



1 **Integrating coupled surface–subsurface modeling and 2 field measurements: insights for rewetting a degraded 3 fen peatland**

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8
9 **Abstract.** Peatlands play a crucial role in regional water balance and carbon dynamics but are often degraded due
10 to drainage and agricultural use. In Germany, many drained peatlands have shifted from carbon sinks to CO₂ sources.
11 Rewetting these ecosystems is therefore essential to restore their ecological functions and mitigate greenhouse gas
12 emissions. However, effective rewetting requires a detailed understanding of peatland hydrology and its response
13 to climatic and management conditions. To address this need, this study employs a fully coupled surface–subsurface
14 hydrological model (HydroGeoSphere) to analyze the complex hydrological functioning of a typical degraded fen
15 peatland site (11.6 ha) in Brandenburg, Germany. The model-based quantification of hydrological fluxes is basis for
16 assessing peatland vulnerability to climate variability and land use while informing potential rewetting strategies
17 aimed at reducing CO₂ emissions. The studied peatland is connected to a regional aquifer and intensively drained by
18 a system of ditches. Simulations used daily meteorological inputs and detailed field measurements from 2015 to
19 2023. Evapotranspiration (ET) was parameterized using field-measured vegetation dynamics (seasonal leaf area
20 index and management schedules), while measured ditch water levels served as hydraulic boundary conditions. The
21 site was spatially divided into different management units with distinct vegetation parameters. The peat profile was
22 represented by two layers (a 0.3 m highly degraded surface peat overlying a 0.7 m less degraded layer) overlying
23 sand (aquifer) and till (aquifer base). The model was evaluated from different angles against eddy covariance ET and
24 groundwater table dynamics during a calibration period (2016–2020) and a validation period (2021–2023) using a
25 multi-metric approach. The model successfully reproduced seasonal water-table fluctuations and ditch–peatland
26 interactions, including ET-driven hydraulic gradient dynamics between summer and winter. Simulated ET closely
27 matched eddy covariance measurements, with RMSE values of 64 mm yr⁻¹, 10.2 mm month⁻¹, and 1.01 mm d⁻¹, and
28 showed only minor biases during dry conditions, while over the year seasonal dynamics of ET were also well captured
29 by the model. The model reproduced groundwater variations with sufficient accuracy, achieving KGE values of 0.80–
30 0.85, NSE of 0.83–0.86, and RMSE of 0.15 m during calibration and validation. The analysis of seasonal and
31 interannual water-storage changes showed pronounced shifts between hydrological surplus and deficit,
32 demonstrating that drained fens are highly sensitive to evapotranspiration demand and prolonged drought. The
33 modeling approach captured key hydrological processes with high robustness. The model's water balance analysis
34 provides an initial assessment of potential management measures, under the given climatic and hydrological
35 conditions, that could enable effective rewetting of the peatlands. These findings support ongoing peatland
36 restoration initiatives on drained peatlands in Europe.



37

38 **1. Introduction**

39 Peatlands rank among the most efficient terrestrial ecosystems for long-term carbon sequestration, holding nearly
40 one-third of global soil carbon despite representing only about 3% of the Earth's land surface. Around 10% of
41 Europe's land area, and approximately 5% of Germany's land area are covered by peatlands (Page et al., 2011; Xu et
42 al., 2018; Tanneberger et al., 2021; BMUV, 2021). In Germany, peatlands have stored carbon amounts comparable
43 to those in the country's forests. However, more than 90% of these peatlands have been drained, and over 95% are
44 now considered degraded, primarily due to agricultural land use, making them significant sources of greenhouse gas
45 emissions (Joosten & Tanneberger, 2017; Tanneberger et al., 2021). The regional hydrological effects caused by
46 peatland drainage are rarely quantified, despite their recognized importance for climate change mitigation and
47 adaptation strategies. These drained ecosystems currently release an estimated 53 million tons of CO₂-equivalents
48 annually, corresponding to roughly 7% of Germany's total anthropogenic emissions (BMUV, 2021). This hydrological
49 degradation not only amplifies greenhouse gas emissions but also disrupts essential ecological functions. At the
50 process level, biogeochemical cycles in peatlands are tightly linked to water-table dynamics which controls oxygen
51 availability and drives organic matter decomposition (Tfaily et al., 2013; Khaledi et al. 2024). In addition to the carbon
52 storage functions, peatlands play an essential role in regulating local and regional hydrological cycles. Peatlands in
53 Europe help to maintain water tables, and buffer hydrological extremes (Ahmad et al., 2021, Karimi et al., 2025).
54 They influence groundwater discharge and streamflow dynamics through their water-retentive capacity. Despite the
55 central role of water storage dynamics in controlling peatland functioning, quantitative assessments of seasonal and
56 interannual storage change in peatlands remain rare (Bourgault et al., 2017). Most studies focus on water-table
57 variability or individual flux components, while the full water-balance partitioning, including lateral exchanges with
58 ditches and groundwater systems, has received limited attention. The specific microclimatic conditions, organic-rich
59 soils and high water tables within peatlands are the basis for a specific flora and fauna (Millar et al., 2018; Li et al.,
60 2019). Besides drainage and land use, climate change, including rising temperatures and shifting precipitation
61 patterns intensified peatland vulnerability. Simulations by McLaughlin et al. (2021) indicate that under severe
62 warming scenarios, peatlands may lose a significant portion of their carbon sink capacity due to drying and enhanced
63 peat decomposition. Swindles et al. (2025) demonstrate that higher summer temperatures across European
64 peatlands increase decomposition and carbon release unless water tables are maintained near the surface, further
65 highlighting the vulnerability of these systems under warming and water scarcity. Restoring the natural hydrological
66 function of peatlands through rewetting has therefore become a central goal of environmental policy and research
67 (Ekardt et al., 2020; Chen et al., 2023; Meyer-Jürshof et al., 2025).

68 Although climate variables, particularly precipitation, temperature and solar radiation (through evapotranspiration),
69 are primary drivers of peatland hydrology, understanding and managing the hydrological behavior of drained and
70 rewetted peatlands requires accurate, spatially explicit representations of surface–subsurface water interactions
71 and boundary conditions of the water balance. However, modeling such ecosystems remains challenging due to
72 several factors such as the high spatial variability in vegetation, water management and anthropogenic interventions
73 (drainage, pumping, and water level regulators such as weir), complex feedback between evapotranspiration and
74 shallow groundwater, seasonally shifting ditch–peatland interactions, and soil heterogeneity with compacted
75 organic layers over mineral substrates. Recent lysimeter modeling in degraded peat has shown that ignoring such
76 heterogeneity (e.g., by using unimodal parameterizations) leads to overestimation of water storage and damped



77 water-table dynamics, whereas dual-porosity representations better capture observed fluctuations (Davies et al.,
78 2024). Modelling tools are essential for disentangling these drivers. Many previous studies have used statistical or
79 conceptual models to relate water levels to meteorological variables (e.g., Okkonen & Klöve, 2010; Ballard, et al.,
80 2011; Binet et al., 2013; Bertrand et al., 2021). While these models can predict typical seasonal fluctuations, they
81 struggle to simulate extreme droughts or capture feedback among vegetation, soil hydraulics and surface water.
82 Traditional physically based hydrological models often decouple surface and subsurface processes or oversimplify
83 vegetation–atmosphere interactions, limiting their ability to simulate water dynamics in such complex settings
84 (Paniconi & Putti, 2015). Hence, fully coupled surface–subsurface models are necessary to resolve the complex
85 interactions governing peatland hydrology.

86 HydroGeoSphere (HGS) is a fully integrated, physically based model that simultaneously solves the
87 three-dimensional Richards equation for variably saturated subsurface flow together with two-dimensional surface
88 flow equations, enabling explicit representation of vertical heterogeneity, lateral exchanges with rivers or ditches,
89 and spatially variable vegetation and management practices (Ala-aho et al., 2017, Hwang et al., 2018). Furthermore,
90 HGS enables spatially distributed parameterization of vegetation and land cover, making it particularly suitable for
91 peatlands where agricultural management practices such as mowing or grazing strongly influence
92 evapotranspiration and soil moisture dynamics. Despite its strengths, HGS remains underutilized in peatland studies
93 as its application requires long-term hydroclimatic data and high-resolution land use information for validation.
94 Recently, Renaud et al. (2025), has used a 2D cross-sectional transect to model water-table dynamics, which
95 simplified the computational setup but inherently neglected three-dimensional fluxes parallel to the river.
96 Moreover, vegetation parameters such as leaf-area index and root depth were not measured on site but derived
97 from literature, so the model did not capture dynamic vegetation growth or land-management variations. By
98 addressing these limitations through field-measured vegetation data and a fully parameterized 3D model setup, this
99 study offers a more realistic simulation of peatland hydrological processes.

100 This study addresses a hydrologic model setup that can be used for a robust assessment of possible rewetting
101 measures under given climatic and hydrological conditions in a typical degraded fen peatland in Brandenburg,
102 Germany. Our objectives are (i) providing optimized parametrization for a better understanding of peatland
103 hydrological processes under different climate conditions and land management pressures (ii) considering all
104 relevant water fluxes between groundwater, surface water and atmosphere within a fully-integrated modeling, and
105 (iii) quantification of the specific contributions of precipitation, evapotranspiration, inflow, outflow and water
106 storage change to the water balance of the peatland site. The insights support the design and adaptation of effective
107 rewetting strategies in the context of a sustainable preservation of peatlands in the future.

108

109 **2. Methodology**

110 **2.1. Case Study**

111 The study site represents a degraded fen peatland located near Paulinenaue in the federal state of Brandenburg,
112 northeastern Germany (Fig. 1a). The peatland, including its surrounding drainage ditches, covers an area of

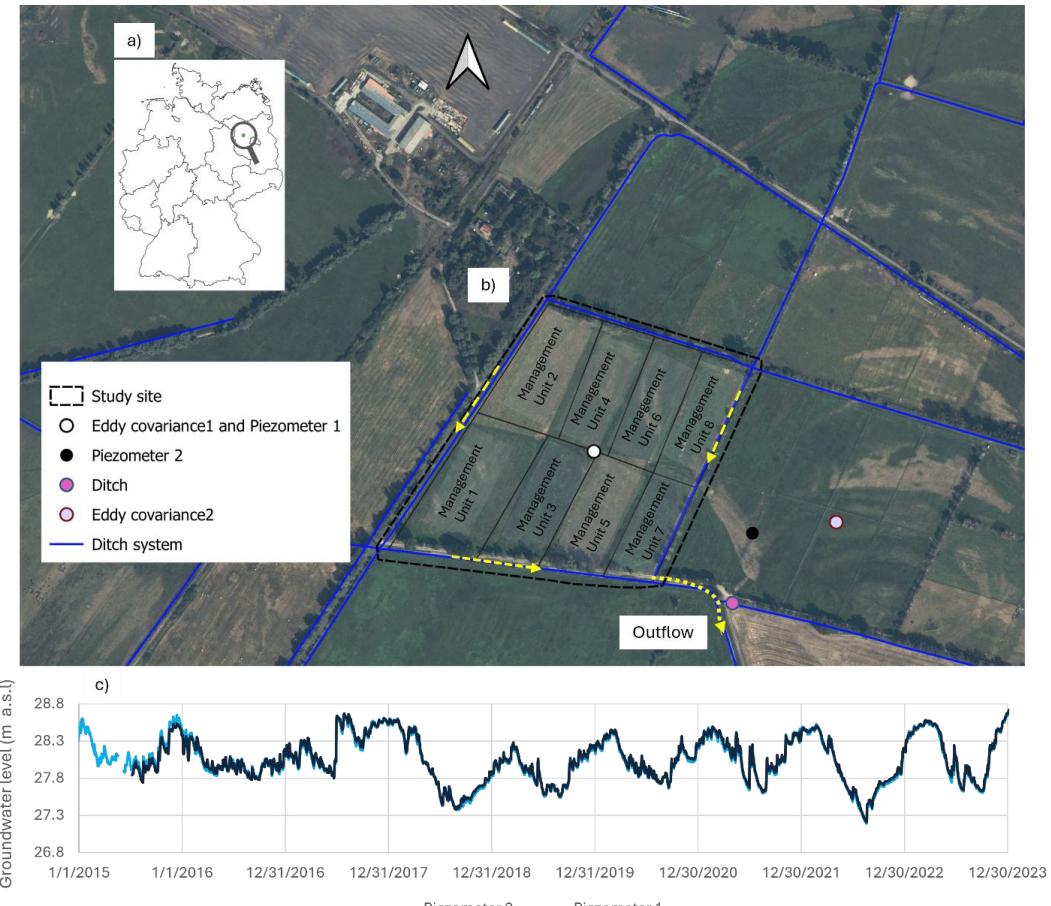


113 approximately 11.6 ha. It has been extensively drained through a network of ditches that maintain water levels below
114 the surface during the vegetation period for agricultural grassland production.

115 The land cover is permanent grassland, composed primarily of a few grass species such *Phalaris arundinacea*, *Elymus*
116 *repens*, *Alopecurus pratensis*, *Poa* spp., and *Juncus* spp., which were identified through botanical surveys. Vegetation
117 dynamics and management patterns were characterized by in-situ Leaf Area Index (LAI) measurements collected
118 seasonally from 2015 to 2020. The peatland is divided into eight management units, each with distinct mowing and
119 grazing schedules (as an example, see Table S1 in appendix, showing the management units and plant growth stages
120 in year 2015). Groundwater levels inside and outside the study site have been recorded daily since 2015.
121 Observations suggest that the peatland site and its surrounding groundwater are hydraulically connected to a larger
122 regional aquifer system (Fig 1c). An eddy covariance station in the middle of the study site has been operational since
123 2014 to measure land-atmosphere exchanges, including evapotranspiration (Fig. 1b).

124 At the surface, the site is covered by a peat sequence consisting of a more compact, strongly decomposed upper
125 layer underlain by a thicker layer of less degraded peat, both of which play a central role in regulating water storage
126 and flow (Fig. 3). Beneath the peat, the topsoil consists of silt to fine sand, followed by a saturated zone of medium
127 to coarse sand with an average thickness of about 10 m. According to the Brandenburg geological database
128 (Geologischer Dienst, Landesamt für Bergbau, Geologie und Rohstoffe Brandenburg), the geological base (bedrock)
129 is composed of till with clay lenses, which forms the lower hydrological boundary (Fig. 5).

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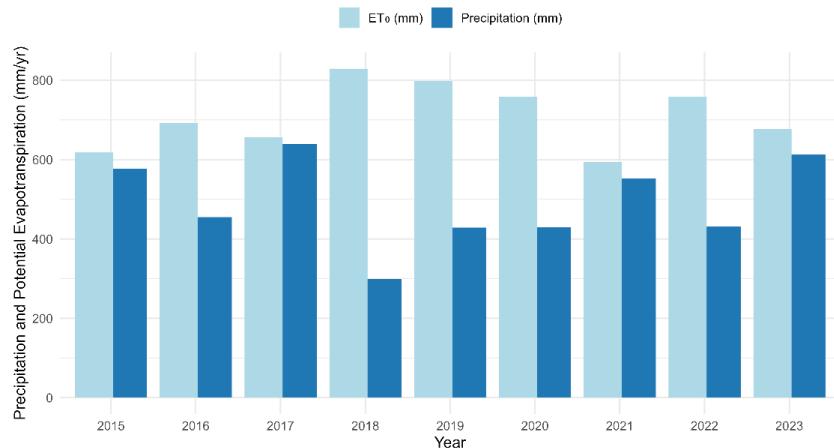
132 **Fig. 1.** Location of the study site in northeastern Germany (a), and positions of the eddy covariance towers,
133 piezometers, and ditch water level measurement within the site, and the order of management units (b),
134 groundwater fluctuations in and outside of the study site since 2015 (c). Background satellite imagery: © Microsoft
135 Bing Maps.

136

137 Daily values of precipitation, net radiation, soil heat flux, temperature, wind speed, and relative air humidity were
138 measured as part of the station eddy covariance2 at the study site (Fig.1b, eastern site). The data were used for
139 calculating the potential evapotranspiration (ET_0) with the FAO Penman–Monteith method (Allen et al., 1998).
140 Annual precipitation over the study period (2015–2023) ranged from a minimum of 299 mm in 2018 to a maximum
141 of 640 mm in 2017 (Fig. 2). Annual ET_0 averaged is roughly 710 mm, with the highest value of 829 mm in 2018 and
142 the lowest of 594 mm in 2021 (Fig. 2). The pronounced water deficit in dry years (e.g., 2018, 2019, 2020, and 2022)
143 reflects meteorological drought conditions where ET_0 substantially exceeded precipitation during the growing
144 season. Such deficits are known to promote peat desiccation, increase oxidation, and accelerate carbon losses. These
145 climatic patterns are particularly critical for the hydrology of drained peatlands, as summer deficits exacerbate
146 groundwater drawdown.

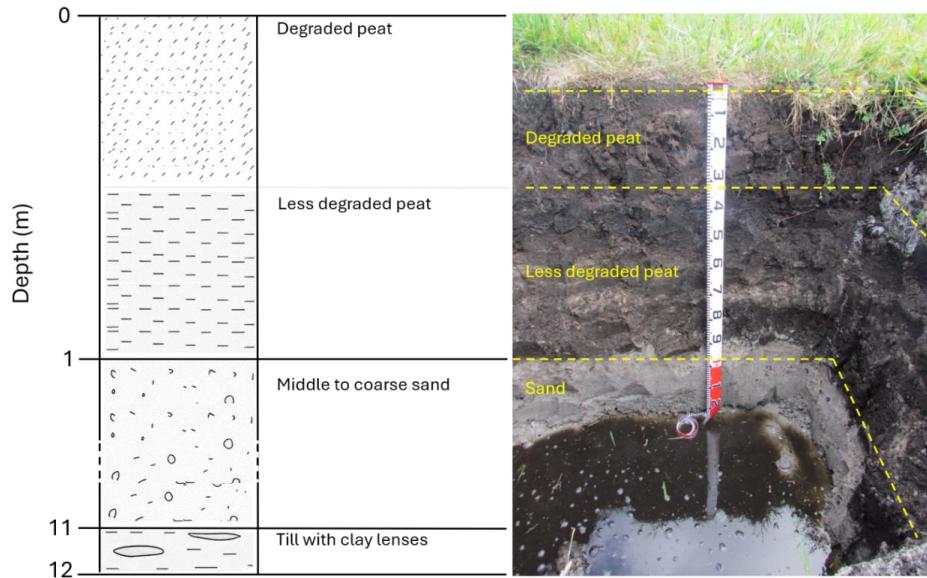


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149 **Fig. 2.** Annual precipitation and potential evapotranspiration (ET₀) highlighting water deficit.



150

151 **Fig. 3.** Cross-section of the model domain showing the main geological layers (Photo: © Axel Behrendt).

152

153 **2.2. Eddy Covariance Tower**

154 The eddy station on the study site (eddy covariance1 in Fig. 1b) is equipped with an Irgason CO₂/H₂O open path
155 analyser (Campbell) for measuring the wind components and the water content of the air with a frequency of 10 Hz.
156 A CNR4 sensor (Kipp & Zonen) measured the shortwave and longwave downwell and upwell radiation components



157 so that the net radiation can be calculated as the difference between the components. Three soil heat flux plates
158 (Huxeflux) measured the ground heat flux at 8 cm depth. The heat stored in the layer above the heat flux plates was
159 calculated on the basis of the measured soil temperature and volumetric water content above the plates and soil
160 properties (organic matter content, volumetric heat capacity) by the approach of de Vries (1963) described in
161 Liebenthal and Foken (2006). All data were stored on a CR3000 data logger (Campbell).

162 The software package TK3 was used for the post-processing of the 10 Hz raw data (Mauder and Foken, 2015). All
163 data were checked for their quality and validity following the Foken evaluation system (Foken and Wichura, 1996,
164 Foken, 2008). The data were aggregated to 30-minutes values. Only data sets fulfilling the quality requirements were
165 used in the following steps of the data processing. The gap of the energy balance as sum of net radiation, soil heat
166 flux, latent and sensible heat flux was closed using the Bowen ratio method (Foken, 2008; Foken et al, 2006; Mauder
167 et al., 2018; Twine et al., 2000). At the end of the post-processing the data were aggregated to daily values and the
168 actual evapotranspiration was calculated based on the latent heat flux. Data gaps were filled with data from a
169 neighboring station (eddy covariance2 in Fig. 1b). The data of the neighboring station were processed in the same
170 way as the data of station 1.

171 **2.3. Leaf Area Index (LAI)**

172 Leaf Area Index (LAI) measurements were collected in the study area using a SunScan Canopy Analysis System (Delta-
173 T Devices, UK) at multiple locations within each of the eight management units. Measurements were taken across
174 different months and seasons between 2015 and 2020 to capture seasonal vegetation dynamics and the influence
175 of distinct management practices such as mowing and grazing. The resulting LAI time series provided spatially and
176 temporally distributed information on canopy development, which was directly integrated into the HydroGeoSphere
177 model to parameterize vegetation-dependent processes such as evapotranspiration and interception. LAI is a key
178 variable in peatland hydrological modeling because it governs canopy resistance, influences transpiration rates, and
179 modulates the partitioning of energy fluxes between latent and sensible heat. For the simulation period 2021–2023,
180 when direct LAI measurements were unavailable, management schedules and seasonal patterns from the
181 corresponding earlier years were assumed, based on consistent agricultural practices at the site.

182 **2.4. Hydrologic Numerical Model**

183 The HydroGeoSphere (HGS) model was used to develop a fully integrated, physically based hydrological model of the
184 fen peatland. HGS solves the three-dimensional Richards equation for variably saturated porous media together with
185 the diffusion-wave approximation of the Saint-Venant equations for surface water flow, using a control-volume finite
186 element approach (Therrien et al., 2010). This integration allows a dynamic representation of surface–subsurface
187 exchanges, interception, infiltration, evapotranspiration, and lateral groundwater–ditch fluxes. The model domain
188 represents the degraded fen peatland and its surrounding drainage ditches (Fig. 5). The vertical domain extends to a
189 depth of nearly 12 m, representing three geological layers: (i) two peat layers showing different degrees of peat
190 degradation (around 1 m), (ii) a middle to coarse sand layer (saturated/unsaturated), and (iii) glacial till with clay
191 lenses as the geological base. The model mesh consisted of 2,490 nodes and 4,523 triangular elements with refined
192 vertical discretization near the surface. Mesh quality indicators confirmed robustness, with aspect ratios (mean 1.6),
193 edge ratios (mean 1.2), minimum internal angle (mean 30.8°, SD 2.6°) and maximum angle (mean 70.0°, SD 4.7°).
194 Avoiding high aspect ratio and extreme angles is critical for numerical convergence in finite element modeling. Along
195 the land surface, daily precipitation and potential evapotranspiration (ET₀) were applied as Neumann fluxes. Along



196 the perimeter of the model domain, observed ditch water levels were imposed as fixed head (Dirichlet boundary) to
197 represent the regionally controlled drainage network. A no-flow condition was applied at the till–clay base, which
198 acts as the lower geological boundary. To establish realistic initial states, the year 2015 was used as a spin-up phase
199 to bring the model to hydrostatic equilibrium before the analysis period starts.

200 **2.5. Hydraulic Properties of Peat Layers**

201 Peat soils' hydraulic conditions differ markedly from mineral soils due to their high organic content, large pore space,
202 and variable degree of decomposition. Intact peat typically has very high porosity and strong water-holding capacity,
203 but drainage and agriculture use cause compaction and oxidation, which reduce porosity and reduced saturated
204 hydraulic conductivity. The soil profile in Paulinenaue is represented by two peat layers with contrasting degrees of
205 degradation (Fig. 3). The upper degraded peat (≈ 30 cm depth) was parameterized with a lower porosity ($\theta_s = 0.70$)
206 and reduced saturated conductivity ($K_x = K_y = 0.5 \text{ m d}^{-1}$; $K_z = 0.1 \text{ m d}^{-1}$), reflecting compaction and reduced vertical
207 flow pathways (Fig. 3). A relatively steep van Genuchten α (1.5 m^{-1}) was applied to capture its rapid desaturation
208 when the water table drops. In contrast, the less degraded lower peat layer was assigned a higher porosity ($\theta_s =$
209 0.80) and higher conductivity ($K_x = K_y = 3 \text{ m d}^{-1}$; $K_z = 0.3 \text{ m d}^{-1}$), consistent with its fibrous structure and larger pores.
210 The van Genuchten α (0.75 m^{-1}) and β (1.7) values represent its higher water retention capacity, keeping water
211 available under moderate suctions (Fig. 4, ranges based on Wallor et al., 2018a, and Wallor et al., 2018b; Liu, H. and
212 Lennartz, B., 2019; Renaud et al., 2025).

213

214 Table 1. Boundary and initial conditions applied in the HydroGeoSphere model.

Boundary / Condition	Representation in Model	Description / Notes
<i>Top boundary</i>	<i>Neumann flux</i>	<i>Daily precipitation and ET_o (from DWD station)</i>
<i>Ditches (lateral)</i>	<i>Dirichlet boundary (fixed head)</i>	<i>Observed ditch water levels applied around perimeter</i>
<i>Bottom boundary</i>	<i>No-flow</i>	<i>Till/clay base considered impermeable</i>
<i>Spin-up / Initialization</i>	<i>Hydrostatic equilibrium (year 2015)</i>	<i>One warm-up used before analysis period</i>

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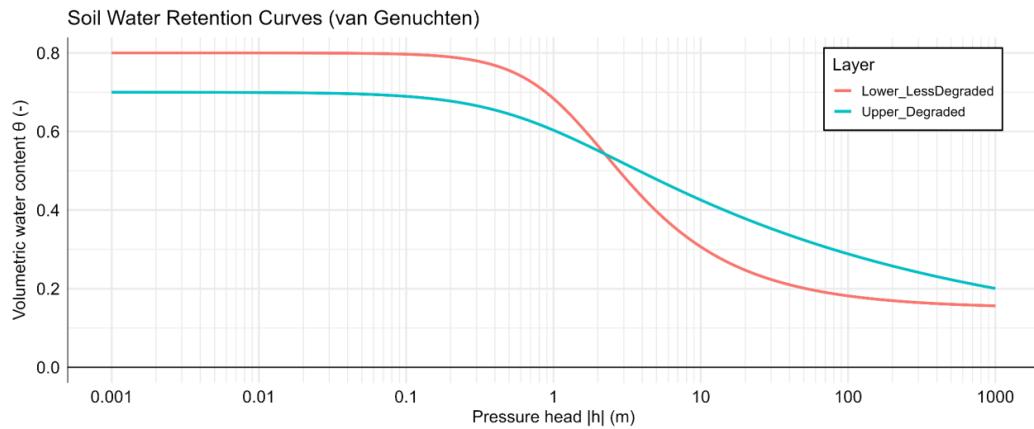
216 Table 2. Peat hydraulic properties



Layer	K (m d^{-1})		$\Theta_s [\text{m}^3 \text{m}^{-3}]$	$\alpha (\text{m}^{-1})$	$\beta (-)$	Residual Saturation
	$K_{x,y}$	K_z				
<i>Upper degraded peat</i>	0.5	0.1	0.7	1.5	1.2	0.05
<i>Lower less degraded peat</i>	3	0.3	0.8	0.75	1.7	0.15

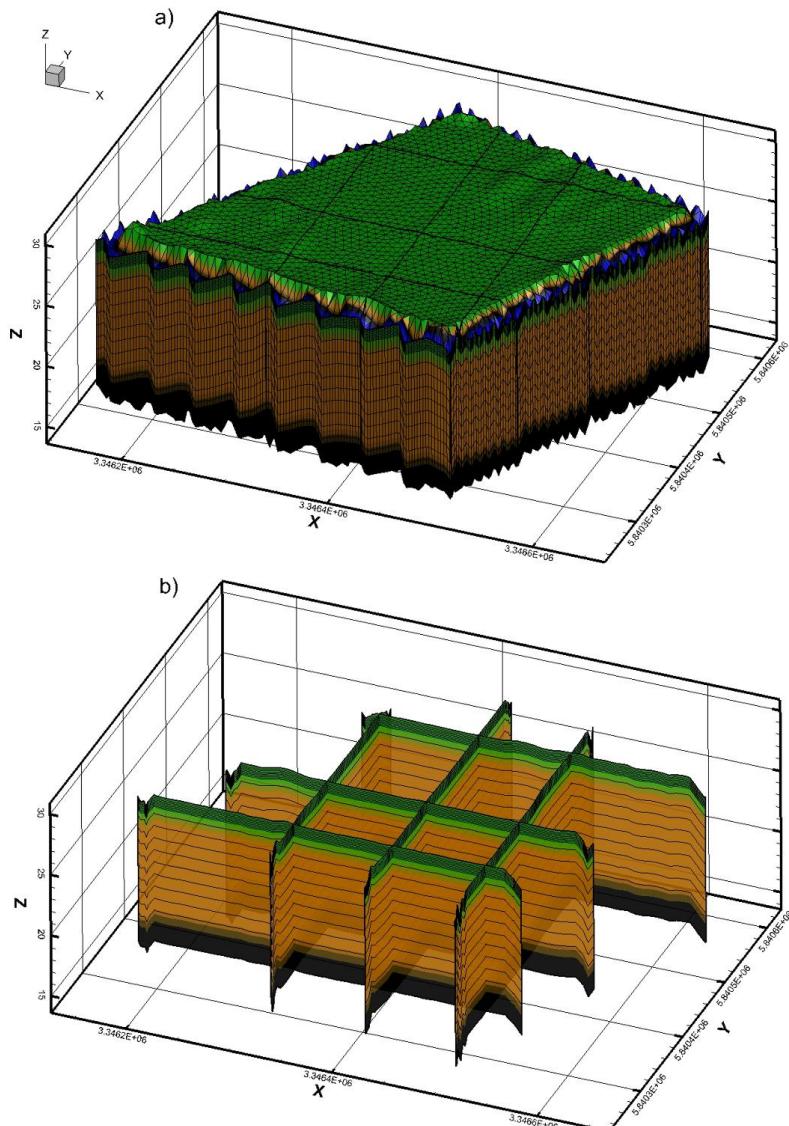
217
218 • K: Hydraulic conductivity
219 • Θ_s : Porosity
220 • α : Van Genuchten parameter
 • β : Van Genuchten parameter

221



222

223 **Fig. 4.** Soil water retention curves for upper degraded peat ($\alpha = 1.5$, $\beta = 1.2$) and lower less degraded peat ($\alpha = 0.75$,
224 $\beta = 1.7$).



225

226 **Fig. 5.** 3D view of the HydroGeoSphere model domain. (a) Subset of the irregular triangular surface mesh
227 representing the peatland topography, with colors indicating layers of different hydraulic properties. (b) Cross-
228 sectional view of the model layers illustrating the internal mesh and geological structure.

229 **2.6. Evapotranspiration Parameterization**

230 Evapotranspiration in the model is represented through parameterizations for transpiration, evaporation, and
231 interception processes. Transpiration was governed by three fitting constants (C1–C3) that control the ratio of actual
232 to potential evapotranspiration (AET/ET₀) following the Kristensen & Jensen (1975) framework. With C1 = 0.4 and C2
233 = 0.10, AET/ET₀ increases linearly with LAI from a baseline of 0.10 and reaches 1.0 at LAI ≈ 2.3–2.5. Although the
234 observed seasonal maximum LAI in the fen grassland is around 7, this parameterization reflects the assumption that



235 full ET_0 is typically achieved at moderate canopy densities and further increases in LAI do not substantially raise
236 transpiration. $C_3 = 2.0$ defines the soil moisture stress response, producing a gradual increase in AET/ET_0 between
237 the wilting point ($\theta = 0.1$) and field capacity ($\theta = 0.4$), so that transpiration is reduced under drought but recovers
238 smoothly after rewetting (Figure 6). Transpiration limiting saturations were set to 0.1 (wilting point), 0.4 (field
239 capacity), 0.95 (oxic limit), and 1.0 (anoxic limit), reflecting the ability of fen grasses to remain physiologically active
240 in wet soils but to become constrained under both drought and oxygen stress. Evaporation limiting saturations were
241 0.1 (minimum) and 0.4 (maximum). Thus, evaporation is suppressed under dry conditions but occurs at full potential
242 once the upper soil layer contains moisture. An evaporation depth of 0.40 m was specified, representing the active
243 soil layer contributing to evaporation, while root uptake was parameterized with a maximum depth of 0.40 m and a
244 quadratic decay function to simulate dense surface rooting and sparse rooting at depth. A canopy storage parameter
245 of 0.01 and initial interception storage of 0.0 were included to account for short-term retention of precipitation on
246 plant surfaces. This parameterization ensures that the model captures the interaction between vegetation structure,
247 soil water availability, and atmospheric demand, while distinguishing between vegetation-driven transpiration, soil
248 evaporation, and open-water evaporation from ditches.

249

250 Table 3. Evapotranspiration parameterization used in the HydroGeoSphere model.

Parameter	Value (unit)	Notes / Role in Model
Transpiration fitting constants		
C_1	0.4	Controls slope of AET/ET_0 increase with LAI
C_2	0.1	Baseline evaporation fraction, present even at low LAI
C_3	2.0	Soil-moisture stress sensitivity
Transpiration limiting saturations		
Wilting point	0.1	Below this, transpiration ceases
Field capacity	0.4	Maximum transpiration occurs between FC and oxic limit
Oxic limit	0.95	Plants remain active under near-saturated conditions
Anoxic limit	1.0	Plants inactive due to oxygen stress
Evaporation limiting saturations		
Minimum saturation (e_s)	0.1	Below this, evaporation ceases
Maximum saturation (e_s)	0.4	Above this, full evaporation occurs

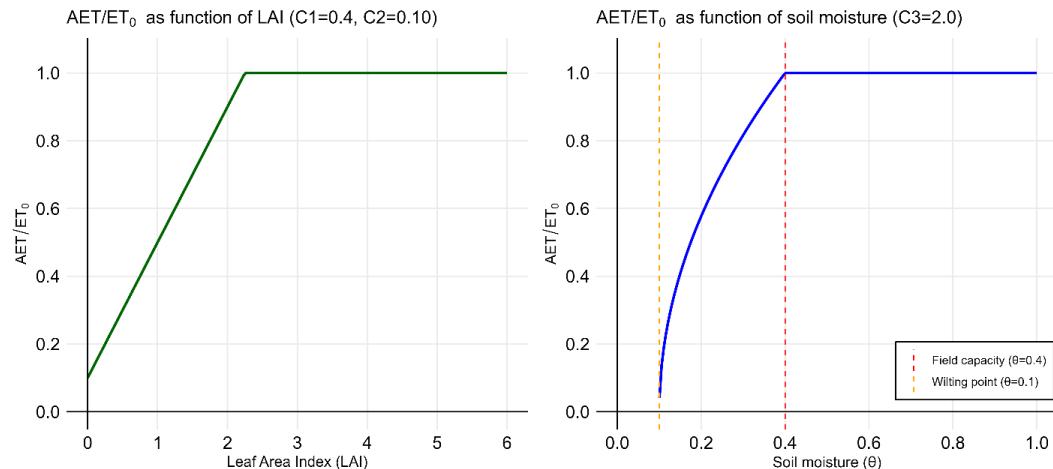


<i>Evaporation depth</i>	0.40 (m)	<i>Depth of soil layer contributing to evaporation</i>
Root uptake		
<i>Root depth</i>	0.40 (m)	<i>Maximum rooting depth of fen grassland</i>
<i>Root distribution</i>	<i>Quadratic decay</i>	<i>Denser roots near surface, fewer roots at depth</i>
Canopy & interception		
<i>Canopy storage parameter</i>	0.01	<i>Maximum temporary canopy water storage</i>
<i>Initial interception storage</i>	0.0	<i>No initial interception storage</i>

251
 252 • LAI: Leaf Area Index
 253 • ET: Evapotranspiration
 254 • AET: Actual Evaporation
 255 • ET_0 : Potential Evaporation
 • FC: Field Capacity

256

257



258
 259 **Fig. 6.** Reduction functions for AET/ ET_0 used in HydroGeoSphere with C1 = 0.4, C2 = 0.10 and C3 = 2.0. Left: effect of
 260 LAI; Right: soil-moisture stress between wilting point ($\theta = 0.1$) and field capacity ($\theta = 0.4$).

261

262 **2.7. Water Storage Changes**

263 HGS applies evapotranspiration as a fully distributed process, with contributions from surface and canopy
 264 evaporation as well as subsurface evaporation and transpiration acting on all nodal layers within the specified



265 evaporation and rooting depths. Because AET is implemented across multiple depth intervals, the resulting water
266 balance cannot be accurately derived from surface fluxes alone. Therefore, a volumetric, node-based accounting
267 approach was applied to quantify all inflow, outflow, and storage terms consistently across the 3D model domain.
268 Storage change (ΔS) was obtained from the nodal mass-balance output, which reports changes in fluid mass within
269 each control volume at each time step. It is calculated as follows:

270
$$\Delta S = PCP + HI - AET - HO - Resid \quad (1)$$

271 PCP: Precipitation

272 AET: Evapotranspiration

273 HI: Horizontal inflows

274 HO: Horizontal outflows

275 Resid: Nonlinear Newton residual

276 The sum of all vertical inflow and outflows between all layers is zero. Cumulative storage changes were then
277 calculated from these time-step values and plotted to evaluate seasonal and interannual storage dynamics. The
278 volumetric fluxes were divided by the total contributing nodal-control-volume area to provide depth-equivalent units
279 (i.e., mm day⁻¹).

280 **3. Results**

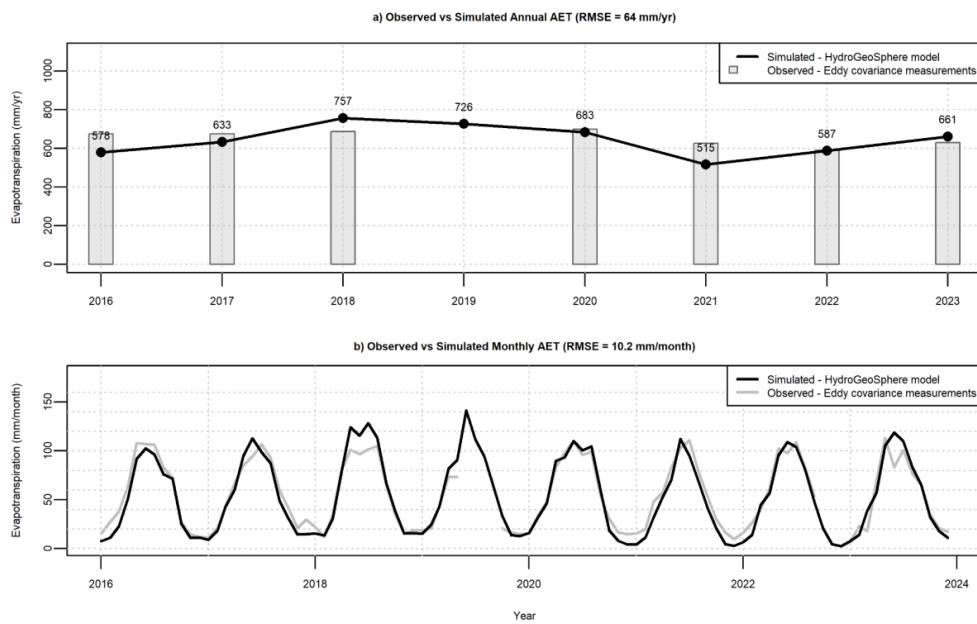
281 **3.1. Actual Evapotranspiration**

282 Figures 7a,b compares simulated AET from the HydroGeoSphere (HGS) model with observations derived from eddy
283 covariance measurements for the period 2015–2023 (excluding 2019 due to data gaps). At the annual scale, the
284 model captures the interannual variability of AET reasonably well, with simulated values ranging between 515 and
285 757 mm yr⁻¹ compared to measured values of 589–720 mm yr⁻¹. Years with relatively high and low observed
286 evapotranspiration (e.g., 2020 and 2022) are also reflected in the model output, resulting in a root mean square
287 error (RMSE) of about 64 mm yr⁻¹. At the monthly scale, the model reproduces both the magnitude and seasonal
288 dynamics of AET, closely tracking the observed annual cycles with higher evapotranspiration during the growing
289 season (May–September) and lower values during winter months. The agreement is strong (RMSE = 10.2 mm
290 month⁻¹), indicating that the model adequately represents both seasonal variability and interannual differences in
291 evapotranspiration, although slight underestimations are apparent in some autumn and winter months (e.g., year
292 2021). Notably, during the extremely dry year 2018 (annual precipitation around 300 mm yr⁻¹), the largest
293 discrepancies occurred in the summer months, with the model overestimating evapotranspiration relative to the
294 eddy covariance observations.

295 Figure 8 shows model results of monthly variations of main AET components which are transpiration and evaporation
296 (from soil and canopy storage) for 2016–2023. Both components are lowest in winter and increase through spring,
297 peaking in summer. Transpiration dominates the growing season, often reaching 80–100 mm month⁻¹ in June–
298 August, whereas evaporation is smaller (about 5–30 mm month⁻¹) with short pulses after wet periods. Interannual
299 variability is evident in the amplitude and timing of both components. The strongest summer transpiration occurs in

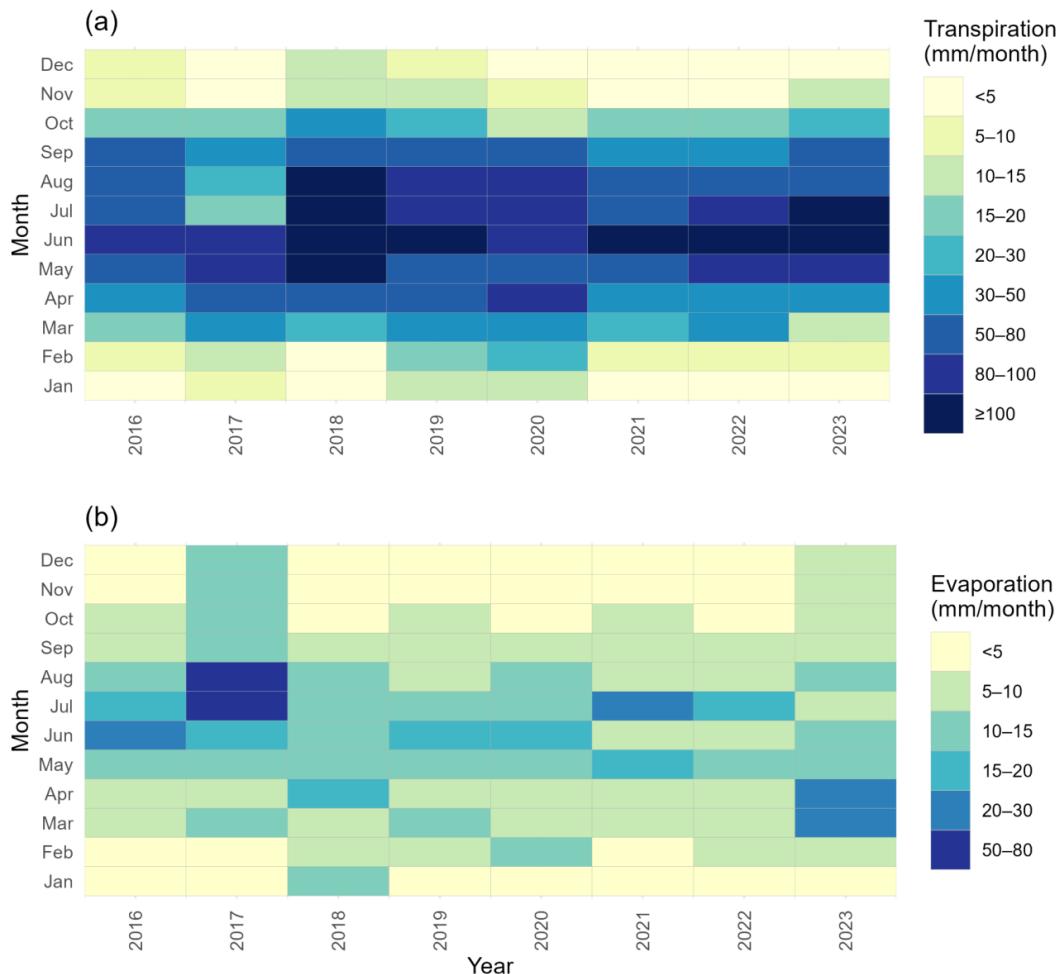


300 2018, consistent with warm, dry conditions and high canopy demand. The high precipitation in June and July of 2017
301 and winter of 2018 increased water availability into the following growing season and supported canopy
302 development, with LAI around 7, yielding the highest transpiration rates in summer 2018. In contrast, 2021 shows
303 subdued transpiration for much of the year and several moderate evaporation pulses, indicating wetter surface
304 conditions. Evaporation peaks are most pronounced in early to mid-summer 2017 and in July 2021–2022, reflecting
305 precipitation events and enhanced soil or canopy wetness. In 2017 precipitation totals were high and the water table
306 often reached or exceeded the surface, indicating saturated peatland conditions; these conditions limited root
307 aeration and reduced transpiration, while sustained soil and canopy wetness enhanced evaporation. Overall, actual
308 evapotranspiration is transpiration-dominated in summer, while evaporation contributes a larger share during cooler
309 or wetter months and immediately following rainfall.



310
311 **Fig. 7.** Comparison of simulated and observed actual evapotranspiration (AET) using HydroGeoSphere calculations
312 and eddy covariance measurements: (a) annual values (2019 excluded due to missing observations), and (b) seasonal
313 variability based on monthly data.

314



315

316 **Fig. 8.** Seasonal dynamic of actual evapotranspiration components, (a) Transpiration, (b) Evaporation
 317 from soil water and canopy storage (interception).

318

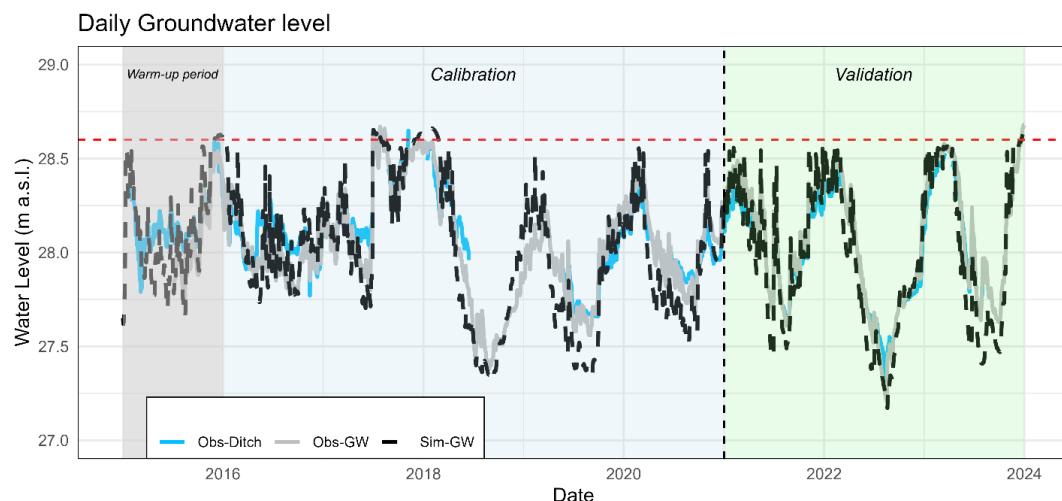
319 **3.2. Groundwater Level**

320 The HydroGeoSphere model successfully reproduced the daily dynamics of groundwater levels in the Paulinenaue
 321 fen site for the period 2015–2023 (Fig. 9). Simulated water levels closely followed observed groundwater levels in
 322 the central piezometer, capturing both seasonal fluctuations and interannual variability at daily resolution. The fit is
 323 strong, with model performance metrics of $NSE = 0.83$, $KGE = 0.80$, and $RMSE = 0.15$ m during calibration period
 324 (2016–2020), and $NSE = 0.86$, $KGE = 0.85$, and $RMSE = 0.15$ m, while during validation (2021–2023). Observed ditch
 325 water levels were consistently higher than simulated peatland groundwater levels during summer months and
 326 slightly lower during winter, reflecting the seasonal inversion of hydraulic gradients between the ditch and peatland
 327 aquifer. This pattern was well reproduced by the model, which captured both the summer drawdown of the



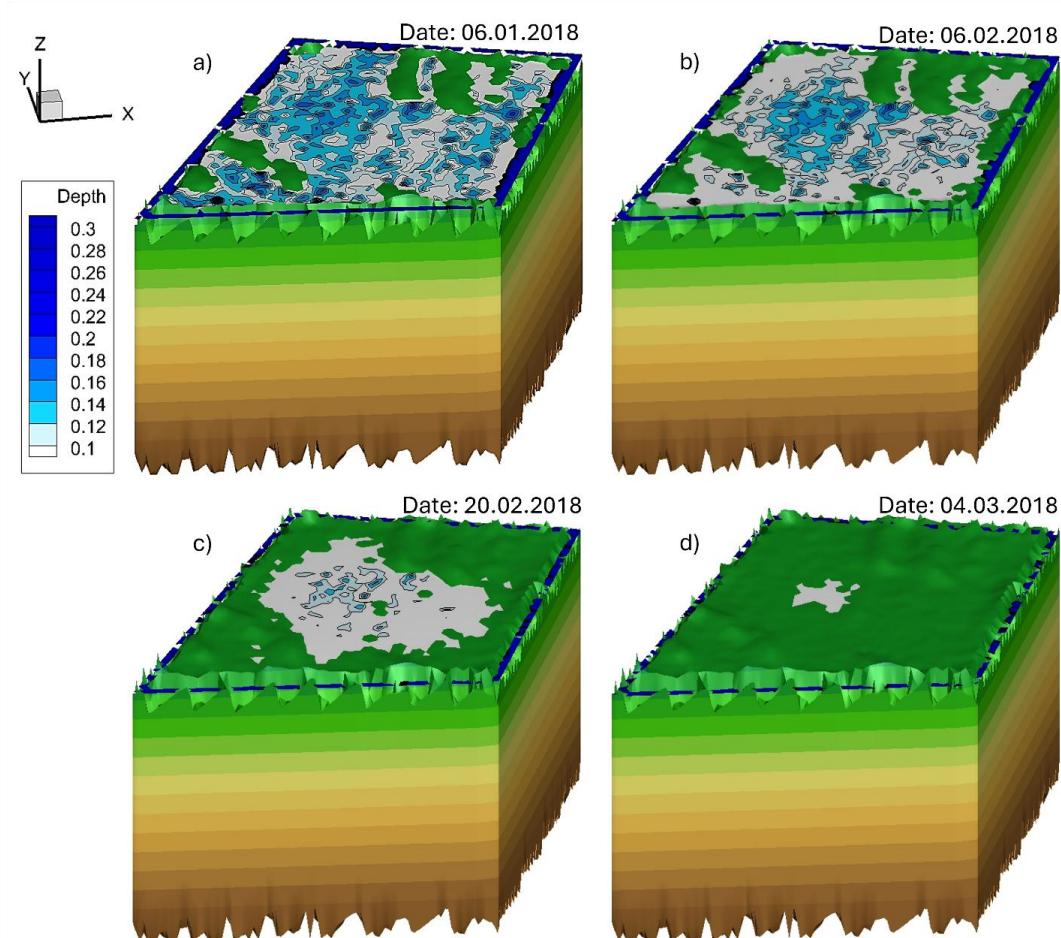
328 groundwater table caused by high evapotranspiration and the winter recharge-driven rise. In addition, during the
329 flood years in summer 2017, in winter 2022, and in winter 2023, when water levels rose above the peat surface, the
330 model correctly reproduced the occurrence of surface inundation, confirming its ability to simulate both
331 groundwater and overland flow processes Conversely, in summers where ditch water level data were missing (e.g.,
332 2018), the simulated groundwater level deviated from observations, underlining the system's sensitivity to accurate
333 ditch boundary conditions. This sensitivity is further illustrated in summer 2021, when two consecutive groundwater
334 drawdowns occurred: the one with observed ditch data was captured better by the model than the one without
335 ditch data. These results demonstrate that the parameterization of evapotranspiration, ditch boundary conditions,
336 and soil–surface coupling adequately represents the main hydrological controls on groundwater and surface water
337 dynamics in this drained fen system. To demonstrate that model performance extends beyond the observation well,
338 Fig. 10 shows the simulated spatial dynamics of surface inundation across the peatland during winter 2018. It
339 captures the progression from widespread inundation in January (Fig. 10a) to its disappearance by March (Fig. 10b),
340 confirming the model's ability to reproduce site-wide hydrological responses including the extent, depth, and
341 temporal evolution of surface inundation, not just point-scale fluctuations.

342



343

344 **Fig. 9.** Observed and modeled dynamics of groundwater and ditch water levels in the Paulinenauen fen site for the
345 period 2015–2023.



346

347 **Fig. 10.** Simulated water depth (m) dynamics on top of the peatland during winter 2018, showing gradual reduction
348 of surface water extent and depth from (a) 06 January to (d) 04 March. The sequence illustrates initial flooding and
349 the slow drop in water level over time.

350

351 **Table 4.** Model performance in simulating groundwater dynamic for calibration (2016-2020) and validation period
352 (2021-2023).

Performance measures	Calibration (2016-2020)	Validation (2021-2023)
KGE	0.80	0.85
NSE	0.83	0.86
RMSE	0.15	0.15

353



354 3.3. Water Fluxes

355 The year 2017 represents a wet year (precipitation 640 mm year^{-1}) and the year 2018 an extremely dry year
356 (precipitation 299 mm year^{-1}) for the study region. Daily water fluxes (inflow, outflow, precipitation, and actual
357 evapotranspiration) from 1st January 2017 to 31st December 2018 are shown in Figure 11. Daily inflow and outflow
358 show event-driven dynamics, with sharp outflow pulses after rainfall, especially in autumn and winter when
359 evapotranspiration is low (Fig. 11a). Inflow occurs mainly during summer, when AET-driven water deficits create
360 hydraulic gradients that generate flow from the ditches and adjacent aquifer into the peatland site. Outflow is higher
361 and more frequent in 2017 and smaller in 2018. The positive netflow (inflow – outflow) indicates that lateral
362 exchange via ditches and the aquifer supplied additional water to the domain during summer drawdown.

363 Simulated AET reproduces the timing and magnitude of the observed seasonal cycle, with summer maxima and
364 winter minima (RMSE = 1.01 mm day^{-1} , KGE = 0.70; Fig. 11b). The largest deviations occur during the extreme drought
365 of 2018, when the model slightly overestimates summer AET.

366 Figure 11c shows both simulated and observed groundwater levels, allowing a direct interpretation of how flux
367 variations shape water-table dynamics. Groundwater levels are well reproduced, with simulated heads closely
368 tracking observations and capturing seasonal rises, recessions, and most event-scale fluctuations (RMSE = 0.12 m,
369 KGE = 0.88, NSE = 0.92). During warm seasons, ditch water levels are generally higher than the peatland water table,
370 reflecting AET-driven drawdown of the groundwater level; the gradient between ditch and groundwater level and
371 the associated lateral inflow which are accurately captured by the model. During winter months the groundwater
372 levels are higher than the ditch water levels and outflow from the site to the ditch occurs (Fig 11a).

373 Figure 12 illustrates the strong coupling between the seasonal characteristic of the meteorological components of
374 the water balance and the water storage behavior of the study site reflected in groundwater dynamic. The wet
375 periods (summer/autumn 2017, autumn 2023) and the dry summer 2018 deviate significantly from the average
376 annual pattern. Within times with water levels above surface, inundated depressions act as important buffers
377 providing short-term storage (Fig. 12a).

378 The water storage of the site (Fig. 12a) follows this seasonal hydrological cycle. From December to February, water
379 storage increases, as precipitation and inflow exceeds evapotranspiration and outflow. From April onward, water
380 storage decreases, with the strongest decline in June–August (-60 to $-120 \text{ mm month}^{-1}$), reflecting high evaporative
381 demand. Deviations from this normal behaviour occur after periods of heavy rainfall. For example, the strongest
382 positive storage pulses, reaching values up to $+80$ to $+100 \text{ mm month}^{-1}$, occurring during months with high
383 precipitation (June/July 2017) resulted in an inundation of large parts of the site.

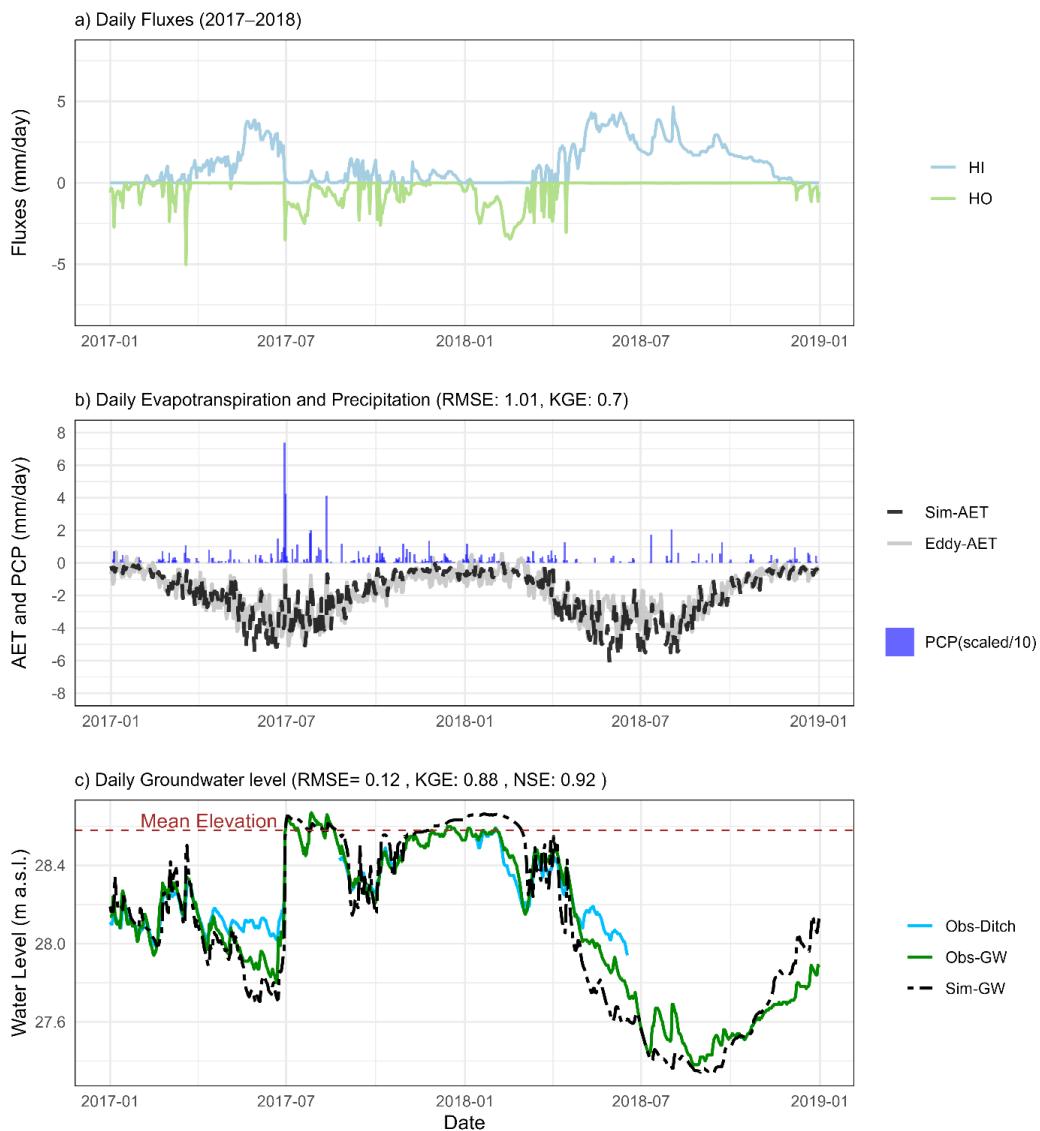
384 The groundwater hydrograph (Fig. 12b) reflects the dynamic of the water storage of the study site. It shows the same
385 seasonal pattern as the water storage (Fig. 12a), with highest water levels occurring between December and
386 February and the lowest levels typically in June–September. Winter water levels frequently rise above 28 m a.s.l.,
387 while summer minima often drop 0.5–1 m below the winter peak (e.g., see the variation in 2022). The deepest
388 groundwater levels occur in the drought years of 2018 and 2022, whereas 2017, and 2023 show elevated water
389 tables associated with wetter conditions.

390 The average monthly water storage changes across all years (Fig. 12c) highlights the consistency of this seasonal
391 pattern. Autumn and winter months display predominantly positive storage changes with low variability, whereas



392 spring and summer months exhibit consistently negative medians and a wide spread of values. June and July show
393 the strongest negative extremes due to the high stochasticity of precipitation. While the other water balance
394 components exhibit clear seasonal patterns, summer precipitation is irregular and dominated by long dry periods
395 without any precipitation and heavy rainfall events, causing a wider spread of water storage changes in these
396 months.

397 Water management measures can also lead to extreme water storage changes. For example, the pronounced
398 negative water storage change in February is the result of water management operations. Following the heavy
399 rainfall in June-July 2017, large parts of the study region remained inundated for several months. To reduce flooding,
400 the water authority lowered the water level in the main drainage channel (Großer Havelländischer Hauptkanal),
401 which subsequently caused a regional decline in groundwater levels. These effects are also well reflected by the
402 model.



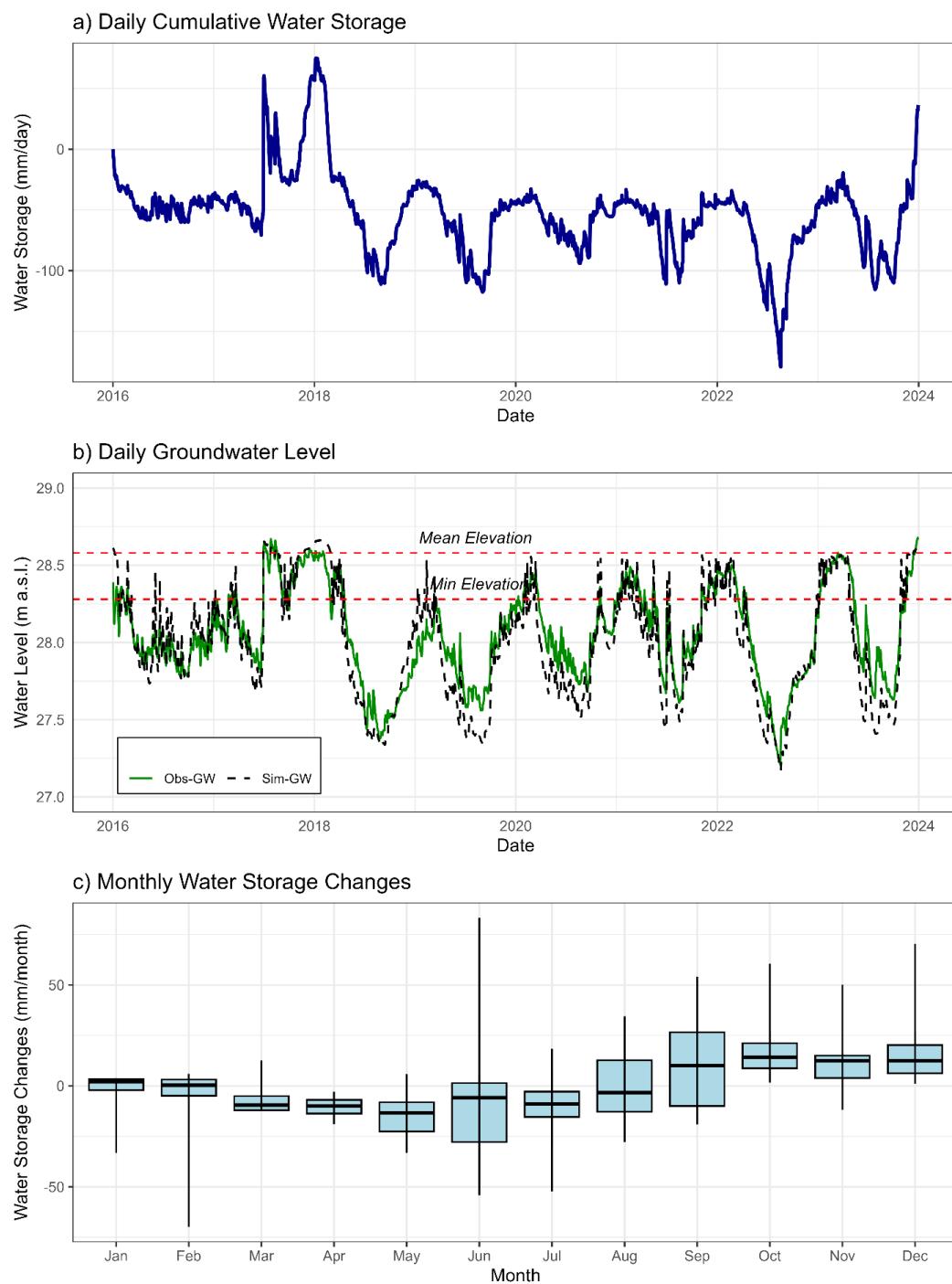
403

404 **Fig. 11.** Hydrologic dynamics of water balance components in extreme conditions (2017–2018): (a) lateral fluxes (HI
 405 – inflow, HO – outflow); (b) daily AET (sim vs. eddy covariance) and precipitation (PCP scaled by dividing by 10); (c)
 406 ditch and groundwater levels (simulations vs. observations).

407

408

409



410

411 **Fig. 12.** Simulated water storage changes relative to the 1st of January 2016 (a) observed and simulated groundwater
412 hydrograph (a) and boxplot of the simulated monthly water storage changes between 2016 and 2023 (c)

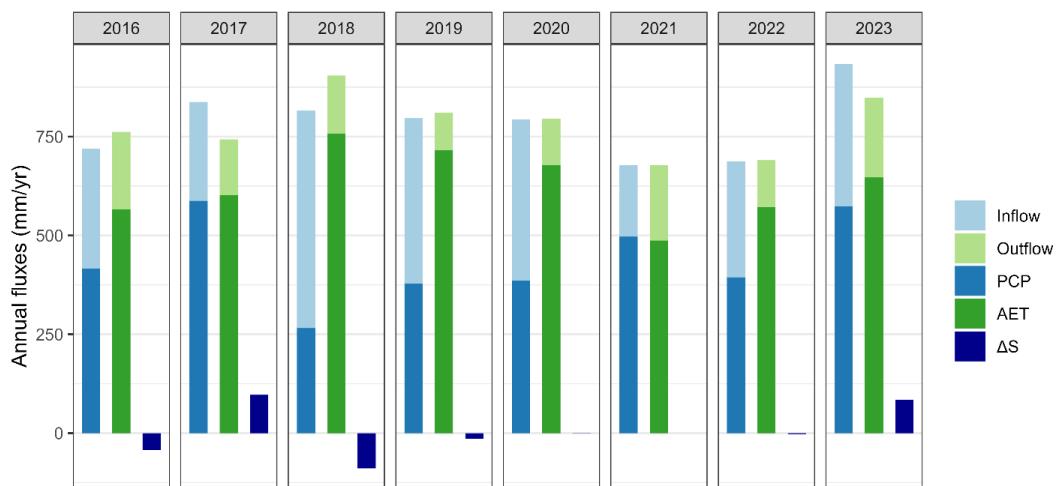


413

414 Figure 13 summarizes the annual values of all water balance components (AET, precipitation, inflow, outflow, and
415 water storage change) for the period 2016–2023. Precipitation and inflow represent the water supply and AET and
416 outflow the water consumption. A negative annual water-storage change indicates that end-of-year storage is lower
417 than at the beginning, implying that stored water was used to supply dry-year conditions. In contrast, a positive
418 storage change means that additional water was retained at the site, which typically occurs in wetter years. The
419 results illustrate the clear contrasts between wet and dry years, reflecting the combined influence of precipitation,
420 AET, and lateral exchange with the ditch–aquifer system. Years with above-average precipitation such as 2017 and
421 2023 resulted in positive annual storage changes, indicating that the peatland gained water over the hydrologic year
422 and maintained elevated groundwater levels. In contrast, the consecutive dry years of 2016, 2018 and 2019 show
423 negative annual storage, confirming a cumulative moisture deficit, progressive drying of the peat profile, and decline
424 in the groundwater level consequently. The years 2020–2022 have a neutral annual water balance, as groundwater
425 levels at the beginning and end of each year are nearly identical.

426 AET represents the dominant annual loss term in all years and varies strongly between wet and dry years. Only in
427 wet years like 2017 and 2023 do the precipitation have a similarly high level as AET. The highest AET totals occur in
428 2018, when atmospheric water demand was elevated. Although lateral inflow from the surrounding ditch–aquifer
429 system partially counterbalances these losses during summer drawdown, it is insufficient to compensate for the large
430 climatic deficits during extreme drought period. Conversely, in wet years, lateral outflow becomes more prominent
431 as high groundwater levels and winter recharge promote export of water from the peatland. Comparison of inflow
432 and outflow shows that the inflow is greater than the outflow. It underlines the dependency of the water balance of
433 the site from its catchment. This is typically for peatland sites under similar climatic conditions as our study area.

434



435
436 **Fig. 13.** Simulated annual sums of water balance components (AET: actual evapotranspiration, PCP: precipitation,
437 inflow, and outflow, ΔS : water storage change)



438

439 **4. Discussion**

440 This study demonstrates that a fully coupled surface–subsurface model, based on extensive field measurements can
441 reproduce and explain the hydrological behavior of a degraded fen peatland under varying climatic and management
442 conditions. By calibrating and validating HydroGeoSphere over nine years of data, the model reproduced
443 groundwater fluctuations and evapotranspiration dynamics with high accuracy. The integration of field-based Leaf
444 Area Index (LAI) measurements and detailed land management schedules enhanced the realism of
445 evapotranspiration parameterization, which is still rarely implemented in hydrological modeling applications. The
446 explicit representation of vegetation dynamics is therefore a distinctive aspect of this model setup. Nevertheless,
447 slight seasonal mismatches occurred, particularly an overestimation of AET during the extremely dry summer of
448 2018. This highlights limitations of parameterization under extreme weather conditions. Accuracy problems of
449 measured AET can also be a problem. But they were not in the focus of the investigation. A reason can be an
450 insufficient water supply to the plants due to very low groundwater levels (Dietrich et al., 2021). Future work could
451 explore dynamic vegetation parameterizations to reduce such discrepancies.

452 The parametrization of transpiration in this study was tailored to reflect the behavior of fen grass species that
453 tolerate high water tables and oxygen stress. By allowing plants to remain physiologically active under near-saturated
454 conditions, the model reproduced a characteristic feature of peatland vegetation that is often absent in standard
455 hydrological models, which typically reduce transpiration under such conditions. This was particularly evident in
456 summer of 2017, when water levels were above the surface, yet transpiration still occurred, reflecting the ability of
457 fen species to maintain activity under saturated conditions. The current evapotranspiration parametrization applied
458 here works well for the existing vegetation composition and successfully reproduces observed fluxes under saturated
459 conditions. However, if rewetting is sustained over the long term, vegetation composition may shift, as shown by
460 Dietrich (2024) and Dietrich and Kaiser (2017), where wetter conditions favored the establishment of sedges with
461 different evapotranspiration dynamics compared to grassland species. This indicates that future hydrological
462 modeling should allow for updating or dynamic parametrization of vegetation to reflect such ecological changes and
463 ensure that evapotranspiration feedback remain realistically represented under prolonged rewetting.

464 The two-layer representation of degraded versus less degraded peat was critical for simulating vertical water storage
465 and fluxes and therefore groundwater dynamics. The compacted upper peat dried out quickly and drove rapid water-
466 table fluctuations, while the deeper, less degraded layer buffered water losses and sustained groundwater levels
467 during droughts. This pattern is consistent with field studies showing that the upper 20–30 cm of peat, which has
468 been significantly altered by drainage, is associated with an increase in bulk density and a decrease in porosity,
469 specific yield, and saturated hydraulic conductivity, while deeper peat often retains more of its original storage
470 capacity. (Price et al., 2009; Menberu et al., 2021). If this vertical contrast is ignored, models tend to overestimate
471 water storage in the surface layer and keep too much water available for plants, which leads to an underestimation
472 of evapotranspiration sensitivity to drought. Similarly, homogeneous parameterizations often dampen simulated
473 water-table rebounds, whereas layered or dual-reservoir approaches can reproduce the fast-rewetting response
474 seen in drained peatlands after rainfall (Binet et al., 2013). This effect has also been observed in lysimeter-based
475 modeling of degraded peat soils, where unimodal (homogeneous) parameterizations tended to overestimate soil
476 water storage and dampen water-table fluctuations compared to bimodal approaches (Davies et al., 2024). Capturing



477 this heterogeneity is therefore essential for realistically assessing evapotranspiration dynamics and the rewetting
478 potential of degraded peatlands.

479 Another important consideration relates to peat degradation effects on hydraulic properties. In this study,
480 anisotropic saturated hydraulic conductivity was implemented for both, the degraded upper and less degraded lower
481 peat layers. These parameters were kept constant throughout the nine-year simulation period. While this approach
482 is appropriate for capturing short- to medium-term dynamics, it is important to acknowledge that peat hydraulic
483 properties are not static, as long-term use of agricultural machinery leads to soil compaction. In addition, numerous
484 studies have shown that peat degradation processes — such as oxidation, compaction, and subsidence — lead as
485 well to marked changes in bulk density, porosity, and saturated hydraulic conductivity over time (Sherwood et al.,
486 2013; Morris et al., 2022). Typically, intact peat with high macroporosity exhibits relatively high saturated hydraulic
487 conductivity, whereas degraded peat has much lower conductivities, in some cases several orders of magnitude.
488 Studies in tropical and temperate climate confirm that long-term drainage and land use reduce near-surface
489 saturated hydraulic conductivity by 60–70% or more, effectively transforming highly conductive peat into a low-
490 permeability medium. For long-term simulations in connection with testing different scenarios, it would therefore be
491 valuable to incorporate the dynamic evolution of peat properties, either through layered parameterizations
492 reflecting different degradation phases or through pedotransfer functions linking saturated hydraulic conductivity to
493 bulk density and humification degree. Including such time-dependent parameter changes could improve the realism
494 of long-term projections and help anticipate how ongoing degradation might alter rewetting effectiveness.

495 The study confirms that ditch water levels consistently exceed groundwater levels during summer, leading to a water
496 flow from the ditch into the site. During winter, the direction of the gradient and consequently, the flow direction is
497 often reversed, reflecting seasonal variability. Capturing these shifts is only possible if the parametrization of
498 hydraulic exchange processes is appropriately implemented. The correct representation of ditch–aquifer
499 connectivity determines whether the model can reproduce both summer and winter dynamics. In addition,
500 evapotranspiration parametrization plays a central role in shaping these gradients over the year, since vegetation
501 water use strongly influences groundwater levels during summer. Therefore, accurate representation of AET demand
502 supports the model's ability to simulate seasonal shifts in ditch–groundwater interactions. These dynamics, including
503 the occurrence of inundation during wet years, were well captured by the model. However, the sensitivity of
504 simulations to ditch boundary conditions (e.g., data gaps in 2018 and 2021) underscores the importance of reliable
505 monitoring networks. By quantifying inflow, outflow, and evapotranspiration, the model provided a detailed water
506 budget. The results highlight the vulnerability of degraded peatlands to climatic water deficits, especially during
507 consecutive drought years (2018–2020). Future rewetting strategies must account not only for precipitation
508 variability but also for ditch–aquifer connectivity and vegetation-driven demand.

509 The modeled water balance shows that the hydrological functioning of the drained fen is highly sensitive to both
510 seasonal and interannual climate variability. The system alternates between periods of net storage gain and net
511 storage loss because of the balance between precipitation inputs, evaporative demand, and the hydraulic
512 connections to the surrounding drainage network. Similar to observations in other degraded peatlands in Germany
513 (e.g., Ahmad et al., 2021), evapotranspiration emerges as the dominant control on water-table drawdown during
514 the growing season, even when groundwater levels remain shallow (e.g., April 2023). This strong atmospheric control
515 creates persistent hydraulic gradients toward ditches that reinforce lateral inflows during dry periods. The
516 fluctuations between hydrological surplus and deficit reflect a characteristic vulnerability of drained fens: their long-



517 term water balance is not simply a function of annual precipitation, but of the interaction between evaporative
518 demand, soil hydraulic properties (such as porosity), and drainage boundary conditions. As noted in previous studies
519 (Price & Ketcheson, 2009; Sherwood et al., 2013), degraded surface peat with reduced specific yield intensifies water-
520 table variability, amplifying seasonal drying and limiting the system's capacity to retain water during drought. This
521 behavior underscores a key challenge for rewetting efforts such as raising ditch water levels that alone may not be
522 sufficient to counterbalance increased evaporative demand under warmer climates, as also emphasized by Dietrich
523 (2024) and Davies et al. (2024). Effective rewetting strategies therefore require measures that reduce drainage and
524 enhance local water retention, particularly during periods of high atmospheric demand.

525 The detailed water balance analysis shows that evapotranspiration and precipitation dominate the hydrological
526 behavior of the site, while lateral inflow and outflow play a smaller but seasonally important role. Under the
527 dominant climatic conditions, with AET commonly exceeding precipitation, maintaining sufficient inflow from the
528 surrounding peatland catchment is essential to prevent persistent summer drawdown and to support rewetting.
529 Conversely, outflow is largely controlled by drainage infrastructure and increases during wetter periods, illustrating
530 the sensitivity of the system to water-level management in the ditch network. As the model realistically captures
531 these seasonal shifts and the contrasting storage behavior of degraded peat versus inundated areas, it can be applied
532 to explore management options, such as raising ditch water levels, modifying drainage control, or supplying
533 additional water from the wider catchment. In this way, the modelling framework provides a basis for designing and
534 evaluating rewetting strategies that aim to reduce drainage losses, enhance water retention, and stabilize
535 groundwater levels under future climatic conditions.

536 However, the model results also show that mesh quality and refinement influenced convergence and stability,
537 especially in the heterogeneous peat–sand–till sequence. The finite element mesh avoided extreme aspect ratios,
538 yet further adaptive refinement could improve representation of sharp hydraulic gradients at ditch–peat interfaces.
539 Although topography in our peatland site remains largely stable, long-term degradation and compaction can lead to
540 subsidence and localized depressions as a result of human activities. In such cases, applying dynamic meshing
541 schemes, as recently proposed in HydroGeoSphere applications (Hwang et al., 2025), could be valuable. These
542 approaches adjust the computational grid to evolving surface conditions and may therefore improve long-term
543 peatland simulations by maintaining numerical stability and better capturing altered hydrological pathways caused
544 by subsidence. Similarly, the Kristensen–Jensen framework used for evapotranspiration captured LAI effects well,
545 but its saturation at $LAI \approx 2.5$ may oversimplify canopy effects at higher densities. Wetland species are known to have
546 high LAI values connected with high AET (Peacock & Hess, 2004, Drexler et al., 2008, Dietrich, 2024). Exploring
547 alternative canopy-resistance formulations could improve ET estimates under lush summer conditions.

548 One of the model's great strengths is that it integrates all factors involved in the water balance of a peatland area
549 into a single model. It takes into account evaporation processes by vegetation parameters, vertical and horizontal
550 flow processes in the unsaturated and saturated soil zones, and the integration of water management facilities such
551 as ditches. The direct coupling of ditches and groundwater makes the model ideal for scenario analyses of water
552 management measures to improve water retention or peatland restoration, as these measures focus primarily on
553 raising or controlling water levels in ditches or the supply and drainage of water via ditch systems. Additionally,
554 uncertainties in peat hydraulic properties and vegetation responses to water management interventions deserve
555 closer attention. As rewetting initiatives expand in Germany and across Europe, models that integrate both the



556 physical complexity and the degradation-driven evolution of peat properties will be central to designing effective
557 peatland restoration and climate mitigation strategies.

558 While the model performed well at the field scale, scaling these insights to regional levels remains challenging due
559 to computational demand and the need for site-specific data (LAI, peat properties, ditch management, groundwater
560 dynamics). The approach nevertheless provides a benchmark for developing hybrid modeling frameworks that
561 combine physically based simulations with machine learning to extend results across larger peatland landscapes.

562

563 **5. Conclusion**

564 This study applied a fully coupled HydroGeoSphere model to investigate the hydrological functioning of a degraded
565 fen peatland under present climatic and management conditions. The model successfully reproduced groundwater
566 fluctuations, evapotranspiration dynamics, and ditch–groundwater interactions, which represent the key
567 hydrological processes governing the peatland water balance essential for accurately quantifying their contributions
568 to the overall water budget. Key to this achievement is the optimized parametrization of evapotranspiration
569 processes, which not only accounted for vegetation dynamics through measured LAI and management schedules but
570 also incorporated the physiological behavior of peatland plant species, capable of maintaining growth and
571 transpiration under fully saturated conditions. In addition, the explicit representation of the hydraulic properties of
572 the layered peat structure (degraded and less degraded) plays an important role, which allowed the model to capture
573 rapid water-table fluctuations and storage changes. The results emphasize that ditch–aquifer connectivity and water-
574 consuming vegetation strongly shape the dominant seasonal gradients between surface water and groundwater.
575 Therefore, correct parametrization of both exchange processes and evapotranspiration is essential to capture the
576 observed seasonal reversals in hydraulic gradients. Moreover, the assessment of seasonal and interannual storage
577 changes offers new insight into the long-term hydrological sensitivity of degraded fens. The marked alternation
578 between water-storage gains in wet years and cumulative deficits during drought demonstrates the limited capacity
579 of drained peatlands to buffer increases in evaporative demand, even when shallow groundwater and lateral inflow
580 provide temporary support. Overall, this work provides a detailed framework for representing peatland hydrology
581 and highlights the importance of integrating vegetation–water interactions to accurately quantify water balance
582 components. It establishes a robust baseline for future scenario analyses of management interventions required for
583 sustainable rewetting aimed at reducing CO₂ emissions and peatland conservation.

584

585 **Code and data availability**

586 The current version of the HGS model and all the data used to process, set up, and evaluate the model will be
587 available upon request.

588 **Author contributions**

589 NM: data collection, methodology, software and model setup, model input–output analysis and visualization, writing
590 (original draft, review, and editing). OD: data collection and computations, methodology, Writing (review and



591 editing). JP: field work and providing LAI data, Writing (review and editing). CM: supervision, methodology, Writing
592 (review and editing).

593 **Competing interests**

594 The contact author has declared that none of the authors has any competing interests.

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603

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