

Full Coupling Between the Atmosphere, Surface and Subsurface for Integrated Hydrologic Simulation

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Key Points:

- Coupling of HydroGeoSphere (HGS) to Weather Research and Forecasting (WRF).
- Demonstration of the coupled HGS-WRF model with the California Basin.
- Comparison of the coupled HGS-WRF model to standalone WRF.

12 **Abstract**

13 An ever increasing community of earth system modelers are incorporating new physical pro-
14 cesses into numerical models. This trend is facilitated by advancements in computational
15 resources, improvements in simulation skill, and the desire to build numerical simulators that
16 represent the water cycle with greater fidelity. In this quest to develop a state-of-the-art
17 water cycle model, we coupled HydroGeoSphere (HGS), a 3D control-volume finite ele-
18 ment surface and variably-saturated subsurface flow model that includes evapotranspiration
19 processes, to the Weather Research and Forecasting (WRF) Model, a 3D finite difference
20 nonhydrostatic mesoscale atmospheric model. The two-way coupled model, referred to as
21 HGS-WRF, exchanges the actual evapotranspiration fluxes and soil saturations calculated
22 by HGS to WRF; conversely, the potential evapotranspiration and precipitation fluxes from
23 WRF are passed to HGS. The flexible HGS-WRF coupling method allows for unique meshes
24 used by each model, while maintaining mass and energy conservation between the domains.
25 Furthermore, the HGS-WRF coupling implements a sub-time stepping algorithm to mini-
26 mize computational expense. As a demonstration of HGS-WRF's capabilities, we applied it
27 to the California Basin and found a strong connection between the depth to the groundwater
28 table and the latent heat fluxes across the land surface.

1 Introduction

Numerical modelers typically subdivide the terrestrial environment into multiple domains, the atmosphere, land surface, and subsurface, and commonly simulate each one independently. The most prevalent approach is to first run the atmospheric model, and then apply its output to a hydrologic simulation. Although sequential modeling may be seen as an acceptable practice, current climate models do not rigorously handle the near surface water balance, especially in the groundwater domain, which may produce a large simulation bias once forecasts exit their calibration envelope.

Current climate models coupled to land surface models (LSMs) are often utilized to predict the risks to water resources [?]. Typically, LSMs are one-dimensional vertical columns that include shallow vertical vadose zone flow, biogeophysics, heat transport, and snow processes [?]. However, LSMs lack physics-based lateral surface/subsurface water flow, groundwater storage dynamics, and the critical feedbacks between surface and subsurface hydrology [??].

To overcome these limitations, land-surface modelers have replaced their simple one-dimensional hydrologic models with more advanced water balance approaches, although these more advanced models still contain shortcomings [???]. For example, the Weather Research and Forecasting (WRF) Model, a 3-D mesoscale nonhydrostatic atmospheric model, has been coupled to several groundwater flow models, including Noah Disturbed, a 2-D areal Boussinesq groundwater flow model [??], which fails when groundwater tables are steep (e.g. near pumping wells or around mountainous topography). In a separate study, WRF was coupled to ParFlow, a 3-D finite difference surface/subsurface flow model [?]. Additionally, ParFlow was coupled to the COSMO (Consortium for Small-scale Modeling) atmospheric model via the OASIS (Ocean-Atmosphere-Sea-Ice-Soil) coupling package [???]. However, ParFlow relies on the Noah LSM or CLM for representing the evapotranspiration process, which limits root-zones to the near-surface.

Coupled hydrologic atmospheric simulations have been applied to a large host of problems. WRF-Hydro [?] simulated a three year period over the Crati River Basin in Southern Italy, and found that the fully coupled simulation outperformed the uncoupled precipitation patterns [?]. WRF-HMS, an LSM with a lateral 2-D Boussinesq groundwater flow model [?], was applied to Poyang Lake basin (160,000 km²) and observed a 5% change in spatial precipitation patterns for coupled vs. uncoupled models [?]. WRF-ParFlow modeled the San Joaquin watershed and correlated a connection between the depth of water table with

an increased planetary boundary height [?].

A framework fully coupling the atmosphere, surface, and subsurface should have more skill than previous serial methods because the hydrologic model would replace the climate model’s shallow earth assumption, producing a dynamically linked atmosphere, surface and subsurface system. For this reason, we coupled an advanced hydrologic model, HydroGeoSphere, to a mesoscale meteorological model, the Weather Research and Forecasting Model (WRF Model Version 3.7.1, August 14, 2015) [??].

2 Numerical Models

2.1 HydroGeoSphere Terrestrial Model

HydroGeoSphere (HGS) is a fully-integrated, globally implicit, finite difference or control-volume finite element, surface and variably-saturated subsurface flow model with evapotranspiration processes, solved by a Newton-Raphson parallelized solver and an adaptive time-stepping scheme [?????]. HGS has been implemented over a large range of scales including small-scale test catchments [????], regional flow systems [???], large drainage networks [???], and continental scale basins [?]. Furthermore, HGS was adapted to include heat transport processes [??] and was used as an LSM coupled to an atmospheric boundary layer model [?].

HGS implements the non-linear three-dimensional variably-saturated Richards’ equation for subsurface flow:

$$S_w S_s \frac{\partial \psi}{\partial t} + \theta_s \frac{\partial S_w}{\partial t} = \nabla \cdot (\mathbf{K} \cdot k_r \nabla (\psi + z)) + \sum \Gamma_{ex} + Q \quad (1)$$

where Γ_{ex} [T^{-1}] is the internal fluid exchanges between domains (surface, subsurface, fractures, macropores, pumping wells, and tile drains) and Q [T^{-1}] is the external fluid exchanges (e.g. evapotranspiration, snow melt). The parameters S_s [L^{-1}], z [L], θ_s [-], and \mathbf{K} are the specific storage [LT^{-1}], elevation, saturated water content, and hydraulic conductivity, respectively. The pressure head, ψ [L], and relative permeability, k_r [-], are functions on the water saturation, S_w [-], which is approximated by lookup tables or by numerical parameterizations (e.g., ?, ?).

The two-dimensional surface domain is draped over the subsurface flow regime, and the two domains are directly linked by applying either the common node or dual-node techniques. The common node method enforces the exact same head values for each shared node, while the dual-node approach estimates a flux between the two domains. Overland

92 flow in the surface domain is based on the diffusion-wave equation:

$$\frac{\partial(d_o + z)}{\partial t} = \nabla \cdot (\mathbf{K}_o \cdot \nabla(d_o + z)) - d_o\Gamma + Q \quad (2)$$

93 which assumes mild slopes, depth-integrated velocities, and neglects inertial effects. The
 94 surface hydraulic conductivity \mathbf{K}_o [LT^{-1}] is approximated as a function of depth, d_o [L], by
 95 the Manning, Chezy, or Darcy-Weisbach equations.

96 HGS implements a process-based framework to calculate evapotranspiration (ET) based
 97 on the soil saturation, potential evapotranspiration (PET), soil type and vegetation param-
 98 eters. ET processes are an implicit component of HGS' flow simulation and are simultane-
 99 ously solved within the flow solution. Actual evapotranspiration, AET [LT^{-1}], in HGS is
 100 comprised of three components (canopy evaporation E_{can} [LT^{-1}], transpiration T_p [LT^{-1}],
 101 and bare soil (or open water) evaporation E_s [LT^{-1}):

$$AET = E_{can} + T_p + E_s \quad (3)$$

102 where each component is always positive and the sum of the components can never exceed
 103 the PET [LT^{-1}]. The transpiration and evaporation functions are implemented as:

$$T_p = f_1(LAI)f_2(S)RDF[PET - E_{can}] \quad (4)$$

104

$$E_s = \alpha^*(PET - E_{can} - T_p)EDF \quad (5)$$

105 where LAI [-] is the leaf area index, S [-] is the soil moisture content, RDF is the root
 106 density function, EDF is the energy density function, and f_1 [-], f_2 [-], and α^* [-] are fitting
 107 functions. The reader is referred to the HydroGeoSphere user manual of the processes
 108 represented in HGS as well as the numerical solution procedures employed [?].

109 **2.2 Weather Research and Forecasting Atmospheric Model**

110 The Weather Research and Forecasting (WRF) Model is a non-hydrostatic mesoscale
 111 finite difference atmospheric model. The WRF modeling suite hosts two separate dynamical
 112 cores (for the purpose of this study we used only the Advanced Research WRF (ARW) core),
 113 data assimilation, advanced physics-based parameterization, numerous radiative schemes,
 114 and multiple land surface models. The WRF model implements the terrain following flux-
 115 based Euler equations solved by the third-order Runge-Kutta temporal discretization with a
 116 second-order split-time acoustic wave. A detailed description of WRF's development can be
 117 found in the NCAR Technical Note, *A Description of the Advanced Research WRF Version*

118 \mathcal{P} [?].

119 The Noah LSM [?] is one of the most popular land surface schemes in the WRF model.
 120 It simplifies the near surface as a shallow 2 m thick series of one-dimensional columns that
 121 incorporate vadose zone hydrology and heat transport. The subsurface domain is modeled
 122 with four vertical layers that range between 10 to 100 cm thick. The benefits of using
 123 the Noah LSM coupled to WRF are that it includes vadose zone hydrology, subsurface
 124 heat transport, plant physics, and it is computationally efficient and easy to use. However,
 125 the Noah LSM fails to include three-dimensional subsurface flow, surface water flow, and
 126 saturated groundwater flow.

127 **3 Coupling Method**

128 **3.1 Spatial Coupling**

129 Atmospheric and hydrologic models are inherently different because of their drastically
 130 contrasting fluid properties, physical equations, time-scales, geometrical arrangement, and
 131 grid sizes. Hydrologic models require high spatial resolution meshes to capture the local
 132 topographic and hydrogeologic properties, with discretizations ranging from centimeters
 133 (vertical resolution in the near-surface) to kilometers (horizontal resolution in large-scale
 134 regional systems). In contrast, mesoscale atmospheric models cover a greater surface area
 135 and implement significantly coarser meshes that are between several kilometers to tens of
 136 kilometers.

137 HydroGeoSphere, a temporally implicit model, implements three types of meshing al-
 138 gorithms: finite difference hexahedra (8-point elements), finite element triangular prism
 139 (6-point elements), or finite element tetrahedra (4-point elements). However WRF, a tem-
 140 porally explicit model, only employs a regular finite difference hexahedral elemental mesh.
 141 Linking HGS to WRF required the development of a custom coupling framework that in-
 142 dependently correlates the communication of information between the two model's unique
 143 meshes.

144 The HGS-WRF coupling framework allows for independent model meshing and projec-
 145 tion characteristics by comparing the geographic coordinates (i.e., latitude and longitude)
 146 between the two domains. Our coupling method, shown in Figure 1, handles overlapping
 147 grid cells by computing the spatially-weighted area-based arithmetic mean, which maintains
 148 energy and mass conservation. The coupled model scheme internally projects each element
 149 to its geographic coordinate and then passes the data to the receiving model. This data is

150 then reprojected and interpolated to the model's mesh. For instance, the HGS model may
151 implement an Albers projection with a horizontal discretization of 4 km, while the WRF
152 simulation will use the Lambert conformal projection with a 10 km discretization.

Figure 1. Model linking framework between the HGS and WRF models. This example illustrates the passage of data (soil saturation and evapotranspiration fluxes) from HGS to WRF. Each color (green, red, blue, and orange) represents an HGS element that intersects the WRF element of interest.

153 Initially, the WRF model internally calculates the potential evapotranspiration (PET)
 154 and precipitation (I) rates and passes them to HGS:

$$PET_{hgs_j} = \frac{\sum_i^n A_i \cdot PET_{wrf_i}}{\sum_i^n A_i} \quad (6)$$

$$I_{hgs_j} = \frac{\sum_i^n A_i \cdot I_{wrf_i}}{\sum_i^n A_i} \quad (7)$$

155 where A is the overlapping elemental or cell area, and the subscripts wrf_i and hgs_j are the
 156 indices for the Weather Research and Forecasting and HydroGeoSphere models, respectively.
 157 After calculating the fluxes from WRF to HGS, our modeling framework passes HGS' actual
 158 evapotranspiration and soil saturation back to the WRF model:

$$S_{wrf_i} = \frac{\sum_j^n A_j \cdot S_{hgs_j}}{\sum_j^n A_j} \quad (8)$$

$$AET_{wrf_i} = \frac{\sum_j^n A_j \cdot AET_{hgs_j}}{\sum_j^n A_j} \quad (9)$$

159 where S is the soil water saturation and AET is the actual evapotranspiration. The current
 160 numerical implementation of HGS-WRF is for finite difference meshes, as shown in Figure 1.
 161 However, since HGS can also use unstructured element meshes (e.g. prisms or tetrahedra),
 162 Equations 6-9 can be readily adapted for the finite element method.

163 Hydrogeologic models typically use basin-divide lateral boundaries, which eliminates
 164 interflow from upstream catchments. The only water fluxes left are flows out of the basin
 165 and exchanges between the atmosphere via precipitation and evapotranspiration. Atmo-
 166 spheric models, on the other hand, implement rectangular domains that overlap multiple
 167 basins, water bodies, and political boundaries. Combining the two models together, re-
 168 quired a domain splitting algorithm that allows for separate boundaries for each individual
 169 model. The smaller HGS domain is a subset of the larger WRF simulation. Inside of the
 170 WRF model, the HGS portion overrides the internal LSM with HGS' soil saturation and
 171 evapotranspiration calculations. However, outside of the HGS portion, the WRF model uses
 172 its own land surface scheme (Noah LSM).

173 Furthermore, to aid with the linkage between models, HGS implements the same near-
 174 surface layering used in the Noah LSM. In both models the first, second, third, and fourth
 175 layers are 10, 30, 60, and 100 cm thick, respectively. The fourth layer in the Noah model
 176 is the last layer, while the HGS model further discretizes the subsurface deeper than these
 177 four upper layers. As described earlier, HGS passes its soil saturation values to Noah and
 178 overwrites the values per layer.

179 **3.2 Temporal Coupling**

180 Integrated hydrology traditionally involves time scales ranging from hourly (surface
 181 water) to millennial processes (groundwater flow). In contrast, atmospheric physics have
 182 much faster time scales and require small time steps (seconds to minutes) to capture acous-
 183 tic waves, radiative energy, and convective flow; for example, WRF recommends 6 second
 184 global time steps per kilometer of horizontal resolution (e.g. a WRF simulation with a grid
 185 spacing of 10 km would use a 60 second time step). Combining atmospheric and hydrologic
 186 models together creates a major discrepancy in time scales that need to be resolved.

187 The simplest temporal coupling method is to run both models at the same time step
 188 and directly exchange boundary information between the models at every time step. How-
 189 ever, running both models at the same time step results in wasted computational resources,
 190 because the hydrologic model’s moisture balance does not rapidly change over the course
 191 of seconds. For this reason, we implemented a sub-time stepping routine such that the
 192 atmospheric model can run at a much smaller time step, while the hydrologic model runs
 193 a coarser temporal resolution. This assumption is acceptable because WRF’s radiative en-
 194 ergy balance routine does not run every time step; the recommended WRF radiative time
 195 step is 10 times the global time step. This means that the potential evapotranspiration, a
 196 component of the radiative energy balance, is only updated during the larger time steps.

197 Our explicit temporal method, shown in Figure 2, remains first-order accurate which
 198 requires small time steps to resolve the diurnal PET forcing. HGS passes saturation and
 199 actual evapotranspiration fluxes to WRF. Meanwhile WRF transfers its potential evapo-
 200 transpiration and precipitation fluxes to HGS.

201 **3.3 Parallelization**

202 Both the HGS and WRF models are extremely complex software packages that are im-
 203 plemented with advanced numerical solvers. HydroGeoSphere implements a shared memory
 204 OpenMP approach, and is optimized for multi-core workstations [?]. The HGS paralleliza-
 205 tion efforts were focused on optimizing the matrix construction and sparse-matrix solver,
 206 while minimizing error between serial and parallel solutions (difference in head between
 207 solutions is less than 10^{-3} m). In contrast, WRF has multiple parallel choices including
 208 OpenMP, MPI (a distributed memory algorithm better suited for multi-processor cluster
 209 computing), and a hybrid OpenMP + MPI option (for shared and distributed memory
 210 computing systems).

Figure 2. HGS-WRF coupling schematic. HGS passes the saturation and actual evapotranspiration to WRF. WRF passes the precipitation and potential evapotranspiration fluxes to HGS.

211 Ensuring parallelization in HGS-WRF is not a luxury, but rather a requirement to ef-
 212 ficiently solve complex problems. Our method, which passes data between the two models,
 213 implements a quasi-parallel scheme. Each model is running as a standalone parallel process
 214 that alternate compute cycles. Initially, the WRF model will compute the PET and precip-
 215 itation fluxes and pass them to HGS. Once WRF outputs these fluxes, the WRF simulation
 216 is placed on pause until HGS computes the AET and saturation values. After HGS outputs
 217 its values to WRF, the HGS simulation is placed on pause. This cycle continues until the
 218 end of the numerical simulation. Currently the parallelization is written for both models to
 219 only run OpenMP (i.e. on Linux workstations and IBM Power Systems), but several simple
 220 additions to the parallelization routine could allow the HGS-WRF model to be extended to
 221 a hybrid scheme, where HGS uses OpenMP and WRF implements OpenMP + MPI. These
 222 hybrid simulation approaches with optimal model communications will drastically improve
 223 the computational runtimes [?].

224 **4 Model Demonstration**

225 To illustrate the capabilities of HGS-WRF, we developed a prototype California Basin
 226 Model that covers the entire state of California. The hydrogeological California Basin Model
 227 is 14 layers thick with 400,000 nodes at a 4 km resolution (see Figure 5 for the extents of
 228 the WRF and HGS-WRF domains). The HGS-WRF model was built using the HYDRO1K
 229 digital elevation model [?], STATSGO2 soil database [?], and the ? sediment thickness
 230 map. The model extends to 6,000 m below sea level, and the consolidated rock unit starts
 231 below the unconsolidated sediments calculated from ?. A 3D contiguous database of the
 232 subsurface stratigraphy was unavailable, and the deeper rocks were simply assumed to be
 233 homogeneous. For illustrative purposes, the consolidated rocks saturated-zone properties
 234 K_x , K_z , S_s , and θ_s were set to $2.5 \cdot 10^{-6}$ m/s, $2.5 \cdot 10^{-7}$ m/s, $1.0 \cdot 10^{-6}$ 1/m, and 0.05,
 235 respectively. The WRF model was discretized to a 12 km horizontal resolution with 42
 236 vertical layers (2.8 million nodes), and we implemented the ERA-Interim six-hour global
 237 reanalysis data, for the lateral boundary conditions [?].

238 Initially, the stand-alone HGS California Basin Model was spun up to current day
 239 conditions (with local water use [?]), then the coupled HGS-WRF model was executed. The
 240 coupled model ran for 10 days (January 1st to January 10th, 2011), and was computed on
 241 a 6-core Intel i7-3960X workstation with 32 GB of memory running the Ubuntu OS. WRF
 242 implemented a fixed 50 second time step, while HGS used an adaptive time step set with

243 a maximum of 300 seconds. The data was exchanged between the two models every 300
244 seconds. The 10 day model demonstration took three days to compute and the results for
245 the first day are shown in Figure 3. The plots include surface water in log-meter depth,
246 precipitation, evapotranspiration, and change in soil saturation over three time intervals.
247 Naturally, the ET undergoes a diurnal cycle due to the sun rising and setting during the day.
248 As the ET increases, it removes water from the land surface and shallow subsurface, into
249 the atmospheric domain, thus increasing the atmospheric humidity. Once water is removed
250 from the surface and subsurface, the ET rates decline because it takes more energy to move
251 deeper subsurface water into the atmosphere.

252 Conversely, as more water enters the atmosphere, the PET values decline due to a build
253 up of humidity and a decrease in atmospheric temperature. Once the atmospheric humidity
254 exceeds saturation, the water leaves the atmosphere and re-enters the surface and subsurface
255 domains as precipitation. During heavy precipitation events, water quickly moves over the
256 surface, collects into larger rivers, and discharges into the Pacific Ocean. Additionally,
257 groundwater discharge replenishes the ET water loss and can continue supplying surface
258 water flows and ET during low precipitation conditions.

259 The domain mass balance, shown in Figure 4, illustrates the inherent diurnal cycle of the
260 system. During the day, evapotranspiration rates peak to a maximum value of $12,000 \text{ m}^3/\text{s}$,
261 while at night the ET fluxes approach zero. Over the first two days, the large precipitation
262 event in Northern California drastically increases river discharge to over $1,700 \text{ m}^3/\text{s}$. For
263 the next eight days, the river fluxes follow the diurnal evapotranspiration pattern, due to
264 an increase in groundwater discharge at night from the decrease in ET.

265

Figure 3. HGS-WRF simulation for the first day. The hours are listed in UTC time.

266 **Figure 4.** HGS-WRF mass balance showing actual evapotranspiration (AET), precipitation,
267 and river outflow.

268 **4.1 Comparison to Standalone WRF**

269 The primary difference between the coupled HGS-WRF and the WRF model are the
270 inclusion of two-dimensional surface water and deep three-dimensional groundwater flow,
271 which influence the distribution of latent heat fluxes. A spatial comparison of the latent heat
272 fluxes between HGS-WRF and standalone WRF for January 2011, is shown in Figure 5.
273 Over the majority of the domain, both models use the Noah LSM and the latent heat
274 fluxes between the HGS-WRF and WRF simulations are very similar. However, within
275 the California Basin, where HGS-WRF implements the HydroGeoSphere model, there is
276 a drastic difference in evapotranspiration. In the northern regions, the HGS-WRF model
277 produces significantly more evapotranspiration than WRF, while in the southern portion of
278 California, WRF yields more ET than HGS-WRF.

279 The main cause for this contrast is the influence of groundwater, shown in Figure 6,
280 where each individual dot in the figure indicates a single node from HydroGeoSphere's
281 surface layer. The shallower the groundwater table the more negative the divergence in
282 latent heat fluxes between the two models, such that HGS-WRF produces more ET than
283 WRF. Conversely for deeper groundwater tables, WRF produces more latent heat flux
284 bias than the HGS-WRF model. Furthermore, a spatial trend in the pattern is apparent,
285 where the northern regions, shown by green and red dots, cluster towards the bottom left
286 portion (shallow groundwater table with negative LE difference). The southern regions in
287 the blues and purples are grouped towards the right and middle section of the figure (deeper
288 groundwater tables with positive LE difference). The influence of watertable depth on latent
289 heat fluxes across the land surface has been previously noted by ??.

290 **Figure 5.** Comparison between the latent heat fluxes between WRF and HGS-WRF models.
291 The entire WRF model coverage area is shown above, whereas the HGS domain is restricted to
292 the California Basin. Red regions indicate zones that the standalone WRF model produced more
293 latent heat. The blue regions are regions that HGS-WRF produced more latent heat. Simulation
294 comparison is averaged over January 2011.

Figure 6. Depth to groundwater versus latent heat flux comparison between WRF and HGS-WRF models. Each colored dot represent one HGS node with corresponding latitude. The depth to groundwater table is from the simulated HGS model results.

5 Discussion

There have been several studies that have successfully coupled groundwater flow to atmospheric models, and each of these previous coupling methods has had limitations. Currently, all coupled models require that the groundwater flow component be included over the entire land surface domain of the atmospheric model. The required atmospheric domain is much larger than the domain of interest in the hydrologic model, due to the necessity of having the atmospheric boundaries much farther away than the hydrologic domain (to eliminate influence of the model boundary). In some cases, as with the 2-D Noah Distributed groundwater model [?], including the coupled groundwater flow simulation over the entire atmospheric domain is acceptable because the addition of the saturated and linearized groundwater flow equation is such a small component of the simulation. Nonetheless, more comprehensive 3-D surface/subsurface models such as ParFlow still require the surface/subsurface domains to cover the complete atmospheric domain, which may cost additional computational resources and requires significantly more effort to construct the large-scale basin model [?].

The next main limitation that other coupling methods have is their reliance on a constrained land surface model for calculating actual evapotranspiration. Existing methods export the near surface soil moisture values (typically the first 2 meters) to the atmospheric model, then the atmospheric model's land surface routine internally calculates the evapotranspiration. However, root zones often extend past the shallow 2 m subsurface, and well-draining sandy-soils with shallow water tables may have dry near-surface conditions limiting actual evapotranspiration, in zones where high ET legitimately occurs.

The simplest method to couple atmospheric and hydrologic models is to directly overlay the meshing so that each model's node overlaps. In this method, no mass/energy interpolation is required between domains, thus simplifying continuity conditions. However, this assumption forces the atmospheric model to run at the same grid spacing as the hydrologic model. Depending on the system, either the atmospheric model will require an excessively tight model mesh (extra computational expense) or the hydrologic model will be overly coarse (not properly resolving the physical problem). In our coupling method, we implemented a custom domain-splitting framework such that each model utilizes its own separate model mesh. This allows the hydrologic and atmospheric models to use separate projection methods, different mesh resolutions, and independent numerical methods (i.e. finite difference or finite element). An additional benefit of using independent model meshes is

328 that our method can take existing HGS and WRF models that have been calibrated and
329 tuned, and then they can be quickly coupled by running the HGS-WRF code. Integrated
330 basin-scale modeling is an extremely time intensive process, which can require months to
331 develop a well-tuned model.

332 The ten-day simulation of the California Basin successfully demonstrates the strong
333 atmospheric, surface and subsurface connections within the HGS-WRF model. The diurnal
334 signal is apparent in both the evapotranspiration and river outflow signal indicating the
335 connections between the domains. Furthermore, the comparison of the coupled HGS-WRF
336 model to the standalone WRF simulation indicates a strong correlation between depth to
337 groundwater table and latent heat fluxes. A detailed explanation of the California Basin
338 Model will be provided in a subsequent paper, which will include the details of spin-up
339 process, an extended 200 day HGS-WRF simulation, and a comprehensive description of
340 the setup of the HGS portion of the California Basin.

341 **5.1 Future Work**

342 Deep subsurface heat transport was not included in the current version of the HGS-
343 WRF coupling, although HGS has the capability to simulate heat transport over the land
344 surface and subsurface. The water balance was the primary focus for the first version of
345 HGS-WRF, and it was desired to minimize the number of independent variables in the
346 coupling scheme to reduce computational demand. In future releases, it is planned to incor-
347 porate heat transport coupling as an option in the HGS-WRF model. From experience with
348 the HGS-ABL (atmospheric boundary layer) model, the depth of the subsurface may play a
349 critical role for temperature regulation, especially during prolonged drought conditions [?].

350 The current HGS-WRF model also does not include snow accumulation and melting,
351 and sediment freeze-thaw processes would be the next logical advancement to the coupled
352 model. Currently, HGS-WRF treats all water as liquid precipitation, which may artificially
353 increase stream flow during the winter months and decrease streamflow during the summer.
354 HydroGeoSphere already incorporates snowmelt and soil freeze-thaw, and the WRF simu-
355 lation would provide the solid-phase precipitation to simulate winter processes.

356

357 **Acknowledgments**

358 Jason H. Davison was partially supported by an Ontario Early Researcher Award awarded
359 to John C. Lin and by the Natural Sciences and Engineering Research Council of Canada
360 (NSERC) awarded to Edward A. Sudicky. Support was also provided by the Canada Re-
361 search Chair held by Edward A. Sudicky. The data used to create this study is listed in the
362 references and in Section 4 Model Demonstration.