



1 Reinforce lake water balance components estimations by integrating water isotope

2 compositions with a hydrological model

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10 Abstract:

11 Accurate estimation of water balance components of groundwater-fed lakes, including 12 subsurface inflow, as well as actual evaporation from lakes, poses a complex task for 13 hydrologists employing hydrological models. Hence, an alternative approach was used to 14 capture the dynamic behavior of the hydrological groundwater/surface water system, which 15 can be used for integration with the hydrological model and serves as a validation for the 16 hydrological model estimates of the water balance components. The approach, based on measurements of the stable isotopes (δ^{18} O and δ D) enables the quantitative estimation of the 17 18 individual water flux and evapotranspiration rates. An isotope-mass-balance model was used 19 to quantify lake water balances over a one-year sampling period. The approach is based on the global relationship between the δ^{18} O and δ D values of the precipitation and kinetic isotopic 20 21 fractionation in the lake water during evaporation. Assuming that the lake is hydrostatically 22 connected to the groundwater the isotope mass-balance model accounts for the quantification 23 of the evapotranspiration rate considering the groundwater inflow compensating the 24 evaporation loss. The study addresses the model-based quantification of subsurface-25 groundwater inflow and evaporation losses of a young glacial groundwater lake (Lake Gross 26 Glienicke (GG), southwest of Berlin in the Havel catchment), over the period from 2015 to 27 2023 with the integrated hydrological model HydroGeoSphere. Utilizing the isotopic mass 28 balance model, HydroCalculator, under steady-state hydrologic regime conditions, the 29 evaporation-to-inflow (E/I) ratio is determined for the period of one year spanning August 30 2022 to September 2023. Employing the fully integrated hydrological model, calibrated and 31 validated under monthly normal transient flow conditions from 2008 to 2023 for the lake





32 catchment, subsurface, and groundwater inflows to the lake are calculated and compared to 33 the calculated E/I ratios based on the isotopic measurement of the lake water. Isotopic signatures of surface water, groundwater, and rainwater (δ^{18} O and δ D) confirm a flow-34 35 through type for the lake. The calculated E/I ratio for GG Lake is around 40%. The calculated 36 evaporation for the years 2022 and 2023, within the isotopic mass balance model framework 37 $(E_{iso22} = 601 \text{ mm}, E_{iso23} = 553 \text{ mm})$, aligns well with the actual evaporation from the lakes 38 calculated by the HGS model ($E_{HGS22} = 688$ mm, $E_{HGS23} = 659$ mm). The change in the ratio 39 of evaporation to inflow (E/I) leads to a significantly improved estimation of evaporation 40 rates after correction for temperature fluctuations and inflow data from previous years (2015-41 2021). With a correlation coefficient of 0.81, these revised estimates show a high degree of 42 agreement with the evaporation rates predicted by the HydroGeoSphere (HGS) model for the 43 corresponding years. Despite the uncertainties associated with the analysis of the water 44 isotope signature, its integration into the hydrological model serves to validate the 45 hydrological model calculations of the water balance components.

Keywords: HydroGeoSphere, HydroCalculator, Lake Water Exchange, Evaporation Loses,
Stable Isotope, Gross Glienicke Lake

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49 **1. Introduction**

50 Hydrological models have undergone substantial advancements in past decades (Singh, 2018; 51 Herrera et al., 2022), but still face unsolved problems and uncertainties in depicting 52 hydrological processes (Liu and Gupta, 2007; Renard et al., 2010). Recently, a wide range of 53 monitoring and modeling techniques have emerged for investigating water fluxes at different 54 scales (Fekete et al., 2006, Windhorst et al., 2014). However, limitations in the hydrological 55 model parametrizations lead to insufficient quantification of water flows, and therefore a 56 quantitatively and qualitatively incorrect interpretation of hydrodynamic processes and, as a result, inaccurate assumptions for water management purposes (Müller Schmied et al., 2014). 57 58 Improving the informative value of the models, representing complex hydrological processes 59 is needed to enhance the applicability of the models for future estimations and scenario 60 analyses. The optimal adaptation of hydrological models to real conditions for a precise determination of water balance components is often considered unattainable within the 61 62 current technical possibilities due to the overwhelming amount of work associated with field 63 measurements. Particularly the quantification of groundwater-surface water exchange using





64 the hydrological models faces pronounced uncertainties in considering geostructural 65 heterogeneities in different scales. To validate the model results, arduous monitoring surveys 66 are required to measure for example the groundwater-surface water interactions along rivers' 67 banks or lakes' shorelines (Partington, 2020). Hence, a combination of hydrogeological modeling, field measurements, and innovative isotopic-based studies along with appropriate 68 69 linkage between these approaches will be a concrete way to achieve optimal parameterization 70 and validation capabilities for modeling hydrological processes in complex geohydraulic 71 systems.

72 Recent studies show that the stable water isotope mass balance on different water sources 73 together with the numerical models improve the model performance in simulating the 74 interactions between groundwater and surface water (e.g., Jafari et al., 2021). The isotopic 75 insights enhance hydrogeologists' efforts to calculate water balance components such as the 76 groundwater inflow to the surface water resources and water losses due to evaporation 77 (Skrzypek et al., 2015; Vyse et al., 2020). This method is based on the fractionation of heavier 78 isotopes caused by evaporation from surface water, provoking a disparity in isotopic 79 composition between groundwater and surface water. However, a limitation of this approach 80 is that the stable water isotope mass balance is restricted to the sampling time, making it 81 incapable of accounting for the transient behavior of groundwater-surface water exchange 82 over different time periods. Although water isotope tests can describe the changes in water 83 fluxes in specific time frame intervals (e.g. monthly), deficiency in providing spatial 84 exchanges between groundwater and surface water can be also introduced as their limitation. 85 These gaps can be addressed by integrating water isotope analyses with physical-based 86 hydrogeological models.

Water isotope analyses are helpful techniques to evaluate the model's performance. For
instance, Ala-Aho et al. (2015) assess the hydrogeological model performance by comparing
the simulated groundwater inflow to lakes in the middle of Finland with calculated recharge
by water isotope analyses.

In the northeast of Germany, groundwater levels and landscape runoff have largely been in decline for over three decades (Lahmer 2003; Germer et al., 2011; Merz and Pekdeger, 2011); regional climate studies suggest further decreases over the next decades (Gerstengarbe et al., 2003, 2013). Thus, water resource management for this region requires a thorough assessment of possible adaptions and measures to counteract or mitigate severe consequences, such as decreasing groundwater heads and surface water levels and declining groundwater and surface





97 water quality. The development of integrated management schemes for groundwater-98 dependent ecosystems such as lakes under climate changes, requires a more comprehensive 99 understanding of hydrological dynamics and a better estimation of ecologically relevant water 90 fluxes are of prime importance. Therefore, an alternative approach was developed to capture 91 the complex behavior of the hydrological groundwater-surface water system. This approach 92 can be integrated with hydrological models to improve parameterization and validate model 93 calculations in water balance components.

104

105 **2. Materials and methods**

106 2.1 Study area

Gross Glienicke Lake (GGS) with an area of 0.59 km^2 and a maximum depth of 10 meters is 107 located in Berlin-Brandenburg state, Germany (30-87 m.s.l., Fig. 1a). It is a young glacial 108 lowland lake that is exclusively fed by groundwater. The lake's water levels have shown 109 110 significant seasonal variability (around 0.4 m) over recent decades (Fig. 2). Since 2014 the 111 lake's water level has faced severe drops (Fig. 2) around 1.28 meters. GGS as a seepage lake 112 (lake without an outlet) is surrounded by low-density residences (Fig. 1). In the north and 113 northwestern areas, grassland and farmland are the dominant surface cover and directly overly sandy soils. The aquifers are recharged in the Döberitzer-Heide region where fine sandy soil 114 with a hydraulic conductivity ranging from 8×10^{-5} to 12×10^{-5} m/s can be found according to 115 the lab analysis. The root zone within the grassland ranges in depth from the surface to 30 cm. 116 117 The groundwater monitoring measurements illustrate a smooth hydraulic head gradient from West to East, highlighting connections between the lake and the aquifer recharge area from 118 119 the Döberitzer-Heide region (Fig. 3). On the east side of the lake catchment, a regulated river, 120 the Havel, has been flowing from northeast to southwest. Continuous stratigraphic units have 121 been delineated throughout the lakes' catchment based on the geological features information 122 collected from 480 boreholes (Fig. 2, The Federal Institute for Geosciences and Natural 123 Resources - BGR 2022). The geology in the study area is formed by a series of layered Pleistocene and Tertiary sediments that are approximately 150 to 200 m thick, with a lower 124 125 confining bed of Oligocene marine Rupel clay. The series consists of a complex interplay of glacial deposits from the Pleistocene and permeable marine and limnic sediments of the 126 127 Upper Oligocene and Miocene. The series can be divided into an upper unconfined aquifer 128 system of shallow Weichselian and late Saalian sediments. In general, a shallow (i.e., 5 to 10





129 m) unconfined aquifer is separated from the thick (140 to 190 m) lower confined aquifer by a 130 15 to 20 m thick layer of Saalian sediments. The confined and unconfined aquifers consist of multiple permeable sediment layers partially disconnected by layers of till, but still 131 132 hydraulically connected. The hydraulic connection to the lakes is mainly controlled by these 133 aquifer layers. Underneath these sediments is a thick confined aquifer system of the early 134 Saalian and Elster layers, and Upper Oligocene and Miocene sediments. The first shallow, 135 unconfined aquifer in the catchment area is characterized by highly permeable glacial sand and gravel deposits (Holocene and Weichselian). A till layer (Weichselian age, Fig. 3) is 136 137 found in the underlying layers. The till is underlain by late Warthe sandy sediments forming 138 the second aquifer.

139

140 2.2 Available meteorological data

141 From 1990 to 2023, radar-based CER v2 data (The Central Europe Refined Analysis version 142 2, details on the data pre- and past-processing are provided by Jänicke et al., 2017) generated a mean daily temperature of 10.4 °C and average annual precipitation of 612 mm, with an 143 144 average annual actual evapotranspiration (PET) of 639, 646, and 670 mm for farmland and 145 grassland, forest, and urban areas respectively. The annual mean humidity in Gatow station 146 varies from 50% to 70% over the last two decades (2000-2023). During the hydrological year 147 of the survey (August 2022 to September 2023), the mean precipitation, temperature, and 148 humidity in the study region were 51.9 mm, 12.3 °C (Potsdam weather station, DWD, 2024;), 149 and 69 % respectively (recorded in Gatow weather station of The Berlin Measurement 150 Network (MEVIS, Fig. 1). Precipitation data as one of the boundary conditions in the 151 modeling work has been obtained from the Potsdam station of the German Weather Service 152 (DWD, 2024).







Fig. 1. Location of the study area, highlighting Gross Glienicke Lake (GGS), Sacrow Lake (SAS), Havel
channel, piezometers on the east (E-GGS) and west (WD-GGS and WS-GGS) side of GGS, Berlin waterworks,
and land use classifications. © OpenStreetMap contributors 2023. Distributed under the Open Data Commons
Open Database License (ODbL) v1.0..



158

159 Fig. 2. Lakes water level fluctuations and precipitation variations during the period of 2008 to 2023.

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162

Fig. 3. Conceptual cross-section of the aquifer demonstrating the geological structure, the hydraulic head, and water flow direction from the groundwater recharge area towards Gross Glienicke Lake (GGS) and the groundwater discharge area with a lower altitude. WD-GGS, WS-GGS, and E-GGS are the piezometers on the west and east sides of GGS.

167

168 2.2 Stable isotope analysis

169 Surface water samples were collected for one year (from August 2022 to September 2023) on 170 a monthly time interval from GGS and three piezometers installed in the first two aquifers, 171 encompassing the lakes, and from two rainwater samplers. The total depth and well water 172 volume of the monitoring well, and the stability of in-situ parameters such as temperature, pH, and electrical conductivity (EC) were monitored as guidance of appropriate timing for 173 174 water sampling from the piezometers. All water samples collected in clusters within two-day excursions were filtered through a membrane filter (0.2 and 0.45- μ m pore) and stored at 6 °C 175 to prevent evaporation before laboratory analysis. Stable isotope ratios of oxygen (¹⁸O/¹⁶O) 176 177 and hydrogen (²H/¹H) in H²O in water samples were measured with a PICARRO L1102-i 178 isotope analyzer. The L1102-i is based on the WS-CRDS (wavelength-scanned cavity ring-





179 down spectroscopy) technique (Gupta et al., 2009). Measurements were calibrated by the 180 application of linear regression of the analyses of IAEA calibration material VSMOW, VSLAP, and GISP. The stable isotope ratios of oxygen and hydrogen are expressed in the 181 conventional delta notation ($\delta^{18}O$, δD) per mil (∞) versus VSMOW. For each sample 6 182 183 replicate injections were performed and arithmetic average and standard deviations (1 sigma) 184 were calculated. The reproducibility of replicate measurements is generally better than 0.1 ‰ 185 for oxygen and 0.5 % for hydrogen.

186

187 2.3 Isotope mass balance model

188 The evaporation loss from a lake such as GGS Lake can be calculated by knowing the 189 transient stable water isotope compositions of inflow and moisture in ambient air and climate 190 data (air temperature and humidity) for the specific period of time and considering a steadystate hydrologic condition (no additional water inflows, Tweed et al. 2009; Gibson and Reid, 191 192 2010). The isotope mass balance model (Hydrocalculator) whose capability has been verified 193 through various field experiments globally (Skrzypek et al. 2015; Vyse et al., 2020) was 194 applied to estimate evaporation over inflow ratio (E/I) for the GGS Lake in the steady state 195 condition. The differences between stable isotope compositions of water samples reflect the 196 isotopic phases: enrichment (heavier isotope) or dilution (lighter isotope). Hence, a series of 197 time-based analyses enables the assessment of evaporation progress. Climate data from 198 nearby weather stations (Gatow and Potsdam) were utilized to address uncertainties arising 199 from the distance to the points of water samplings (Gibson and Reid, 2014; Skrzypek et al. 200 2015). The stable isotope composition of moisture in ambient air (δ_{air}) is estimated from the 201 mean monthly weighted averages from the stable isotope composition of precipitation (δ_{pcp}) 202 of GNIP station (GNIP/Berlin (DWD, BFG, BGR & HHZM, Stumpp et al. 2014 and Schmidt 203 et al. 2020) which were corrected by local precipitation stable isotope composition. The δ_{air} is 204 calculated based on the rain and rain-LEL as follows (Gibson and Reid, 2014):

205
$$\delta_{air} = (\delta_{pcp} - \varepsilon^+)/(1 + \varepsilon^+ \times 10^{-3})$$
 Eq. 206 1

206

207 where ε^+ is an isotope fractionation factor that is solely temperature-dependent. ε is the total 208 fractionation factor, and equals the sum of the equilibrium isotope fractionation factor ε^+ , as 209 given above plus the kinetic isotope fractionation factor ε_k (Gibson and Reid, 2010):





210	$\varepsilon = \varepsilon^+ / (1 + \varepsilon + \times 10^{-3}) + \varepsilon_k $ Eq.		
211	2		
212	The kinetic fractionation ε_k is defined as (Gat 1995):		
213	$\varepsilon_k = (I - h) \times C_k$ Eq.		
214	3		
215	According to Gonfiantini, 1986 and Araguas-Araguas et al., 2000, the kinetic fractionation		
216	constant (C_k) is 12.5 percent for δD and 14.2 percent for $\delta^{18}O.$ Air relative humidity (h) is		
217	given as a fraction.		
218	Based on the local climate conditions the enrichment of stable isotope compositions can be		
219	limited. According to Gat and Levy (1978) and Gat, (1981), this limitation threshold (δ^*) can		
220	be estimated by considering air humidity (h), δ_{air} , and a total enrichment factor (ϵ).		
221	$\delta^* = (h \times \delta_{air} + \varepsilon) / (h - \varepsilon \times 10^{-3})$ Eq.		
222	4		
223	When this limitation exceeds, further evaporation does not result in isotope enrichment.		
224	The ratio of evaporation over inflow (E/I) can be calculated using the following reformulated		
225	equation (e.g. as by Mayr et al. (2007)) under steady-state hydrological conditions. E/I is the		
226	fraction of inflowing water evaporated from GGS Lake:		
227	$E/I = ((\delta_{inflow} \cdot \delta_{outflow}) / (\delta^* \cdot \delta_{inflow}) \times E_s) $ Eq. 5		
228	enrichment slope (E_s) is defined by Welhan and Fritz, 1977 and Allison and Leaney, 1982		
229	accordingly:		
230	$E_s = (h - (\varepsilon \times 10^{-3})) / (1 - h + (\varepsilon \times 10^{-3}))$ Eq.		
231	6		
232	The model calculates the evaporative losses based on the theory behind the Craig-Gordon		
233	model (Gibson and Reid (2014)). The variables used in the Hydrocalculator model are listed		
234	in Table 1.		
235	Table 1. The list of variables used in the Hydrocalculator model		

Variable	Description	Unit
Т	temperature	°C
h	air relative humidity	-





δ_{air}	stable isotope composition of moisture in ambient air	%
δ_{pcp}	stable isotope composition of precipitation	%
LEL	slope of the local evaporation line	
ε	total isotope fractionation	%
\mathcal{E}^+	equilibrium isotope fractionation factor	%
Ek	kinetic isotope fractionation factor	%
C_k	kinetic fractionation constant	
δ^*	limiting isotopic composition	%
E/I	Evaporation over inflow ratio	%
δ_{inflow}	stable isotope composition of inflow (groundwater)	%
$\delta_{outflow}$	stable isotope composition of outflow (lake)	%
E_s	enrichment slope	-

236

237 2.4 Model domain configuration and boundary conditions

238 2.4.1 Surface – subsurface flows

The HydroGeoSphere (HGS) modeling code (Aquanty Inc, 2023) was used to simulate the hydrological processes in the GGS Lake catchment. HGS is a 3-D, fully integrated, and physically-based model with the capacity to simulate the interwoven flow mechanisms of subsurface and surface water by coupling solutions obtained from the diffusion-wave of the two-dimensional, depth-integrated diffusion-wave of the Saint Venant equation governing surface water flow (Eq. 8, Viessman Jr. and Lewis, 1996) and the Richards' equation governing three-dimensional unsaturated and saturated subsurface flows (Eq. 9).

Eq. 7

247
$$\frac{\partial \phi_0 h_0}{\partial_t} - \frac{\partial}{\partial_x} \left(d_0 K_{0x} \frac{\partial h_0}{\partial_x} \right) - \frac{\partial}{\partial_y} \left(d_0 K_{0y} \frac{\partial h_0}{\partial_y} \right) + d_0 \Gamma_0 \pm Q_0 = 0$$

248 ϕ_0 represents the porosity (dimensionless) of the surface flow domain, which varies based on 249 the presence of rills and obstructions. h_0 stands for the water surface elevation (L). *t* denotes 250 time (T). d_0 indicates the depth of flow (L). K_{0x} and K_{0y} represent surface conductance. Γ_0 is 251 the water exchange rate (L³ L⁻³ T⁻¹) occurring between the surface and subsurface systems. 252 Q₀ represents external sources or sinks.

253 The interaction between the two flow domains is facilitated by the exchange term Γ_0 through:

254

Eq. 8





$$d_0 \Gamma_0 = \frac{k_r K_{zz}}{l_{exch}} (h - h_0)$$

256 k_r symbolizes the exchange's relative permeability. K_{zz} represents the saturated hydraulic 257 conductivity in the vertical direction. l_{exch} corresponds to the coupling length.

258

Eq. 10

259
$$\nabla \cdot (W_m q) + \sum \Gamma_{ex} \pm Q = W_m \left(\frac{\partial}{\partial_t}\right) (\theta_s S_w)$$

In the given context: *Wm* (dimensionless) represents the volumetric porosity fraction within the porous media domain. Γ_{ex} stands for the volumetric exchange rate (L³ L⁻³ T⁻¹) occurring between the porous media and other flow domains. *Q* denotes the source or sink term. *t* signifies time (T). θ_s corresponds to porosity (dimensionless). *S_w* refers to the degree of water saturation (dimensionless).

265 The flow rate q (L T⁻¹) is portrayed as:

266

$267 \quad \boldsymbol{q} = -\boldsymbol{K}.\,\boldsymbol{k_r}\,\boldsymbol{\nabla}\boldsymbol{h}$

268 *K* signifies the hydraulic conductivity (L T⁻¹). k_r represents the relative permeability 269 (dimensionless), which is dependent on water saturation. *h* corresponds to the hydraulic head 270 (L), calculated as the sum of the elevation head and pressure head.

The three-dimensional surface-subsurface flows in porous media and saturated zones were solved with the control volume finite element method. Nonlinear equations were linearized using Newton-Raphson and solved iteratively at each time step for the entire hydrologic system.

275

276 2.4.2 Evapotranspiration

The evapotranspiration process needs specific prerequisites for accurate parameterization as it is treated to play a dual role in the HGS as both a boundary condition and a distinct domain. Within this framework, the evapotranspiration fluxes encounter a restriction governed by a potential evapotranspiration flux (PET) which is defined by the modeler. The PET values are designated a boundary condition, serving its purpose on the surface domain. With each subsequent time step, a condition emerges, if the calculated actual evapotranspiration (AET)





283 surpasses the PET, then the PET value is employed as a flux directed toward the relevant 284 model faces. Conversely, if the calculated AET falls short of the PET, the computed AET 285 value itself becomes the applied flux. Additional details on the evapotranspiration process 286 formulations within the HGS model are presented in Kristensen and Jensen (Kristensen and 287 Jensen, 1975). The two-dimensional PET database used in this research is calculated using the 288 energy balance method and covers the period 2000 - 2022. The method is a balance of the 289 energy terms which are the net radiation, the change in the heat content of the lake, and the 290 latent and sensible heat fluxes. The equation is based on measurements of global radiation, air 291 and water temperature, cloud cover, and vapor pressure. The latent heat flux, which represents 292 the energy used for evaporation, was determined by subtracting the sensible heat fluxes and 293 the change in the heat content of the lake from the net radiation. The evaporation rate was 294 then calculated by dividing the latent heat flux by the latent heat of vaporization of the water. 295 A maximum evaporation threshold of 15 mm/day was set. More details are given by Ölmez et 296 al., 2024.

297 The simulation domain encompasses the entire GGS Lake catchment (Fig. 4) which is defined 298 based on the surface topography considering the equipotential lines derived from the lakes' 299 levels and measurements of the hydraulic head surrounding the lakes (piezometers). The surface topography across the catchment was produced by stitching a digital elevation model 300 301 (DEM) from The Shuttle Radar Topography Mission (SRTM) with a resolution of 30 m, and 302 the bathymetry data of GGS (Wolter, 2010) and Sacrow Lake (SAS) (Lüder et al., 2006, 303 Bluszcz et al., 2008). Due to the high vegetation density, flat elevations, and the substantial 304 hydraulic conductivity of the predominantly sandy soil, the absence of river formation is currently observed in the catchment. The foundational 2-D triangular mesh supporting the 305 306 comprehensive 3-D triangular prism grid within the HGS model was created using AlgoMesh 307 (Merrick, 2017). Each 2-D mesh layer encompasses a total of 2837 mesh nodes and 5300 308 triangular finite elements (Fig. 4). The complete 3-D model (Fig. 4) grid extends the 2-D 309 mesh across 15 subsurface layers, broadly categorized as one soil layer, 14 Quaternary 310 material layers, and one competent bedrock layer (Rupel clay).







311

Fig. 4. Hydrostratigraphic units and enlarged view of the mesh within the Gross GlienickeLake catchment

314 Spatially distributed land cover data (Fig. 1) were utilized to capture a broad range of factors 315 influencing evapotranspiration and overland flow, including evaporation depth, root depth, leaf area index (LAI), surface roughness, rill storage height, and obstruction storage height. 316 317 Specific parameters for evapotranspiration (ET) and overland flow are tailored to each land 318 cover type. To accurately reflect the impact of vegetation growth on water demand through 319 evapotranspiration, the Leaf Area Index (LAI) during winter (January) and summer (July) 320 using the Sun Sacan device type SS1 were measured, capturing both maximum and minimum 321 current LAI values. The measured LAI indices were then compared with data from the 322 MCD15A2H Version 61 Moderate Resolution Imaging Spectroradiometer (MODIS), which 323 provides a 4-day composite with a pixel size of 500 m for January and July (Myneni et al., 324 2021). Corrected monthly average MODIS LAI values for each land cover type, spanning 325 from 2000 to 2023, were subsequently integrated into the HGS model.

326 2.4.3 Unsaturated zone





327 The top subsurface layer in the 3-D mesh with spatial varying depths shows the distribution of 328 soil materials across the catchment. The soil data were obtained from the soil map with a 329 scale of 1:200,000 (BUEK200) which was prepared by the Federal Institute for Geosciences 330 and Natural Resources (BGR, 2007). The soil samples were collected from various depths, 331 extending up to 3 meters, at 10 different sites, primarily in the groundwater recharge area 332 (Döberitzer-Heide region) and natural conservation zones. The sampling locations were 333 selected based on soil types. The percentages of sand, silt, and clay for each soil type were 334 determined in the laboratory to classify the soil textural types, using the United States 335 Department of Agriculture (USDA) soil textural calculator. A set of 2 soil textural types, 336 sand, and loamy sand has been recognized. Unsaturated soil hydraulic parameters and soil 337 moisture retention properties required for the van Genuchten application with the HGS model 338 were uniquely estimated for each soil type using the ROSETTA program, version 1 (Schaap et al., 2001). 339

340 Underlying the soil layers are 14 Quaternary geology layers that overlie bedrock. To represent 341 the topography of the subsurface in the lake catchment, relevant data was extracted from the 342 groundwater model provided by Berliner Wasserbetriebe. This model (software FEFLOW) 343 was calibrated in 2012 and updated in 2013 (BWB, 2012 and 2013) using measured data from 344 the year 2010. The model focuses on analyses of the waterworks at Beelitzhof, Tiefwerder, and Kladow. It, therefore, covers areas along both banks of the river Havel. Additional 345 346 datasets from boreholes were merged into a single surficial geology dataset using the 347 Rockware model setup for the GGS Lake catchment (Hermanns, 2022). The initial hydraulic 348 conductivity values for each type of Quaternary material were taken from the FEFLOW 349 model. The hydraulic properties of the unsaturated zone were manually adjusted during 350 manual model calibration.

351 2.4.4 Groundwater – lake levels loggers

352 A total of 8 groundwater monitoring wells scattered within the catchments, along with two 353 loggers set on the lakes, provide a well spatially distributed dataset of groundwater-lake level 354 dynamics for evaluating model simulations. The groundwater monitoring wells were selected 355 based on location, catchment area, and data availability spanning from 2000 to 2023. GGS has been monitored since 1964. Moreover, within the study catchment, the regulated flow 356 357 system of the river Havel is maintained to facilitate water conveyance. Since 1980, an 358 established logger has been operational to meticulously monitor the water level dynamics 359 within this river. For the presented study particular interest lies in loggers No. 51, 52 (WD-





GGS and WS-GGS), and 3154 (E-GGS). WD-GGS(50) and WS-GGS(51) belong to two
different aquifers (shallow(WS-GGS) and deep aquifer (WD-GGS) and are situated on the
western side of GGS (Brandenburg, Fig. 3). These loggers in the recharge area of the GGS
lake consistently maintain water levels averaging 15-20 cm higher than the GGS.
Additionally, logger E-GGS, located close to the eastern shoreline downstream of GGS
(Berlin), consistently registers water levels averaging 30-40 cm lower than the GGS lake (Fig. 3).

367 2.4.5 Groundwater abstractions

368 Two major drinking water supply systems, Kladow and Beelitzhof, located alongside Havel on the southwest side of Berlin have been in operation since 1888 and 1932 respectively by 369 370 BWB. Kladow comprises 16 wells up to 93 meters deep and a maximum pumping rate of 371 30,000 m³/day, while Beelitzhof has 85 wells up to 170 meters deep and a maximum pumping 372 rate of 160,000 m^3 /day. To assess the impact of groundwater withdrawals from deeper layers, 373 the model domain was extended to a depth of 150 meters below sea level. According to 374 studies BWB, upto 80 percent of the water extracted by Kladow originates from bank 375 filtration along the river Havel. The remaining 20 percent of the extracted water originates 376 directly from groundwater recharge as well as outflow from GGS. As various water resources 377 contribute to the overall drinking water production in the main waterworks in this area 378 (Beelitzhof, located on the western side of the Havel), a detailed analysis was conducted to 379 assess the share of the GGS Lake catchment. The analysis involved the implementation of 380 distinct scenarios within the hydrologic model.

381 2.4.6 Model evaluations

382 This study emphasizes the importance of a multifaceted approach to evaluate hydrological 383 model performance, utilizing both traditional and innovative methodologies. Initially to 384 evaluate model performance the simulated seasonal and long-term groundwater and lake level 385 fluctuations will be compared to observed water levels of the lakes and piezometers around the lakes. The performance evaluation of hydrological models commonly relies on various 386 387 metrics such as the Nash-Sutcliffe efficiency (NSE), percent bias (PBIAS), root mean squared 388 error (RMSE), and the Kling-Gupta efficiency (KGE). The KGE, introduced by Gupta et al. 389 (2009), offers a comprehensive assessment by considering bias, correlation, and variability 390 separately. Given the specific hydrological focus of each metric, a multi-metric approach was 391 adopted for calibrating the HGS model parameters, as demonstrated to efficiently balance





392 model performance by previous studies (Pfannerstill et al., 2014; Mahmoodi et al., 2020). For 393 model assessment, NSE, PBIAS, RMSE, and KGE were employed as performance metrics on 394 a monthly basis. Calibration runs were evaluated based on predefined thresholds for NSE 395 (0.65), PBIAS (-25% to 25%), and KGE (0.65) to identify the most suitable configurations. 396 The calibration process was carried out manually due to the model's long execution time and 397 limited computational capacity. Emphasis was placed on calibrating model parameters of the 398 unsaturated zone which governs water movement into the soil and subsequently into or out of 399 the aquifer. The initial hydraulic conductivity values for each type of soil were determined 400 from existing literature (Steidl et al 2023) and lab analysis and later manually adjusted during 401 model calibration. The model parameterization for the saturated zone was initially derived 402 from the FEFLOW model calibrated by the BWB (BWB, 2012/2013). The calibration and 403 validation periods chosen for the simulation runs were 2008-2018 and 2019-2023, 404 respectively, preceded by an eight-year spin-up phase before 2008 to reach quasi-steady state 405 condition fitting to the conditions in 2008.

406 To evaluate the performance of the HydroGeoSphere (HGS) model on different angles, a 407 detailed assessment involving the simulation of the inflow to GGS Lake (denoted as I_{HGS}) was 408 undertaken. This parameter (I_{HGS}) was subsequently used as a testing parameter to evaluate 409 the model's performance in calculating evaporation rates for the years 2022 and 2023 using an 410 independently determined E/I ratio. The evaporation rate (E_{ISO}) can be expressed as:

411

Eq. 11

$$E_{ISO} = \frac{1}{A} \cdot I_{HGS} \cdot P$$

Where A is the area of the lake water body (m^2) , I_{HGS} represents the annual inflow to the lake 413 414 (m^3) and P is the percentage of losses of inflow due to evaporation derived from the isotope 415 analysis (E/I). Figure 5 shows the methodology used to evaluate the model performance 416 across different dimensions. The underlying assumption for this evaluation is that an accurate 417 simulation of inflow to the lake by the HGS model (I_{HGS}) would yield evaporation rates (E_{ISO}) 418 comparable to those calculated by the HGS model (E_{HGS}). Thus, the consistency between 419 evaporation estimates derived from both approaches serves as a validation of the HGS 420 model's capability to simulate other water balance components precisely.







422

423 Fig. 5. Flow chart of the methodology employed to evaluate the model performance

424 The well-captured inflow and subsequent evaporation rates for the years 2022 and 2023 by 425 the HGS model, allow us to extend this approach for estimating evaporations during the 426 earlier period from 2015 to 2021. This period is crucial as it encompasses years without water 427 isotope analyses, during which significant drops in lake and groundwater levels were observed. However, the E/I ratio derived from recent years (2022 and 2023) cannot be 428 429 directly applied to earlier years due to variations in temperature and inflow, which are key 430 factors influencing isotopic signatures (dilution and enrichment). To adjust the E/I ratio for 431 earlier years, we incorporated annual temperature variations and inflow data into our model. Specifically, we compared the temperature and inflow of each specific year (γ_x) to the 432 433 corresponding values from 2022 and 2023(y22-23). This comparison yielded ratios for 434 temperature (T_{Yx}/T_{Y22-23}) and inflow (I_{Y2023}/I_{Yx}) , which were used to modify the E/I ratio accordingly. For instance, to apply the E/I_{Y22-23} to the year 2015, we multiplied the E/I_{Y22-23} 435 436 ratio by the temperature ratio (T_{Y15}/T_{Y2-23}) and the inflow ratio (I_{Y2023}/I_{Y15}). A temperature 437 ratio greater than 1 indicates higher temperatures in 2015 compared to 2023, suggesting a higher E/I ratio, greater evaporation, and enrichment. An inflow ratio greater than 1, 438 439 indicating lower inflow in 2015 compared to 2023, would lead to a greater E/I ratio, reflecting 440 greater evaporation and enrichment. The adjusted E/I ratios were then applied to refine the 441 initial evaporation estimates from the isotopic mass balance model. These revised evaporation 442 estimates were subsequently compared to the evaporation rates calculated by the HGS model 443 for the period 2015-2021.

444

445 **3. Results**





446 Isotopic analysis

447 Alterations in the mean monthly isotopic compositions of lake water and groundwater, along 448 with temperature and precipitation data, from August 2022 to August 2023 is presented in 449 Figure 6. The δD values for GGS (Fig. 6a), show significant variability, showing a 450 pronounced drop in δD values from around -8‰ in August 2022 to -17‰ in January 2023, 451 followed by a gradual decrease (except February) to approximately -16‰ by June 2023, which represent a strong dilution phase. The period from July to September demonstrates 452 enriched values of δD alongside rising temperature and evaporation as a consequence. The 453 454 δ^{18} O values for GGS (Fig. 6b) record a fluctuating pattern, ranging from -0.27% to 1.4%, 455 with peaks observed in August 2022 and July 2023, but experienced noticeable drops in April and May 2023. Overall, the isotopic data indicate that GGS experiences great isotopic 456 457 enrichment (heavier isotopic composition).

458 The δ^2 H values of groundwater on the east side of GGS (E-GGS, Fig. 6c) range from 459 approximately -40% to -50%, with notable fluctuations throughout the year. The isotopic 460 composition of groundwater on the west side of GGS (W-GGS, Fig. 6c) has less variability, with δD values mostly remaining between -55‰ and -60‰, suggesting a rather stable 461 isotopic environment. Despite the fluctuations in δD values, the $\delta^{18}O$ values (Fig. 6d) show 462 less variation, indicating some degree of isotopic stability in the oxygen isotopes in the 463 groundwater of both sides of lakes. E-GGS presents a relatively stable trend with δ^{18} O values 464 465 fluctuating between -5‰ and -6‰. W-GGS, with a seasonal pattern similar to E-GGS, shows a consistent range of δ^{18} O values between -8‰ and -8.5‰ with minimal fluctuations (Fig. 466 467 6d). Overall, the E-GGS with heavier isotopic signatures experiences greater isotopic variability, meanwhile, the W-GGS site maintains a more consistent isotopic signature, 468 469 indicative of a more stable hydrological regime. These observations (Fig. 6a,b,c,d) indicate 470 that the isotopic composition of both lake water and groundwater was generally heavier 471 (stronger enrichments) during the summer of 2022 compared to the summer of 2023.

The monthly average temperature (TMP, Fig. 6e), follows a clear seasonal pattern. It drops from around 20°C in August 2022 to a low of 5°C in January 2023, then rises again to about 20°C by July 2023. Alongside temperature, precipitation values fluctuate significantly, with peaks exceeding 90 mm in August 2022 and June 2023, and lower values around 20 mm observed in Oct and November 2022 and May and September 2023.







477

478 Fig. 6. Monthly averages of lake water isotopic compositions (a: δD and b: $\delta^{18}O$), groundwater isotopic 479 compositions (c: δD and d: $\delta^{18}O$), temperature (e), and precipitation (f) data from August 2022 to September 480 2023.

481 The relationship between δD and $\delta^{18}O$ values for lake water (GGS) and groundwater (WGGS 482 and EGGS) from August 2022 to September 2023 are illustrated in Figure 7. The isotopic 483 values for lake water are significantly clustered, with δD values between -25‰ and -5‰ and 484 $\delta^{18}O$ values from 2‰ to -2‰ and are isotopically heavier compared to precipitation and 485 groundwater, suggesting significant evaporative enrichment. WGGS exhibits $\delta^{2}H$ values from





486 -50‰ to -65‰ and δ^{18} O values from -9‰ to -7‰. EGGS δ^{2} H values range from 487 approximately -50‰ to -40‰, with δ^{18} O values between -7‰ and -5‰, indicating less 488 enrichment compared to lake water and higher enrichment compared to the groundwater on 489 the west side.



Fig. 7. Isotopic composition of lake water and groundwater measured on the west (WGGS) and east (EGGS)
sides of Groß Glienicke from August 2022 to September 2023. The local meteoric water line (LMWL) is driven
from (GNIP/Berlin (DWD, BFG, BGR & HHZM, Stumpp et al. 2014).

494

490

495 Variables used for the calculation of evaporative losses and evaporation over inflow ratio 496 (E/I) ratios calculated for GGS during the August 2022–September 2023 period are presented 497 in Table 2. The winterwater isotopic compositions (dilution phase) served as the initial 498 sampling point for calculating the E/I ratio in both years 2022 and 2023. The δ_A -value of the 499 ambient air moisture was calculated based on the stable isotope composition of local precipitation sampled in the Groß Glienicke region and the Lankwitz campus of the Freie 500 501 University Berlin. The calculated evaporative losses over inflow were equal to 43.4% and 502 42.3% based on δD and 30.11% and 29.4% based on $\delta^{18}O$ in 2022 and 2023 respectively. The 503 E/I ratio calculated based on δD is around 12% higher compared to the E/I based on $\delta^{18}O$. 504 Therefore, as a mean ratio, an average of 37% will be used for further analyses.

505

506**Table 2.** Variables used for calculation of evaporative losses and the ratio of total evaporation to inflow (E/I) as507a function of the measured δD and $\delta^{18}O$ isotope enrichments for Lake Gross Glienicke (GGS) surveyed during508the August 2022–September 2023.





Parameters	Description	δD	δ ¹⁸ Ο
3	Kinetic isotope fractionation factor [‰] (h dependent)	-887.5	-1008.2
ε*	Equilibrium isotope fractionation factor [‰] (T dependent)	78.7465	9.3468
3	Total isotope fractionation [%]	-814.5018	-998.9398
Ck	Kinetic isotope fractionation constant [‰]	12.5	14.2
α*	Equilibrium isotope fractionation factor [‰] (T dependent)	1.0787	1.0093
δ*	Limiting isotope composition	-134.7724	-30.2151
m	Enrichment slope	-1.0129	-1.0138
δΑ	Ambient air moisture	-124.9844	-16.7601
E/I _{Y2022}	Evaporation over inflow ratio [‰] of Groß Glienicke Lake in 2022	43.37	29.63
E/I _{Y2023}	Evaporation over inflow ratio [‰] of Groß Glienicke Lake in 2023	42.28	29.07

509

510

511 Hydrological modeling and model evaluations

512	Figure 8 illustrates the simulated vs. observed hydraulic heads (meters above sea level: m
513	a.s.l.) at West-GGS Piezometer (W-GGS), Lake Gross Glienicke (GGS), and East-GGS
514	Piezometer (E-GGS) over the period from January 2008 to December 2023. The model's
515	performance is evaluated using several metrics, including the Kling-Gupta Efficiency (KGE),
516	Percent Bias (PBIAS), and Root Mean Square Error (RMSE), as shown in Table 3. A strong
517	alignment is evident between simulated and observed hydraulic heads, both in terms of
518	magnitude and seasonality.

The simulated groundwater levels on the west side of the lakes, despite some over- and underestimations, showed very good agreement with the observed data. For the calibration period (2008-2018), the performance metrics are KGE of 0.86, PBIAS of 0.0%, and RMSE of 0.13 m. During the validation period (2019-2023), the model maintained high performance with a KGE of 0.82, PBIAS of -0.1%, and RMSE of 0.07 m. Both observed and simulated data exhibit a general declining trend over the study period, with hydraulic heads decreasing from approximately 30.15 m to 30 m.

For GGS, similar to W-GGS, both observed and simulated values show a decreasing trend from around 31 m in 2008 to approximately 29.80 m in 2023. The model simulations closely follow the observed data, with minor deviations. The performance metrics for GGS during the calibration period are KGE of 0.78, PBIAS of -0.2%, and RMSE of 0.13 m. In the validation period, the metrics are KGE of 0.75, PBIAS of 0.0%, and RMSE of 0.06 m, indicating high accuracy in representing the hydraulic behavior of the lake.

The observed and simulated groundwater dynamics on the east side of GGS show good
agreement. During the calibration period, the performance metrics are KGE of 0.84, PBIAS
of 0.1%, and RMSE of 0.09 m. In the validation period, the metrics are KGE of 0.70, PBIAS





- of 0.2%, and RMSE of 0.09 m. Overall, the high-performance metrics confirm the model's reliability and accuracy in capturing both the long-term trends and seasonal variations of groundwater-surface water dynamics within the study area, providing valuable insights into
- 538 groundwater-surface water interactions.

539



541 Fig. 8. Time series of observed and simulated hydraulic heads at three locations: (a) West Gross Glienicke
542 Piezometer (W-GGS), (b) Lake Gross Glienicke (GGS), and (c) East Gross Glienicke Piezometer (E-GGS) from
543 January 2008 to December 2023.

544 Table 3. Model performance evaluation using several metrics for both calibration (2008-2008) and validation

545 (2018-2023) periods

		West Gross Glienicke	East Gross Glienicke	Gross Glienicke Lake
Calibration	KGE	0.86	0.84	0.78





2008-2018	PBIAS	0.0	0.1	-0.2
_	RMSE	0.13	0.09	0.13
Validation	KGE	0.82	0.7	0.75
2019-2023	PBIAS	-0.1	0.2	0
_	RMSE	0.07	0.09	0.06

546

547 The calculated annual water exchange for GGS from 2008 to 2023, detailing the inflow to the 548 lake, outflow from the lake, and their differences (net flow = inflow - outflow) are presented 549 in Figure 9. The inflow to GGS demonstrates substantial annual variability. For instance, years such as 2008, 2011, 2012, and 2018 show relatively higher inflows compared to other 550 551 years. Notably, there is a discernible decreasing trend in inflow from 2011 to 2021, with 2018 552 being an exception. This trend indicates a progressive reduction in the hydrological inputs to 553 the lake over the decade. Simultaneously, the outflow from GGS shows significant annual 554 variability, with the highest outflow occurring between 2012 and 2017. This increased 555 outflow, coupled with the decreasing inflow, points to a period of significant net water loss 556 for GGS, potentially impacting the lake's water levels. Positive net flow values in years 2011, 557 2018, and 2022 indicate years when inflow exceeds outflow, contributing to the lake's water gain. Variation in both water gain and loss over the study period, reflects the complex 558 interplay of natural hydrological processes governing the lake's water balance. 559

560 Figure 10 illustrates the water exchange dynamics of GGS over the period from August 2022 to September 2023, highlighting seasonal patterns in inflow, outflow, and net flow. During 561 562 the warmer months, particularly June and July, the inflow to the lake peaks at approximately 563 90,000 m³, indicating a significant increase in water input during this period. Among the summer months (July to September), August has the lowest amount of inflow, with an 564 average of approximately 40,000 m³. In contrast, the inflow is pronouncedly lower during the 565 months of March, April, and May, averaging around 20,000 m³. The lake experiences positive 566 net flow, reflecting water gain, during the months of June and July. From December to May 567 568 (except February), the net flow is predominantly negative, indicating that outflow exceeds 569 inflow. During these months, the lake loses water, with outflow reaching its highest levels.







Fig. 9. Annual water exchange of Lake Groß Glienicke (GGS) from 2008 to 2023, showing the inflow to the
lake, outflow from the lake, and net flow (Outflow - Inflow)



573

570

574 Fig. 10. Monthly water exchange of Lake Gross Glienicke (GGS) from August 2022 to September 2023





576 To compare the lake evaporation values calculated by the HGS model and the results of the 577 isotopic signature of lake water actual evaporation (E) from GGS for the years 2022 and 2023 was calculated using Eq. 11, which incorporates simulated annual water inflow to GGS (I_{HGS}) 578 579 from the HGS model and the E/I ratio derived from isotope analysis (mean ratio: 37%). These 580 results were compared to the evaporation rates from GGS simulated by the HGS model (Fig. 581 11). The annual evaporation estimates from the isotope analysis for the years 2022 and 2023 582 show good agreement. The values calculated by the HGS model (E_{HGS}) are slightly lower (around 80 mm). The general agreement between the evaporation rates simulated by the HGS 583 584 model and the values derived from the isotope approach indicates accurate inflow simulation 585 by the HGS model. Transferring the E/I ratios from 2022 and 2023 to calculate evaporation 586 for earlier years (2015-2021, Fig. 11) results in notable discrepancies, particularly evident in 587 2015 and 2018, where evaporation reaches 450 and 1000 mm, respectively, compared to E_{HGS} values of 590 and 614 mm for those years. However, modified isotope analysis (considering 588 589 annual variations in inflow and temperature, see Eq.11) demonstrates good agreement with 590 the HydroGeoSphere model, emphasizing our approach of incorporating temperature and 591 inflow data for accurate evaporation predictions for earlier years. This suggests that while the 592 E/I ratio obtained from 2022 and 2023 can be applied to estimate evaporation in previous 593 years, adjustments for differences in inflow and temperature between those years and 2022-594 2023 are crucial for enhanced estimations of evaporations in years without isotope analysis 595 (E/I).









597 598

Fig. 11. Comparison of actual evaporation from Lake Gross Glienicke (GGS) estimated by Hydroeosphere
model (HGE) and isotope analyses (E/I ratio), and modification of isotope analyses (E/I ratio) considering
annual temperature and inflows to the lake. Annual temperatures are indicated, corresponding to the secondary
y-axis.

602

603 **4. Discussion**

604 The comparison of actual evaporation (E) from GGS for 2022 and 2023, derived from isotope 605 analysis (E/I) and HydroGeoSphere (HGS) model calculations, demonstrates the 606 comparability of these methods. Despite slight differences in the results, likely due to the 607 influence of direct precipitation with lighter isotope signatures diluting the lake water, the general consistency between the methods underscores the accurate simulation of inflows by 608 609 the HGS model. Seasonal variations in water isotopic compositions offer a valuable perspective for evaluating hydrological model simulations. Higher outflow observed from 610 611 December 2022 to May 2023 provides insights into the dilution phase of groundwater isotopic compositions on the eastern side of GGS during these months. Notably, GGS water 612 613 isotopes reached their heaviest form in August 2023, contrasting with other summer months 614 with similar temperatures and precipitation, suggesting lower inflow during that particular 615 month as simulated by the HGS model. This approach highlights the value of using isotopic 616 data to evaluate model simulations, providing a complementary angle to traditional methods. 617 Considering the higher precipitation and warmer temperatures in the summer of 2022 618 compared to the summer of 2023, the isotopic composition of both lake water and east





619 groundwater was heavier during the summer of 2022. This indicates that isotopic "light" rainwater, which directly recharged the lakes in the summer of 2022, was insufficient to 620 621 counterbalance the influence of the higher temperatures, resulting in a heavier isotopic 622 composition in the lakes. Despite the high precipitation in summer 2022, which would 623 typically dilute the isotopic signature of the lake, the higher temperatures (evaporation) led to 624 an enrichment of water isotopes. This phenomenon can be attributed to the large water body 625 of the lake (approximately 4 million m³ considering the average water depth of 6.5 m reported by Wolter and Ripl (1999) compared to the rain amount (0.05 million m³), where the 626 627 relatively small volume of rainwater was not enough to significantly alter the isotopic 628 composition of the much larger lake volume. This is in line with the findings of the study by 629 Vyse et al.2020 where they discovered a more significant influence of rain on shallow water 630 bodies compared to larger water bodies in the State of Brandenburg (NE-Germany). The low 631 impacts of rainwater on the isotope composition of the lake can also be interpreted to mean 632 that GGS is a groundwater-dominated lake with a very good hydraulic connection to the 633 groundwater, which keeps the lake water fresh and diluted by providing a source of 634 isotopically-depleted water. This underlines the robustness of the isotopic approach against 635 the variations of meteorological factors influencing the isotopic signature.

636 The isotopic differences between lake water and groundwater on both side of the GGS, highlight the complex interactions and distinct hydrological processes occurring within the 637 638 study area. Considering that the groundwater samples from the east and west of GGS belong 639 to the same depth (9-10 m below surface) and same aquifer, the isotopically-enriched 640 groundwater in the east side can only be explained by a well-mixed interaction of lake water 641 and groundwater in this area. This underlines that GGS is a flow-through lake, a conclusion 642 supported by the E/I ratio being up to 37% (e.i., mean ratio) in 2022 and 2023. The E/I ratio 643 of GGS aligns with the E/I ratio reported for wetlands and lake water bodies downstream of 644 the Spree catchment showing similar climate conditions. In 2021, the E/I ratio in this area was 645 up to 34% (Chen et 625 al., 2023). Vyse et al. (2020) reported that the wetlands with lower 646 landscape elevations located in northen Brandenburg typically possessed higher E/I ratios 647 than the ones with higher elevations. This is due to the hydrological function of small 648 waterbodies in the Pleistocene landscape. Higher wetlands have a more recharge/flow 649 through character whereas lower positions show a discharge character. Moreover, Cluett and 650 Thomas (2020) highlighted that the sensitivity of lake water isotopes to inflow and 651 evaporation can vary significantly over time, influenced by regional hydroclimate (e.g., direct 652 precipitation and humidity) and local hydrology (e.g., type of the lake). The uncertainty in





evaporative loss calculated using the code embedded in the Hydrocalculator is mainly due to uncertainties in the required inputs, temperature, and humidity, which can cause variations of up to 2% (E/I) according to Skrzypek et al. (2015). Despite these potential measurement variations, the calculated E/I ratio for GGS provides a reliable estimate that aligns with known hydrological behaviors in similar regions.

658 For the period 2015-2021, evaporation estimates derived from the isotope analysis (E/I) of 659 2022 and 2023 generally show lower values compared to the HGS model's evaporation 660 estimates, except 2018. This year saw a strong inflow peaking due to heavy rainfall events in the summer 2017. This suggests that using the E/I ratios from 2022 and 2023 for earlier years 661 without adjustments can lead to significant inaccuracies. The modified isotope analysis (E/I), 662 663 which incorporates annual variations in inflow and temperature, shows a better agreement 664 with the HGS model evaporation calculates, especially in the earlier years. This finding underscores the importance of including both temperature and inflow data for more accurate 665 evaporation predictions. Comparing these results with previous studies, Herbst and Kappen 666 (1999) reported evaporation from the Bornhöved Lake, which covers 1.1 km² in northern 667 668 Germany and has a maximum depth of 26 m, to be around 650 mm for the years 1992-1995. 669 The evaporation estimates for GGS from both the HGS model and the modified isotope 670 analysis fall within a similar range reported by Herbst and Kappen (1999), suggesting that 671 despite differences in geographical and hydrological characteristics, the annual evaporation 672 rates for German lowland lakes are comparable. These findings support the use of 673 comprehensive, multi-faceted approaches in hydrological studies to improve the precision of 674 evaporation estimates and enhance water resource management.

675

676 **5. Conclusion**

This study has addressed the challenge of accurate estimation of water balance components 677 678 comprehensively through the quantification of subsurface-groundwater inflow and 679 evaporation losses to Gross Glienicke Lake (GG), located in northeast Germany. Through the 680 combined use of the isotopic mass balance model, HydroCalculator, alongside the fully integrated hydrological model, HydroGeoSphere (HGS), a detailed understanding of the 681 682 hydrological dynamics governing GG Lake was attained. The calculated evaporation rates 683 derived from the isotopic mass balance model, exhibit strong alignment with the actual 684 evaporation rates calculated by the HGS model. This alignment underscores on one hand the 685 reliability and efficacy of the integrated hydrologic modeling approach in predictions of water





686 balance components such as inflow to the lake in a complex hydrogeological setting. On the 687 other hand, incorporating evaporation rate estimations given by isotope analysis corrected by 688 temperature variations and historical inflows leads to an improvement of the inflow results 689 even for the years without measured isotope data. Despite inherent uncertainties associated 690 with water isotope signature analyses, the integration of isotopic data with hydrological 691 modeling has provided valuable validation for the estimation of water balance components. 692 Moving forward, this integrated approach holds promise for enhancing the robustness of 693 hydrological models and facilitating more accurate assessments of water resources and 694 ecosystem dynamics in similar lake environments.

695

Data availability: All data (except the data provided by Berliner Wasserbetriebe) used to
process and set up the models will be available upon request. Data provided by Berliner
Wasserbetriebe can be requested through a separate usage agreement with Berliner
Wasserbetriebe.

Author contributions: Data collection, fieldwork, HGS model setup, and code development,
model input-output analysis, writing (original draft, review, and editing). JS: Isotopic
laboratory analysis, writing (review and editing). MS: Model input-output analysis, writing
(original draft, review, and editing). CM: Model input-output analysis, writing (original draft,
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