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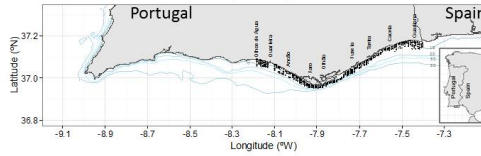
## Abstract

Most marine species are known to be aggregated in spatial patches. Those patches vary through time, both in biomass and location, as a response to a set of drivers. Thus, to understand the ecological, environmental and anthropological drivers of species biomass, long-term spatial analyses of species dynamics are essential. In particular, target species of small-scale fisheries are further influenced by governance and market drivers, with a significant impact on their dynamics, which in turn have considerable economic and social impacts. In the Algarve coast (southern Portugal) an important small-scale fishery targeting four commercial bivalve species takes place on sandy bottoms. In the current work, almost three decades (1986-2014) of data on the biomass of the striped venus clam (*Chamelea gallina*) were analysed using spatial-temporal geostatistical methods. The study area comprises 119 km<sup>2</sup> of coast, distributed between -8.1° to -7.4°W longitude and 36.95° to 37.17°N latitude, and from 3 to 15 m depth. Species biomass was modelled in function of space and time using a spatial-temporal variogram. The model was then used to produce interpolations over the study area and period using spatial-temporal kriging. The striped venus clam (*C. gallina*) showed a simple sum metric covariance spatial-temporal model, explaining 53% of the covariance. The model showed that this bivalve species is aggregated in 2.9 km spatial patches, whose location and dimension varied throughout the sampling period.

Spatial-temporal kriging prediction maps were used to extract the main summarizing statistical features, such as the mean, maximum and variance, for all time-series stacked, thus determining the prevailing favourable areas for this species and corresponding spatial patterns of variation through time. The main spatial-temporal patterns of variation were further explored using Empirical Orthogonal Function analysis (EOF), which is a particular type of principal component analysis that decomposes the spatial-temporal variability in the time series into principal spatial orthogonal components. This analysis allows identifying the main types of spatial distribution and temporal evolution of a given variable.

The stock estimates produced and the corresponding spatial distribution will be used in the framework of the European research project SAFI (<http://www.safiservices.eu/>; REA Grant Agreement 607155), that aims to develop indicators between species and aspects of their environment to support fisheries operations and help in management decisions, thus contributing for the sustainable exploitation of the living marine resources by using satellite-derived information.

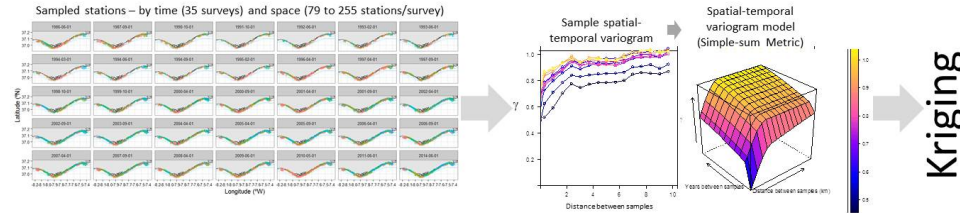
## Location & sampling



- **Temporal coverage:** 1986-2014 (28 years)
- **Bathymetry:** 3-15m
- **Substrate:** Sandy bottoms – fine-sand/gravel
- **Area covered:** Olhos d' Água to the river Guadiana
- **Aim:** IPMA's regular monitoring fishing surveys for estimation of the status of the stocks of commercial bivalves
- **Sampling stations:** 92 transects perpendicular to the coast, crossed at crossed at eight bathymetric levels: 3 m, 4.8 m, 6.6 m, 8.4 m, 10.2 m, 12 m, 13.8 m and 15.6 m.
- **Sampling methods:** bivalve dredge towed for 5 minutes at a constant speed of 1.5 knots, sweeping an area of 144m<sup>2</sup>



## Spatial-temporal analyses



## Examples of results that can be achieved with the kriging predictions

