



1	Enhancing Winter Climate Simulations of the Great Lakes:
2	Insights from a New Coupled Lake-Ice-Atmosphere (CLIAv1)
3	Model on the Importance of Integrating 3D Hydrodynamics with
4	a Regional Climate Model
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20 Abstract

21	The Laurentian Great Lakes significantly influence the climate of the Midwest and Northeast
22	United States, due to their vast thermal inertia, moisture source potential, and unique heat and
23	moisture flux dynamics. This study presents a newly developed coupled lake-ice-atmosphere
24	(CLIAv1) modeling system for the Great Lakes by coupling the National Aeronautics and
25	Space Administration (NASA)-Unified Weather Research and Forecasting (NU-WRF) regional
26	climate model (RCM) with the three-dimensional (3D) Finite Volume Community Ocean
27	Model (FVCOM) and investigates the impact of coupled dynamics on simulating the Great
28	Lakes' winter climate. By integrating 3D lake hydrodynamics, CLIAv1 addresses the
29	limitations of traditional one-dimensional (1D) lake and demonstrates superior performance in
30	reproducing observed LSTs, ice cover distribution, and the vertical thermal structure of the
31	Great Lakes compared to the NU-WRF model coupled with the default 1D Lake Ice Snow and
32	Sediment Simulator (LISSS). CLIAv1 also enhances simulation of over-lake atmospheric
33	conditions, including air temperature, wind speed, and sensible and latent heat fluxes,
34	underscoring the importance of resolving complex lake dynamics for reliable climate
35	projections. More importantly, this study addresses the crucial question about what are the key
36	processes influencing lake thermal structure and ice cover that are missed by 1D lake models
37	but effectively captured by 3D lake models. Through process-oriented numerical experiments,
38	we identify key 3D hydrodynamic processes-ice transport, heat advection, and shear
39	production in turbulence—that explain the superiority of 3D lake models over 1D lake models,
40	particularly in cold season performance and lake-atmosphere interactions. Properly resolving
41	these processes using 3D hydrodynamic model is crucial for successfully simulating the lake-
42	ice-atmosphere coupled Great Lakes winter system. This research underscores the necessity of
43	incorporating 3D hydrodynamic models in RCMs to improve our predictive understanding of
44	the Great Lakes' response to climate change. The findings advocate for a shift towards high-
45	resolution, physics-based modeling approaches to ensure accurate future climate and
46	limnological projections for large freshwater systems.

47





48 1 Introduction

49	The Laurentian Great Lakes, with a surface area of 246,000 km ² , represent Earth's largest
50	surface freshwater resources, containing 21% of the world's surface freshwater and 84% of
51	North America's surface freshwater (Botts and Krushelnicki, 1987; EPA, 2014; Notaro et al.,
52	2015; Xue et al., 2022). Over 55 million people live within the Great Lakes' megaregion
53	(Todorovich, 2009; Sharma et al., 2018). The lakes support the United States' and Canadian
54	economies by impacting drinking water supply, shipping, fishing, power production,
55	transportation, manufacturing, wastewater treatment, agriculture, and recreation (Vaccaro and
56	Read, 2011). The Great Lakes' support of these vital industries sustains approximately 1.3
57	million jobs and \$82 billion in annual wages (Rau et al., 2020). As an invaluable resource to
58	wildlife and society, the ecologically diverse Great Lakes Basin is home to over 3,500 animal
59	and plant species, including over 170 fish species (Botts and Krushelnicki, 1987; Crossman and
60	Cudmore, 1998; EPA, 2014). The basin's wetlands serve as spawning and nesting habitat,
61	reduce erosion, and protect water quality (Notaro et al., 2015).
62	The Great Lakes are critically important in terms of their impacts on the climate of the Midwest

and Northeast United States and southern Ontario, Canada. The regional climate is highly

64 sensitive to the Great Lakes due to the lakes' vast thermal inertia, potential source of moisture

- 65 to the overlying atmosphere, and contrasts in heat, moisture, roughness, and albedo compared
- 66 to surrounding land (Changnon and Jones, 1972; Scott and Huff, 1996; Chuang and Sousounis,
- 67 2003; Notaro et al., 2013a; Briley et al., 2021; Wang et al., 2022). During late autumn through
- 68 winter, when cold, dry continental air masses from Canada pass over the relatively mild Great
- 69 Lakes, the air masses are destabilized and moistened, leading to enhanced cloud cover and
- 70 precipitation downwind of the lakes (Niziol et al., 1995; Ballentine et al., 1998; Kristovich and
- 71 Laird, 1998; Notaro et al., 2013b; Shi and Xue, 2019). During the broader unstable lake season,
- vhich spans from September to March and is characterized by amplified lake-effect cloud
- 73 cover and precipitation due to lake surface temperatures typically exceeding overlying air
- 74 temperatures, lake-effect snowfall typically peaks during December-January, and lake ice cover
- 75 is most extensive during February-March (Assel, 1990; Niziol et al., 1995; Kristovich and
- 76 Laird, 1998; Lam and Schertzer, 1999; Notaro et al., 2013b). The establishment of extensive
- 77 lake ice cover usually by mid-late winter dampens over-lake turbulent fluxes of heat and





- 78 moisture, subsequently reducing resulting lake-effect precipitation (Brown and Duguay, 2010; 79 Notaro et al., 2021). Specifically, increasing lake ice cover leads to a linear reduction in latent 80 heat fluxes and nonlinear reduction in sensible heat fluxes (Gerbush et al., 2008). When 81 relatively cool (warm) air masses pass over the Great Lakes during winter (summer), the 82 relatively warm (cool) lake surface reduces (enhances) atmospheric stability and increases 83 (decreases) deep convection, cloud cover, and precipitation (Scott and Huff, 1996; Holman et 84 al., 2012; Bennington et al., 2014). The lakes' relatively low roughness compared to the 85 surrounding land leads to strengthened over-lake wind speeds and potential shoreline 86 convergence in support of enhanced lake-effect precipitation. Due to the lakes' large thermal 87 inertia and resulting seasonal evolution in lake-air temperature contrast, the Great Lakes 88 typically strengthen wintertime cyclones and summer anticyclones and weaken summertime 89 cyclones and wintertime anticyclones (Notaro et al., 2013a). The basin is a preferred zone of
- 90 wintertime cyclogenesis due to the relative warmth of the lake surfaces and consequential
- 91 enhancement in low-level convergence (Petterssen and Calabrese, 1959; Colucci, 1976;
- 92 Eichenlaub, 1978).

Given the aforementioned substantial influence of the Great Lakes on regional climate, their
representation and evaluation in both global and regional climate models have been the focus of
several studies in the past decade. There is a wide spectrum among climate models regarding
the treatment of large lakes. Due to their coarse spatial resolution, most global climate models
(GCMs), including those from various phases of the Coupled Model Intercomparison Project
(CMIP), either omit the Great Lakes entirely or offer a crude representation using wet soil,
wetlands, ocean grid cells, or 1D lake models (Briley et al., 2021; Minallah and Steiner, 2021).

- 100 Among regional climate models (RCMs) without lake models, many apply a rudimentary
- 101 approach to estimate lake surface temperatures (LSTs) by extrapolating the closest ocean grid
- 102 cell's sea-surface temperatures (SSTs), likely from Hudson Bay or the North Atlantic Ocean,
- 103 from the initial and lateral boundary conditions datasets to the lake grid cell, potentially
- 104 inducing vast biases and intra-lake discontinuities in LST and ice cover (Gao et al., 2012;
- 105 Mallard et al., 2015; Spero et al., 2016; Xiao et al., 2016; Hanrahan et al., 2021). This approach
- 106 is the default treatment of LSTs in the Weather Research and Forecasting (WRF) model
- 107 (Hanrahan et al., 2021; Wang et al., 2022). Alternatively, the WRF Preprocessing System can





- 108 designate time-averaged 2-m air temperatures to the underlying lake surfaces to provide
- 109 estimated lower boundary conditions of LST based on the user-specified time window for
- 110 temporal averaging and time lag for addressing thermal inertia (Wang et al., 2012; Mallard et
- al., 2015; Hanrahan et al., 2021; Wang et al., 2022). However, this approach still produces
- 112 unrealistic LSTs and ice cover as the lakes cannot achieve equilibrium with the overlying
- atmosphere due to the lack of interactive lake-atmosphere feedbacks (Bullock et al., 2014;
- 114 Spero et al., 2016).
- 115 For those GCMs and RCMs that aim to incorporate coupled lake-atmosphere interactions, most
- 116 apply 1D lake models (Perroud et al., 2009; Martynov et al., 2010; Stepanenko et al., 2010;
- 117 Subin et al., 2012). Those include 2-layer bulk models founded in similarity theory such as the
- 118 Freshwater Lake (FLake) model (Mironov et al., 2010), thermal diffusion models which
- 119 parameterize eddy diffusivity such as the Minnesota Lake Water Quality Management Model
- 120 (MINLAKE, Riley and Stefan, 1988) and the Hostetler model (Hostetler and Bartlein, 1990),
- 121 Lagrangian turbulence models such as the Dynamics Reservoir Simulation Model (DYRSM,
- 122 Yeates and Imberger, 2003), and $k-\epsilon$ turbulence closure models with horizontally averaged
- 123 velocity such as LAKE (Stepanenko and Lykossov, 2005; Stepanenko et al., 2011) and Simstrat
- 124 (Goudsmit et al., 2002). Each of these different categories of 1D lake models has its own
- advantages and disadvantages (Perroud et al., 2009; Martynov et al., 2010; Stepanenko et al.,
- 126 2010; Subin et al., 2012). As demonstrated in these studies, the deficiencies include struggles
- 127 with simulating seasonal stratification in FLake, insufficient mixing for deep lakes in the
- 128 Hostetler model, and excessive mixing for shallow lakes in the computationally expensive
- 129 turbulence models.
- 130 Multiple modeling studies have assessed the performance of coupling RCMs to 1D lake models
- 131 in the Great Lakes region. While this coupling permits the representation of key lake-
- 132 atmosphere interactions and the heterogeneous spatiotemporal patterns of LSTs and lake ice
- 133 cover, 1D lake models typically perform poorly at reproducing the lake thermal structure and
- 134 seasonal ice evolution of large, deep lakes, such as Lake Superior, due to the overly simplified
- 135 hydrodynamic processes. Common biases in 1D lake models include an anomalously early
- 136 timing of both spring-summer stratification and autumn turnover, with positive biases in
- 137 summer LST and negative biases in winter LST (Bennington et al., 2014; Mallard et al., 2014).





- 138 The International Centre for Theoretical Physics (ICTP) Regional Climate Model version 4 139 (RegCM4), coupled to the 1D Hostetler lake model, yields a prolonged lake ice season with 140 excessive ice cover due to the neglect of horizontal heat advection within the lakes (Notaro et 141 al., 2013b). The coupling of a thermal diffusion lake model, the Lake, Ice, Snow and Sediment 142 Simulator (LISSS, Subin et al., 2012) to the WRF model (available starting with version 3.6 of 143 WRF) results in an early warm-up and overly rapid cool-down in the seasonal evolution of 144 LSTs for deep lakes, along with an early onset of lake ice cover in support of its excessive 145 abundance (Xiao et al., 2016). Mallard et al. (2014) found that WRF, coupled to FLake, 146 produced the best performance for Lake Erie (the smallest and shallowest Great Lake) and the 147 worst performance for Lake Superior (the largest and deepest Great Lake) among the Great 148 Lakes in terms of simulated LST and ice cover biases. Often, modelers aim to reduce biases in 149 the simulated vertical temperature profile of deep lakes in 1D models by artificially enhancing 150 the vertical eddy diffusivity to crudely compensate for the absence of a dynamic circulation and 151 vertical mixing processes (Subin et al., 2012; Bennington et al., 2014; Lofgren, 2014; Gu et al., 152 2015; Mallard et al., 2015), although such a non-physics based approach may only yield limited 153 benefits to minimizing these biases (Xiao et al., 2016). The lack of fully resolved lake 154 hydrodynamics in models, including dynamic 3D lake circulation, upwelling and downwelling, 155 thermal bar formation, explicit horizontal mixing, and ice motion, along with overly simplified 156 stratification processes, unrealistic treatment of eddy diffusivity, and the assumption of 157 instantaneous mixing of thermal instabilities (Song et al., 2004; Martynov et al., 2010, 2012; 158 Stepanenko et al., 2010; Bennington et al., 2014; Gu et al., 2015; Mallard et al., 2015; Sharma 159 et al., 2018; Sun et al., 2020; Notaro et al., 2021; Hutson et al., 2024) has been the main 160 obstacle in further improving climate simulations for the Great Lakes Basin.
- 161 In recent years, a limited number of Great Lakes studies have aimed to enhance the
- 162 representation of three-dimensional (3D) lake hydrodynamical processes and reduce the
- 163 substantial biases in LST and ice cover associated with 1D lake models by coupling RCMs with
- 164 3D hydrodynamic models (Xue et al., 2017, 2022; Sun et al., 2020; Kayastha et al., 2023).
- 165 These studies have responded to the urgent call for continued progress in coupling high-
- 166 resolution RCMs with 3D lake models that address the complex processes and features of large,
- 167 deep lakes, as highlighted in previous research (Martynov et al., 2010; Bennington et al., 2014;
- Briley et al., 2021; Notaro et al., 2021). Xue et al. (2017) developed a two-way coupled 3D





- 169 lake-ice-climate modeling system, known as the Great Lakes-Atmosphere Regional Model 170 (GLARM), by coupling RegCM4 with a 3D unstructured-grid hydrodynamic model, the Finite 171 Volume Community Ocean Model (FVCOM, Chen et al., 2012). The resulting coupled 3D 172 modeling system exhibited notable skill in reproducing the mean, variability, and trends in 173 regional climate across the Great Lakes Basin and the physical characteristics of the Great 174 Lakes, including their thermal structure and ice cover, significantly improving upon previous 175 RCM experiments coupled with 1D lake models. The updated version, GLARM-V2, has been 176 utilized to generate future climatic and limnological projections for the Great Lakes region 177 (Xue et al., 2022). Similarly, Sun et al. (2020) developed a lake-atmosphere-hydrology 178 modeling system by coupling the Climate-WRF (CWRF) model with 3D FVCOM and 179 compared its performance against CWRF coupled with the 1D LISSS. They found that the 180 former configuration outperformed the latter in simulating LST, ice cover, and the vertical 181 thermal structure in the Great Lakes. Kayastha et al. (2023) developed and validated the WRF-182 FVCOM Two-way Coupling (WF2C) model, showing WF2C improved upon past 1D lake 183 model-based studies by significantly reducing the simulated summer LST bias, and revealing 184 how coupled lake-atmosphere dynamics can influence summer LST by modifying surface heat 185 fluxes through impacts on meteorological state variables. These studies underscore the 186 advantages of coupling an RCM with a 3D lake hydrodynamic model for accurately depicting 187 lake physical processes and lake-atmosphere feedbacks in the Great Lakes Basin. However, 188 there is a notable absence of research dedicated to identifying the fundamental processes 189 resolved in 3D lake models that contribute to these improvements, which is important to 190 optimize effort allocation in future model development and improve our predictive 191 understanding of the system. This knowledge gap is particularly significant for the Great Lakes 192 during the winter seasons.
- 193 This paper attempts to address this knowledge gap, by developing a new coupled lake-ice-
- 194 atmosphere (CLIA version 1 or CLIAv1) modeling system for the Great Lakes by coupling the
- 195 National Aeronautics and Space Administration (NASA)-Unified Weather Research and
- 196 Forecasting (NU-WRF) regional climate model (RCM) with the three-dimensional (3D) Finite
- 197 Volume Community Ocean Model (FVCOM). Note that CLIAv1 is hereinafter referred to as
- 198 NU-WRF/FVCOM for the sake of particular attention given to comparing NU-WRF's
- 199 performance during the cold season when two-way coupled with 3D FVCOM (NU-





- 200 WRF/FVCOM) versus 1D LISSS (NU-WRF/LISSS). After a thorough validation of the
- 201 coupled model, we conduct a series of process-oriented numerical experiments to identify the
- 202 most important hydrodynamic processes that contribute to the superiority of the 3D lake model
- 203 over the 1D lake model in enhancing lake-atmosphere coupling for the Great Lakes.

204 2 Model, Data, and Numerical Experiment Design

205 2.1 Atmosphere Model

206 NU-WRF is an observation-driven integrated regional modeling system, developed at NASA's

207 Goddard Space Flight Center (GSFC), that resolves chemistry, aerosol, cloud, precipitation and

208 land processes at satellite-resolvable scales (roughly 1-25 km) to improve the continuity

209 between microscale, mesoscale and synoptic processes. Developed as a superset of the

210 community WRF, NU-WRF unifies the NCAR - Advanced Research version of WRF model

211 (WRF-ARW) with the GSFC Land Information System (LIS, Kumar et al., 2006; Peters-Lidard

212 et al., 2007, 2015), the Goddard Chemistry Aerosol Radiation and Transport (GOCART) model

213 (Chin et al., 2000), the Goddard radiation and microphysics schemes (Shi et al., 2014), and the

214 Goddard Satellite Data Simulator Unit (G-SDU, Matsui et al., 2013, 2014). NU-WRF

215 simulations here utilize the Noah Land Surface Model, which simulates soil moisture and

216 temperature, skin temperature, snowpack depth and the energy flux and water flux terms of the

217 surface energy balance and surface water balance (Mitchell, 2005). Currently, by default, the

two-way lake-atmosphere interactions in NU-WRF are represented using the embedded 1D

LISSS (Subin et al., 2012) from the Community Land Model version 4.5 (Oleson et al., 2013)

220 with modifications by Gu et al. (2015).

221 Notaro et al. (2021) conducted 20 simulations to identify the regionally optimal NU-WRF

222 configuration and schemes for the cold season period of November 2014-March 2015 in the

223 Great Lakes region. The best model configuration was referred to as the "Morrison

224 combination" and is used in this study. The "Morrison combination" includes Morrison

225 microphysics (Morrison et al., 2009), Rapid Radiative Transfer Model (RRTM, Mlawer et al.,

226 1997) longwave radiation physics, Community Atmosphere Model (CAM, Collins et al., 2004)

227 shortwave radiation physics, Mellor-Yamada-Nakanishi-Niino Level 2.5 (MYNN2.5,

228 Nakanishi and Niino, 2006, 2009) planetary boundary layer physics, and Mellor-Yamada-





- 229 Nakanishi-Niino (MYNN, Nakanish, 2001) surface layer schemes. The improved simulations
- 230 of air temperature and surface insolation using the Morrison combination primarily benefits
- from the Community Atmosphere Model's shortwave radiation scheme (Notaro et al., 2021).
- 232 The Morrison combination is essentially the WRF configuration determined by Mooney et al.
- 233 (2013) to produce the best simulated wintertime temperature simulation over Europe, who
- 234 found that winter air temperatures are highly sensitive to the choice of radiation physics.
- 235 The NU-WRF one-way nested configuration consists of an outer domain with 15-km grid
- spacing for the majority of North America and an inner domain with 3-km grid spacing for the
- 237 Great Lakes region (Fig. 1), with the atmospheric vertical resolution assigned to 61 levels. The
- 238 initial and lateral boundary conditions are provided by the Global Data Assimilation System 0-
- 239 hour analysis. The cumulus parameterization option used for the outer domain is the Kain-
- 240 Fritsch scheme (Kain and Fritsch, 1990; Kain, 2004) with resolved, unparameterized
- convection in the inner domain.







- 243 Figure 1. NU-WRF nested domains (upper panel) and unstructured mesh used in FVCOM to represent
- the Great Lakes in FVCOM (lower panel). The two dots denote the locations of Granite Island (87.4°W,
- $245 \qquad 46.7^\circ N) \text{ on Lake Superior and Spectacle Reef (84.1^\circ W, 45.7^\circ N) on Lake Huron.}$

246 2.2 Hydrodynamic Model

247 The hydrodynamic model, FVCOM, is a free-surface, primitive equation hydrodynamic model 248 that solves the momentum, continuity, temperature, salinity, and density equations and is closed 249 physically and mathematically using turbulence closure submodels (Chen et al., 2012). 250 Numerically, FVCOM employs the finite-volume method over an unstructured triangular grid 251 and vertical sigma layers, optimizing flexibility and accuracy for complex terrains. The grid 252 resolution adjusts from 1–2 km near coasts to resolve coastal geometry complexity, to 2-4 km 253 offshore to improve computational efficiency (Fig. 1), with the model comprising 35,000 grid 254 cells and 40 sigma layers. Vertical mixing processes are modeled using the Mellor-Yamada 255 level-2.5 (MY25) turbulence closure model (Mellor and Yamada, 1982), while horizontal 256 diffusivity is derived from velocity shear and grid resolution through the Smagorinsky (1963) 257 formulation.

258 FVCOM also includes an unstructured-grid, finite-volume version of the Los Alamos

259 Community Ice Code (CICE), which describes ice thickness distribution in time and space.

260 CICE includes a thermodynamic model to compute local growth rates of snow and ice due to

261 vertical conductive, radiative, and turbulent fluxes. It also features an ice dynamics model to

simulate the ice pack velocity due to wind and ice-water stress, Coriolis effects, sea surface

263 slope, and internal stress, estimated with elastic-viscous-plastic rheology (Hunke and

264 Dukowicz, 1997). The transport model in CICE calculates the advective process of the areal

265 concentration, ice volumes, and other state variables. The ridging parameterization in CICE

addresses mechanical redistribution, which transfers ice among thickness categories (Hunke et

267 al., 2010).

268 In contrast, the default 1D lake model, LISSS, embedded in NU-WRF solves the 1D thermal

269 diffusion equation (i.e. lake thermal dynamics only) by segmenting the vertical stratification of

the lake into multiple distinct levels that include: snow (applicable when the snow's thickness

271 surpasses a specified minimum value); the combined section of lake water and ice, collectively





- 272 identified as the "lake body"; and the bottom layers consisting of sediment, soil, and bedrock
- 273 (collectively termed "sediment" unless specified differently). This structured division allows for
- simulating thermal dynamics within each segment, facilitating a prediction of temperature
- distribution and variations across the lake's depth (Subin et al., 2012).

276 3 Two-way Coupling of NU-WRF/FVCOM

277 The development of interactively coupled model systems [see review by Giorgi and Gutowski 278 Jr. (2015)] emerged quickly in the late 2000s driven by rapid technological advancement and 279 the increase in computational capability. Model coupling is essential to multi-physics 280 simulations representing various components of the Earth system. Over the past two decades, 281 several coupling technologies for earth system modeling have been developed. Examples 282 include the Earth System Modeling Framework (ESMF), the Model Coupling Toolkit (MCT), 283 and the OASIS-MCT coupler, which is the latest version of the OASIS3 coupler interfaced with 284 the Model Coupling Toolkit (MCT) that offers a fully parallel implementation of coupling field 285 regridding and exchange (Valcke et al., 2012; Craig et al., 2017). Although coupling 286 implementations can follow different approaches, their applications in geophysical simulations 287 typically carry out several key functions, including interpolating and transferring the coupling 288 fields between different model grids, managing data transfer between constitutive models at a 289 desired coupling frequency, and coordinating the execution of the constituent models in a 290 parallel computational environment (Valcke et al., 2012). In general, coupling data must be 291 interpolated and transferred between the constituent models under several constraints, such as 292 conservation of physical properties, numerical stability, consistency with physical processes, 293 and computational efficiency.

- 294 In the study, NU-WRF and FVCOM are run simultaneously, exchanging information
- 295 bidirectionally at 1-hour intervals through the OASIS3-MCT coupler. FVCOM dynamically
- 296 calculates the LST and ice cover, providing these as overlake surface boundary conditions to
- 297 NU-WRF. Meanwhile, NU-WRF calculates and supplies the atmospheric forcings required by
- 298 FVCOM, including surface air temperature, surface air pressure, relative and specific humidity,
- 299 total cloud cover, surface winds, and downward shortwave and longwave radiation.





300 3.1 Data for Model Validation

301	The average daily LST, obtained from composite images taken by the Advanced Very High
302	Resolution Radiometer, is sourced from version 2 of the Great Lakes Surface Environmental
303	Analysis (GLSEA) LST Dataset, developed by the National Oceanic and Atmospheric
304	Administration's (NOAA) Great Lakes Environmental Research Laboratory (GLERL). A
305	comprehensive evaluation carried out by Schwab et al. (1999) shows that LST measurements
306	from GLSEA and the buoy-based LSTs had an average discrepancy of less than 0.5°C across
307	all buoys, with a root-mean-square difference (RMSD) between 1.10°C and 1.76°C. The Great
308	Lakes Ice Cover Dataset, compiled by GLERL, has also been added to the GLSEA product.
309	The dataset incorporates daily average ice cover data across the lakes, which draws from ice
310	products produced by the United States National Ice Center and the Canadian Ice Service, and
311	is detailed in studies by Assel et al. (2002, 2013), Assel (2005), and Wang et al. (2012).
312	In-situ lake thermistor measurements for vertical lake thermal structure were obtained from
313	Spectacle Reef on Lake Huron (Fig. 1). Measurements for over-lake atmospheric variables,
314	including air temperature, wind velocity, downward shortwave radiation, and sensible and
315	latent heat fluxes, were obtained from Granite Island on Lake Superior and Spectacle Reef on
316	Lake Huron through the Great Lakes Evaporation Network (GLEN) (Blanken et al., 2011;
317	Spence et al., 2011; Lenters et al., 2013; Spence et al., 2013; Spence et al., 2019). These level-1
318	eddy covariance data received minimal adjustments, notably the elimination of heat spikes and
319	a basic visual quality assessment. This dataset was compared with an independent dataset of
320	Great Lakes' turbulent fluxes developed by Moukomla and Blanken (2017), revealing a "good
321	statistical agreement" between them, with RMSD ranging from 4.5 to 7 $W\!/\!m^2$ for latent and
322	sensible heat fluxes (Moukomla and Blanken, 2017).

323 **3.2 Design of Numerical Experiments**

324 We designed numerical experiments in two categories. In category 1, we evaluate the cold

325 season performance of the NU-WRF/FVCOM two-way coupling (case C1-1) against the NU-

326 WRF/LISSS 1D lake model (case C1-2). To ensure the objectivity of the comparison, both C1-

- 327 1 and C1-2 utilize an identical NU-WRF configuration (except for differences in lake
- 328 treatment) as described in Section 2.1, following the optimal NU-WRF configuration for the
- 329 study region as determined by Notaro et al. (2021). The comparison of C1-1 and C1-2 aims to





- 330 examine the overall impact of using a 3D versus a 1D lake model configuration on simulating
- 331 lake hydrodynamic conditions and the subsequent impact on the atmospheric state through
- 332 lake-ice-atmosphere interactions from November 2014 to March 2015. The initial lake
- 333 conditions of November 2014 were obtained from multiple years of FVCOM standalone
- 334 simulations driven by Climate Forecast System Reanalysis (CFSR) forcing Xue et al. (2015).
- 335 In category 2, a set of process-oriented numerical experiments is designed to identify the
- 336 impact of various 3D hydrodynamical processes critical to the coupled Great Lakes system.
- 337 These processes are either neglected or oversimplified by the NU-WRF/LISSS 1D lake model
- 338 while being resolved by the NU-WRF/FVCOM 3D lake model. Case C2-1 (NoIceTransp) is
- designed to examine the impact of ice transport associated with currents (Section 5.1). In this
- 340 scenario, FVCOM is configured identically to C1-1, except that ice dynamics, ice velocity
- 341 fields, and ice pack transport are disabled in FVCOM. Instead, only ice thermal dynamics are
- 342 simulated to account for the spatio-temporal evolution of ice thickness distribution through
- thermodynamic growth and melting processes (Bitz and Lipscomb, 1999). Consequently, the
- 344 ice model is simplified to function as an energy-conserving thermodynamic model, akin to that
- 345 used in the 1D lake model.
- 346 Case C2-2 (NoHeatAdv) analyzes the impact of 3D heat transport associated with lake
- 347 circulation. FVCOM is configured identically to C1-1, except that the advective heat transport
- 348 associated with current movement is disallowed in C2-2. This is realized by turning off the
- 349 advection terms in the temperature equation in FVCOM, which is essentially an advection-
- 350 diffusion equation that governs the distribution and evolution of temperature (Section 5.2).
- 351 Therefore, the temperature calculation is simplified to imitate the 1D vertical diffusion equation
- 352 used in the 1D lake model.
- 353 Case C2-3 (NoShearProd) aims to assess the influence of 3D currents on calculation of
- 354 turbulent mixing, a crucial factor in controlling the heat redistribution and thermal structure in
- 355 the lakes. In this case, we exclude the turbulence shear production term that depends on
- currents in the turbulent kinetic equation (Section 5.3). In summary, the three cases in category
- 357 2 collectively reveal the significant impacts of currents in elements that are not accounted for in
- 358 the LISSS 1D lake model, i.e. on ice transport, heat transport, and turbulent mixing intensity,





- 359 respectively. These experiments are summarized in Table 1.
- 360 Table 1. A summary of the numerical model experiments. The "3D currents" column shows if the
- 361 experiment resolves the 3D currents of the Great Lakes. The "Ice transport" column shows if the
- 362 experiment resolves the ice transport associated with currents in the Great Lakes. The "Heat advective
- 363 transport" column shows if the experiment resolves the 3D heat transport associated with Great Lakes
- 364 circulation. The "Shear production in turbulence" column shows if the experiment uses the turbulence
- 365 shear production term that depends on currents in the turbulent kinetic equation. The "Lake model"
- 366 column shows the lake model used in the experiment.

Experiment	3D	Ice	Heat	Shear	Lake
	currents	transport	advective	production in	model
			transport	turbulence	
C1-1 (Lake3D)	Yes	Yes	Yes	Yes	FVCOM
C1-2 (Lake1D)	No	No	No	No	LISSS
C2-1 (NoIceTransp)	Yes	No	Yes	Yes	FVCOM
C2-2 (NoHeatAdv)	Yes	Yes	No	Yes	FVCOM
C2-3 (NoShearProd)	Yes	Yes	Yes	No	FVCOM

367

368 4 Results

369 4.1 Lake Temperature and Ice Coverage

370 The NU-WRF/FVCOM model (case C1-1) accurately captures the seasonal evolution of LSTs

371 across all of the lakes with lake-mean LST root-mean-square-error (RMSE) less than 0.4°C

372 (Fig. 2 upper panels). During November, the lakes are in the middle of their cooling period and

373 the LSTs decrease rapidly, yet at different paces, largely due to variations in the lakes' depth





374 and latitude, which leads to strong spatial heterogeneity in LST (Fig. 3, left panels). The 375 GLSEA data and the 3D lake model closely align in terms of the spatial LST patterns, with 376 warmer waters of 10-12°C in the central and eastern basins of Lakes Erie and Ontario and 8-377 10°C in the southern basins of Lakes Michigan and Huron, while much cooler temperatures are 378 found across Lake Superior, ranging between 4-6°C. The most notable underestimation of LST 379 by the 3D lake simulation occurs in the southern basin of Lake Huron, while the model well 380 captures the LSTs in the northern basin of Lake Huron. Transitioning to January 2015 (Fig. 3, 381 right panels), at the onset of the ice season, NU-WRF/FVCOM accurately reflects the seasonal 382 cooling of the lakes, showing a significant reduction in LSTs, while also well delineating the 383 detailed temperature differences between the colder nearshore and relatively warmer offshore 384 waters, in good agreement with the observational data. On the other hand, NU-WRF/LISSS 385 (case C1-2) fails to capture the spatial heterogeneity in LSTs, but also generates a systematic 386 cold bias of 2-3°C during January across nearly all of the lakes (Fig. 3, bottom panels). Such a 387 cold bias was persistent in the NU-WRF/LISSS (Lake1D) simulation throughout the cold 388 season, as detailed in Notaro et al. (2021).



Figure 2. Time series of daily lake-averaged LST (°C, upper panels) and percent ice cover (lower panels)
for the five lakes from GLSEA data (black lines) and NU-WRF/FVCOM 3D lake model simulations (red
lines) during the simulation period of November 2014-March 2015. Both the temporal correlation and
RMSE are reported in each panel.





394



395 Figure 3. Spatial patterns of monthly mean LSTs (°C) from GLSEA data (top panels), NU-

396 WRF/FVCOM 3D lake model simulations (middle panels), and NU-WRF/LISSSS 1D lake model

397 simulations (bottom panels) for November 2014 (left panels) and January 2015 (right panels).

398 NU-WRF/FVCOM (Lake3D) also demonstrates its skill in capturing the evolution of the 399 vertical thermal structure within the lake, which is particularly challenging in large and deep 400 lakes (Bennington et al., 2014; Xue et al., 2017). As exemplified in Fig. 4, the in-situ thermistor 401 measurement at Spectacle Reef on Lake Huron is located in a deep region with a water depth 402 greater than 200 meters. The 3D model reproduces the conclusion of the summer stratification 403 process until the end of November. The following turnover, a seasonal process where the 404 surface water cools, becomes denser, and sinks-mixing with the warmer water from below-405 is also represented in the 3D lake model between December and January. Subsequently, the 406 winter inverse stratification, where colder water ($\leq 4^{\circ}$ C) lies above warmer water due to the fact 407 that freshwater's density peaks at 4°C, is captured by the 3D model as it develops from 408 February onward, although the model shows a stronger winter inverse stratification and earlier 409 onset than observed. In contrast, NU-WRF/LISSS falls short of these detailed observations. Not 410 only does it mispredict the occurrence of turnover and winter stratification much earlier than





- 411 observed, but it also substantially underestimates the extent of mixing between the surface and
- 412 deeper waters. This underestimation results in a flawed representation of excessive surface
- 413 cooling and a substantial overestimation of the warming of the deep waters.



414

415 Figure 4. Seasonal evolution of daily vertical temperature (°C) profiles from the thermistor observations

416 (top panel), NU-WRF/FVCOM 3D lake model (middle panel), and NU-WRF/LISSS 1D lake model

417 (bottom panel) at Spectacle Reef in Lake Huron during November 2014-March 2015.

418 Correspondingly, NU-WRF/FVCOM resolves the spatiotemporal evolution of lake ice cover

419 very well across all of the lakes with RMSE of percent ice cover less than 8% for Lakes Huron,

420 Michigan, and Ontario and 11% and 18% for Lakes Superior and Erie, respectively (Fig. 2

- 421 lower panels). The 3D lake model and GLSEA data exhibit similar seasonal trends both in
- 422 timing and magnitude, with ice cover typically starting to rapidly increase in January, peaking
- 423 in February and early March, and declining thereafter (Fig. 2). Lake Erie shows the earliest and
- 424 sharpest increase in ice cover, peaking near 100% in early February and throughout mid-March,
- 425 indicative of its shallower depth and weaker thermal inertia. Lakes Huron and Superior show a





- 426 persistent increase in ice cover through February, with peak coverage of >90% occurring at the
- 427 beginning of March. Lakes Michigan and Ontario exhibit more gradual increases and lower
- 428 peaks in ice cover. The model appears to capture the general seasonal trends of the GLSEA
- 429 data with high fidelity, although some discrepancies are evident, particularly over Lakes Erie
- 430 and Superior (Fig. 2).

431 NU-WRF/FVCOM performs reasonably well in mirroring the general spatial patterns of lake 432 ice cover (Fig. 5, top and middle panels). For January, the GLSEA data shows a pronounced ice 433 formation in the nearshore regions across the lakes, with the greatest ice concentration visible 434 along the coastlines and very limited ice cover in offshore waters. The model captures this 435 nearshore ice development quite well, although it suggests less ice cover in the offshore areas, 436 particularly over Lake Erie. In February, the extent of ice cover varies dramatically across the 437 lakes, including nearly full ice cover on Lake Erie and significant ice-free areas on Lake 438 Ontario, as well as for Lakes Michigan and Huron, which have distinctly less ice cover in their 439 southern and central basins, respectively. The model captures this variability very well, while 440 slightly overestimating the ice cover in the central regions of Lake Superior. For March, the 441 model successfully replicates the patterns of significant declines in ice cover in the western 442 sections of the lakes, with much higher ice coverage in the eastern sections of the lakes.

- 443 On the other hand, NU-WRF/LISSS (Lake1D) generates excessive ice cover during January,
- 444 when both observations and NU-WRF/FVCOM suggested that the majority of the lakes were
- 445 ice-free. In February, the excessive ice cover simulated by the NU-WRF/LISSS model persists,
- 446 with near 100% ice coverage over all of the lakes, and the model fails to depict the large spatial
- 447 variability across the lakes. Such a persistent overestimation of ice cover throughout the cold
- season by NU-WRF/LISSS was also reported in Notaro et al. (2021).







449

Figure 5. Spatial patterns of mean percent lake ice cover from GLSEA data (top panels), NUWRF/FVCOM 3D lake model simulations (middle panels), and NU-WRF/LISSSS 1D lake model
simulations (bottom panels) for January 2015 (first column), February 2015 (second column), and March

453 2015 (third column).

454 4.2 Over-lake Latent and Sensible Heat Fluxes

455 The improved LST and ice simulation by the 3D lake model translates to an improvement in the 456 simulated over-lake latent and sensible heat fluxes, particularly for the ice-cover season (Fig. 457 6). The observations for upward latent and sensible heat fluxes from two eddy covariance flux 458 towers at Granite Island on Lake Superior and Spectacle Reef on Lake Huron are compared 459 against the simulated fluxes from NU-WRF/FVCOM (Lake3D) and NU-WRF/LISSS 460 (Lake1D). The two lakes are selected for demonstration as they have the highest ice coverage 461 during the simulation period. NU-WRF/LISSS reasonably simulates the magnitude and 462 variability of the heat fluxes from November until mid-December, similar to the observations 463 and NU-WRF/FVCOM, although with larger biases. However, it grossly underestimates the 464 fluxes during the ice-cover season (January-March) by simulating a nearly constant near-zero 465 flux. This is mainly due to the excessive ice cover simulated by the 1D lake model, which 466 creates a physical barrier for air-lake energy fluxes. Since the 3D lake model more accurately





- 467 simulates the LST and ice cover, it successfully captures the magnitude and variability of the
- 468 heat fluxes, even during the ice-cover season, with RMSEs that are 50% lower than those from
- 469 the 1D lake model (Fig. 6). Latent heat in Spectacle Reef is the only exception, where NU-
- 470 WRF/FVCOM struggles to capture the magnitude of the upward latent heat flux due to the
- 471 overestimated ice cover at the site. However, it still outperforms NU-WRF/LISSS in terms of
- 472 capturing the seasonal trend in latent heat fluxes.



474 Figure 6. Time series of daily sensible (upper panels) and latent (lower panels) heat fluxes (W/m²) from 475 GLEN observations (black lines), NU-WRF/FVCOM 3D lake model simulations (red lines), and NU-476 WRF/LISSS 1D lake model simulations (blue lines) at Granite Island on Lake Superior (left) and 477

- Spectacle Reef on Lake Huron (right). The RMSE and temporal correlations between the simulations and
- 478 GLEN observations are provided in each panel.

479 4.3 **Over-lake Air Temperature and Wind**

- 480 Along with the improved simulation of the Great Lakes' physical characteristics and surface
- 481 heat fluxes, NU-WRF/FVCOM improves the simulated over-lake atmospheric state across the
- 482 Great Lakes, including air temperature and wind speed. The cold air temperature biases
- 483 produced over the lakes by NU-WRF/LISSS are significantly reduced (Fig. 7) with better





484 simulated, more intense upward heat fluxes in January. This improvement in the simulated air 485 temperature at the two sites, Granite Island and Spectacle Reef, is clearly evident. Similar to the 486 fluxes, NU-WRF/LISSS modeled air temperature diverges from the observations in January 487 and February, with a noticeable cold bias. This cold bias is the result of significant suppression 488 of the upward heat fluxes during those months in the 1D lake model due to excessive simulated 489 ice cover. NU-WRF/FVCOM, on the other hand, produces a much warmer and more accurate 490 over-lake air temperature for January and February due to its reasonable representation of 491 upward heat fluxes. The simulated wind speed over the lakes is also improved, especially in 492 January-February (Fig. 7). This advancement is attributed to the refined simulation of surface 493 roughness (i.e., ice versus water), the water-air temperature gradient, and associated instability 494 over the lakes due to decreased ice cover. Large wind spikes (16 m/s) in January-February are 495 more accurately captured by NU-WRF/FVCOM.



497 Figure 7. Time series of daily air temperature (°C, upper panels) at 2-m height (T2) and wind speed
498 (m/s, lower panels) from GLEN observations (black lines), NU-WRF/FVCOM 3D lake model

- 499 simulations (red lines), and NU-WRF/LISSSS 1D lake model simulations (blue lines) at Granite Island
- 500 on Lake Superior and Spectacle Reef on Lake Huron during November 2014-March 2015. The RMSE
- 501 and temporal correlations between the simulations and GLEN observations are provided in each panel.





502 **5 Discussion**

503	The Great Lakes modeling community has agreed on the pressing need to integrate 3D lake
504	models instead of conventional 1D lake modeling in the Great Lakes regional climate studies
505	(Delaney and Milner, 2019). However, no studies have yet detailed the key 3D hydrodynamic
506	processes that explain the superiority of 3D lake models over 1D lake models, especially
507	regarding cold season performance and lake-atmosphere interactions. The primary goal of this
508	study is to identify the crucial processes influencing lake thermal structure and ice cover that
509	are missed by 1D lake models but effectively captured by 3D lake models, through a series of
510	process-oriented experiments presented below.

511 5.1 Impact of Ice Movement

512 The 3D hydrodynamic model, FVCOM, includes an embedded unstructured-grid ice model 513 capable of resolving several components for atmosphere-ice-water interactions (Gao et al., 514 2011). It includes a thermodynamic model that computes the local growth rates of snow and ice 515 due to vertical conductive, radiative, and turbulent fluxes, aligning with features typically 516 included in 1D lake models (Bitz and Lipscomb, 1999). More importantly, it features an ice 517 dynamics model that predicts the ice pack's velocity field based on its material strength; a 518 transport model that describes the advection of areal concentration, ice volumes, and other state 519 variables; and a ridging parameterization that facilitates the transfer of ice among thickness 520 categories (Hunke et al., 2010).

521 Case C2-1 (NoIceTransp) is designed to examine the impact of ice transport on LSTs and

522 overlying atmospheric conditions, compared to standard case C1-1 (Lake3D). In case C2-1, ice

523 dynamics, velocity fields, and ice pack transport are disabled in FVCOM. Instead, only ice

524 thermal dynamics are simulated, as in the 1D lake model. Figure 8 compares cases C1-1

525 (Lake3D) and C2-1 (NoIceTransp), illustrating their performance in simulating the observed

526 spatial pattern of ice coverage in March 2015, characterized by open water on the western side

527 of the Great Lakes and predominant ice cover on the eastern side (Fig. 8a). Utilizing a 3D lake

528 model that only accounts for ice thermal dynamics results in an overestimation of ice cover,

529 with near 100% lakewide ice cover in Lakes Superior, Huron, and Erie (Fig. 8b). However,

530 integrating ice dynamics, including transport influenced by wind and water-ice stress, results in





531 excellent agreement with observations, highlighting the critical role of ice transport in accurate 532 ice modeling (Fig. 8c). This pattern aligns with the modeled ice velocities, which attribute the 533 eastward ice cover distribution to dominant eastward ice transport (Fig. 8d). Under cold winter 534 conditions characterized by strong westerly winds, ice is driven eastward, maintaining open 535 water in the lake's western part. This facilitates ongoing atmospheric interactions, allowing for 536 heat release. Neglecting these dynamics leads to unrealistic ice accumulation by diminishing 537 the influence of wind on surface water movement and mixing. This overaccumulation of ice 538 cover hampers the efficiency of vertical turbulent mixing, which is essential for maintaining a 539 warmer surface layer, thereby exacerbating ice formation and accumulation. The incorporation 540 of ice dynamics into 3D lake models is thus essential for accurately simulating ice distribution, 541 emphasizing the necessity of resolving ice transport to replicate observed patterns accurately.



543 Figure 8. Spatial patterns of mean percent lake ice cover from GLSEA data (a), case 2-1 (NoIceTransp)
544 simulations (b), and case 1-1 (Lake3D) standard simulations (c), along with simulated mean ice

545 velocities (m/s) during (d) March 2015.





546 5.2 Impact of Heat Transport

547 The 3D lake model also resolves the advective transport of heat associated with the simulated 548 circulation. The advective transport and turbulent mixing of temperature in the 3D lake model 549 are governed by following equation:

550
$$\frac{\partial T}{\partial t} + u \frac{\partial T}{\partial x} + v \frac{\partial T}{\partial y} + w \frac{\partial T}{\partial z} = \frac{\partial}{\partial z} \left(K_h \frac{\partial T}{\partial z} \right) + F_T \tag{1}$$

551 with the surface heat flux boundary condition:

552
$$\frac{\partial T}{\partial t} = \frac{1}{\rho c_p K_h} [LW(x, y, t) - LH(x, y, t) - SH(x, y, t)]$$
(2)

where *T* is the water temperature and u, v, and w are the x, y, and z components of the water

velocity, respectively. K_h is the vertical thermal diffusivity coefficient and F_T is the horizontal

555 diffusion term. ρ is water density, c_p is the specific heat capacity of water and

556 LW(x, y, t), LH(x, y, t), and SH(x, y, t) are net longwave radiation, upward latent heat and

sensible heat fluxes varying in space and time, respectively.

558 Case C2-2 (NoHeatAdv) analyzes the impact of 3D heat transport. In this case, the 3D 559 temperature advection terms $(u\frac{\partial T}{\partial x}, v\frac{\partial T}{\partial y}, w\frac{\partial T}{\partial z})$ are turned off.

560 Comparing the standard simulation C1-1 (Lake3D) to case C2-2 (NoHeatAdv), Figure 9 561 demonstrates that, in the absence of advective heat transport by lake currents, the surface 562 temperatures can remain consistent with the basic patterns observed in the standard 3D lake 563 simulation throughout the entire simulation period. The differences in the time series of lake-564 wide average LSTs for the five lakes are small, with a maximum difference of 0.4°C between 565 the two cases. The spatial patterns of LST biases, when compared with GLSEA, are generally 566 more noticeable, with the most significant positive biases (~ 2° C) concentrated around the 567 coastal waters of the Great Lakes and eastern Lake Erie from January to March 2015 and larger 568 negative biases (~ 3°C) in the central basin of Lake Huron in November 2014 in the 569 NoHeatAdv case.





- 570 In addition, the vertical transport associated with upwelling, resolved by the 3D model, brings
- 571 relatively warmer water from deep in the lake to the surface. This vertical transport mechanism
- 572 cannot be represented in 1D lake models that only account for vertical diffusion. This can
- 573 create significant local-scale differences along the coast, as shown on the western shore of Lake
- 574 Superior in March 2015 [Fig. 9, bottom panels. Notice that the GLSEA is not able to well
- 575 capture coastal upwelling (Ye et al., 2020)]. This underscores the importance of including
- 576 advective heat transport to accurately resolve the redistribution of heat within the lake. The
- 577 inclusion of advective dynamics, by facilitating both lateral and vertical redistribution, enables
- 578 a more realistic simulation of the complex spatial heat patterns within large lake systems.





580 Figure 9. Spatial patterns of mean LSTs (°C) from GLSEA data (first column), case C1-1 (Lake3D)

- 581 standard simulations (second column), and case C2-2 (NoHeatAdv) simulations (third column) from
- 582 November 2014 (top row) to March 2015 (bottom row). Their monthly biases relative to GLSEA data are
- 583 presented in the fourth and fifth columns, respectively.
- 584 Capturing the evolution of the vertical thermal structure within the deep water is particularly
- 585 challenging in lake models. As previously shown in Fig. 4, the in-situ thermistor measurement





- at Spectacle Reef on Lake Huron is located in a deep region with a water depth greater than 200
- 587 meters. Case C2-2 (NoHeatAdv) generally reproduced the thermal patterns from case C1-1
- 588 (Lake3D) in terms of both timing and intensity of summer stratification, fall turnover, and
- 589 winter inverse stratification (Fig. 10a,b). While the comparison shows that the overall thermal
- 590 structures are similar in both simulations, there is a noticeable difference within the subsurface
- 591 layer, specifically between 50 to 100 meters in depth (Fig. 10c), suggesting that heat advection
- 592 might have a more significant impact on temperature distribution in the subsurface layer of the
- 593 water column in this case. Without accounting for heat advective transport, there appears to be
- 594 artifacts of stepwise vertical thermal gradients in case C2-2.



- 596 Figure 10. Mean vertical temperature (°C) profiles from a): case C1-1 (Lake3D) standard run and b):
- 597 case C2-2 (NoHeatAdv) and c): their difference at Spectacle Reef in Lake Huron during November 2014-
- 598 March 2015.



604



- 599 To gain a deeper understanding of the results, we analyzed the heat balance to identify the
- 600 contributions of different physical processes. This analysis involved examining each term in the
- 601 temperature governing equation (Eq. 1) that is directly computed in FVCOM over the
- 602 simulation period. The temperature change is driven by 3D advective heat transport, horizontal
- 603 heat diffusion, and vertical diffusion due to turbulent mixing.



605 Figure 11. Monthly averaged vertical profile of each term of the temperature equation in the C1-1

606 (Lake3D) simulation from November 2014 to March 2015, output from location at Spectacle Reef in

607 Lake Huron. The temperature change rate $\left(\frac{\partial T}{\partial t}\right)$ is determined by 3D advection (blue), horizontal diffusion

608 (red), and vertical diffusion (purple).

609 The analysis revealed the relative impact of physical processes on thermal changes during the 610 winter months (Fig. 11). Significant cooling and decreasing temperatures were observed within 611 the upper 100 meters of the water column, as indicated by the negative temperature change over time $\left(\frac{\partial T}{\partial t}\right)$ in this zone. In contrast, the water below 100 meters in depth remained largely 612 613 unchanged and stable in temperature. In November and December, vertical turbulent mixing 614 processes primarily controlled the cooling rate in the upper 100 meters, during which surface 615 heat fluxes served as net losses from the lake along with vigorous turbulent mixing in the lake. 616 Advection played a much less important role in temperature changes during this period. 617 However, starting in January, 3D advection played an important role in redistributing heat in 618 the 25-100 meter layer, offsetting some of the cooling induced by surface heat loss through 619 mixing. In February and March 2015, advection proved to be significant at the lower boundary





- 620 of the surface mixed layer. These observations explain the larger temperature difference in the
- subsurface layer between cases C1-1 (Lake3D) and C2-2 (NoHeatAdv) (Fig. 10), highlighting
- 622 the evolving balance between vertical diffusion and advection in controlling the epilimnetic
- heat budget and temperature changes in large lakes during the cold season.

624 5.3 Impact of Vertical Mixing

625 The analysis above (Fig. 11) highlights the dominant factor, vertical turbulent mixing, in 626 determining seasonal lake temperature change. Note that we have already discussed the 627 importance of ice transport associated with currents as well as the impact of advective heat 628 transport. To understand the mechanism responsible for the differing performance between the 629 1D and 3D lake models in simulating vertical mixing, we examine how vertical turbulent 630 mixing is calculated in these two types of models. The intensity of vertical mixing in both 631 models is represented by vertical eddy diffusivity, which is determined by turbulent kinetic energy (q^2) . In the 3D hydrodynamic lake model, a sophisticated 3D turbulence closure model 632 is used, in which a prognostic equation predicts the change rate of q^2 based on its advection, 633 634 and its turbulence production, including both shear-induced production (P_s) and buoyancy-635 induced production (P_b) , and its dissipation rate (ϵ) , as well as its diffusion. This equation is 636 complemented by either a separate prognostic equation for dissipation rate (k- ϵ ; Launder and 637 Spalding, 1974) or a diagnostic equation for turbulent mixing length (Meller and Yamada, 638 1982).

639 The equation governing the evolution of turbulent kinetic energy (q^2) is

640
$$\frac{\partial q^2}{\partial t} + u \frac{\partial q^2}{\partial x} + v \frac{\partial q^2}{\partial y} + w \frac{\partial q^2}{\partial z} = P_s + P_b - \epsilon + \frac{\partial}{\partial z} \left(K_q \frac{\partial q^2}{\partial z} \right) + F_q \tag{3}$$

641 where $q^2 = (\langle u'^2 \rangle + \langle v'^2 \rangle + \langle w'^2 \rangle)/2$, with u', v', w' represent the fluctuating components of 642 velocity in the *x*, *y*, *z* directions, respectively. The $\langle \rangle$ denotes averaging over time or space to 643 obtain the mean. Shear production is often approximated as $P_s = K_m((\frac{\partial u}{\partial z})^2 + (\frac{\partial v}{\partial z})^2)$, where 644 K_m is the vertical eddy viscosity coefficient. Buoyancy production is computed as $P_b =$ 645 $-\frac{g}{\rho_0}K_h\frac{\partial \rho}{\partial z}$, where *g* is acceleration due to gravity. ρ_0 is reference density of the fluid (e.g.,





- 646 ocean water or air). K_h is thermal diffusivity, $\frac{\partial \rho}{\partial z}$ is vertical gradient of density, indicating
- 647 stratification. The turbulent kinetic energy dissipation rate is represented as $\epsilon = q^3/Bl$, where
- 648 *l* is the turbulence length scale and B is an empirical constant. $K_{m,h,q} = qlS_{m,h,q}$, where $S_{m,h,q}$
- 649 are stability functions for $K_{m,h,q}$, respectively. K_q is the vertical diffusivity coefficient for
- turbulent kinetic energy and F_q is horizontal diffusion of the turbulent kinetic energy.
- Figure 12 reveals that in the Great Lakes, shear production—induced by the vertical gradient of
- 652 horizontal velocity in the water column—is the primary driver of subsurface turbulent mixing.
- 653 Conversely, buoyancy production plays a secondary role, being at least one order of magnitude
- smaller than shear production in the first 50 meters of depth. This underscores the importance
- of including accurate current simulation when estimating the vertical turbulent mixing, which is
- 656 crucial for accurately simulating heat exchange in the water column and ultimately determining
- 657 the lake's thermal structure and ice formation.



658

659 Figure 12. Monthly averaged vertical profile of each term of the turbulence kinetic equation in the C1-1

660 (Lake3D) simulation from November 2014 to March 2015, at Spectacle Reef on Lake Huron. The change

- rate of turbulent kinetic energy (blue) is based on the 3D advection (red), and the turbulence production,
- 662 including both shear-induced production (green) and buoyancy-induced production (cyan), the
- 663 dissipation rate (black), and vertical diffusion (purple).





- Figure 13 compares the vertical temperature profiles between the standard simulation C1-1
- 665 (Lake3D) and case C2-3 (NoShearProd). The NoShearProd case shows much stronger
- stratification, particularly from January to March. The absence of shear production leads to
- significantly reduced turbulent mixing and limiting heat exchange between surface and deeper
- 668 waters, which results in a much colder surface layer (0-40 m) in January and much warmer
- deep waters (50-150 m) in February and March compared to the standard run. Consequently,
- 670 the colder surface water temperature favors ice formation, leading to overestimated ice cover in
- 671 the NoShearProd case compared to the standard simulation and observations, particularly in
- 672 January and February (Fig. 14).







- 674 **Figure 13.** Mean vertical temperature (°C) profiles from a): case C1-1 (Lake3D) standard run and b):
- 675 case C2-3 (NoShearProd) and c): their difference at Spectacle Reef on Lake Huron during November
- 676 2014-March 2015.



677

Figure 14. Spatial patterns of mean percent lake ice cover from GLSEA data (first column), case C1-1

- 679 (Lake3D; second column) and case 2-3 (NoShearProd; third column) for January 2015 (first row),
- February 2015 (second row), and March 2015 (third row).

LISSS, as true of other 1D lake models, was originally designed for small and shallow inland
lakes and was not designed to resolve water currents (Subin et al., 2012; Notaro et al., 2021).

- Some other 1D lake models (Stepanenko and Lykossov, 2005; Stepanenko et al., 2011) employ
- a crude representation of average flow fields. Therefore, 1D lake models rely on empirical or
- 685 semi-empirical relationships to estimate how wind stress affects the lake's turbulence and
- 686 mixing without explicitly resolving 3D velocity fields. These thermal diffusion-based models
- 687 often employ a latitude-dependent Ekman decay, accompanied by an empirical modification
- 688 factor, to estimate a lumped eddy diffusivity coefficient as an approximation for surface wind-
- 689 induced mixing (Xiao et al., 2016). Thus, the lack of accurate simulation of turbulent mixing





- 690 processes makes the 1D model of limited capacity in accurately simulating the Great Lakes'
- 691 thermal structure.

692 6 Summary and Conclusion

693 In summary, a two-way coupled NU-WRF/FVCOM model (CLIAv1) has been developed 694 toward the next generation of a regional climate model for the Great Lakes Basin for accurate 695 representations of lake-ice-atmosphere interactions. NU-WRF/FVCOM significantly improved 696 on the performance of NU-WRF coupled with an optimized 1D lake model, and accurately 697 reproduced the physical characteristics of the Great Lakes (e.g., LST, ice cover, and thermal 698 structure). This led to further improvements in simulated over-lake atmospheric conditions 699 (e.g., air temperature, wind, latent and sensible heat) through two-way lake-atmosphere 700 interactions.

701 While 1D column lake models have been widely used in the simulations of inland lakes 702 worldwide, small inland lakes and the Great Lakes exhibit fundamental differences in their 703 physical characteristics, such as size and depth, which in turn influence their mixing behaviors, 704 thermal structures, and circulation patterns. Inland lakes, generally much smaller (with a typical 705 average area of 1-10 kilometers) and much shallower (with a typical average depth of ~10m), 706 respond more rapidly to atmospheric conditions. This leads to a fairly uniform horizontal 707 pattern and a simpler mixing process in response to surface wind, due to their shallow depth 708 and small thermal inertia. Therefore, 1D column lake models serve as an appropriate and 709 efficient tool for simulating inland lake processes, particularly when the lake depth is shallower 710 than 20 meters. In contrast, the vast size (e.g., Lake Superior alone covers about 82,100 square 711 kilometers) and significant depth (e.g., the average depth of Lake Superior is 147 m, with a 712 maximum depth of 400 m) of the Great Lakes result in complex hydrodynamic and thermal 713 dynamics. This complexity causes the Great Lakes to exhibit many sea-like characteristics.

714 This study has highlighted key physical processes that differentiate the large, deep Great Lakes

- from small, shallow inland lakes, and how these processes impact lake simulations.
- 716 Specifically, we identified that ice dynamics, particularly ice transport, are vital in the Great
- 717 Lakes, influencing ice cover formation and heat exchange between the lake and the atmosphere.





718 Secondly, we show that advective heat transport, facilitates both lateral and vertical 719 redistribution, enables a more realistic simulation of the complex spatial temperature patterns, 720 particularly the predominance of advective heat transport in the subsurface layers. Thirdly, we 721 identified the critical role of resolving shear production in turbulent mixing in the Great Lakes, 722 which is the most influential factor that determines heat transfer and, subsequentially, lake 723 thermal structure. Ice transport, heat transfer, and shear production in turbulence mixing are 724 fundamentally linked to the 3D lake currents, which are missing or crudely represented in 1D 725 lake models. Our findings underscore that circulation currents are pivotal in the physical 726 limnology of the Great Lakes. Given the ongoing impact of climate change on these aquatic 727 systems (Zhong et al., 2016; Woolway et al., 2021; Cannon et al., 2024), accurately 728 incorporating 3D lake dynamics becomes crucial for projecting future thermal structures and 729 ecosystem effects.

730 Lastly, we acknowledge that there are multiple ways to tune the 1D lake column model or build 731 an accurate empirical relationship between atmospheric conditions and the strength of mixing 732 to improve 1D model simulations. However, the major challenge with this approach is that any 733 empirical or simplified physical relationship carries significant risks of not holding in the 734 future, especially in the context of climate change. While it may work well to calibrate the 735 model based on a substantial amount of validation data, this approach has a much larger risk 736 and lacks reliability if the model is used for climate projections where conditions change 737 significantly. Therefore, we advocate for the complete integration of 3D hydrodynamic lake 738 models in a two-way coupled fashion to project future changes in large freshwater systems. 739 This method ensures that projections are based on physical processes, reducing the risk 740 associated with empirical relationships and increasing the model's reliability for future climate 741 scenarios.

742 Code and data availability

- 743 The source codes of CLIAv1 with the two-way coupled FVCOM and NU-WRF used in this
- study are available at https://doi.org/10.5281/zenodo.12746348 (Huang, 2024a) and
- 745 <u>https://doi.org/10.5281/zenodo.12746306</u> (Huang, 2024b) respectively. The GLSEA data were
- 746 obtained from the NOAA Coastwatch website (https://coastwatch.glerl.noaa.gov/glsea/doc/)





- 747 (GLSEA, 2023). The GLEN data were from the Lake Superior Watershed Partnership website
- 748 (<u>https://superiorwatersheds.org/GLEN/</u>), with data compilation and publication provided by
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751 Author contributions

- 752 PX conceived the study. PX and CH developed the model code. PX designed the experiments.
- 753 PX, MN, XZ, CH, MBK, and CZ conducted the analyses. PX, MN, MBK, and CZ wrote the
- 754 original manuscript. All others contributed to revising the manuscript. All authors have read
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756 **Competing interests**

757 The authors declare that they have no conflict of interest.

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