



1 **Analyzing the impact of groundwater flow and storage changes on Budyko**
2 **relationships across the continental US**

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10 **Abstract:**

11 We use a transient, integrated groundwater surface water simulation of the majority of
12 the continental US to evaluate the impacts of groundwater surface water exchanges on
13 Budyko relationships for roughly 25,000 nested watersheds ranging in sized from 100
14 km² to over 3,000,000 km². The simulation spans a single water year with varying
15 degrees of storage change and trans-watershed lateral flow throughout the domain.
16 This provides a test bed for evaluating the role of realistic groundwater storage changes
17 across a range of climates and physical settings within a rigorous computational
18 framework in which every watershed's water balance component can be fully defined.
19 Budyko curves are compared using (1) the standard simplification where
20 evapotranspiration is calculated as the difference between precipitation and outflow,
21 (2) the case in which evapotranspiration is directly measured (or simulated) and (3) a
22 modified approach where groundwater contributions are accounted for with the
23 incoming precipitation. Results show that for all three cases, large watersheds still
24 converge around the Budyko curve despite the fact that the equilibrium assumption is
25 violated. However, the best results are found when groundwater interactions are
26 directly accounted for. While the two standard approaches for inferring or directly
27 measuring evapotranspiration still generate Budyko type behavior, the parameters of
28 the curves vary systematically with groundwater contributions and the relationship
29 between curve parameters and groundwater surface water exchanges is fundamentally



30 different between approaches. This results in contradictory spatial patterns depending
31 on the calculation method and highlights the importance of evaluating uncertainty in
32 the equilibrium assumption for any Budyko analysis, especially those that seek to
33 attribute variability in curve parameters to physical characteristics. Finally, the shape
34 parameter is evaluated as a function of scale and large, continental scale basins are
35 shown to converge onto a single Budyko relationship.

36

37 **1. Introduction:**

38 The Budyko hypothesis states that the fraction of precipitation (P) that leaves a
39 watershed through evapotranspiration (E), as opposed to runoff, can be well predicted
40 by the aridity of the watershed [Budyko, 1958; 1974]. Budyko [1974] compared long-
41 term evapotranspiration fractions to aridity for 1,200 large watersheds around the globe
42 and showed that 90% of the variance in evapotranspiration ratio (E/P) could be
43 described by a single empirical curvilinear equation dependent only on aridity, often
44 referred to as the “Budyko Curve”. Budyko noted that this consistent relationship is a
45 reflection of the dominance of macroclimate over large drainage areas and long time
46 periods where it can be assumed that a watershed is in steady state.

47 The simplicity of this relationship has since garnered much interest within the
48 hydrologic community for its potential to predict watershed behavior using only climate
49 variables, which are often easier to observe than many hydrologic variables, and
50 without relying on computationally expensive or heavily parameterized numerical
51 models. In recent years the Budyko hypothesis has also been put forward as a way of
52 predicting hydrologic sensitivity to climate change especially in ungauged basins [e.g.
53 Randall J. Donohue et al., 2011; Jones et al., 2012; Renner et al., 2014]. However,
54 application of this method has been partially limited by spatial variability between
55 watersheds and the required steady state assumption.

56 The original Budyko curve presented a universal relationship between
57 evapotranspiration and aridity [Budyko, 1974]. Subsequent work has shown that, while
58 the Budyko curve is generally robust, the shape of the curve can vary between locations



59 and climate alone is not sufficient to predict watershed partitioning, especially for
60 smaller watersheds. Differences in behavior between river basins have been attributed
61 to seasonal lags in water and energy supply, vegetative and soil properties [R. J.
62 *Donohue et al.*, 2007]. The original Budyko curve has been reformulated multiple times
63 to incorporate additional free parameters [Choudhury, 1999; Fu, 1981; Milly, 1994; L.
64 *Zhang et al.*, 2001; L. *Zhang et al.*, 2004], and numerous studies have used these
65 modified formulations to relate curve parameters to physical basin characteristics in
66 many settings [e.g. *Li et al.*, 2013; *Shao et al.*, 2012; *Williams et al.*, 2012; *Xu et al.*, 2013;
67 *Yang et al.*, 2009]. For example, *Li et al.* [2013] and *Yang et al.* [2009] evaluated
68 relationships between the shape of the Budyko curve and vegetation coverage.
69 Similarly, *Williams et al.* [2012] and L. *Zhang et al.* [2004] found distinct shape
70 parameters when comparing forested watersheds to grasslands, although it should be
71 noted that they reached the opposite conclusion about their relative magnitudes.
72 Others have focused on the role of soil moisture and noted differences in behavior
73 based on plant water availability and seasonal lags in supply and demand [e.g. *Milly*,
74 1994; *Yang et al.*, 2007; *Yokoo et al.*, 2008].

75 Many previous studies have demonstrated good predictive abilities using
76 modified Budyko formulations even when applied to smaller watersheds and shorter
77 time scales than those originally intended. However, poor performance in some
78 locations, especially over annual or seasonal time periods, has been attributed to the
79 influence of storage changes that violate the steady state assumption [*Milly and Dunne*,
80 2002; *Lu Zhang et al.*, 2008]. *Istanbulluoglu et al.* [2012] and *T Wang et al.* [2009]
81 showed interannual storage changes can produce a negative correlation between
82 evapotranspiration ratio and aridity that is counter to the Budyko curve for baseflow
83 dominated basins in the Nebraska Sand Hills. *D Wang* [2012] evaluated inter-annual
84 storage changes for twelve watersheds in Illinois and showed that, on an annual
85 timescale, variability in runoff and storage is larger than evapotranspiration, and
86 accounting for storage can improve the performance of Budyko predictions. *Du et al.*
87 [2016] presented a method for explicitly accounting for storage changes within the



88 Budyko framework and demonstrated that this approach can greatly improve
89 performance in arid regions or over shorter time scales where the steady state
90 assumption is not valid. These studies all indicate the potential importance of
91 groundwater surface water interactions within the Budyko framework, however they
92 are based on a limited number of catchments. Further research is needed to evaluate
93 the impact of storage changes across spatial scales and physical settings and to provide
94 a more general understanding of this behavior.

95 Here we evaluate the role of groundwater surface water exchanges in the
96 Budyko framework using an physically based hydrologic model to directly simulate the
97 integrated groundwater surface water system over a large spatial domain at high
98 resolution. This approach provides detailed quantitative information about subsurface
99 fluxes, which has been a limiting factor for previous research on this topic. For example,
100 in some previous studies, the impact of groundwater storage changes have been
101 inferred from variability around the Budyko relationship, but the storage changes
102 themselves were not measured [*Milly and Dunne, 2002; Lu Zhang et al., 2008*]. Others
103 have addressed interactions more directly using baseflow separation techniques that
104 require only streamflow observations [*T Wang et al., 2009*] or lumped watershed
105 models that parameterize baseflow and recharge [*Du et al., 2016*]. However, with both
106 of these approaches the groundwater system is still not directly simulated or observed.
107 *Istanbulluoglu et al. [2012]* and *D Wang [2012]* did use observations of water table
108 depth to directly quantify storage changes and demonstrate the impact of this change
109 within the Budyko framework; but the study areas with this approach were relatively
110 limited (four watersheds for *Istanbulluoglu et al. [2012]*, and twelve for *D Wang [2012]*).
111 Groundwater observations sufficient to precisely characterize watershed storage
112 changes are not widely available. Therefore, for this analysis we rely on simulated
113 watersheds to evaluate the role of storage changes across a broad range of spatial
114 scales and physiographic settings with different Budyko approaches.

115

116



117 2. Methods

118 2.1 The use of hydrologic modeling for Budyko Analysis

119 Previous work has evaluated the Budyko curve using hydrologic models of varying levels
120 of complexity. The abcd model employed by *Du et al.* [2016], among others, is a lumped
121 water balance model that includes baseflow and groundwater recharge using calibrated
122 parameters. *Yokoo et al.* [2008] used a different water balance model with a more
123 complex groundwater formulation that includes saturated and unsaturated zones, but
124 the authors noted limitations in simulating infiltration excess overland flow with this
125 approach. *Gentine et al.* [2012] applied a water balance model that includes a soil
126 bucket and can simulate infiltration excess overland flow; however it did not include
127 topography and was only applied at the plot scale. While these approaches do account
128 for storage in the subsurface and varying levels of complexity in groundwater surface
129 water exchanges, they all take a lumped approach and rely on calibrated parameters
130 that are not physically based.

131 Increasing in sophistication, *Troch et al.* [2013] used a semi-distributed model
132 that included shallow perched aquifers as well as root zone and soil moisture dynamics;
133 and *Koster and Suarez* [1999] evaluated a global circulation model that simulated land
134 surface and atmospheric processes using physically based equations. The trade-off for
135 increasing model sophistication is computational expense especially for large high-
136 resolution domains. *Koster and Suarez* [1999] used a global simulation but at low spatial
137 resolution (4° by 5°), while *Troch et al.* [2013] limited their analysis to the hillslope scale.
138 Furthermore, both of these approaches are focused on the land surface and shallow
139 subsurface and neither included lateral groundwater flow.

140 To the authors' knowledge, no one has evaluated Budyko behavior over large
141 spatial scales using a hydrologic model that integrates lateral groundwater flow with
142 surface processes. So called, integrated hydrologic models that incorporate physically
143 based lateral groundwater flow with overland flow and land surface processes are a
144 relatively new development in hydrologic modeling. These tools are ideal for capturing
145 dynamic behavior and interactions throughout the terrestrial hydrologic cycle and they



146 have been increasingly applied over the last decade. To achieve this level of complexity
 147 requires significant computational resources, and detailed model inputs. These
 148 requirements have generally limited the application of integrated tools to regional scale
 149 domains. Continental-scale high-resolution simulations have only recently become
 150 technically feasible.

151 **2.2 Integrated Hydrologic Model**

152 For this analysis, we use the first high-resolution integrated groundwater surface water
 153 simulation of the majority of the continental US (CONUS) [Maxwell and Condon, 2016;
 154 Maxwell et al., 2015]. The CONUS simulation was developed using the integrated
 155 hydrologic model ParFlow-CLM [Kollet and Maxwell, 2006; 2008; Maxwell and Miller,
 156 2005]. ParFlow simulates three-dimensional variably saturated groundwater flow using
 157 Richards' equation:

$$158 \quad S_s S(\psi_p) \frac{\delta \psi_p}{\delta t} + \phi \frac{\delta S(\psi_p)}{\delta t} = \nabla \cdot [-K_s(x) k_r(\psi_p) \cdot \nabla(\psi_p - z)] + q_s \quad (1)$$

160

161

162 where S_s is the specific storage [L^{-1}], S is the relative permeability [-], which varies with
 163 pressure head ψ_p [L] based on the *Van Genuchten* [1980] relationships, t is time [T], ϕ is
 164 the porosity of the subsurface [-], $K_s(x)$ is the saturated hydraulic conductivity tensor [LT^{-1}],
 165 k_r is the relative permeability [-], which also varies with pressure head according to
 166 the *Van Genuchten* [1980] relationships, z is the depth below the surface [L] and q_s is a
 167 source/sink term [T^{-1}].

168 Overland flow is included in the groundwater flux term of Eq. (1) (i.e. in the first
 169 term on the right hand side) using a free surface overland flow boundary condition that
 170 applies continuity of pressure and flux across the boundary between the land surface
 171 and the subsurface. Overland flow is solved using the kinematic wave approximation of
 172 the momentum equation where the diffusion terms are neglected and it is assumed that
 173 the bed slope, S_o [-] is equivalent to the friction slope. Flow varies as a function of
 174 ponded depth according to Manning's equation:



175

$$v = \frac{\sqrt{s_0}}{n} \psi_p^{2/3} \quad (2)$$

177 where n [$\text{TL}^{-1/3}$] is the Manning's roughness coefficient. Using this approach ParFlow is
178 able to solve variably saturated groundwater flow and overland flow simultaneously.

179 Practically this means that (1) the location of surface water bodies does not need to be
180 specified a priori and will develop wherever water ponds in the domain, and (2) two-
181 way groundwater surface water exchanges can evolve dynamically based on head
182 gradients and subsurface properties.

183 ParFlow is also coupled with a land surface model derived from the Common
184 Land Model (CLM) [Dai *et al.*, 2003]. In the combined ParFlow-CLM model [Kollet and
185 Maxwell, 2008], ParFlow solves the water balance in the subsurface and CLM solves the
186 combined water energy balance at the land surface. At the land surface, the energy
187 balance (R_{net}) is comprised of sensible (H), latent (LE) and Ground (G) heat fluxes [Wm^{-2}]:

188

$$R_{net} = H + LE + G \quad (3)$$

190

191 All of the energy fluxes listed in Eq. (3) vary with soil moisture. CLM uses
192 pressure head and saturation values for the upper subsurface layers (in this case the top
193 2m) simulated by ParFlow and passes infiltration fluxes back to ParFlow. Land surface
194 processes are also driven by atmospheric forcing variables, which are provided as inputs
195 to the model. Forcing variables include short and longwave radiation, precipitation, air
196 temperature, atmospheric pressure, specific humidity and wind. Using these inputs,
197 CLM simulates multiple land surface processes including canopy interception,
198 evaporation from the canopy and the ground surface, plant transpiration, ground and
199 sensible heat fluxes as well as snow dynamics.

200 This study focuses on simulated evapotranspiration E [LT^{-1}], which is the sum of
201 evaporation, E_v and plant transpiration T . CLM uses a mass transfer approach with
202 mean variables where evaporation is calculated using the gradient between the specific
203 humidity at the ground surface, q_g [MM^{-1}], and the specific humidity at a reference



204 height, q_a [MM⁻¹], scaled by a soil resistance factor β [-], air density ρ_a and the
205 atmospheric resistance, r_d [-] as follows:

206

$$207 \quad Ev = -\beta \rho_a \frac{q_g - q_a}{r_d} \quad (4)$$

208

209

210 The soil resistance factor is calculated based on the saturation relative to the residual
211 saturation and the saturation in the uppermost soil column (refer to *Jefferson and*
212 *Maxwell* [2015] for the complete formulation).

213 Similarly transpiration is calculated by scaling the potential evapotranspiration to
214 account for stomatal and aerodynamic resistance as follows

215

$$216 \quad T = (R_{pp,dry} + L_w) L_{SAI} \left(\rho_a \frac{q_{sat} - q_a}{r_d} \right) \quad (5)$$

217

218 Here $R_{pp,dry}$ [-] is a scaling parameter, L_w [-] is the fraction of the canopy that is covered in
219 water, L_{SAI} is leaf and stem area index and q_{sat} [-] is the saturated specific humidity [mm-
220 1]. $R_{pp,dry}$ is a function of light and moisture limitations. Parameters that are used to
221 determine leaf area index, reflectance and transmittance and root distributions vary by
222 land cover type and are provided as inputs to the model using the 18 land cover classes
223 defined by the International Geosphere Biosphere Program (IGBP). For additional
224 details on the numerical approach and analysis on the sensitivity of evaporation and
225 transpiration within CLM the reader is referred to [*Ferguson et al.*, 2016; *Jefferson and*
226 *Maxwell*, 2015; *Kollet and Maxwell*, 2008; *Maxwell and Condon*, 2016].

227

228 **2.3 Model domain and simulations**

229 The analysis presented here is based on a previously developed transient ParFlow-CLM
230 simulation of the majority of the Continental US (CONUS) documented in *Maxwell and*
231 *Condon* [2016]. The CONUS domain covers the majority of eight major river basins,
232 shown in Fig. 1 and spans roughly 6.3M square kilometers at 1km lateral resolution. As



233 detailed in in *Maxwell and Condon* [2016] and *Maxwell et al.* [2015], the model extends
234 102 m below the subsurface with five vertical layers that contour to the land surface
235 using a terrain following grid formulation [*Maxwell*, 2013]. The vertical resolution of the
236 domain decreases with depth to better resolve the shallow subsurface. Layer
237 thicknesses are 0.1, 0.3, 0.6, 1 and 100m moving from the land surface down. Spatially
238 heterogeneous physical parameters for the subsurface include porosity, saturated
239 hydraulic conductivity and van Genuchten parameters. Subsurface spatial units were
240 determined using a national permeability map developed by *Gleeson et al.* [2011] for
241 the bottom 100 m of the domain and the soil survey geographic database (SSURGO) for
242 the top two meters. Maps of the subsurface units and their properties are available in
243 *Maxwell and Condon* [2016] and *Maxwell et al.* [2015]. The land surface was derived
244 from the Hydrologic data and maps based on the Shuttle Elevation Derivatives at
245 multiple Scales (HydroSHEDS) digital elevation model using a topographic processing
246 algorithm to ensure fully connected drainage network [*Barnes et al.*, 2016]. Vegetation
247 types were extracted from the USGS land cover dataset using the IGBP land cover
248 classifications.

249 The model was first initialized to a steady state groundwater configuration using
250 the ParFlow model without CLM starting from a completely dry domain and providing a
251 constant recharge forcing over the land surface to achieve a dynamic equilibrium.

252 Development of this steady state simulation and evaluation of the resulting
253 groundwater configuration are provided in [*Condon and Maxwell*, 2015; *Condon et al.*,
254 2015; *Maxwell et al.*, 2015]. Using the steady state groundwater configuration as a
255 starting point, the coupled ParFlow-CLM model was used to simulate the fully transient
256 system including land surface processes for water year 1985 (i.e. Oct. 1, 1984 through
257 Sep., 30 1985), which was chosen as it is the most climatologically average within the
258 past 30 years. The transient simulation was driven by historical hourly meteorological
259 forcings from the North American Land Data Assimilation System Phase 2 (NLDAS 2)
260 [*Cosgrove et al.*, 2003; *Mitchell et al.*, 2004]. Anthropogenic activities such as
261 groundwater pumping and surface water storage are not included in the transient



262 simulation. Therefore the simulation represents a pre-development interpretation of
263 water year 1985, which is ideal for Budyko analysis. Complete details of the
264 development of the transient simulation are available in *Maxwell and Condon* [2016].

265 The integrated physically based approach described above requires significant
266 computational resources. The benefit of this computationally intensive approach is that
267 it provides high-resolution (1 km^2) gridded outputs that fully define water and energy
268 fluxes from the groundwater through the land surface without calibration and with a
269 minimal number of empirical parameters. The transient simulation has been validated
270 against publically available observations for the water year 1985 period. This includes
271 transient observations at varying frequencies from 3,050 stream gauges, 29,385
272 groundwater wells and 378 snow stations for a total of roughly 1.2 million comparisons
273 points. Flux tower observations were not available over this period, but latent heat
274 fluxes were also compared to the Modern Era Retrospective-analysis for Research and
275 Application (MERRA) dataset. Complete details of the model validation are provided in
276 the supplemental information of *Maxwell and Condon* [2016]. Although there are of
277 course limitations to the model, overall comparisons between simulated and observed
278 values demonstrate that the modeling approach is robust; stream-flow timing and
279 magnitude are generally well matched in undeveloped basins, snowpack timing and
280 melt is accurate and spatial patterns in latent heat flux are reasonable.

281

282 **2.4 Water Balance**

283 Within ParFlow-CLM, incoming precipitation can infiltrate to the subsurface, contribute
284 to runoff or pond. Evaporation occurs from ponded water, bare soil and canopy
285 interception. Additionally, roots pull water from the subsurface to support transpiration
286 for plants and lateral groundwater flow redistributes moisture within the subsurface
287 and can further support overland flow. All of these processes occur within every 1 km^2
288 grid cell in the domain. The focus of this work is on watershed function and therefore
289 the gridded results are aggregated to more hydrologically relevant units. The domain is
290 divided into 33,454 subbasins each containing a single stream. Subbasin areas, outlined



291 in Fig. 1, vary but are generally on the order of 100 km². The total drainage area for
292 every subbasin, henceforth referred to as the watershed, is defined by tracing up the
293 river network to encompass the entire upstream contributing area. This results in
294 33,454 nested watersheds ranging in drainage area from about one hundred square
295 kilometers to over three million. For all of the following analysis we will focus on the
296 24,235 watersheds that are contained within the highlighted regions of Fig. 1. Similarly,
297 while the simulation uses an hourly time step, here we evaluate annual values.

298 At the watershed scale, precipitation P [L³] is balance by surface water outflows,
299 Q_{out} [L³], evapotranspiration, E [L³], and groundwater surface water exchanges, referred
300 to as groundwater contributions, G [L³].

301

$$302 \quad P = Q_{out} + E + G \quad (6)$$

303

304 Equivalently this can be expressed in terms of ratios relative to incoming precipitation
305 where the sum of the outflow ratios sum to one:

306

$$307 \quad 1 = \frac{Q_{out}}{P} + \frac{E}{P} + \frac{G}{P} \quad (7)$$

308

309 As noted above, every watershed fully encompasses its contributing area, and
310 therefore surface water inflows are zero. P is the sum of the gridded annual
311 precipitation over the drainage area. Every watershed is defined to have a single outlet
312 point. Q_{out} is the overland flow calculated hourly at the outlet using the ponded water
313 depth and Eq. (2) and summed over the simulation period. E is the total evaporation and
314 transpiration simulated by ParFlow-CLM summed fore every grid cell in the drainage
315 area over the year. Groundwater contributions are calculated by using Eq. (6) as, $P - Q_{out}$
316 - E . This approach results in a perfect water balance for every watershed. However, it is
317 important to note that here G encompasses both exchanges between groundwater and
318 surface water, which can be either positive fluxes from the surface to the subsurface or
319 negative fluxes from subsurface to the surface, as well as changes in surface water



320 storage. The assumption within this notation is that, over the annual simulation,
321 changes in ponded water are small relative to groundwater surface water exchanges.
322 Therefore we refer to the combined value as groundwater contributions and follow the
323 convention that a positive groundwater contribution denotes water that is infiltrating
324 from the land surface to the subsurface whereas a negative value indicates groundwater
325 discharge which can either occur from groundwater supported E or baseflow
326 contributions to streams.

327 In addition to the simulated evapotranspiration (E), potential evaporation E_p is
328 calculated using Eq. (4), the hourly meteorological forcing data used to drive the
329 simulations (air temperature, atmospheric pressure, specific humidity, and wind speed),
330 and simulated ground temperatures in the uppermost layer of the model. To calculate
331 potential evaporation, as opposed E , the β parameter is set to one to eliminate soil
332 resistance and q_g is the saturated specific humidity calculated based on the ground
333 temperature and atmospheric pressure. As with E , hourly gridded E_p values are
334 summed over the entire simulation period for every watershed drainage area.

335 Fig. 2 maps the aridity index (E_p/P) as well as each component of the water
336 balance from Eq. (7) expressed as ratios of precipitation. Subplots b and c show regional
337 trends in the relative importance of evapotranspiration as opposed to overland flow. In
338 the more arid western portions of the domain (shown in red on subplot a), Q_{out} is small
339 compared to E whereas in the more humid eastern portions of the domain (blue and
340 orange values in subplot a) the relative magnitude of Q_{out} increases. Within this annual
341 simulation, subplot d shows that groundwater surface water exchanges can be a
342 substantial portion of the water balance in much of the domain, indicating than an
343 assumption of steady state is not valid for this simulation. The importance of
344 groundwater contributions over this time frame is not surprising because the steady
345 state assumption is generally applied over much longer time periods. The purpose of
346 this work, however, is not to evaluate the conditions under which a steady state
347 assumption will be valid or to predict changes in storage. Rather, we are taking
348 advantage of the ability to directly calculate groundwater surface water exchanges



349 within the simulation to evaluate the impact of these exchanges across a range of
350 spatial scales and climates.

351 The groundwater contribution ratio map also illustrates the importance of lateral
352 groundwater flow at multiple spatial scales within the system. Groundwater storage
353 gains (i.e. positive values of G/P) are prevalent in the western arid portion of the domain
354 and groundwater discharge to surface water is more common in the humid eastern
355 portion of the domain. Within large basins like the Missouri, positive groundwater
356 contributions occur in the headwater regions and transitions to negative values
357 downstream. This is an illustration of lateral groundwater convergence and regional
358 flow systems. Note that results are mapped by subbasin, but all water balance
359 calculations are carried out for the complete watershed draining to a subbasin outlet.
360 Therefore, Fig. 2 should be viewed as a system of nested subbasins with values
361 representing progressively larger drainage areas as you move downstream. With this in
362 mind, it is also intuitive that some of the largest groundwater contribution ratios occur
363 in headwater basins while in downstream reaches on major rivers the values are smaller
364 indicating a regional balance between local groundwater surface water exchanges.

365

366 **2.5 Budyko analysis**

367 We evaluate watershed behavior within the Budyko framework across a wide range of
368 scales using the fully defined water balance components from the simulation, and the
369 aridity index values calculated above. Budyko's original formulation expressed
370 evapotranspiration ratio as a function of aridity index (E_p/P) as follows [Budyko, 1974]:

371

$$372 \quad \frac{E}{P} = \left\{ \frac{E_p}{P} \left[1 - \exp \left(-\frac{E_p}{P} \right) \tanh \left(\frac{P}{E_p} \right) \right] \right\}^{0.5} \quad (8)$$

373

374 Although the original analysis by Budyko did show some scatter around the curve, Eq.
375 (8) defines a universal relationship that does not include any free parameters to account
376 for spatial differences [Budyko, 1974]. Subsequent work has observed systematic
377 variability between watersheds that can be related to climate, land cover and soil



378 properties [e.g. *R. J. Donohue et al.*, 2007]. To reflect this, the original universal Budyko
379 formulation has been refined multiple times to include additional free parameters
380 [*Choudhury*, 1999; *Fu*, 1981; *Milly*, 1994; *L. Zhang et al.*, 2001; *L. Zhang et al.*, 2004]. For
381 a summary of these formulations refer to *Du et al.* [2016] and *L. Zhang et al.* [2004].

382 Here we apply the commonly used Budyko formulation from *Fu* [1981] and *L.*
383 *Zhang et al.* [2004]:

384

$$385 \quad \frac{E}{P} = 1 + \frac{E_p}{P} - \left(1 + \left(\frac{E_p}{P} \right)^\omega \right)^{1/\omega} \quad (9)$$

386

387 Eq. (9) includes one free parameter, ω which can range from one to infinity, henceforth
388 referred to as the shape parameter. ω is an empirical parameter that has not been
389 ascribed a specific physical meaning, but is generally conceptualized as an integrated
390 catchment property that reflects characteristics such as land cover, soil properties,
391 topography and seasonality [*L. Zhang et al.*, 2004]. Fig. 3 plots Eq. (8) for a range of ω
392 values. The bold line ($\omega=2.6$) is roughly equivalent to the original Budyko equation (Eq.
393 8) [*L. Zhang et al.*, 2004]. Following the original Budyko assumption of no change in
394 storage, in humid locations where potential evaporation is less than precipitation, the
395 system is energy limited and the maximum value of E is E_p . Conversely, when the aridity
396 index is greater than one evapotranspiration the system is water limited and the
397 maximum E/P value is one (indicating that all incoming precipitation is evaporated). As
398 the shape parameter increases the curves move progressively closer to the water
399 ($E/P=1$) and energy ($E/P=E_p/P$) limitations of the system.

400

401 **3. Results and discussion**

402 **3.1 Assessing the impacts of groundwater contributions using multiple water balance** 403 **approaches**

404 Fig. 3 follows the standard Budyko assumption that any incoming precipitation that does
405 not contribute to evapotranspiration will leave the watershed as overland flow (i.e. E/P
406 $+ Q_{out}/P=1$). This is equivalent to assuming no change in storage, or zero groundwater



407 contribution in Eq. (7). However, this may not always be the case for many reasons. For
408 example, in the simulation evaluated here, the groundwater contribution fractions (G/P)
409 mapped in Fig. 2d illustrate that groundwater is an important contributor to the water
410 balance in many parts of the domain over the one year simulation period; often the
411 groundwater contribution is greater than 10% of precipitation especially in small
412 headwater basins. In this section we evaluate the impact of assuming that a basin is in
413 equilibrium when groundwater contribution ratios are not zero using three different
414 approaches to estimate the evapotranspiration ratio.

415 Precipitation and runoff are generally much easier to measure at the watershed
416 scale than evapotranspiration or groundwater storage changes. Therefore, in many
417 Budyko analyses evapotranspiration is not actually measured directly, but is calculated
418 as the difference between precipitation and surface outflow [e.g. Greve *et al.*, 2015;
419 *Istanbulluoglu et al.*, 2012; Jones *et al.*, 2012; Renner *et al.*, 2014; T Wang *et al.*, 2009;
420 Xu *et al.*, 2013; Yang *et al.*, 2009]. This approach relies on the assumption that changes
421 in storage are negligible. We refer to this as the *inferred evapotranspiration* approach
422 and mimic it by approximating the evapotranspiration ratio as simulated $(P - Q_{out})/P$. The
423 inferred approach is compared to the less common *direct evapotranspiration* approach
424 where Budyko relationships are evaluated using observed evapotranspiration (or in this
425 case the evapotranspiration that is simulated by our integrated hydrologic model). For
426 the direct evapotranspiration approach Budyko relationships are evaluated using
427 simulated E/P .

428 An alternate method that takes groundwater into account directly is to replace
429 precipitation with effective precipitation, defined as precipitation minus groundwater
430 contribution ($P - G$). This is the approach that was used by Du *et al.* [2016] in their study
431 of Budyko relationships in arid basins. It should be noted here that the first two
432 approaches (i.e. inferred and direct evapotranspiration) are commonly used in analyses
433 that use the standard equilibrium assumption while the final method is designed for
434 situations where this is not the case. By comparing results between all three, we



435 consider the impact of nonzero groundwater contributions both for approaches that
436 assume it is negligible, and for when it is accounted for.

437 Fig. 4 plots every watershed in the domain shown in Fig. 1 using the three
438 approaches. All figures follow the Budyko curves, demonstrating that Budyko
439 relationships are recreated with the integrated hydrologic model. However, there are
440 some notable differences between methods. For example, with the inferred E approach
441 shown in subplot a, the points are focused near the water limit line (i.e. $(P - Q_{out}) / P = 1$)
442 for high aridity values. Conversely, with the direct approach (subplot b), the
443 evapotranspiration ratios are generally lower at high aridity values. Also, with the direct
444 approach, there are points with evapotranspiration ratios are greater than one and fall
445 above the water limit. This would appear to violate the water balance and will be
446 discussed more later. When groundwater contributions are adjusted for using the
447 effective precipitation approach show in subplot c, the points collapse more closely onto
448 the Budyko curves.

449 Variability between approaches can be explained by the differences in the way
450 groundwater contributions influence each method. This is illustrated in Fig. 5. In cases
451 where G is zero, Fig. 5 demonstrates that the three formulations are equivalent. This is
452 not the case when groundwater contributions are nonzero. When evapotranspiration is
453 inferred from $P - Q_{out}$ [as in: Greve *et al.*, 2015; Istanbuloglu *et al.*, 2012; Jones *et al.*,
454 2012; Renner *et al.*, 2014; T Wang *et al.*, 2009; Xu *et al.*, 2013; Yang *et al.*, 2009], Eq. (7)
455 shows that the result is actually $E + G$ (and not E) so instead of evaluating,

456

457
$$\frac{E}{P} = f\left(\frac{E_p}{P}\right) \quad (10)$$

458

459 as intended in the Budyko formulation, this approach actually results in

460

461

462
$$\left(\frac{E}{P} + \frac{G}{P}\right) = f\left(\frac{E_p}{P}\right) \quad (11)$$

463



464 The direct approach avoids the limitations of the inferred approach by evaluating
465 the evapotranspiration as intended in Eq. (10). However, groundwater storage changes
466 will still influence the resulting behavior because the partitioning shown in Fig. 5b is now
467 between evaporation and runoff plus groundwater contributions, not just runoff. This
468 means that the maximum evapotranspiration fraction (i.e. the upper water limit) is not
469 one, but one minus the groundwater contribution fraction. This shift in the limit
470 explains the values greater than one in Fig. 4b; in these watersheds groundwater
471 contributions are negative (i.e. groundwater supplying water to the land surface) and
472 this allows for evapotranspiration values that are greater than the incoming
473 precipitation. Similar shifts in the upper limits of the system for arid locations were
474 found by *Potter and Zhang* [2009] who noted that evapotranspiration was actually
475 approaching a fixed portion of potential evapotranspiration for high rainfall years in arid
476 basins in Australia. Finally, in the effective precipitation case, the intent is to maintain
477 focus on partitioning between evapotranspiration and overland flow so groundwater
478 contributions are removed from the denominator of both ratios. This ensures that the
479 modified outflow and evapotranspiration ratios will sum to one even when groundwater
480 surface water exchanges are occurring.

481 For all three approaches we calculate the shape parameter for every point in the
482 domain using Eq. (9). For the direct evapotranspiration approach we use Eq. (9) directly.
483 For the inferred evapotranspiration approach we substitute in $(P-Q_{out})/P$ for E/P . For the
484 effective precipitation approach we substitute in $E/(P-G)$ for E/P and $Ep/(P-G)$ for Ep/P .
485 Fig. 6 a-c plot the resulting shape parameters as a function of groundwater contribution
486 fraction colored by aridity. Both the inferred (6a) and direct approaches (6b) show
487 clear, but contradictory, relationships with groundwater surface water exchanges. There
488 is a positive relationship between the shape parameter and groundwater contribution
489 fraction for the inferred evapotranspiration approach at the lower limits of the system
490 as delineated by the dashed line in Fig. 6a. Because the groundwater contribution is
491 included in the evapotranspiration fraction when evapotranspiration is inferred from
492 precipitation and runoff (i.e. $P-Q_{out} = E + G$), nonzero groundwater contributions will



493 contribute directly to the evapotranspiration fraction as shown in Fig. 5a. If a constant
494 positive groundwater contribution is applied across aridity values this would result in a
495 vertical shift the Budyko curve relative to a zero groundwater contribution system as
496 illustrated in Fig. 6d. Moving vertically within the Budyko plot results in higher shape
497 parameters (as shown in Fig. 3). Therefore, the vertical shifting shown in Fig. 6d will
498 result in a positive correlation between groundwater contribution fraction and shape
499 parameter.

500 When the direct evapotranspiration approach is applied (6b), groundwater
501 contributions are now lumped in with the outflow fraction (Fig. 5b). In this case, positive
502 groundwater contributions can be conceptualized as shifting the limits of what is
503 available for evapotranspiration. This is illustrated by the shifting limit lines in Fig. 6e
504 where the upper water limitation on the system shifts from 1 to $1-G/P$ and the energy
505 limitation shifts from E_p/P to $E_p/P-G/P$. This creates a nonlinear inverse relationship
506 between curve number and groundwater contributions. As the groundwater
507 contribution fraction increases, the decreasing upper bounds on evapotranspiration
508 fraction will bias the system towards lower curve numbers (refer to Fig. 3). This is a
509 nonlinear relationship which can be shown by calculating the shape parameter as a
510 function of groundwater contribution fraction in Eq. (9) for the limiting case where there
511 is no outflow (i.e. $G/P=1-E/P$). The dashed line on Fig. 6b shows the resulting
512 relationship for a relatively high aridity value of 6. The curve provides a good
513 approximation for the upper limit of Fig. 5b.

514 The effective precipitation case shows the least correlation with groundwater
515 contribution fraction. This is to be expected because the purpose of this approach is to
516 adjust for groundwater contributions before calculating shape parameters. However, it
517 should be noted that some dependence on groundwater contribution is still to be
518 expected the extent that groundwater surface water exchanges are also correlated with
519 other watershed properties. This is true for the other approaches too. While the effect
520 of groundwater contributions within each space can be precisely determined using Eq.
521 (7) and Eq. (9), it is important to note that the watersheds evaluated here are also



522 heterogeneous in land cover, topography and seasonality. Therefore, in the scatter plots
523 shown in Fig. 6, the relationships between shape parameters and groundwater
524 contribution explained by subplots d and e appear as limits rather than strong
525 predictors. This point is also made by *Istanbulluoglu et al.* [2012] who used the inferred
526 evapotranspiration approach. They provide a similar conceptual model to Fig. 6d for the
527 shifting effect of storage changes on behavior. However, for the four basins in Nebraska
528 that they evaluated they found a negative relationship between inferred
529 evapotranspiration ratios and aridity that was attributed to a strong negative correlation
530 between groundwater contribution fraction and aridity index.

531 Fig. 7 compares the shape parameters calculated with each approach to
532 illustrate the way that different assumptions can bias derived Budyko relationships.
533 Subplot a shows the differences between the inferred and direct evapotranspiration
534 approaches, which are commonly used in studies that assume no change in storage.
535 Because groundwater contributions are incorporated into different components of the
536 water balance with these methods Fig. 7a shows that, for positive groundwater
537 contributions (green points), the inferred shape parameters are systematically higher
538 than the direct shape parameters, while the inverse is true when groundwater
539 contributions are negative (purple points). Furthermore, when groundwater
540 contributions are large (i.e. the dark green circles in subplot a), the direct method has
541 uniformly low shape parameters, but the inferred method still shows a range of shape
542 parameters. This is to be expected from Figs. 6 b and e that show that for high
543 groundwater contributions the upper limit of the evapotranspiration ratio is limited by
544 surface water 'losses' to groundwater and as a result all of the points will be pushed to
545 low curve numbers.

546 The direct and inferred evapotranspiration methods are also compared to the
547 effective precipitation approach, which does account for groundwater contributions
548 (Fig.7 b &c). As would be expected, the direct and inferred methods have inverse
549 biases; shape parameters are systematically higher with the inferred approach relative
550 effective precipitation and lower for the direct approach. Here too the trends with



551 groundwater contributions are reversed with positive contributions creating a positive
552 bias for the inferred case and a negative bias for the effective precipitation case. Also,
553 there is a much stronger correlation between curve numbers between the inferred
554 evapotranspiration and effective precipitation approaches than with the direct
555 evapotranspiration approaches. This is partially due to the loss of curve number
556 sensitivity for high groundwater contribution fractions with the direct approach which is
557 also shown in subplot a. For all three cases, Fig. 7 demonstrates systematic variability in
558 the shape parameter even for relatively small groundwater contributions.

559

560 **3.2 Spatial patterns and scaling**

561 The influence of catchment characteristics such as vegetative properties, topography
562 and climate on Budyko relationships has been well established by previous studies[e.g.
563 *Li et al.*, 2013; *Milly*, 1994; *Shao et al.*, 2012; *Williams et al.*, 2012; *Xu et al.*, 2013; *Yang*
564 *et al.*, 2009; *Yokoo et al.*, 2008]. As such, spatial variability in shape parameters is to be
565 expected even after adjusting for groundwater contributions. Fig. 8 maps shape
566 parameters for the nested watersheds calculated using three different approaches. All
567 three maps demonstrate local variability and regional trends in the shape parameters.

568 Spatial patterns are consistent between approaches in the more humid eastern
569 portion of the domain, where groundwater contribution ratios are generally smaller
570 (Fig. 2d), but in the more arid western portion of the domain significant differences are
571 observed. For both the inferred evapotranspiration and effective precipitation
572 approaches there are large red areas indicating curve numbers greater than four where
573 the evapotranspiration ratio is falling very close to the water limitation. The areas with
574 curve numbers greater than four are generally consistent between the inferred
575 evapotranspiration and effective precipitation approaches, but the inferred
576 precipitation approach results in higher curve numbers throughout the western portion
577 of the domain than the effective precipitation approach. This is consistent with the
578 biases shown in Fig. 5b. Conversely, with the direct evapotranspiration approach the
579 western portion of the domain has much lower shape parameters and less spatial



580 variability. Again, this finding is consistent with Fig. 6b and e, which show that when
581 groundwater contributions are high, the curve numbers are uniformly low because the
582 flux from the surface water system to the groundwater shifts the upper limit of the
583 evapotranspiration fraction down. The systematic differences in Fig. 8, both with
584 respect to the shape parameter values and the spatial patterns, where groundwater
585 surface water exchanges are occurring indicate the potential to arrive at fundamentally
586 different conclusions depending on what approach is used.

587 *Xu et al.* [2013] built a neural network model to predict shape parameters using
588 long-term observations from 224 watersheds with drainage areas ranging from 100 to
589 10,000 km². They then predicted shape parameters globally using a variety of catchment
590 characteristics. Excluding the small drainage areas with shape parameters greater than
591 four, the spatial patterns calculated here with the effective precipitation approach
592 match well with the global map presented by *Xu et al.* [2013]. Even though the one-year
593 transient simulation used for the analysis presented does not meet the equilibrium
594 criteria, Figs. 4c and 8c show that realistic Budyko relationships are still found when
595 groundwater contributions are accounted for using the effective precipitation approach.

596 Furthermore, for our system we can compare results by drainage area. Budyko
597 originally limited analysis to large basins (which he defined as drainage areas greater
598 than 10,000 km²) where he argued that macroclimate can be expected to dominate
599 partitioning [*Budyko, 1974*]. Indeed subsequent work has shown that for smaller areas
600 vegetation dynamics becoming increasingly important [*R. J. Donohue et al., 2007*]. Fig. 9
601 plots Budyko relationships for every watershed grouped by drainage area using the
602 effective precipitation formulation as an example. In this figure, the drainage area is
603 increased from watersheds less than 100 km² (9a) to watersheds greater than 100,000
604 km² (9d). This figure shows that the scatter decreases as drainage area increases and
605 the points converge around a single curve. This behavior illustrates increased
606 importance of local watershed characteristics for smaller drainage areas consistent with
607 previous studies.



608 This is further demonstrated by Fig. 10, which shows the interquartile range of
609 shape parameters for each approach with increasing drainage area. In all three cases,
610 we observe a decrease in variance moving from small to large basins and a trend of
611 decreasing shape parameters with increasing area. Again this indicates increased
612 importance of watershed characteristics at smaller scales. In the case of the inferred
613 and direct evapotranspiration approaches, because groundwater contributions are not
614 accounted for in the calculations, some of this variability can also be attributed to
615 spatial patterns in groundwater surface water exchanges and lateral groundwater flow.
616 The inferred evapotranspiration and effective precipitation approaches are the most
617 similar while the direct evapotranspiration formulation is systematically lower and less
618 variable (this is also consistent with the maps shown in Fig. 9). For the largest drainage
619 areas, the median shape parameter is 1.8 for the inferred evapotranspiration approach,
620 1.5 for the direct evapotranspiration approach and 1.7 using effective precipitation.

621 The shape parameters estimated with the effective precipitation approach are
622 arguably the most relevant to other long-term studies that have assumed equilibrium
623 conditions. Our simulated median value is slightly lower than the original Budyko value
624 of 2.6 and the median value of 2.56 found by [Greve *et al.*, 2015] using the 411 Model
625 Parameter Estimation Experiment (MOPEX) catchments in the US. However, it compares
626 well with 1.8 median value for large MOPEX basins in the US reported by Xu *et al.*
627 [2013]. Although, it should be noted that Xu *et al.* [2013] report a higher 2.6 median
628 value for small basins, and the median small basin value reported found here is 2.0.
629 Part of this bias can likely be attributed to the concentration of MOPEX basins in the
630 eastern portion of the US where Fig. 9 shows that shape parameters are generally
631 higher. Overall, the consistency in spatial patterns and convergence around the Budyko
632 curve for large drainage areas indicates that the ParFlow-CLM model recreates Budyko
633 relationships even over a relatively short annual time. However, for smaller watersheds
634 variability in catchment characteristics is still an important consideration.

635

636 **4. Summary and Conclusions**



637 For this study, we use a one-year integrated groundwater surface water simulation of
638 the majority of the continental US to evaluate Budyko relationships. We focus on the
639 implications of assuming equilibrium in systems where we can demonstrate that
640 systematic groundwater surface water exchanges are occurring over many spatial
641 scales. Using the integrated modeling approach allows us to directly calculate every
642 component of the water balance. Taking advantage of this we recreate multiple
643 approaches used in Budyko analysis: (1) inferring evapotranspiration from precipitation
644 and runoff assuming no storage changes, (2) using the evapotranspiration directly
645 simulated by the model but not accounting for groundwater contributions in the Budyko
646 formulation and (3) accounting for groundwater contributions using a modified
647 approach where the total available water is calculated by subtracting groundwater
648 contributions from precipitation to determine ‘effective precipitation’.

649 Results show that with both the inferred and direct evapotranspiration
650 approaches even small groundwater contribution fractions can influence the result.
651 Over large watersheds, the two evapotranspiration methods converged to different
652 shape parameters, 1.8 for inferred evapotranspiration and 1.5 for direct
653 evapotranspiration. We also demonstrate that the same groundwater contribution ratio
654 will alter Budyko curve differently depending on whether evapotranspiration is
655 evaluated directly or inferred. With the inferred evapotranspiration approach,
656 groundwater contributions are inherently included in the evapotranspiration ratio, so it
657 shifts points vertically on the Budyko plot; whereas when evapotranspiration is used
658 directly, groundwater contributions essentially shift the limits of what is available to be
659 partitioned into overland flow or evapotranspiration.

660 This fundamental difference means that comparisons between basins can be
661 significantly altered depending on what approach is used if groundwater surface water
662 exchanges are occurring. We show that for both methods the shape parameter varies as
663 a function of groundwater contribution ratio, especially in arid watersheds. As a result,
664 the two approaches produce different spatial patterns in shape parameters. In both
665 cases it is simple to demonstrate how a constant groundwater contribution value will



666 shift the Budyko curve for all aridity values. However, because groundwater surface
667 water exchanges are spatially heterogeneous and depend on aridity as well as other
668 watershed characteristics, the impact of variable groundwater contributions across a
669 large sample of watersheds is difficult to predict. For example, some recent studies that
670 have evaluated the impact of groundwater storage changes in a small number of arid
671 basins have shown that the correlation between recharge to groundwater and aridity
672 leads to an inverse relationship between evapotranspiration fraction and aridity
673 [*Istanbulluoglu et al.*, 2012; *T Wang et al.*, 2009]. For larger studies that include a more
674 diverse set of watersheds correlations between groundwater contribution ratios may
675 not be as clear; still, our results indicate that groundwater contributions are an
676 important consideration when evaluating scatter between watersheds within the
677 Budyko framework.

678 Using the effective precipitation approach demonstrated by *Du et al.* [2016] we
679 are able to directly account for groundwater contributions within the Budyko
680 framework. Results with this formulation demonstrate that consistent Budyko
681 relationships can be developed even when groundwater surface water exchanges violate
682 the steady state assumption, as long as they can be quantified and appropriately
683 adjusted for. After accounting for groundwater contributions, there is still significant
684 variability in the shape parameters between small watersheds, which is to be expected
685 given the other heterogeneities in watershed characteristics across the domain. We find
686 that for larger basins, the points converge around a single curve. Interestingly, the two
687 methods that do not account for groundwater contributions also converge around a
688 single curve for large watersheds although they result in different shape parameters.

689 One of the primary assumptions of the Budyko hypothesis is that watersheds are
690 in equilibrium and there are no changes in storage. This means that all incoming
691 precipitation will either leave the watershed as evapotranspiration or overland flow.
692 While the Budyko curve has been well verified with observations from around the globe,
693 it is now widely accepted that the relationship between evapotranspiration ratios and
694 aridity indices is not universal and that some additional curve parameters are need to



695 account for spatial variability between watersheds. Many subsequent studies have
696 related curve parameters from various Budyko formulations to catchment properties
697 such as vegetation, topography and seasonality. However, very few studies evaluate the
698 validity of the equilibrium assumption over their study period. Furthermore, because
699 long-term evapotranspiration estimates at the watershed scale are seldom available
700 many infer that evapotranspiration is the difference between precipitation and outflow.

701 Using an integrated model, we demonstrate that changes in groundwater
702 storage and lateral flow can also contribute to variability between watersheds in
703 systematic ways and that different approaches based on the equilibrium assumption will
704 yield different results when this assumption is not met. While these results indicate the
705 potentially significant implications of ignoring groundwater contributions within Budyko
706 analyses, it should be noted that both groundwater surface water exchanges and
707 evapotranspiration are difficult to measure over large scales and long-term records
708 similar to stream gauges are often not available. We acknowledge that the common
709 approach of developing a water balance using precipitation and streamflow alone is
710 generally necessitated by data availability not ignorance. For this study, we take
711 advantage of a sophisticated hydrologic simulation that allows us to define every flux in
712 the domain and balance water perfectly, but this is not the case for most observational
713 studies. Furthermore, even within this fully defined approach, we acknowledge that
714 even the most sophisticated numerical model is a gross simplification of the natural
715 system. Therefore, the intention of these comparisons is not to discredit previous
716 approaches, rather to illustrate the potential impacts of incorrectly assuming
717 equilibrium conditions across a broad range of physiographic settings and spatial scales.
718 Our results show that even when changes in storage are occurring, large watersheds still
719 roughly follow Budyko curve; however the shape parameter and scatter will vary with
720 groundwater contribution and depending on how evapotranspiration is quantified. This
721 indicates that, even if groundwater contributions are not included directly into analyses,
722 it is vital provide some quantification of uncertainty in water balance closure in order to



723 determine whether differences in shape parameters are actually resulting from unique
724 basin characteristics or uncertainty in storage.

725

726 **Data Availability:**

727 All data from this analysis are available upon request. Instructions for accessing the
728 ParFlow simulations used here are provided in [Maxwell *et al.*, 2016].

729

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737

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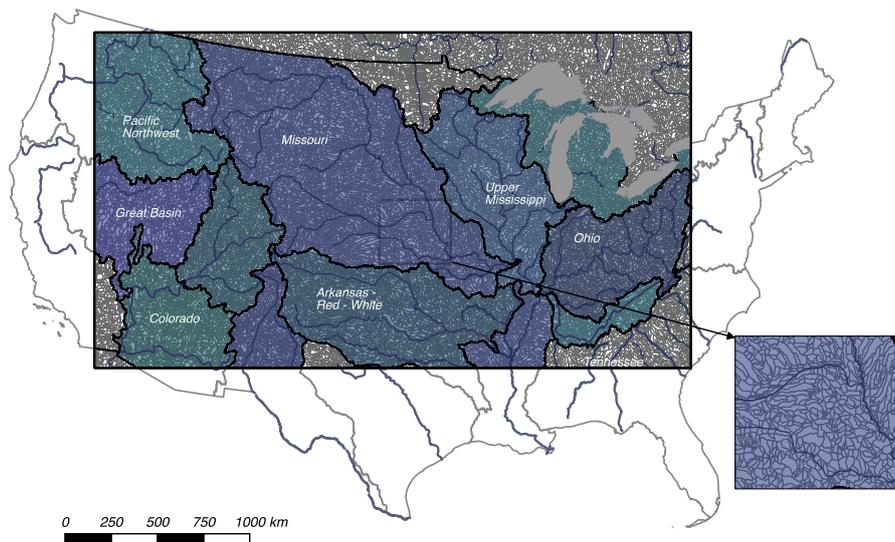
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898 **Figures:**

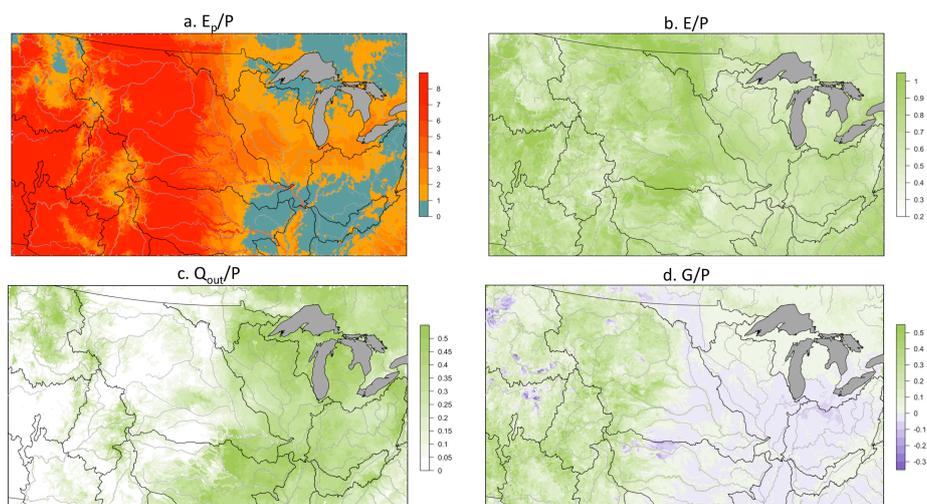


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900 **Figure 1:** Map of the simulation domain extent (black box) with major river basins
901 highlighted and labeled. Subbasins within the domain are outlined in grey. Major rivers
902 are show in blue for reference (Note that the simulated river network is much more
903 highly resolved as illustrated in [Maxwell et al., 2015])

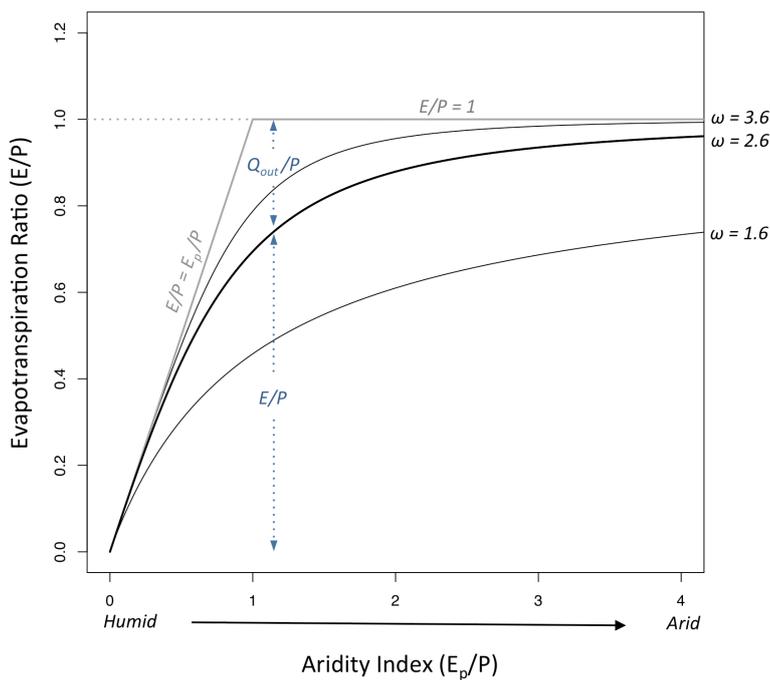
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907 **Figure 2:** Maps of (a) aridity index (E_p/P) and the ratios of (b) evapotranspiration (c)
908 outflow and (d) groundwater contributions compared to precipitation. Major river
909 basins are outlined in black. Note that ratios are mapped according to the subbasins
910 shown in Fig. 1 but the values reflect the water balance for the entire watershed. This is
911 a system of nested watersheds so the value for each watershed is reported at its outlet
912 subbasin.



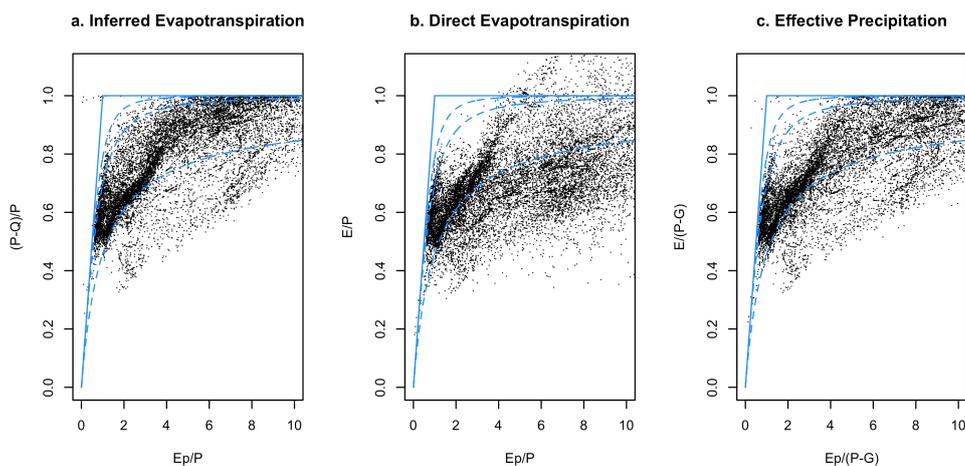
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914 **Figure 3:** Illustration of the Budkyyo framework showing curves with three different
915 shape parameters (black lines, $\omega=1.6, 2.6$ and 3.6) in relation to the water ($E/P=1$) and
916 energy ($E/P=E_p/P$) limits of the system, grey lines.

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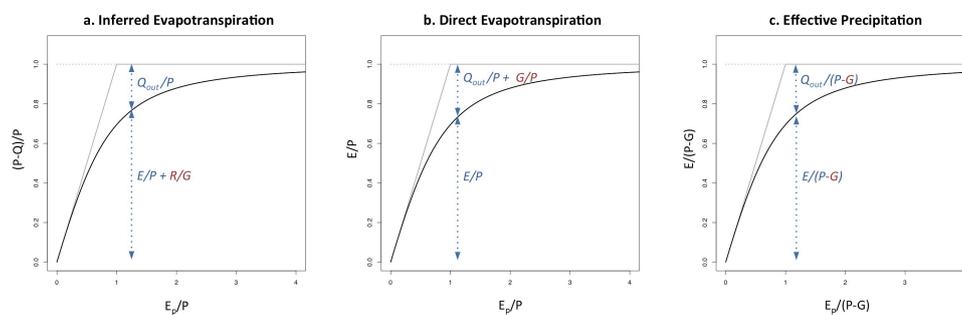


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920 **Figure 4:** Budyko plots for the three approaches (a) inferred evapotranspiration, (b)
921 direct evapotranspiration and (c) effective precipitation with points for every watershed
922 in the domain. Dashed blue lines are Budkyko curves with ω values of 1.6, 2.6 and 3.6
923 and the solid blue lines are the water and energy limits (refer to Fig. 3).

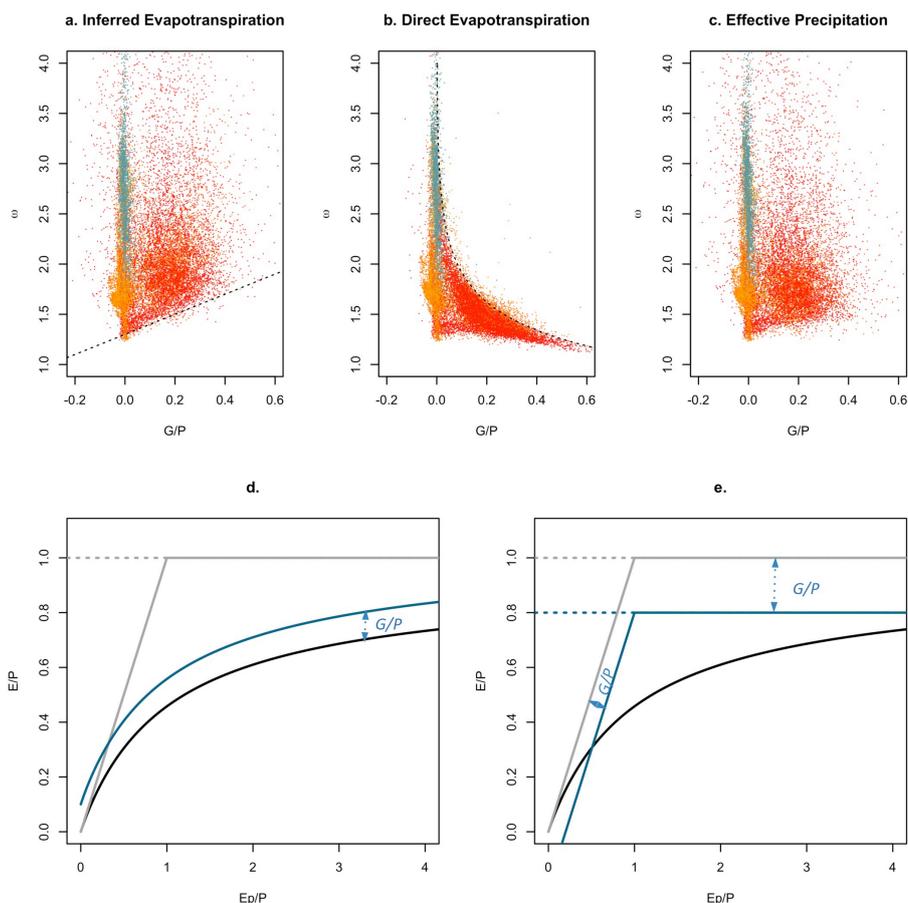


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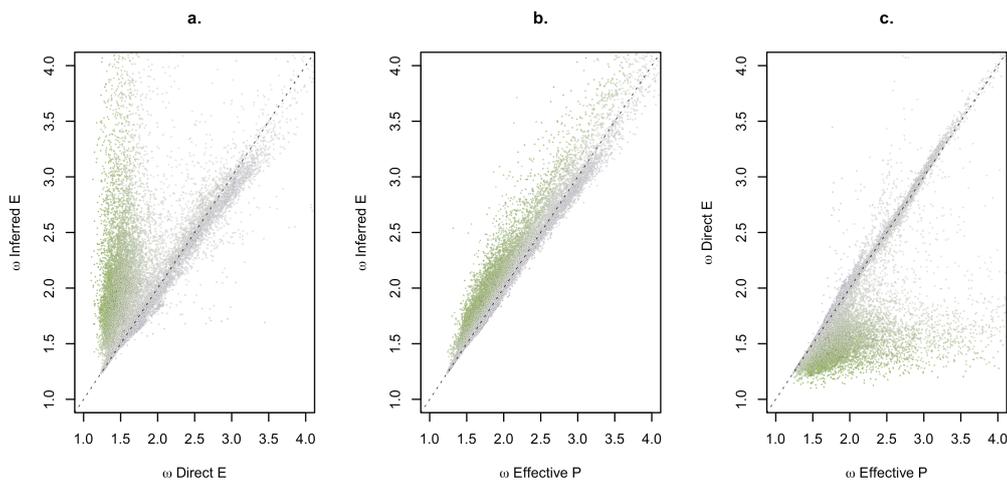
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926 **Figure 5:** Illustration of the treatment of groundwater contributions for each of the
927 three approaches. The black lines show the water and energy limits and an example
928 Budyko curve, similar to Fig. 3. Arrows indicate the water balance component
929 represented above and below the curve in each case.



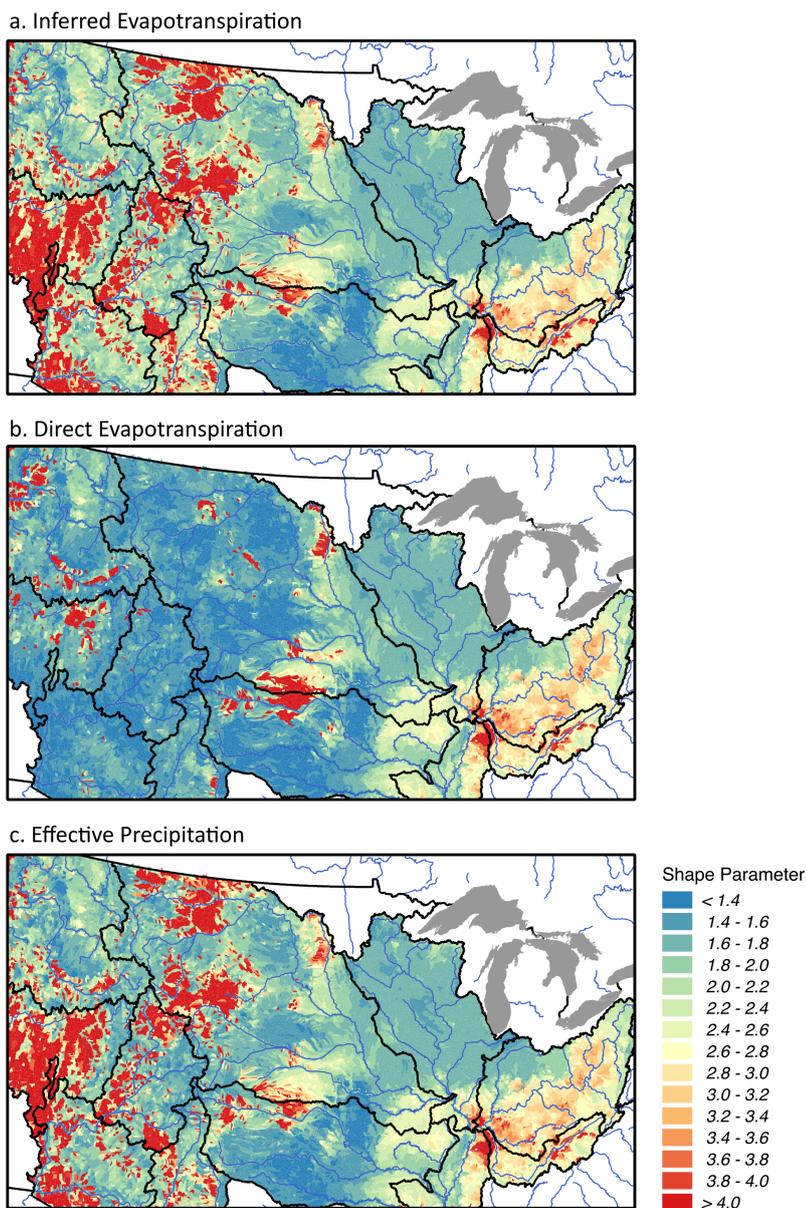
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931 **Figure 6:** Comparison of shape parameters to groundwater contribution ratios for the
 932 three approaches in every watershed (a-c). Points are colored by aridity as shown in Fig.
 933 2a. A dashed line with a slope of one is included on (a) for reference. The dashed line on
 934 (b) shows the relationship between the shape parameter and groundwater contribution
 935 fraction for an example aridity value of six in the limiting case where outflow is zero. The
 936 conceptual figures below illustrate the impact of a positive groundwater contribution
 937 (i.e. a net flux from the surface to the subsurface) for (d) the inferred evapotranspiration
 938 and (e) direct evapotranspiration approaches.



939

940 **Figure 7:** Comparison of shape parameters between the three approaches for every
941 watershed. Points are colored by groundwater contribution fraction as shown in Fig. 2d.
942 The dashed line on each plot is a one to one line for reference.



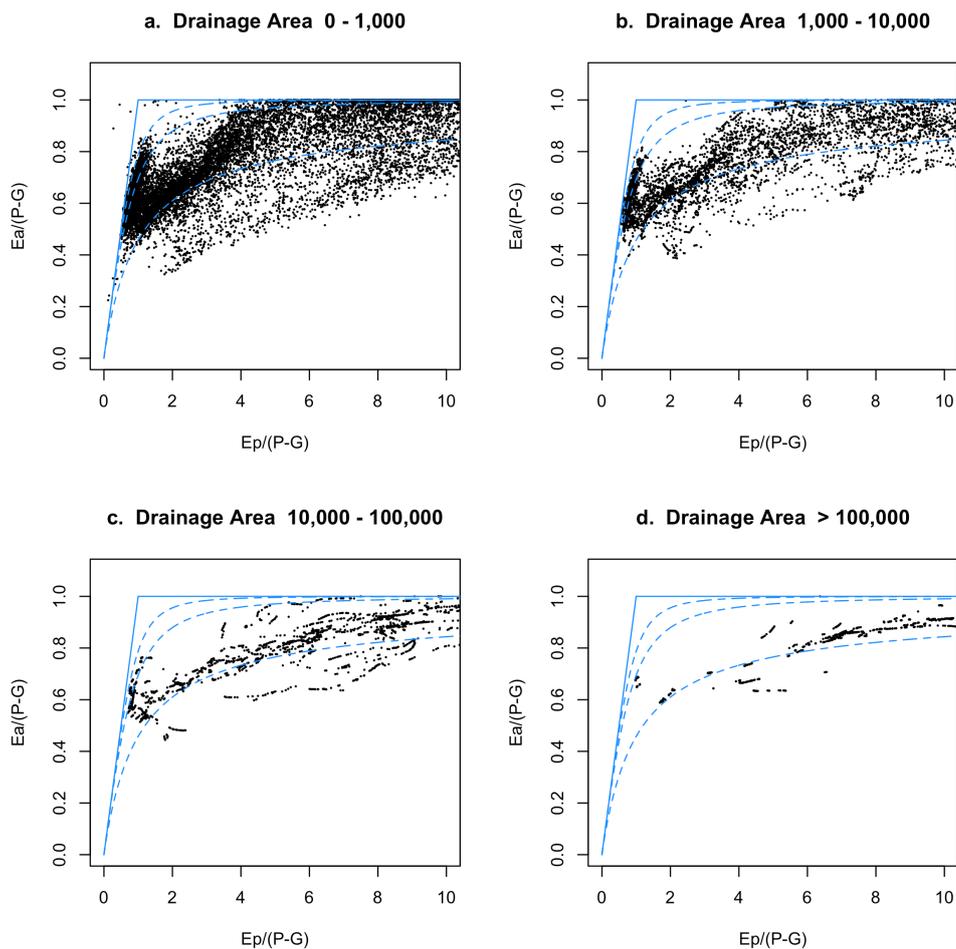
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944 **Figure 8:** Map of shape parameters calculated for the 24,235 nested watersheds using

945 the (a) inferred evapotranspiration, (b) direct evapotranspiration and (c) effective

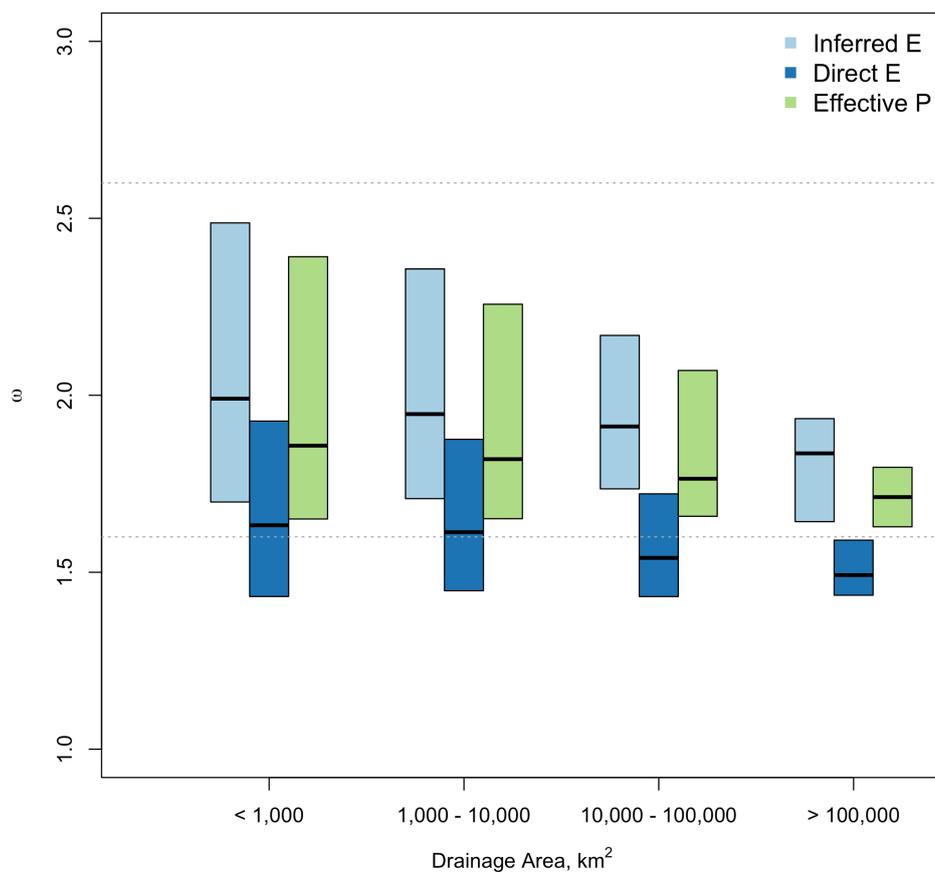
946 precipitation. Major rivers are outlined in blue and regional boundaries in black.

947



948

949 **Figure 9:** Budyko plots of evapotranspiration ratio versus aridity index using the
950 effective precipitation method with watersheds grouped by drainage area [km²]. Blue
951 dashed lines are Budkyo curves with shape parameters of 1.6, 2.6 and 3.6 (refer to Fig.
952 3) and the solid blue lines show the water and energy limits.



953

954 **Figure 10:** Boxplots showing the interquartile range (i.e. 25-75th percentile values) of
955 shape parameters for all three approaches grouped by drainage area. Dashed lines are
956 at 1.6 and 2.6 for reference.

957