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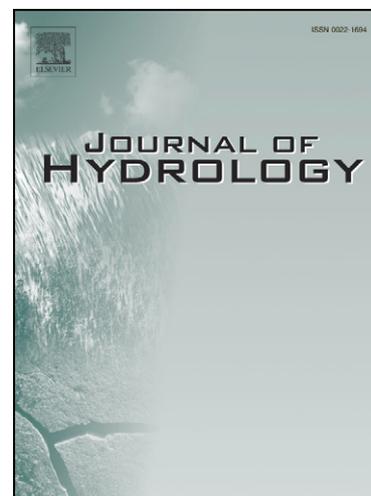
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**Modeling the effect of clay drapes on pumping test response in a
cross-bedded aquifer using multiple-point geostatistics**

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25 **Abstract**

26 This study investigates whether fine-scale clay drapes can cause an anisotropic pumping test
27 response at a much larger scale. A pumping test was performed in a sandbar deposit
28 consisting of cross-bedded units composed of materials with different grain sizes and
29 hydraulic conductivities. The measured drawdown values in the different observation wells
30 reveal an anisotropic or elliptically-shaped pumping cone. The major axis of the pumping
31 ellipse is parallel with the strike of cm to m-scale clay drapes that are observed in several
32 outcrops. To determine (1) whether this large-scale anisotropy can be the result of fine-scale
33 clay drapes and (2) whether application of multiple-point geostatistics can improve
34 interpretation of pumping tests, this pumping test is analysed with a local 3D groundwater
35 model in which fine-scale sedimentary heterogeneity is modelled using multiple-point
36 geostatistics. To reduce CPU and RAM demand of the multiple-point geostatistical simulation
37 step, edge properties indicating the presence of irregularly-shaped surfaces are directly
38 simulated. Results show that the anisotropic pumping cone can be attributed to the presence
39 of the clay drapes. Incorporating fine-scale clay drapes results in a better fit between observed
40 and calculated drawdowns. These results thus show that fine-scale clay drapes can cause an
41 anisotropic pumping test response at a much larger scale and that the combined approach of
42 multiple-point geostatistics and cell edge properties is an efficient method for integrating fine-
43 scale features in larger scale models.

44
45 **Keywords:** Multiple-point geostatistics; Groundwater Flow; Heterogeneity; Pumping test;
46 Upscaling; Cross-bedding

47
48 **Highlight 1:** Fine-scale clay drapes can cause anisotropic pumping test response at larger
49 scale

50 **Highlight 2:** An approach using multiple-point geostatistics and edge properties is proposed

51 **Highlight 3:** Incorporating fine-scale clay drapes results in a better fit of drawdowns

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55 1. Introduction

56 Clay drapes are thin irregularly-shaped layers of low-permeability material that are often
57 observed in different types of sedimentary deposits (Reineck and Singh 1973). Their
58 thicknesses are often only a few centimetres (Houthuys 1990; Stright 2006). Despite their
59 limited thicknesses, several studies indicate that they may influence subsurface fluid flow and
60 solute transport at different scales (Ringrose et al. 1993; Willis and White 2000; Morton et al.
61 2002; Mikes 2006; Stright 2006; Li and Caers 2011; Huysmans and Dassargues 2009). It
62 seems that structural heterogeneity (such as clay drapes) at fine scale might yield anisotropy
63 at large scale, whereas "random" heterogeneity may yield an isotropic behavior at large scale.
64 However, many studies show that the effect of fine-scale heterogeneity is limited to fine
65 scales and averaged out on larger scales and that consequently the type of geological
66 heterogeneity that needs to be taken into account depends on the scale of the problem under
67 consideration (Schulze-Makuch and Cherkauer 1998; Schulze-Makuch et al. 1999; Beliveau
68 2002; Neuman 2003; Eaton 2006). Van den Berg (2003) found for example that anisotropy
69 caused by lamination is small compared to the influence of larger scale heterogeneities so that
70 these sedimentary structures only cause anisotropy on a smaller scale. It is therefore unclear
71 whether centimeter-scale clay drapes can influence groundwater flow and solute transport at
72 scales exceeding the meter-scale. This study therefore investigates whether fine-scale clay
73 drapes can cause an anisotropic pumping test response at a much larger scale. This study is
74 based on measured drawdown values from a pumping that reveal an anisotropic or elliptical-
75 shaped pumping cone: the major axis of the pumping ellipse is parallel with the strike of the
76 centimetre to meter-scale clay drapes that are observed and measured in several outcrops and
77 quarries. This study quantitatively investigates whether this large-scale anisotropy can be the
78 result of fine-scale clay drapes.

79

80 It is very difficult to incorporate clay drapes in aquifer or reservoir flow models, because of
81 their small size and the complexity of their shape and distribution. In standard upscaling
82 approaches (Renard and de Marsily 1997; Farmer 2002), the continuity of the clay drapes is
83 not preserved (Stright 2006). Multiple-point geostatistics is a technique that has proven to be
84 very suitable for simulating the spatial distribution of such complex structures (Strebelle
85 2000; Strebelle 2002; Caers and Zhang 2004; Hu and Chugunova 2008; Huysmans and
86 Dassargues, 2009; Comunian et al. 2011; dell'Arciprete et al. 2012). Multiple-point
87 geostatistics was developed for modelling subsurface heterogeneity as an alternative to
88 variogram-based stochastic approaches that are generally not well suited to simulate complex,
89 curvilinear, continuous, or interconnected structures (Koltermann and Gorelick 1996; Fogg et
90 al. 1998; Journel and Zhang 2006). Multiple-point geostatistics overcomes the limitations of
91 variogram by directly inferring the necessary multivariate distributions from training images
92 (Guardiano and Srivastava 1993; Strebelle and Journel 2001; Strebelle 2000; Strebelle 2002;
93 Caers and Zhang 2004; Hu and Chugunova 2008). In this way, multiple-point geostatistics
94 provides a simple mean to integrate a conceptual geological model in a stochastic simulation
95 framework (Comunian et al., 2011). In the field of groundwater hydrology, application of
96 multiple-point geostatistics to modeling of groundwater flow and transport in heterogeneous
97 media has become an active research topic in recent years. Feyen and Caers (2006) apply the
98 method to a synthetic two-dimensional case to conclude that the method could potentially be a
99 powerful tool to improve groundwater flow and transport predictions. Several recent studies
100 apply the method to build realistic (hydro)geological models based on field observations on
101 geological outcrops and logs (Huysmans et al. 2008; Ronayne et al. 2008; Huysmans and
102 Dassargues 2009; Bayer et al. 2011; Comunian et al. 2011; Le Coz et al. 2011; dell'Arciprete
103 et al. 2012). In large-scale three-dimensional grids multiple-point geostatistics may be
104 computationally very intensive. Several studies focus on improved implementations of the

105 multiple-point statistics techniques to make the algorithms more powerful and
106 computationally efficient (e.g., Mariethoz et al. 2010; Straubhaar et al. 2011). Huysmans and
107 Dassargues (2011) developed the method of “direct multiple-point geostatistical simulation of
108 edge properties” which enables simulating thin irregularly-shaped surfaces with a smaller
109 CPU and RAM demand than the conventional multiple-point statistical methods. This method
110 has been applied on simple test cases (Huysmans and Dassargues 2011) and the present study
111 is the first to apply the method of “direct multiple-point geostatistical simulation of edge
112 properties” to a full-scale three-dimensional groundwater model. In this way, this study
113 investigates whether the combined approach of using multiple-point geostatistics and edge
114 properties is an efficient and valid method for integrating fine-scale features in larger scale
115 models.

116

117 A last scientific goal of this paper is to determine the added benefits of explicitly
118 incorporating clay drape presence for inverse modelling of pumping tests. Several authors
119 have shown that incorporating heterogeneity can result in improved correspondence between
120 calculated and observed hydraulic heads (e.g., Herweijer 1996; Lavenue and de Marsily 2001;
121 Kollet and Zlotnik 2005; Ronayne et al. 2008; Harp and Vesselinov 2011). However, some
122 authors show that incorporating additional data about heterogeneity does not always result in
123 better calibration (e.g., Hendricks Franssen and Stauffer 2005). This paper quantifies the
124 change in calibration error when clay drapes are incorporated in groundwater flow models.

125

126

127 **2. Material and methods**

128 The methodology followed in this study consists of the following steps. First, field data are
129 obtained in an extensive field campaign mapping sedimentary heterogeneity and fine-scale air

130 permeability. Secondly, a training image displaying clay drape occurrence is constructed based
131 on the geological and hydrogeological field data obtained from this field campaign. Thirdly,
132 this training image with small pixel size is converted into an upscaled edge training image
133 which is used as input training image to perform multiple-point SNESIM (Single Normal
134 Equation SIMulation) simulations. The SNESIM algorithm (Strebelle 2002) allows borrowing
135 multiple-point statistics from the training image to simulate multiple realizations of facies
136 occurrence. SNESIM is a pixel-based sequential simulation algorithm that obtains multiple-
137 point statistics from the training image, exports it to the geostatistical numerical model, and
138 anchors it to the actual subsurface hard and soft data. In this study, the resulting simulations
139 indicate at which cell edges horizontal or vertical clay drapes are present. This information is
140 incorporated in a local 3D groundwater model of the pumping test site by locally adapting
141 vertical leakance values and by locally inserting horizontal flow barriers. All hydraulic
142 parameters including the clay drapes properties are calibrated using the measured drawdown
143 time series in six observation wells.

144

145 **2.1 Geological setting**

146 The pumping test site is situated in Bierbeek near Leuven (Belgium) as shown in Figure 1.
147 The subsurface geology in this area consists of a 4m-thick cover of sandy loam from
148 Pleistocene age, 35m of Middle-Eocene Brussels Sands and 12 m of low permeable Early-
149 Eocene Ieper Clay (Figure 2). At the pumping test site the Brussels Sands aquifer acts as an
150 unconfined aquifer. All pumping and observation wells of the pumping test are screened in
151 the Brussels Sands. The Brussels Sands formation is an early Middle-Eocene shallow marine
152 sand deposit in Central Belgium (Fig. 1). Its geological features are extensively covered in
153 Houthuys (1990) and Houthuys (2011). This aquifer is a major source of groundwater in
154 Belgium and was studied at the regional scale by Peeters et al. (2010). The most interesting

155 feature of these sands in terms of groundwater flow and transport is the complex geological
156 heterogeneity originating in its depositional history. The Brussels Sands are a tidal sandbar
157 deposit. Its deposition started when a strong SSW-NNE tidal current in the early Middle-
158 Eocene produced longitudinal troughs, that were afterwards filled by sandbar deposits. In
159 these sandbar deposits, sedimentary features such as cross-bedding, mud drapes and
160 reactivation surfaces are abundantly present (Houthuys 1990; Houthuys 2011). The
161 orientation of most of these structures is related to the NNE-orientation of the main tidal flow
162 during deposition.

163

164 **2.2 Pumping test**

165 In February 1993, a pumping test was performed in Bierbeek (Belgium) under the authority of
166 the company TUC RAIL N.V. in the framework of high-speed train infrastructure works. One
167 pumping well (PP1) and six observations wells were drilled in the 35m-thick coarse facies of
168 the Brussels Sands (Figure 3). The observation wells are situated between 4m and 75m from
169 the pumping well and are located in different orientations. Before pumping the water table
170 was at 49.8 m. During the pumping test, there was 72 hours of pumping in well PP1 with a
171 flow rate of 2120 m³/day. Water level in six observation wells was continuously monitored
172 during pumping and during an additional 24 hours of recovery after pumping. The pumping
173 test was interpreted by inverse modelling using a numerical method described in Lebbe and
174 De Breuck (1995). This analysis showed that the best calibration was obtained assuming
175 horizontal anisotropy in the coarse facies of the Brussels Sands. The maximal horizontal
176 hydraulic conductivity was found to be 28.3 m/day while the minimal horizontal hydraulic
177 conductivity was 13.4 m/day. The principal direction of maximal horizontal hydraulic
178 conductivity corresponds to N 115°48' E (TUC RAIL N.V., 1993). This principal orientation

179 is exactly perpendicular to the SSW-NNE orientation of the main tidal flow during deposition
180 and the mud drapes in the Brussels Sands.

181

182

183 **2.3 In situ mapping and measurement of clay drape properties**

184 The Brussels Sands outcrops in the Bierbeek quarry are used as an analog for the Brussels
185 Sands found in the subsurface at the pumping test site. This quarry is located at approximately
186 500 m from the pumping test site (Figure 1). This outcrop of approximately 1200 m² was
187 mapped in detail with regard to the spatial distribution of sedimentary structures and
188 permeability in Huysmans et al. (2008). A total of 2750 cm-scale air permeability
189 measurements were carried out in situ on different faces of the Bierbeek quarry to
190 characterize the spatial distribution of permeability. From the hydrogeological point of view
191 in the present study, the main interest lies in the occurrence and geometry of structures with
192 high and low hydraulic conductivity. In this perspective, the Brussels Sands can be regarded
193 as consisting of horizontal permeable sand layers of approximately 1m thick intercalated by
194 horizontal low-permeable clay-rich bottomsets and inclined low-permeable clay drapes.
195 Figure 4 shows a field picture and a geological interpretation of the typical clay-sand patterns
196 in the Bierbeek quarry. More details about the spatial distribution of the fine-scale
197 sedimentary structures and measured permeability in the Brussels Sands can be found in
198 Huysmans et al. (2008).

199

200 **2.4 Training image construction**

201 The observed spatial patterns of clay drape occurrence are explicitly represented in a training
202 image. Training images are essential to multiple-point geostatistics. In multiple-point
203 geostatistics, "training images" are used to characterize the patterns of geological

204 heterogeneity. A training image is an explicit grid-based representation of the expected
205 geological patterns. In the simulation step, patterns are borrowed from the training image and
206 reproduced in the simulation domain. (Guardiano and Srivastava 1993; Strebelle and Journel,
207 2001; Caers and Zhang 2004). More information about the theory behind multiple-point
208 geostatistics can be found in Strebelle (2000) and Strebelle (2002). Description of the
209 different multiple-point algorithms can be found in the following papers: SNESIM (Strebelle
210 2002; Liu 2006), FILTERSIM (Zhang et al. 2006; Wu et al. 2008), SIMPAT (Arpat and Caers
211 2007), HOSIM (Mustapha and Dimitrakopoulos 2010) and the Direct Sampling method
212 (Mariethoz et al. 2010).

213
214 In this study, a two-dimensional fine-scale training image of clay and sand occurrence of the
215 Brussels Sands was constructed based on the in situ mapping in the Bierbeek quarry. In the
216 third dimension perpendicular to the 2D training image, layering and clay drapes are very
217 continuous as shown on figure 5 which shows quarry wall pictures in the NNE-direction and
218 the perpendicular orientation. While the picture of the NNE-oriented face displays cross-
219 bedding and inclined mud drapes, the picture of the perpendicular face shows continuous
220 horizontal layering (Figure 5). Therefore all layers and sedimentary structures are assumed to
221 be continuous in that direction. The incorporation of 3D simulations based on several 2D
222 training images in different directions following the approaches discussed in Comunian et al.
223 (2012) could be interesting future work. The two-dimensional fine-scale training image along
224 the NNE-direction (Figure 6) shows an alternation of sand-rich and clay-rich zones. More
225 details about construction of this training image can be found in Huysmans and Dassargues
226 (2009). This training image will be used in section 2.6 where multiple-point statistics are
227 borrowed from this training image to simulate realizations of clay drape occurrence to be used
228 as input for the local groundwater flow model.

229

230 **2.5 Groundwater flow model**

231 The groundwater flow model is a three-dimensional local model of 600m x 600 m x 30.4m
232 including all pumping and observation wells from the pumping test in Bierbeek described in
233 section 2.2. The size of the model in the horizontal direction was chosen to be 600 m since
234 analysis of the pumping test data showed that no drawdown is measured at 300 m from the
235 pumping well (TUCRAIL N.V., 1993). The model is oriented along the N22.5°E direction
236 which is parallel to the direction of the main geological structures and the main anisotropy
237 axis of the observed drawdowns from the pumping test. The top of the Ieper Clay deposits
238 represents the impermeable bottom of the model due to the low permeability of this unit
239 (Huysmans and Dassargues, 2006). The top of the model corresponds to an elevation of 49.8
240 m, which corresponds to the initial groundwater level before pumping. This means that only
241 one geological layer is present in the model, i.e., the Brussels Sands. The model consists of a
242 central inner zone of 55m x 100m x 15.3m including all well screens and an outer zone
243 (Figure 7). This inner zone consists of 51 cells in the x-direction, 183 cells in the y-direction
244 and 51 layers. In the central inner zone where all the well screens are situated, a very small
245 grid cell size of 0.3m x 0.3m x 0.3m is adopted so that individual clay drapes can be explicitly
246 incorporated in the model in this zone. In the outer zone, larger grid cell sizes between 0.45m
247 and 82m in the horizontal direction and layer thicknesses between 3m and 6m are chosen. For
248 numerical reasons, the dimensions of the grid cells do not exceed 1.5 times the dimensions of
249 their neighboring cells. The total model consists of 213 cells in the x-direction, 361 cells in
250 the y-direction and 54 layers. The total number of grid cells in the model is thus 4,152,222
251 cells. The model is run in transient conditions with a total time length of 4510 minutes
252 subdivided into 99 time intervals. Piezometric heads are prescribed at the lateral boundaries of
253 all model layers. In the pumping well PP1, a pumping rate of 2120 m³/day is applied during

254 72 hours. Initial hydraulic conductivity and storage parameters were taken from a previous
255 interpretation (TUCRAIL N.V., 1993) and calibrated afterwards. A total of 594 observed
256 heads measured in six observation wells (Figure 3) from two minutes after the start of
257 pumping until 240 minutes after stopping of pumping are available for calibration. The
258 differential equations describing groundwater flow are solved by PMWIN (Chiang and
259 Kinzelbach 2001), which is a pre- and post-processor for MODFLOW (McDonald and
260 Harbaugh 1988), using a block-centered, finite-difference, method.

261
262 Two model variants are run and calibrated separately. First, a homogeneous and horizontally
263 isotropic model without clay drapes is run. In this model, different values for horizontal and
264 vertical hydraulic conductivity are allowed, but no anisotropy of hydraulic conductivity in the
265 horizontal direction is introduced. Calibration is performed for adapting values of horizontal
266 hydraulic conductivity, vertical hydraulic conductivity and specific storage. The second
267 model incorporates a random clay drape realization as described in section 2.6. In this model,
268 calibration of the following additional parameters is performed: clay drape thickness and clay
269 drape hydraulic conductivity in model layers 4 to 54 and anisotropy factor of shallow layers 1
270 to 3 in which clay drapes are not explicitly incorporated. In layers 1 to 3 clay drapes are not
271 explicitly incorporated since these belong to the outer zone of the model The spatial
272 distribution of the clay drapes is not changed during calibration. Storage is assumed identical
273 in the sand and the clay drapes. In this model, the only heterogeneity and anisotropy of
274 hydraulic conductivity is related to the presence of clay drapes. Background hydraulic
275 conductivity of layers 4 to 54 is homogeneous and isotropic so that the effect of clay drapes
276 on hydraulic heads can be determined without influence of other heterogeneity or anisotropy
277 effects. By comparing the results of these two model variants, the effects of clay drapes on the
278 piezometric depression cone and on the calibration results can be quantified. This approach of

279 comparing a heterogeneous model with a homogeneous equivalent was for example also
280 applied in Mariethoz et al. (2009). For both models, a two-step calibration procedure is
281 adopted. First, a sensitivity analysis and trial-and-error subjective calibration is performed and
282 second, the optimal model from manual calibration is further calibrated using PEST (Doherty
283 et al. 1994). The sensitivity analysis consists of varying the adjustable parameters (horizontal
284 hydraulic conductivity, vertical hydraulic conductivity, specific storage and clay drape
285 parameters) and assessing their effect on simulated drawdowns.

286

287 **2.6 Clay drapes simulation using multiple-point geostatistical simulation of edge** 288 **properties**

289

290 In order to incorporate clay drapes showing patterns similar to the training image of Figure 6
291 (left) in the groundwater flow model, the technique of direct multiple-point geostatistical
292 simulation of edge properties (Huysmans and Dassargues 2011) is used. This technique was
293 designed to simulate thin complex surfaces such as clay drapes with a smaller CPU and RAM
294 demand than the conventional multiple-point statistical methods. Instead of pixel values, edge
295 properties indicating the presence of irregularly-shaped surfaces are simulated using multiple-
296 point geostatistical simulation algorithms. The training image is upscaled by representing clay
297 drapes as edge properties between cells instead of representing them as objects consisting of
298 several cells. The concept of the edge of a flow model and the associated edge properties was
299 introduced in the work of Stright (2006) as an additional variable. The edge properties are
300 assigned to the cell faces. The cell property used in this study is the presence of clay drapes
301 along cell faces. More details about the method can be found in Huysmans and Dassargues
302 (2011). Figure 6 shows how the fine-scale pixel-based training image (left) is converted into
303 an upscaled edge-based training image (right). The fine-scale training image has a grid cell

304 size of 0.05 m and represents the clay drapes as consisting of pixels with a different pixel
 305 value than the background material. The upscaled edge-based training image has a grid cell
 306 size of 0.30 m and represents the clay drapes as edge properties that indicate the presence of
 307 clay drapes along the edges of all grid cells.

308

309 In this study, the upscaled 30 m by 30 m training image from Figure 6 (right) is used as input
 310 to SNESIM from SGeMS (Remy et al. 2009) to simulate clay drape realizations to be
 311 imported in the inner central zone of the model where individual clay drapes are incorporated.
 312 Vertical 2D realizations of 54.9m by 15.3m are generated. Figure 7 shows a random clay
 313 drape realization that is incorporated in the groundwater flow model.

314

315 The realizations of clay drape presence can be imported in the groundwater flow code
 316 PMWIN using the Horizontal-Flow Barrier (HBF) package and the vertical leakance array
 317 (VCONT array). The HBF package simulates thin vertical low-permeability geological
 318 features, which impede horizontal groundwater flow. They are situated on the boundaries
 319 between pairs of adjacent cells in the finite-difference grid (Hsieh and Freckleton, 1993). A
 320 horizontal-flow barrier is defined by assigning the barrier direction, which indicates the cell
 321 face where the barrier is located, and barrier hydraulic conductivity divided by the thickness
 322 of the barrier (Chiang and Kinzelbach, 2001). Horizontal edges are inserted into PMWIN by
 323 adapting vertical leakance (VCONT array) between two model layers. The VCONT matrix
 324 for every model layer is calculated as

$$325 \quad VCONT = \frac{2}{\frac{\Delta z_u}{(K_z)_u} + \frac{\Delta z_c}{(K_z)_c} + \frac{\Delta z_l}{(K_z)_l}} \quad (1)$$

326 where K is hydraulic conductivity, Δz is thickness and u , c and l respectively represent the
 327 upper layer, semi-confining unit and lower layer as indicated on figure 8. In case a horizontal

328 edge is present in a model cell, the edge is inserted in the model as a semi-confining unit.
329 Initially, it is assumed that all clay drapes in the groundwater flow model have a thickness of
330 0.02 m and a hydraulic conductivity of 0.283 m/d. As mentioned previously, these values are
331 optimized during calibration.

332

333 3. Results

334 Figures 9 and 10 show the resulting model outputs from the homogeneous model and the clay
335 drape model. For both models, automatic calibration using PEST did not result in lower
336 calibration errors, possibly as a result of the large size (4,152,222 grid cells) and complexity
337 of the models. Manual calibration was apparently able to identify a good local minimum,
338 possibly due to the small number of calibrated variables. In the homogeneous and horizontally
339 isotropic model, the best calibration results were obtained with the following set of calibrated
340 parameter values: horizontal hydraulic conductivity of 22.2 m/day, vertical hydraulic
341 conductivity of 4.8 m/day and specific storage of $3 \times 10^{-5} \text{ m}^{-1}$. In the clay drape model, the
342 best calibration results were obtained with the following set of calibrated parameter values:
343 horizontal hydraulic conductivity of 23 m/day vertical hydraulic conductivity of 4 m/day,
344 specific storage of $3 \times 10^{-5} \text{ m}^{-1}$ and a clay drape parameter (hydraulic conductivity of drapes
345 divided by drape thickness) with values between 0.175 and 9.905 day^{-1} . This clay drape
346 parameter varied between the different sublayers of the inner zone of the model reflecting the
347 alteration between layers with thicker or less permeable clay drapes and layers with thinner or
348 more permeable clay drapes. In every layer, a single clay drape parameter was assigned, thus
349 not allowing spatial variation of the clay drape parameter within the model layers.

350 Figure 9 shows piezometric maps at $z = 27 \text{ m amsl}$, i.e., located at the depth of the centre of
351 the pumping well screen, for (1) the homogeneous and horizontally isotropic model and (2)
352 the clay drape model. Figure 9A shows circular hydraulic head contours indicating an

353 isotropic piezometric pumping depression cone resulting from the homogeneity and isotropy
354 of hydraulic conductivity in the horizontal direction. Figure 9B shows the hydraulic head
355 contours for the second model which incorporates clay drapes. These contours are elliptical
356 demonstrating an anisotropic pumping depression cone. The vertical clay drapes cause
357 bending of the hydraulic head contours. Since no other K heterogeneity than the clay drape
358 presence is incorporated in the model, these results show that anisotropic pumping cones at
359 large-scale can be attributed to the presence of fine-scale clay drapes.

360

361 Figure 10 shows calculated versus observed drawdown graphs for (1) the homogeneous and
362 horizontally isotropic model variant and (2) the clay drape model variant. Figure 11 shows
363 calculated and observed drawdown versus time for (1) the homogeneous and horizontally
364 isotropic model variant and (2) the clay drape model variant. The error variance for the
365 isotropic model is $1.24 \text{ E-}2 \text{ m}^2$, while the error variance for the clay drape model is as low as
366 $7.292\text{E-}3 \text{ m}^2$. The residual mean of the homogeneous and isotropic model is $-3.08 \text{ E-}2 \text{ m}$,
367 while the residual mean for the clay drape model is $-2.12 \text{ E-}2 \text{ m}$. Correlation coefficient
368 between observed and simulated drawdowns increases from 0.9906 (homogeneous/isotropic
369 model) to 0.9916 (clay drape model). Incorporating the clay drapes thus results in a better
370 fitting between calculated and observed drawdown values for this pumping test. Especially
371 the larger drawdowns are better reproduced in the clay drape model. This is most obvious on
372 figure 11 which shows that time behaviour is generally well reproduced by the model but that
373 larger drawdowns are better reproduced in the clay drape model. These large drawdowns are
374 measured in observation well PB1.2 which is located close to the pumping well (Figure 3). In
375 the clay drape model, a clay drape is present in the pumped layer between the pumping well
376 and observation well PB1.2 which acts as a flow barrier between those two wells. The

377 presence of this barrier results in a better reproduction of the measured drawdowns by the
378 model.

379

380 **4. Discussion and conclusion**

381 This study has investigated the effect of fine-scale clay drapes on pumping test response. For
382 this purpose, spatial distribution and geometry of clay drapes observed in a cross-bedded
383 aquifer were explicitly incorporated in a local groundwater model of a pumping test site. Clay
384 drape parameters were calibrated in order to reproduce hydraulic head measurements
385 observed during a pumping test. Best calibration results were obtained with a zoned clay
386 drape parameter or drape leakage coefficient, defined as hydraulic conductivity of drapes
387 divided by drape thickness, with values between 0.175 and 9.905 day⁻¹. If the clay drape
388 thickness is assumed to be 0.02 m, this means that hydraulic conductivity of the clay drapes is
389 between 0.0035 m/day (= 4.05x10⁻⁸ m/s) and 0.1981 m/day (= 2.29x10⁻⁶ m/s). These values lie
390 in the hydraulic conductivity interval of silt and silty sand respectively according to Fetter
391 (2001), so these values are realistic and are certainly not chosen unrealistically low. This
392 implies that the anisotropic pumping cone can be reproduced and explained with realistic
393 values for clay drape thickness and hydraulic conductivity and that fine-scale clay drapes can
394 cause an anisotropic pumping test response at a much larger scale.

395

396 Incorporating clay drapes in groundwater models is challenging since they are often irregular
397 curvilinear three-dimensional surfaces which may display a very complex spatial distribution.

398 In this paper, a combined approach of multiple-point geostatistics and edge properties was
399 used to incorporate the clay drapes in the flow model. Clay drapes were represented as grid

400 cell edge properties instead of representing them by pixels. This allowed modelling with a

401 larger grid cell size and thus a smaller CPU and RAM demand. A realistic spatial distribution

402 of clay drape occurrence was simulated using multiple-point geostatistics based on a field-
403 based training image. This combined approach of multiple-point geostatistics and edge
404 properties has shown to be an efficient and valid approach since realistic spatial patterns and
405 geometry of the clay drapes can be preserved in the model without having to represent each
406 clay drape by pixels.

407
408 In order to determine the added value of explicitly incorporating clay drape presence in the
409 flow model for pumping test interpretation, the model was also compared with a
410 homogeneous and isotropic model calibrated on the same pumping test data. Incorporating the
411 clay drapes resulted in a better fit between calculated and observed drawdown values than the
412 homogeneous model.

413

414

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419

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619

620 **Figure captions**

621

622 **Figure 1** Map of Belgium showing Brussels Sands outcrop and subcrop area (shaded part)
623 (modified after Houthuys (1990) and inset showing the location of the pumping test site and
624 the Bierbeek quarry

625

626 **Figure 2** Geological EW-profile through the study area, modified after Houthuys (1990)

627

628 **Figure 3** Pumping test configuration showing the pumping well (white circle) and
629 observation wells (black circles) and the orientation and delineation of the central inner zone
630 of the local groundwater model. Observation well PB1.1 is not screened in the Brussels Sands
631 and is therefore not used in this study.

632

633 **Figure 4** Raw (left) and interpreted (right) field picture, showing foresets, bottomsets and
634 clay drapes in the Brussels Sands observed in the Bierbeek quarry

635

636 **Figure 5** (A) Photomosaic of Bierbeek quarry wall in NNE direction showing cross-bedding
637 and quasi-horizontal clay-rich bottomsets. Height of quarry wall is approximately 4–5 m and
638 (B) photomosaic of N45°W oriented Bierbeek quarry wall showing continuous horizontal
639 layers. Length of quarry wall shown on picture is approximately 22 m.

640

641 **Figure 6** (A) Vertical two-dimensional training image of 30 m by 30 m in NNE direction:
642 sand facies (white), clay-rich facies (black) modified from Huysmans and Dassargues (2009)
643 and (B) the corresponding edge training image modified from Huysmans and Dassargues
644 (2011)

645

646 **Figure 7** Groundwater flow model grid and edge realization

647

648 **Figure 8** Grid configuration used for the calculation of VCONT in the presence of a
649 horizontal clay drape between two cells (modified after Chiang and Kinzelbach 2001)

650

651 **Figure 9** Piezometric maps at $z = 27$ m corresponding to the central level of the pumping well
652 screen, for (A) the homogeneous and horizontally isotropic model and (B) the clay drape
653 model showing drawdown after two days

654

655 **Figure 10** Calculated versus observed drawdown graphs for (1) the homogeneous and
656 horizontally isotropic model variant and (2) the clay drape model variant

657

658 **Figure 11** Calculated and observed drawdown versus time for (A) the homogeneous and
659 horizontally isotropic model variant and (B) the clay drape model variant

660

Figure 1

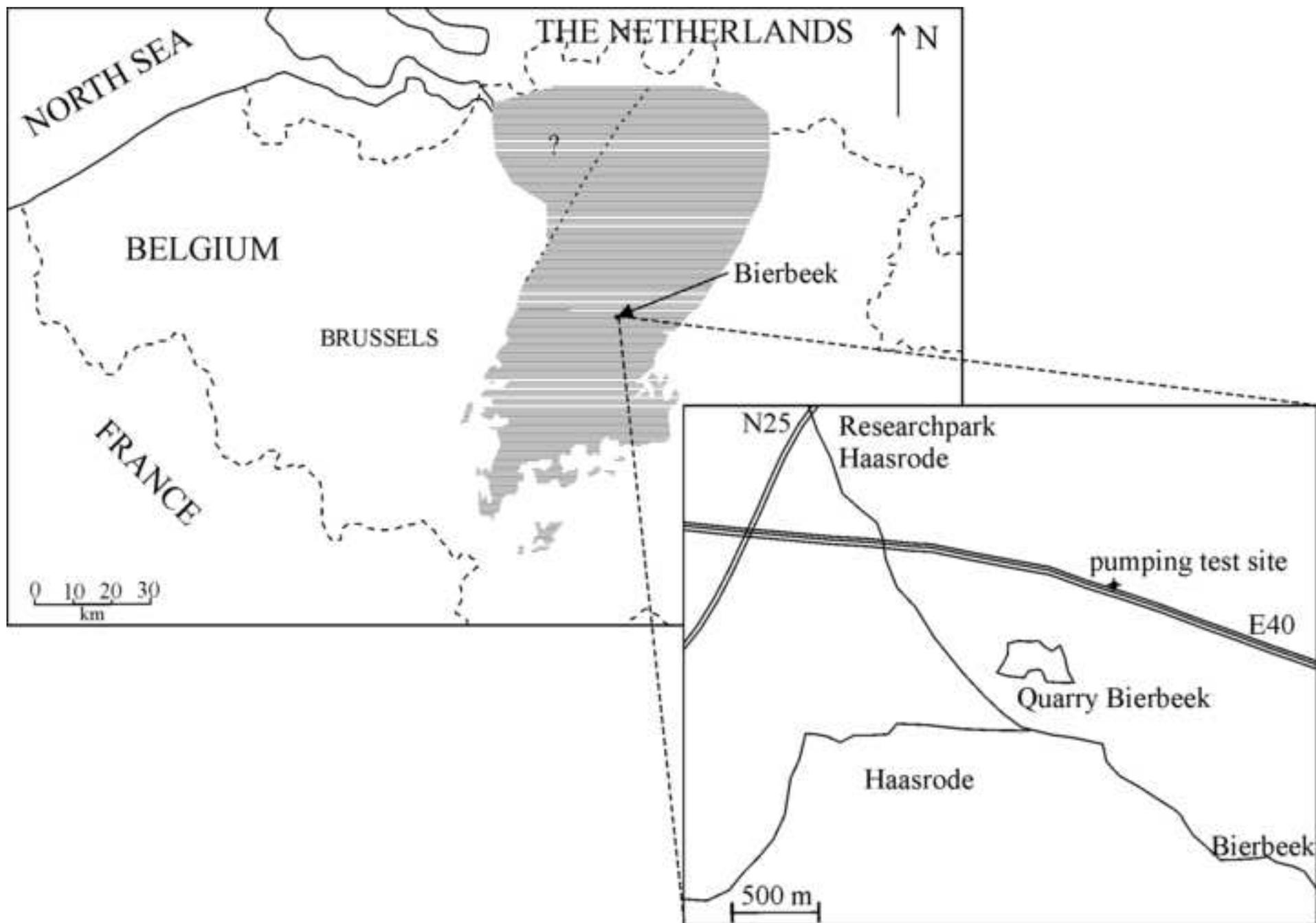
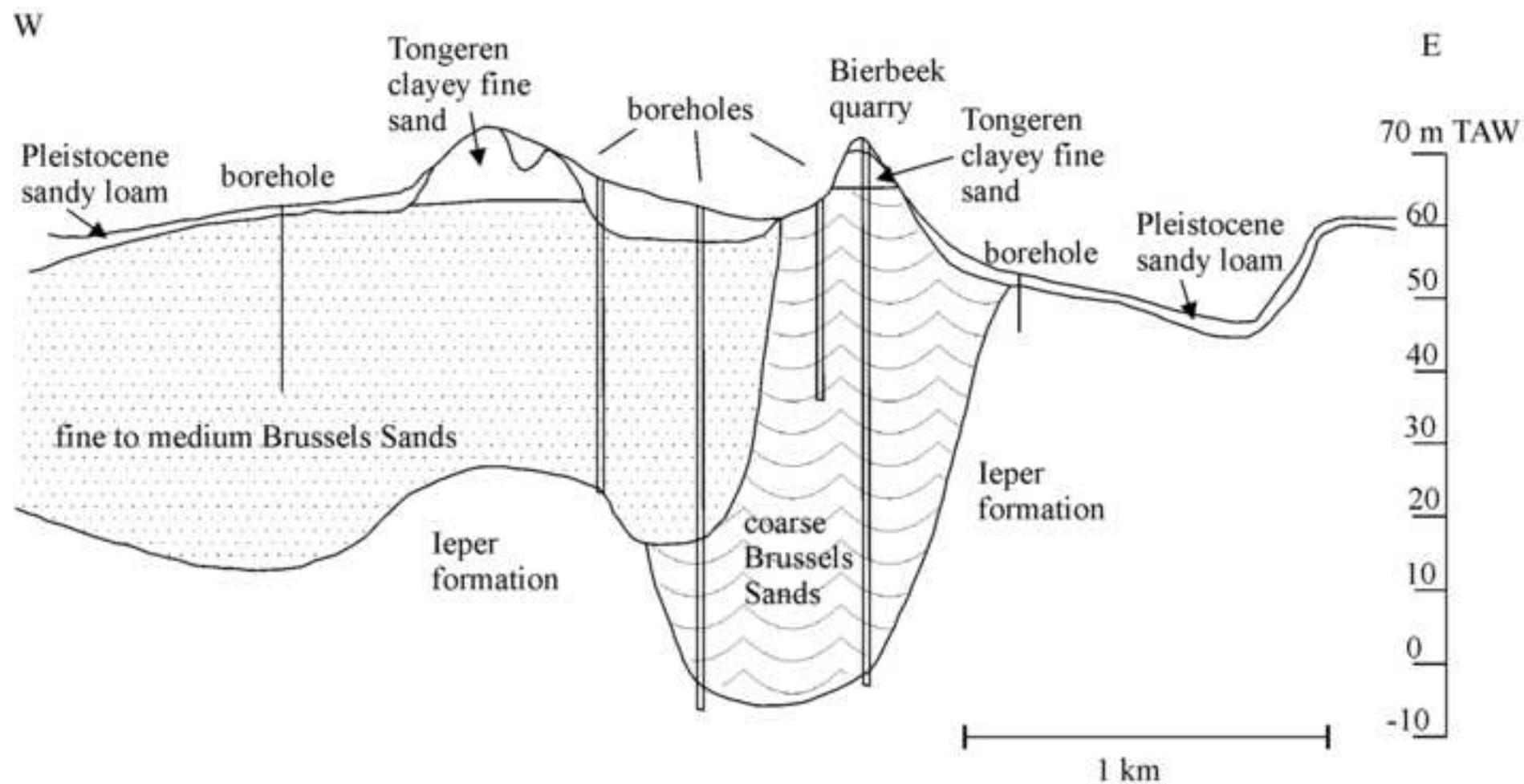


Figure 2



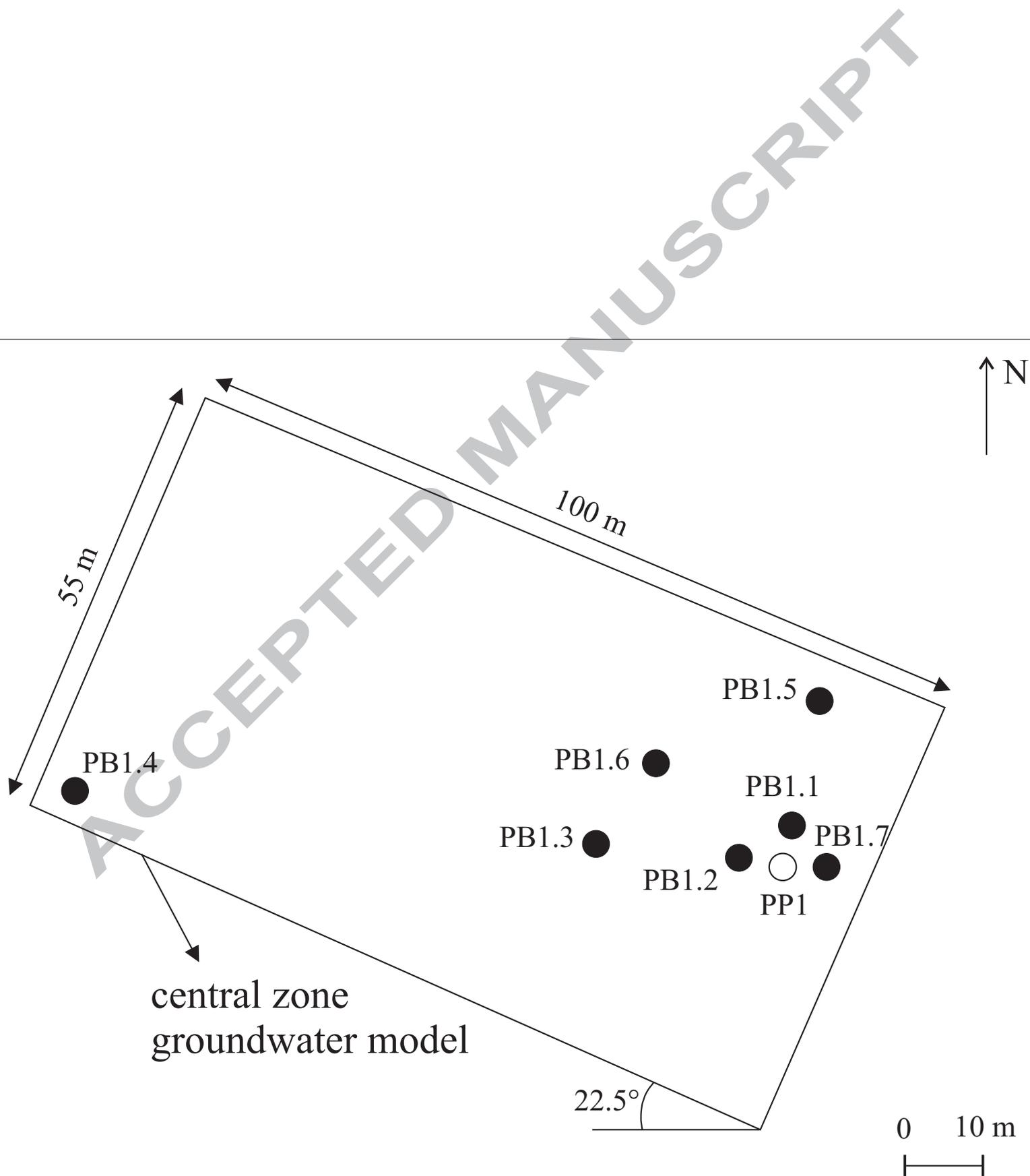
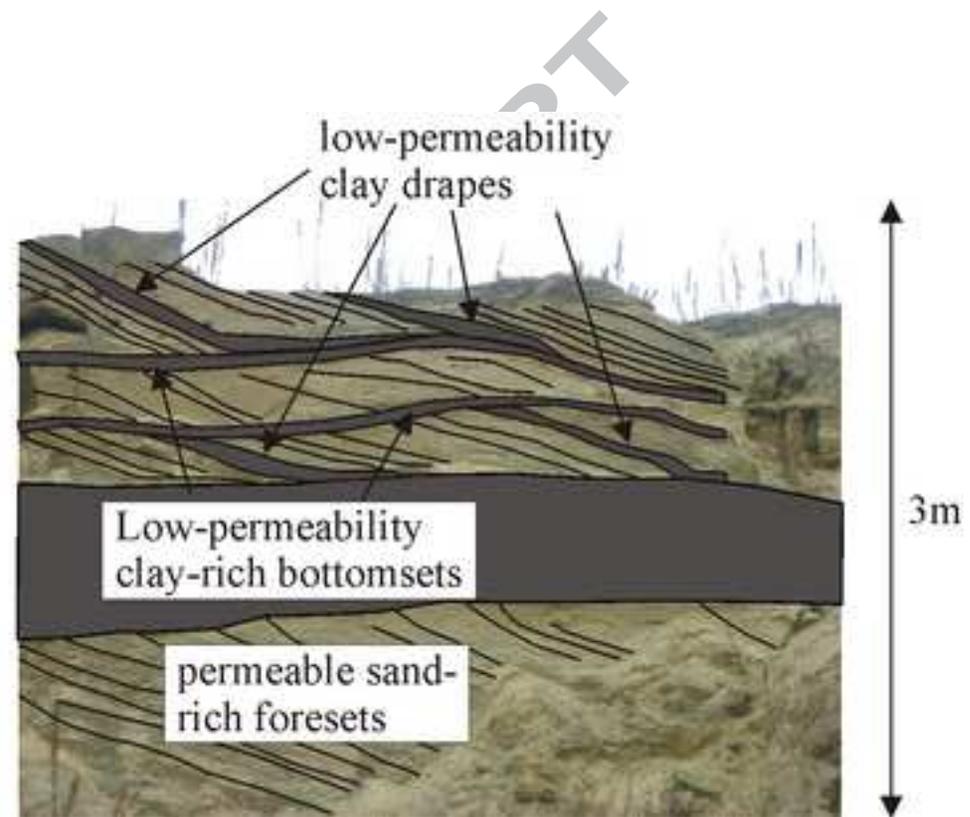


Figure 4



(A)



(B)



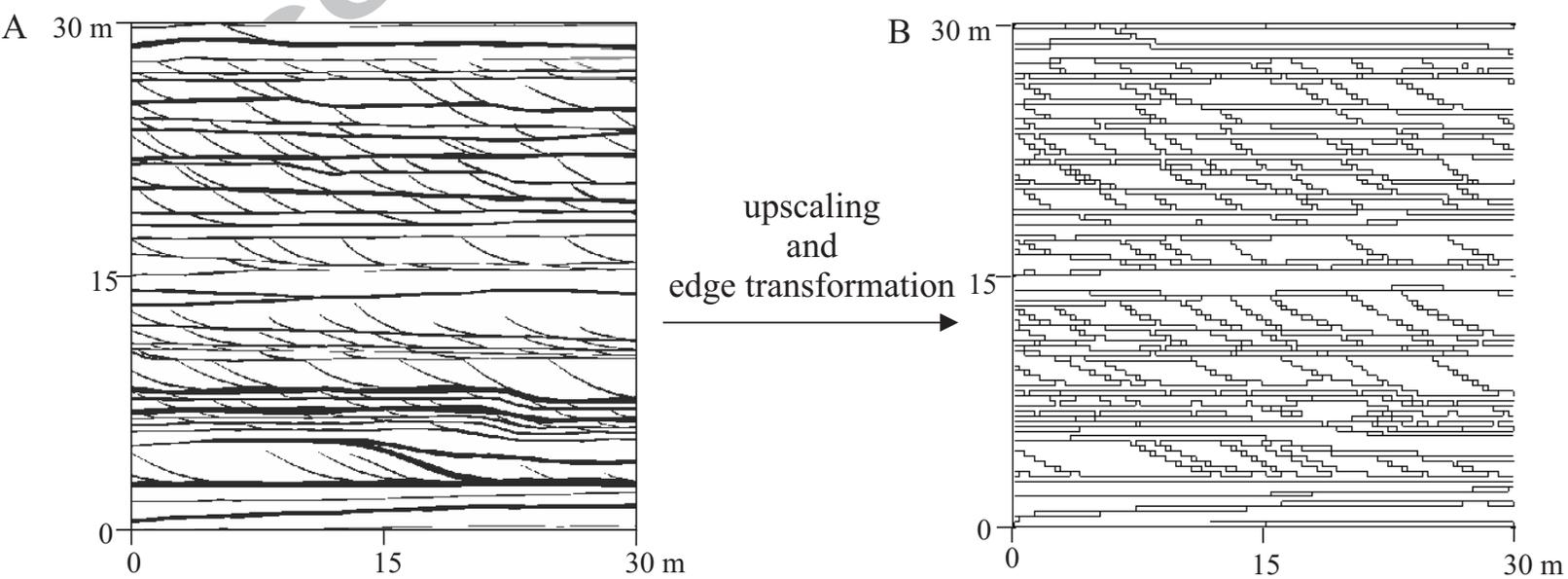
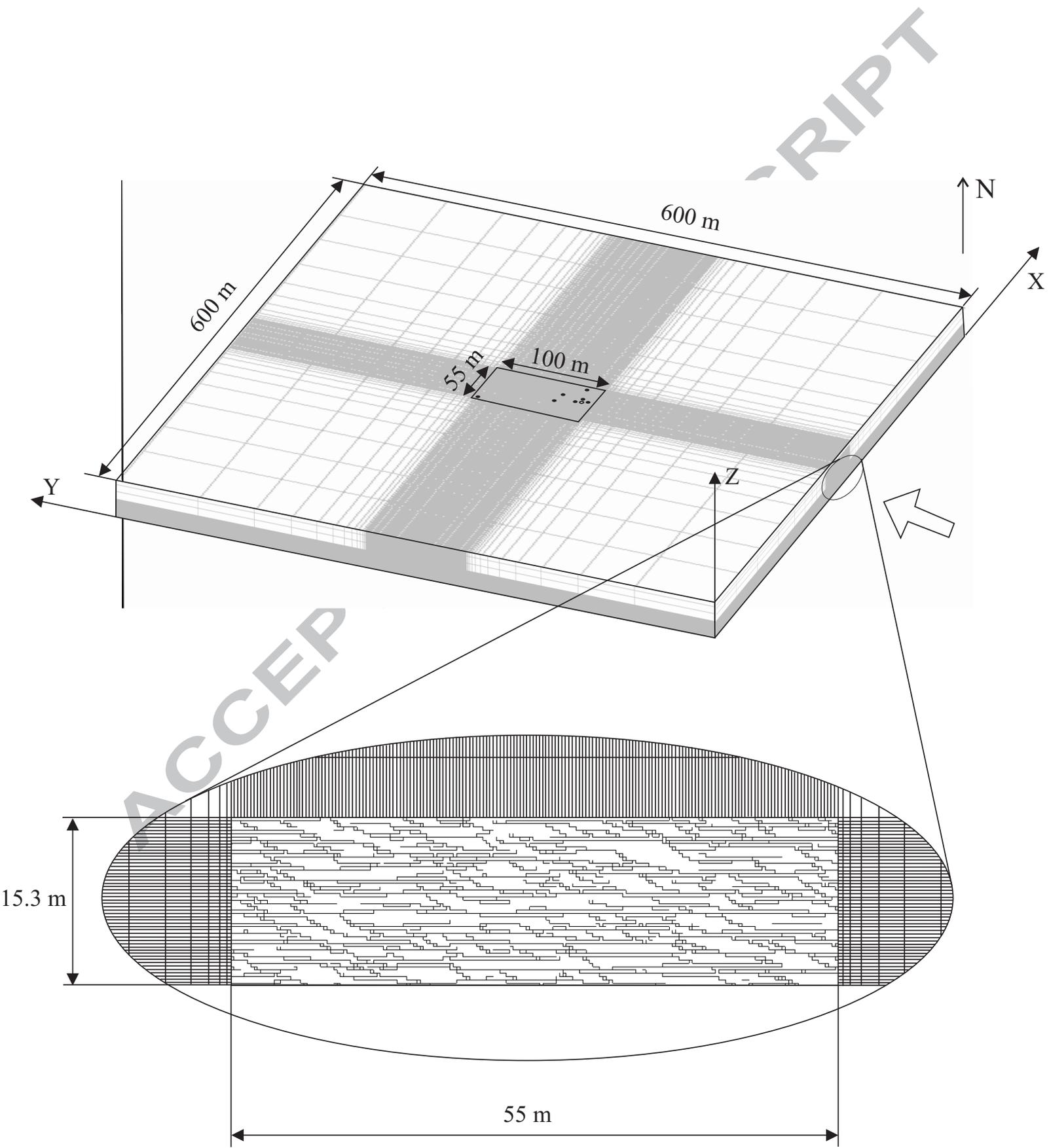
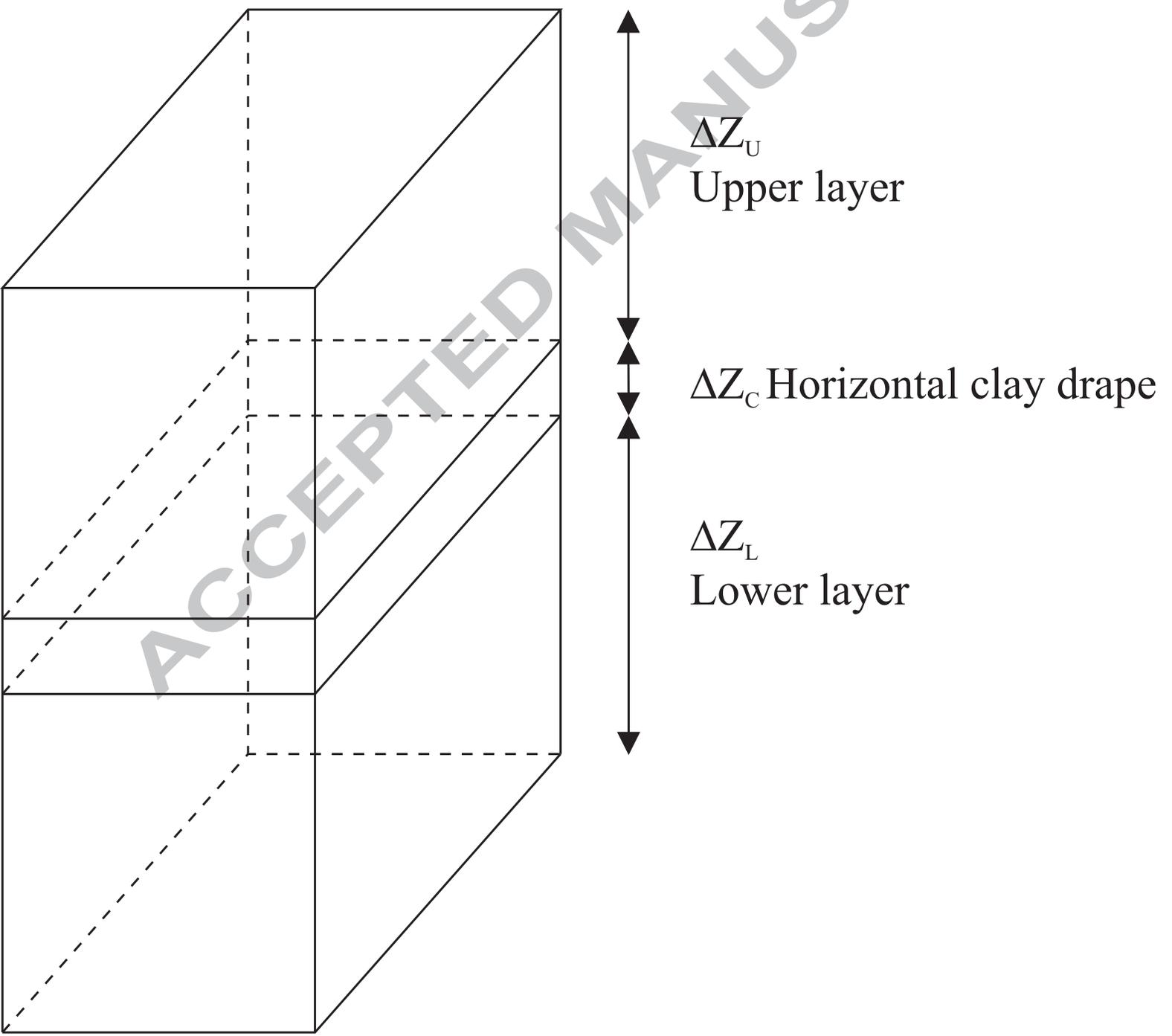
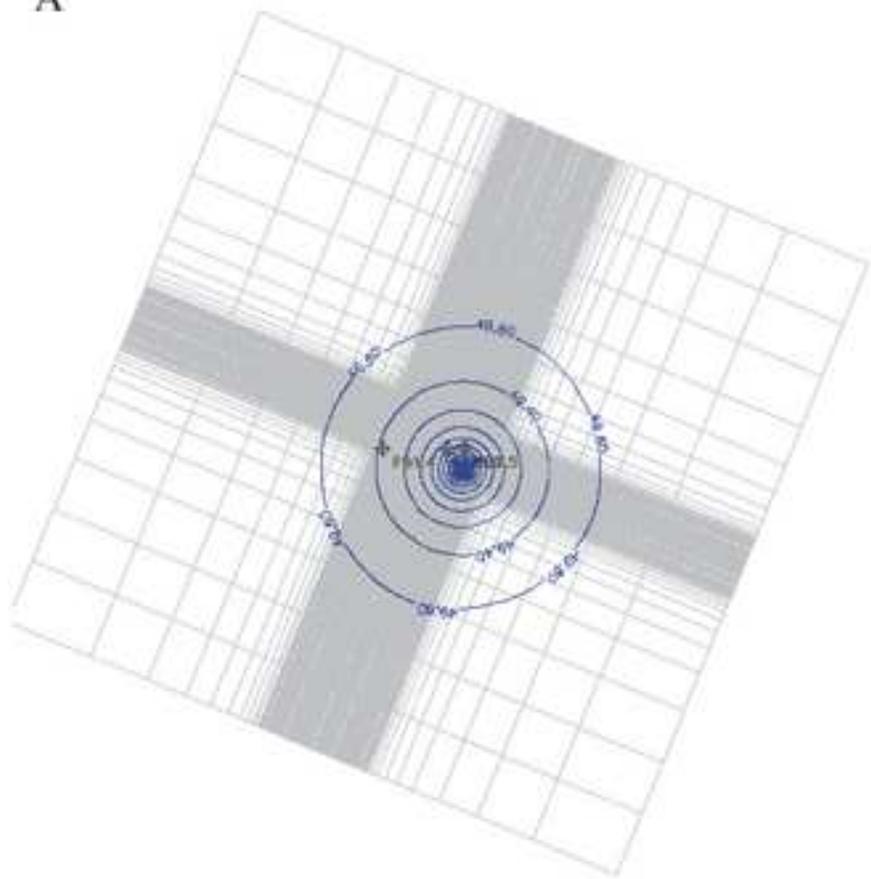


Figure 7

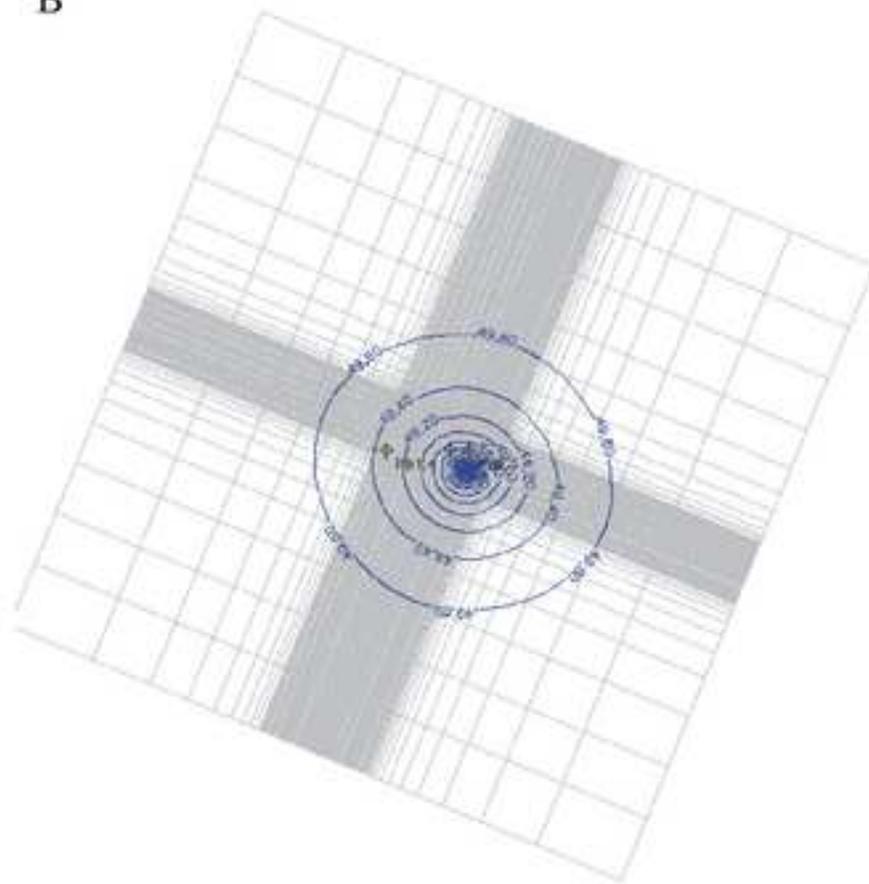




A

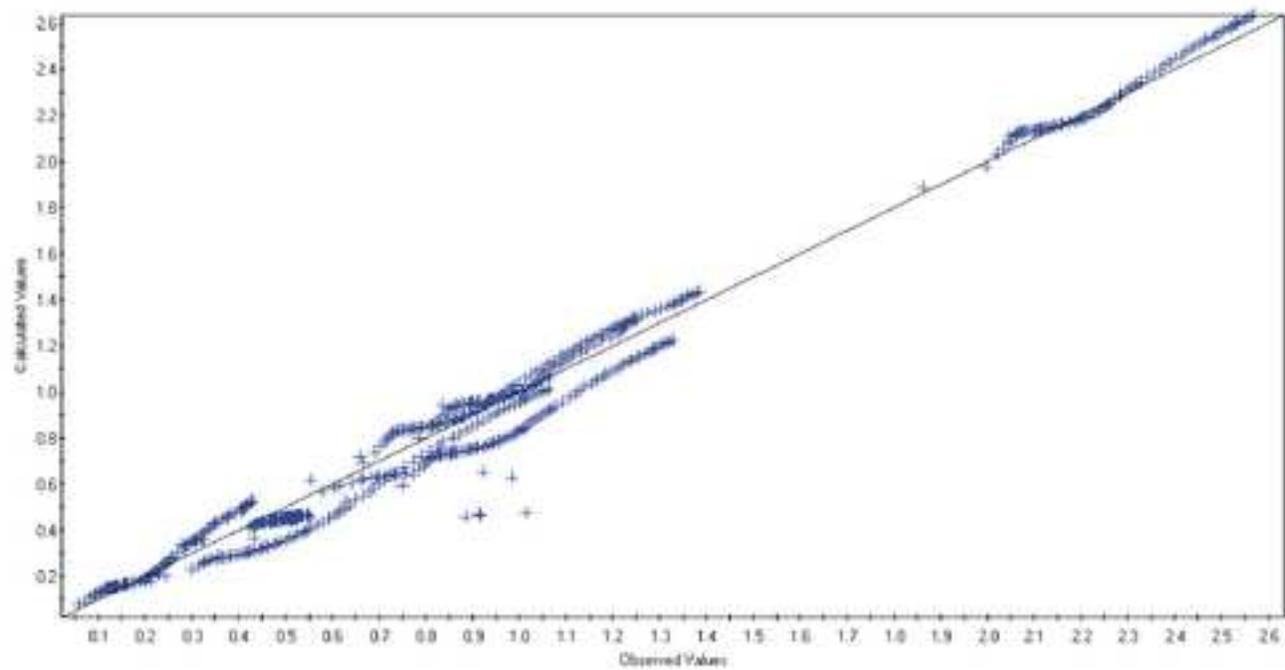
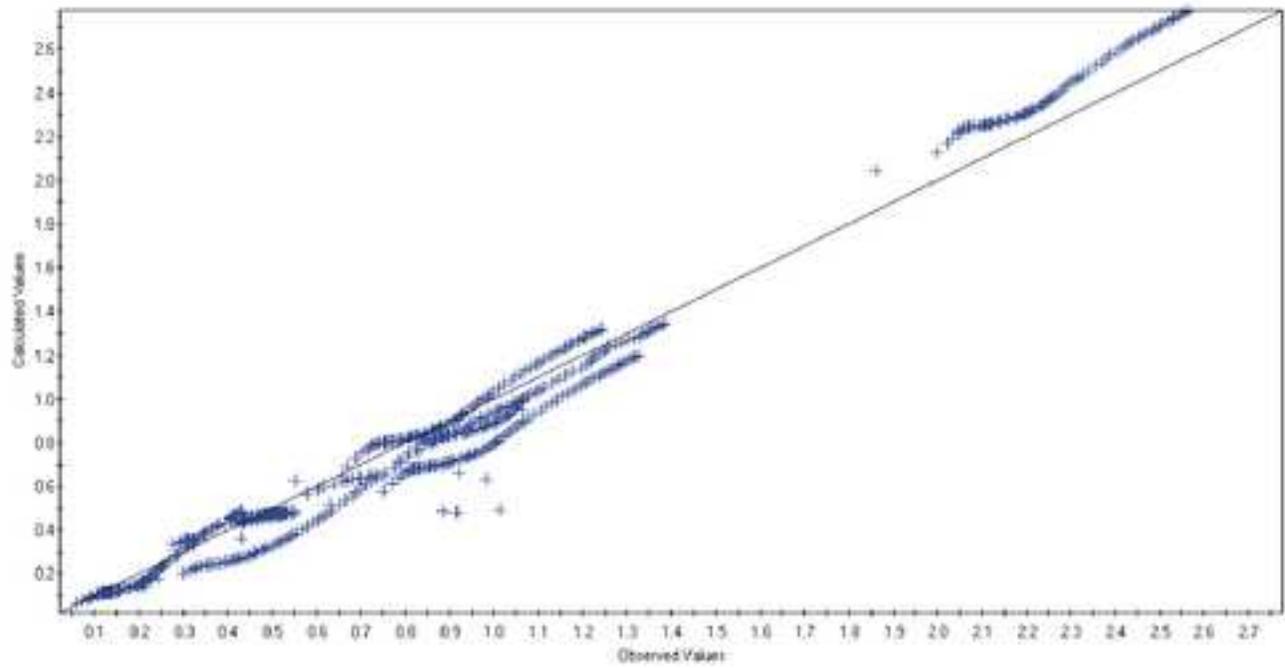


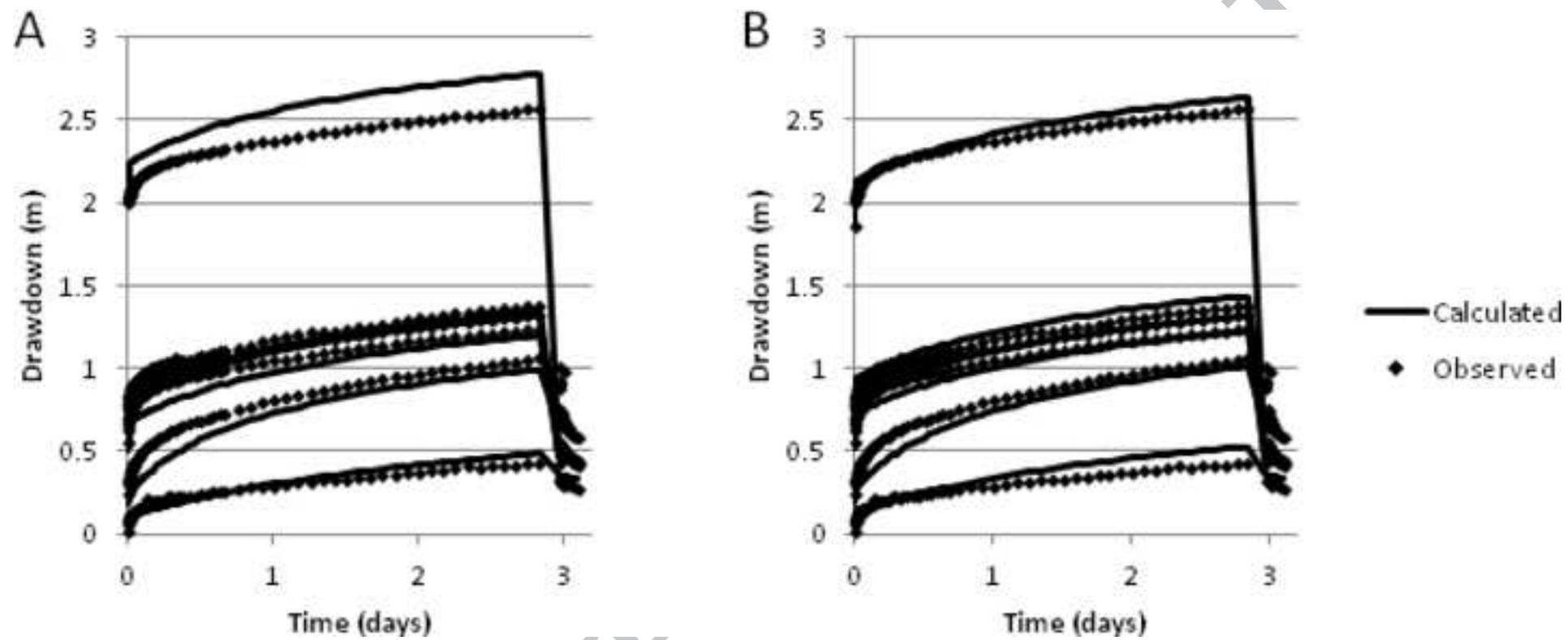
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Figure 10





Highlight 1: Small-scale clay drapes can cause anisotropic pumping test response at larger scale

Highlight 2: An approach using multiple-point geostatistics and edge properties is proposed

Highlight 3: Incorporating small-scale clay drapes results in a better fit of drawdowns

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