Investigating the relationship between biodiversity and ecosystem multifunctionality: Challenges and solutions. *Methods in Ecology and Evolution*, *5*(2), 111–124. https://doi.org/10.1111/2041-210X.12143

Carboni, M., Santoro, R., & Acosta, A. T. R. (2010). Are some communities of the coastal dune zonation more susceptible to alien plant invasion? *Journal of Plant Ecology*, *3*(2), 139–147. https://doi.org/10.1093/jpe/rtp037

Carboni, M., Santoro, R., & Acosta, A. T. R. (2011). Dealing with scarce data to understand how environmental gradients and propagule pressure shape fine-scale alien distribution patterns on coastal dunes: Drivers of alien patterns on coastal dunes. *Journal of Vegetation Science*, *22*(5), 751–765. https://doi.org/10.1111/j.1654-1103.2011.01303.x

Carmona, C. P., Bueno, C. G., Toussaint, A., Träger, S., Díaz, S., Moora, M., Munson, A. D., Pärtel, M., Zobel, M., & Tamme, R. (2021). Fine-root traits in the global spectrum of plant form and function. *Nature*, *597*(7878), 683–687. https://doi.org/10.1038/s41586-021-03871-y

Cascone, S., Sperandii, M. G., Cao Pinna, L., Marzialetti, F., Carranza, M. L., & Acosta, A. T. R. (2021). Exploring Temporal Trends of Plant Invasion in Mediterranean Coastal Dunes. *Sustainability*, *13*(24), 13946. https://doi.org/10.3390/su132413946

Castro, A. J., Martín-López, B., López, E., Plieninger, T., Alcaraz-Segura, D., Vaughn, C. C., & Cabello, J. (2015). Do protected areas networks ensure the supply of ecosystem services? Spatial patterns of two nature reserve systems in semi-arid Spain. *Applied Geography*, *60*, 1–9. https://doi.org/10.1016/j.apgeog.2015.02.012

Chen, Y., Wang, J., Jiang, L., Li, H., Wang, H., Lv, G., & Li, X. (2023). Prediction of spatial distribution characteristics of ecosystem functions based on a minimum data set of functional traits of desert plants. *Frontiers in Plant Science*, *14*, 1131778. https://doi.org/10.3389/fpls.2023.1131778

Commission, E. (2009). Directive 2009/147/EC of the European Parliament and of the Council of 30 November 2009 on the conservation of wild birds. *Official Journal of the European Union*, *20*, 7–25.

Convention on Biological Diversity. (2022). *Decision adopted by the Conference of the Parties to the Convention on Biological Diversity: 15/4. Monitoring framework for the Kunming-Montreal Global Biodiversity Framework. Https://www.cbd.int/doc/decisions/cop-15/cop-15-dec-04-en.pdf.*

Cornwell, W. K., Cornelissen, J. H. C., Amatangelo, K., Dorrepaal, E., Eviner, V. T., Godoy, O., Hobbie, S. E., Hoorens, B., Kurokawa, H., Pérez-Harguindeguy, N., Quested, H. M., Santiago, L. S., Wardle, D. A., Wright, I. J., Aerts, R., Allison, S. D., Van Bodegom, P., Brovkin, V., Chatain, A., … Westoby, M. (2008). Plant species traits are the predominant control on litter decomposition rates within biomes worldwide. *Ecology Letters*, *11*(10), 1065–1071. https://doi.org/10.1111/j.1461-0248.2008.01219.x

Cribari-Neto, F., & Zeileis, A. (2010). Beta Regression in *R*. *Journal of Statistical Software*, *34*(2). https://doi.org/10.18637/jss.v034.i02

de Bello, F., Lavorel, S., Díaz, S., Harrington, R., Cornelissen, J. H. C., Bardgett, R. D., Berg, M. P., Cipriotti, P., Feld, C. K., Hering, D., Martins da Silva, P., Potts, S. G., Sandin, L., Sousa, J. P., Storkey, J., Wardle, D. A., & Harrison, P. A. (2010). Towards an assessment of multiple ecosystem processes and services via functional traits. *Biodiversity and Conservation*, *19*(10), 2873–2893. https://doi.org/10.1007/s10531-010-9850-9

DeCock, E., Moeneclaey, I., Schelfhout, S., Vanhellemont, M., De Schrijver, A., & Baeten, L. (2023). Ecosystem multifunctionality lowers as grasslands under restoration approach their target habitat type. *Restoration Ecology*, *31*(1), e13664. https://doi.org/10.1111/rec.13664

Defeo, O., McLachlan, A., Schoeman, D. S., Schlacher, T. A., Dugan, J., Jones, A., Lastra, M., & Scapini, F. (2009). Threats to sandy beach ecosystems: A review. *Estuarine, Coastal and Shelf Science*, *81*(1), 1–12. https://doi.org/10.1016/j.ecss.2008.09.022

Del Vecchio, S., Prisco, I., Acosta, A. T. R., & Stanisci, A. (2015). Changes in plant species composition of coastal dune habitats over a 20-year period. *AoB PLANTS*, *7*. https://doi.org/10.1093/aobpla/plv018

Díaz, S., Kattge, J., Cornelissen, J. H. C., Wright, I. J., Lavorel, S., Dray, S., Reu, B., Kleyer, M., Wirth, C., Colin Prentice, I., Garnier, E., Bönisch, G., Westoby, M., Poorter, H., Reich, P. B., Moles, A. T., Dickie, J., Gillison, A. N., Zanne, A. E., … Gorné, L. D. (2016). The global spectrum of plant form and function. *Nature*, *529*(7585), Article 7585. https://doi.org/10.1038/nature16489

Díaz, S., Lavorel, S., De Bello, F., Quétier, F., Grigulis, K., & Robson, T. M. (2007). Incorporating plant functional diversity effects in ecosystem service assessments. *Proceedings of the National Academy of Sciences*, *104*(52), 20684–20689. https://doi.org/10.1073/pnas.0704716104

Diaz, S., Lavorel, S., de Bello, F., Quetier, F., Grigulis, K., & Robson, T. M. (2007). Incorporating plant functional diversity effects in ecosystem service assessments. *Proceedings of the National Academy of Sciences*, *104*(52), 20684–20689. https://doi.org/10.1073/pnas.0704716104

Dietze, M. (2017). Ecological forecasting. In *Ecological Forecasting*. Princeton University Press.

Dornelas, M., Antão, L. H., Bates, A. E., Brambilla, V., Chase, J. M., Chow, C. F. Y., Fontrodona-Eslava, A., Magurran, A. E., Martins, I. S., Moyes, F., Sagouis, A., Adu-Acheampong, S., Acquah-Lamptey, D., Adam, D., Ajani, P. A., Albaina, A., Almaraz, P., An, J., Anderson, R. S., … Zvuloni, A. (2025). BioTIME 2.0: Expanding and Improving a Database of Biodiversity Time Series. *Global Ecology and Biogeography*, *34*(5), e70003. https://doi.org/10.1111/geb.70003

Drius, M., Bongiorni, L., Depellegrin, D., Menegon, S., Pugnetti, A., & Stifter, S. (2019). Tackling challenges for Mediterranean sustainable coastal tourism: An ecosystem service perspective. *Science of The Total Environment*, *652*, 1302–1317. https://doi.org/10.1016/j.scitotenv.2018.10.121

Drius, M., Jones, L., Marzialetti, F., De Francesco, M. C., Stanisci, A., & Carranza, M. L. (2019). Not just a sandy beach. The multi-service value of Mediterranean coastal dunes. *Science of The Total Environment*, *668*, 1139–1155. https://doi.org/10.1016/j.scitotenv.2019.02.364

Dudley, N. (2008). *Guidelines for applying protected area management categories*. IUCN.

Duffy, J. E., Godwin, C. M., & Cardinale, B. J. (2017). Biodiversity effects in the wild are common and as strong as key drivers of productivity. *Nature*, *549*(7671), Article 7671. https://doi.org/10.1038/nature23886

Durán, O., & Moore, L. J. (2013). Vegetation controls on the maximum size of coastal dunes. *Proceedings of the National Academy of Sciences*, *110*(43), 17217–17222. https://doi.org/10.1073/pnas.1307580110

EC. (2025). *Https://ec.europa.eu/environment/nature/natura2000/index_en.htm retrived on 24/03/2025*.

Elith, J., & Leathwick, J. R. (2009). Species Distribution Models: Ecological Explanation and Prediction Across Space and Time. *Annual Review of Ecology, Evolution, and Systematics*, *40*(1), 677–697. https://doi.org/10.1146/annurev.ecolsys.110308.120159

Fantinato, E., Del Vecchio, S., Silan, G., & Buffa, G. (2018). Pollination networks along the sea-inland gradient reveal landscape patterns of keystone plant species. *Scientific Reports*, *8*(1). https://doi.org/10.1038/s41598-018-33652-z

Farnsworth, K. D., Albantakis, L., & Caruso, T. (2017). Unifying concepts of biological function from molecules to ecosystems. *Oikos*, *126*(10), 1367–1376. https://doi.org/10.1111/oik.04171

Fenu, G., Carboni, M., Acosta, A. T. R., & Bacchetta, G. (2013). Environmental Factors Influencing Coastal Vegetation Pattern: New Insights from the Mediterranean Basin. *Folia Geobotanica*, *48*(4), 493–508. https://doi.org/10.1007/s12224-012-9141-1

Fischer, M., Bossdorf, O., Gockel, S., Hänsel, F., Hemp, A., Hessenmöller, D., Korte, G., Nieschulze, J., Pfeiffer, S., Prati, D., Renner, S., Schöning, I., Schumacher, U., Wells, K., Buscot, F., Kalko, E. K. V., Linsenmair, K. E., Schulze, E.-D., & Weisser, W. W. (2010). Implementing large-scale and long-term functional biodiversity research: The Biodiversity Exploratories. *Basic and Applied Ecology*, *11*(6), 473–485. https://doi.org/10.1016/j.baae.2010.07.009

Freschet, G. T., Roumet, C., Comas, L. H., Weemstra, M., Bengough, A. G., Rewald, B., Bardgett, R. D., De Deyn, G. B., Johnson, D., Klimešová, J., Lukac, M., McCormack, M. L., Meier, I. C., Pagès, L., Poorter, H., Prieto, I., Wurzburger, N., Zadworny, M., Bagniewska-Zadworna, A., … Stokes, A. (2021). Root traits as drivers of plant and ecosystem functioning: Current understanding, pitfalls and future research needs. *New Phytologist*, *232*(3), 1123–1158. https://doi.org/10.1111/nph.17072

Funk, J. L., Larson, J. E., Ames, G. M., Butterfield, B. J., Cavender-Bares, J., Firn, J., Laughlin, D. C., Sutton-Grier, A. E., Williams, L., & Wright, J. (2017). Revisiting the Holy Grail: Using plant functional traits to understand ecological processes. *Biological Reviews*, *92*(2), 1156–1173. https://doi.org/10.1111/brv.12275

Galasso, G., Conti, F., Peruzzi, L., Alessandrini, A., Ardenghi, N. M. G., Bacchetta, G., Banfi, E., Barberis, G., Bernardo, L., Bouvet, D., Bovio, M., Castello, M., Cecchi, L., Del Guacchio, E., Domina, G., Fascetti, S., Gallo, L., Guarino, R., Gubellini, L., … Bartolucci, F. (2024). A second update to the checklist of the vascular flora alien to Italy. *Plant Biosystems - An International Journal Dealing with All Aspects of Plant Biology*, *158*(2), 297–340. https://doi.org/10.1080/11263504.2024.2320129

Garland, G., Banerjee, S., Edlinger, A., Miranda Oliveira, E., Herzog, C., Wittwer, R., Philippot, L., Maestre, F. T., & Heijden, M. G. A. (2021). A closer look at the functions behind ecosystem multifunctionality: A review. *Journal of Ecology*, *109*(2), 600–613. https://doi.org/10.1111/1365-2745.13511

Garnier, E., Navas, M.-L., & Grigulis, K. (2015). *Plant Functional Diversity: Organism traits, community structure, and ecosystem properties*. Oxford University Press. https://doi.org/10.1093/acprof:oso/9780198757368.001.0001

Geldmann, J., Coad, L., Barnes, M. D., Craigie, I. D., Woodley, S., Balmford, A., Brooks, T. M., Hockings, M., Knights, K., Mascia, M. B., McRae, L., & Burgess, N. D. (2018). A global analysis of management capacity and ecological outcomes in terrestrial protected areas. *Conservation Letters*, *11*(3). https://doi.org/10.1111/conl.12434

Gelman, A., Su, Y.-S., Yajima, M., Hill, J., Pittau, M. G., Kerman, J., Zheng, T., Dorie, V., & Su, M. Y.-S. (2015). Package 'arm'. *Data Analysis Using Regression and Multilevel/Hierarchical Models*.

Hagan, J. G., Vanschoenwinkel, B., & Gamfeldt, L. (2021). We should not necessarily expect positive relationships between biodiversity and ecosystem functioning in observational field data. *Ecology Letters*, *24*(12), 2537–2548. https://doi.org/10.1111/ele.13874

Hanisch, M., Schweiger, O., Cord, A. F., Volk, M., & Knapp, S. (2020). Plant functional traits shape multiple ecosystem services, their trade-offs and synergies in grasslands. *Journal of Applied Ecology*, *57*(8), 1535–1550. https://doi.org/10.1111/1365-2664.13644

Hartig, F. (2017). Package 'DHARMa'. *R Package*.

Heslenfeld, P., Jungerius, P. D., & Klijn, J. A. (2008). European Coastal Dunes: Ecological Values, Threats, Opportunities and Policy Development. In M. L. Martínez & N. P. Psuty (Eds.), *Coastal Dunes* (Vol. 171, pp. 335–351). Springer Berlin Heidelberg. https://doi.org/10.1007/978-3-540-74002-5_20

Ho, D. E., Imai, K., King, G., & Stuart, E. A. (2007). Matching as Nonparametric Preprocessing for Reducing Model Dependence in Parametric Causal Inference. *Political Analysis*, *15*(3), 199–236. https://doi.org/10.1093/pan/mpl013

Hochkirch, A., Schmitt, T., Beninde, J., Hiery, M., Kinitz, T., Kirschey, J., Matenaar, D., Rohde, K., Stoefen, A., Wagner, N., Zink, A., Lötters, S., Veith, M., & Proelss, A. (2013). Europe Needs a New Vision for a Natura 2020 Network. *Conservation Letters*, *6*(6), 462–467. https://doi.org/10.1111/conl.12006

Janssen, J., Rodwell, J., García Criado, M., Gubbay, S., Haynes, T., Nieto, A., Sanders, N., Landucci, F., Loidi, J., Ssymank, A., Tahvanainen, T., Valderrabano, M., Acosta, A., Aronsson, M., Arts, G., Attorre, F., Bergmeier, E., Bijlsma, R.-J., Bioret, F., & Gubbay, S. (2016). *European Red List of Habitats Part 2. Terrestrial and freshwater habitats*. European Commission. https://doi.org/10.2779/091372

Jax, K. (2005). Function and "functioning" in ecology: What does it mean? *Oikos*, *111*(3), 641–648. https://doi.org/10.1111/j.1600-0706.2005.13851.x

Joppa, L. N., & Pfaff, A. (2009). High and Far: Biases in the Location of Protected Areas. *PLoS ONE*, *4*(12), e8273. https://doi.org/10.1371/journal.pone.0008273

Jucker, T., Avăcăriței, D., Bărnoaiea, I., Duduman, G., Bouriaud, O., & Coomes, D. A. (2016). Climate modulates the effects of tree diversity on forest productivity. *Journal of Ecology*, *104*(2), 388–398. https://doi.org/10.1111/1365-2745.12522

Kapfer, J., Hédl, R., Jurasinski, G., Kopecký, M., Schei, F. H., & Grytnes, J.-A. (2017). Resurveying historical vegetation data – opportunities and challenges. *Applied Vegetation Science*, *20*(2), 164–171. https://doi.org/10.1111/avsc.12269

Karger, D. N., Conrad, O., Böhner, J., Kawohl, T., Kreft, H., Soria-Auza, R. W., Zimmermann, N. E., Linder, H. P., & Kessler, M. (2021). *Climatologies at high*

resolution for the earth's land surface areasCHELSA V2.1 (current) (Version 2.1, p. 2.1 KB) [Geotiff]. EnviDat. https://doi.org/10.16904/ENVIDAT.228.V2.1

Kati, V., Hovardas, T., Dieterich, M., Ibisch, P. L., Mihok, B., & Selva, N. (2015). The challenge of implementing the European network of protected areas Natura 2000. *Conservation Biology*, *29*(1), 260–270. https://doi.org/10.1111/cobi.12366

Kattge, J., Bönisch, G., Díaz, S., Lavorel, S., Prentice, I. C., Leadley, P., Tautenhahn, S., Werner, G. D., Aakala, T., & Abedi, M. (2020). TRY plant trait database–enhanced coverage and open access. *Global Change Biology*, *26*(1), 119–188.

Klumpp, K., & Soussana, J.-F. (2009). Using functional traits to predict grassland ecosystem change: A mathematical test of the response-and-effect trait approach. *Global Change Biology*, *15*(12), 2921–2934. https://doi.org/10.1111/j.1365-2486.2009.01905.x

Knollová, I., Chytrý, M., Bruelheide, H., Dullinger, S., Jandt, U., Bernhardt-Römermann, M., Biurrun, I., De Bello, F., Glaser, M., Hennekens, S., Jansen, F., Jiménez-Alfaro, B., Kadaš, D., Kaplan, E., Klinkovská, K., Lenzner, B., Pauli, H., Sperandii, M. G., Verheyen, K., … Essl, F. (2024). ReSurveyEurope: A database of resurveyed vegetation plots in Europe. *Journal of Vegetation Science*, *35*(2), e13235. https://doi.org/10.1111/jvs.13235

Kremen, C., & Ostfeld, R. S. (2005). A call to ecologists: Measuring, analyzing, and managing ecosystem services. *Frontiers in Ecology and the Environment*, *3*(10), 540–548. https://doi.org/10.1890/1540-9295(2005)003[0540:ACTEMA]2.0.CO;2

La Bella, G., Acosta, A. T. R., Jucker, T., Bricca, A., Ciccarelli, D., Stanisci, A., Migliore, M., & Carboni, M. (2024). Below-ground traits, rare species and environmental stress regulate the biodiversity–ecosystem function relationship. *Functional Ecology*, *38*(11), 2378–2394. https://doi.org/10.1111/1365-2435.14649

La Bella, G., Carboni, M., Sperandii, M. G., De Bello, F., Stanisci, A., & Acosta, A. T. R. (2023). Drivers of plant community (in)stability along a sea–inland gradient. *Journal of Ecology*, *111*(10), 2296–2309. https://doi.org/10.1111/1365-2745.14183

Laughlin, D. C., & Messier, J. (2015). Fitness of multidimensional phenotypes in dynamic adaptive landscapes. *Trends in Ecology & Evolution*, *30*(8), 487–496. https://doi.org/10.1016/j.tree.2015.06.003

Lavorel, S., & Garnier, E. (2002). Predicting changes in community composition and ecosystem functioning from plant traits: Revisiting the Holy Grail. *Functional Ecology*, *16*(5), 545–556. https://doi.org/10.1046/j.1365-2435.2002.00664.x

Lavorel, S., Grigulis, K., Lamarque, P., Colace, M., Garden, D., Girel, J., Pellet, G., & Douzet, R. (2011). Using plant functional traits to understand the landscape distribution of multiple ecosystem services. *Journal of Ecology*, *99*(1), 135–147. https://doi.org/10.1111/j.1365-2745.2010.01753.x

Lazzaro, L., Tondini, E., Lombardi, L., & Giunti, M. (2020). The eradication of Carpobrotus spp. in the sand-dune ecosystem at Sterpaia (Italy, Tuscany): Indications from a successful experience. *Biologia*, *75*(2), 199–208. https://doi.org/10.2478/s11756-019-00391-z

Le Bagousse-Pinguet, Y., Soliveres, S., Gross, N., Torices, R., Berdugo, M., & Maestre, F. T. (2019). Phylogenetic, functional, and taxonomic richness have both positive and negative effects on ecosystem multifunctionality. *Proceedings of the National Academy of Sciences*, *116*(17), 8419–8424. https://doi.org/10.1073/pnas.1815727116

Lecina-Diaz, J., Alvarez, A., De Cáceres, M., Herrando, S., Vayreda, J., & Retana, J. (2019). Are protected areas preserving ecosystem services and biodiversity? Insights from Mediterranean forests and shrublands. *Landscape Ecology*, *34*(10), 2307–2321. https://doi.org/10.1007/s10980-019-00887-8

Lenth, R. (2022). *Emmeans: Estimated marginal means, aka least-squares means. R package version 1.7. 2*.

Levin, N., Kidron, G. J., & Ben-Dor, E. (2008). A field quantification of coastal dune perennial plants as indicators of surface stability, erosion or deposition. *Sedimentology*, *55*(4), 751–772. https://doi.org/10.1111/j.1365-3091.2007.00920.x

Liao, C., Peng, R., Luo, Y., Zhou, X., Wu, X., Fang, C., Chen, J., & Li, B. (2008). Altered ecosystem carbon and nitrogen cycles by plant invasion: A meta-analysis. *New Phytologist*, *177*(3), 706–714. https://doi.org/10.1111/j.1469-8137.2007.02290.x

Liquete, C., Zulian, G., Delgado, I., Stips, A., & Maes, J. (2013). Assessment of coastal protection as an ecosystem service in Europe. *Ecological Indicators*, *30*, 205–217. https://doi.org/10.1016/j.ecolind.2013.02.013

Madani, N., Kimball, J. S., Ballantyne, A. P., Affleck, D. L. R., Van Bodegom, P. M., Reich, P. B., Kattge, J., Sala, A., Nazeri, M., Jones, M. O., Zhao, M., & Running, S. W. (2018). Future global productivity will be affected by plant trait response to climate. *Scientific Reports*, *8*(1), 2870. https://doi.org/10.1038/s41598-018-21172-9

Maes, D., Collins, S., Munguira, M. L., Šašić, M., Settele, J., Van Swaay, C., Verovnik, R., Warren, M., Wiemers, M., & Wynhoff, I. (2013). Not the Right Time to Amend the Annexes of the European Habitats Directive. *Conservation Letters*, *6*(6), 468–469. https://doi.org/10.1111/conl.12030

Maestre, F. T., Quero, J. L., Gotelli, N. J., Escudero, A., Ochoa, V., Delgado-Baquerizo, M., Garcia-Gomez, M., Bowker, M. A., Soliveres, S., Escolar, C., Garcia-Palacios, P., Berdugo, M., Valencia, E., Gozalo, B., Gallardo, A., Aguilera, L., Arredondo, T., Blones, J., Boeken, B., … Zaady, E. (2012). Plant Species Richness and Ecosystem Multifunctionality in Global Drylands. *Science*, *335*(6065), 214–218. https://doi.org/10.1126/science.1215442

Malavasi, M., Bartak, V., Carranza, M. L., Simova, P., & Acosta, A. T. R. (2018). Landscape pattern and plant biodiversity in Mediterranean coastal dune ecosystems: Do habitat loss and fragmentation really matter? *Journal of Biogeography*, *45*(6), 1367–1377. https://doi.org/10.1111/jbi.13215

Malavasi, M., Santoro, R., Cutini, M., Acosta, A. T. R., & Carranza, M. L. (2016). The impact of human pressure on landscape patterns and plant species richness in Mediterranean coastal dunes. *Plant Biosystems - An International Journal Dealing*

with All Aspects of Plant Biology, *150*(1), 73–82. https://doi.org/10.1080/11263504.2014.913730

Manning, P., Van Der Plas, F., Soliveres, S., Allan, E., Maestre, F. T., Mace, G., Whittingham, M. J., & Fischer, M. (2018). Redefining ecosystem multifunctionality. *Nature Ecology & Evolution*, *2*(3), 427–436. https://doi.org/10.1038/s41559-017-0461-7

Marzialetti, F., Lozano, V., Große-Stoltenberg, A., Carranza, M. L., Innangi, M., La Bella, G., Bagella, S., Rivieccio, G., Bacchetta, G., Podda, L., & Brundu, G. (2025). Assessing eco-physiological patterns of Ailanthus altissima (Mill.) Swingle and differences with native vegetation using Copernicus satellite data on a Mediterranean Island. *Ecological Informatics*, 103080. https://doi.org/10.1016/j.ecoinf.2025.103080

Maun, M. A. (2009). *The Biology of Coastal Sand Dunes*. OUP Oxford.

McLachlan, A., & Defeo, O. (2017). *The ecology of sandy shores*. Academic press.

Meyer, S. T., Koch, C., & Weisser, W. W. (2015). Towards a standardized Rapid Ecosystem Function Assessment (REFA). *Trends in Ecology & Evolution*, *30*(7), 390–397. https://doi.org/10.1016/j.tree.2015.04.006

Millennium Ecosystem Assessment. (2005). *Ecosystems and human well-being* (Vol. 5). Island Press, Washington, DC.

O'Connor, L. M. J., Pollock, L. J., Renaud, J., Verhagen, W., Verburg, P. H., Lavorel, S., Maiorano, L., & Thuiller, W. (2021). Balancing conservation priorities for nature and for people in Europe. *Science*, *372*(6544), 856–860. https://doi.org/10.1126/science.abc4896

Olmo, V., Petruzzellis, F., Alberti, G., Sigura, M., Balbi, S., & Bacaro, G. (2025). Assessing ecosystem services in protected areas: Trade-offs and hotspots in Friuli Venezia Giulia region (northeastern Italy). *Biodiversity*, 1–14. https://doi.org/10.1080/14888386.2025.2460759

Pakeman, R. J., & Quested, H. M. (2007). Sampling plant functional traits: What proportion of the species need to be measured? *Applied Vegetation Science*, *10*(1), 91–96.

Ploton, P., Mortier, F., Réjou-Méchain, M., Barbier, N., Picard, N., Rossi, V., Dormann, C., Cornu, G., Viennois, G., Bayol, N., Lyapustin, A., Gourlet-Fleury, S., & Pélissier, R. (2020). Spatial validation reveals poor predictive performance of large-scale ecological mapping models. *Nature Communications*, *11*(1), 4540. https://doi.org/10.1038/s41467-020-18321-y

Prisco, I., Acosta, A. T. R., & Stanisci, A. (2021). A bridge between tourism and nature conservation: Boardwalks effects on coastal dune vegetation. *Journal of Coastal Conservation*, *25*(1), 14. https://doi.org/10.1007/s11852-021-00809-4

Prisco, I., Angiolini, C., Assini, S., Buffa, G., Gigante, D., Marcenò, C., Sciandrello, S., Villani, M., & Acosta, A. T. R. (2020). Conservation status of Italian coastal dune habitats in the light of the 4th Monitoring Report (92/43/EEC Habitats Directive). *Plant Sociology*, *57*, 55–64. https://doi.org/10.3897/pls2020571/05

Quétier, F., Thébault, A., & Lavorel, S. (2007). Plant Traits in a State and Transition Framework as Markers of Ecosystem Response to Land-Use Change. *Ecological Monographs*, *77*(1), 33–52. https://doi.org/10.1890/06-0054

Quijas, S., Schmid, B., & Balvanera, P. (2010). Plant diversity enhances provision of ecosystem services: A new synthesis. *Basic and Applied Ecology*, *11*(7), 582–593. https://doi.org/10.1016/j.baae.2010.06.009

Ratcliffe, S., Wirth, C., Jucker, T., van der Plas, F., Scherer-Lorenzen, M., Verheyen, K., Allan, E., Benavides, R., Bruelheide, H., Ohse, B., Paquette, A., Ampoorter, E., Bastias, C. C., Bauhus, J., Bonal, D., Bouriaud, O., Bussotti, F., Carnol, M., Castagneyrol, B., … Baeten, L. (2017). Biodiversity and ecosystem functioning relations in European forests depend on environmental context. *Ecology Letters*, *20*(11), 1414–1426. https://doi.org/10.1111/ele.12849

Ricci, L., Musciano, M. D., Sabatini, F. M., Chiarucci, A., Zannini, P., Gatti, R. C., Beierkuhnlein, C., Walentowitz, A., Lawrence, A., Frattaroli, A. R., & Hoffmann, S. (2023). A multitaxonomic assessment of Natura 2000 effectiveness across European biogeographic regions. *Conservation Biology*, cobi.14212. https://doi.org/10.1111/cobi.14212

Richards, S. A. (2005). Testing ecological theory using the information-theoretic approach: Examples and cautionary results. *Ecology*, *86*(10), 2805–2814. https://doi.org/10.1890/05-0074

Roig, F. X., Rodríguez-Perea, A., Martín-Prieto, J. A., & Pons, G. X. (2009). Soft Management of Beach-Dune Systems as a Tool for their Sustainability. *Journal of Coastal Research*, 1284–1288. JSTOR.

Roumet, C., Birouste, M., Picon-Cochard, C., Ghestem, M., Osman, N., Vrignon-Brenas, S., Cao, K., & Stokes, A. (2016). Root structure–function relationships in 74 species: Evidence of a root economics spectrum related to carbon economy. *New Phytologist*, *210*(3), 815–826. https://doi.org/10.1111/nph.13828

Rout, M. E., & Callaway, R. M. (2009). An Invasive Plant Paradox. *Science*, *324*(5928), 734–735. https://doi.org/10.1126/science.1173651

Sarmati, S., Angiolini, C., Sperandii, M. G., Barták, V., Gennai, M., Acosta, A. T. R., Bertacchi, A., Bonari, G., Foggi, B., Maccherini, S., Viciani, D., & Bazzichetto, M. (2025). A complex interplay between natural and anthropogenic factors shapes plant diversity patterns in Mediterranean coastal dunes. *Landscape Ecology*, *40*(1), 20. https://doi.org/10.1007/s10980-024-02025-5

Schlacher, T. A., Dugan, J., Schoeman, D. S., Lastra, M., Jones, A., Scapini, F., McLachlan, A., & Defeo, O. (2007). Sandy beaches at the brink: Sandy beach conservation crisis. *Diversity and Distributions*, *13*(5), 556–560. https://doi.org/10.1111/j.1472-4642.2007.00363.x

Scurlock, J. M. O., Johnson, K., & Olson, R. J. (2002). Estimating net primary productivity from grassland biomass dynamics measurements. *Global Change Biology*, *8*(8), 736–753. https://doi.org/10.1046/j.1365-2486.2002.00512.x

Simberloff, D. (2011). How common are invasion-induced ecosystem impacts? *Biological Invasions*, *13*(5), 1255–1268. https://doi.org/10.1007/s10530-011-9956-3

Sperandii, M. G., Barták, V., & Acosta, A. T. R. (2020). Effectiveness of the Natura 2000 network in conserving Mediterranean coastal dune habitats. *Biological Conservation*, *248*, 108689. https://doi.org/10.1016/j.biocon.2020.108689

Sperandii, M. G., Barták, V., Carboni, M., & Acosta, A. T. R. (2021). Getting the measure of the biodiversity crisis in Mediterranean coastal habitats. *Journal of Ecology*, *109*(3), 1224–1235. https://doi.org/10.1111/1365-2745.13547

Sperandii, M. G., Bazzichetto, M., Gatti, F., & Acosta, A. T. R. (2019). Back into the past: Resurveying random plots to track community changes in Italian coastal dunes. *Ecological Indicators*, *96*, 572–578. https://doi.org/10.1016/j.ecolind.2018.09.039

Sperandii, M. G., De Bello, F., Valencia, E., Götzenberger, L., Bazzichetto, M., Galland, T., E-Vojtkó, A., Conti, L., Adler, P. B., Buckley, H., Danihelka, J., Day, N. J., Dengler, J., Eldridge, D. J., Estiarte, M., García-González, R., Garnier, E., Gómez-García, D., Hallett, L., … Lepš, J. (2022). LOTVS: A global collection of permanent vegetation plots. *Journal of Vegetation Science*, *33*(2), e13115. https://doi.org/10.1111/jvs.13115

Sperandii, M. G., Prisco, I., Stanisci, A., & Acosta, A. (2017). RanVegDunes – A random plot database of Italian coastal dunes. *Phytocoenologia*, *47*. https://doi.org/10.1127/phyto/2017/0198

Spohn, M., Bagchi, S., Biederman, L. A., Borer, E. T., Bråthen, K. A., Bugalho, M. N., Caldeira, M. C., Catford, J. A., Collins, S. L., Eisenhauer, N., Hagenah, N., Haider, S., Hautier, Y., Knops, J. M. H., Koerner, S. E., Laanisto, L., Lekberg, Y., Martina, J. P., Martinson, H., … Yahdjian, L. (2023). The positive effect of plant diversity on soil carbon depends on climate. *Nature Communications*, *14*(1), 6624. https://doi.org/10.1038/s41467-023-42340-0

Stallins, J. A. (2005). Stability domains in barrier island dune systems. *Ecological Complexity*, *2*(4), 410–430. https://doi.org/10.1016/j.ecocom.2005.04.011

Świerkosz, K., & Reczyńska, K. (2022). VESTA – resurvey of natural, non-forest vegetation (Central Europe). *Vegetation Classification and Survey*, *3*, 221–222. https://doi.org/10.3897/VCS.96011

Tsiafouli, M. A., Apostolopoulou, E., Mazaris, A. D., Kallimanis, A. S., Drakou, E. G., & Pantis, J. D. (2013). Human Activities in Natura 2000 Sites: A Highly Diversified Conservation Network. *Environmental Management*, *51*(5), 1025–1033. https://doi.org/10.1007/s00267-013-0036-6

UNEP-WCMC and IUCN. (2025). *Protected Planet: The World Database on Other Effective Area-based Conservation Measures (WD-OECM) [On-line], Cambridge, UK: UNEP-WCMC and IUCN. Available at: Www.protectedplanet.net accessed on 24/03/2025*.

United Nations. (2015). *Transforming our world: The 2030 agenda for sustainable development. United Nations General Assembly, A/RES/70/1. Https://sdgs.un.org/2030agenda*.

Valencia, E., Maestre, F. T., Bagousse-Pinguet, Y. L., Quero, J. L., Tamme, R., Börger, L., García-Gómez, M., & Gross, N. (2015). Functional diversity enhances the resistance of ecosystem multifunctionality to aridity in Mediterranean drylands. *New Phytologist*, *206*(2), 660–671. https://doi.org/10.1111/nph.13268

Van Der Biest, K., De Nocker, L., Provoost, S., Boerema, A., Staes, J., & Meire, P. (2017). Dune dynamics safeguard ecosystem services. *Ocean & Coastal Management*, *149*, 148–158. https://doi.org/10.1016/j.ocecoaman.2017.10.005

Van der Maarel, E. (2003). Some remarks on the functions of European coastal ecosystems. *Phytocoenologia*, *33*, 187–202. https://doi.org/10.1127/0340-269X/2003/0033-0187

van der Plas, F. (2019). Biodiversity and ecosystem functioning in naturally assembled communities. *Biological Reviews*, *94*(4), 1220–1245. https://doi.org/10.1111/brv.12499

Van Der Plas, F., Schröder-Georgi, T., Weigelt, A., Barry, K., Meyer, S., Alzate, A., Barnard, R. L., Buchmann, N., De Kroon, H., Ebeling, A., Eisenhauer, N., Engels, C., Fischer, M., Gleixner, G., Hildebrandt, A., Koller-France, E., Leimer, S., Milcu, A., Mommer, L., … Wirth, C. (2020). Plant traits alone are poor predictors of ecosystem properties and long-term ecosystem functioning. *Nature Ecology & Evolution*, *4*(12), 1602–1611. https://doi.org/10.1038/s41559-020-01316-9

Villéger, S., Mason, N. W. H., & Mouillot, D. (2008). New multidimensional functional diversity indices for a multifaceted framework in functional ecology. *Ecology*, *89*(8), 2290–2301. https://doi.org/10.1890/07-1206.1

Waldén, E., Queiroz, C., Plue, J., & Lindborg, R. (2023). Biodiversity mitigates trade-offs among species functional traits underpinning multiple ecosystem services. *Ecology Letters*, *26*(6), 929–941. https://doi.org/10.1111/ele.14220

Williams, J. W., Seabloom, E. W., Slayback, D., Stoms, D. M., & Viers, J. H. (2005). Anthropogenic impacts upon plant species richness and net primary productivity in California. *Ecology Letters*, *8*(2), 127–137. https://doi.org/10.1111/j.1461-0248.2004.00706.x

Wilsey, B., Martin, L., Xu, X., Isbell, F., & Polley, H. W. (2023). Biodiversity: Net primary productivity relationships are eliminated by invasive species dominance. *Ecology Letters*, ele.14342. https://doi.org/10.1111/ele.14342

Zuur, A. F., Ieno, E. N., & Elphick, C. S. (2010). A protocol for data exploration to avoid common statistical problems: *Data exploration*. *Methods in Ecology and Evolution*, *1*(1), 3–14. https://doi.org/10.1111/j.2041-210X.2009.00001.x