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1 **Assessing the water balance of the Upper Rhine Graben**
2 **hydrosystem**

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15

16 **Abstract**

17 The Upper Rhine alluvial aquifer is an important transboundary water resource. However, as
18 in many alluvial systems, the aquifer inflows and outflows are not precisely known, due to the
19 difficulty in estimating the river infiltration flux and the boundary subsurface flow. To
20 provide a thorough representation of the aquifer system, a coupled surface-subsurface model
21 was applied on the whole aquifer basin, and several parameter sets were tested to investigate
22 the uncertainty due to poorly known parameters (e.g. aquifer transmissivity computed by an
23 inverse model, river bed characteristics). Twelve simulations were run and analysed using
24 standard statistical criteria, as well as a more advanced statistical method, the Karhunen

25 Loève Transform (KLT). This analysis showed that although the model performs reasonably
26 well, some piezometric levels underestimations persist in the south of the basin. An accurate
27 representation of the aquifer behaviour requires taking into account river infiltration and the
28 functioning of irrigation canals in the Hardt area. It also appeared that increasing the
29 maximum river infiltration flow leads to altered results.

30 River infiltration to the aquifer was estimated to represent about 80 % of the aquifer inflows
31 with a mean annual value around $115 \pm 16.5 \text{ m}^3/\text{s}$, therefore with an uncertainty of 14 %.
32 This quantity is larger than estimated in previous studies, but also in agreement with some
33 results obtained during low water periods. This important conclusion highlights the
34 vulnerability of the Upper Rhine Graben aquifer to pollution from the rivers and to climate
35 change since it is highly probable that the rivers' regime from the neighbouring mountain
36 ranges will be affected by a reduced snow cover.

37

38 **Keywords:** groundwater-surface water relations; groundwater recharge/water budget; Upper
39 Rhine alluvial aquifer; hydrogeological model parameters sensitivity.

40 **1 Introduction**

41 Alluvial hydrosystems have been recognised as important water resources but also very
42 vulnerable systems to pollution or change in water availability (Allen et al., 2004).
43 Knowledge of the different components of the water balance is therefore of significant
44 importance to optimize water management and/or to estimate the impacts of anthropogenic
45 activities or climate changes. These water balance components are difficult to measure, as the
46 observations of water fluxes provide only local information (Sanford 2002, Kalbus et al.,
47 2006). Thus there is a need to assess these fluxes by modelling, as they are linked to the
48 boundary conditions, the sink/source terms and the hydrodynamic parameters, which are
49 mostly unknown and estimated through model calibration. Model calibration leads to non-

50 unique solutions and therefore, the evaluation of some components of the water balance is
51 uncertain (Konikow and Bredehoeft 1992, Beven 2006).

52 The components of the water balance for unconfined groundwater are recharge due to
53 precipitations, exchanges with surface water, in/out-flows through the boundaries (either
54 prescribed heads or prescribed fluxes), storage in the aquifer, and water uptake by pumping
55 wells. Recharge and surface-groundwater interactions constitute key processes in the
56 behaviour and the evolution of these systems (Sanford, 2002, Woessner, 2000, Sophocleous,
57 2002). These fluxes can be estimated by fully coupled surface-subsurface models including
58 the unsaturated zone such as ParFlow (Kollet and Maxwell, 2006), Hydrogeosphere (Therrien
59 et al. 2007), and CATHY (Camporese et al., 2008) among others. However, applications of
60 these fully coupled models are generally limited to small areas. For larger systems, coupled
61 surface-subsurface models are usually simplified (Hu et al. 2007, Rushton 2007) and the
62 recharge is estimated separately using simplified hydrological models based on 1D Richards
63 equation (e.g. MIKE SHE, Reefsgard 1997) or more conceptual schemes (e.g. MARTHE,
64 Noyer and Elsass 2006).

65 This work aims at estimating the different components of the water balance and its associated
66 uncertainties for the Upper Rhine aquifer. The studied hydrosystem is one of the most
67 important water resources in Western Europe, and several regional scale models were
68 previously developed, as for example in the LIFE project in 1996 (LfU 1996, LfU 2005), and
69 in the INTERREG project Monit (LUBW 2006) focusing on nitrate contamination of
70 groundwater. These two studies led to different findings about aquifer recharge: river
71 infiltration was found to be the main recharge component in LIFE, whereas in Monit it was
72 effective rainfall. However, these discrepancies may be linked to the fact that the models were
73 limited to the alluvial plain, and used simple methods to assess the subsurface flows from the
74 Vosges and the Black Forest surrounding mountains.

75 In the present work a modelling of the whole Rhine basin between Basel and Lauterbourg is
76 performed (figure 1). This enables to better constrain the lateral water inputs to the aquifer,
77 and thus, to have a homogenous representation of the whole Upper Rhine graben basin.
78 Moreover it allows assessing the impacts of climate change on the hydrosystem, which could
79 lead to altered conditions over the mountainous catchments.

80 The Upper Rhine alluvial aquifer groundwater dynamics is modelled using different
81 assumptions and calibration methods based on trial and errors or inverse approach. The use of
82 several likely parameter values for a given model leads to different likely water balances and
83 provides an estimate of the associated uncertainty. Since several possible calibrated models
84 are obtained, we associated classic statistical criteria (average error, Nash criteria) with a
85 more advanced statistical method, the Karhunen Loève transform (KLT) to evaluate the
86 model performance (Wilks, 1995, Longuevergne et al., 2007). The KLT allows the
87 decomposition of piezometric time series into several independent temporal vectors,
88 corresponding to the main influential processes. This gives a different insight into the
89 performances of the simulations, focused on the representation of the main influential
90 processes over the piezometric head temporal evolution, and helps in discriminating and
91 quantifying the model ability to represent these processes. Using the results of this analysis,
92 the aquifer water budget and recharge components are assessed, as well as the uncertainty in
93 their computation.

94 This article is organised as follows: first, the characteristics of the basin are presented. Then,
95 the models used in this work are described (section 3), and the calibration of the surface water
96 budget is presented (section 4). In section 5, the different parameter sets are presented, and the
97 sensitivity of model results is analysed using the classic statistical criteria and the Karhunen
98 Loève transform. Eventually, section 6 presents the estimation of the aquifer water budget and
99 stock variations and the uncertainty on the recharge/discharge fluxes.

100

101 **2. Basin characteristics**

102

103

104 **Figure 1 : a) Situation of the basin and topography with main rivers and gauging stations used for model**
105 **calibration; b) Mean annual precipitation in mm/yr computed on the period 1983-2006 as estimated by**
106 **the SAFRAN analysis (cf section 2.2).**

107

108 **2.1 Geographical and geological setting**

109 The Upper Rhine graben hydrosystem, situated at the French-German border (Figure 1), is
110 composed of three units: i) a tectonic graben containing Secondary to Tertiary sediments,
111 several kilometres thick (Figure 2) and forming the substratum of the Rhine alluvial plain
112 (Illies, 1972), ii) the graben shoulders with much older materials such as crystalline,
113 metamorphic and ancient sedimentary rocks constituting the Black Forest mountains in the
114 East and the Vosges mountains in the West, and iii) the Sundgau hills in the South, formed by
115 marly Oligocene substratum of the Quaternary sediments which was relatively less subsided
116 than in the rest of the graben during the continental rifting.

117 The graben contains Tertiary marls and clays, covered by Quaternary alluvium deposited by
118 the Rhine River and forming the studied alluvial aquifer (cf. figure 2) (Duprat et al., 1979,
119 Bauer et al. 2005). The hydrographic network is very dense in the plain, due to the presence
120 of many groundwater-fed streams, allowing significant exchange of water between rivers and
121 the aquifer (Schmitt, 2001).

122 **Figure 2 : Geological cross section of the Rhine Graben, showing the sediment succession in the plain and**
123 **relative thickness (modified from Illies (1972))**

124

125 The main Rhine's tributary in this part of its basin, the Ill River (figure 1), has its source in the
126 Sundgau, while some small rivers having their sources in the same area fully infiltrate towards
127 the aquifer. The other most important rivers have their source in the Vosges and Black Forest
128 mountains West and East of the plain.

129 The material of the Rhine alluvial aquifer is mainly coarse quaternary gravels and sands with
130 good hydrogeological properties. The hydraulic conductivities are in the order of $10^{-4} - 10^{-3}$
131 m/s (Duprat et al., 1979; LfU, 1996; LUBW, 2006b). On the other hand the rocks forming
132 the mountains are rather impervious material containing only local aquifer formations of
133 small extension. The alluvial aquifer thickness reaches more than 200 m in its centre, East of
134 Colmar (LUBW, 2006). Mineral composition and thus structure and texture of the sediments
135 deposited by the Rhine River are marked by their Alpine origin, whereas towards the edges of
136 the graben, Vosgian and Black Forest rivers have deposited alluvial material having different
137 hydrogeological properties. These elements contribute to the heterogeneity of the aquifer
138 material, which in turn impacts the geometry of the river-aquifer interactions.

139 The aquifer substratum consists of the Oligocene marls considered impervious. The
140 groundwater flows from South to North and is closer to the surface in its northern part. Water
141 table depths range from 0 to 20 m (Hardt area), and groundwater fed wetlands are present in
142 the middle and the northern parts of the plain.

143 **2.2 Climatic conditions**

144 Precipitations show large contrasts over the basin, with annual values ranging from 550
145 mm/year in the plain, to more than 2000 mm/year in the Vosges and Black Forest mountains
146 (Figure 1). Snowfalls account for around 3 % of total precipitations in the plain, and up to
147 37 % at the mountains tops. Thus snow accumulation and melting are important processes for
148 the dynamics of river flows in these catchments. Moreover potential evapotranspiration,
149 computed with Penman formula on the 1983-2006 time period, is weaker on the mountains

150 tops (around 600 mm/year) than in the plain (around 800 mm/year). As a consequence the
151 rivers flowing from the mountainous catchments carry an important proportion of the water
152 involved in the hydrosystem budget and play a crucial role in the recharge of the Rhine
153 alluvial aquifer.

154 Daily precipitations and potential evapotranspiration (PET) for the modelling are provided by
155 the SAFRAN analysis performed by Météo France, based on the definition of climatologically
156 homogeneous zones and on altitudinal variation of the meteorological variables (Quintana-
157 Seguì et al 2008, Vidal et al., 2010). They are used here at a daily time step from 1985 to
158 2003, which is about the period of time used in the previous Monit project (LUBW 2006). For
159 the present study the SAFRAN analysis that is usually provided on a regular grid of 8x8 km²
160 was projected over a specific grid with cells varying from 1 to 8 km size in order to better
161 account for the topographic gradient of precipitation (Figure 1).

162 **2.3 Available data**

163 Daily river discharges are observed in about 20 to 30 river gauges within the studied area,
164 depending on the time period. These gauging stations are managed mainly by the DREAL-
165 Alsace (Direction Régionale de l'Environnement, de l'Aménagement et du Logement in
166 Alsace) on the French side, and by the LUBW-Baden-Württemberg (Landesanstalt für
167 Umwelt, Messungen und Naturschutz in Baden-Württemberg) on the German side.
168 Piezometric levels data consist in numerous observation wells (more than 200 points) sampled
169 weekly, only a few of them having daily data. These observation wells are managed mainly
170 by the APRONA (Association pour la PROtection de la Nappe d'Alsace) on the French side
171 and by the LUBW on the German side.

172 Seasonal piezometric levels fluctuations are generally limited to one meter or less, except in
173 the South and near the Eastern and Western borders where the water table is deeper. No long
174 term tendency appears for the piezometric levels over the simulated period, from 1985 to

175 2003. Nevertheless piezometric levels are rather low between 1990 and 1995, and rather high
176 between 2000 and 2003. The same holds true for the mountain rivers discharges (cf. figure 7).

177

178 Land use is characterized by large agricultural areas in the plain, with extensive cultures of
179 cereals (some of them being irrigated), vineyards in the piedmont, whereas the mountains are
180 largely covered by coniferous and deciduous forests (Figure 3). Agricultural lands cover
181 approximately 40 %, and forests 45 % of the basin. Artificial areas represent almost 10 % of
182 the area, which is densely populated with more than 200 inhabitants/km². The main cities are
183 Basel, Colmar, Mulhouse and Freiburg in the South, Strasbourg in the North. Important water
184 withdrawals, around $3.6 \cdot 10^8$ m³/yr, are located close to these cities for drinking and industrial
185 purpose. Irrigated agricultural areas represent around 6 % of the region, mostly in the
186 Southern central part of the plain.

187

188 **3. Models presentation**

189 Two different models were used: one classical hydrogeological model (MODCOU, Ledoux et
190 al., 1989), and one hydrogeological model including an inversion of the model parameters
191 capacity (HPP-INV, Majdalani and Ackerer, 2010).

192 In this section, the models are presented briefly, with a special focus on MODCOU which is
193 mainly used in this study.

194 **3.1. MODCOU**

195 MODCOU is a spatially distributed coupled hydrogeological modelling tool that was already
196 used in several large basins in France (Golaz-Cavazzi et al., 2001, Ledoux et al., 2007,
197 Korkmaz et al., 2009).

198 It is built up on several interconnected modules. The first one is dedicated to the estimation of
199 the surface water budget computed with a simple reservoir scheme, the production functions

200 (Figure 4). Inputs to this conceptual model are the daily precipitation and potential
201 evapotranspiration (PET) data on a spatial grid. Outputs are actual evapotranspiration (AET),
202 infiltration and runoff on each grid cell and at a daily time step. More details on the
203 production functions are given in the next section where the calibration of the surface water
204 budget is presented.

205 The direct infiltration of water from effective rainfall through the unsaturated zone is
206 simulated with a Nash cascade (Philippe et al., 2010).

207 The aquifer representation is based on a square grid discretization, with a finite differences
208 resolution of the diffusivity equation. The rivers flows are simulated according to a
209 Muskingum scheme (David et al., 2011).

210 There are two kinds of interactions between the surface and the groundwater: 1) the aquifer
211 can be drained where piezometric level reaches the soil level, and 2) drainage or infiltration of
212 water towards the aquifer can occur on river cells. Owing to the major role of the river-aquifer
213 interactions on the Upper Rhine Graben aquifer, the description of these processes in
214 MODCOU is further detailed here.

215 On river cells, the river-aquifer exchange flow is proportional to the head difference between
216 the river and the aquifer, to permeability and width of the riverbed, and inversely proportional
217 to its thickness. As these characteristics of the river bed are mostly unknown, they are
218 integrated in two parameters called the transfer coefficient (T_p), for the case when the aquifer
219 and the river are connected, and the maximum river infiltration flow (Q_{lim}) for the
220 disconnected case (Rushton 2007). The exchange flow is therefore computed as follows:

$$221 \quad Q_{exch} = \max(T_p(H_{gw} - H_{riv}); -Q_{dis}), \quad (1)$$

222 With Q_{exch} the exchange flow between the river and the aquifer which is negative when water
223 flows from the river to the aquifer, H_{gw} the piezometric head in the aquifer, H_{riv} the water head
224 in the river, and Q_{dis} the available flow in the river cell. Q_{lim} is a maximum infiltration flow

225 from the river to the aquifer, corresponding to the case when the aquifer head is below the
226 river bed (Rushton, 2007). A steady river head H_{riv} is usually considered in MODCOU,
227 however, due to the impact of river-aquifer interactions in the Rhine hydrosystem, time
228 varying heads based on observations were included for the Rhine River, as well as for the
229 Hardt Canal. This waterway is used for irrigation purposes, and preliminary results have
230 shown it was important to explicitly simulate its losses to the aquifer.

231 MODCOU uses a daily time step to allow a better estimation of the river-aquifer exchange
232 dynamics.

233 MODCOU can also be coupled with the soil-vegetation-atmosphere transfer scheme ISBA
234 (Noilhan and Planton 1989) to form the SIM model (Habets et al., 2008). SIM is used here to
235 have another estimation of the effective rainfall infiltration over the aquifer.

236 **3.2 HPP-INV**

237 The HPP-INV model (Chardigny 1999, Majdalani and Ackerer, 2010) is a modelling tool
238 using an inversion algorithm in order to assess aquifer hydrodynamic parameters. It is based
239 on a classical hydrogeological model, and includes an adaptative multiscale triangulation for
240 the inversion algorithm (Majdalani and Ackerer, 2010).

241 Piezometric levels simulation is based on a finite elements scheme and the river flows are
242 simulated with a Muskingum scheme (Majdalani and Ackerer, 2010). Rivers-aquifer
243 interactions are computed from the difference between river and aquifer heads, with transfer
244 coefficients values calibrated during the inversion process (LUBW, 2006). In the present
245 application, this model is run at a monthly time step.

246 There is no computation of a surface water budget, and groundwater recharge by rainfall is
247 provided by MODCOU. Moreover, the model is limited to the aquifer in the plain, and the
248 lateral inflows from bordering catchments are calibrated.

249 The inversion method used to derive the spatial distribution of permeability is based on
250 minimization of the error between simulated and observed piezometric heads and river flow
251 data. Several distributions can be obtained according to different inversion conditions.

252

253 **4. Model set up**

254 **4.1 Discretization**

255 The area modelled with MODCOU corresponds to the Rhine basin between Basel in the
256 South and Lauterbourg in the North. The basin total area is approximately 13900 km², and the
257 area of the alluvial aquifer is 4655 km² (Figure 3). Two different boundary conditions are
258 imposed: the Rhine river discharge at Basel, and fixed piezometric heads at the northern and
259 southern boundaries.

260 The basin was discretized with square cells ranging in size from 200 m for the river cells to
261 1600 m. There are 99715 surface cells, and 34180 aquifer cells (Figure 3). The alluvial
262 aquifer is represented with a single layer.

263 Subsurface flow at the western and eastern borders of the aquifer is known to be a significant
264 part of the recharge (LfU, 1996, LUBW, 2006). It is represented by specific boundary
265 conditions in the model: the surface runoff reaching the aquifer surface on the bordering cells
266 infiltrates to the water table and thus contributes to the aquifer recharge. This process is called
267 hereafter lateral subsurface flow.

268

269

270 **Figure 3 : a) Discretization of the surface (light red) and the aquifer parts (green) of the model. The**
271 **border cells where the surface runoff is infiltrating are plotted in dark brown ; b) Land use data**
272 **extracted from Corine Land Cover 2000 data set.**

273

274

275 Uncertainty on the components of the water budget is evaluated by exploring several equally
276 probable hydrodynamic parameter sets (permeability, transfer coefficient, maximum river
277 infiltration flows) as well as two estimations of the water budget from the SIM land surface
278 scheme, and the MODCOU production functions. Consequently, only the best parameter set
279 for the production functions, obtained through calibration, was considered to run MODCOU.
280 The calibration method is presented in the following section.

281 **4.2 Calibration of the surface water budget model**

282 The parameter calibration process can be problematic, because of non-uniqueness (see for
283 example Beven, 1993; Gupta et al, 1998; Refsgaard, 1997). To avoid this kind of issues
284 during the calibration process, the number of parameters to be adjusted is usually reduced. A
285 manner to do this is to assess values for parameters based on their physical significance in the
286 hydrosystem. However the spatial distribution of these physical parameters is often not well
287 known, and the point measurement values may not be representative of grid cell scale
288 parameters. Moreover, model parameters are not always easily linked to physical properties of
289 the system. This is the case for the production function parameters of the MODCOU surface
290 model (Figure 4), which allow partitioning the incoming rainfall between actual
291 evapotranspiration and effective rainfall, and of effective rainfall between infiltration and
292 surface run-off. The values of the parameters are not connected to observed variables and are
293 most often calibrated.

294

295 **Figure 4 : schematization of a production function in MODCOU for surface water budget computation**

296 **(Ledoux et al., 1989)**

297

298 Calibration of the water budget is based on the reduction of errors between observed and
299 simulated rivers flows. River gauges out of the Rhine alluvial plain were used to calibrate the
300 production functions, considering that the contribution of aquifer-rivers interactions at these
301 gauging stations is negligible.

302 This calibration was carried out in a two steps process, with first calibration of the water
303 volumes flowing at the gauging stations, and then of river flow dynamics.

304 Ten gauging stations in the Vosges Mountains and in the Sundgau area were selected, based
305 on the repartition of the main production functions on their drainage areas. Twenty one
306 production functions were defined on this area, crossing information on land use (Corine
307 Land Cover) and geological soil types (from geological maps). The parameters were
308 calibrated for the fourteen production functions representing more than 20 km² on the gauging
309 stations watersheds, as the influence of the other production functions was not sufficient over
310 these areas. The parameters from the ten remaining production functions were set according
311 to similar soil and vegetation types in previous applications (Golaz et al.2001, Gomez et al,
312 2003, Korkmaz et al., 2009).

313 Calibration was performed in a semi automatic way, by testing several predefined values for
314 each parameter. The calibration period was from August 2000 to December 2003, as this
315 period presents contrasted climatic conditions, and has almost complete daily river flow
316 measurements for the selected gauging stations.

317 For each of the fourteen production functions, five reservoir parameters were adjusted. First
318 the parameters determining the total volume of water flowing at a gauging station (CRT and
319 DCRT) were calibrated based on the relative difference between simulated and observed
320 water volumes over the calibration period following equation 2.

$$321 \sum_{i=1}^{n_{stations}} \left| \frac{V_{obs,i} - V_{sim,i}}{V_{obs,i}} \right| \times \frac{1}{n_{stations}}, \quad (2)$$

322 $n_{stations}$ being the number of gauging stations used for calibration, $V_{sim, i}$ the simulated, and
323 $V_{obs, i}$ the observed volumes of water flowing at a station i .

324 Then the parameters affecting only the dynamics of river flow (FN, CQI and CQR) were
325 calibrated by optimizing the Nash efficiency (see § 5.1).

326 Only the best set of parameters was kept, as the sensitivity of the model to groundwater
327 recharge is assessed by using the water budget from the SIM model, as explained below.

328 **5. Assessment of the hydrogeological simulations**

329 In order to address the uncertainty in the aquifer water budgets and specifically recharge
330 components, several simulations using different equiprobable parameter sets were performed.

331 Some well constrained parameters were fixed (e.g. calibrated production functions
332 parameters, specific storage), while unconstrained hydrodynamic parameters that influence 1/
333 the transfer dynamic, 2/ the computation of rivers-aquifer interactions, and 3/ the aquifer
334 recharge were investigated.

335

336 1/ Transmissivity: Several distributions were obtained with HPP INV, by using various initial
337 conditions or criteria. These distributions are in good agreement with experimental data from
338 pumping tests (Majdalani and Ackerer, 2010). Four contrasted distributions were tested in the
339 present study to assess the sensitivity of the model to this parameter (Figure 5 and Table 1).

340 Two sets have similar mean values, but contrasted standard deviations, and the two others
341 have higher and lower mean values.

342 2/ To investigate the uncertainty on the river aquifer interactions, the sensitivity of the model
343 to three parameters was tested:

344 a) The maximum river infiltration rate to the aquifer, for river cells of 200 m: $Q_{lim} = 0$ (i.e. no
345 infiltration), 25, 50 and 100 L/s (see Equation 1). This parameter influences surface /

346 groundwater interactions, exclusively in the direction of river infiltration towards the water
347 table.

348 b) The transfer coefficients between the surface and the groundwater parts of the model: the
349 initial value of 0.05 m²/s was multiplied by 2 and by 10. This parameter has an influence over
350 the surface-aquifer interactions in both directions.

351 c) The Rhine River and the Hardt canal heads: the impact of the variation with time of these
352 river heads on the aquifer and streamflows behaviours was assessed. Although the Hardt canal
353 has far smaller discharges than the Rhine River, preliminary results and KLT analysis have
354 shown that it was important to represent this irrigation waterway. For this canal it was also
355 tested not taking into account variable water flows.

356

357 3/ Uncertainty in the surface water balance is taken into account by using two estimates of the
358 infiltration rates: one from MODCOU (conceptual model), and the other from SIM (ISBA
359 land surface scheme). The latter leads to an infiltration reduced by about 23 %.

360

361 **Figure 5 : Two of the four transmissivity distributions tested, obtained through inversion with the HPP**
362 **INV model.**

363 The specific storage distribution used is the same as in Majdalani and Ackerer (2010), with
364 values of 0.05 everywhere, except along a band a few kilometres wide around the southern
365 part of the Rhine River, where the value is 0.12.

366 A reference simulation, based on mean range parameters (Table 1), is considered in order to
367 assess the impact of different parameters variation independently: eleven other simulations
368 were run by modifying a single parameter values at a time. Table 1 summarizes the different
369 parameters values tested.

370

Simulation	Transmissivities (mean value and standard deviation in m ² /s)	Maximum infiltration flow Q _{lim}	Transfer coefficient T _p	Infiltration rate Q _i
Reference	Hpp19 (0.415; 0.426)	50 L/s	0.05 m ² /s	MODCOU
T hpp18	Hpp18 (0.415; 0.457)	50 L/s	0.05 m ² /s	MODCOU
T hpp47	Hpp47 (0.438; 0.465)	50 L/s	0.05 m ² /s	MODCOU
T hpp103	Hpp103 (0.377; 0.412)	50 L/s	0.05 m ² /s	MODCOU
Q _{lim} = 0	Hpp19	0 L/s	0.05 m ² /s	MODCOU
Q _{lim} = 100 L/s	Hpp19	100 L/s	0.05 m ² /s	MODCOU
Q _{lim} = 25 L/s	Hpp19	25 L/s	0.05 m ² /s	MODCOU
T _p = 0.1 m ² /s	Hpp19	50 L/s	0.1 m ² /s	MODCOU
T _p = 0.5 m ² /s	Hpp19	50 L/s	0.5 m ² /s	MODCOU
SIM	Hpp19	50 L/s	0.05 m ² /s	SIM
Steady H Rhine	Hpp19	50 L/s	0.05 m ² /s	MODCOU
Steady H & Q Hardt	Hpp19	50 L/s	0.05 m ² /s	MODCOU

371

372 **Table 1 : Recapitulation of the different model parameterizations tested**

373

374 **5.1 Standard analysis**

375 Results obtained with the different parameterizations are quantitatively evaluated in terms of
376 river flows using Nash criterion, and of piezometric levels using biases.

377 *5.1.1. Comparison of the riverflows*

378 Nash criterion (Nash and Sutcliffe, 1970) is a measure of fit between observed and computed
379 values (Equation 3). Its optimal and maximal value is 1 and its minimum value is $-\infty$. A zero
380 value indicates that the model does not perform better than the mean value of the
381 observations.

$$Nash = 1 - \frac{\sum_{i=1}^n (Q_{obs,i} - Q_{sim,i})^2}{\sum_{i=1}^n (Q_{obs,i} - \bar{Q}_{obs})^2}, \quad (3)$$

with n the number of observed values, $Q_{obs,i}$ the observed river flows, $Q_{sim,i}$ the simulated river flows, with i denoting the time step, and \bar{Q}_{obs} the arithmetic mean of the observed river flows.

The distribution of Nash criteria for the twelve simulations is shown at each gauging station (Figure 6). General agreement between simulated and observed river flows is found on most mountainous catchments, although some stations have low criterion values. Main discrepancies are located in the plain, which can be explained by the difficulty to reproduce the complex interactions between rivers and the aquifer in this area, and also by an important regulation of discharges in some rivers. The poor results on the German side can be linked to the poorer reconstitution of the atmospheric forcing, as less local climatic data were used in the SAFRAN analysis in Germany. There is no variation between the different simulations in the mountainous catchments as the different parameters tested have no influence on these catchments. The criterion presents some variability in the plain, especially on the Ill River.

396

Figure 6 : Spatial distribution of the daily Nash criterion values. At each gauging station a pie is plotted representing the distribution of the Nash criterion values for the 12 simulations.

399

However, the parameters have a more pronounced impact on low flows. Indeed, the 5-year return period low flow (QMNA5) can almost double from a simulation to another, especially in the southern part of the plain (not shown).

5.1.2. Piezometric levels comparison

403

404 Piezometric levels biases are computed as the average difference between simulated and
405 observed piezometric heads.

$$406 \quad Bias = \frac{1}{n} \sum_{i=1}^n (H_{sim,i} - H_{obs,i}), \quad (4)$$

407 Here n is the number of observed values, $H_{sim,i}$ the simulated piezometric levels, $H_{obs,i}$ the
408 observed piezometric levels, i denoting the time step.

409

410 **Figure 7 : Comparison between observed and simulated piezometric levels at the 190 observation wells**
411 **over the 1986-2002 period. Left: bias [m] computed for the reference simulation, middle: mean bias [m]**
412 **for the 12 simulations, right: standard deviation of the bias over the 12 simulations [m].**

413

414 Figure 7 shows the spatial distribution of the piezometric levels biases for the reference
415 simulation as well as the mean bias and its standard deviation for the 12 simulations in meter.
416 The piezometric levels are underestimated in the South Western part and rather well estimated
417 in the centre of the plain. The sensitivity of the parameters tested is higher in the Southern
418 part of the aquifer, with large standard deviations. Indeed in this area piezometric levels are
419 strongly influenced by river infiltration. Some simulations better represent the observed
420 piezometric levels in the South, as illustrated with the reference simulation results.

421

422 **Figure 8 : Bias histogram computed on the 190 observation wells, for the twelve simulations**

423

424 Figure 8 shows the statistical distributions of the 190 biases computed for each of the twelve
425 simulations. The distributions show that most observations are well reproduced by the
426 models, the largest number of observation wells being associated with the lowest errors. Only
427 the simulation with no infiltration of rivers to the aquifer ($Q_{lim}=0$) shows a distinct pattern,
428 with more negative biases. This underlines the importance of the river to aquifer flux.

429 Compared to the reference simulation, a maximum infiltration rate set to $Q_{lim}=100$ L/s leads to
430 an overestimation of the piezometric levels, with more observation wells having bias values
431 ranging from 1 to 5 m, while the reduced value of this maximum infiltration rate ($Q_{lim}=
432 25$ L/s) leads to slightly less values close to 0. The modification of the transfer coefficient for
433 surface/groundwater exchange (T_p) leads to a greater number of negative biases values, but
434 mainly in the range between -2.5 and -1 m.

435 It is rather difficult to distinguish the results obtained by the four transmissivity distributions,
436 and by the two infiltration rates.

437 Not taking into account the variations of the Rhine water levels leads to more overestimated
438 piezometric levels, whereas steady values for the Hardt Canal water levels and flow lead to
439 more underestimated levels.

440 Similar results were obtained for the root mean square error (not shown) for which the
441 maximum dispersion is reached for the $Q_{lim} = 0$ simulation.

442 From these comparisons, it is clear that classic statistical results are not sufficient to
443 discriminate the various sets of parameter and that all the simulations except the one with no
444 river infiltration give rather comparable results. In order to go further in the analysis of the
445 simulations, we used a more advanced tool, the Karhunen Loève Transform.

446 **5.2 Analysis using the Karhunen-Loève transform method**

447

448 *5.2.1 Principle of the piezometric data analysis*

449 Longuevergne et al. (2007) have performed a statistical analysis (Karhunen-Loève transform,
450 hereafter KLT) to separate the main influent processes on piezometric variability from
451 observed time series. KLT, similar to Principal Components Analysis (Wilks, 1995), involves
452 summing up a set of possibly correlated time series into a few uncorrelated variables called
453 eigenvectors Y_k , related to the original variables by an orthogonal transformation maximizing

454 explained variance. Piezometric head time series, expressed as the variation around the mean
455 level, may then be simply reconstructed as a sum of the temporal behaviours (eigenvectors

456 $Y_k(t))$ with associated spatial weight a_k^{obs} :
$$\Delta H_{obs}(t) = \sum_k a_k^{obs} Y_k(t)$$

457

458 This method was already applied on the Upper Rhine aquifer piezometric observations
459 (Longuevergne et al., 2007). Several eigenvectors describing the global temporal behaviour of
460 the aquifer were extracted and interpreted as different contributions to groundwater variations,
461 namely mountain rivers interactions, effective rainfall and Rhine river interactions for the
462 three more important ones. The analysis of the weights a_k^{obs} at 190 observation wells allows
463 describing the spatial distribution of these different influences. The results showed that nearly
464 70% of the aquifer behaviour is determined by interactions with rivers, with a distinction
465 between the Rhine River and the other ones.

466 In the present study, the Karhunen Loève method is also applied to each simulation in order to
467 assess how well it reproduces the processes involved in the observed piezometric time series.

468 For this purpose, the model results were projected on the eigenvectors Y_k obtained from
469 observations to compute spatial information a_k^{sim} for each model and compare them with
470 spatial information from observations, for each separate process described by the
471 eigenvectors, allowing a multi-criterion analysis of the model quality.

472

473 ***5.2.2. Decomposition of the simulated piezometric time series***

474 The five main eigenvectors Y_k (interpreted as specific processes) determined by KLT of the
475 piezometric data were able to explain more than 80% of the aquifer behaviour (Longuevergne
476 et al., 2007).

477 The associated weights a_k^{obs} for each well are compared to those obtained by projection of the
478 model results on the eigenvectors Y_k , a_k^{sim} , for each simulation. As eigenvectors are
479 orthogonal, the root mean square error (RMSE) between model and observations may be
480 simply written as $\|H_{obs}(t) - H_{sim}(t)\|^2 = \sum_{k=1}^5 (a_k^{obs} - a_k^{sim})^2 \|Y_k(t)\|^2 + \varepsilon$, where ε is a truncation
481 error (only the five first eigenvectors are considered here). Minimization of the root mean
482 square error may be carried out on the spatial information, for each eigenvector associated
483 with highlighted processes.

484 The first eigenvector (Figure 9), related to the contribution of mountain rivers, explains nearly
485 50% of the aquifer behaviour. Specifically, variations in annual river discharge at the entrance
486 of the plain where the aquifer lies are closely related to long-term signature of the first
487 eigenvector (Figure 9 a). Spatially, the contribution of the first eigenvector to observations is
488 more important on the mountain piedmont than along the Rhine River (Figure 9 b). The same
489 map can be drawn for the reference simulation, considering the same eigenvector (Figure 9 c),
490 and the difference between the two maps highlights the model error (Figure 9 d). Globally, the
491 agreement is quite good and differences are localized in specific areas. For example, the
492 impact of mountainous rivers seems to be underestimated on the mountain piedmont north of
493 Colmar only. On contrary, the impact is overestimated along the Rhine River and in the areas
494 which are known to be highly anthropized (canals north of Basel, Mulhouse area affected by
495 potash mines and artificial lakes south of Strasbourg).

496 The method allows pointing out areas where the contribution of each process to the head
497 variability is not properly reproduced by the model. This spatial information may also be
498 interpreted globally as statistical information to assess the ability of the different simulations
499 to capture the global behaviour of the aquifer system (Figure 10). Ideally, the distribution of

500 $a_k^{sim} - a_k^{obs}$ for the 190 observation wells and for a given simulation should be centred on zero
501 with reduced dispersion, for each eigenvector.

502 For example, considering no river infiltration drives to significant underestimation of the
503 impact of the first eigenvector, corresponding to the contribution of mountainous rivers.
504 Figure 10 shows that the maximum river infiltration flow of 100 L/s, gives larger errors, and
505 thus, that this parameter set is less probable than others.

506 The largest contribution to RMSE for all simulations is concentrated on the second
507 eigenvector (Figure 10) previously interpreted as the contribution of effective rainfall. The
508 examination of spatial errors highlights strong differences in the Southern and Eastern parts of
509 the domain. However, the error is reduced when taking into account even partially impact of
510 irrigation in the Hardt area (Noyer and Elsass, 2006), since the error is larger for the
511 simulation without varying water flows and levels in the Hardt irrigation canal (Figure 10)

512 Analysis of the differences on the third eigenvector, related to the impact of the Rhine River,
513 clearly indicates that both infiltration and Rhine river level variations are important processes
514 to describe the interactions between the Rhine River and the aquifer. It also underlines that
515 varying the transfer coefficient values (T_P) can be interesting for a better representation of the
516 Rhine river influence over the aquifer.

517

518

519 **Figure 9: Spatial and temporal analysis with the KLT method for the first eigenvector, associated with the**
520 **influence of mountain rivers over groundwater variability. a) First eigenvector time series; b) and c) Maps**
521 **of the weights associated with the first eigenvector for the observations and reference simulation**
522 **respectively; d) Map of the differences between simulated and observed weights.**

523

524 **Figure 10: Errors in the reproduction of the eigenvectors weights at the 190 observation wells for the**
525 **twelve simulations a) Mean total errors for the five eigenvectors and standard deviations ; b) Contribution**
526 **to the total error of eigenvectors 1, 2 and 3.**

527

528 The four transmissivity distributions tested perform equally well to describe the impact of the
529 mountainous rivers. Hpp103 shows slightly larger underestimation North of Colmar, but also
530 smaller overestimation in Strasbourg area. In the meantime this distribution, which has the
531 lowest mean value, generates less error on the third eigenvector, which has more energy along
532 the Rhine River, and more error on the second eigenvector, representing the rainfall influence.
533 The statistical analysis of the simulations has shown that one of the twelve simulations,
534 “ $Q_{lim}=0$ ”, failed to reproduce the behaviour of the basin, and that two others, “Steady H&Q
535 Hardt” and “ $Q_{lim} = 100 \text{ L/s}$ ”, are less probable although they cannot be completely rejected.
536 The nine remaining simulations cannot really be distinguished statistically, and, although
537 there are still some errors in the representation of the hydrosystem, these nine simulations
538 represent equally well the basin functioning.

539

540 **6. Analysis of the water fluxes**

541 **6.1 Surface-aquifer exchange flux**

542 Surface-aquifer exchange is known to be one key component of the aquifer water budget,
543 although only few quantified estimations based on observations are available.

544 Figure 11 shows the average and the standard deviation of the simulated mean exchange
545 flows on the period for the nine acceptable simulations. This map highlights the main
546 drainage and river infiltration zones of the hydrosystem. Simulated river infiltration flows
547 (blue cells on Figure 11) are high in the Southern part of the aquifer, mainly along the Rhine
548 River. They are also important in Strasbourg area. These correspond to zones where the
549 infiltration of rivers is known to be important, especially in the Southern part of the plain
550 where several studies have given quantitative estimations of the recharge from the rivers to
551 the aquifer (Esteves, 1989, George et al., 1995).

552 In most part of the domain, the surface-aquifer exchange present low variability (standard
553 deviation values close to 0) except in some parts near the centre of the aquifer (Figure 11).

554

555

556 **Figure 11: Mean exchange fluxes between the surface and the aquifer for the reference simulation (left)**
557 **and on average on the 9 simulations (centre), and its relative standard deviation (right)**

558

559 Water table drainage zones (green to red cells on the left and middle maps of Figure 11) occur
560 in the Southern part of the aquifer, near its borders and along the Rhine River. Drainage zones
561 extensions are quite important in the centre of the plain between Colmar in the South and
562 Strasbourg in the North. This corresponds to the Ried area, an important wetland with shallow
563 water table, leading to an important aquifer overflow and the presence of groundwater-fed
564 streams.

565 In the Northern part of the aquifer, important drainage flows from the aquifer occur, due to
566 the proximity of the water table to the topographic surface.

567 Comparisons were made with values found in the literature (PIREN Eau Alsace, 1984;
568 Esteves, 1989; George et al., 1995; Schmitt, 2001). These values often correspond to limited
569 space and time extents, but it is interesting to know whether the orders of magnitude of the
570 simulated exchange are plausible.

571 Esteves (1989) indicates an average infiltration flow of 200 to 300 L/s on the Fecht River
572 between its entry into the plain and its confluence with the Weiss River. On this reach an
573 infiltration flow of 1000 L/s is simulated on average. Downstream the drainage flow varied
574 from 100 to 300 L/s on average (Esteves, 1989), whereas MODCOU results give an average
575 of 1635 L/s. Thus for this river the intensity of the exchange is certainly overestimated by the
576 model.

577 Along the Ill River, Esteves (1989) gives a value of 200 L/s/km infiltrating from the river to
578 the aquifer between Ensisheim and Colmar. The PIREN Eau Alsace (1984) gives a value of
579 130 L/s/km infiltrating between Ensisheim and Oberhergheim, and 37 L/s/km between
580 Oberhergheim and Horbourg-Wihr, for the period 1971-1974. George et al. (1995) estimate
581 the infiltration between Mulhouse and Ensisheim amounts to 65 L/s/km in average.
582 The simulated exchange flows give an average infiltration of 124 L/s/km between Mulhouse
583 and Ensisheim, 93 L/s/km between Ensisheim and Oberhergheim, 87 L/s/km between
584 Oberhergheim and Horbourg-Wihr and 79 L/s/km between Ensisheim and Colmar.
585 Thus simulated infiltration of the Ill River between Mulhouse and Colmar is less variable than
586 estimated by observations, and it is overestimated upstream of Ensisheim and downstream of
587 Oberhergheim, while it seems underestimated between Ensisheim and Oberhergheim, i.e. on
588 the section just downstream of the confluence with the Thur River. However most of these
589 estimates are made by punctual gauging, a method lacking precision, and giving results with
590 time limited representativeness. Similar results are found on the Thur and Doller rivers, at the
591 South western end of the basin (George et al. 1995, Schmitt 2001).
592 It comes to the conclusion that the MODCOU model with the selected parameter sets is able
593 to represent the main spatial pattern of the aquifer-surface interactions. However, it has some
594 difficulties capturing the spatial variability of the exchange even by using several parameter
595 sets. On average this flux is overestimated in some sections of the hydrographic network. This
596 is probably due to the use of spatially homogeneous distribution for the parameters T_p and
597 Q_{lim} directly influencing the river-aquifer exchange.

598 **6.2 Water Budgets**

599 In order to address uncertainties in the functioning of the Upper Rhine aquifer, an analysis of
600 the water budgets computed by the different simulations is performed.

601

602 **Figure 12: Mean annual partition of the aquifer inflows between lateral subsurface flow, effective rainfall**
603 **and river infiltration (right bars), and of the aquifer outflows between pumping and rivers drainage (left**
604 **bars).**

605

606 Figure 12 shows the water budget for the aquifer recharge and discharge for the nine plausible
607 simulations. All the simulations, except the one with the effective rainfall infiltration scaled
608 according to SIM have the same effective rainfall and lateral subsurface flow recharge rates.

609 The river infiltration rate is the most important and most variable flux between the different
610 simulations. This flux almost doubles from the simulation $Q_{lim} = 25$ L/s to the simulation $T_p =$
611 0.5 m²/s. For the nine simulations, it represents about 80 % of the aquifer recharge, with a
612 maximum of 82 % for the simulations $T_p = 0.5$ m²/s and $Q_i * 0.77036$ and a minimum of 74 %
613 for the simulation $Q_{lim} = 25$ L/s. The nine simulations allow estimating the river infiltration
614 flux to 115 ± 16.5 m³/s, thus with an uncertainty of 14 %. Considering the twelve
615 simulations, the flux estimates would have been more scattered, with an average value of 105
616 ± 37 m³/s. Therefore, the KLT analysis allowed reducing the uncertainty from 35% to 14%.

617 These results are consistent with those obtained in the framework of the LIFE project. A value
618 of 85 % was found with a steady state simulation for a low water situation (LfU, 1996). On
619 the contrary in the Monit project the main contribution to the recharge of the aquifer was
620 found to be the effective rainfall infiltration over its surface (LUBW, 2006).

621 The lateral subsurface flow is about two times lower than the effective rainfall infiltration, and
622 represents approximately 7% of the water budget.

623 The main output process in all cases is the drainage of the aquifer by the rivers and by
624 overflow.

625 Another interesting variable is the aquifer storage variations. To analyse the uncertainty on
626 this variable according to the parameter sets, the evolution of the aquifer storage were
627 converted into water height variations over the entire surface of the aquifer. Thus, a variation

628 of 1 mm over the whole aquifer represents an evolution of the aquifer volume of 4,655,000
629 m³.

630 These variations are computed by reference to the aquifer stock at a given date, namely the
631 30th July 1986. This date was chosen in order not to take into account variations during the
632 first year of simulations, which could be due to inaccurate estimates of initial piezometric
633 levels.

634

635

636 **Figure 13: Monthly aquifer stock evolution from 1986 to 2003 as estimated by the 9 acceptable simulations**

637

638 The nine simulations present similar evolution of the aquifer stock. The largest discrepancy is
639 associated to the simulation with reduced maximum river infiltration rate. For the outlier
640 simulation without river infiltration ($Q_{lim} = 0$ L/s) the stock decreases by about 500 mm (not
641 shown). When the maximum river infiltration rate is set to $Q_{lim} = 25$ L/s, the aquifer stock is
642 reduced by approximately 25 mm at most.

643 The stock variations are not very sensitive to the transmissivity distributions although the
644 differences between the four distributions can reach 10 mm at some periods (Figure 13).

645 The different transfer coefficient values (T_p) lead to quite similar stock variations, even for the
646 higher value of $T_p = 0.5$ m²/s, which was shown to be the case having the highest river
647 infiltration flow. In this case, the stock is increased by only a few millimetres. Indeed there is
648 compensation between the river infiltration and drainage rates leading to a similar evolution
649 of the aquifer storage.

650 The reduced rainfall infiltration computed by SIM also has a limited impact on the aquifer
651 storage variations. Indeed the maximum impact reaches 10 mm at most between the reference
652 simulation and the reduced infiltration simulation.

653 To summarize, the aquifer stock variations are rather sensitive to the maximum river
654 infiltration rate, whereas they are less sensitive to rainfall infiltration and to transfer
655 coefficients. This can be explained by the fact that a modification of the infiltration over the
656 aquifer surface is balanced by a modification of the exchange with rivers flowing from the
657 mountains. Increasing the transfer coefficients leads to larger exchange between the aquifer
658 and the surface in both directions. This phenomenon can explain why the aquifer stock is not
659 really affected. On the other hand when the maximum river infiltration rate is modified, it
660 only affects the exchange from the rivers to the aquifer, leading to increased aquifer stock
661 variations.

662 **Conclusion**

663 The objective of the study was to increase understanding of the Upper Rhine Graben aquifer
664 functioning. Modelling of the Upper Rhine graben aquifer was performed with the MODCOU
665 coupled hydrogeological model computing the surface water budget, groundwater flow and
666 interactions between the aquifer and the surface. Distributions of aquifer transmissivities were
667 provided by the inversion algorithm of the HPP-INV hydrogeological model. In the present
668 study, the sensitivity of the MODCOU model results to some model parameters governing the
669 surface / groundwater interactions was assessed, specifically in terms of aquifer behaviour.
670 Although the developed model is not suited to represent fine scale processes and fully coupled
671 interactions between surface water and groundwater, it represents the main processes that
672 impact the aquifer water budget, and thus, allows an analysis of their spatial distribution.

673

674 Twelve model runs were performed using several parameter sets for the transmissivity, the
675 surface/groundwater transfer coefficient, the maximum river infiltration flow, Hardt Canal

676 water levels and flows and Rhine river water levels, and two estimations of the surface water
677 budget in the plain.

678 These twelve runs were assessed using classic statistical criteria: the daily Nash efficiency for
679 the river flows, and biases for the piezometric levels. These analyses show that for each run,
680 the model had some problems reproducing the daily discharges on the German side, probably
681 due to some weaknesses of the atmospheric analysis using fewer meteorological stations in
682 Germany than in France. The biases on the simulated piezometric heads are rather weak
683 except on the southern part of the basin, where the discrepancy between the twelve
684 simulations is large. These classical analyses were able to highlight that one set of parameters
685 was not coherent with the functioning of the aquifer system, and thus that it is required to take
686 into account the possibility for rivers to infiltrate.

687 In order to go further in the assessment of the simulations results, a mathematical analysis
688 method based on a Karhunen Loève transform (KLT) of piezometric time series was used to
689 analyse the simulated piezometric levels temporal variations, as well as the observed ones.
690 This method allows separating the different processes influencing the water table evolution, in
691 the form of eigenvectors representing river infiltration and the effective rainfall influences.
692 The spatial distribution of the projections stemming from the decomposition of piezometric
693 chronicles as linear combinations of the eigenvectors allows having an idea of the area of
694 dominating influence of the three main processes. Comparison of the projections obtained
695 with simulated and observed piezometric time series also highlights the main areas where
696 difficulties arise to reproduce the water table behaviour. KLT showed that all model runs
697 failed to represent the annual component of the second eigenvector, related to the impact of
698 effective rainfall, mainly due to large errors in the Hardt area in the south of the domain, and
699 in the Eastern part of the plain. This error is however reduced when taking into account the
700 irrigation practises in the Hardt area, with irrigation water flowing through channels with high

701 bed permeability and thus large river-aquifer exchanges. The KLT method underlines the
702 need to better take into account these processes in the model. This method also proved an
703 interesting and original technique to assess the model performances in reproducing the spatial
704 and temporal variability of water table levels, as well as the recharge/discharge processes
705 inducing these variations. It might be considered using this method in order to test spatial
706 patterns of some parameters taken homogeneous for now, such as the transfer coefficient T_p
707 and the maximum river infiltration rate Q_{lim} .

708 Comparison of the aquifer water budgets obtained with different parameter sets shows that
709 there is a balance between the magnitude of river to aquifer infiltration and drainage of the
710 aquifer so that the aquifer stock is rather stable for each run.

711 The intensity of the simulated river-aquifer exchange was compared to available observations,
712 mainly on the Ill River. Estimated values are in the same order as the observed ones. However
713 the model presents smoother spatial variations of the simulated fluxes than observed, which
714 could be due to the use of spatially homogeneous river bed parameters.

715 The KLT method has shown that nine over twelve simulations were able to have a similarly
716 satisfying representation of the aquifer behaviour. This allows reducing the uncertainty of
717 aquifer recharge by river infiltration from 35 % to 14 % with an average flux of $115 \text{ m}^3/\text{s}$.
718 This represents about 80% of the aquifer inflow.

719 This proportion compares well to the LIFE project estimate for a low water situation, but is
720 more important than the one from the Monit project for a transient state simulation.

721 This study brings increased comprehension of the functioning of an alluvial aquifer, which is
722 strongly influenced by groundwater /surface water interactions. It shows that modelling,
723 together with inversion and advanced statistical methods are powerful tools to quantify the
724 water budget of an alluvial aquifer.

725 Future work will include a climate change impact study on this vulnerable aquifer, by taking
726 into account the uncertainty associated with emission scenarios, climate models, and
727 hydrogeological model parameters.

728

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738

739

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Figure 1

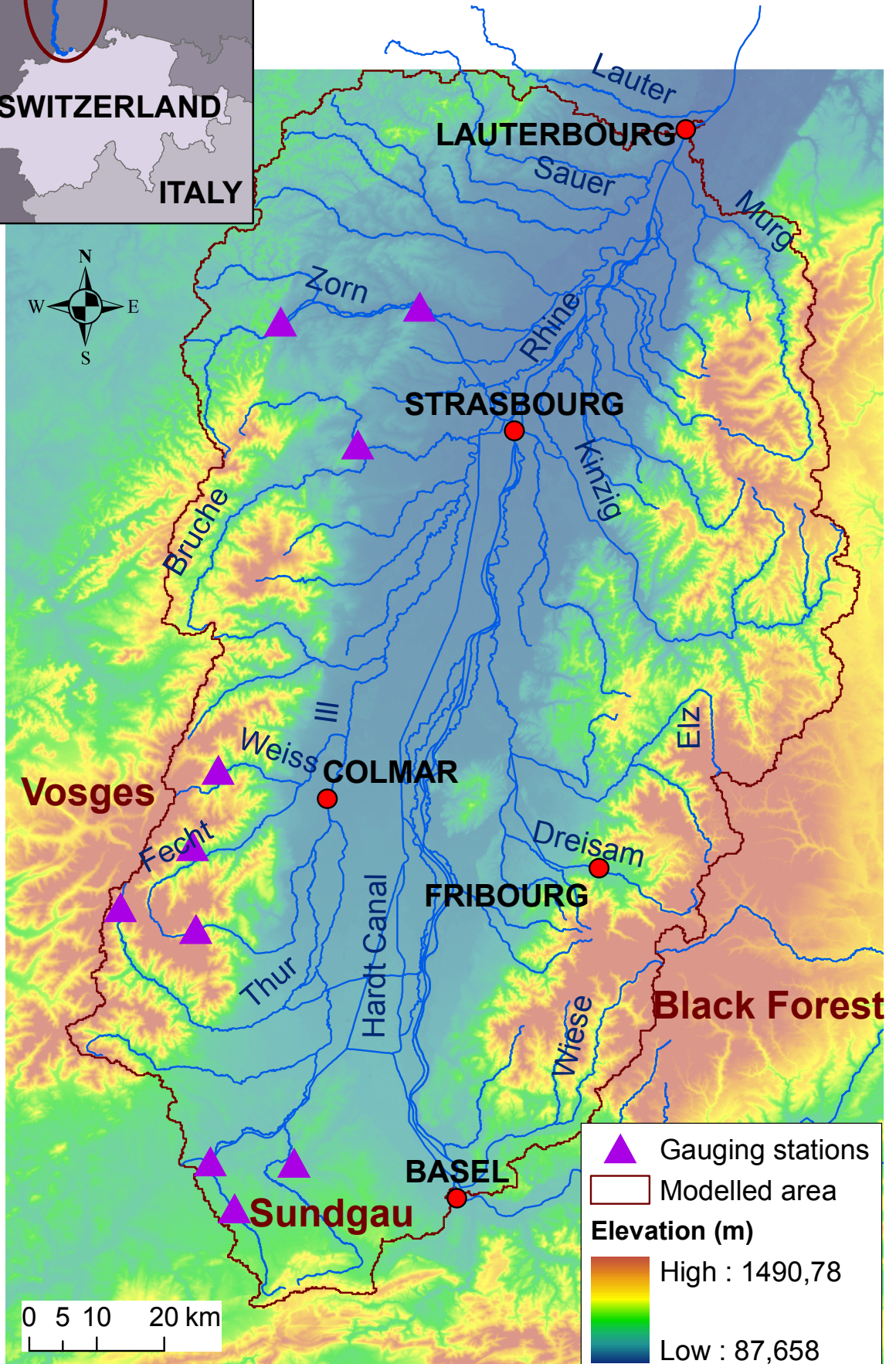
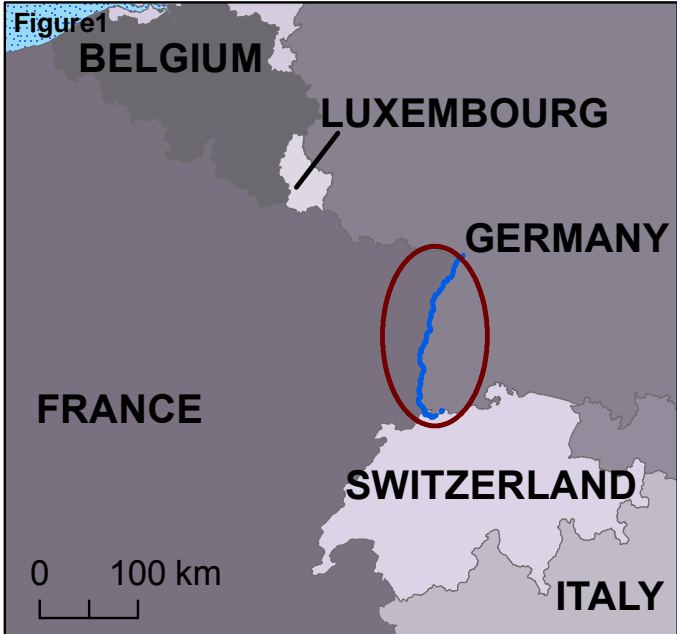


Figure1
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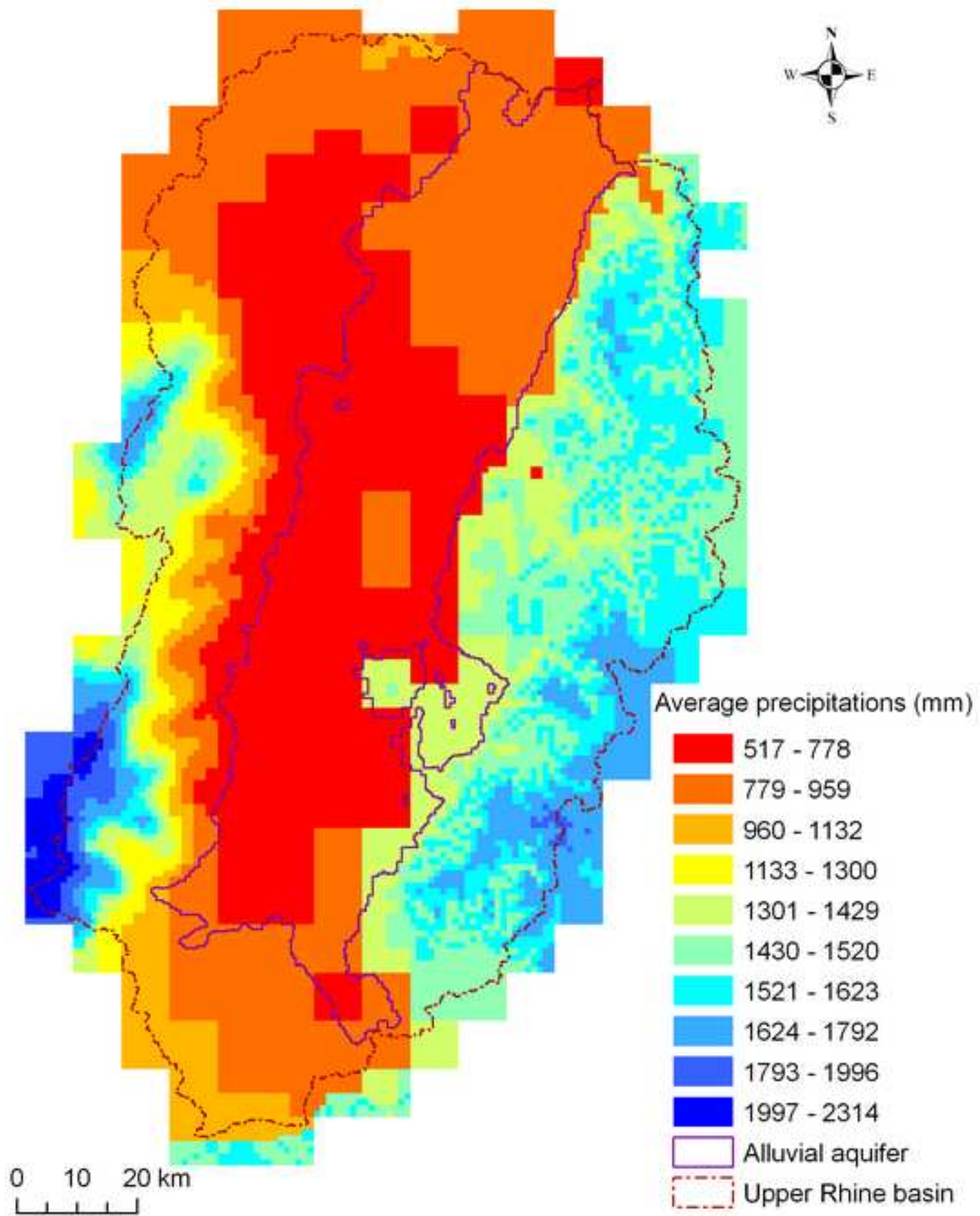


Figure2

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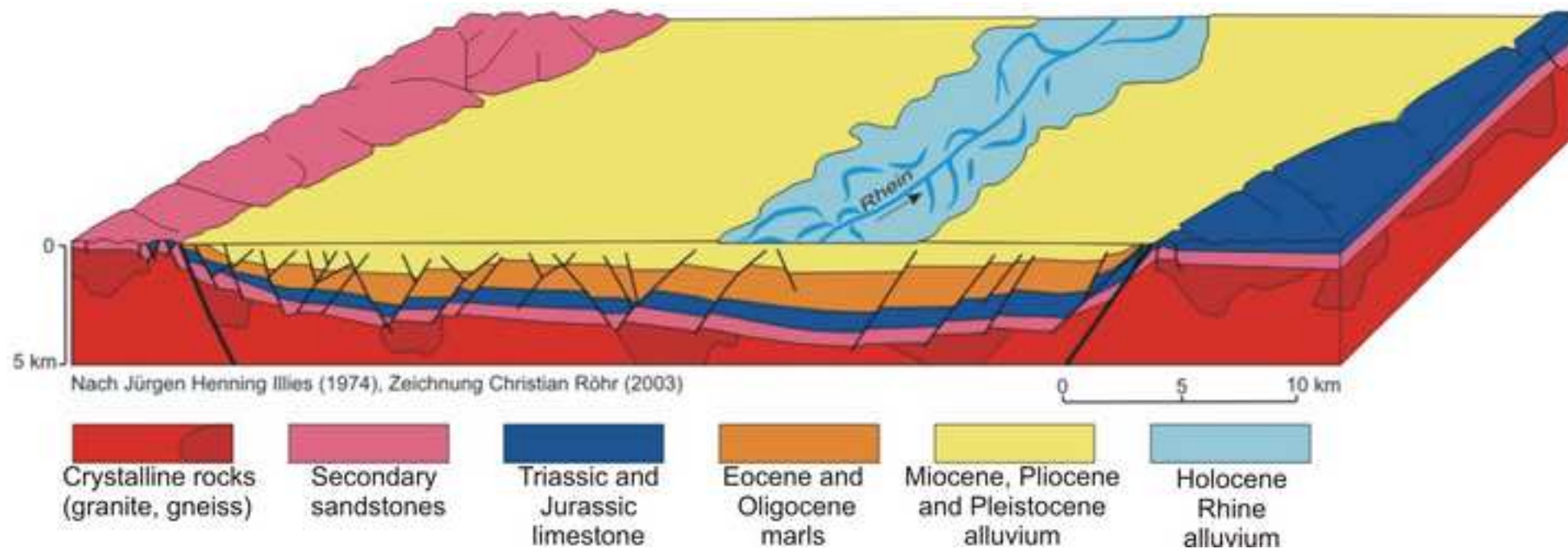


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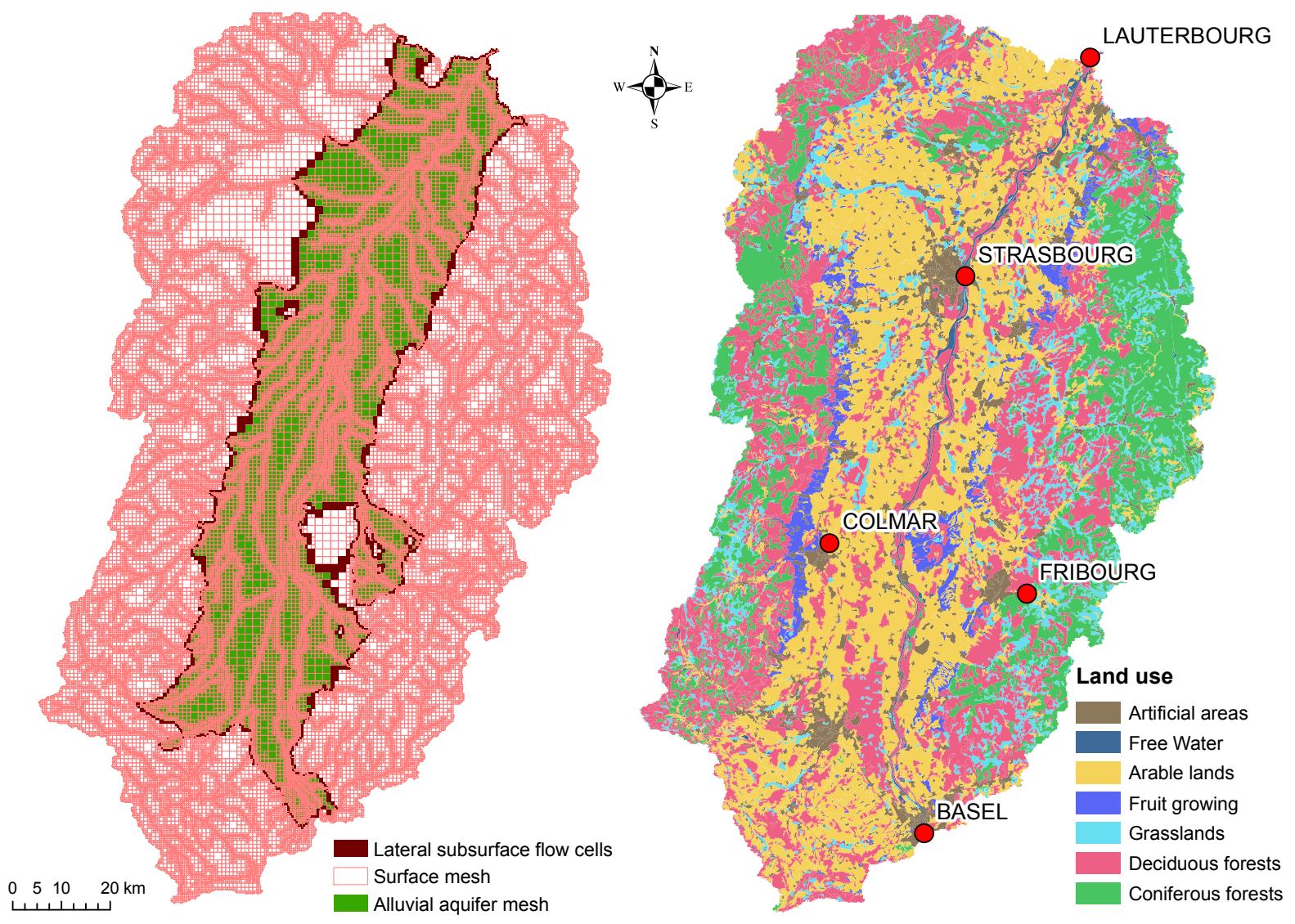
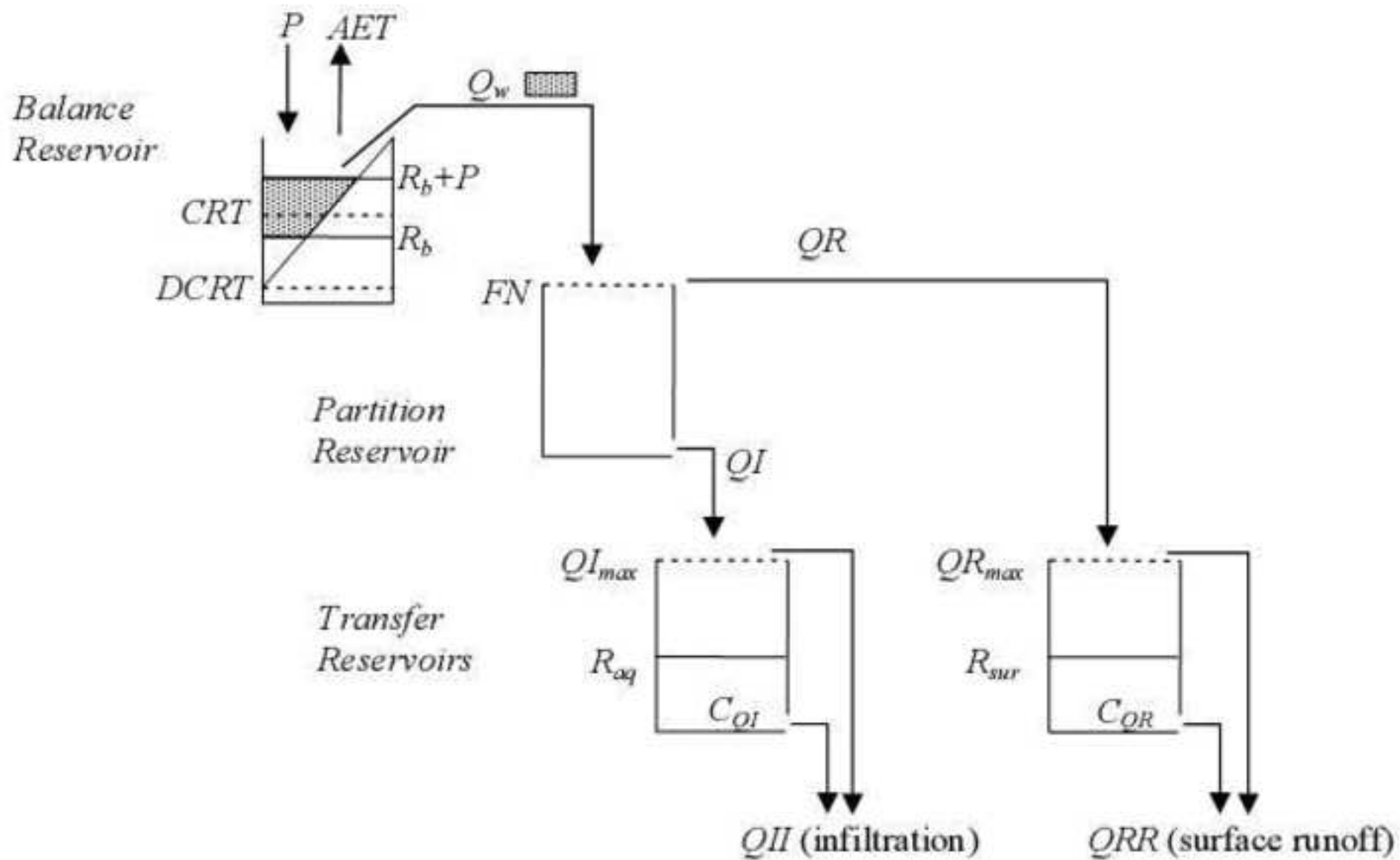


Figure 4

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T hpp103

Mean : $3.77e-1 \text{ m}^2/\text{s}$

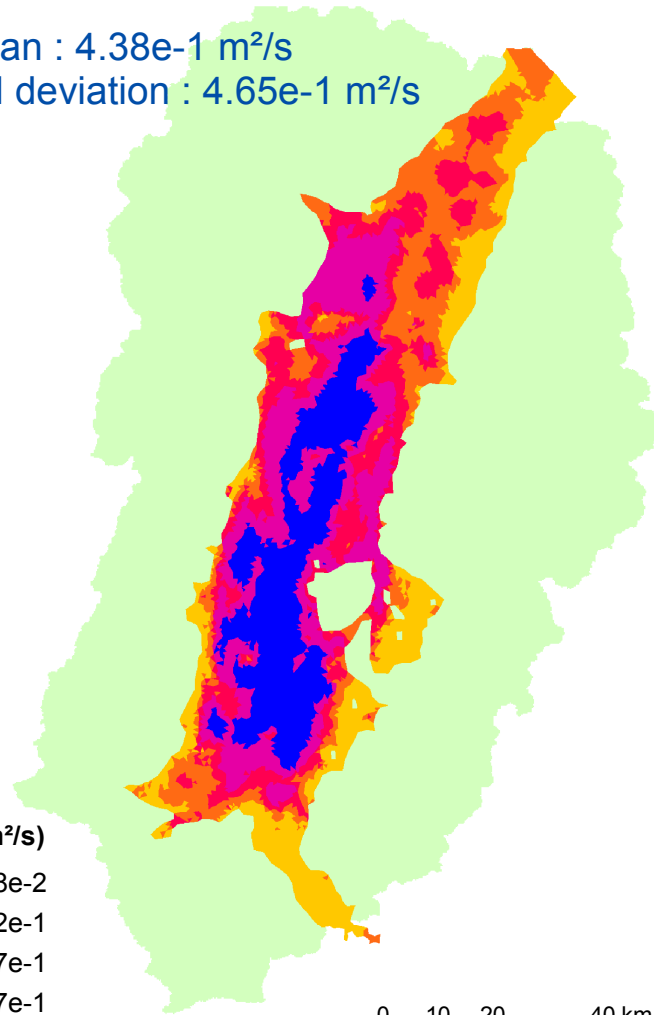
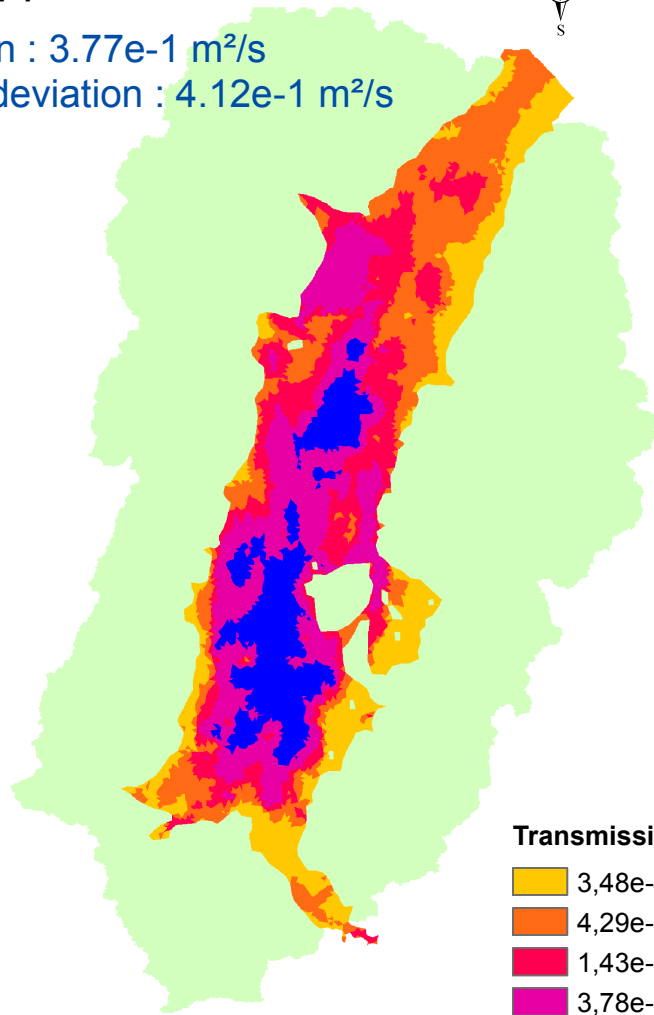
Std deviation : $4.12e-1 \text{ m}^2/\text{s}$



T hpp47

Mean : $4.38e-1 \text{ m}^2/\text{s}$

Std deviation : $4.65e-1 \text{ m}^2/\text{s}$



Transmissivity (m^2/s)

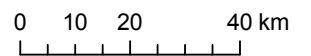
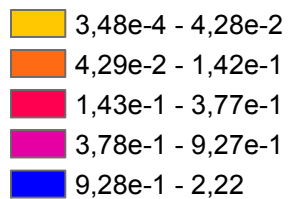


Figure6

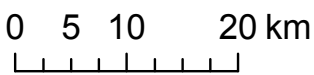
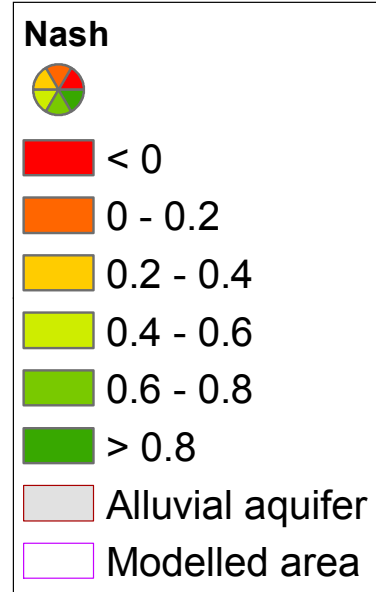
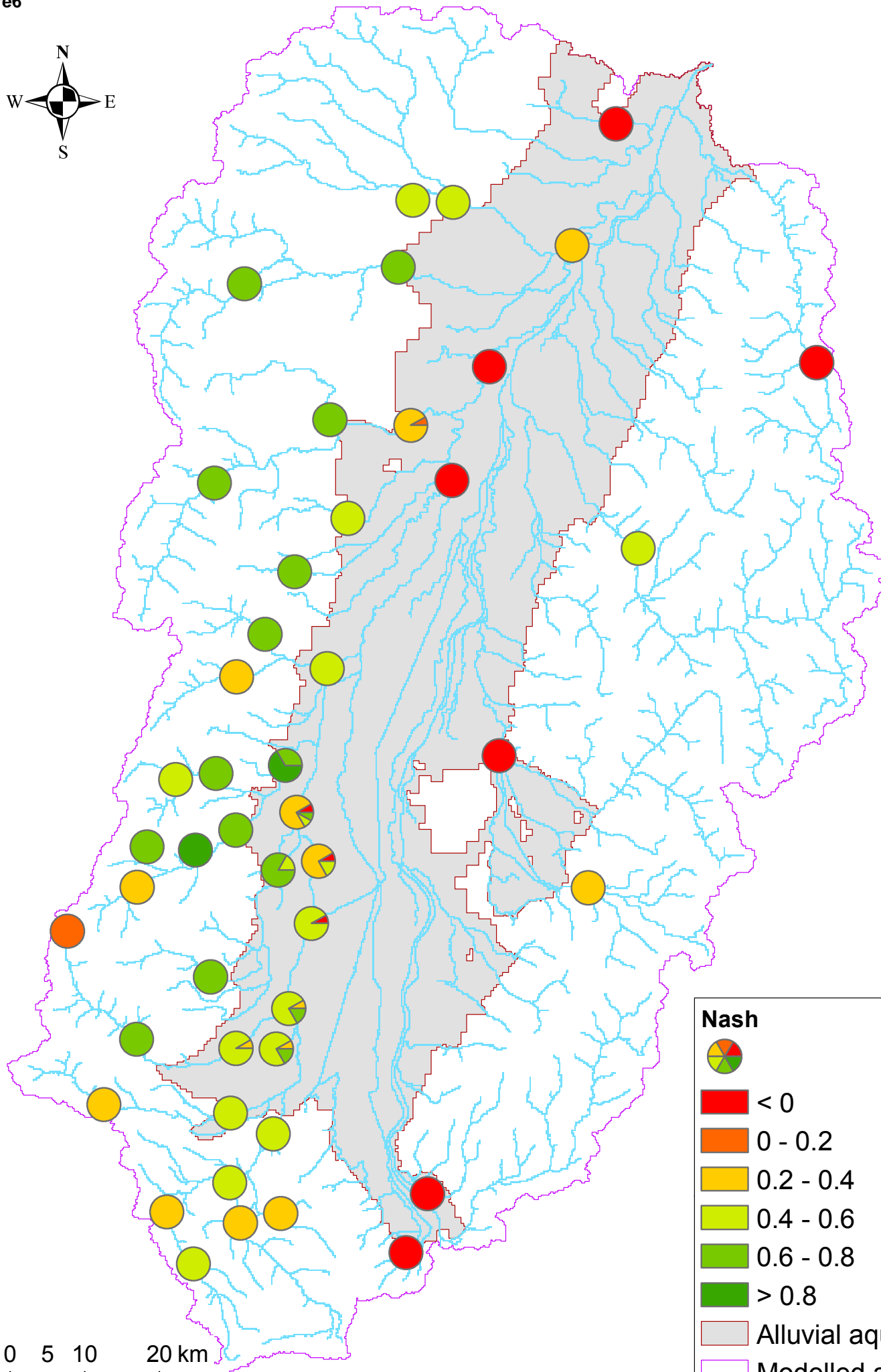
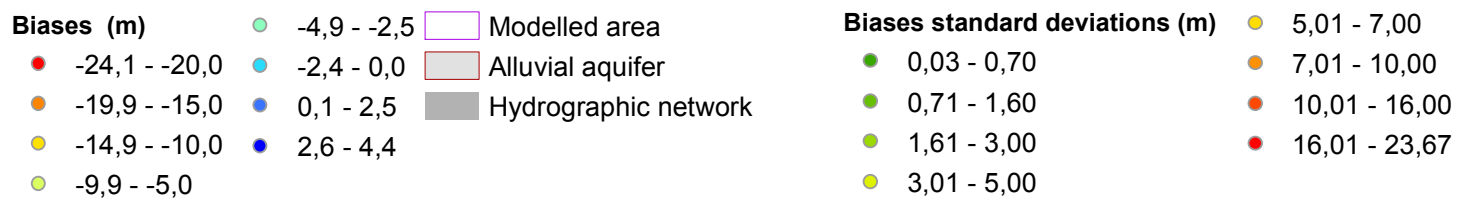


Figure7



Values for the reference simulation

Mean values for the 12 simulations

Standard deviation over the 12 simulations

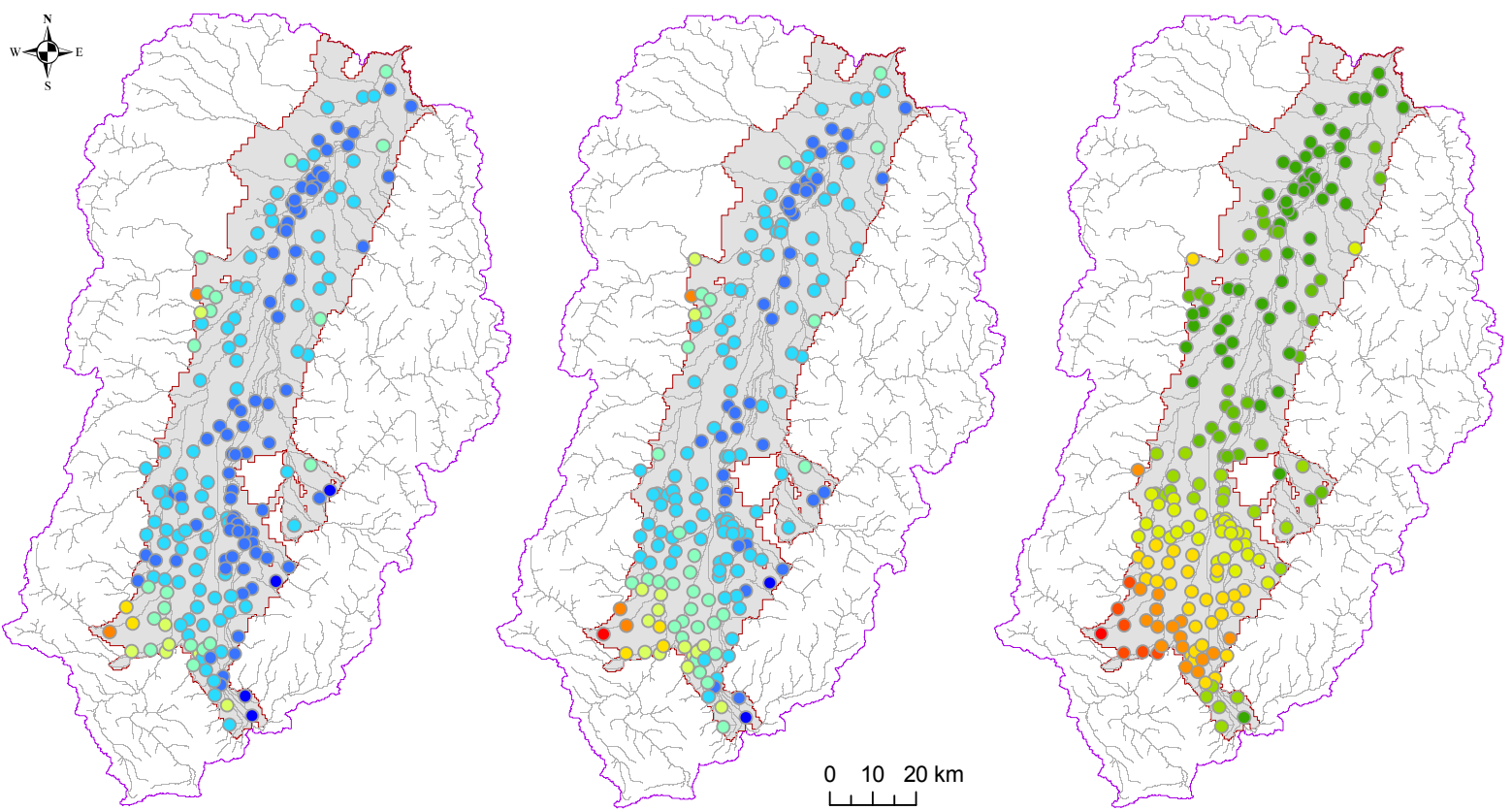


Figure8

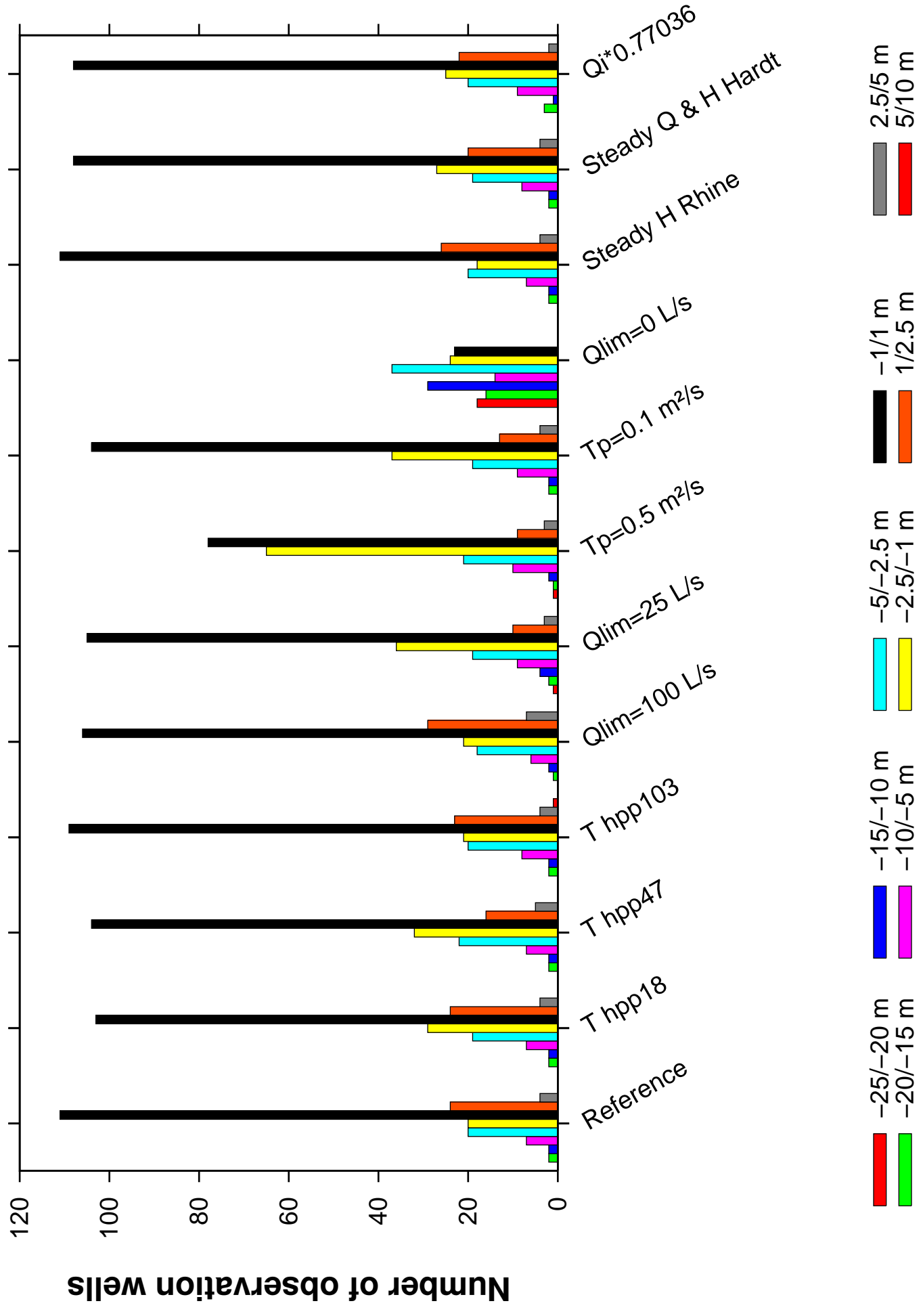


Figure9

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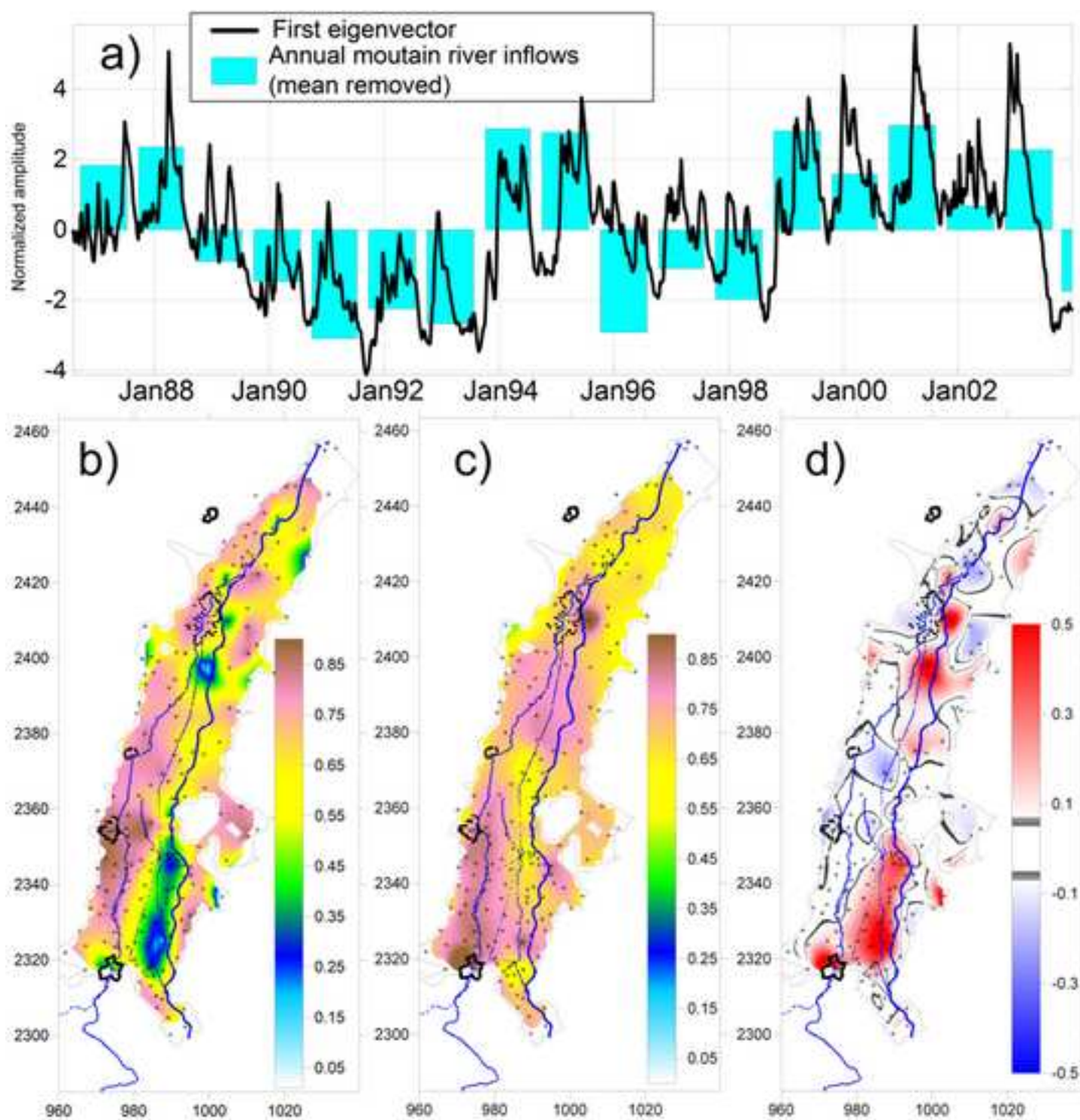


Figure10
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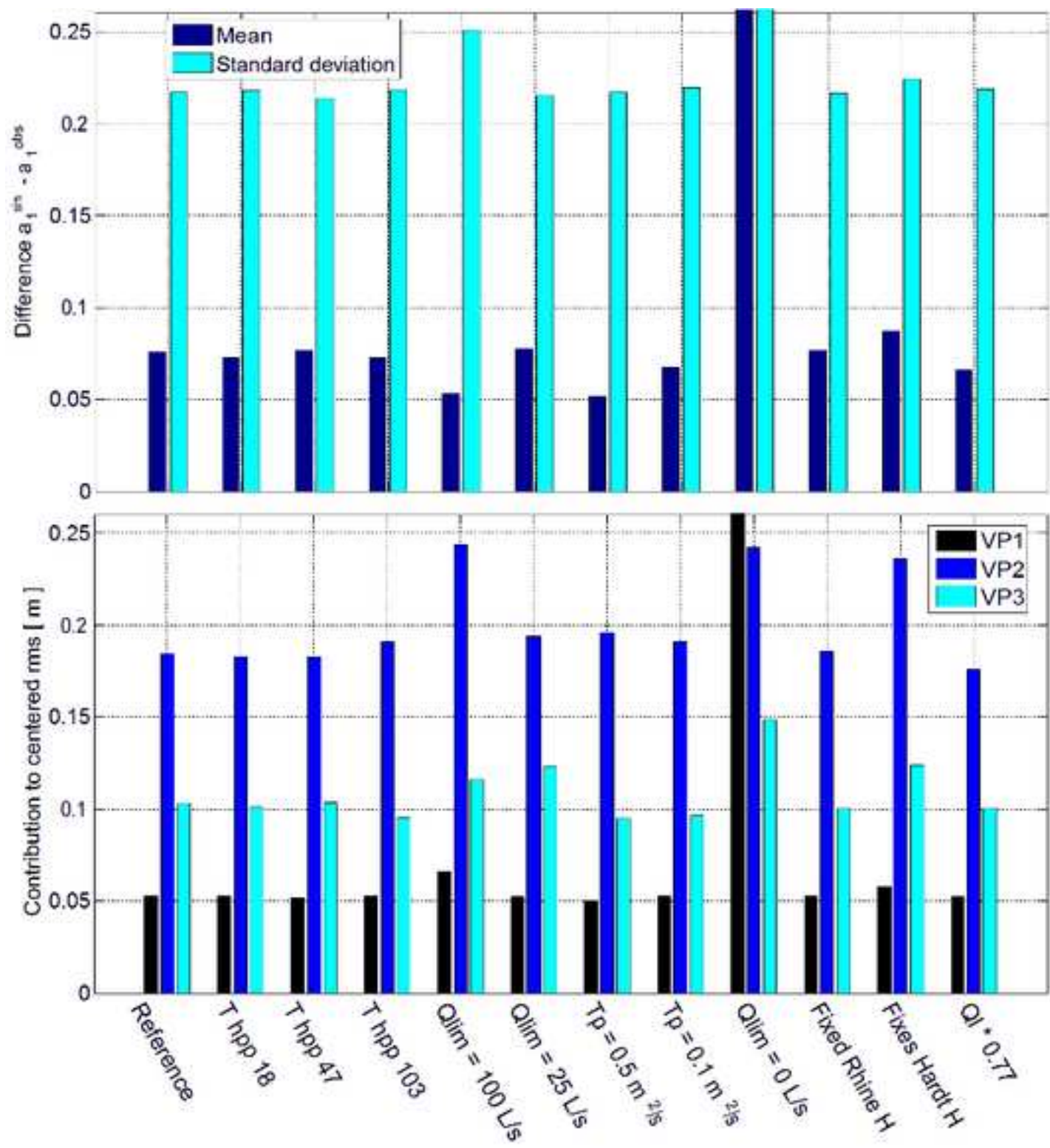
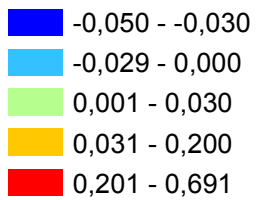
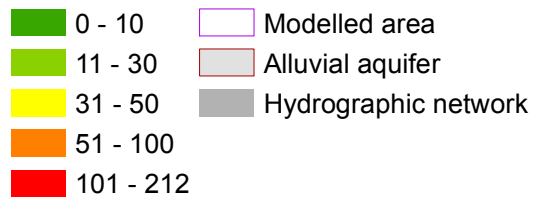


Figure11

Average exchange flow values (m3/s)



Average exchange flow standard deviation (m3/s)



Values for the reference simulation

Mean values for the 9 simulations

Standard deviation over the 9 simulations

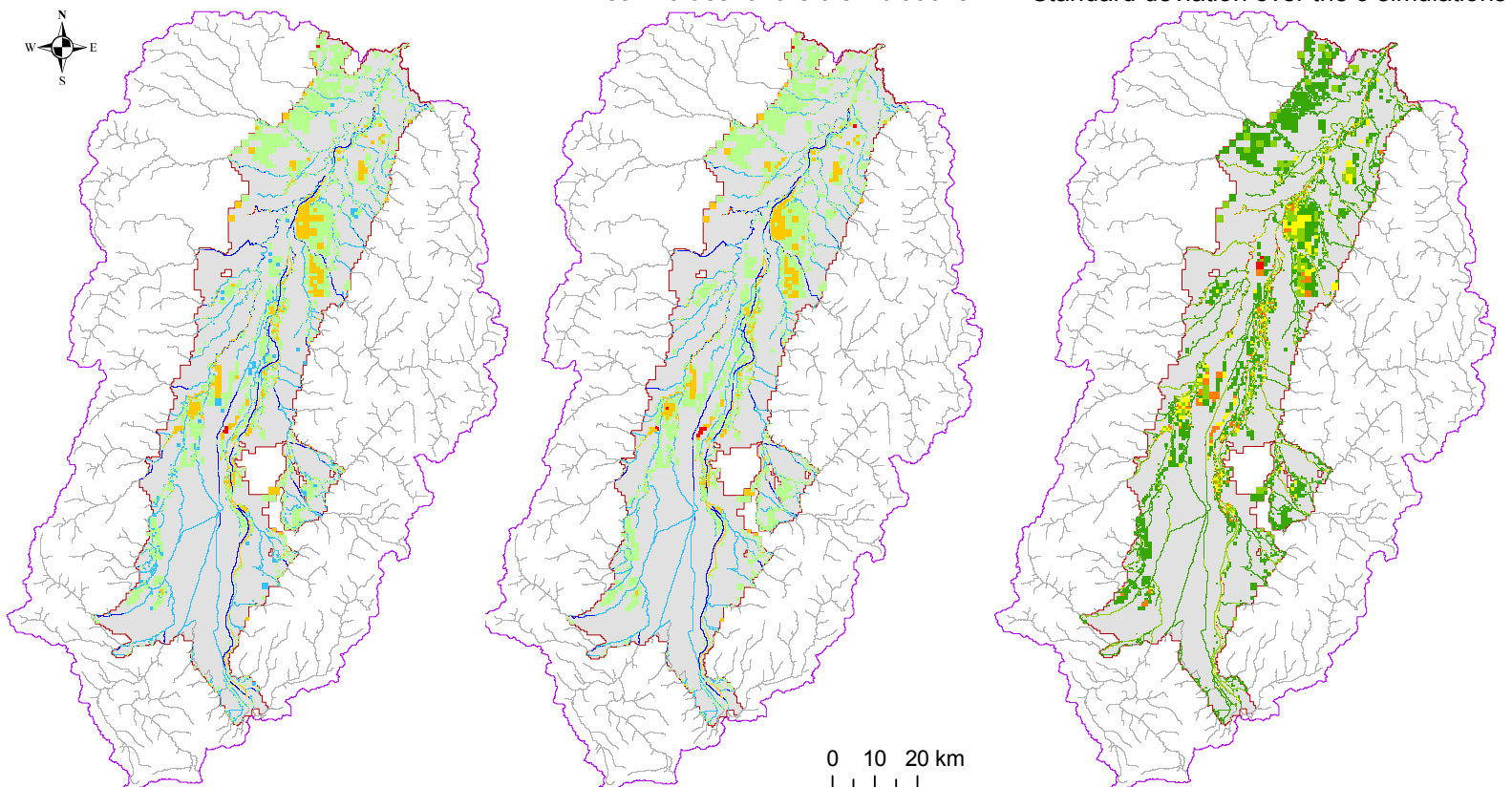


Figure12

