Modelling seabed sediment physical properties and organic matter content in the Firth of Clyde

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Abstract. High quality quantitative maps of seabed sedimentary physical and geochemical properties have numerous research and conservation applications, including habitat and ecosystem modelling, marine spatial planning and ecosystem service mapping. However, such maps are lacking for many ecologically and economically important marine areas. Using legacy data supplemented by measurements from recent benthic surveys, modelled hydrodynamic variables and ~~high~~ high-resolution bathymetry, quantitative maps for the top 10 cm of seabed sediment were generated via a combination of statistical and machine-learning techniques for the Firth of Clyde, a semi-enclosed coastal sea on the west coast of Scotland. The maps include sediment fractions of mud, sand and gravel, whole-sediment median grain size, sediment permeability and porosity, rates of natural seabed abrasion, and sediment particulate organic carbon and nitrogen content. Properties were mapped over an unstructured grid, so that very high resolutions were achieved close to the coastlines, where sediments may be expected to be spatially heterogeneous.

Overall, the maps reveal extensive areas of very low sediment permeability coupled with low rates of natural seabed disturbance. Moreover, muddy sediments in the inner Firth of Clyde, Inchmarnock Water and the sea lochs are enriched in organic carbon and nitrogen relative to the sediments of the outer Firth of Clyde. As a demonstration of the value of these maps, the standing stock of organic carbon and nitrogen in the surficial sediments of the Clyde was calculated. The Clyde stores 3.42 and 0.33 million tonnes of organic carbon and nitrogen in the top 10 cm of seabed sediment, respectively, substantially contributing to Scotland's coastal and shelf blue carbon stocks. Data products are available from: https://doi.org/10.15129/2003faa2-ee93-4c11-bb16-48485f5f136d.

1   Introduction

Seabed surficial sediment properties are a defining feature of shelf-sea ecosystems, determining organic matter storage, nutrient recycling and contaminant degradation, turbidity, habitat availability for, and productivity of benthos and fish (Ehrenhauss et al., 2004; Janssen et al., 2005; Kamann et al., 2007; Neumann et al., 2017b; Serpetti et al., 2016). Maps of

sediment properties are therefore important for marine planning and conservation (e.g. Schiele et al., 2015), and assessments of marine service provision (Cowling et al., 2008; Townsend et al., 2014), especially in coastal waters where anthropogenic disturbance is often greatest (Eigaard et al., 2017) and sediments can be extremely heterogeneous (Serpetti et al., 2011).

35   However, maps of adequate resolution are lacking for many regions.

For many decades the standard procedure for characterising seabed sediment samples involved sieving through standard meshes to separate fractions designated mud (sometimes divided into silt and clay), sand and gravel. The proportions by weight of these fractions were then used to classify the sediment usually according to either the Wentworth or Folk scheme (Folk, 1954; Wentworth, 1922). In reality the grain size boundaries between these three fractions are arbitrary, the classification

40   schemes themselves are a compromise between simplicity and detail, and they were originally designed for specific purposes – for example the Folk scheme was designed to resolve the importance of transport on sediment grain distributions (Valentine, 2019).

Recent seabed sediment samples are often analysed by laser diffraction instruments which provide a finely resolved size distribution of grain sizes based on particle volume. Summary measurements of sediment properties from these These

45   instruments offer measurements of sediment propertiesare typically reported on a continuous scale rather than by class, e.g. median grain size, skewness, kurtosis. These properties can be related to other more-difficult-to-measure features such as porosity, permeability and organic matter content. However, it is often the case that laser size distributions are aggregated to the traditional sieve based mud, sand and gravel size classes so that sediments can be classified according to the traditional Folk or Wentworth schemes, or similar (e.g. Lepland et al., 2014), and the continuous variable data are not reported. Note

50   however that there may not be a simple 1:1 relationship between laser size classes and sieve-based classes (Fisher et al., 2017; Wilson et al., 2018).

Most seabed sediment maps display classification data, e.g. https://www.bgs.ac.uk/data/services/offprodwms.html; http://www.mareano.no/kart/mareano_en.html?language=en. The reason is that for the main part they rely on sample data accumulated over many decades, most of which is class-based. However, classification maps have a number of limitations. In

55   particular, they are not easily scalable to finer resolution. For example the EMODnet seabed habitat classification map is gridded at a resolution of approximately 100 m (https://www.emodnet-seabedhabitats.eu/), but the sediment class descriptors are so broad that areas of shelf and coastal seabed are portrayed as homogeneous at scales of 10's of km. In reality, there may be important heterogeneity in whole-sediment grain size compositions which cannot be resolved by the classification scheme. Misiuk et al. (2019) compared and contrasted categorical and continuous variable approaches to mapping seabed sediments

60   and for one particular case study found that both had a similar performance. However, while this may be true generally, it remains the case that it is extremely difficult to meaningfully recover whole-sediment continuous properties such as median grain size, or mud content from class-based data. These properties are the key to imputing other hard-to-measure variables such as carbon content or permeability, based on statistical relationships developed from more limited sampling. Several authors have attempted to extract estimates of such variables from categorical data (Burrows et al., 2014; Smeaton et al., 2020),

65 but this approach may lead to over- or under-predictions especially for properties like permeability which vary by several orders of magnitude within a single grain size class (Serpetti et al., 2016).

Sediment carbon inventories are a case in point with regard to class-based data. Smeaton et al. (2020) estimated the marine sedimentary carbon stock in Scottish shelf seas from the British Geological Survey 1:25000 seabed sediment classification, but resorted to assigning a mean carbon content to each class based on statistical summaries by Diesing et al. (2017) for

70 samples collected only in the North Sea (Diesing et al., 2017), and integrating the areas of each sediment class. Diesing et al. (2017) were the first to apply Random Forests algorithms (Breiman, 2001) to the prediction of sediment organic carbon content, relying on categorical Folk sediment classes to extrapolate predictions and estimate carbon stocks within the NW European shelf. More recently, Smeaton et al. (2021) combined geostatistical models and sedimentary carbon statistical summaries to map carbon stocks across the UK Exclusive Economic Zone (EEZ). Their approach leveraged a large volume of available

75 marine data and delineated characteristic sedimentary organic carbon content differences in coastal and offshore areas yet retained a dependence on mapping to sediment classes. In contrast, using methods similar to Stephens and Diesing (2015), Wilson et al. (2018) used Random Forest algorithms to generate synthetic maps of sediment carbon content across the whole NW European shelf by blending interpolated and predicted maps (0.125-degree resolution - approximately 10km). Interpolation was confined to the alphahull – a convex shape defined around a set -of raw samples assembled from international

80 databases within which sample density exceeded a given threshold. Outside this hull the distributions were predicted by a Random Forest with inputs of bathymetric and hydrodynamic data.

Wilson et al. (2018) also used the Random Forest approach to map mud, sand and gravel compositions, porosity, permeability, and the probability of encountering exposed bedrock. Mitchell et al. (2019) used a similar approach to modelling mud, sand and gravel composition at higher spatial resolution (approximately 0.2 km). Data products from studies such as these have

85 proved to be extremely valuable and have been used in a range of different applications including biodiversity - ecosystem functioning relationships (Maureaud et al., 2019), benthic community sensitivity to trawling disturbance (Rijnsdorp et al., 2018), and the economic cost of disrupting sedimentary carbon storage (Avelar et al., 2017; Luisetti et al., 2019, 2020).

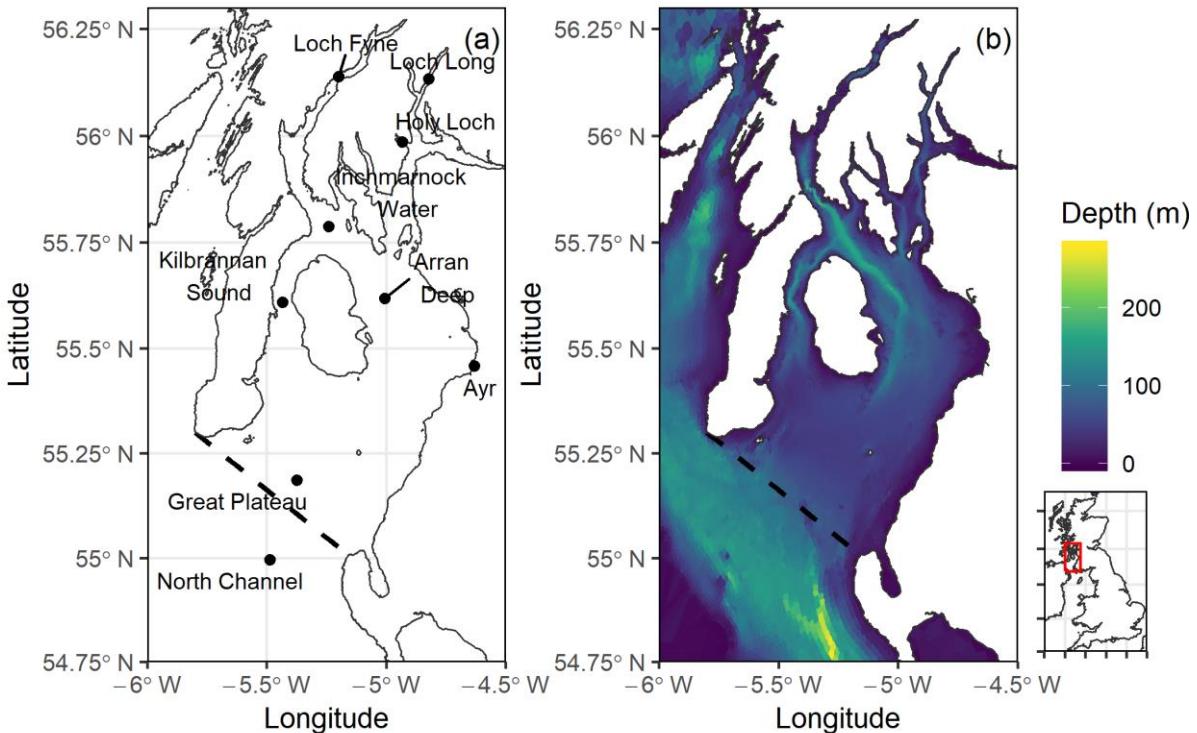
Here, the approach taken by Wilson et al. (2018) was followed and applied at very high spatial resolution (down to 50 m) in a semi-enclosed coastal sea (Firth of Clyde, SW Scotland) where there are strong gradients of bathymetry, tidally dominated

90 hydrodynamics, and sediment properties. The Firth of Clyde and associated sea-lochs feature several deep post-glacial basins lined with very fine, organic-rich mud which has been heavily impacted by trawling (McIntyre et al., 2012). To-date, sediments in the Clyde have only been mapped at a coarse-resolution or as part of larger continental shelf mapping exercises (e.g. Diesing et al., 2017; Mitchell et al., 2019; Smeaton et al., 2021; Wilson et al., 2018). In the following sections, the compilation of legacy data and the collection of additional field samples and their laboratory analysis is first reported. Subsequently, statistical

95 modelling methods used to generate the maps from the source data and the subsequent validation of these data products is described. Finally, an application of these data products to calculate the standing stock of organic carbon and nitrogen in the surficial sediment of the Firth of Clyde is presented and the implications for future research discussed.

2 Methods

2.1 Study area

100 The scope of this study is the generation of high resolution full-coverage maps for sediment fractions of mud, sand and gravel, whole-sediment median grain size, sediment permeability and porosity, rates of natural seabed abrasion, and sediment particulate organic carbon and nitrogen content in the Firth of Clyde. The Firth of Clyde is situated on the west coast of Scotland and covers an area of approximately 3600 km^2 (Fig. 1).

105 **Figure 1 Maps of the Firth of Clyde showing (a) locations referred to in text and, (b) bathymetry mapped over an unstructured grid. Seabed depth is shown in metres and the dashed line shoes the outer geographical limits of the Firth of Clyde. The location of the Firth of Clyde on the west coast of Scotland is shown inset. Bathymetry data source: Sabatino et al. (2016).**

2.2 Data centres and legacy data

The compilation and initial processing of data from the Firth of Clyde is outlined in the following sections. Where
110 measurements of key variables for the Firth of Clyde were lacking or sparse, measurements were supplemented with data from surrounding coastal seas. The sources of data used in the analyses described below are listed in Table 1.

Table 1 Sources of seabed data from data centres and published literature, and the number of records used from each source. Sources include British ~~Geographical~~ Geological Survey (BGS), Marine Scotland Science (MSS), Scottish Environmental Protection Agency (SEPA) and the Centre for Environment, Fisheries and Aquaculture Science (CEFAS).

Property	Symbol	Units	Data source	Number of records	Temporal coverage	Region	Definition
Presence of hard substrate	-	-	BGS	-	-	Firth of Clyde	Presence of rocky outcrops and hard substrate within 0.5 m of the seabed
Sediment grain size fractions	-	%	BGS	930	1969 - 1980	Firth of Clyde	Percentage by volume of mud, sand and gravel fractions of dried sediment
			MSS	447	1997 - 2011	Firth of Clyde	
			SEPA	77	2005 - 2016	Firth of Clyde	
Median grain size	D_{50}	mm	CEFAS	1873	1998 - 2010	Celtic Sea, English Channel, North Sea	Grain diameter that is larger than 50% of other particles
			MSS	1214	2003 - 2004	North Sea	
Permeability	κ	m^2	Serpetti et al. (2016)	163	2008 - 2009	North Sea	Connectedness of fluid-filled pore-space within the sediment
Porosity	ϕ	%	Lohse et al. (1993)	7	1991 - 1992	North Sea	Proportion of pore-space volume within the sediment
			Ruardij and van Raaphorst (1995)	49	-	North Sea	
			Serpetti et al. (2016)	166	2008 - 2009	North Sea	
Particulate organic carbon	POC	%, mg g^{-1} dw	SEPA	168	2005 – 2006	Firth of Clyde	Particulate organic carbon in milligrams per gram dried sediment
Particulate organic nitrogen	PON	%, mg g^{-1} dw	SEPA	118	2005 - 2006	Firth of Clyde	Particulate organic nitrogen in milligrams per gram dried sediment

2.2.1 Rock distribution

Although rock outcrops and hard substrata cover a relatively small proportion of the total seabed area in the Firth of Clyde, they play a functionally distinct role as benthic habitats and in the cycling of nutrients within marine ecosystems. Hence, for a complete characterisation of seabed properties, the distinction between areas of sediment and hard substrate is important. The most extensive available dataset on the distribution of bare-rock and hard substrate features in Scottish coastal waters was compiled by the British ~~Geographical~~ Geological Survey (BGS) from the interpretation of archived acoustic data and seabed samples (http://www.bgs.ac.uk/discoverymetadata/13605550.html). These data were obtained as polygons representing areas of hard substrate, defined by BGS as rock and sediments with a grain size of larger than 64 mm within 0.5 m of the seabed (Gafeira et al., 2010). Additional point observations of seabed substrate in the Clyde estuary were digitised from the Admiralty chart no. 2007 covering the River Clyde.

2.2.2 Sediment grain size fractions

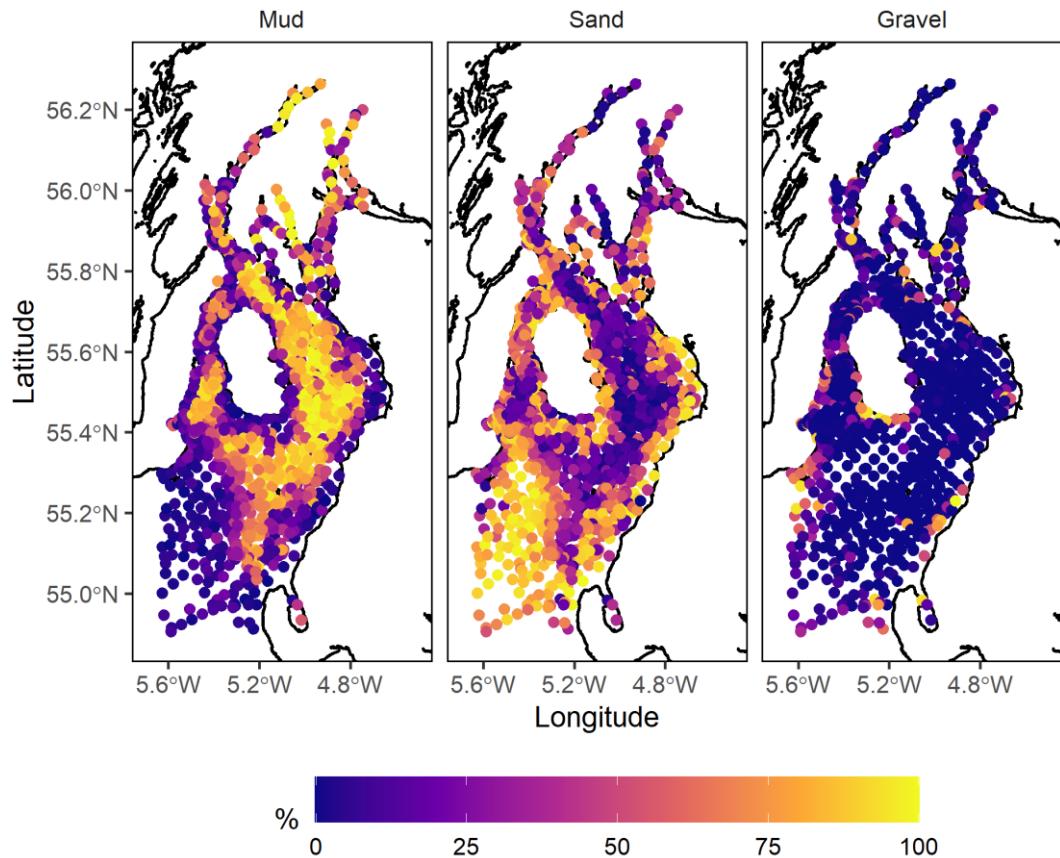
Data on the percentages mud, sand and gravel in Firth of Clyde sediments from BGS and Marine Scotland were inherited from Wilson et al. (2018), and additional legacy measurements were requested from the Scottish Environmental Protection Agency (SEPA). Marine Scotland data comprised percentages of mud and sand, excluding the gravel fraction. BGS data were derived from samples collected during cruises carried out in the late 1960s and 1970s with detailed information on sample collection and processing in Chesher et al. (1972) and Deegan et al. (1973). SEPA measurements were made from grab samples collected annually from a network of 18 stations to fulfil the Clean Safe Seas Environmental Monitoring Programme (CSEMP) reporting obligations as well as additional surveys. Sampled locations were well-distributed across the Firth of Clyde and in the associated sea lochs with a high density of points in the outer Firth (Fig. 2).

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Figure 2 Locations sampled by BGS, MSS and SEPA for the measurement of percentage mud, sand and gravel in seabed sediments in the Firth of Clyde.

The data were filtered to retain only grab samples and maintain a consistent sampling method. Analysis techniques differed between sources. Sediment fractions reported by BGS were measured from sieved samples, whereas Marine Scotland and

140 SEPA data were obtained via laser diffraction analysis. Inconsistency between measurements derived from laser and sieve methods have long been reported in the literature. A correction formula by Wilson et al. (2018) was used to correct sieve diffraction-derived mud and sand fraction measurements in samples containing less than 60% mud to laser-measured equivalents. The numbers of records used in the analysis from each data source are shown in Table 1.

2.2.3   Median grain size

145 Whole-sediment median grain size data for the Firth of Clyde are sparse. Marine Scotland sediment data for the Firth of Clyde include measurements of median grain size for only the combined mud and sand fractions and these were therefore omitted from the analysis. Fresh measurements of median grain size were carried out from samples collected during recent research cruises in the Firth of Clyde (section 2.3); yet these data were insufficient for the mapping predicted median grain size from

environmental variables. However, whole-sediment median grain size is related to the proportions of mud, sand and gravel

150   fractions (Wilson et al., 2018) permitting prediction of this property from full-coverage sediment grain size fraction maps, given an appropriate fitted statistical relationship. Hence, to fit a statistical model predicting the whole-sediment median grain size for a wide range of marine sediments, the sparse measurements of whole-sediment median grain size and fractions mud, sand and gravel in the Firth of Clyde were supplemented by data comprising these variables and collected by the Centre for Environment, Fisheries and Aquaculture Science (CEFAS) from the Celtic Sea, English Channel and North Sea, and data

155   collected by Marine Scotland from the North Sea were used (Fig. 3).

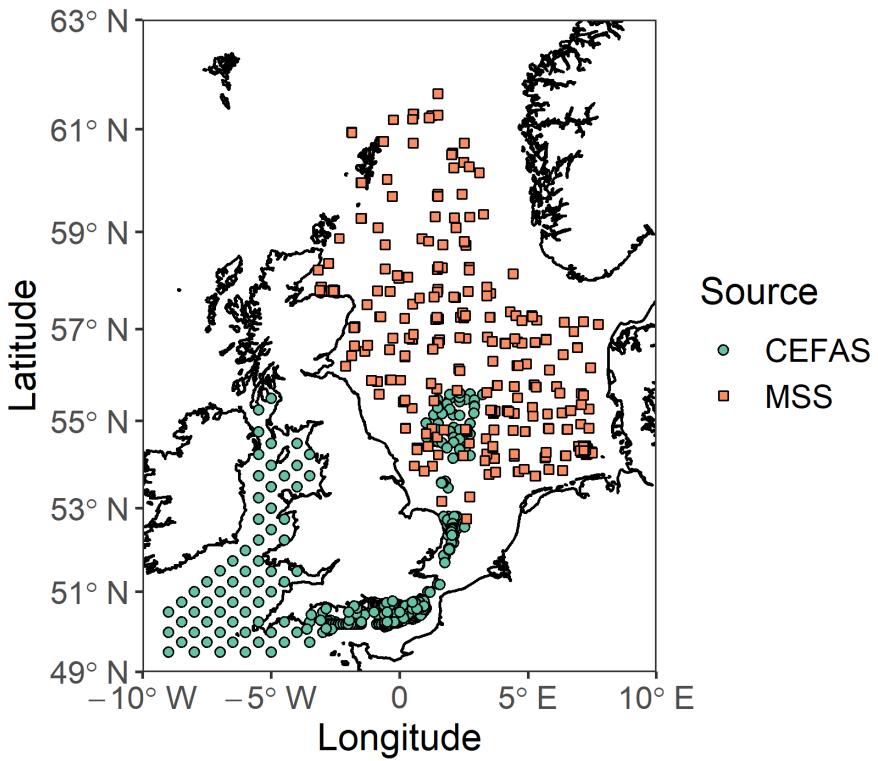



Figure 3 Sampling locations for sediment data obtained from CEFAS and MSS for the statistical modelling of relationships between median grain size and fractions of mud, sand and gravel.

2.2.4   Permeability and porosity

160   Measurements of seabed sediment permeability and porosity in the Firth of Clyde are lacking, and these properties were prioritised in the collection of additional field data. (see section 2.3). Given the difficulty of carrying out permeability measurements on fresh samples of undisturbed seabed sediment (Wilson et al., 2008), there are few published datasets on the

permeability of marine sediments available globally. Extensive work has been carried out to empirically relate permeability to measurements of grain size distribution for a range of sediment types (Qi et al., 2015; see review and meta-analysis by

165   Shepherd, 1989). Recent parameterisations of empirical expressions predicting marine sediment permeability from grain size variables have been carried out by Wilson et al. (2018) and Neumann et al. (2017a). However, there remains considerable uncertainty regarding the suitability of these expressions for predicting the permeability of muddy seabed given the lack of data for fine-grained sediments. Similarly, Ssediment porosity is inversely related to grain size (Moore and Keller, 1984), and several authors have derived expressions relating sediment porosity to median grain size and mud content in natural marine

170   sediments (e.g. Silburn et al., 2017; Wilson et al., 2018). Therefore, sediment permeability and porosity data from neighbouring coastal seas were used to supplement fresh measurements for the Firth of Clyde and parameterise relationships that extended to the finer-grained sediments of the Clyde. The best available dataset on seabed sediment permeability was collected by Serpetti et al. (2016) for seven stations off Stonehaven in the North Sea between 2008 and 2009. This spans 163 measurements of permeability, median grain size and percentages of mud, sand and gravel in the top 10 cm of seabed sediment from

175   undisturbed cores collected from a range of sediment types.
Sediment porosity is inversely related to grain size (Moore and Keller, 1984). Sediment porosity and median grain size data for the North Sea representative of the top 10 cm of sediment and published by Lohse et al (1993), Ruardij and van Raaphorst (1995) and Serpetti et al (2016) were used.

2.2.5   Fishing abrasion pressure

180   In recent years, the use of seabed swept-area ratio (SAR) as a quantitative metric of seabed fishing pressure has gained traction in the published literature (Bolam et al., 2017; e.g. Hiddink et al., 2016; Rijnsdorp et al., 2016). SAR is the ratio between the total area swept by the gears of a fishing fleet and the area of grid cell that the values were aggregated over (Eigaard et al., 2016, 2017). To briefly summarise the process of computing SAR, Vessel Monitoring System (VMS) positions are merged with logbook data to obtain locations of fishing operations. These are then combined with footprint estimates for given gear

185   arrangements to obtain the area swept by the gear and then aggregated over a regular grid.
Data for the seabed-surface swept-area ratio calculated over a regular grid at a resolution of 0.05 ° x 0.05 ° was obtained from the Joint Nature Conservation Committee (JNCC) for the period 2009 – 2016. These data cover UK vessels only (D. Edwards, pers. comm.), and it is therefore assumed that the number of non-UK vessels fishing in the Clyde is negligible. Given that these data were already spatially gridded, a climatology of mean annual swept area ratio was compiled (Fig. 4).

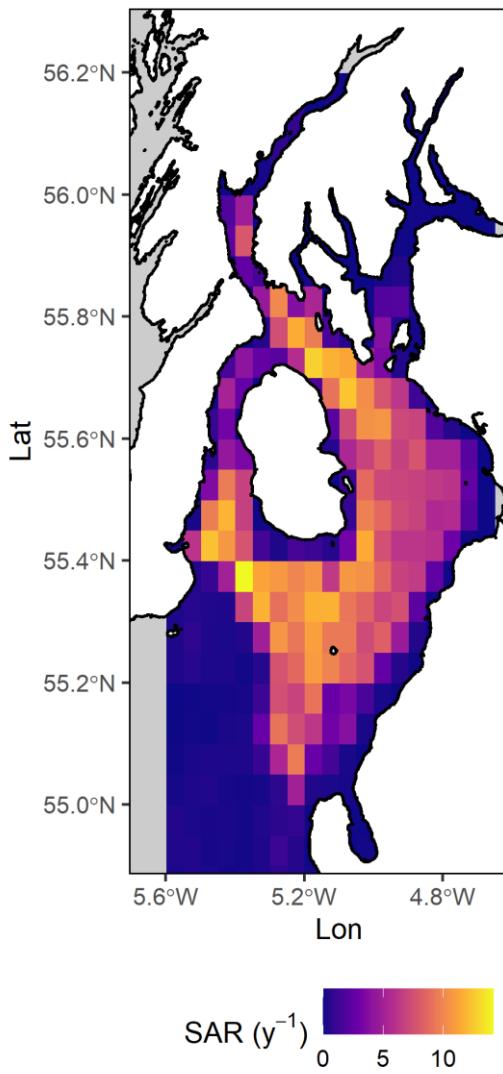



Figure 4 Mean annual swept area ratio (SAR) for the years 2009 - 2016. Spatial distribution of seabed abrasion pressure from fishing vessels with an overall length greater than 12 m. Areas where no data is available are shown in grey.

2.2.6   Seabed organic matter content

Data on sediment organic carbon and nitrogen content and sediment fractions of mud, sand and gravel were requested from SEPA for the Firth of Clyde. SEPA measurements of organic carbon and nitrogen content were taken from the <63 μm grain size fraction rather than the whole sediment. Using these values directly would over-predict organic carbon and nitrogen as organic matter is expected to be overwhelmingly associated with the mud fraction (Wiesner et al., 1990).

The organic matter content associated with the gravel fraction (> 2000 µm) was assumed to be negligible and the average density of carbon and nitrogen associated with the sand fraction for a range of marine sediments was calculated (see Appendix A). These estimates were then used to scale SEPA measurements of organic carbon and nitrogen to the whole sediment.

2.3 Additional data collection

Research cruises in the Firth of Clyde were carried out between 24 – 28 April and 13 - 17 October 2017 for the collection of additional seabed sediment geotechnical and nutrient content data. Respectively, 35 stations in the outer Firth of Clyde and 14 stations in the inner Firth of Clyde and sea lochs were sampled over the course of the two cruises. Measurements of seabed sediment grain size, and particulate organic carbon and nitrogen were carried out on seabed sediment samples collected using a 0.1 m² Day grab. Sediment subsamples were taken from each grab sample using a 5 by 3.5 cm acrylic minicore tube and frozen immediately following collection aboard the research vessel at -18 °C in sealable polyethylene bags for subsequent laboratory analysis. Sediment particulate organic carbon and nitrogen content were measured from freeze-dried Day grab subsamples using an EAS Costech 4020 Elemental Combustion Analyser. The remainder of the freeze-dried subsample was sieved for 10 minutes through a stack of 11.20, 8.00, 5.56, 3.35, 2.36, 2.00- and 1.40-mm mesh sieves mounted on a sieve shaker. The sediment fractions retained on each sieve were weighed to the nearest 0.1 milligram and further analysed using a Beckman Coulter LS230 Laser Diffraction Particle Size Analyser. The sieve and laser diffraction data were blended and summary statistics, including median grain size, % fractions of mud, sand and gravel, skewness and kurtosis, were calculated using GRADISTAT v8 (Blott and Pye, 2001).

Measurements of sediment permeability, porosity and grain size were made from undisturbed seabed sediment cores sampled using a Craib corer (Craib, 1965). Following collection, seabed sediment cores were refrigerated in the dark with loosely fitted caps, and analysis for permeability was carried out within eight hours of collection. Cores were trimmed to the top 10 cm of seabed sediment prior to analysis. The hydraulic conductivity (K) of the top 10 cm of seabed sediment was measured via the falling head method (Klute and Dirksen, 1986). The decline in water level between two different hydraulic heads was recorded over a 2-hour interval and the hydraulic conductivity was calculated via:

$$K = \log_e \left(\frac{H_1}{H_2} \right) \frac{a\,L}{A\,t} \tag{1}$$

where t is the time (s) taken for the level of the percolating fluid to decline between hydraulic head differences H_1 and H_2 (m), L and A are the length (m) and cross-sectional area (m²) of the sediment sample respectively, and a is the cross-sectional area (m²) of the water reservoir. The intrinsic permeability is related to the hydraulic conductivity via the expression:

$$\kappa = \frac{K\,\eta}{\rho\,g} \tag{2}$$

where g is the acceleration due to gravity (9.81 m s⁻²), ρ is fluid density (1027 kg m⁻³) and η is dynamic viscosity of the fluid (1.48×10^{-3} kg m⁻¹ s⁻¹). Following the measurement of hydraulic conductivity, sediment cores were extruded from the acrylic tubes, sealed in polythene bags and frozen at -12 °C for subsequent porosity and grain size analysis. Sediment was weighed

before and after drying in an oven at 105 °C for 24 hours (Moore and Keller, 1984), and sediment porosity was calculated for each core slice via:

$$\phi = \frac{V_w}{V_T} \tag{3}$$

where V_T is the volume of saturated sediment (cm^3) calculated from the thickness of the slice and the internal cross-sectional area of the acrylic core tube, and V_w is the pore volume calculated with the expression:

$$V_w = \frac{ww - dw}{\rho} \tag{4}$$

where dw and ww are the dry and wet weights (g) of the sample, and ρ is the water density taken to be 1 g cm^{-3} given that solutes are retained in the dried sediment. Grain size analysis was then carried out as previously described for sediment grab samples.

2.4   Mapping seabed properties

All data processing and statistical modelling was carried out using the R statistical program (R Core Team, 2019). Several seabed properties were modelled using Random Forest models, which outperformed generalised additive and linear models in the prediction of sediment grain size fractions and organic carbon and nitrogen content. Random Forests models were fitted in the present study using the *ranger* package (v0.11.2) (Wright and Ziegler, 2017). It has the advantage of capturing complex non-linear relationships in the data and reducing the variance in model prediction, whilst being relatively insensitive to irrelevant explanatory variables (Hastie et al., 2009). Random Forest models have been applied to numerous marine mapping challenges, including the distribution of vulnerable benthic species (Portela et al., 2015), global patterns in seafloor biomass (Wei et al., 2010), the diversity and biomass of reef fish communities (Knudby et al., 2010), and the particle size fractions (Li et al., 2011; Stephens and Diesing, 2015) and organic carbon content (Diesing et al., 2017; Wilson et al., 2018) of seabed sediments.

Although implementations of Random Forest models typically carry out an internal cross-validation using 'out-of-bag' data, this does not account for spatial autocorrelation. To provide a more robust estimate of model prediction accuracy, a spatial cross-validation was run where data are binned at a resolution of 0.125 °Latitude and 0.25 °Longitude, increments of approximately 13.9 and 15.75 km respectively, into spatial blocks. A third of the data ~~bins~~ blocks are randomly assigned as test data, the model is trained on the remaining data and evaluated against the test data. This process is repeated, and the model ~~prediction~~ performance ~~error~~metric is averaged between successive iterations until the error stabilises. The suitability of the binning resolution was determined for each model separately by evaluating residual autocorrelation in semivariograms generated for models fitted to the full data, following recommendations by Roberts et al. (2017). One thousand iterations were used for the cross-validation of rock and sediment grain size fraction models, whereas 2000 iterations were used for the organic carbon and nitrogen models.

A suite of environmental variables were selected as candidate predictors of seabed sedimentology and organic matter based on their expected importance in the Firth of Clyde, evidence from published research, and availability. The grain size

distribution of surficial seabed sediments may be predicted by topographic features such as bathymetry ~~to predict seabed properties, and capturing aspects of bathymetry~~ (Bockelmann et al., 2018), seabed ~~topography~~roughness and distance to the coast (Stephens and Diesing, 2015; Wilson et al., 2018). ~~and hydrodynamic flow~~Tidal current speeds and bed shear stress are important drivers of sediment mobilisation and transport (Soulsby, 1997).~~.~~These variables are also predictors of seabed organic

265   carbon (Diesing et al., 2017), in part due to the strong relationship between the sediment grain size and organic matter content (Wiesner et al., 1990). Environmental variables were calculated from a high-resolution unstructured-grid Finite-Volume Community Ocean Model (FVCOM) hydrodynamic model configured for the Firth of Clyde (Sabatino et al., 2016). The model was forced using a combination of tidal, atmospheric, meteorological and hydrological data to capture tidal, wind and density-driven flow in the Firth of Clyde for the years 2005 and 2006, generating hourly estimates of current velocity over an irregular

270   grid of triangular cells that provides higher spatial resolution close to the coast. A detailed description of the calibration and validation of the Clyde FVCOM hydrodynamic model is given by Sabatino et al. (2016). The full list of explanatory environmental variables used to predict seabed physical properties and sedimentary organic matter is given in Table 2.

Table 2 Explanatory environmental variables used to predict seabed physical properties and particulate organic carbon (POC) and nitrogen (PON) content. For each predicted sedimentary property, the explanatory variables used are listed (x). In addition, for
275   **properties where model selection was carried out, the explanatory variables that were evaluated but not retained in the optimal model are also listed (o).**~~The annual minimum, mean and maximum modelled sea surface salinity and the mean annual swept area ratio by the fishing fleet were evaluated as explanatory variables for POC and PON but were not retained in the optimal model.~~

Variable name	Units	Rock	Grain size fractions	Median grain size	Permeability and porosity	POC	PON
Latitude	°		x				
Longitude	°		x				
Depth	m	x	x			o	o
Topographic roughness	-	x	x			x	o
Seabed slope	-	x	x			o	o
Standard deviation of slope	-	x	x			o	o
Distance to coastline	m	x	x			x	x
Annual minimum current speed	m s^{-1}	x	x			o	o
Annual mean current speed	m s^{-1}	x	x			o	o
Annual maximum current speed	m s^{-1}	x	x			o	o
Annual minimum bed shear stress	N m^{-2}	x	x			o	o
Annual mean bed shear stress	N m^{-2}	x	x			o	x
Annual maximum bed shear stress	N m^{-2}	x	x			x	x
Mean annual swept area ratio	-					o	o
Annual minimum surface salinity	-					o	o
Annual mean surface salinity	-					o	o
Annual maximum surface salinity	-					o	o

Mud content	%	x	x		x	x
Sand content	%	x				
Gravel content	%	x				
Median grain size	mm		x			

Distance to coastline is taken to be the minimum Euclidean distance in metres between a given point and any point on the
280   coastline and was calculated using high-resolution polygons of the Scottish coast and islands and the *dist2Line* function from
the *geosphere* package (v1.5-10) (Hijmans, 2019). The annual minimum, mean and maximum bed shear stress and the annual
minimum, mean and maximum depth-averaged current speed were calculated from predicted hourly values extracted from the
Firth of Clyde hydrodynamic model run for two consecutive years (2005 and 2006). The seabed slope, standard deviation of
the slope and topographic roughness were calculated following the methods used by Wilson et al. (2018). Seabed slope was
285   calculated with the *slope* function from the *SDMTools* package (v1.1-221.1) (Van Der Wal et al., 2019), and the standard
deviation of the slope and topographic roughness were calculated from slope and bathymetry respectively following the method
of Cavalli et al. (2008).

These explanatory variables were mapped to an irregular grid of 39449 points with a spatial arrangement that corresponded to
the locations of velocity estimates in the FVCOM unstructured grid. This grid provided the basis for the prediction of modelled
290   sedimentary properties, whilstThis eliminated eliminating the need to interpolate depth-averaged current velocity and bed
shear stress outputs from the hydrodynamic model to a different grid arrangement. Explanatory variables were then
interpolated to sample and legacy data locations via pointwise bilinear interpolation using the *akima* package (Akima and
Gebhardt, 2020). Given the irregular arrangement of grid points, predicted maps are visualised using Dirichlet cells generated
from the packages *ggvoronoi* (Garrett et al., 2019) and *deldir* (v0.1-16) (Turner, 2019). Maps and figures are generated in R
295   using the *ggplot2* package (Wickham, 2016). Figure 5 illustrates how mapped explanatory variables and empirical data
informed sedimentary models, and how fitted models were then used to map out seabed sediment properties.

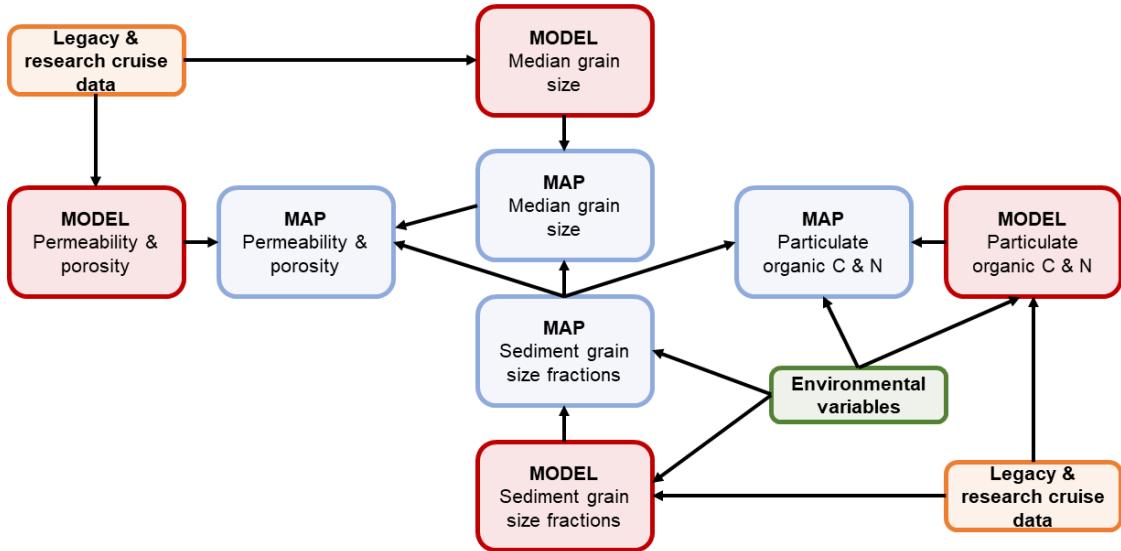



Figure 5 Illustration of the workflow described in this chapter. Legacy and research cruise data (orange boxes) were used in conjunction with mapped environmental variables (green boxes) to fit models of seabed sediment grain size and geotechnical properties and organic carbon and nitrogen content. Full-coverage quantitative maps of these properties (blue boxes) were generated by using mapped environmental variables and sedimentary maps created earlier in the workflow as the substrate from which to generate predictions from the fitted models.

2.4.1   Rock distribution

Coverage of BGS data did not extend into the Clyde estuary and sea lochs, and a Random Forest binary classification model was used to predict the distribution of hard substrate in these areas. To assemble a training data set, the irregular grid points containing the explanatory variables were classified as rock or non-rock within areas covered by BGS data. These were augmented by rock outcrop observations in the Clyde estuary digitised from Admiralty Charts. The environmental variables used to predict the presence of rock substrate are given in Table 2. Model performance was assessed via spatial cross-validation with the mean square error (MSE) of the averaged classification tree predictions, and the overall accuracy, sensitivity and specificity of predictions used as performance metrics following recommendations by Lawson et al. (2014). Prediction sensitivity refers to the percentage of rock observations that were correctly classified as rock, whereas specificity refers to the percentage of non-rock observations that were correctly predicted to have no rock present. A total of 39 spatial blocks were generated, with an average of 596 observations in each block. Given the sparsity of rock observations, the bootstrapping sampling of training and test data was constrained so that at least one instance of rock is present in both subsets. One thousand iteratively sampled training and test datasets were generated for spatial cross-validation. Further detail on the code used to implement spatial cross-validation may be found in Supplementary Material.

2.4.2 Sediment grain size fractions

Sediment particle size fractions were mapped in terms of the proportion of mud, sand and gravel. As compositional data carrying relative rather than absolute information, it is possible to capitalise on the fact that the whole composition information may be retrieved from n-1 components of the data (Aitchison, 2003). Following methods given by Stephens and Diesing (2015), a coupled additive log-ratio transformation (Eq. 5 and Eq. 6) was used to couple the sediment fractions and ensure that the proportions summed to 1.

$$\log\left(\frac{sand}{mud}\right) = \log(sand) - \log(mud) \tag{5}$$

$$\log\left(\frac{gravel}{mud}\right) = \log(gravel) - \log(mud) \tag{6}$$

This approach had two important benefits. Firstly, two, rather than three, models were needed to predict the complete compositional information of sediment grain size fractions. Secondly, it was possible to accommodate Marine Scotland data, where gravel fraction measurements were lacking, into the analysis without information loss by using mud as the denominator within the transformation. This is because spatial predictions were carried out for the separate models and only subsequently back-transformed. Mud was used as the denominator within this transformation to accommodate MSS data that lacked measurements of the gravel fraction.

Random Forest models were used to predict the log-ratios of sand to mud and gravel to mud from environmental variables listed in Table 2. Back-transformation of log-ratios into separate sediment fractions was carried out via a multinomial logit function. Spatial cross-validation of the model was carried out to test the accuracy of back-transformed predictions of sediment mud, sand and gravel fractions. A total of 40 spatial blocks were generated with an average of 24 samples within each block. The mean R^2 metric across cross-validation iterations computed through the *caret* package (v6.0.86)(Kuhn, 2020) was used to assess model performance. This metric takes the popular and robust formulation: $R^2 = 1 - (\sum(y - \hat{y})^2)/(\sum(y - \bar{y})^2)$, where y is a vector of observation data, $\hat{y}$ is a vector of model-generated predictions, and $\bar{y}$ is the observation data mean (Kvalseth, 1985).

Produced maps were further validated against independent sediment data collected during the April and October 2017 research cruises. Bilinear interpolation was used to predict mud, sand and gravel content at sampled stations and the ~~variance explained~~predictive performance of the map was ~~by the model was~~calculated for the station-averaged observed data using R^2.

2.4.3 Median grain size

A generalised additive model (GAM) of the form: $\log_{10}(D_{50}) \sim s(mud, sand, gravel)$ was used to relate the whole-sediment median grain size to sediment fractions for data from the Firth of Clyde and surrounding shelf seas. This expression outperformed other GAM formulations and Random Forests models. The incorporation of data external to the Firth of Clyde

within this model is supported by the relative robustness among quantitative measures of grain size for sediments derived from contrasting tidal regimes and seabed depths (Aldridge et al., 2015; Bockelmann et al., 2018). The predictive accuracy of the GAM fit was assessed through Leave-One-Out Cross-validation (LOOCV). A map of median grain size for the Firth of Clyde was predicted from the fitted GAM using the previously generated maps of sediment mud, sand and gravel content.

2.4.4 Permeability and porosity

Measurements of sediment permeability and porosity for the Firth of Clyde were supplemented by literature data from the North Sea. This followed an initial assessment to ensure that grain size relationships with permeability and porosity for these two datasets overlapped satisfactorily; however, it is assumed North Sea measurements are a suitable approximation for unsampled higher-energy and coarser-grained Firth of Clyde sediments. The relationships between sediment permeability and median grain size for the top 10 cm of sediment was expressed via a log-log form of the equation described by Wilson et al. (2018), modified to incorporate an intercept:

$$\log_{10}(permeability) = \log_{10}\left(10^a D_{50}^b + c\right) \tag{7}$$

Relationships were also defined between sediment permeability and mud content in the top 10 cm of seabed sediment using a log-log function. Sediment porosity measurements were supplemented by data from literature and fitted to the logistic function given by Wilson et al. (2018):

$$\log_{10} \phi = p_1 + p_2 \left(\frac{1}{1+e^{\frac{-(\log_{10} D_{50} - p_3)}{p_4}}}\right) \tag{8}$$

Equations were fitted to data using the Nelder-Mead algorithm to minimise the root sum of square error (RSS). Relationships were also defined between sediment permeability and porosity and mud content in the top 10 cm of seabed sediment using a log-log function. These linear models were fitted using Maximum Likelihood and a Gaussian error distribution. Maps of sediment permeability and porosity for the Firth of Clyde were then generated by applying the fitted equations to the appropriate sedimentary grain size maps of median grain size.

2.4.5 Natural seabed abrasion

Following Wilson et al. (2018), sediment mobilisation by near-bed water movement was predicted by the Shields relationship. Rates of natural seabed abrasion were then expressed in terms of the annual percentage of time that sediment motion occurs (hereafter referred to as sediment mobility). The hourly near-bed shear stress (τ_b) and occurrence of sediment movement were calculated for each point on the unstructured grid of the Clyde using the *bedshear* package (Wilson and Heath, 2019). This hourly binary classification for sediment movement was then summarised to an annual proportion for each point and mapped.

2.4.6   Seabed organic matter content

The particulate organic carbon and nitrogen content of seabed sediments in the Firth of Clyde were modelled using Random Forests. To assess the potential influence of organic matter loading in river freshwater input on the organic carbon and nitrogen content of seabed sediments, estimated annual minimum, mean and maximum surface salinity was extracted from the Firth of Clyde hydrodynamic model and used as a potential explanatory variable. Mean annual fishing swept area ratio was also included as a potential explanatory variable to account for the possible winnowing effect of frequent trawling disturbance on sedimentary organic matter (Palanques et al., 2014; Pusceddu et al., 2014; Sala et al., 2021). Although Random Forests are generally robust to the presence of highly correlated explanatory variables, model performance may be impacted in smaller datasets (e.g. Li et al., 2011). Recursive Feature Elimination (RFE) is a popular method to select a subset of relevant uncorrelated variables from a large number of noisy or highly-correlated variables by iteratively fitting a Random Forest model and removing the least important ranked variable (Gregorutti et al., 2017). Given the relatively small number of sediment organic carbon and nitrogen measurements available, a variation of Recursive Feature Elimination (RFE) was used to improve model prediction. This was implemented as follows: ~~Spatial~~ spatial cross-validations were run for the full model and for single removals of each environmental variable. The variable whose removal led to the greatest reduction in mean squared error in model prediction was dropped and the process repeated for the reduced suite of variables. This was repeated until the no further minimisation of model error was achieved through further ~~the~~ removal of ~~further~~ environmental variables. Spatial blocking yielded 22 organic carbon blocks containing an average of 7.6 samples in each block and 22 nitrogen blocks containing an average of 5.3 samples in each block. Two thousand iteratively sampled training and test datasets were generated. The distribution of R^2 across cross-validation iterations was highly asymptotic due to instances where $R^2 < 0$ given noise in the sampled data (Kvalseth, 1985). Hence, the median R^2 across iterations was used to avoid the overestimation of model misfit.

2.4.6.1   Standing stock of organic carbon and nitrogen in surficial sediments

Using the map of porosity for the Firth of Clyde, the dry bulk density of the sediment (ρ_d) was calculated from sediment porosity (ϕ) assuming a sediment density (ρ_s) of 2.65 g cm^{-3} (Blake and Hartge, 1986) via:

$$\rho_d = (1 - \phi)\rho_s \tag{9}$$

The masses of organic carbon and nitrogen were then calculated as the product of the proportion by mass of the organic element in dry sediment, dry bulk density, sediment depth and polygon area.

3   Results

3.1   Rock distribution

The distribution of hard substrate in the Firth of Clyde is shown in Fig. 6. Random Forest model predictions suggest that the distribution of hard substrate within the inner Firth of Clyde and sea lochs is limited and largely confined to coastlines. Spatial cross-validation of model predictive performance showed an overall accuracy of 86.5 %, and a sensitivity and specificity of

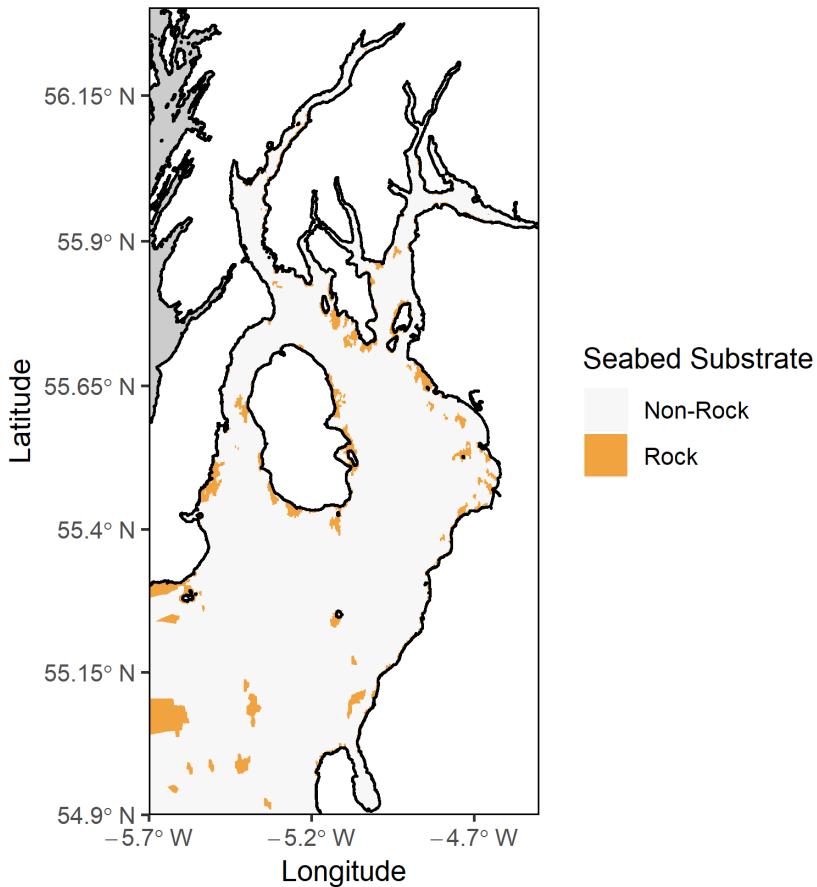



Figure 6 Distribution of hard substrate in the Firth of Clyde according to BGS data layers and Random Forest model predictions generated for areas lacking data - the inner Firth of Clyde and sea lochs.

410   **3.2   Sediment grain size fractions**

Maps of predicted percentages of mud, sand and gravel in the Firth of Clyde are shown in Fig. 7. Extensive areas in the outer Firth of Clyde, including the Arran Deep and Kilbrannan Sound, consist of sediments dominated by the mud fraction, with close to 100 % of particles having a diameter $< 63 \mu m$. In the inner Firth of Clyde and sea lochs, the mud fraction dominates in the sediments of upper Loch Fyne, Holy Loch, and Loch Long. In contrast, sand dominates in coastal areas, especially along
415   the Ayrshire coast, and on the Great Plateau, whereas the percentage of gravel in Clyde Sea sediments is almost universally low.

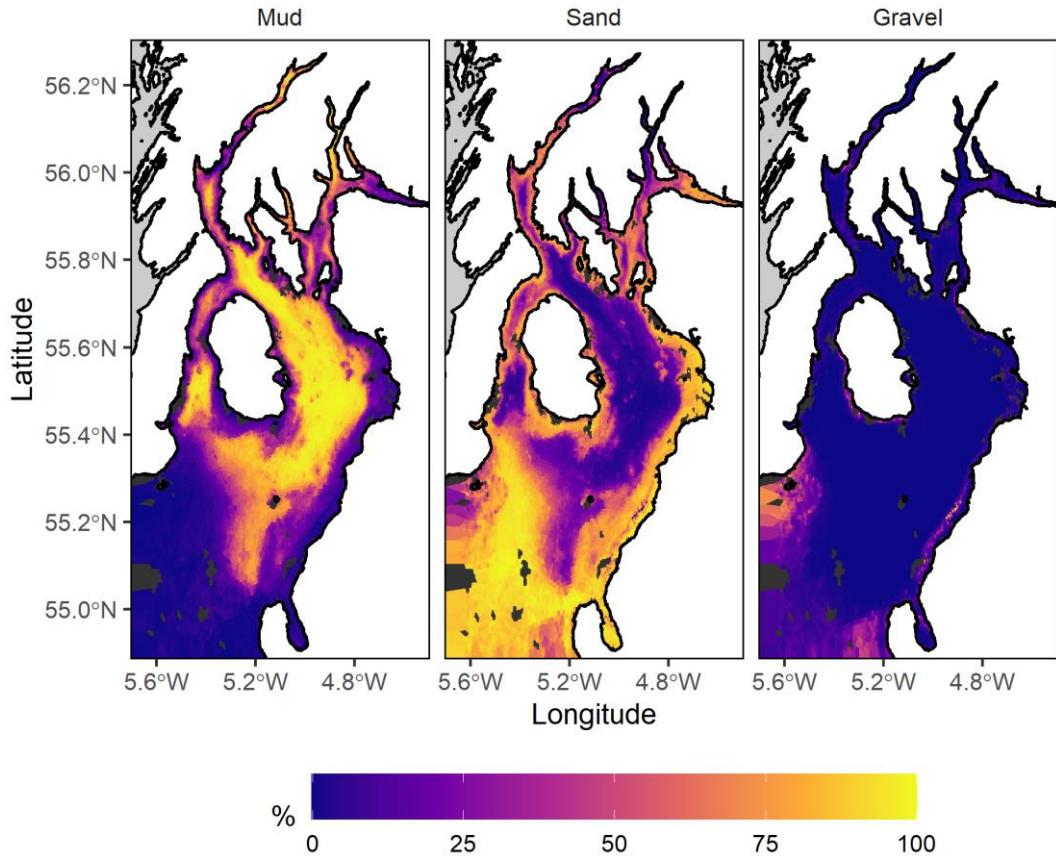



Figure 7 Mapped percentages of mud, sand and gravel within Firth of Clyde sediments predicted from Random Forest models.

Random Forest models capture the variation in mud and sand content well. Spatial cross-validation results yield R^2 values of 0.611, 0.382 and 0.321 for mud, sand and gravel, respectively. Results from the map validation against independent data are shown in Fig. 8 and demonstrate good predictability of mud and sand percentages. R^2 for mud and sand content from map validation against independent data were higher (R^2 mud = 0.635, R^2 sand = 0.611). However, prediction of gravel content is considerably poorer (R^2 gravel = 0.08).

420

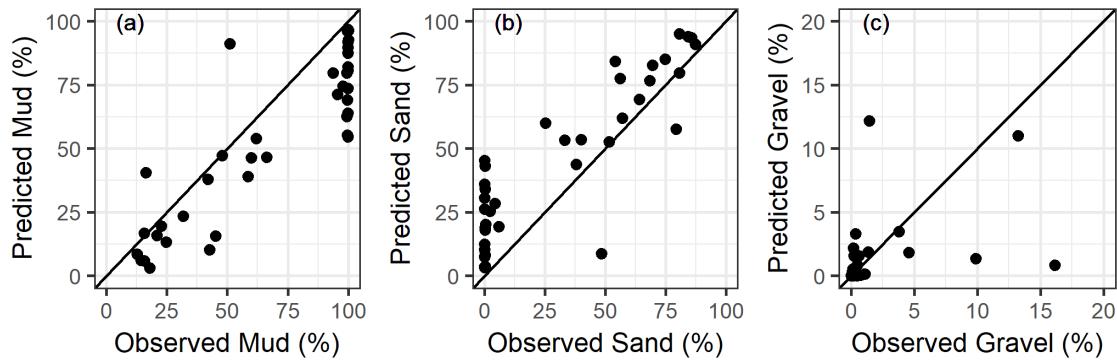



Figure 8 Comparison of predicted and observed percentages of (a) mud, (b) sand and (c) gravel in Firth of Clyde seabed sediments. Observation data are recent measurements carried out in 2017 that were not used during model fitting. Predictions were generated from the sediment maps presented in Fig. 7. This is therefore an independent test of model and map predictive performance.

3.3 Median grain size

Leave-one-out cross-validation (LOOCV) of the GAM shows overall good agreement between predicted and observed measurements of sediment median grain size (Fig. 9). However, there is a mismatch between observations and predictions for sediments where the median grain size < 0.02 mm. The model accounted for 90.76% of the variance in the data.

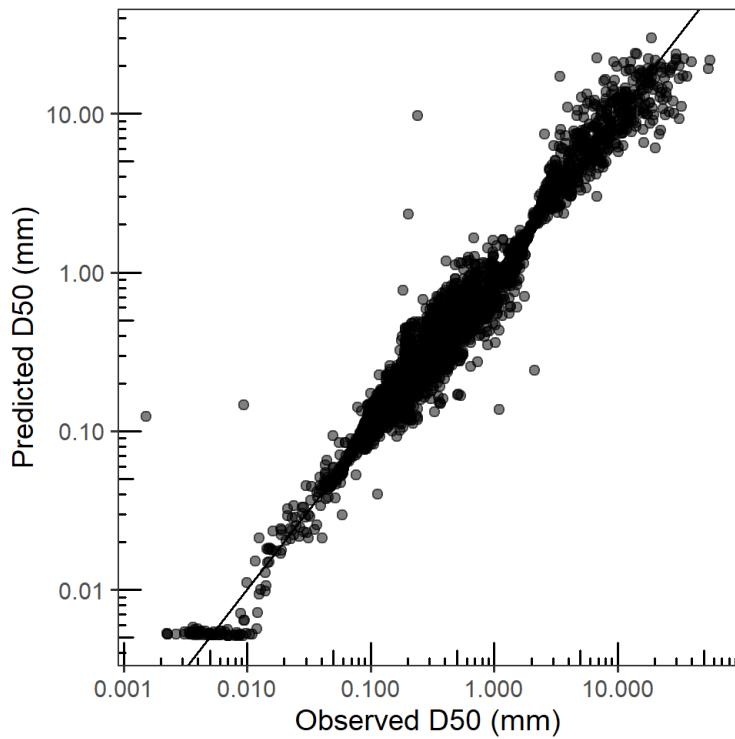



Figure 9 Leave-one-out cross-validation results showing predicted and measured values of median grain size for a dataset comprising samples collected from cruises in the Firth of Clyde in April and October 2017 and supplemented with data from the North Sea, Celtic Sea and English channel collected by CEFAS and MSS (n = 3244). Note $\log_{10}$ scale.

435

Following spatial variation in the fractions of mud and sand, the median grain size of seabed sediments in the Firth of Clyde is lowest in smallest in the Arran Deep and Inchmarnock Water and largest in the shallow waters along the Ayrshire coast and the Great Plateau (Fig. 10).

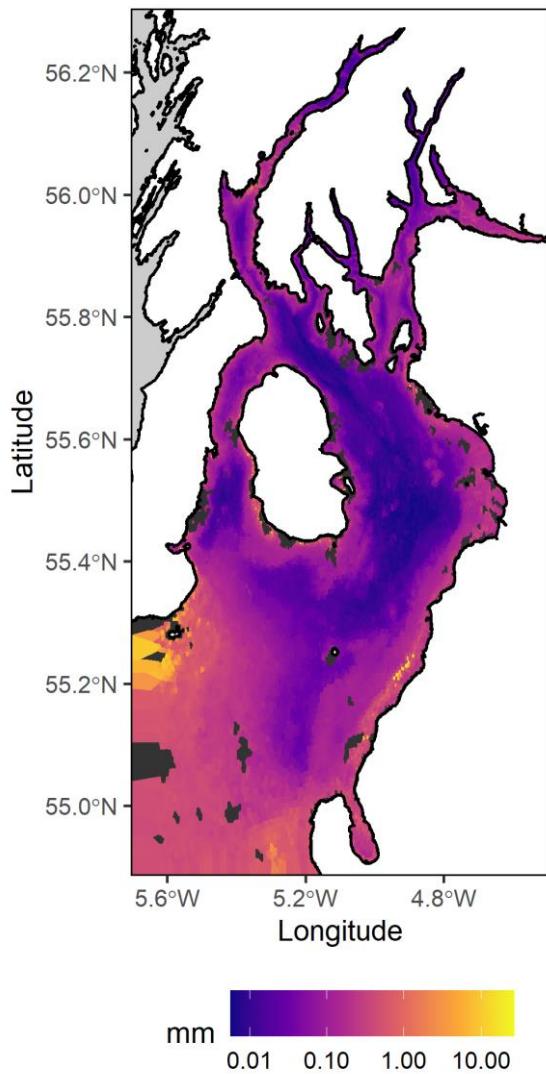



Figure 10 Mapped predictions of median grainsize for seabed sediments in the Firth of Clyde.

3.4   Permeability and porosity

The combined permeability dataset suitably covered a range of sediment types ranging from very fine-grained sediment to sandy sediments. Sediment permeability in the Firth of Clyde is well predicted by both the median grain size and % mud content (Fig. 11). The fitted relationships with mud and median grain size capture 82.3% and 91.5% of the variance in the data, respectively.

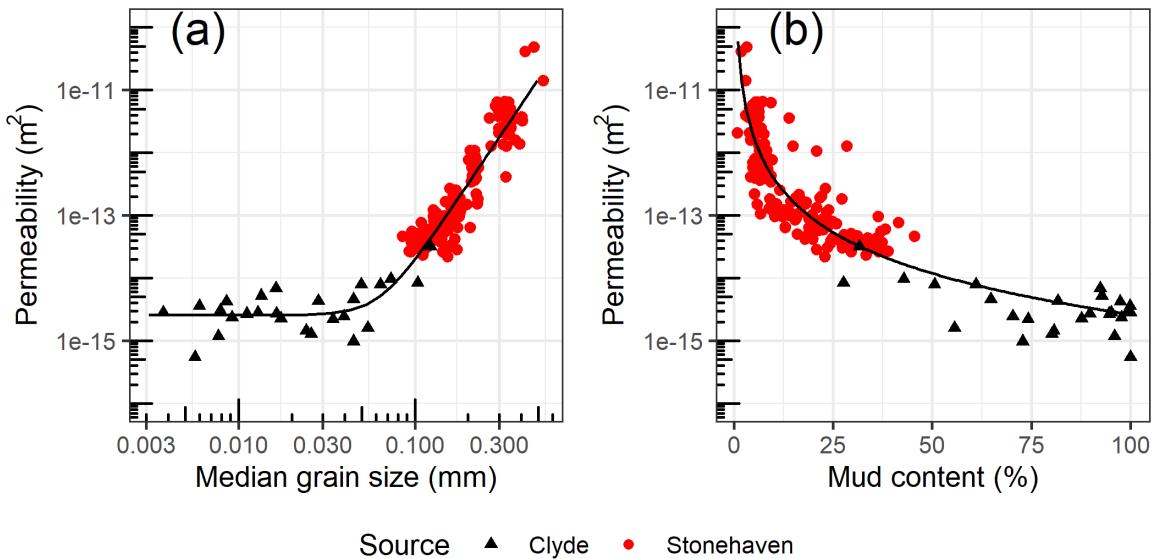



Figure 11 Relationship between sediment permeability (m²) and (a) median grain size (mm) and (b) % mud content for measurements from the Firth of Clyde (black triangles) and samples from the North Sea analysed by Serpetti et al. (2016) (red points). Fitted relationships for the combined data are shown by a solid line. Note log₁₀ scale.

The fitted sediment permeability relationships with median grain size and mud content are given by:

$$\log_{10}(permeability) = \log_{10}(10^{-9.5659}D_{50}^{4.1877} + 2.5634 \times 10^{-15}) \tag{10}$$

$$\log_{10}(permeability) = -2.171 \log_{10}(mud) - 10.232 \tag{11}$$

Porosity is predicted by median grain size and % mud content (Fig. 12) with fitted relationships capturing 77.6% and 72.2% of the variance in the data, respectively. The parameters for the fitted logistic function for the relationship with median grain size in the top 10 cm of seabed sediment are shown in Table 3. The fitted porosity relationship with % mud content is given by:

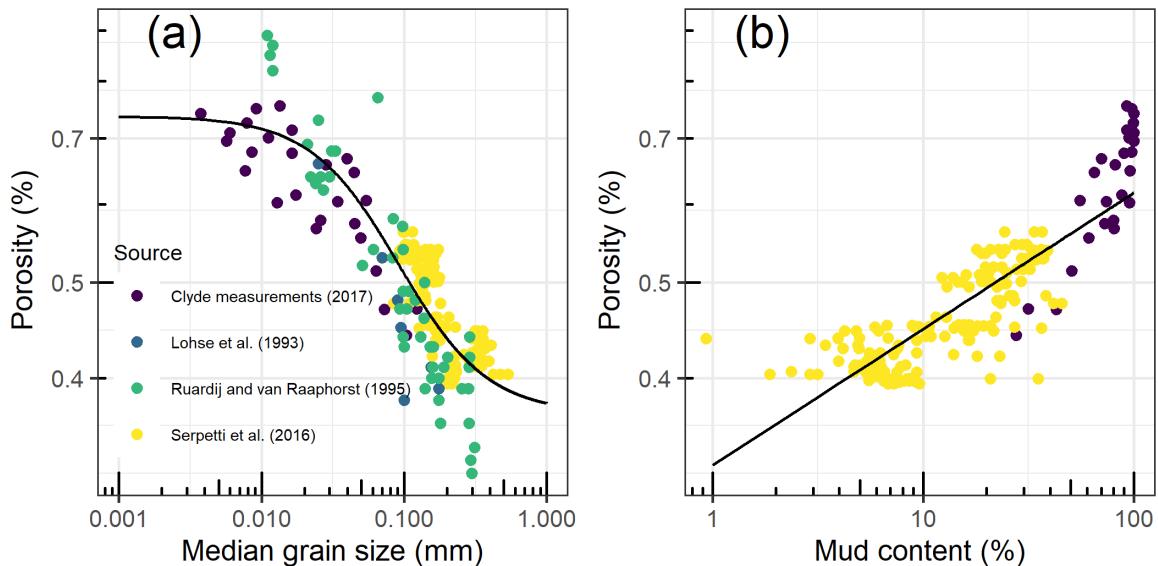
$$\log_{10}\phi = 0.138 \log_{10}(mud) - 0.486 \tag{12}$$

460 **Figure 12 Relationship between sediment porosity (%) and (a) median grain size (mm) and (b) % mud content for measurements for the top 10 cm seabed sediment on a log$_{10}$ scale.**

Table 3 Fitted parameter values for the logistic relationship between sediment porosity and median grain size in the top 10 cm of seabed sediment.

Parameter	Value
p_1	-0.435
p_2	0.302
p_3	-1.035
p_4	-0.314

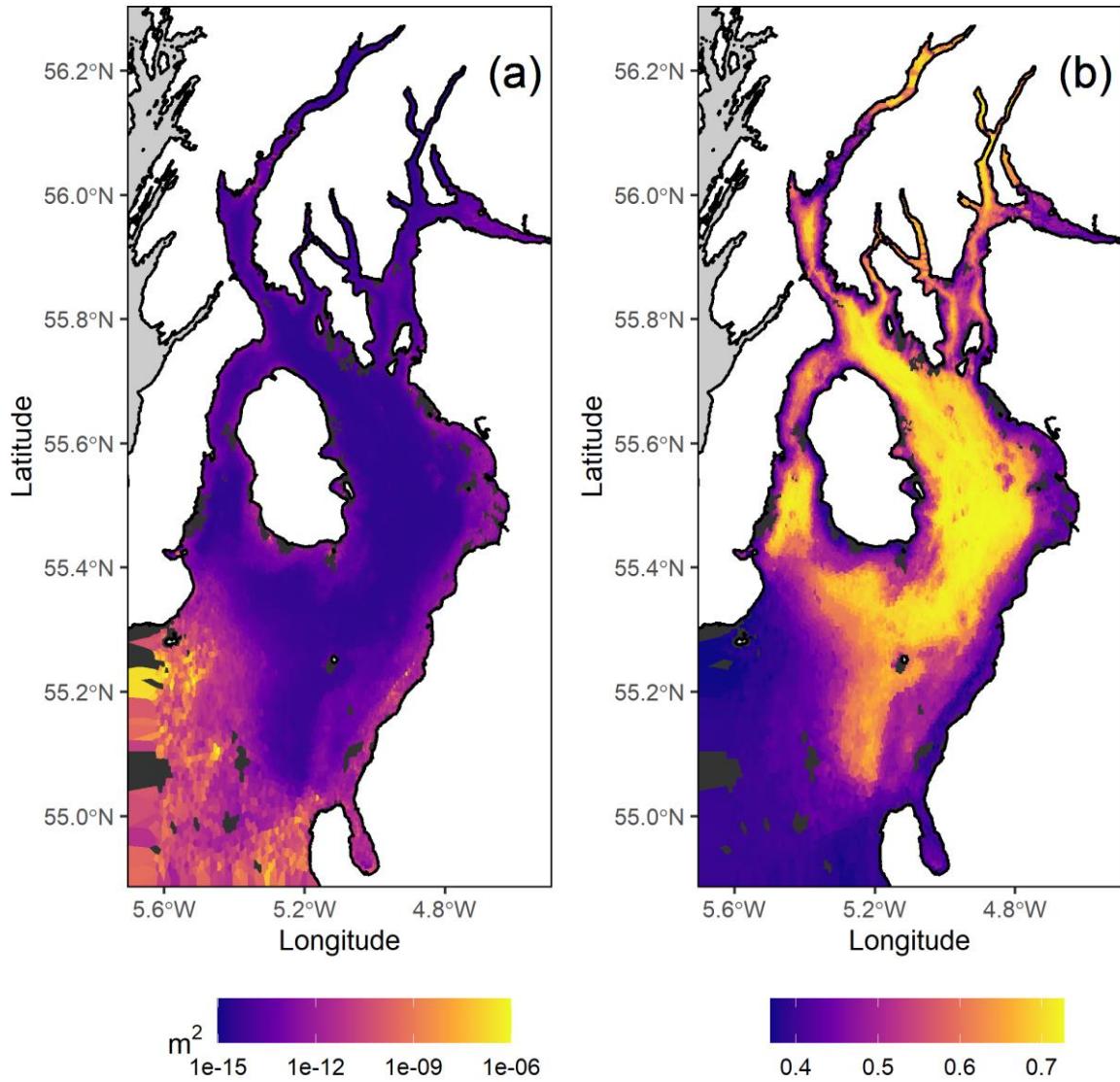
465 The relationship between sediment permeability and mud content, rather than median grain size, was chosen to map the distribution of seabed sediment permeability in the Firth of Clyde (Fig. 13a) because the prediction dataset is one step closer to measured data. The vast majority of the Firth of Clyde is dominated by sediments having a permeability $< 10^{-12}$ m^2. This implies that sediments are impervious and solute exchange between the seabed and water column is diffusion-mediated (Huettel et al., 2003).

470

Figure 13 Mapped predictions of (a) sediment permeability and (b) sediment porosity in the Firth of Clyde. Seabed consisting of hard substrate or sediments with a predicted permeability greater than 1×10^{-6} m^2 are shown as black areas. Permeability predictions were generated from the map of percentage mud content in seabed sediments. Porosity was predicted from the map of median grain size in the Firth of Clyde.

475   **3.5   Natural seabed abrasion**

Extensive areas of the Clyde Sea, particularly Inchmarnock Water, the Inner Firth of Clyde and Loch Fyne appear to experience negligible annual sediment movement (Fig. 14). Sediment moves for up to 25% of the year in some areas of the Outer Firth.

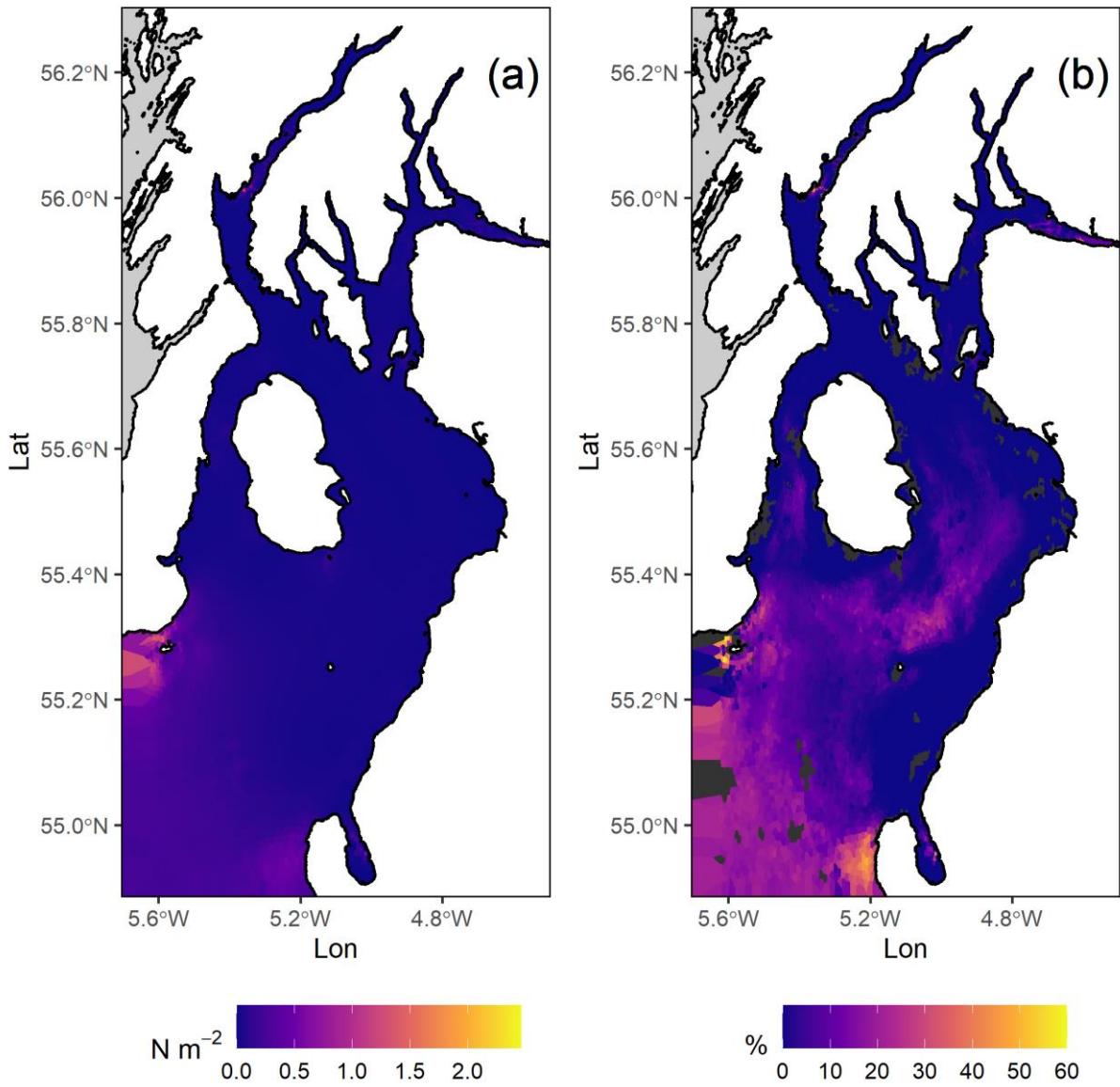
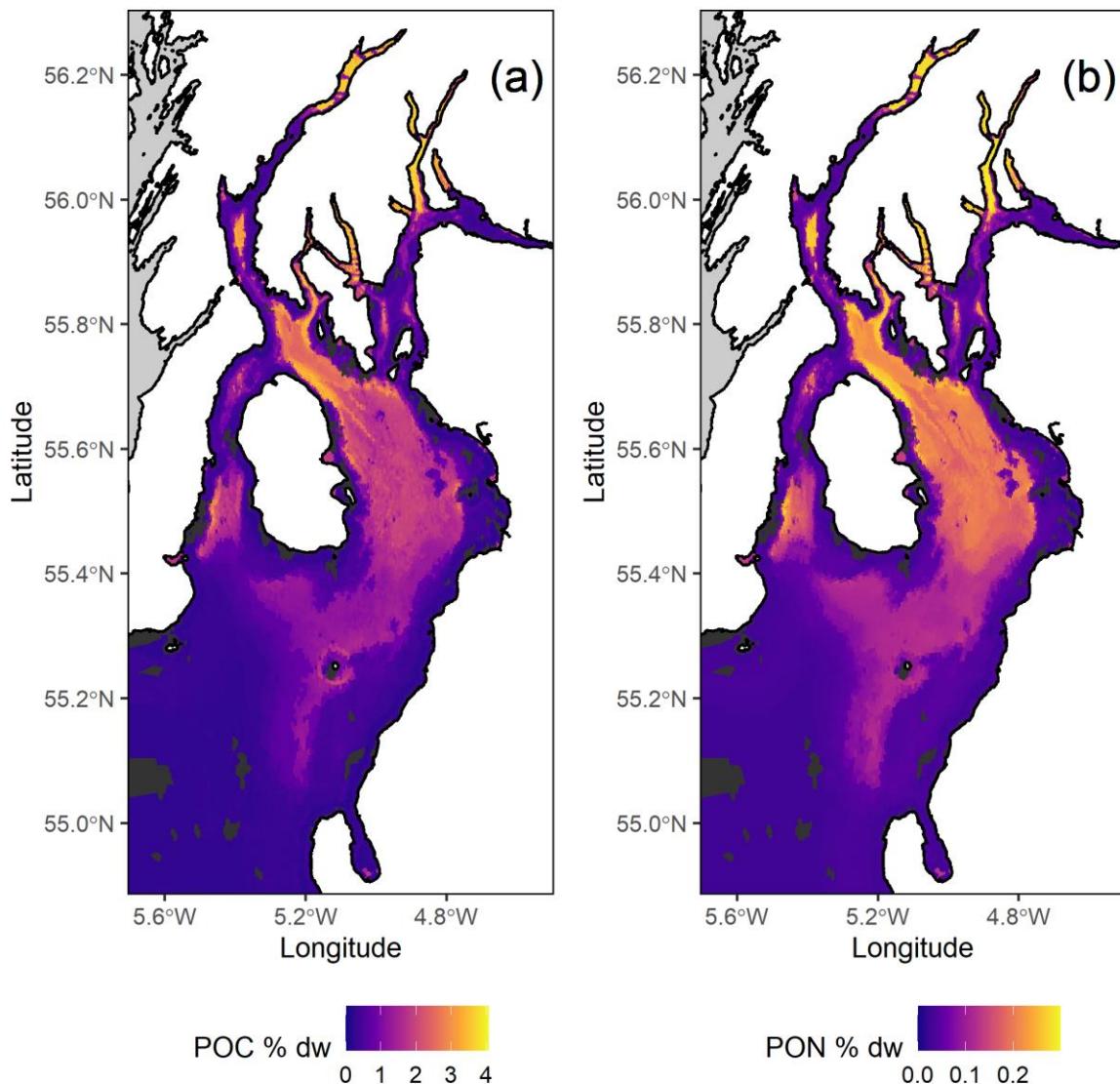



Figure 14 Mapped predictions of (a) annual mean bed shear stress and (b) sediment mobility in terms of the annual percentage of time that sediment motion occurs in the Firth of Clyde.

3.6 Sediment organic matter content

Cross-validated RFE substantially improved model predictions for both organic carbon and nitrogen. The initial models of organic carbon and nitrogen each contained 16 predictors and had a cross-validation R^2 of 0.~~557~~ 387 and 0.~~685~~561, respectively. Optimal models for particulate organic carbon and nitrogen each comprise four environmental variables, shown in Table 2, and have an R^2 of 0.~~705~~ 566 and 0.~~827~~789, respectively.

Maps of predicted % organic carbon and nitrogen per gram dry weight sediment are shown in Fig. 15. Although salinity was selected out of the optimal model, there remains a pattern of increased organic carbon and nitrogen enrichment in fine-grained inner firth sediments in proximity to sources of freshwater run-off.

 Figure 15 Maps of predicted % dry weight sediment (a) organic carbon and (b) organic nitrogen content within the top 10 cm of seabed in the Firth of Clyde.

Deviations from the Redfield ratio (molar ratio C/N = 6.6) are seen in areas close to sources of terrestrial organic matter input (Inner Firth, Loch Fyne, Ayrshire coast) (Fig. 16). By calculating the bulk density of the sediment, the total mass of organic carbon and nitrogen in the top 10 cm of Firth of Clyde sediments estimated as 3.42~~24~~ and 0.3~~3~~277 million tonnes, respectively.

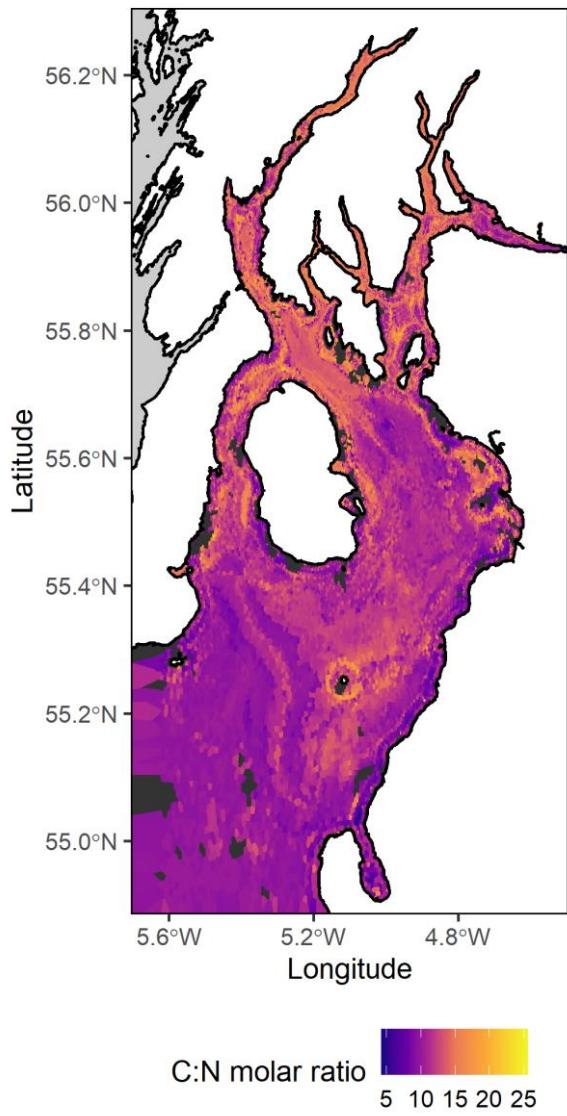



Figure 16 Variation in the predicted organic carbon-nitrogen molar ratio.

4   Discussion

The maps of surficial sediment grain size fractions, whole-sediment median grain size, permeability, porosity, whole-sediment organic carbon and nitrogen content, and seabed natural disturbance presented in this study fill a current data gap for the Clyde. The variable spatial resolution allowed by the irregular grid arrangement suitably captures the fine-scale transition from coarser-grained sediment close to the coast to fine-grained sediments in deeper offshore waters.

Confidence in the predicted distributions of seabed mud, sand, organic carbon and nitrogen content is high, with validation procedures showing that the maps accounted for ~ 60% of the variation in the independent data. However, the fraction of gravel in samples was low or negligible in most samples, and this is reflected in the poor confidence implied by the validation results. Conversely, mud content is as a key predictor of several biogeochemical and geotechnical properties, including sediment permeability and organic carbon content, and given the abundance of muds within the Firth of Clyde, accurate predictions of sediment mud content and the discrimination of mud-rich and sandy areas are important benchmarks for these data products.

Comparison of mud content predictions generated from sediment grain size fraction maps with measurements from recently collected seabed sediment samples shows credible discrimination of mud-rich sediments from sands and muddy sands. However, maps consistently underpredict the mud content of seabed clay samples. This may be a limitation of the sample processing methodology adopted for the collection of independent validation data. These data were derived from subsamples collected from each grab (section 2.3), whereas the BGS data that largely drives mud content predictions were derived from the stacked-sieve processing of the whole grab sample (Deegan et al., 1973). Nevertheless, this approach is a robust test of the performance of these maps as predictions are also impacted by mapping resolution and local-scale sediment heterogeneity. Sedimentology data from future survey efforts will help refine estimates of the predictive performance of the presented data products as well as contributing to improved iterations of seabed sediment maps for the Firth of Clyde.

The spatial distribution of organic carbon in the surficial sediments of the Clyde and the predicted molar ratios of carbon and nitrogen agree remarkably well, at a coarse scale, with patterns reported by Pearson et al. (1986). This may indicate that spatial patterns of sediment organic carbon and nitrogen are a relatively stable feature of the Clyde. The organic enriched muddy sediments in upper Loch Fyne, Loch Striven and the Kyles of Bute, the inner Clyde sea lochs, and the Inchmarnock basin coincide with areas of very low seabed disturbance. This may indicate that enrichment patterns are due to long-term organic matter deposition over eutrophic seabed. Although mud content was the most important predictor of organic carbon and nitrogen, corroborating previous work (Diesing et al., 2017; Wilson et al., 2018), this relationship was strongly dependent on other environmental variables, particularly the proximity to terrestrial areas and local bed shear stress, indicating that organic matter – sediment type relationships cannot be generalised across different regions.

Conversely, measurements of sediment permeability in the Clyde map onto median grain size relationships established for the North Sea and extend these relationships to finer-grain sediments. The updated relationships may be applied to a wide range of sediment types collected from diverse marine coastal regions. A new relationship with the percentage mud content in sediment is presented to allow the prediction of permeability in mud-rich sediments.

Contrasting with patterns found by Wilson et al. (2018) in the North Sea, the overall high mud content of Clyde marine sediments leads to large seabed areas with a predicted sediment permeability well below 10^{-12} m^2. Despite the expectation that shallow coastal sands would have permeabilities exceeding this threshold (Serpetti et al., 2016; Wilson et al., 2008), this was only observed for limited areas of the South Ayrshire coast.

535 The mass of organic carbon in the surficial sediments of the Clyde was calculated to be 3.42 million tonnes within a 3739 km^2 area. This complements recent larger-scale and detailed mapping of carbon stocks in coastal and shelf sediment within the UK Exclusive Economic Zone (EEZ) by Smeaton et al. (2021) by better resolving transitions in the carbon storage characteristics of fjordic sediments and the muds of the outer Firth. Smeaton et al. (2021) characterised organic carbon content for a range of sediment types within broadly delineated fjordic and open shelf areas to address underestimates in the organic carbon stock

540 of inshore and coastal mud within previous works. Their mapped sediment organic carbon content for fjordic sea lochs of the Clyde compare well with the data products presented here. However, our work shows organic carbon enriched sediments in the Inchmarnock basin that are not captured by Smeaton et al. (2021), and provides a more credible transition in the carbon storage characteristics between Clyde sea loch and open firth sediments. suggests that the density of organic carbon in the Clyde within the top 10 cm of marine sediments may have been underestimated in work carried out to date to catalogue stocks

545 of organic carbon in Scottish coastal and shelf sediments (Burrows et al., 2014; Smeaton et al., 2020). For instance, the area covered by the Firth of Clyde is < 1 % of Scotland's mapped extended Exclusive Economic Zone (EEZ; 554,755 km^2), yet surficial sediment organic carbon stocks calculated in the present study are 1.5% of the total organic carbon stock in the EZZ calculated by Smeaton et al. (2020; 221 Mt). Given that the produced maps do not provide any estimate of sediment volume beyond the top 10 cm, direct comparisons with studies that estimate total carbon stocks are difficult (e.g. Smeaton et al., 2016).

550 The frequency of seabed trawling disturbance, estimated via the mean annual swept area ratio, was not a predictor of sediment of organic carbon or nitrogen in the Clyde. This suggests that present patterns of demersal trawling do not contribute to measurable differences in sediment organic carbon and nitrogen within the Firth of Clyde. However, this finding does not account for the potential historical Hence, large-scale redistribution of organic matter (sensu Pusceddu et al., 2014) due to transport of resuspended sediments with the onset of demersal trawling in this shallow coastal marine system reported by

555 Pusceddu et al. (2014) for deep sea continental slopes is not observed in this shallow coastal marine system. NeverthelessMoreover, this does not imply that sedimentary organic matter is unaffected by the periodic resuspension of seabed sediments by the passage of demersal trawl gear. More detailed geochemical studies addressing the deposition, remineralisation and burial rates of organic matter are needed to provide a more complete assessment of the impact of trawling activity on carbon sequestration and storage in the marine sediments of the Firth of Clyde. Nevertheless, the maps presented

560 here will contribute to these efforts. Coupled with mapped distributions of anthropogenic and natural seabed abrasion pressure, these maps are expected to be particularly useful for the assessment of seabed sensitivity to changes in disturbance regime and the maintenance of seafloor integrity.

4.1 Limitations

A key challenge in mapping Clyde sediment was the large variation in the availability of high-quality data on measurements

565 sediment properties and of sediment properties across the range of environmental gradients represented in the region. Therefore, whole-sediment median grain size measurements from surrounding shelf seas were used to parameterise a statistical relationship between the whole-sediment median grain size and fractions of mud, sand and gravel. This approach provided

credible predictions of whole-sediment median grain size for a wide range of sediment types, but the model was insensitive to variation in median grain size < 0.02 mm. This is a methodological limitation; the model depends on information from the mud-sand ratio to track the variation in median-grain size, and mud-only sediments do not provide this information. However, the impact of this model mismatch for derived maps of sediment geotechnical properties is limited as the data suggested a decoupling of relationships with median grain size below 0.02 mm.

Additionally, tThere is limited good data on the fine-scale distribution of rocky outcrops and hard substrata for coastal seabed around the UK, including the Firth of Clyde. The available data were in the form of polygons with limited information to distinguish between rocks, boulders and rock covered by a layer of sediment. Cross-validation results suggest high overall accuracy in the discrimination hard and soft substrata, but the coarse thematic resolution of the source data yields little information on structural complexity and heterogeneity of areas of rocky seabed that may be valuable for ecological modelling applications (Elliott et al., 2017). Although rock and gravel areas comprise a small proportion of the overall Firth of Clyde, mMore work is needed to refine these maps.

Random Forests models performed well, capturing the critical features in the spatial distributions of properties investigated in this paper. However, Random Forests models may perform poorly when extrapolating predictions to new geographic areas (Misiuk et al., 2019), in part because predictions do not extrapolate beyond the range of observation data used in model fitting. Accordingly, maps of data-limited properties, such as sediment organic carbon and nitrogen, may not capture the extreme values in that may occur in the understudied muds of fjordic sea-lochs (Smeaton and Austin, 2019).

The maps assume that no changes in distribution of sediment properties occurred as a result of any changes in disturbance regime over time. This assumption is supported by very low tidal currents over much of the Clyde (Davies et al., 2004), and fact that fishing pressure was not retained as a predictor of organic carbon and nitrogen concentration in Clyde sediments. Additional work is needed to define permeability and grain size relationships in coarser sediments with a median grain size greater than 0.6 mm. Hence, sediment permeability estimates greater than 10^{-11} m^2 may be unreliable. Nevertheless, this constitutes a small fraction of the mapped area.

Data availability

Mapped data products are available in csv and netCDF format from:
https://doi.org/10.15129/2003faa2-ee93-4c11-bb16-48485f5f136d (Heath and Pace, 2021).

Author contributions

MP, MH, HS, DB and DS designed the analysis. MP, DB and HS collected samples and carried out laboratory analysis of sediments. MP and HS carried out compilation of data and modelling. MP drafted the manuscript, and all authors contributed to the writing of the manuscript.

Competing interests

The authors declare that no competing interests are present.

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610 HR09011) and contributing institutions.

Appendix A: Scaling mud-fraction organic matter content to the whole-sediment

If the organic matter content associated with the gravel fraction (>2 mm) is negligible, the % particulate organic carbon content (POC) of the whole sediment is given by:

615 $$POC = \frac{m_{mud} + m_{sand}}{p_{mud} + p_{sand}} \cdot 100 \qquad \text{(A1)}$$

Where m_{mud} and m_{sand} are the masses of organic carbon respectively associated with the proportions of mud (p_{mud}) and sand (p_{sand}). If the POC and mass fraction of carbon associated with mud (gC g^{-1}) are known, then the mass of carbon associated with sand may be determined via:

$$m_{sand} = \frac{POC}{100} \cdot mud \left(1 + \frac{sand}{mud}\right) - w_{mud} \cdot mud \qquad \text{(A2)}$$

620 The mass fraction of carbon associated with sand (gC g^{-1} sand) was calculated for CEFAS data collected from the North Sea (n = 148) and 5 grab samples collected from the Firth of Clyde in April and October 2017.

Figure A1 shows that there is no relationship between the mass fractions of carbon associated with sand and mud, suggesting that a single value for the mass fraction of carbon in sand may be applied to any sediment. Taking the mean carbon in sand mass fraction for the available data yields 0.000301 gC g^{-1} sand.

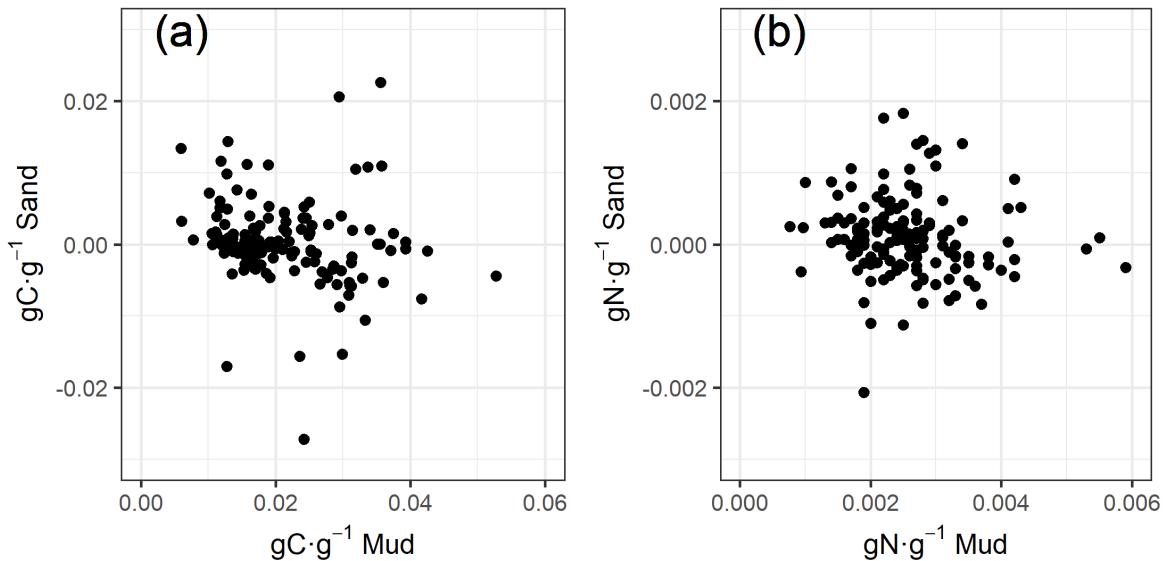
625 These formulae were also used to calculate the mass fraction of nitrogen associated with sand (gN g^{-1} sand) for CEFAS (n = 138) and Clyde Sea (n = 5) data. Similarly, no relationship between the mass fractions of nitrogen associated with sand and mud were observed (Figure A1). The mass fraction of nitrogen in sand may potentially vary as a function of the sand content. However, noise in the data prevents parametrisation of this relationship, and hence a mean nitrogen in sand mass fraction is used (0.000117 gN g^{-1} sand).

630

Figure A.1 Relationship between (a) the mass fraction of organic carbon in mud (gC g⁻¹ mud) and in sand (gC g⁻¹ sand) and (b) the mass fraction of organic nitrogen in mud (gN g⁻¹ mud) and in sand (gN g⁻¹ sand) for sediment collected from the North Sea by CEFAS and the Clyde Sea.

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