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Supplementary information

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Long-term carbon storage in shelf sea sediments reduced by intensive bottom trawling

In the format provided by the authors and unedited

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Supplementary Figure 1. Spatial distribution of mud, OC content and bottom trawling intensity in the North Sea. a, Mean mud content (%) in the upper 10 cm sediment, with marked large-scale mud depocenters. **b**, Mean Total Organic Carbon (TOC) content (%) in surface 10 cm sediment interpolated from existing field data^{48,49}. **c**, Synthesized annual aggregated bottom trawling intensity in terms of swept area ratio (SAR, yr⁻¹) averaged over

the period of 2015-2020 in surface sediments (0-2 cm). **d**, similar to (c) but for sub-surface sediments (2-10 cm). The contour line of SAR=1 yr⁻¹ is indicated in (c) and (d). All color plots are in logarithmic scale.



Supplementary Figure 2. Correlation between mud content and OC content, between multiyear (2015-2020) averaged SAR and mud content and between SAR and OC content referred from the North Sea dataset (n = 2380).



Supplementary Figure 3. Number of field data points in each aggregated interval.



Supplementary Figure 4. Distribution of field sampling stations in the North Sea. Data from two densely sampled local areas (indicated by the rectangles) representing sandy (German Bight, n = 167) and muddy seabed (Fladen Ground, n = 450) are shown in Figure 2 (main text).



Supplementary Figure 5. Simulated change of the major source and sink terms of sedimentary OC as well as the resultant net stock change in the North Sea. The upper and lower edge of the grey zone for the trawling result refer to the mean ± standard deviation

of the model results from the 18 trawling scenarios, respectively. The mean values are indicated by the red curve.



Supplementary Figure 6. Comparison of simulated macrobenthos biomass and OC stock in surface sediments between scenarios with trawling and without trawling (No-Trawling) in the North Sea. The reference of OC stock in (b) is the model result after the first-loop of simulation. Annual trawling effort is added in (a). Time series of simulated net annual primary production is indicated by the grey line with y-axis on the right in (b). The upper and lower edge of the grey zone in (a) and (b) refer to the mean \pm standard deviation of the model results from the 18 trawling scenarios, respectively. The mean values are indicated by the blue curve. Model projection of recovery of macrobenthic biomass and OC stock in

the highest trawling impact scenario (Trawling_M3_10pct_phymix1) with trawling stopped is also shown in both plots.



Supplementary Figure 7. Distribution of modeled OC/mud in relation with the multiyear averaged SAR in the entire North Sea after 50 years' trawling. a, Distribution in weakly or untrawled areas (SAR < 1 yr⁻¹, n = 59900). b, Similar to (a) but for intensely trawled areas (SAR ≥ 1 yr⁻¹, n = 25030). The means \pm standard deviations are indicated by dots and error bars.



Supplementary Figure 8. Comparison of simulated relative change of mud and OC content in surface sediments of intensively trawled areas (SAR > 1 yr⁻¹) in the North Sea after 50 yrs' trawling. The diagonal line indicates unchanged OC/mud.



Supplementary Figure 9. Vertical distribution of simulated overall remineralization rate (k) of organic carbon in North Sea sediments. The black curves refer to the mean value and the space between the 10th and 90th percentiles is marked by grey colour.



Supplementary Figure 10. Simulated distribution of labile OC content (%) in the North Sea surface sediments (averaged over the upper 10 cm) in summer (a) and winter (b) of the No_trawling scenario.



Supplementary Figure 11. Comparison of multi-year (2015-2020) averaged bottom trawling effort in the North Sea between the daily-resolved AIS dataset²³ and annually aggregated ICES dataset⁵³.



Supplementary Figure 12. Monthly distribution of bottom trawling effort in the North Sea for the period 2016-2020 derived from the daily-resolved AIS dataset²³.



Supplementary Figure 13. Spatial distribution of annual aggregated bottom trawling effort in the North Sea for the period 2015-2020 derived from the daily-resolved AIS dataset²³.



Supplementary Figure 14. Annual aggregated bottom trawling effort by two dominant métiers (beam and otter trawlers) for the year 2015 and 2020 derived from the daily-resolved AIS dataset²³.



Supplementary Figure 15. Reconstructed time series of annual trawling effort for 1950-1984 using three different scaling methods. a, Linear regression analysis between landings and total trawling effort in the North Sea for the period 1985-2020. **b**, Reconstructed total trawling effort (1950-1984) based on the linear regression relationship in (a). **c**, Linear regression analysis between UK trawling effort and total trawling effort in the North Sea for the period 2003-2015. **d**, Reconstructed total trawling effort (1950-1984) based on the linear regression relationship in (c). **e**, Technological creep multiplier in scaling the trawling effort with landings for the period 1950-1984. **f**, Reconstructed total trawling effort (1950-1984) based on the linear regression relationship in (a) with additional increase of technological efficiency. The monitored total trawling effort for 1985-2020 is also shown in (b), (d) and (f).



Supplementary Figure 16. Comparison between field-derived and simulated values. a. Macrobenthic biomass at sampling stations (n = 56) with sources listed in Supplementary Table 2. The RMSE is 2.23 g m⁻², equivalent to 40% of the spatially averaged value in the entire North Sea. **b**, Mean biomass over each degree in latitude. **c**, Bioturbation rate. Field data sources are distinguished in (c). Red triangles indicate data from the southern North Sea estimated mainly from Chlorophyll profiles (data from this study), green circles are from Solan et al.⁷⁰ and blue diamonds (data from this study) are from the northern North Sea estimated mainly from 210Pb profiles. The Normalized RMSE for bioturbation rate is 0.176.

Supplementary Table 1. Numerical experiments representing scenarios of varying impacts of bottom trawling in the study.

Name of scenario	Trawling	Trawling	Mortality in	Physical
	included	reconstruction	macrobenthos	mixing
		method for	scaled by	coefficient
		1950-1984	SAR	scaled by
				SAR
				$(cm^2 d^{-1})$
No_trawling	No			
Trawling_M1_10pct_phymix1	Yes	Method_1	11%	0.24
Trawling_M1_10pct_phymix10	Yes	Method_1	11%	2.4
Trawling_M1_50pct_phymix1	Yes	Method_1	20%	0.24
Trawling_M1_50pct_phymix10	Yes	Method_1	20%	2.4
Trawling_M1_90pct_phymix1	Yes	Method_1	30%	0.24
Trawling_M1_90pct_phymix10	Yes	Method_1	30%	2.4
Trawling_M2_10pct_phymix1	Yes	Method_2	11%	0.24
Trawling_M2_10pct_phymix10	Yes	Method_2	11%	2.4
Trawling_M2_50pct_phymix1	Yes	Method_2	20%	0.24
Trawling_M2_50pct_phymix10	Yes	Method_2	20%	2.4
Trawling_M2_90pct_phymix1	Yes	Method_2	30%	0.24
Trawling_M2_90pct_phymix10	Yes	Method_2	30%	2.4
Trawling_M3_10pct_phymix1	Yes	Method_3	11%	0.24
Trawling_M3_10pct_phymix10	Yes	Method_3	11%	2.4
Trawling_M3_50pct_phymix1	Yes	Method_3	20%	0.24
Trawling_M3_50pct_phymix10	Yes	Method_3	20%	2.4
Trawling_M3_90pct_phymix1	Yes	Method_3	30%	0.24
Trawling_M3_90pct_phymix10	Yes	Method_3	30%	2.4

Nr stations	Sampling time	Data type	Source
	(mm/yy)		
6	03/99, 08/00	TOC, mud & Benthos	(1)
15	08/92, 05/93	TOC, mud & Benthos	(2)
35	06-10/21	TOC, mud & Benthos	(3)
2380	1990-2022	TOC & mud	(4)

Supplementary Table 2. Summary of field data in the North Sea

Note. Sources: (1) Ståhl et al.⁷²; (2) Rosenberg et al.⁷³; (3) Zhang et al.²⁹; (4) This study.

Supplementary Table 3. Maximum penetration depths of different gear components used in the model. Layers above the maximum impacted depth are considered impacted. Ground gear is separated into a surface and a subsurface component based on the métier using the ratios given in Eigaard et al.⁵⁴

Gear component	Maximum penetration depth (cm)			
	Sand (<10% mud content)	Mud (≥10% mud content)		
Otter trawl doors	5	10		
Beam trawl shoes	10	10		
Sweeps, chains and bridles	2	5		
Tickler chains	5	10		
Ground gear subsurface	5	10		
Ground gear surface	2	2		

Supplementary Table 4. Data sources used in this study.

Data	Reference	URL
description		
Gridded data	Bockelmann, F.	https://doi.org/10.1594/WDCC/coastMap_Substrate_M
of surface	D. Mud content	ud
mud content	of Noarth Sea	
in the North	surface	
Sea	sediments. Worl	
sediments	d Data Center for	

	Climate	
	(WDCC) at	
	DKRZ (2017).	
Gridded data	Bockelmann,	https://doi.org/10.1594/WDCC/coastMap_Substrate_T
of surface	F.D. Total	OC
organic	organic carbon	
carbon	content of North	
content in	Sea surface	
the North	sediments. Worl	
Sea	d Data Center for	
sediments	Climate	
	(WDCC) at	
	DKRZ (2017).	
Global surface mud and OC content	Jenkins, C. Building offshore soils databases. <i>Sea</i> <i>Technol.</i> 38 , 25– 28 (1997).	https://instaar.colorado.edu/~jenkinsc/dbseabed/
Global surface sediment OC content	Atwood, T. B., Witt, A., Mayorga, J., Hammill, E. & Sala, E. Global Patterns in Marine Sediment Carbon Stocks. <i>Front. Mar. Sci.</i> 7 , 165 (2020).	https://figshare.com/articles/marine_soil_carbon/99418 16
Global annual surface sediment SAR	Kroodsma, D.A. et al. Tracking the global footprint of fisheries. <i>Science</i> 359 , 904–908 (2018).	https://globalfishingwatch.org/datasets-and-code/

Annual aggregated bottom trawling effort from	ICES. OSPAR request 2018 for spatial data layers of fishing intensity/pressur e. Data Outputs (2019).	https://doi.org/10.17895/ices.data.4686
the ICES		
data		
Field and	Zhang, W. Field	https://doi.org/10.5281/zenodo.8297751
model data	and Model Data	
of OC and	for Bottom	
macrobentho	Trawling	
S	Impacts in the	
	North Sea	
	(Version 1)	
	(2023).	
Daily	Zhang, W. Field	https://doi.org/10.5281/zenodo.8297751
aggregated	and Model Data	
spatial data	for Bottom	
on bottom	Trawling	
trawling in	Impacts in the	
the North	North Sea	
Sea for the	(Version 1)	
period 2015-	(2023).	
2020		
Multi-year	Zhang, W. Field	https://doi.org/10.5281/zenodo.8297751
averaged	and Model Data	
surface	for Bottom	
sediment	Trawling	
swept area	Impacts in the	
ratio (SAR	North Sea	
2015-2020)	(Version 1)	
	(2023).	

Model code	Zhang, W. Field	https://doi.org/10.5281/zenodo.8297751
and	and Model Data	
simulation	for Bottom	
results	Trawling	
	Impacts in the	
	North Sea	
	(Version 1)	
	(2023).	
Annual	Zhang, W. Field	https://doi.org/10.5281/zenodo.8297751
aggregated	and Model Data	
trawling-	for Bottom	
induced	Trawling	
resuspension	Impacts in the	
rate in the	North Sea	
North Sea	(Version 1)	
2015-2020	(2023).	
Point data	Zhang, W. Point	https://doi.org/10.5281/zenodo.13322571
and R code	data and R code	
	for multivariate	
	analysis of	
	Bottom Trawling	
	Impacts in the	
	North Sea	
	(Version 1)	
	(2023).	

Supplementary Text 1. Regression analysis using Instrumental Variables

The instrumental variable (IV) approach isolates the exogenous variation of the explanatory variable to estimate its causal effect on the dependent variable⁷⁵. In our case, water depth is used as an instrument for SAR. Specifically, a binary depth dummy variable with a cutoff at 150 m is created, based on the observation that fishing activities significantly decrease beyond

this depth (see Supplementary Supplementary Figure 16). Therefore, *depth* is converted to a binary variable which is equal to 0 if *depth* is smaller than 150 m and 1 otherwise:

$$depth_dummy_i = \begin{cases} 1 \text{ if } depth_i \ge 150 \text{ m} \\ 0 \text{ if } depth_i < 150 \text{ m} \end{cases}$$



Supplementary Figure 17. SAR versus depth (m). The dashed line at 150 m indicates the cut-off point for the instrument which is used in the regression analysis.

The IV approach proceeds in two stages: In the first stage, SAR is regressed on *depth_dummy* and the other covariates (*mud*, *tau*, *dist*, *phy*, *temp*) to obtain the estimated predicted values $S\widehat{AR}_i$:

 $S\widehat{AR}_{i} = \alpha_{0} + \alpha_{1} \cdot depth_{dummy_{i}} + \alpha_{2} \cdot mud_{i} + \alpha_{3} \cdot tau_{i} + \alpha_{4} \cdot dist_{i} + \alpha_{5} \cdot phy_{i} + \alpha_{6} \cdot temp_{i} + \varepsilon_{i},$

with coefficients α_0 to α_6 and error term ε_i .

In the second stage, OC is regressed on the obtained values SAR_i and the same covariates to estimate the treatment effect:

$$OC_{i} = \beta_{0} + \beta_{1} \cdot SAR_{i} + \beta_{2} \cdot mud_{i} + \beta_{3} \cdot tau_{i} + \beta_{4} \cdot dist_{i} + \beta_{5} \cdot phy_{i} + \beta_{6} \cdot temp_{i} + \vartheta_{i},$$

with coefficients β_0 to β_6 and error term ϑ_i .

In the first stage of the IV regression, the F-statistic indicates how well the instrument variables explain the endogenous variable. A high F-statistic result (typically above 10) in the first stage suggests that the instrument variable (*depth_dummy*) is well-suited to explain the endogenous variable. In our case, the F-statistic of 81.25 indicates that the instrument variable is a strong instrument for estimating SAR. The 'Kleibergen-Paap rk Wald F'-statistic of 13.55 indicates that the instrument used is strong and relevant, addressing the potential endogeneity of SAR⁷⁶.

Regression results are provided in Supplementary Supplementary Table . The statistical model shows a highly significant (p<0.05) negative effect of SAR on OC when all relevant covariates are included. The sign of the effect is consistent with the OLS regression, but the effect size of SAR is increased in the IV approach, suggesting that the instrumented values of SAR overestimate the negative impact on OC.

Supplementary Table 5. Results of the multivariate linear regression on OC using the IV approach. Robust standard errors are in parentheses. Three outliers identified by Cook's Distance were removed. The p-values are based on two-sided tests. No adjustments were made for multiple comparisons. Kleibergen-Paap rk Wald F statistic: 13.55; First-stage Fstatistic: 81.25. ***p<0.01, **p<0.05, *p<0.1.

Variables	Reg. 1	Reg. 2	Reg. 3	Reg. 4	Reg. 5
SAR	-1.856	-1.4987*	-0.9968**	-1.0515**	-0.943**
	(1.2592)	(0.8182)	(0.4058)	(0.4694)	(0.3766)
Mud (%)	0.0584*	0.0535**	0.0414***	0.0441***	0.0368***
	(0.0315)	(0.0226)	(0.0117)	(0.0143)	(0.0102)
Tou (Po)		1.3235	-1.3175	-2.5619	-2.0884
		(1.6266)	(1.1127)	(1.7846)	(1.5168)
			-0 004***	-	-
Dist (km)			(0.0012)	0.0037***	0.0042***
			(0.0012)	(0.0011)	(0.0011)
Dhy(ma/m3)				0.0041	0.0134**
Phy (mg/m³)				(0.0037)	(0.0063)
					-
Temp (°C)					0.2255***
					(0.0832)
SAR (95% CI Lower)	-4.3252	-3.1033	-1.7926	-1.9721	-1.6816
SAR (95% CI Upper)	0.6131	0.1059	-0.201	-0.1309	-0.2044
Kleibergen-Paap rk Wald F	5.09	7.8928	13.481	11.2826	13.5526
statistic					
First-stage F-statistic	187.2627	139.0133	110.1923	91.0809	81.2519

Reference

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