



Bioaccumulation as a driver of high MeHg in the North and Baltic Seas

David J. Amptmeijer¹, Elena Mikheeva¹, Ute Daewel¹, Johannes Bieser¹, and Corina Schrum^{1,2}

¹Matter Transport and Ecosystem Dynamics, Helmholtz-Zentrum Hereon, Geesthacht, Germany

²Universität Hamburg, Institute for Marine Sciences, Mittelweg 177, 20146 Hamburg, Germany

Correspondence: David J. Amptmeijer (davidamptmeijer@gmail.com)

Abstract. Mercury (Hg) is a toxic pollutant that poses significant risks to marine ecosystems and human health as a result of bioaccumulation. Despite its known hazards, the processes that govern Hg bioaccumulation within the marine food web are poorly understood. This study examines the role of the marine ecosystem in Hg cycling in highly productive coastal seas. We integrate Hg biotic uptake, release and transformation into the ECOSMO E2E marine ecosystem model, coupled with the MERCY v2.0 marine Hg cycling model. Our results show that bioaccumulation can increase total methylmercury (tMeHg) in coastal pelagic waters from 0.059 to 0.092pM, a 44% increase. Bioaccumulation and binding of Hg to organic matter contribute to elevated Hg levels in surface waters. Furthermore, cyanobacteria-driven reduction of Hg²⁺ to Hg⁰ reduces average marine Hg concentrations by up to 9% above the mixed layer depth in the Gotland Deep and 20% in shallow Baltic Sea regions, and increases Hg⁰ evaporation in the Baltic Sea, reducing Hg inflow into the North Sea. We quantify a 1% increase in tMeHg per 4.5 mgC m⁻³ biota biomass. Finally, we show that bioaccumulation decreases the burial of Hg by 13 kg y⁻¹ increasing Hg export to the Atlantic Ocean and the English Channel. These findings highlight the importance of ecosystem feedback on marine Hg cycling and demonstrate the need to integrate biological processes into Hg cycling models.

1 Introduction

Mercury (Hg) is a naturally occurring toxic element that is extremely persistent in the environment (Driscoll et al., 2013). It can be methylated to methylmercury (MeHg) by marine microorganisms, a dangerous neurotoxin that can bioaccumulate in the marine food chain (Trevors, 1986; Mason et al., 1995). Hg can reach levels 10⁶ times higher in fish than in the surrounding water and can reach levels in seafood that are unsafe for human consumption, especially during pregnancy or for developing children (Hsi et al., 2016; Counter and Buchanan, 2004; Outridge et al., 2018). The toxicity and emission of Hg received worldwide attention in 1956 when Minamata Bay in Japan was polluted with large amounts of Hg. This Hg was methylated by microorganisms and bioaccumulated in the marine ecosystem. Due to the consumption of polluted marine wildlife, more than 1000 people died and more were permanently disabled (Harada, 1995). Efforts to control Hg emissions culminated in the Minamata Convention on Mercury, which is a pledge to reduce Hg emissions and is currently signed by 128 countries (Outridge et al., 2018).



To ensure the effectiveness of the Minamata Convention in controlling the level of Hg pollution, the status of Hg emissions is periodically reviewed in the effectiveness evaluations of the Minamata Convention. In the first draft of the 2023 effectiveness evaluation, it is stated that Hg measurements in the air, biota, and humans are the key parameters to monitor the current risk of Hg (UNEP, 2021). Because samples of high-trophic-level animals are easily available in the form of commercial fish and their Hg levels are comparatively high and thus easier to measure accurately, Hg biomonitoring is often focused on assessing the Hg levels in high trophic levels. In the effectiveness evaluation, it is stated that Hg levels in water can be insightful, but because of the complexity of sampling and the variability in the data, it does not receive the same recommendation for sampling as fish and wildlife receive. This can lead to insufficient data to understand the cycling and bioaccumulation of marine Hg at the base of the food web, although these processes are essential in linking Hg emissions to (Me)Hg concentrations in seafood. Modeling studies are a perfect tool to improve our understanding of these complex interactions and can help evaluate the effectiveness of Hg reduction strategies.

Because MeHg formation and subsequent bioaccumulation in seafood are the dominant source of Hg exposure to humans, Hg levels in the world's oceans are of special concern. Oceanic Hg levels are influenced by a variety of sources. Hg can enter the ocean from the atmosphere through atmospheric exchange of Hg^0 or wet deposition of oxidized Hg^{2+} . Along with the atmosphere, Hg can enter the global ocean; it can enter through rivers, sea ice melt, coastal erosion, and hydrothermal vents (Zagar et al., 2006). Hg can be released from the ocean by the evaporation of volatile Hg^0 and dimethylmercury (DMHg), or it can be buried in the sediment, removing it from the biosphere (Van Veen et al., 2002; Zagar et al., 2006; Outridge et al., 2018). The dominant species of Hg in surface water is inorganic Hg^{2+} . Hg^{2+} and Hg^0 are in a dynamic equilibrium, but in water, this equilibrium favors Hg^{2+} . Although Hg^0 can evaporate, Hg^{2+} can be methylated into 2 forms of organic Hg - monomethyl mercury (MMHg^+) and double methylated DMHg. Both forms are highly toxic, but only DMHg is volatile and can evaporate from surface water, while MMHg^+ can bioaccumulate (Morel et al., 1998). The role DMHg in bioaccumulation is unknown. The sum of MMHg^+ and DMHg is referred to as MeHg. Since only MMHg^+ bioaccumulates, the term MeHg, in this paper, refers to the total methylated fraction of Hg in seawater. Under anoxic conditions, Hg^{2+} can be bound to S^{2-} to form cinnabar (HgS), which is considered a sink due to its low solubility (Oliveri et al., 2016). In seawater, due to an abundance of chloride, Hg^{2+} and MMHg^+ exist primarily in the form of inorganic chlorine complexes. The neutral forms of these complexes, HgCl_2 and MMHgCl compounds, can diffuse through cell membranes due to their lipophilic nature or bind to organic matter (Zhong and Wang, 2009). The speciation of Hg with organic carbon in the marine ecosystem, such as detritus and DOM, is a complex interaction that can influence the speciation, solubility, mobility, membrane permeability, and toxicity of Hg (Ravichandran, 2004). In summary, there are three fractions of Hg^{2+} and MMHg^+ in our model. HgCl and MMHgCl , referred to as Hg^{2+} and MMHg^+ in this paper, Hg -DOM and MMHg -DOM and Hg -detritus and MMHg -detritus.

Bioaccumulation of Hg occurs when species take up Hg at a rate higher than that at which it is excreted (Bryan, 1979). The most important step in bioaccumulation is the uptake of Hg directly from the water through respiration, absorption, or swallowing (Lee and Fisher, 2016). This can lead to a concentration of Hg inside organisms up to 100,000 times higher than in surrounding water. This process is called bioconcentration and is especially important at the base of the food web. The second process is the increase of Hg with increasing trophic position, which is called biomagnification. As a result of biomagnification,



already high bioconcentrated Hg values in phytoplankton can be amplified to extremely high values in high trophic animals, such as predatory fish, marine mammals, and seabirds (Lavoie et al., 2013).

Bioconcentration can be measured in laboratory studies as it is the equilibrium between dissolved Hg and Hg concentrated in biota. In the case of Hg, several chemical forms of Hg, such as Hg^{2+} and MMHg^+ can bioconcentrate at the same time and have different bioconcentration rates (Mason et al., 1996). Biomagnification is more variable because it depends on the trophic transfer efficiency of both biomass and Hg and their excretion rates, which can be influenced by various factors such as life cycle, water temperature and diet (Borgå et al., 2004). Biomagnification can be estimated in nature by sampling stable carbon and nitrogen isotopes with Hg to assess both the Hg content and the trophic position of a series of species (Lavoie et al., 2013). The biomagnification factor can then be calculated based on the increase in Hg per increase in the trophic position (Mackay and Fraser, 2000). The bioconcentration of Hg^{2+} and MMHg^+ can occur at every trophic level. However, it is mainly dependent on the surface area of organic membranes that are in contact with water and therefore dominated by microorganisms such as phytoplankton (Mason et al., 1996).

The uptake of Hg^{2+} and MMHg^+ into phytoplankton is a complex 2-step process in which Hg binds first to the phytosphere before it is absorbed into the cell. Recent data suggests that MMHg^+ uptake is influenced by cell-dependent factors such as phytosphere thickness and availability of transmembrane channels for MMHg^+ transport, while this is not the case for Hg^{2+} (Garcia-Arevalo et al., 2024). When phytoplankton is grazed, protein-bound MMHg^+ uptake is much more efficient than lipid-bound Hg^{2+} . This leads to an increased biomagnification factor for MMHg^+ compared to Hg^{2+} (Mason et al., 1996). This results in much higher levels of MMHg^+ in high trophic animals compared to Hg^{2+} , although the dissolved concentration of Hg^{2+} is generally higher than the concentration of MMHg^+ by an order of magnitude (Bieser et al., 2023).

In addition to Hg^{2+} and MMHg^+ uptake, another role for phytoplankton in Hg cycling is demonstrated by Kuss et al. (2015). Their research showed that certain species of cyanobacteria in the Baltic Sea (notable *Synechococcus* and *Aphanizomenon*) can also react with Hg by reducing dissolved Hg^{2+} to dissolved gaseous Hg^0 . Since Hg^0 is volatile and can evaporate, increasing the fraction of Hg^0 can reduce the Hg. This process is referred to as biogenic reduction.

Marine Hg cycling and bioaccumulation modeling have received attention in the past, with a strong focus on MMHg^+ bioaccumulation modeling. A model using the Shear realistic water column Turbulence Resuspension Mesocosms in the Beaufort Sea was presented by Kim et al. (2008); they incorporated the bioaccumulation and trophic transfer of MMHg^+ in a pelagic food web and the benthic-pelagic coupling to study the effect of sediment resuspension on MMHg^+ bioaccumulation. Schartup et al. (2018) made a non-spatial model of MMHg^+ uptake and trophic transfer; they model bioaccumulation and trophic transfer of MMHg^+ at the base of the food web. Their non-spatial model is driven by observations of aquatic MMHg^+ and successfully reproduces observed bioaccumulated MMHg^+ concentrations in mesozooplankton. Zhang et al. (2020) developed a global model for MMHg^+ uptake in plankton; they modeled bioaccumulation by assuming an instant equilibrium between marine MMHg^+ and phytoplankton, and consequently modeled bioaccumulation into two functional groups of primary consumer zooplankton that were differentiated by their size. Bioaccumulation in these zooplankton species was estimated based on the concentration of MMHg^+ in the phytoplankton they consume and a size-specific MMHg^+ elimination rate. Rosati et al. (2022) published a model for the cycling and bioaccumulation of MMHg^+ in plankton in the Mediterranean Sea; they developed a coupled 3D Hg



biogeochemical transport model to assess the cycling of Hg^{2+} , Hg^0 , MMHg^+ and DMHg . Bioaccumulation was incorporated
95 by modeling the bioconcentration of MMHg^+ in phytoplankton and the consequent biomagnification to zooplankton when
they consume phytoplankton. Together, these models create a strong base for understanding the coupling of MMHg^+ in the
marine environment to bioaccumulation at the base of the food web. This is expanded on by Bieser et al. (2023); which couples
atmospheric Hg concentrations and deposition with MMHg^+ bioaccumulation in the midtrophic level fish Herring (*Clupea*
harengus), while taking into account the bioconcentration of Hg^{2+} and MMHg^+ for all biota. This is done by integrating the
100 MERCY v2.0 model to the output of the 3D ECOSMO-HAMSOM coupled system, which models biogeochemistry and hy-
drodynamics in the North and Baltic Seas on a spatial and temporal scale. In this way, the bioaccumulation of MMHg^+ could
be coupled to atmospheric Hg cycling without having the models interact at run-time.

In the expansion of these approaches, we implemented the bioaccumulation of Hg^{2+} and MMHg^+ in a fully coupled model,
which includes 2 trophic levels of fish. In addition to being fully coupled, it expands on previous models by including high-
105 trophic-level fish while explicitly tracking the trophic level of all biota. Similarly as in Bieser et al. (2023), we incorporate the
Hg cycle and active and passive uptake of both Hg^{2+} and MMHg^+ at every trophic level. We analyze bioaccumulation in the
North and Baltic Seas in idealized 1D watercolumns, as this region provides varied hydrodynamical environments that we can
use to test our model. Then we evaluate the importance of our findings in the same offline coupled 3D ECOSMO-HAMSOM-
MERCY system presented in Bieser et al. (2023) to estimate the effect of bioaccumulation on the Hg cycle and the Hg and
110 MeHg budget.

The North and Baltic Seas are shelf seas in North-Western Europe. The Baltic Sea is a brackish sea of approximately 377,000
 km^2 , which is connected to the 575,000 km^2 large North Sea via the Danish straits. Both seas are important sources of seafood
and 2 million tonnes of metric tonnes are landed annually (ICES, 2022). Hg input into the North and Baltic Seas is dominated
by riverine input and atmospheric deposition (Kwasigroch et al., 2021).

115 In this study, we hypothesize that the ecosystem can influence Hg cycling in several ways, and our aim is to quantify its
effect on Hg cycling. To support this goal, we quantify the feedback of the ecosystem on marine Hg cycling by modeling a fully
coupled Hg speciation and bioaccumulation model with and without bioaccumulation, the complexation of Hg with detritus and
labile DOM, and the biogenic reduction of Hg^{2+} to Hg^0 facilitated by cyanobacteria. Then we analyze the difference between
the scenarios. In this article, we present the results of model runs in the North and Baltic Seas, using three idealized 1D
120 water column setups that represent permanently mixed, seasonally mixed, and permanently stratified water column conditions.
Furthermore, we run the model in a 3D configuration of the North and Baltic Seas to analyze the spatial variation of the
impact of the ecosystem on Hg cycling and quantify the overall effect of these interactions on the Hg budget of these seas. We
investigated the role of the ecosystem on both the total Hg budget and the aquatic Hg fraction. The total Hg (tHg) and total
MeHg (tMeHg) refer to all Hg and MeHg, including what is bioaccumulated. The aquatic Hg and aquatic MeHg refer to all
125 Hg species that are in water but are not bioaccumulated. This is shown in Table 1 for clarity.

The 3D configuration used in this study is based on the model of (Bieser et al., 2023), which is modified to investigate
scenarios in which we change the ecosystem drivers of Hg speciation to assess their impact.



Table 1. Definitions of Hg abbreviations.

Abbreviation	Meaning
Hg	Refers to Hg in general
Hg ²⁺	Dissolved Hg (Bioaccumulates)
Hg ⁰	Elemental Hg (Volatile)
MMHg ⁺	Monomethylmercury (Bioaccumulates, extremely toxic)
DMHg	Dimethylmercury (Volatile, extremely toxic)
MeHg	MMHg ⁺ + DMHg
tHg	All Hg, including what is bioaccumulated
tMeHg	All MeHg, including what is bioaccumulated
Bioaccumulated Hg ⁺	All Hg ²⁺ that is bioaccumulation, does not include what is partitioned to detritus and DOM
Bioaccumulated MMHg ⁺	All Hg ²⁺ that is bioaccumulation, does not include what is partitioned to detritus and DOM
Aquatic Hg	All Hg excluding what is bioaccumulated
Aquatic MeHg	All MeHg excluding what is bioaccumulated

2 Methodology

To evaluate the role of the ecosystem, we quantify the impact of several processes on total and aquatic Hg concentrations. In the first part of this section, we specify how these processes are implemented in the different scenarios that are simulated. The second part focuses on the models, setups, and parameterizations used in this study.

2.1 Processes

To quantify the impact of ecosystem interactions on marine Hg cycling, we evaluated three processes:

- Bioaccumulation
- Biogenic reduction
- Partitioning to detritus and labile-DOM

Bioaccumulation is the uptake of Hg²⁺ or MMHg⁺ from the water. When Hg is bioaccumulated, it cannot evaporate or undergo speciation and photolysis, but it will be transported with the organism that bioaccumulated the Hg. We are interested in this, as the bioaccumulation of Hg removes aquatic Hg²⁺ and MMHg⁺, which undergoes speciation. Aquatic Hg²⁺ and MMHg⁺ is removed during the phytoplankton bloom, and it is released when the bloom period is over. This seasonal removal and release has the potential to influence Hg cycling. Moreover, Hg accumulated in the biota is protected from photolysis, making the accumulated Hg species more stable than that in the surrounding water.



Biogenic reduction is the reduction of Hg^{2+} to volatile Hg^0 by cyanobacteria. It has been shown to be an important interaction during cyanobacterial blooms in the Baltic Sea (Kuss et al., 2015). Here, we attempt to quantify the impact that cyanobacterial blooms have on Hg cycling in the Baltic Sea. Biogenic reduction does not play a role in our North Sea setups, as there are no cyanobacteria in the North Sea in the ECOSMO E2E model (Daewel and Schrum, 2013). That agrees with the research of Kuss et al. (2015), where cyanobacteria-induced biogenic reduction was found in cyanobacteria that are specific to the Baltic Sea.

Partitioning to organic carbon is represented in the model by the binding of Hg^{2+} and MMHg^+ to detritus and labile-DOM. Hg associated with organic carbon can be ingested by scavengers, contributing to the bioaccumulation process. Alternatively, when the detritus sinks, the Hg bound to it is transported to deeper water layers. In this way, the binding of Hg to organic carbon not only facilitates a flux of Hg to deeper water but also can deliver Hg from the upper water layers directly to scavenging animals. The only organic carbon particles for which the 1D setups account are detritus and labile-DOM originating from the ECOSMO E2E model.

The simulation scenarios of our research model are visualized in Fig. 1. Here we run the model with the following interactions enabled: (base case) without bioaccumulation but all other interactions (scenario A), without bioaccumulation and biogenic reduction (scenario B), without partitioning to detritus and labile-DOM (scenario C), and without any of the previously described interactions (scenario D). Since there is no biogenic reduction in the North Sea, the base case is only compared to scenarios B and C in the North Sea setups.

2.2 Models

In this study, we used two model systems with different purposes. The first system is idealized 1D water column setups, which is used to generalize our findings. The second is a 3D setup, which is used to analyze the spatial patterns and estimate the budgets of the North and Baltic Sea.

2.2.1 1D water column model

For the 1D system, the model design is shown in Fig. 1. We use the Generalized Ocean Turbulence Model (GOTM) to simulate the hydrodynamics of the 1D water column setups, the ECOSMO E2E ecosystem model to simulate the marine ecosystem and the MERCY v2.0 model to simulate Hg cycling. (Daewel et al., 2019; Bieser et al., 2023; Burchard et al., 1999). The models are coupled using the Framework for Aquatic Biogeochemical Modeling (FABM) (Bruggeman and Bolding, 2014). ECOSMO E2E, MERCY v2.0, and bioaccumulation models are implemented through FABM and coupled to GOTM using this framework. This coupling allows us to simulate the effect of the marine ecosystem from the ECOSMO E2E model, on the Hg cycling in the MERCY v2.0 model under the influence of the hydrodynamics from the GOTM model.

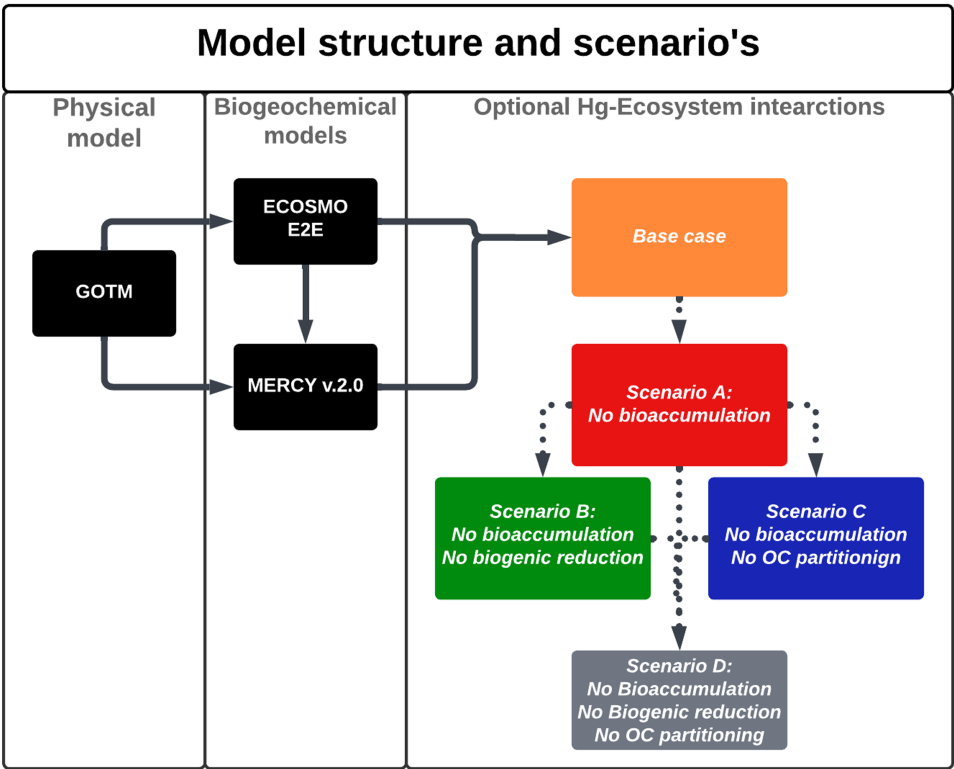


Figure 1. Model setup diagram. The GOTM model drives the ECOSMO E2E Ecosystem model and MERCY v2.0 Hg speciations model. These models drive different setups with optional Hg-Ecosystem interactions. The impact of the ecosystem is evaluated by comparing the base case to a scenario without; bioaccumulation (scenario A), bioaccumulation and biogenic reduction (scenario B), bioaccumulation and partitioning to detritus and DOM (scenario C), and all mentioned ecosystem interactions (scenario D).

175 **2.2.2 GOTM**

The GOTM model is a 1D model that calculates the turbulence of a vertical 1D water column setup by computing the solutions to the one-dimensional version of the transport equation of momentum, salinity and temperature, while being nudged to observational datasets. Therefore, our model has only a vertical resolution and no horizontal resolution. GOTM calculates variables of the physical state (temperature, salinity and density), vertical transport (advection, diffusion, turbulence, and sinking) and surface processes (surface elevation, friction, and velocity). GOTM communicates these state variables to the biochemical models using the FABM interface.

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2.2.3 MERCY

The MERCY v2.0 model is a comprehensive Hg cycling model. It includes the speciation of Hg^0 , Hg^{2+} , HgS , MMHg^+ , and DMHg in water, sediment, and biota. With this model, we implement the partitioning of Hg^{2+} and MMHg^+ to organic carbon and its speciation between the dissolved, particulate and colloidal phases. In addition, atmospheric deposition of Hg, its air-sea and sediment-water exchanges are also resolved in this model. In 1D setups, the MERCY v2.0 model is fully coupled with the ECOSMO E2E-GOTM system.

2.2.4 The ECOSMO E2E ecosystem model

The ECOSMO E2E (ECOSystem Model End-to-End) ecosystem model (Daewel et al., 2019) is an extended version of the ECOSMO II model. ECOSMO E2E includes higher trophic levels, such as macrobenthos and fish. In the 3D setups, this model includes the following 7 biological functional groups: cyanobacteria, flagellates, diatoms, microzooplankton, mesozooplankton, fish, and macrobenthos. In the 1D setup, there are two functional groups of fish, 1 representing lower trophic level pelagic fish such as herring or sprat, and 1 representing a benthic top predator, such as cod. The model simulates nutrient cycles (nitrogen, phosphorus, and silicon), oxygen dynamics, and sedimentation processes.

2.3 Modeled regions

To generalize our findings, the scenarios are tested using the 1D GOTM model in 3 setups with hydrodynamics common for coastal areas. They are designed around physical and biogeochemical regimes representing specific locations in the North and Baltic Seas. For this, we use a 1D ocean water column model to simulate a permanently mixed, a seasonally stratified, and a permanently stratified water column. This allows us to compare our findings across different physical and biogeochemical regimes.

2.3.1 Permanently mixed - Southern North Sea

The first North Sea setup is located in the Southern North Sea ($54^{\circ}15'00.0''\text{N}$ $3^{\circ}34'12.0''\text{E}$) with a depth of 41.5 m. This area is characterized by strong tidal mixing, resulting in high dissolved oxygen concentrations and temperatures throughout the water column. Remineralized nutrients are quickly available for phytoplankton production as they can be mixed with surface water throughout the year. The constant mixing of the water column results in the temperature fluctuating between 5 and 15 $^{\circ}\text{C}$ for the whole water column depending on the season (Van Leeuwen et al., 2015). The setup location is 170 km away from the mainland, so despite being in the Southern North Sea it does have characteristics of the open North Sea. This region is characterized by high primary production between 60 and 80 $\text{g C m}^{-2} \text{y}^{-1}$ (Daewel and Schrum, 2013). The constant mixing allows macrobenthos to feed directly from the phytoplankton bloom, leading to a high macrobenthos stock (Heip et al., 1992). 41.5 m is also deep enough to support larger species of fish, such as herring and cod.

The available concentrations of dissolved silicate limit the growth of diatoms in the North Sea. Diatoms can dominate at the start of the bloom, but other phytoplankton species will take over the bloom after the silicate is depleted. The second flagellate



bloom in the North Sea typically exceeds the first and can continue until light, nitrogen, or phosphorous becomes limiting (Peeters et al., 1991).

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2.3.2 Seasonally mixed - Northern North Sea

The second setup is in the Northern North Sea ($57^{\circ}42'00.0'' N$ $2^{\circ}42'00.0'' E$) with a depth of 110 m. This setup has seasonal stratification in summer and complete mixing of the water column in winter (Van Leeuwen et al., 2015). This results in good growth conditions for phytoplankton in spring on the surface and a subsurface bloom in summer, when nutrients become depleted on the surface. In fall and winter, nutrients from the deeper water layers are mixed back into the surface. During summer, the temperatures in the Northern North Sea setup increase at the surface to $15\text{--}20^{\circ}C$ but remain at $6^{\circ}C$ in deeper water.

Primary production in the Northern North Sea is typically lower than in the Southern North Sea and ranges from 40 to $60\text{ g C m}^{-2}\text{ y}^{-1}$ (Daewel and Schrum, 2013). Furthermore, seasonal stratification means that macrobenthos cannot feed directly from the phytoplankton bloom. This leads to smaller macrobenthos communities than in the Southern North Sea. Fish are better adapted to the Northern North Sea environment, which results in high stocks of fish. The Northern North Sea has a similar nutrient regime as the Southern North Sea, in which silicate limits diatom growth, and flagellates can be limited by light, nitrogen, or phosphorous (Peeters et al., 1991).

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2.3.3 Permanently stratified - Gotland Deep

The third setup is the Gotland Deep ($57^{\circ}18'00.0'' N$ $20^{\circ}00'00.0'' E$) located in the central Baltic Sea. It is 249 m deep and is characterized by a permanent halo and oxycline. During spring and summer, an additional thermal stratification is established. This causes temperatures of $6^{\circ}C$ and preserves anoxia in deep water. The surface water of the Gotland Deep setup remains oxic throughout the year and has temperatures fluctuating between 5 and $20^{\circ}C$. The Gotland Deep setup is also considerably less saline than the North Sea and has a salinity of 12 and 7 g Kg^{-1} in the deep and surface water, respectively (Nausch et al., 2003). This is caused by major freshwater riverine input combined with sporadic and limited inflow from the North Sea (Lehmann et al., 2021). Phytoplankton growth in the Baltic Sea is typically dominated by diatoms, but flagellates can form a large portion of the biomass in open-water areas, such as the Gotland Deep. The autumn bloom is usually dominated by cyanobacteria. Primary production is estimated to be between 20 and $40\text{ g C m}^{-2}\text{ y}^{-1}$ in this region (Daewel and Schrum, 2013). The open Baltic Sea is a productive fishing region with a fish biomass density as the North Sea; however, due to the anoxic bottom water, macrobenthos in the Gotland Deep is extremely limited. Studies in the Gdansk Deep found no macrobenthos with a population density significant for benthopelagic coupling anoxic conditions, and similar conditions can be expected in the Gotland Deep (Kendzierska and Janas, 2024).

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The Gotland Deep has silicate concentrations similar to those of the North Sea but with less nitrogen and phosphorous. Because of this, the available silicate is enough to support diatoms as the most abundant phytoplankton group in the Baltic Sea. The heavily stratified nitrogen-limited surface layer forms an ideal growth environment for cyanobacteria, which can fix nitrogen from the atmosphere. In this way, they can use the available phosphorus, until either phosphorus becomes limiting or
250 the end of the bloom seasons causes light to become limiting (Savage et al., 2010).

2.3.4 3D North and Baltic Seas

In addition to the idealized 1D setups of the different regimes, the total effect of the ecosystem on Hg cycling in the North and
255 Baltic Seas is quantified using a 3D model of the entire region. Both 1D and 3D models are used because the 1D models allow us to analyze different ecosystem interactions in idealized circumstances, while the 3D models allow us to evaluate the total effect of the ecosystem on Hg cycling, sedimentation, and evaporation in the model domain while assessing the effect on the ecosystem on the total Hg budget of the North and Baltic Seas.

2.4 3D spatial model

260 In addition to 1D simulations, the effect of the ecosystem is analyzed by running the MERCY v2.0 3D Hg bioaccumulation and speciation model described in Bieser et al. (2023) in the North and Baltic Seas with and without bioaccumulation and biogenic reduction. MERCY v2.0 model uses the 3D hydrodynamic model HAMSOM (Hamburg Shelf Ocean Model) as a physical model. In this setup, the HAMSOM-ECOSMO hourly output is used to drive the MERCY v2.0 model, making it effectively an offline coupled system. The model is run from the beginning of 2011 to the end of 2015. The first 4 years are
265 used as spinup and the final year is used for the analyses. The model is run in its default setup, without bioaccumulation or biogenic reduction. The effect of bioaccumulation on both tHg and tMeHg is visualized by plotting the relative difference in tHg and tMeHg caused by the ecosystem. The data is visualized using the cartopy package in Python version 3.11.2.0.

2.5 HAMSOM

The HAMSOM (Hamburg Shelf Ocean Model) is a physical hydrodynamic ice-ocean model (Backhaus, 1983; Schrum and
270 Backhaus, 1999). It is directly coupled to the ECOSMO (ECOSystem Model) to form HAMSOM-ECOSMO (Schrum et al., 2006). This coupling allows the ecosystem to directly influence the physical system, for example, through light absorption by the biota. The HAMSOM-ECOSMO system is used to model the North and Baltic Seas. It is a comprehensive simulation of both physical and biogeochemical drivers in the North and Baltic Seas and simulates key drivers such as riverine and atmospheric nutrient input, horizontal advection and sedimentation, and burial of organic matter. This system provides the
275 physical, chemical, and ecosystem data that drive the MERCY v2.0 model.



2.6 Model development

The ECOSMO E2E ecosystem model is originally designed to represent organic matter based on macronutrient fluxes, such as nitrogen, phosphate, and silicate fluxes. This means that certain interactions of the marine ecosystem that could biomagnify Hg, such as predation of species within the same functional group or even cannibalism, which do not alter nutrient fluxes or organic matter stocks, are not explicitly specified in the model (Montagnes and Fenton, 2012; Arrhenius and Hansson, 1996; Schrum et al., 2006). Because of this, the carbon assimilation efficiency of consumers is reduced, compared to the previously published ECOSMO E2E parameterization, so biota needs to take up more carbon for growth and can accumulate more Hg. To compensate for the reduction in carbon intake, the mortality and respiration rate of zooplankton is reduced to a value that is lower than in Daewel et al. (2019), but still within the range used in previously published models (Cruz et al., 2021). This is verified by modeling the trophic level of all biota. This allows us to compare the modeled biomagnification factor to observations and evaluate if our model simulate bioaccumulation in line with observations of animals with the same trohic level. The phytoplankton is parameterized as shown in Table 2. To compensate for the increased zooplankton grazing, the growth rate is increased compared to the previously published ECOSMO E2E version, but remains within the experimentally observed range (Stelmakh and Kovrigina, 2021). All other values are the same as in (Daewel et al., 2019). One notable difference between the ECOSMO E2E model in Daewel et al. (2019) and used in (Bieser et al., 2023) and the version used in this study is the inclusion of a second fish functional group. The fish functional groups are referred to as fish 1 and fish 2. Both are parameterized by the same set of equations as described in the published ECOSMO E2E model, but differ in feeding preferences and rates (Daewel et al. (2019)). Fish 1 preferably feeds on mesoplankton and microzooplankton and feeds on detritus and macrobenthos with lower preference. Fish 1 in the model represents a variety of smaller, mainly pelagic, fish such as herring (*Clupea harengus*) and European sprat (*Sprattus sprattus*). The fish 2 group is modeled as a benthic top predator and prefers to eat macrobenthos, but can also feed on mesozooplankton, fish 1, and detritus. It is mainly representative of large demersal species, such as cod (*Gadus spp.*), but would as a functional group also include other large benthic species such as whiting (*Micromesistius poutassou*) or haddock (*Melanogrammus aeglefinus*). Fish 2 is at the head of the food chain and therefore is not pre-dated in the model. Instead, all loss terms are implicitly included in the mortality term. Table 3 shows the parameterization of the ECOSMO E2E model setup used in this study.

Table 2. Dimensions, shape, and maximum growth and mortality rates of most common phytoplankton species in the North and Baltic Seas, to resemble ECOSMO E2E functional groups and the conversion ratio of mg C to cm² cell membrane and dm³ cell volume. The dimensions and shapes are based on Olenina et al. (2003). For the Cylinder they are radius and height, for the sphere and hemisphere - the radius. The maximum growth rate and the mortality are based on Daewel et al. (2019).

Group	Species	Shape	Dimensions (μm)	μ_{max} (d^{-1})	Mortality (d^{-1})
Diatoms	<i>T. baltica</i>	Cylinder	12.5 x 25	1.4	0.04
Flagellate	<i>P. Catanata</i>	Hemisphere	6	1.2	0.04
Cyanobacteria	<i>A. flos-aquae</i>	Sphere	4	1.0	0.08



Table 3. The parameterization of consumers. The preference and consumption rate (r_{cons}) determine how much of each prey is consumed, the assimilation efficiency (AE) how much of the consumed carbon is assimilated into the predator; the mortality (r_{mort}) and respiration (r_{resp}) rate determine the total loss. Adjustments were made for higher trophic levels compared to Daewel et al. (2019) to enhance the model's suitability for bioaccumulation. Specifically, the AE and grazing rate were lowered. Additionally, fish 2 was introduced as a top predator with modified parameters. It has a higher preference for macrobenthos and consumes fish 1 rather than microzooplankton. As a top predator, fish 2 has a lower AE and consumption rate compared to fish 1. All other rates and equations remain consistent with the ECOSMO E2E model (Daewel et al., 2019).

Group	Prey	Preference (1)	r_{cons} (d^{-1})	AE (1)	r_{mort} (d^{-1})	r_{resp} (d^{-1})
Microzooplankton	Diatom	0.25	0.8	0.75	0.05	0.02
	Flagellates	0.70	0.8	0.75		
	Cyanobacteria	0.30	0.3	0.75		
	Detritus	0.10	0.8	0.8		
Mesozooplankton	Diatom	0.85	0.7	0.6	0.025	0.015
	Flagellates	0.10	0.7	0.6		
	Cyanobacteria	0.30	0.3	0.6		
	Microzooplankton	0.15	0.8	0.6		
	Detritus	0.10	0.7	0.6		
Macrobenthos	Phytoplankton	0.2	0.1	0.6	0.01	0.001
	Zooplankton	0.2	0.2	0.6		
	Sediment	0.15	0.15	0.6		
	Detritus	0.15	0.15	0.6		
	DOM	0.15	0.1	0.6		
Fish 1	Microzooplankton	0.45	0.015	0.5	0.001	0.002
	Mesozooplankton	0.45	0.015	0.5		
	Macrobenthos	0.05	0.015	0.5		
	Detritus	0.05	0.0125	0.5		
Fish 2	Mesozooplankton	0.25	0.012	0.45	0.001	0.002
	Fish 1	0.25	0.012	0.45		
	Macrobenthos	0.45	0.013	0.45		
	Detritus	0.05	0.001	0.45		

Trophic interactions play a crucial role in our model; therefore, the trophic positions of the biota are also explicitly modeled. We modeled the trophic level in two different ways.

- **Trace organic carbon origin from biota** - Initially, we assumed that the trophic level of pelagic detritus and DOM is the same as that of the organisms from which it originates.



305 – **Standardize organic carbon trophic level as 1** - Afterward, we defined the trophic level of detritus and DOM as 1.0.
This is done because Hg associated with detritus and DOM is assumed to have an instantaneous equilibration with Hg,
rather than storing it as other state variables do.

Tracking the trophic level through the detritus and DOM provides the actual trophic level of biota, which can be compared
with observations. The assignment of a fixed trophic level of 1.0 to the detritus and DOM emphasizes the number of trophic
310 interactions that contribute to the biomagnification of Hg. Since Hg in sediment is tracked through organic carbon in our model,
the trophic level of sediment detritus remains consistently modeled.

The modeled trophic levels are shown in Table 4. In the Gotland, Deep macrobenthos is absent because of the anoxic
conditions. Except for fish 2 in the Northern North Sea, all functional groups have trophic levels that are lower than observed
in the North and Baltic Seas. To compensate for this, the carbon uptake efficiency of all zooplankton that feed on detritus and
315 mesozooplankton feeding on plankton was lowered from 0.75 to 0.6, the carbon uptake efficiency for fish 1 from 0.7 to 0.5 and
for fish 2 from 0.7 to 0.45, as mentioned in the model development section and shown in Table 3.

Table 4. Observed and modeled trophic level of functional groups. Observed values for microzooplankton, mesozooplankton, and fish 1 are
based on Nfon et al. (2009), the trophic level for fish 2 is based on Jennings and Van Der Molen (2015), and for macrobenthos on Steger
et al. (2019).

Setup	Gotland Deep		Northern North Sea		Southern North Sea		Observed
OC Trophic Level	modeled	1	modeled	1	modeled	1	-
Microzooplankton	2.1	2.0	2.2	2.0	2.1	2.0	2.00
Mesozooplankton	2.2	2.2	2.8	2.5	2.6	2.5	2.87
Fish 1	2.8	2.6	3.4	2.9	3.5	3.2	3.98
Fish 2	3.7	3.5	4.2	3.7	3.8	3.5	4-4.2
Macrobenthos	-	-	2.8	2.3	2.6	2.3	2-4
Detritus	1.4	1.0	1.5	1.0	1.3	1.0	-
DOM	1.3	1.0	1.6	1.0	1.4	1.0	-
Sediment	1.4	1.3	1.4	1.3	1.4	1.3	-



Bioconcentration

All biota in our model take up Hg^{2+} and MMHg^+ from the water column due to bioconcentration. The bioconcentration rate for phytoplankton depends on the cell surface area and the diffusivity rate of Hg^{2+} and MMHg^+ through the cell membrane (Mason et al., 1996). The surface area is estimated from the most common phytoplankton species in the three phytoplankton functional groups for the North and Baltic Seas. The species and dimensions are shown in Table 2. The dimensions and shapes of phytoplankton are based on Olenina et al. (2003). The carbon content per cell was estimated from the calculated cell volume for diatoms as $pgC = 0.288\mu l^{0.811}$ and for flagellates and cyanobacteria as $pgC = 0.216\mu l^{0.939}$ (Menden-Deuer et al., 2000). Due to limited information on phytoplankton release rates, they are estimated based on uptake rates and observed concentrations $((\text{Hg}^{\text{Aq}} * \text{uptake rate}) / \text{Hg}^{\text{Observed}} = \text{release rate})$ of Hg^{2+} and MMHg^+ .

Based on this, we implemented the change of bioconcentrated pollutant per day for a functional group (Hg^{2+} or MMHg^+) via the following equation:

$$f_{u(g,p)} = b_g * C_p * r_{bc(g,p)} - (r_{rel(p,g)} + r_{bl(g)} + \sum_{g=1}^{n_z} r_{pred(z,g)}) * C_{(p,g)} \quad (1)$$

$f_{u(g,p)}$ – the change in bioconcentrated pollutant p in functional group g [$\text{ng Hg m}^{-3} \text{ d}^{-1}$];

b_g – the biomass of the functional group g [mgC m^{-3}];

C_p – concentration of pollutant p [ng Hg m^{-3}];

$r_{bc(g,p)}$ – bioconcentration rate of group b on pollutant g [$\text{ng Hg mgC}^{-1} \text{ d}^{-1}$];

$r_{rel(p,g)}$ – release rate of pollutant p by the functional group g [d^{-1}];

$r_{bl(g)}$ – biological loss rate due to mortality and respiration of group g [d^{-1}];

n_z – total number of consumers that feed on group g;

z – consumers that feed on group g;

$r_{pred(z,g)}$ – the rate at which predator z feeds on group g [d^{-1}];

$C_{(p,g)}$ – concentration of pollutant p that is in the functional group g [ng Hg m^{-3}].

The bioconcentration, release, and turnover rates are shown in Table 5. In particular, Pickhardt and Fisher (2007) observes that Hg^{2+} accumulates at similar levels in all phytoplankton, while the accumulation of MMHg^+ exhibits an inverse relationship with cell size. Due to this, the uptake / loss rate ratio is the same for all phytoplankton groups for Hg^{2+} , however, only the loss rate is the same for all phytoplankton groups for MMHg^+ . The groups representing phytoplankton species with a smaller size and therefore a higher uptake rate also have a high Hg^{2+} release rate. As a result, all phytoplankton groups reach equilibrium at similar Hg^{2+} concentrations. However, the equilibrium between uptake and release for MMHg^+ is inversely related to cell size. This means that smaller cells will accumulate more MMHg^+ per biomass while accumulating a similar amount of Hg^{2+} , compared to larger cells. The difference in the excretion rates of Hg^{2+} and MMHg^+ can be explained by the different binding behaviors of Hg^{2+} and MMHg^+ in the cell. While Hg^{2+} binds to smaller compounds and adheres to the cell wall, MMHg^+ binds



to larger cytoplasmic proteins. This means that to excrete MMHg^+ , it needs to be demethylated to Hg^{2+} , or excreted using
 350 dedicated peptide membrane channels (Bridges and Zalups, 2010; Takanezawa et al., 2023).

Table 5. The estimated bioconcentration (r_{bc}), release rates (r_{rel}), and turnover rates (r_{to}) for phytoplankton (Mason et al., 1996), zoo-
 plankton (Tsui and Wang, 2004) and fish (Wang and Wong, 2003; Trudel and Rasmussen, 1997). Bioconcentration rates are in $\text{m}^3 \text{mg C}^{-1}$
 d^{-1} and release and turnover rates are d^{-1} (Trudel and Rasmussen, 1997; Pickhardt et al., 2006; Mason et al., 1996; Tsui and Wang, 2004).
 The bioconcentration rates for phytoplankton are based on the cell size, for zooplankton and macrobenthos on *Daphnia pulex*, and for fish on
Plectorhinchus gibbosus.

Functional group	Hg^{2+}			MMHg^+		
	$r_{bc}(\text{m}^3 \text{mg C d}^{-1})$	$r_{rel}(\text{d}^{-1})$	$r_{to}(\text{d}^{-1})$	$r_{bc}(\text{m}^3 \text{mg C d}^{-1})$	$r_{rel}(\text{d}^{-1})$	$r_{to}(\text{d}^{-1})$
Diatoms	3.2E-3	63.1	(-)	3.1E-3	0.75	(-)
Flagellates	3.2E-3	65.5	(-)	3.1E-3	0.75	(-)
Cyanobacteria	5.5E-3	109.7	(-)	5.4E-3	0.75	(-)
Microzooplankton	1.68E-5	0.03	0.03	2.22E-05	7.50E-01	7.50E-01
Mesozooplankton	1.68E-5	3.1E-3	3.1E-3	2.22E-05	1.50E-02	1.50E-02
Macrobenthos	1.68E-5	0.04	0.04	2.22E-05	2.50E-02	2.50E-01
Fish 1	3.90E-7	2.16E-2	4.47E-2	9.07E-6	2.90E-03	3.00E-04
Fish 2	3.90E-7	2.16E-2	4.47E-2	9.07E-6	2.90E-03	3.00E-04

Biomagnification in consumers

The bioaccumulation in consumers depends on both bioconcentration and biomagnification. The release of bioconcentrated
 Hg is referred to as the release rate, and the release of biomagnified Hg is the turnover rate. This difference is caused by a
 different way of accumulation: bioconcentrated Hg can adsorb on the gills of fish where it can be released at different rates,
 355 compared to Hg, absorbed in the gut tissues (Pickhardt and Fisher, 2007; Wang and Wong, 2003). In zooplankton, the same
 release and turnover rate is assumed. This is because zooplankton is so small that the bioconcentrated and biomagnified Hg
 can more easily homogenize throughout the animal (Tsui and Wang, 2004). The functional group-specific release and turnover
 rates are shown in Table 5.

The formulation for biomagnification is based on the consumption rates calculated by the ECOSMO E2E model, multiplied
 360 by an assimilation efficiency based on observations. The assimilation efficiency depends on the type of prey and is for Hg^{2+} 0.2
 when phytoplankton, 0.27 when consumers, and 0.13 when detritus or DOM is consumed. For MMHg^+ , when phytoplankton,
 detritus, or DOM is consumed, assimilation efficiency is set for 0.80, and it is set to 0.96 when consumers are consumed
 (Mason et al., 1995; Tsui and Wang, 2004; Wang and Wong, 2003). The increase in Hg for consumer g per day is implemented
 into the model by the following equation:

$$365 \quad f_{bm(g,p)} = \sum_{g=1}^{n_s} (r_{pred(g,s)} * a_{(s,p)}) - (r_{to(p,g)} + r_{l(g)} \sum_{g=1}^{n_z} r_{pred(z,g)}) * C_{p,g} \quad (2)$$



$f_{bm(g,p)}$ – the change in pollutant p originating from biomagnification in functional group g [$\text{ng Hg m}^{-3} \text{ d}^{-1}$];

s – all functional groups that consumer g predaes on;

$r_{pred(g,s)}$ – consumption rate of group g on group s [d^{-1}];

$a_{s,p}$ – assimilation efficiency of pollutant p when group s is consumed [unitless];

370 $r_{to(p,g)}$ – turnover rate [d^{-1}];

$r_{rel(p,g)}$ – release rate [d^{-1}].

Biomagnification follows the same loss process as bioconcentration, except that the turnover rate ($r_{to(p,g)}$) replaces the release rate ($r_{rel(p,g)}$).

375 It is important to note that extensive studies that separate the effects of bioconcentration and biomagnification are rare. Due to this, the estimated bioconcentration, turnover, and release rate of microzooplankton, mesozooplankton, and macrobenthos are all based on studies performed on pelagic water flea *Daphnia pulex*. This species is abundant in the Baltic Sea, but not in the North Sea. Salinity does not appear to have a direct effect on the bioconcentration of Hg in zooplankton and since *Daphnia* can have a size of 1-5 mm it is average in size for zooplankton (Reinhart et al., 2018). The fish rates are based on a study
380 investigating the uptake, release, and turnover rates in the Indo-Pacific species Harry hotlips *Plectorhinchus gibbosus* (Wang and Wong, 2003).

2.7 Model setup

1D setups

The 1D setups are run using the GOTM model mentioned earlier. The GOTM setups are based on observational data that is
385 used to generate setups using iGOTM (<https://igotm.bolding-bruggeman.com>). All GOTM simulations run for the period 01-01-1989 12:00:00 until 01-01-2011 12:00:00, of which the first 10 years are considered a spin-up period and the period from 01-01-2000 12:00:00 to 01-01-2011 12:00:00 is the actual simulation period and is used for analyses. These setups are based on gridded bathymetry data for water depth ($1/240^\circ$ resolution) (GEBCO Bathymetric Compilation Group, 2020), ECMWF ERA5 data set for meteorological data ($0.25^\circ/\text{hourly}$ resolution) (Wouters et al., 2021), World Ocean Atlas for salinity and
390 temperature profiles (0.25° resolution) (Garcia H.E. et al., 2019), and the TPOX-9 atlas for tides ($1/30^\circ$ resolution) (Egbert and Erofeeva, 2002). The setups have 1 grid cell m^{-1} , and the model is run using a forward Euler time-step ordinary differential equation that solves the state every 60 seconds. The variables are exported as daily mean values before being processed using R v4.4.1. Plots are generated using ggplot v3.5.0.; linear regression and statistics are calculated using ggpubr v0.6.0.

Horizontal boundary exchange of Hg

395 In GOTM, sediment and atmosphere are considered a horizontal surface that interacts with 1 cell of the surface grid. Therefore, we assume that burial will take place in 2 steps. The first step is sedimentation to the shallow sediment layer. From this layer, there is a fixed burial rate of $1.0\text{E-}5 \text{ d}^{-1}$ to deeper sediment that cannot be resuspended. These processes and rates are the



same for Hg, nutrients and organic carbon in the sediment. Sedimentation and resuspension of Hg are coupled with detritus in the ECOSMO E2E model. Hg bound to detritus will sink, sediment, and resuspend at the same rate as the organic matter it is associated with. Macrobenchos can also take up Hg bound to organic carbon when it consumes detritus or DOM. When macrobenchos lose Hg due to respiration or mortality, it is released into the sediment. Hg in the sediment has the same burial and resuspension rates as sediment carbon in the ECOSMO E2E model, as it is assumed to remain bound to organic carbon in the sediment.

Due to its chemical properties, Hg^0 in the surface water layer is constantly exchanged with the atmosphere. This exchange is modeled the same as in Bieser et al. (2023) and modeled after Kuss (2014), which is based on the Henry's law constant determined by Andersson et al. (2008) to estimate the equilibrium between Hg^0 in the air and surface water. This method is extensively evaluated in Bieser et al. (2023) against measurements by Kuss et al. (2018). From all Hg speciations in aquatic environments, two species are volatile - Hg^0 and DMHg. The direction of atmospheric exchange depends on both the aquatic and atmospheric concentrations of those forms of Hg. Furthermore, the atmosphere can be a source of Hg^0 and Hg^{2+} , through direct wet deposition. To simulate the interactions of Hg with the atmosphere in this study, we used the approach previously used in the MERCY v2.0 model (Bieser et al., 2023). Hg^0 is constantly exchanged between the surface layer and the atmosphere. When DMHg is present in the surface layer, it can evaporate from the marine system. In addition, there is the deposition of Hg^0 and Hg^{2+} . Both atmospheric concentration and Hg deposition are provided by the CMAQ model, and the values are the same as in the 3D MERCY v2.0 (Bullock and Brehme, 2002; Bieser et al., 2023).

Atmospheric deposition of nutrients

Horizontal advection of nutrients plays an important role in the North and Baltic Seas. Nutrients are transported into the seas by rivers and are buried in sediment or transported to the Atlantic Ocean, and these processes are different for the three set-up locations (Vermaat et al., 2008). Horizontal transport cannot be accurately captured in a 1D setup and is beyond the scope of this model. To mimic realistic wintertime nutrient concentrations of nitrate, phosphate, and silicate of 3.7, 0.3, and 5.0 μM in the Gotland Deep setup and 7.5, 0.5, and 5.0 μM in the North Sea setups, respectively (Burson et al., 2016). Both regions are characterized by strong horizontal influxes of nutrients that are not present in our 1D setups. To compensate for this limitation of our model, we chose atmospheric deposition so that it would create realistic nutrient conditions during winter in the surface ocean throughout the simulation.

Partitioning to DOM and detritus

The only marine organic carbon particles in the water that interact with Hg in the 1D setups are detritus and DOM, which originate from the ECOSMO E2E model. The partitioning between Hg and detritus and DOM is assumed to be instantaneous, and equilibrium is forced on every model time step. This is based on K_{ow} which is $\log(6.4)$ and $\log(6.6)$ for the partition of Hg^{2+} into detritus and DOM responsibility and $\log(5.9)$ and $\log(6.0)$ for the binding of MMHg^+ to detritus and DOM responsibility. These values are based on Allison et al. (2005) and Tesán Onrubia et al. (2020) and evaluated in Bieser et al. (2023). This



430 mechanism and rates are taken from the MERCY v2.0 model and are described in more detail in Bieser et al. (2023). Hg associated with organic carbon is taken up by organisms when they consume the detritus or DOM to which it is bound.

Organic carbon to dry and wet weight conversion

The biomass in our model is resolved in nutrients according to the redfield ratio. In contrast, the uptake and release estimations are based on studies relying on weighing the samples and are therefore giving in wet or dry weight. We assumed the ratio of milligram carbon to dry weight as: 0.2 for diatoms, 0.33 for flagellates and cyanobacteria and 0.5 for zooplankton and fish (Walve and Larsson, 1999; Sicko-Goad et al., 1984). The weight values in this paper are always given in dry weight unless otherwise specified. Bioaccumulation can also be stated per wet weight in the literature (Nfon et al., 2009). The dry weight to wet weight conversion factor of 0.2 is then used for phytoplankton, 0.3 for fish, and 0.16 for zooplankton and macrobenthos (Cushing, 1958; Ricciardi and Bourget, 1998; ?).

440 3 Model evaluation

Since this is the first publication to use specific 1D GOTM setups in combination with ECOSMO E2E and MERCY v2.0 coupled via FABM, and we made changes to the ECOSMO E2E model, we evaluated whether the 1D setups perform as expected and if the biological carbon and Hg species relevant for bioaccumulation are within the range of published 3D models. Data for higher trophic levels and bioaccumulation is much rarer and therefore this data is evaluated by a quantitative analysis if the modeled values are 1 standard deviation (SD) of observations. The 3D HAMSOM-ECOSMO-MERCY setup is extensively evaluated and shows good agreement with observations for Hg cycling as is discussed in more detail in Bieser et al. (2023), and therefore it was not further evaluated in this study.

3.1 1D Model and Observational Data

For ECOSMO E2E we compared the biomass of the 1D model with the biomass predictions in the published 3D model. This is done because the ECOSMO E2E model is extensively validated in Daewel et al. (2019) and we want to ensure that our 1D setup is consistent with the 3D model setup. The 3D ECOSMO E2E model reports a total primary production of 50 to 90 grams of carbon $\text{m}^{-2} \text{y}^{-1}$ in the open North Sea and between 30 and 50 grams of carbon $\text{m}^{-2} \text{y}^{-1}$ in the open Baltic Sea. In the North Sea, the phytoplankton bloom is initiated by diatoms and is then taken over and exceeded by flagellates. In the Baltic Sea, the bloom is initiated by a large diatom bloom followed by a bloom in flagellates and cyanobacteria. Secondary production ranges between 20 and 40 grams of carbon $\text{m}^{-2} \text{y}^{-1}$ in the open North Sea and 10 to 30 grams of carbon $\text{m}^{-2} \text{y}^{-1}$ in the open Baltic Sea (Daewel et al., 2019). Macrobenthos total biomass estimates range from 1.5 to 40 grams of carbon for the open North Sea, with the higher values closer to the coast (Daan and Mulder, 2001; Heip et al., 1992). The Gotland Deep has anoxic deep water, so there is no macrobenthos (Kendzierska and Janas, 2024). The open Baltic Sea phytoplankton bloom is dominated by diatoms and is followed by flagellates and cyanobacteria (Hjerne et al., 2019). During the autumn, cyanobacteria can become the dominant species with a biomass of up to 50 mg C m^{-3} , but there is a large variety in the intensity of the bloom and the



relative importance of different species (Hjerne et al., 2019).

The results of the simulations of ecosystem biomass are presented in Fig. 2. The total yearly primary production in our model is 50, 62, and 61 $\text{gC m}^{-2} \text{y}^{-1}$, the pelagic secondary production is 24, 42, and 29 $\text{gC m}^{-2} \text{y}^{-1}$, and the average fish stocks are 3.6, 2.6, and 2.2 gC m^{-2} in the Gotland Deep, Northern North Sea and Southern North Sea respectively. The peak macrobenthos population is 12.3 gC m^{-2} in the Southern North Sea, 3.4 gC m^{-2} in the Northern North Sea, and there are no macrobenthos in the Gotland Deep.

Therefore, we conclude that the primary production and biomass of the functional group in the 1D setups align well with observations and remain in line with the 3D ECOSMO-HAMSOM model Daewel et al. (2019). Additionally, the biomass falls within the range of observed ranges for all higher trophic-level functional groups. Because of this, we conclude that the general dynamic of biomass production in 1D setups provides similar results as the 3D ECOSMO and captures the key ecosystem dynamics in all setups.

Hg cycling; 1D Hg model results

Table 6 shows the observed concentration of aquatic Hg and MeHg in the Baltic Sea measured by Kuss et al. (2017), while Fig. 3 shows total and aquatic Hg and MeHg in our 1D model. As shown in Table 1, tHg and tMeHg include bioaccumulated Hg and MeHg, while aquatic Hg and MeHg do not. The aquatic Hg includes Hg^{2+} , Hg^0 , HgS , MMHg^+ , DMHg and all Hg bound to detritus and DOM, as shown in table 1. The modeled tHg in the surface water of the Gotland Deep is 22% lower than the observations but will be within 1 standard deviation of the observed mean. For MeHg our modeled concentration is 50% below the observed mean and 17% below 1 standard deviation of the observed mean. This is hard to evaluate, as the measurement protocol does not lyse phytoplankton cells; we assume that no MeHg associated with biota is measured. However, if we assume that MeHg associated with phytoplankton is measured, our modeled concentration will be above the observed mean with 0.12 ± 0.08 pmol. Because of this uncertainty, we also evaluated the MeHg content indirectly through bioaccumulation. To this extent, Table 6 shows the modeled and observed concentration in diatoms and the observed volume concentration factor (VCF). Since the concentration of MeHg in diatoms in the Gotland Deep (14.4 ± 1.05) is within 1 standard deviation (SD) of the observed mean (10.0 ± 5.0), and the modeled VCF in Gotland Deep ($1.22\text{E}5 \pm 3.8\text{E}4$) is within the observed range ($2.0\text{E}4 - 6.4\text{E}6$), and the observed concentration of MeHg (0.10 ± 0.04) is between our modeled concentration with (0.12 ± 0.08) and without (0.05 ± 0.09) MeHg in phytoplankton, we conclude that our model produces aquatic Hg and MeHg values that are in line with observations, with the caveat that the evaluation of the aquatic MeHg content is based on limited amount of data with a high measurement uncertainty. Figure 3 shows the aquatic and total Hg and MeHg values modeled and demonstrates the difference between the total and aquatic Hg fraction in both depth and time. Notable is the small difference in the total and aquatic fractions of Hg compared to the large difference for MeHg in all three setups. In surface water during the bloom period, phytoplankton take up MeHg, which reduces aquatic MeHg. The removal of MeHg from the water prevents its (photo)demethylation and the dissolved MeHg fraction is replenished by the methylation of Hg^{2+} this increases the tMeHg content, while this is not the case for tHg. An additional noteworthy observation is in the shallow well-mixed Southern North

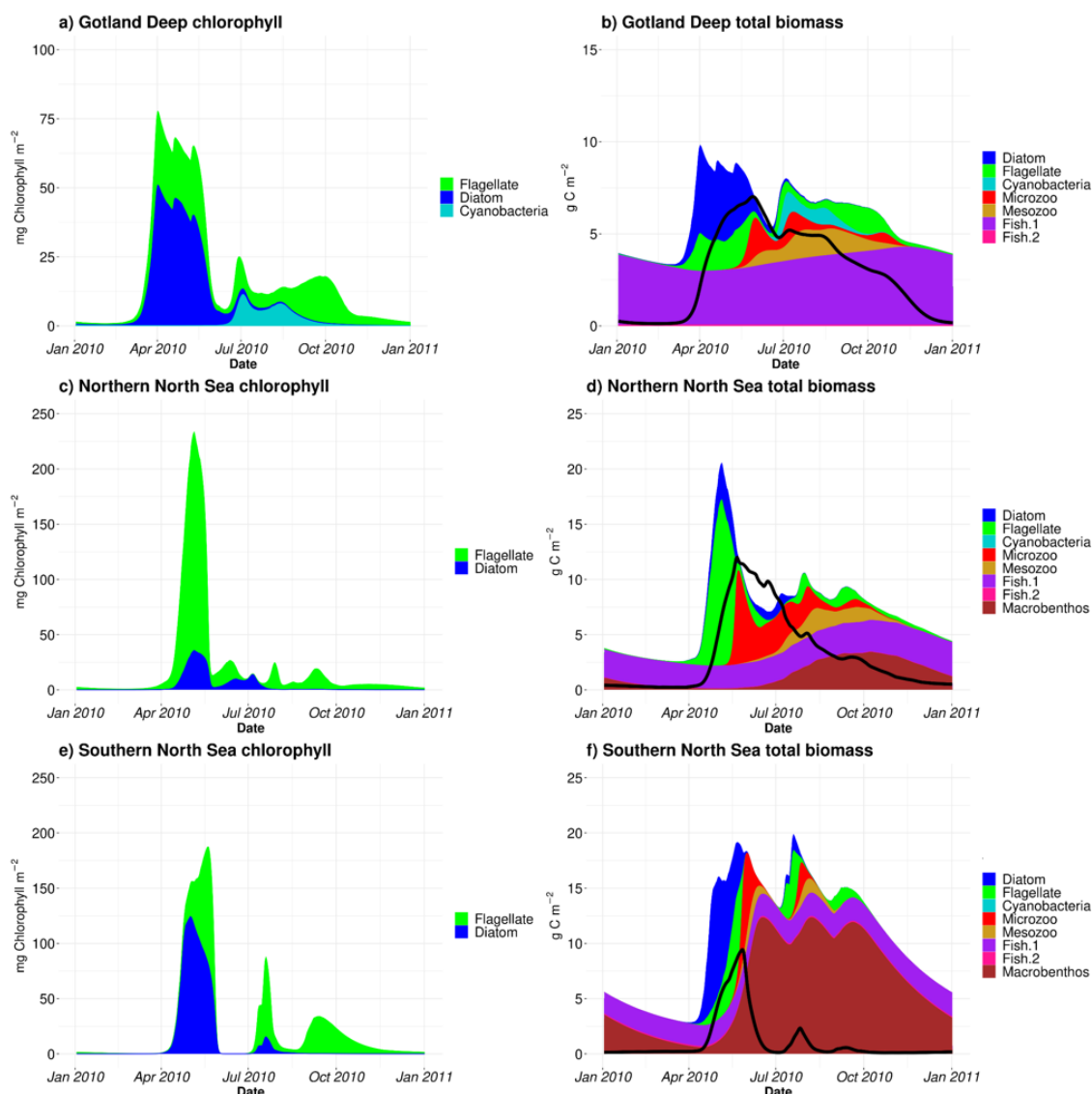


Figure 2. Modeled chlorophyll concentration (left) and organic matter concentration (right). In the last year of the simulation (Jan 2010 to Jan 2011). All living organic material is stacked, and detritus and DOM are plotted in the black line on top. Peak spring bloom chlorophyll concentration varies with location. Gotland Deep (a) has 77.8 mg m^{-2} chlorophyll with succession from diatoms to flagellates to cyanobacteria. The Northern North Sea (c) has 223.5 mg m^{-2} chlorophyll and is dominated by flagellates while the Southern North Sea (e) has 187.3 mg m^{-2} chlorophyll and is initially dominated by diatoms and later succeeded by flagellates. All locations have a succession of zooplankton after phytoplankton which microzooplankton and is taken over by mesozooplankton. Fish biomass ranges Northern North Sea (fish 1: 2.0-3.1, fish 2: 0.033-0.038 g C m^{-2}), Southern North Sea (fish 1: 1.7-2.2, fish 2: 0.10-0.12 g C m^{-2}). Macrobenthos biomass fluctuates seasonally: Northern North Sea (0.080-3.4 g C m^{-2}), Southern North Sea (0.56-12.3 g C m^{-2}), absent in Gotland Deep due to anoxic bottom water.



495 Sea. The mixing allows macrobenthos to feed directly off the spring bloom and thus removes both organic material and Hg from the water column. This leads to a drop in tMeHg during the bloom. In winter, the MMHg⁺ bioaccumulated by macrobenthos is resuspended, leading to a very high wintertime concentration of both aquatic and total MeHg in winter in the Southern North Sea.

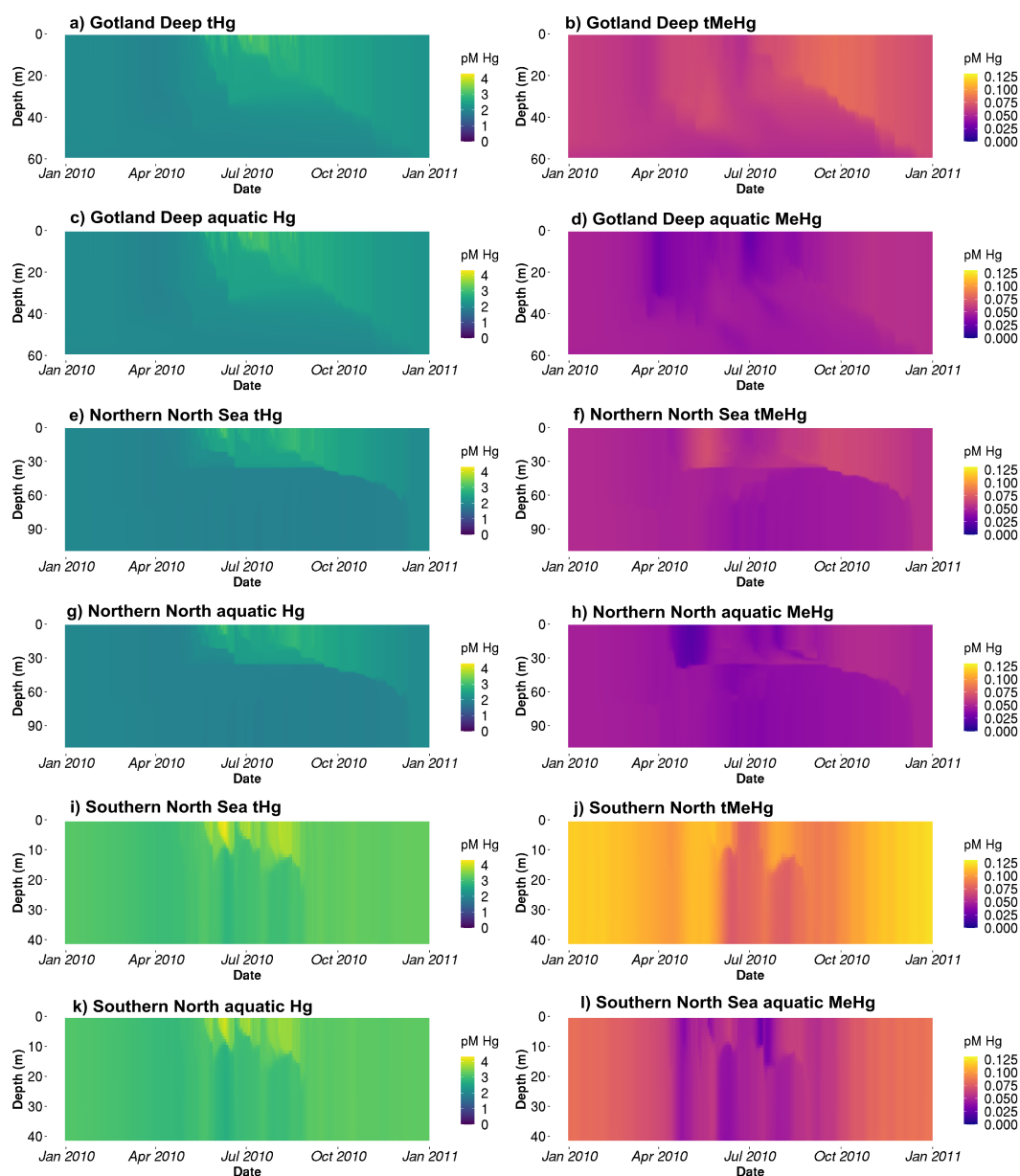


Figure 3. The tHg and tMeHg and the aquatic Hg (Hg^{2+} , Hg^0 , HgS , MMHg^+ , and DMHg) and MeHg (MMHg^+ and DMHg) concentrations in the last year of the simulation (Jan 2010-Jan2011) in all setups. The Gotland Deep and the Northern North Sea have a similar total Hg concentration of 2-3 pM, with an increase in Hg of up to 3.5 pM in the Northern North Sea. The concentration of Hg in the Southern North Sea is 2.5-3.5 pM during most of the year and reaches 4 pM during the spring bloom. The tMeHg content is 0.04-0.06 pM in the surface of the Gotland Deep and Northern North Sea and up to 0.08 pM in the Southern North Sea during winter. A clear reduction in all setups in aquatic MeHg during the bloom period can be seen, while the total concentration remains more stable.



Bioaccumulation model results

500 In Fig. 4 we show the bioaccumulation of Hg^{2+} and MMHg^+ in the Gotland Deep, in Fig. 5 of the Northern North Sea and in Fig. 6 for the Southern North Sea. We compare the modeled Hg bioaccumulation for biota in the Gotland Deep. We only compare when the total water column biomass $> 100 \text{ mg carbon m}^{-2}$ and take the modeled values of the $< 5 \text{ m}$ for phytoplankton and $< 20 \text{ m}$ for other biota to ensure that our bioaccumulation values resemble the modeled values in surface water where biota is active. A comparison between our modeled and observed bioaccumulation is shown in Table 6.

Table 6. Evaluation of the modeled bioaccumulation. Modeled and observed aquatic Hg and MeHg concentrations are shown first based on Kuss et al. (2017), then the observed range of the VCF in diatoms is shown based on Lee and Fisher (2016) and the mean and standard deviation of the modeled VCFs for MeHg into diatoms are shown. The observations for pelagic invertebrates are based on Nfon et al. (2009), the observations for fish on Kwaśniak et al. (2012) and Polak-Juszczak (2018) and the observations for macrobenthos on measurements of *Macoma Baltic* in the Baltic Sea (ICES 25 region) by Polak-Juszczak (2012). Of both model and observations the mean, standard deviation (SD) and coefficient of variation (CV) is shown. The concentration of aquatic tHg and MeHg are shown in pM, the VCF is unitless and the mean and SD of both bioaccumulated Hg^{2+} and MMHg^+ is shown in $\text{ng g}^{-1} \text{ d.w.}$ while the CV is unitless.

Variable	Observed	Model; Gotland Deep	Model; Northern North Sea	Model; Southern North Sea
Aquatic Hg	1.65 ± 2.1	2.37 ± 0.61	2.26 ± 0.22	2.97 ± 0.16
Aquatic MeHg	0.10 ± 0.04	0.05 ± 0.09	0.06 ± 0.01	0.04 ± 0.01
VCF MeHg in diatoms	$2.0\text{E}4\text{-}6.4\text{E}6$	$1.22\text{E}5 \pm 3.8\text{E}4$	$1.0\text{E}5 \pm 4.58\text{E}4$	$1.1\text{E}5 \pm 2.5\text{E}4$
Diatoms (tHg)	10.0 ± 5.0 (0.50)	14.4 ± 1.05 (0.07)	12.9 ± 2.11 (0.16)	16.1 ± 1.62 (0.10)
Microzooplankton (tHg)	37.5 ± 31.3 (0.83)	47.0 ± 12.1 (0.26)	50.3 ± 10.8 (0.22)	54.4 ± 19.7 (0.36)
Mesozooplankton (tHg)	62.5 ± 12.5 (0.20)	66.1 ± 13.5 (0.20)	68.8 ± 10.8 (0.16)	68.3 ± 17.7 (0.26)
Fish 1 (Hg^{2+})	12.7 ± 7.0 (0.55)	7.17 ± 1.72 (0.24)	7.95 ± 0.82 (0.10)	11.8 ± 1.38 (0.12)
Fish 1 (MMHg^+)	49.3 ± 30.7 (0.62)	31.8 ± 1.90 (0.06)	21.2 ± 0.92 (0.04)	44.4 ± 3.62 (0.08)
Fish 2 (Hg^{2+})	34.3 ± 23.7 (0.69)	4.20 ± 0.82 (0.20)	4.81 ± 0.65 (0.14)	10.0 ± 1.84 (0.18)
Fish 2 (MMHg^+)	180 ± 72.3 (0.40)	65.0 ± 2.17 (0.03)	37.0 ± 0.95 (0.03)	64.5 ± 3.43 (0.05)
Macrobenthos (tHg)	25.0 ± 21.4 (0.84)	–	20.0 ± 6.6 (0.22)	30.15 ± 3.2 (0.16)

505 Evaluation of modeled bioaccumulation

Table 6 shows the modeled and observed means, the standard deviation (SD), the coefficient of variation (CV) and the VCF for MeHg in diatoms. The modeled bioaccumulation values from surface water (0-5 m) for diatoms and (0-20 m) for consumers are used for the evaluation. Because there are high-quality observations for the Baltic Sea, but not for the North Sea, we focused the evaluation efforts on the Baltic Sea. The observations are for tHg in invertebrates, while we have separate observations of Hg^{2+} and MMHg^+ in fish. Fish 1 in our model is compared to the observations of herring and fish 2 and to observations in Atlantic Cod. With the exception of the bioaccumulation into fish 2, all bioaccumulation values are within one observed SD of



the observed mean. This, combined with the high CV of the observations, shows that our modeled values are well within the plausible range based on observations, except for the bioaccumulation in fish 2, which is lower than observed. It should be noted that the bioaccumulation modeled in fish 2 is not out of the observed range. Observations for tHg bioaccumulation in Atlantic
515 Cod in the Baltic Sea range between 3 and 1567 ng g⁻¹ d.w. (ICES, 2020). However, the observed mean MMHg⁺ content of all datasets is more than 1 observed SD higher than our modeled mean, leading us to believe that our model underestimates bioaccumulation in fish 2.

Modeled biomagnification

In Fig. 7 we plotted the correlation between the trophic level and the accumulated Hg²⁺ and MMHg⁺. To assess biomagnifica-
520 tion in bioaccumulated MMHg⁺, we assumed an exponential trend between the bioaccumulation of MMHg⁺ and the trophic level, which is forced through the first consumer (microzooplankton). In this way, the exponent is the biomagnification factor. Observed BMF are between 2.0 and 3.4 in the North Sea for MMHg⁺ (Baeyens et al., 2003). For Hg²⁺ an exponential relationship between trophic level and bioaccumulation was neither expected nor present; therefore, we correlated this using a second degree polynomial. The modeled biomagnification factor is between 2.1 and 3.1 for MMHg⁺. This means that our modeled
525 biomagnification factor for MMHg⁺ is within the range of observations in the North Sea.

Summary of the model evaluation

Based on the evaluation of the bioaccumulation, we conclude that the general trend of high bioaccumulation of MMHg⁺ and low Hg⁺ is well reproduced in our model. However, the performance of the model is lower in higher trophic levels and underestimates bioaccumulation into Atlantic Cod. The underestimation in modeled MMHg⁺ values in cod can be attributed to
530 the ecosystem model that underestimates the trophic level of cod by 0.5-0.7. If we extrapolate the equation for biomagnification in the Gotland Deep shown in Fig. 7 ($15.42e^{TL-2}$), in which TL is the trophic level, and estimate bioaccumulation at the observed trophic level Atlantic Cod (4.2), we would expect a MMHg⁺ content of 139 ng Hg g⁻¹ d.w., which would be well within 1 observed SD of the observed mean. Because our model accurately predicts the tHg content of plankton, the content of Hg²⁺ and MMHg⁺ in fish 1, and the MMHg⁺ content at high trophic levels according to their trophic level, it appears that the
535 bioaccumulation model accurately models MMHg⁺ bioaccumulation based on trophic interactions, but that our fish 2 better resembles a mid level trophic predator both in trophic position and MMHg⁺ than a high trophic level predator such as Atlantic Cod.

4 Results & Discussion

4.1 The effect of bioaccumulation on tHg and tMeHg concentrations

540 The 10-year average difference in the daily mean tHg and tMeHg between the base case and the setups without bioaccumulation and biogenic reduction in the Gotland Deep is shown in Fig. 8. Additionally, Table 7 shows the mean difference of tHg

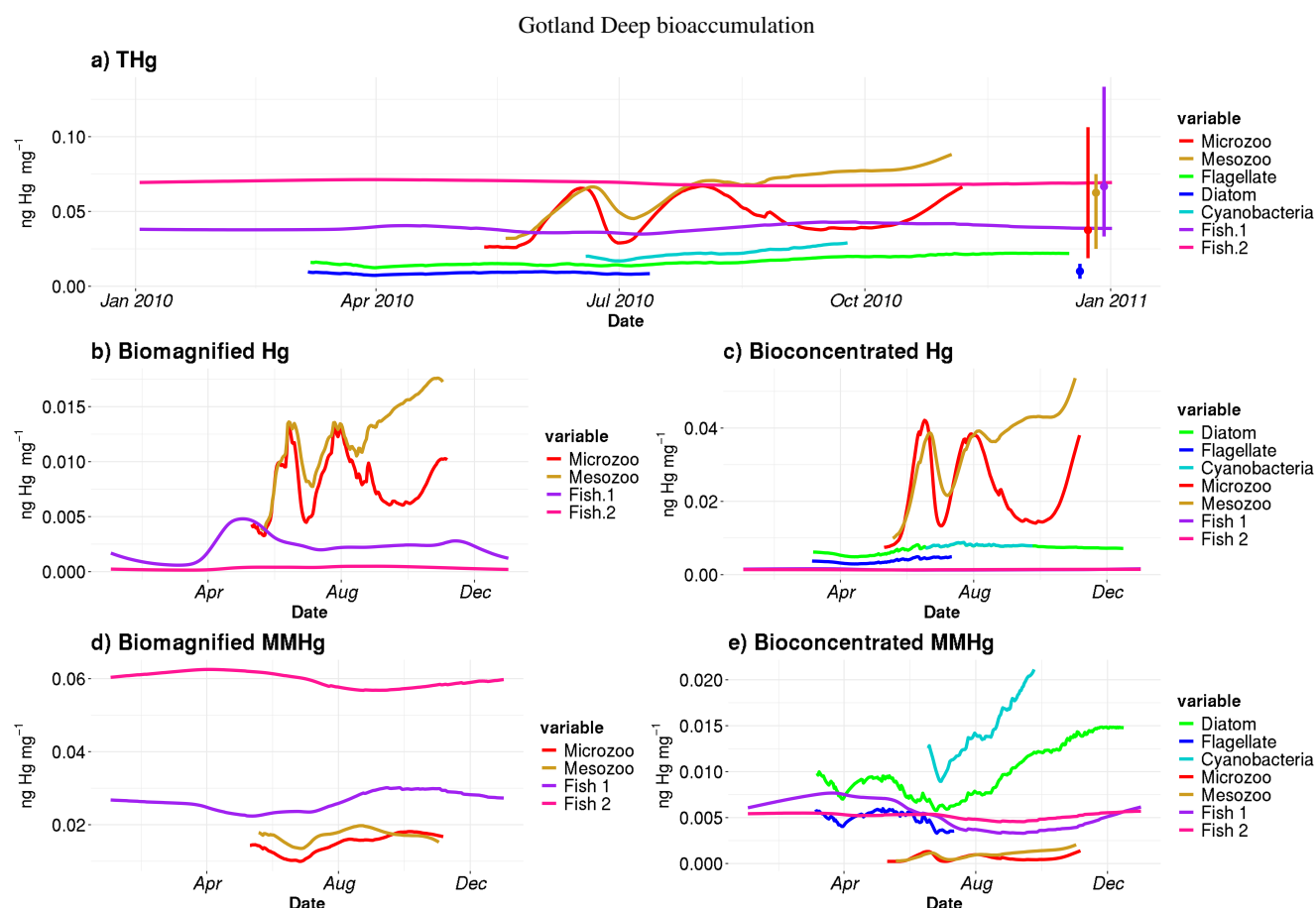


Figure 4. Hg accumulation in the Gotland Deep during the last simulation year (Jan 2010-Jan 2011). Plot 4a shows tHg bioaccumulation with mean and range of observations from Nfon et al. (2009) represented by the point and vertical bar on the right side of the plot. Bioaccumulation is displayed when biomass of the respective functional group exceeds 0.1 gC m^{-2} . tHg bioaccumulation is highest in fish 2, followed by fish 1, microzooplankton, mesozooplankton, cyanobacteria, flagellates, and diatoms with tHg concentrations in observed ranges. The consecutive Fig. show the bioamplification (4b, 4d) and bioconcentration (4c, 4e) of Hg^{2+} (4b, 4c) and MMHg^{+} (4d, 4e). Biomagnified Hg^{2+} is highest in microzooplankton, followed by mesozooplankton and fish, while biomagnified MMHg^{+} increases notably in fish 1 and fish 2. Bioconcentrated Hg^{2+} is very low in fish and highest in zooplankton. Bioconcentrated MMHg^{+} is notably higher in cyanobacteria than in all biota and lowest zooplankton.

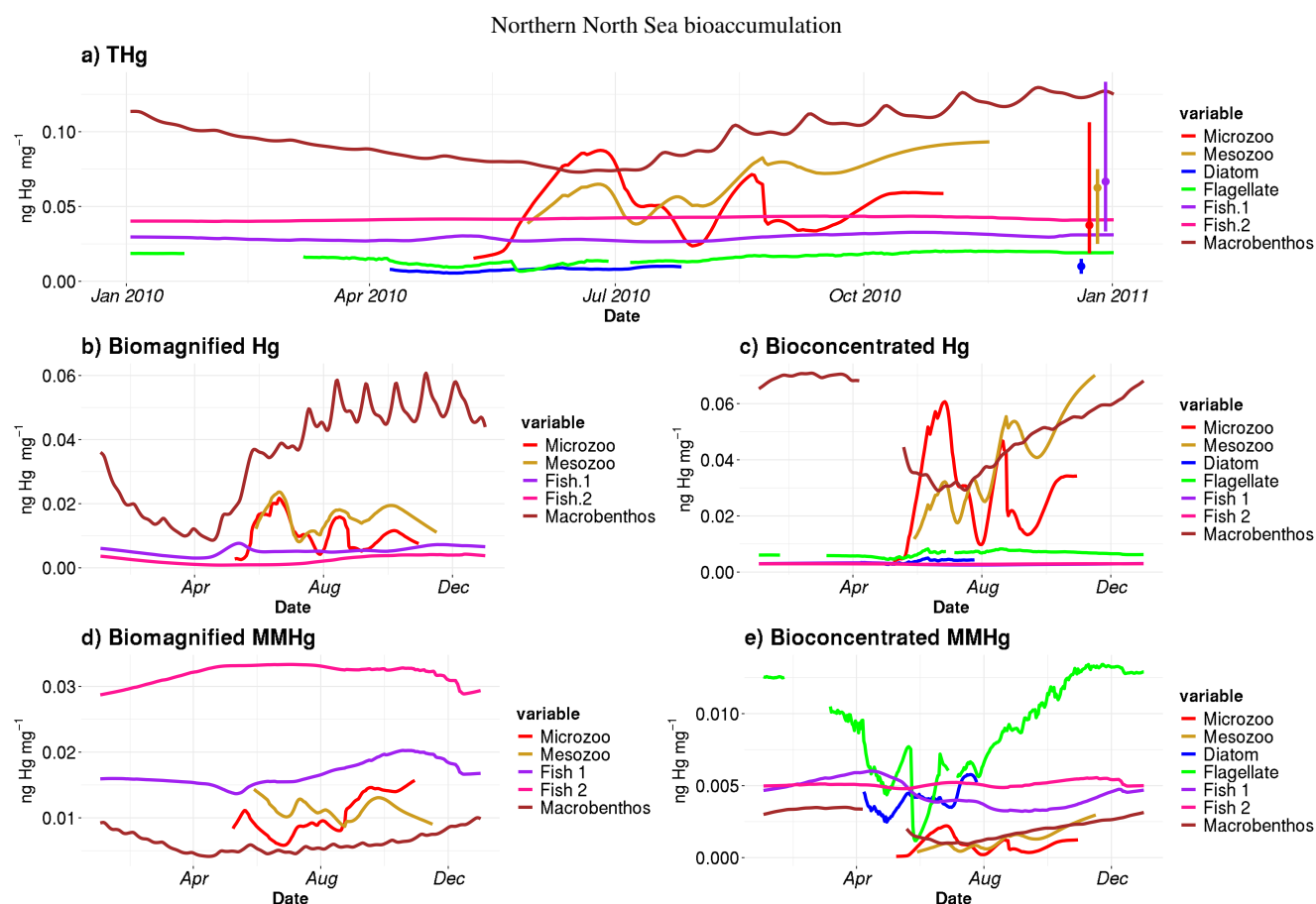


Figure 5. Hg bioaccumulation in the Northern North Sea. Plot 5a shows the tHg bioaccumulation in the Northern North Sea. Phytoplankton and zooplankton are shown if their average biomass is more than 0.1 gC m^{-2} . Phytoplankton has the lowest tHg, which is followed by fish 1, fish 2, mesozooplankton and macrobenthos. Plots 5 b-e show the origin (Biomagnification or Bioconcentration) and species (Hg^{2+} or MMHg^+) of the bioaccumulated tHg. Figure 5b and 5c show that the high tHg in microzooplankton, mesozooplankton and macrobenthos is due to high levels of Hg^{2+} bioconcentration and biomagnification. MMHg^+ Biomagnification follows a pattern in which it is lower in zooplankton and macrobenthos, higher in fish 1, and highest in fish 2. This means that while fish 2 has a lower Hg content than macrobenthos and zooplankton for part of the year, MMHg^+ is higher in fish than in zooplankton and macrobenthos. Figure 5e shows the bioconcentration of MMHg^+ and shows that this is highest in phytoplankton, followed by fish 1 and 2, macrobenthos, and zooplankton.

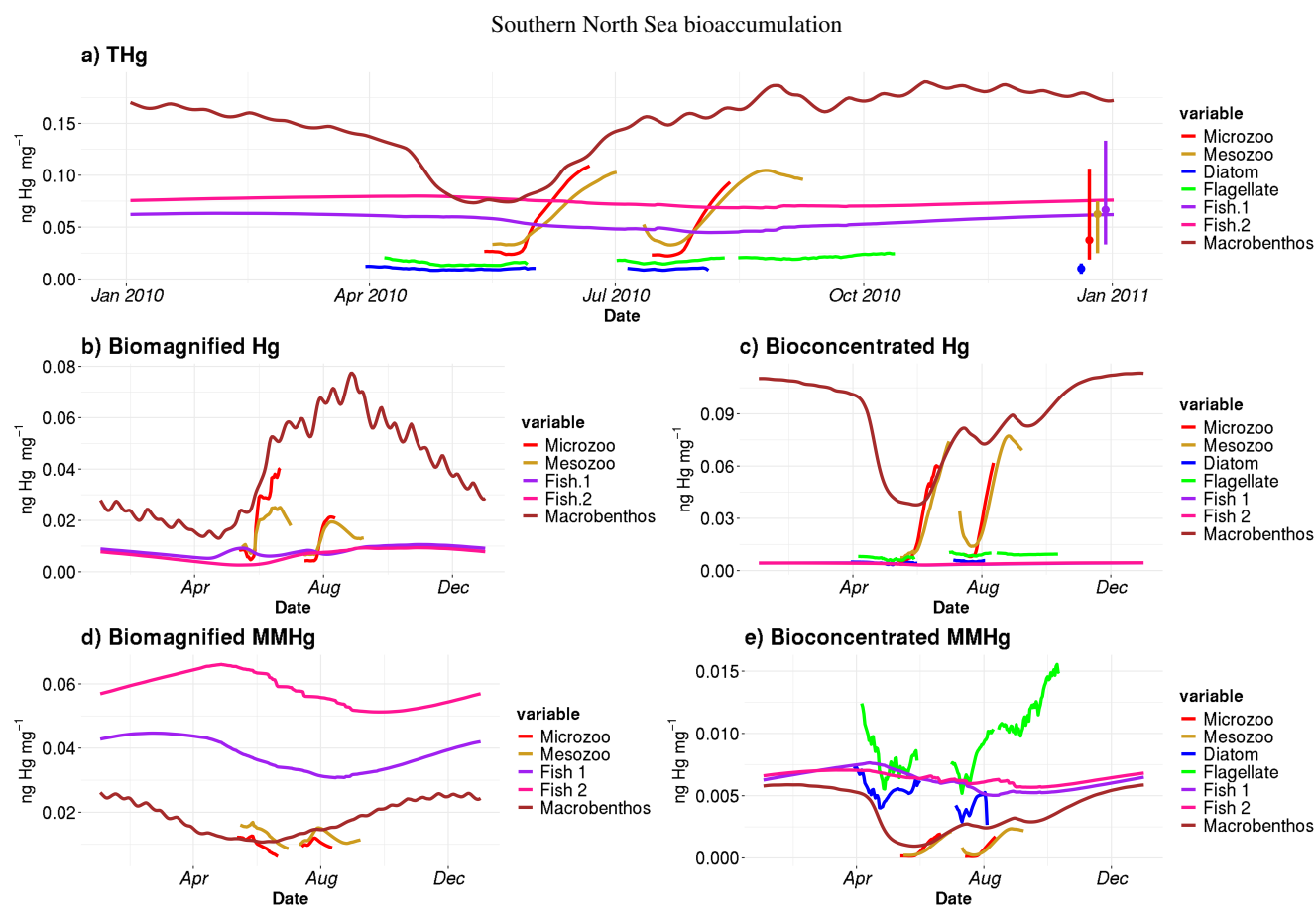


Figure 6. Bioaccumulation of Hg^{2+} and MMHg^+ in the Southern North Sea in the last year of the simulation (Jan 2010-Jan 2011). Figure 6a shows the accumulation of tHg per functional group over an annual cycle. Figures 6b-e display Hg^{2+} and MMHg^+ concentrations, while distinguishing between bioconcentration and biomagnification processes for each species. Phytoplankton and zooplankton are shown if their average biomass is more than 0.1 gC m^{-2} . tHg values for microzooplankton, mesozooplankton, and macrobenthos reach 0.10 and $0.10 \text{ ng Hg mg}^{-1}$. This is higher than fish 1 and fish 2 tHg which is 0.068 and $0.080 \text{ ng Hg mg}^{-1}$. Comparing Fig. 6b-e we see high tHg in microzooplankton, mesozooplankton, and macrobenthos is caused by high Hg^{2+} bioconcentration and biomagnification, while tHg content of fish 1 and fish 2 predominately originates from biomagnification of MMHg^+ .

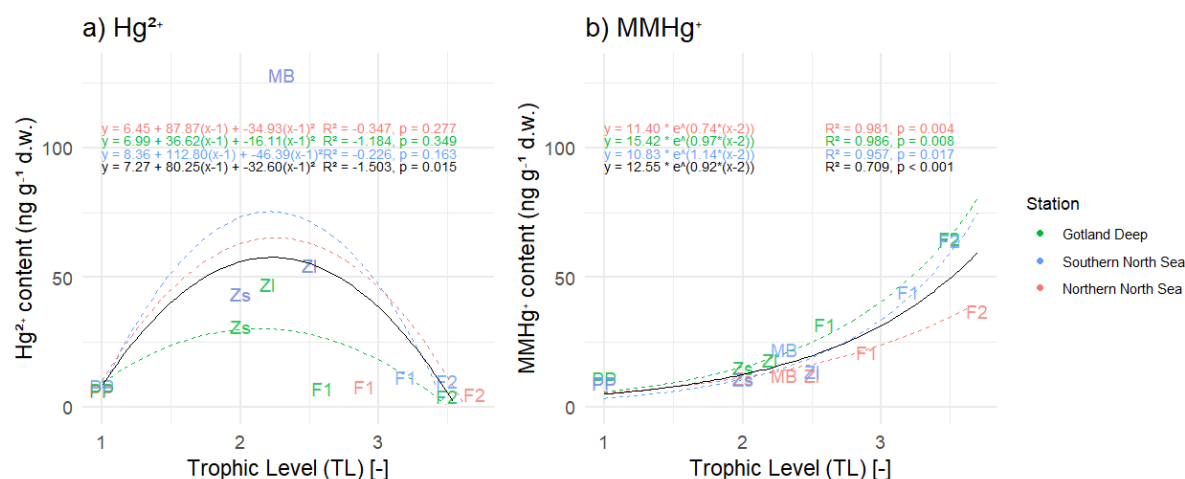


Figure 7. Figure 7 a) shows the correlation between trophic level and Hg^{2+} and Fig. 7 b) shows the correlation between trophic level and bioaccumulation of MMHg^+ in all three setups, the black indicates the average. PP stands for primary producers, Zs for microzooplankton, Zi for mesozooplankton, F1 for fish 1, F2 for fish 2, and MB for macrobenthos. The correlation is fitted that is forced through microzooplankton. The R^2 of the correlation between trophic level and Hg^{2+} shows that there is a weak anti-correlation in the Gotland Deep ($R^2=0.54$) and no correlation ($R^2<0.01$) in the Northern North Sea and Southern North Sea. In all setups, there is a strong correlation between trophic level and MMHg^+ bioaccumulation ($R^2>0.8$). Interestingly, the correlation between bioaccumulation MMHg^+ and trophic level for all setups averages is considerably lower ($R^2=0.51$ for MMHg^+) than for the setups separate.

and tMeHg, the mean and range of the bioaccumulated Hg fraction in the bioaccumulation setup, and the average annual amplitude of the difference in tHg and tMeHg, calculated as half the total range of values. This is shown because, while the average differences provide information on the net effect of bioaccumulation, the amplitude highlights seasonal variations. If the ecosystem causes a large average difference, but this difference is consistent throughout the year, the amplitude will be 0%, but if the ecosystem causes an increase of a 100% and consequently a decrease of a 100% with no average effect, the amplitude will be 100%.

The effect of bioaccumulation on tHg

Bioaccumulation has only a small effect on tHg with a maximum increase of 3% while the average percentage of tHg that is bioaccumulated is 1% in every setup. This increase is caused because the biota takes up Hg, which is consequently protected against reduction to Hg^0 . Since Hg^0 in the water is in exchange with atmospheric Hg^0 , a reduction in Hg^0 will result in a reduction in evaporation until the concentration of Hg^0 in the water is re-equilibrated with the atmospheric concentration. At this point, the atmospheric exchange and the aquatic Hg concentration in the scenarios with and without bioaccumulation are similar, while the tHg concentration is up to 3% higher and in the scenario with bioaccumulation. This is further shown in Fig. 8. which shows the seasonal difference for both aquatic and total Hg and MeHg. Figure 8 shows that even though the



difference is small, there is always more tHg in the base case than in scenario B without bioaccumulation. The only exception is in the permanently mixed shallow Southern North Sea in summer and autumn, which has a large amplitude in the difference in tHg with an increase of 11% at the beginning of the bloom period and a decrease of 6% later in the bloom. This reduction in tHg is caused by macrobenthos, which feeds directly of the plankton bloom and transports Hg from the water column to the benthic via the consumption of organic material. This is corroborated by Fig. 2 which shows that macrobenthos in the Southern North Sea have high biomass (up to 12.3 gC m^{-2}) and by Table 6 which shows a high tHg content in this macrobenthos of $30.1.5 \pm 3.2 \text{ ng g}^{-1} \text{ d.w.}$ During late autumn, winter, and early spring, this tHg is released back into the water resulting in the above-mentioned high increase in tHg of 11% at the onset of the phytoplankton bloom. This causes bioaccumulation to cause a similar mean increase of 3% in the Southern North Sea compared to the other setups, while the amplitude of the difference is 9% compared to 1% in other setups, demonstrating a strong seasonal effect.

The effect of bioaccumulation on tMeHg

tMeHg increases by 41, 23, and 46% for the Gotland Deep, Northern North Sea, and Southern North Sea respectively. This is very similar to the percentage of tMeHg that is bioaccumulated, which is 41, 29 and 43% respectively. Bioaccumulation increases tMeHg, as bioaccumulated MMHg^+ cannot be photodegraded, and aquatic MeHg is replenished by additional methylation. During autumn and winter, the biomass is reduced and bioaccumulated MMHg^+ is released back into the water column. During this period, the light intensity and therefore photodegradation are lower, and detritus and DOM concentrations are higher, which leads to additional shading. This causes the aquatic tMeHg concentration to be higher outside of the bloom period. This means that an increase in bioaccumulated MMHg^+ equals a smaller decrease in aquatic MeHg causing an increase in tMeHg. Additionally, bioaccumulation can increase aquatic MeHg by removing dissolved MeHg when photodegradation is high and re-releasing this MeHg when photodegradation is low.

This increase in tMeHg is driven by organic material and shows a linear relationship with biomass. The correlation of the daily 10-year average values for the increase in tMeHg and tHg with biomass in the surface (0-20 m) is shown in Table 8. If we combine the three setups, we find an increase in tMeHg of $0.028 \text{ ng Hg mgC}^{-1}$ ($r=0.47, p<0.001$) or a relative increase of 1% per 4.5 mgC m^{-3} ($r=0.59, p<0.001$) and an increase in tHg of $0.068 \text{ ng Hg mgC}^{-1}$ ($r=0.07, p<0.001$) or a relative increase of 1% tHg per 71 mgC m^{-3} ($r=0.07, p<0.001$). Although both correlations are significant at the 99% confidence level, the relationship between the difference in tMeHg and biomass is notably stronger ($r=0.47$) than for the difference in tHg ($r=0.07$). Although bioaccumulation is a large amount of tMeHg, it does not constitute a large amount ($< 5\%$) of tHg; therefore, the formation of MeHg due to methylation is not reduced and the bioaccumulated MeHg is replenished due to its rapid equilibrium with other Hg species.

Macrobenthos in the Southern North Sea have a similar, although stronger, effect on tMeHg concentrations to that of tHg. However, this effect is partially altered by the strong increase in tMeHg as a result of the high biomass in this setup. In spring in the Southern North Sea, there is a 77% increase in tMeHg due to bioaccumulation, which is the highest of all setups. However, this rapidly drops as the bloom progresses, and macrobenthos feed on the plankton bloom, causing the increase in tMeHg due to bioaccumulation to briefly be the lowest in the Southern North Sea at the end of the bloom. During winter, the macrobenthos



Table 7. The percentage difference in tHg, and tMeHg average concentrations caused by the ecosystem, and the average seasonal amplitude in surface water (0-20m). Additionally the mean and range of the percentage of tHg and tMeHg that is bioaccumulated (Hg(Bio) and MeHg(Bio)) in the scenario with bioaccumulation is shown. The percentage is calculated as the (scenario - base case)/((base case + scenario)/2)*100. Negative values indicate a reduction caused by the ecosystem effect, and positive values indicate an increase.

Setup	tHg	tHg _{Amp}	tHg(Bio)	tMeHg	tMeHg _{Amp}	MeHg(Bio)
Gotland Deep (no bioaccumulation)	3%	1%	1(0-2)%	41%	16%	41 (26-45)%
Northern North Sea (no bioaccumulation)	1%	1%	1(0-5)%	23%	24%	29 (13-59)%
Southern North Sea (no bioaccumulation)	3%	9%	1(0-4)%	46%	28%	43 (31-64)%
Gotland Deep (no biogenic reduction)	-8%	5%	1(0-2)%	29%	16%	41 (26-45)%

590 population decreases and tMeHg is re-released, causing a large increase in wintertime tMeHg making the increase in tMeHg due to bioaccumulation in the Southern North Sea the largest in all setups.

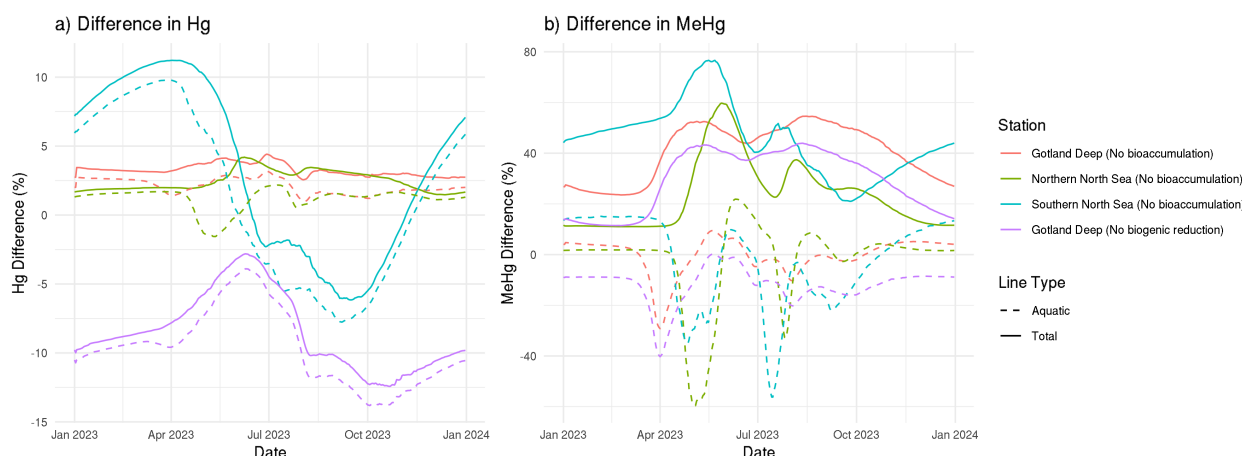


Figure 8. The difference in tHg, aquatic Hg, tMeHg, and aquatic MeHg the base case and the scenario without bioaccumulation and the scenario without biogenic reduction in the Gotland Deep in the top surface 20 meters of the watercolumn. For tHg there is a strong seasonal effect where it is decreased during summer and autumn and increased during spring. This is caused by the consumption of Hg-containing organic material during spring and summer and the release during winter. There is no notable effect ($> 5\%$) in the tHg in the Northern North Sea and Gotland Deep due to bioaccumulation. In the Gotland Deep, however, biogenic reduction causes a strong decrease of up to 12%. There is no large difference ($> 5\%$) between the difference in aquatic Hg and tHg). The difference in tMeHg is much more pronounced. There is a consistent increase in tHg which peaks during spring and summer. These peaks in an increase in tMeHg coincide with a smaller decrease in aquatic MeHg.



Table 8. Correlation, slope, and p-value for the difference in tMeHg and tHg vs the total biomass per setup. There is a much stronger correlation ($r=0.547-0.809$) between the increase in tMeHg then the increase in tHg ($r=0.114-0.149$). The slope for tMeHg ranges from 0.134 and 0.25 and for tHg between -0.0631 and $0.0279 \text{ mgC}^{-1} \text{ m}^3$

Setup	tMeHg			tHg		
	Correlation (r)	Slope	p-value	Correlation	Slope	p-value
Gotland Deep	0.809	0.252	<0.001	0.149	-0.0631	<0.001
Northern North Sea	0.739	0.134	<0.001	0.218	0.00691	<0.001
Southern North Sea	0.547	0.226	<0.001	0.114	0.0279	<0.001

The effect of bioaccumulation on the Hg budget and horizontal transport

In Fig. 9, the difference in the content of tHg and tMeHg and the export by sedimentation and evaporation from the North and Baltic Seas caused by bioaccumulation is evaluated using the 3D the MERCY-HAMSOM-ECOSMO setup. There is a decrease in the evaporation of 9% (Fig. 9b) and a decrease in the burial of 19% (Fig. 9i) caused by bioaccumulation. Figures 9e and Fig. 9f show tMeHg and the difference caused by bioaccumulation. While there is an average increase caused by the bioaccumulation of 10% in tMeHg, this is not uniform. There is a decrease in the Baltic Sea, an increase in the North Sea, and a very strong increase in the Danish Straits. This, combined with the reduction of both evaporation and burial, indicates that bioaccumulation facilitates a flux of Hg out of the Baltic Sea toward the North Sea and the Atlantic Ocean. We see an average reduction of 2.3 and $57 \text{ pmol m}^{-2} \text{ y}^{-1}$ in evaporation and burial, respectively. Our model shows that the bioaccumulation into plankton keeps Hg pelagic, which facilitates transport to the North Sea and consequently the Atlantic Ocean. Without bioaccumulation, a fraction of this Hg will evaporate or be bound to POC and be buried. The reduction in tHg due to bioaccumulation is strongest in areas such as the Wadden Sea and the Bay of Bothnia, where there are both large riverine inputs of Hg and high primary production. However, this local effect is relatively small and we see a total increase in export of 14 kg Hg y^{-1} to the Atlantic Ocean due to bioaccumulation. Of this 14 kg , 13 kg would be buried in sediment, and 1 kg would be evaporated into the atmosphere without bioaccumulation. To put this into perspective, this is 1% of the total modeled river influx into the North Sea, and thus does not significantly alter the the long range transport of Hg.

The effect of the ecosystem on the Hg budget under idealized circumstances

The 1D setup budget is shown in Fig. 10. This shows the modeled mean tHg and tMeHg concentrations, the difference in the run without an ecosystem compared to the setup with an ecosystem and the differences caused by biogenic reduction in the Gotland Deep.

The Southern North Sea setup is permanently mixed. Because of this, Hg can continuously reach the surface area where it can evaporate, and all aquatic MeHg is subject to photodegradation. At the same time, this constant mixing brings the plankton biomass and its bioaccumulated Hg to the benthic boundary layer, where it can be consumed by macrobenthos. This results in



615 a large flux of Hg^{2+} and MMHg^+ from the pelagic to the benthic during the phytoplankton bloom, which is re-released during winter. This leads to high benthic Hg (11.1 pmol m^{-2}) and MeHg (1.6 pmol m^{-2}), but a relatively low burial rate ($1.1 \text{ pmol tHg m}^{-2} \text{ y}^{-1}$) as the Hg is bioaccumulated in macrobenthos rather than associated with organic carbon of sediment. The increase in tMeHg in the Southern North Sea is highest in the 3 settings, due to the high biomass and the ecosystem increases tMeHg by 44%.

620 **The Northern North Sea** setup is only seasonally stratified. This means that during summer, macrobenthos cannot feed directly from the plankton bloom, but Hg-containing organic material can sink below the mixed layer, where it can settle as sediment, be consumed by macrobenthos, or remain until it is remixed to the surface during winter. This results in a low benthopelagic flux of Hg^{2+} and MMHg^+ throughout the year. However, the same average Hg burial as in the Southern North Sea of $1.1 \text{ pmol tHg m}^{-2} \text{ y}^{-1}$. Due to the lower biomass, there is less tMeHg in this setup than in the Southern North Sea, and
625 the effect of the ecosystem causes a smaller increase in tMeHg (13%) than in the Southern North Sea.

The Gotland Deep setup has permanent stratification. Hg can accumulate in the plankton in the surface layer or divide to detritus. When this detritus sinks, it will transport Hg to deeper water. As the Gotland Deep setup is permanently stratified, Hg eventually reaches the sediment and becomes buried. The flux of pelagic Hg^{2+} and MMHg^+ into the sediment is low, but since there is very low resuspension the burial rate is high ($11.7 \text{ pmol tHg m}^{-2} \text{ y}^{-1}$) compared to the the lower burial rate ($1.1 \text{ pmol tHg m}^{-2} \text{ y}^{-1}$) for both North Sea setups. In the Gotland Deep setup, there is a big distinction between the effect of the ecosystem above and below the oxycline. Above the oxycline the ecosystem increases tMeHg by 13%, while below the oxycline this increase is replaced by a small decrease of 3%, as the binding of Hg^{2+} and MMHg^+ to the sinking detritus facilitates the flux of
630 Hg to the sediment.

These differences led to the highest burial of tHg in the Gotland Deep, followed by the Northern North Sea, and the lowest
635 in the Southern North Sea. This shows how ecosystem-induced Hg burial are influenced by local hydrodynamics.

4.2 The cyanobacterial reduction of Hg^{2+}

The model estimates that biogenic reduction can reduce average water column Hg above the mixed layer depth by 7% in the Gotland Deep, due to the transfer of soluble Hg^{2+} to volatile Hg^0 , which increases the evaporation of Hg out of the water to the atmosphere. In Fig. 10 we can see that there is still an increase in tMeHg (13%) above the oxycline without bioaccumulation
640 and biogenic reduction, no increase ($<0.5\%$) below the oxycline. This means that biogenic reduction reduces tMeHg, but the increase caused by bioaccumulation is higher above the mixed layer depth. Figure 11 shows the seasonal difference in the tHg and tMeHg in the 3D MERCY v2.0 model between runs with and without biogenic reduction and the cyanobacterial biomass in the surface layer. Note that we isolated the effect of biogenic reduction in these setups so bioaccumulation does still occur. The MERCY v2.0 model shows a similar reduction in the Gotland Deep as the 1D GOTM setups, but it shows a
645 higher reduction in tHg in coastal areas of up to 20%. Additionally, it shows that while cyanobacteria are most abundant during autumn (Fig. 11 g), they decrease the tHg content throughout the year (11 b, e, h, k), there is an average reduction of -16%, which is highest during the autumn bloom (-17%) and lowest in summer before the bloom (-15%). Finally, cyanobacteria and biogenic reduction only occur in the Baltic Sea in the model, but we can see a decrease in tHg in the Southern North Sea of up

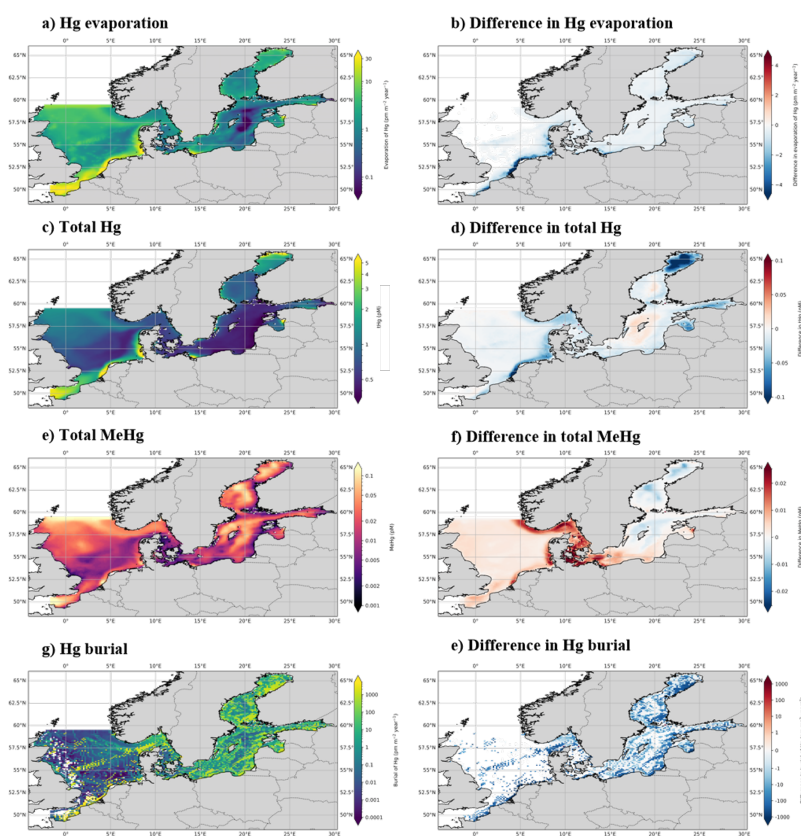


Figure 9. The yearly mean Hg budget of the North and Baltic Seas and the difference caused by the ecosystem, the concentrations and differences of tHg and tMeHg are the mean of the watercolumn. The difference is calculated as the Base case - Scenario A "No bioaccumulation". The average tHg is 0.5 pM, which is reduced by 0.02 pM without bioaccumulation. The average tMeHg is 0.03 pM, which is reduced by 0.003 pM without bioaccumulation. The decrease in dissolved tHg and tMeHg without bioaccumulation is coupled with a decrease in evaporation and burial. Evaporation decreases from 4.3 $\text{pm m}^{-2} \text{y}^{-1}$ to 3.9 $\text{pm m}^{-2} \text{y}^{-1}$ and burial from 323.5 $\text{pm m}^{-2} \text{y}^{-1}$ to 263.2 $\text{pm m}^{-2} \text{y}^{-1}$.



to -7.5% in summer and autumn and -5% in winter and spring. The difference in tMeHg is, however less pronounced with a
650 reduction of up to -3% in tMeHg during autumn and winter with a smaller reduction of -1% during spring and summer in the
Gotland Deep. There appears to be no noticeable reduction in tMeHg in the Southern North Sea (<-1%). This demonstrates
that cyanobacteria can have a very large impact on the tHg budget of the Baltic Sea, even in areas where they are less abundant
and this effect is relevant throughout the year. Kuss et al. (2017) finds that cyanobacterial-induced biogenic reduction causes
approximately 30% of all Hg evaporation during summer, since we have cyanobacteria in our model for 3 months, a year-
655 round average reduction of -9% of tHg above the mixed layer depth and a total reduction of up to -20% during summer and
autumn is in line with these observations. Since Hg evaporation equilibrates the ocean with the atmosphere the annual average
flux is not changed dramatically (-0.3%, or -0.42 nmol m⁻² y⁻¹) as seen in the Fig. 10. Rather the aquatic Hg concentration
that leads to this evaporation is lower, due to a higher Hg⁰/Hg²⁺ ratio. In addition to the reduction of aquatic Hg, it also
changes the seasonality of the evaporation of Hg⁰. In Fig. 12 we show the 10-year average daily evaporation and 30-day
660 running average of the relative difference between the 10-year average of the base case and scenario C (no bioaccumulation
nor biogenic reduction). The cyanobacteria cause an increase in evaporation of Hg⁰ during late summer and autumn facilitated
by the cyanobacterial bloom, which is compensated by a small decrease in evaporation when the bloom is over, resulting in a
similar yearly average evaporation of Hg⁰. The cyanobacteria-facilitated reduction in tHg in our model does not mean that an
abundance of cyanobacteria would automatically lead to a reduced Hg concentration in other regions. Cyanobacteria are a very
665 diverse group, and not all cyanobacteria will reduce Hg²⁺ to Hg⁰ (Kuss et al., 2015).

4.3 Partitioning to detritus and DOM

In Fig. 13, we show that the partition of Hg²⁺ and MMHg⁺ into detritus and DOM causes an increase in tHg and tMeHg above
the mixed layer depth in every setup. Since both North Sea setups are mixed during winter, this leads to a small increase in
tHg (1-2%) in both North Sea setups. In the Gotland Deep, we observe the same increase in tHg caused by the partitioning
670 of Hg to detritus and DOM above the mixed layer depth. However, simultaneously, we notice that partitioning to detritus and
DOM decreases tHg below the mixed layer depth. This occurs because the detritus and the DOM can remove Hg from the
water column since the Hg bound to the detritus can settle as sediment. Since atmospheric Hg wet deposition and atmospheric
concentrations are consistent across all setups, evaporation and burial will equilibrate to the atmospheric influx. If the fraction
of Hg⁰ of tHg is lower, tHg will increase until Hg⁰ is high enough that the evaporation of Hg⁰ reaches equilibrium with
675 atmospheric inputs. Below the mixed layer in Gotland Deep, partitioning to detritus and DOM reduces tHg. This indicates that
below the mixed layer depth, the effect of the detritus and DOM partitioning on the increase in burial is stronger than its effect
on the decrease in evaporation.

4.4 The unique properties of Hg as drivers of these results

These results are partially driven by the unique properties of Hg as a pollutant. Because Hg as an element is stable but can
680 be present in both methylated and nonmethylated forms, it behaves radically differently than a pollutant that would have a
constant total concentration. We find that this is the case for two reasons. First, the concentration of Hg⁰ in the surface layer is

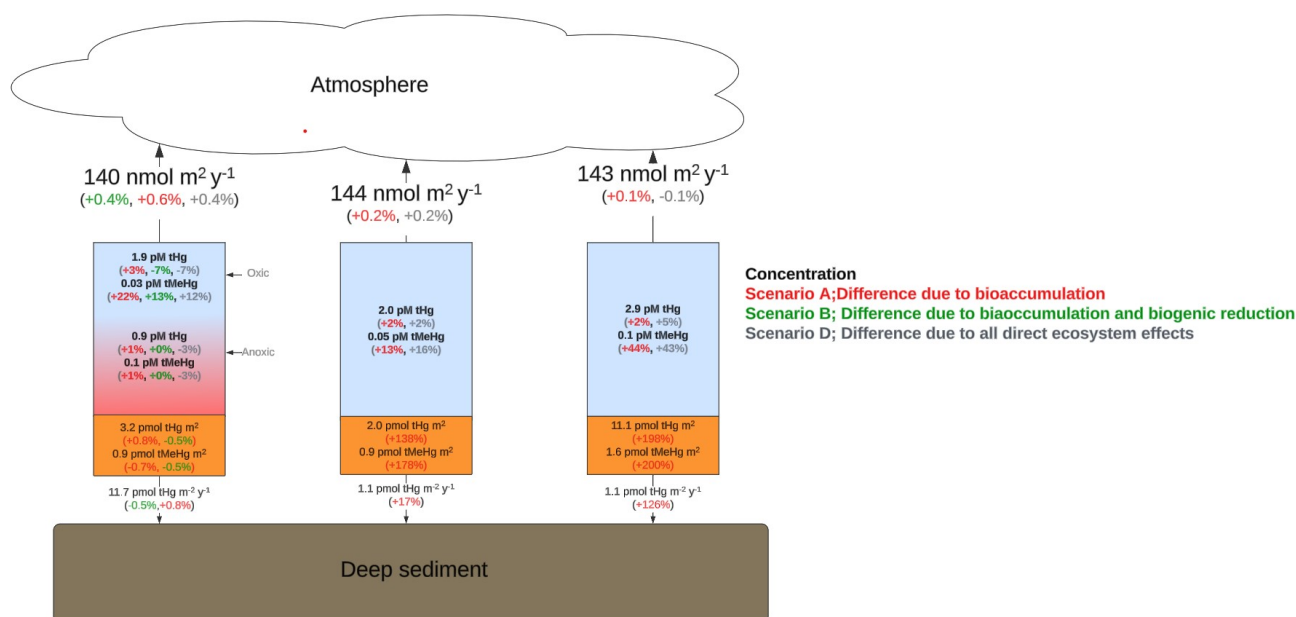


Figure 10. Hg Budget of the 3 Setups. All values are the average values for the final 10 years of the simulation. The blue area indicates oxic water, the red indicates anoxic water, and the brown area indicates sediment. The arrows indicate the direction of the flux. The black values are for the full model that includes all ecosystem effects, and the red indicates the percentage difference in the run without any ecosystem effects. The percentage differences are calculated by $(\text{Base case} - \text{scenario}) / ((\text{Base case} + \text{scenario}) / 2)$. Thus negative numbers correspond to a decrease caused by the ecosystem while positive numbers indicate an increase caused by the ecosystem. In the Gotland Deep, dark green percentages show differences between the full model and setups without biogenic reduction and bioaccumulation. There is net evaporation in all setups of $140\text{--}143 \text{ nmol m}^{-2} \text{ y}^{-1}$. In the North Sea, there is a decrease of both tHg and tMeHg when the ecosystem is not included. This decrease is larger for tMeHg (10–39%) than for tHg (9–2.2%). There is a decrease of 9% of tHg in the oxic layer of the Gotland Deep caused by cyanobacterial biogenic reduction. There is a lower increase in tMeHg (10%) in the oxic water column of the Gotland Deep compared to the northern (+12%) and southern (+39%) North Sea. The effect of biogenic reduction only causes a 0.6–0.5% decrease for tHg and tMeHg in anoxic water respectively in the Gotland Deep and a 0.92% decrease in sediment and burial. Sediment concentrations of both tHg and tMeHg are lowest in the Northern North Sea, followed by the Southern North Sea, and highest in the Gotland Deep.

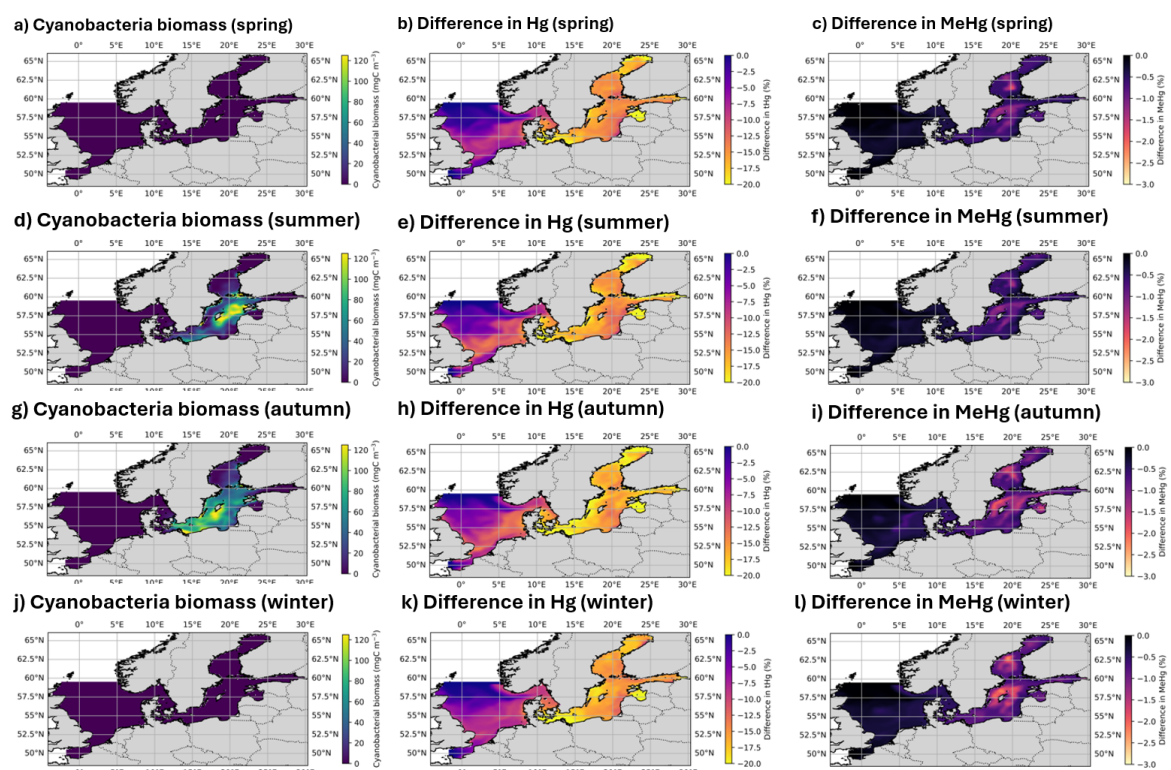


Figure 11. Cyanobacterial biomass and seasonal differences in tHg and tMeHg between the base case and the scenario without biogenic reduction.

in constant exchange with the atmosphere. If the fraction of Hg present as Hg^0 is reduced, because Hg^{2+} is bound to the biota, the evaporation of Hg^0 will be reduced until a new equilibrium with a higher tHg content is reached. Secondly, when MMHg^+ is removed from the water column due to bioaccumulation, only a small amount of tHg is removed ($< 3\%$), which means that
685 methylation of Hg^{2+} into MMHg^+ is reduced by only a small amount. This would favor the net production of MMHg^+ which leads to an increase in tMeHg.

5 Summary

In this study, we hypothesized that the ecosystem is an essential part of the marine Hg cycle. We quantified the impact of the ecosystem on the marine Hg cycle by simulating a 1D water column with and without bioaccumulation, biogenic reduction, and
690 partitioning into detritus and dissolved organic matter (DOM). Furthermore, we ran a 3D model for the North and Baltic Seas with and without bioaccumulation to analyze the spatial effects. Our analysis focused on the effects of these ecosystem interac-

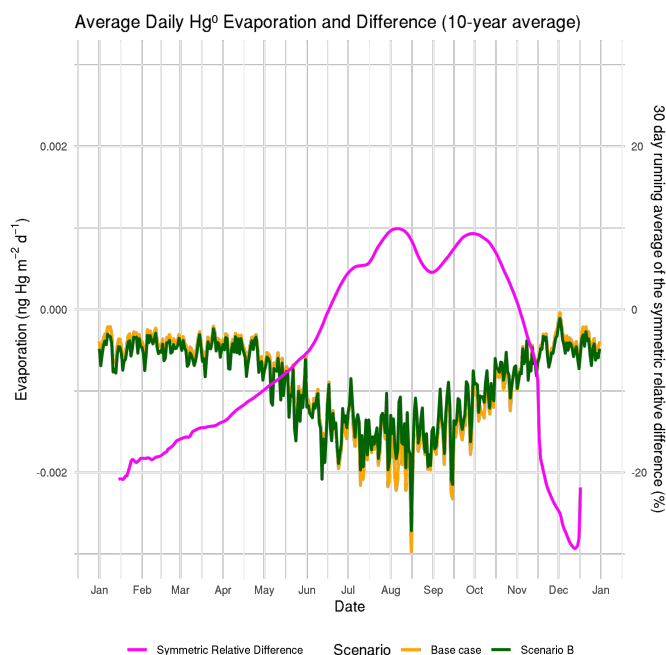


Figure 12. The 10-year average daily atmospheric exchange of Hg^0 between the atmosphere and the sea surface in the base case and scenario c (no bioaccumulation nor biogenic reduction).

tions on total Hg (tHg) and total methylated Hg (tMeHg) concentrations in the water column and the intercompartmental fluxes.

Our model demonstrates the complex differences in bioaccumulation between different species of Hg. The model accurately reproduces bioaccumulation at the base of the food web as shown in Table 6 and models biomagnification to higher trophic levels according to observations for both Hg^{2+} and MMHg^+ as shown in Fig. 7. Although our model underrepresents MMHg^+ in cod compared to the observed mean, this underrepresentation is still within the observations and is explained by an underrepresentation of the trophic level as shown in Table 4. It is important to note that we wanted to implement realistic bioconcentration and trophic transfer rates to not over-tune the model. Several interactions, such as, for example, cannibalism within the functional groups or even the same species, can increase bioaccumulation in ways that are not captured by the model, which would result in both increased bioaccumulation and trophic levels.

The impact of the ecosystem on the MeHg cycling is very strong. In idealized 1D setups, bioaccumulation increases the average tMeHg content by 44% in the Southern North Sea, by 13% in the Northern North Sea, and in the Gotland Deep above the mixed layer depth by 22% as shown in Table Fig. 10. The surge in tMeHg attributed to bioaccumulation is most notable during plankton blooms, where phytoplankton absorbs a large portion of tMeHg, shielding bioaccumulated MMHg^+ from demethylation processes. This MMHg^+ is released to the water column towards the end of the year, where less solar radiation, more particles, more mixing, and more cloud coverage reduce photodemethylation. Our models show that the increase in

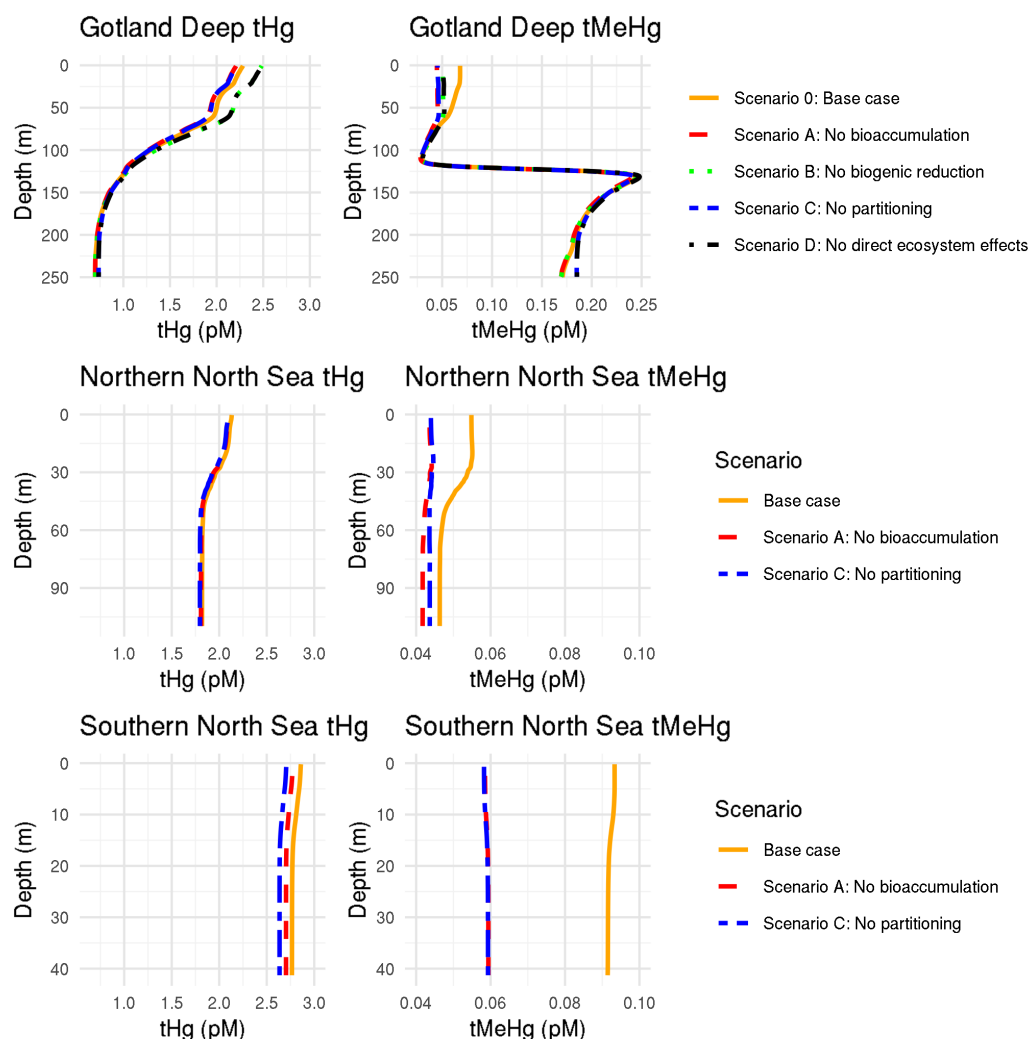


Figure 13. Depth profiles of tHg and tMeHg in the three setups with different model scenarios. In the Gotland Deep, the scenario without biogenic reduction (green) has an increased tHg content. This shows that biogenic reduction can reduce the yearly average water column Hg when cyanobacteria are present for part of the year. Setups without binding to detritus and DOM (blue) have lower tHg in the surface layer. This effect is strongest in the North Sea. Setups without binding to detritus and DOM (blue) also have reduced tMeHg in oxic water, but the effect is smaller than for tHg. In the Gotland Deep anoxic water, the only noticeable effect is caused by binding to detritus and DOM. In this setup, without this, there is both more tHg and tMeHg in the anoxic deep water. In all setups, runs with bioaccumulation (orange) have more tMeHg than without bioaccumulation (red). The Southern North Sea has a slightly lower amount of tHg than the Northern North Sea and Gotland Deep setups.



tMeHg is strongly related to average biomass. And that an increase of 4.5 mgC m^{-3} in the average biomass content will result in a 1% increase in tMeHg due to bioaccumulation.

710 The model reveals regional and seasonal differences in how the ecosystem influences Hg cycling and bioaccumulation across the North and Baltic Seas. The ecosystem increases tHg and tMeHg mostly in highly productive shallow coastal regions, with a reduced effect in deeper, less productive zones. In the Baltic Sea, our findings quantify the average reduction in tHg due to bioaccumulation and cyanobacterial-induced biogenic reduction as a 7% reduction in tHg above the mixed layer depth.

715 The 3D simulation expands on this by visualizing the spatial pattern and allowing us to quantify the effect of bioaccumulation on the Hg budget of the North and Baltic Seas while taking into account spatial variability and horizontal transport. The average tMeHg content increases by 10%, but this increase is caused by an increase in the North Sea and the Danish straits, while there is a decrease in tMeHg in large parts of the Baltic Sea.

5.1 Conclusion

720 This study demonstrates and quantifies the complex role of the ecosystem in shaping Hg speciation and the influence of physical, biochemical, and ecological factors. In addition, it shows the sensitivity of parameterization in modeling and shows that evaluating ecosystem parameters such as the trophic level is essential to comprehend the results of bioaccumulation modeling. Here, we show that bioaccumulation does have notable feedback effects on Hg cycling and therefore should be included in any marine Hg model, even in cases where bioaccumulation is not of direct interest. We conclude that the ecosystem has a direct effect on Hg cycling by;

- 725
- The ecosystem increases marine tMeHg
 - Increase in tMeHg up to 77%
 - Increase of the 10-year average tMeHg up to 44%
 - An increase of 4.5 mgC^{-3} average biomass leads to a 1% increase in tMeHg.
 - Facilitate burial of Hg by transporting Hg to below the thermocline in deep unmixed water via the sinking of detritus
 - 730 – Cause a wintertime increase in both aquatic and tMeHg permanently mixed productive coastal water

Because of this we conclude that the ecosystem has essential feedback on marine Hg cycling.

6 Future outlook

It is important to continue improving our models and to increase our understanding of the mechanisms that drive Hg bioaccumulation and cycling. More efforts should be directed toward understanding the nuanced interactions between Hg cycling and biogeochemical processes, trophic interactions, and ecosystem structure, all of which influence Hg speciation and fate in marine ecosystems. Typically, bioaccumulation is thought of as an add-on to model additional end members, and thus as non-essential

735



when modeling global transport and air-sea exchange. We demonstrate that this is not the case and that bioaccumulation plays an important role in Hg cycling in coastal oceans.

740 To support the Minamata Convention, a solid understanding of the Hg cycle in marine environments is important for designing effective management strategies that aim to mitigate Hg contamination and minimize its impact on aquatic ecosystems and human health. By improving the models and by incorporating new advancements in field observations and experimental studies, we can increase our ability to predict and manage Hg pollution, safeguarding both the environment and human well-being.

Conflict of interest

745 None of the authors declare any competing interest.

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Author contributions

750 The authors contributed to the article as stated in Table 9.

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AI assisted spell check was used in Grammarly and Writefull, while readability suggestions were occasionally provided by an rAI model such as chatGPT (OpenAI). In addition, AI tools helped optimize R and Python visualizations. All suggestions were critically evaluated and implemented only after manual verification. All final text is written and verified by the authors. The
755 sources were searched with Perplexity AI or Google Scholar, but all sources were manually read, verified, and cited.



Table 9. Contributions per Author. Authors are: David Johannes Amptmeijer (DA), Dr. Johannes Bieser (JB), Dr. Ute Daewel (UD), Elena Mikheeva (EM), and Prof. Dr. Corinna Schrum (CS).

Contributor role	Role definition	Authors
Conceptualisation	Identified the need for bioaccumulation in the MERCY v2.0 model	JB, CS
	Conceptualised the study	DA, JB, CS
	Coupling of MERCY v2.0 to FABM	DA, JB
Methodology	Developing the physical setups	EM
	Developed ECOSMO E2E to better suit bioaccumulation	DA, UD
	Design and implement the different scenarios	DA
Validation	Validate if Hg cycling matches in the 1D MERCY v2.0 model	JB, DA
	Validate hydrodynamic conditions	EM, CS, DA
	Validate carbon cycling	DA, UD
Writing	Writing of the original draft	DA, JB
	Reviewing the original draft and quality control	JB, CS, UD, EM, DA
Supervision	Supervising the development of the work	CS, JB
Funding acquisition	Acquired funding via the GMOS-Train ITN	JB

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