

Controls on preferential recharge to Chalk aquifers

A. M. Ireson * A. P. Butler

Department of Civil and Environmental Engineering, Imperial College London, UK

Abstract

There is evidence that, under certain conditions, rapid preferential recharge via the fracture network can occur in Chalk aquifers. This has potentially important implications for contaminant migration through the Chalk unsaturated zone, CUZ, and for groundwater flooding in Chalk catchments. In the case of groundwater flooding, deficiencies in modelling aquifer response have been attributed to inadequate representation of flow processes in the CUZ (Habets et al., 2010). In this paper we consider two complementary approaches for assessing controls on preferential recharge to Chalk aquifers: an empirical approach and a physically-based modelling approach. We show that the main controls on preferential recharge to Chalk aquifers are the characteristics of rainfall events, in terms of duration and intensity, the physical properties of the near-surface, and the antecedent soil moisture in the near surface. We demonstrate a number of deficiencies when past models of the CUZ are applied to the problem of simulating preferential recharge, notably that the assumption of instantaneous equilibrium between fractures and matrix is not valid, particularly during extreme recharge events. In order to simulate preferential recharge, fractures and matrix must be modelled as separate but interacting domains. This was achieved using a dual continua model. The model was computationally demanding, but was able to reproduce observed behaviour, including apparently hysteretic soil moisture characteristic relationships in the near surface, and rapid preferential recharge fluxes in response to high intensity rainfall

events.

Key words: Chalk; Groundwater recharge; Richards equation; Unsaturated zone;

Preferential flow; Fractured porous material

* Corresponding author. E-mail address: andrew.ireson@imperial.ac.uk. Telephone: +44

(0)20 75946120

1 Nomenclature

2	a	matrix block half width, [m]
3	a_0	matrix block half width at the ground surface, [m]
4	a_∞	matrix block half width at depth, [m]
5	C	specific capacity, [m^{-1}]
6	D	event duration, [d]
7	EC	event characteristic, [$mm \cdot d^{-1/3}$]
8	f	(subscript/superscript) a state or flux in the fracture domain
9	I	event intensity, [mm/d]
10	K	hydraulic conductivity, [m/d]
11	K_a	K between matrix and fractures, [m/d]
12	K_s	saturated hydraulic conductivity, [m/d]
13	L	Conductivity exponent parameter, [-]
14	L_{rd}	depth above which 63 % of root density is located, [m]
15	m	(subscript/superscript) a state or flux in the matrix domain
16	Q	water flux, [m/d]
17	$Q_{m,0}$	matrix infiltration capacity, [m/d]
18	Q_T	infiltration of rainfall, [m/d]
19	r_d	root distribution function, [-]
20	r_s	Feddes root stress function, [-]
21	S_s	specific storage, [m^{-1}]
22	S_e	effective saturation, [-]
23	t	time, [d]
24	U	root water uptake, [d^{-1}]

25	V	event volume, [mm]
26	w_f	fracture domain volume fraction, [-]
27	$w_{f,0}$	fracture domain volume fraction at the ground surface, [-]
28	$w_{f,\infty}$	fracture domain volume fraction in the deep Chalk, [-]
29	z	depth below ground level, [m]
30	z_α	CUZ model shape parameter, [m^{-1}]
31	z_β	CUZ model shape parameter, [m]
32	β	matrix block geometry factor, [-]
33	γ_w	empirical coefficient for fracture-matrix exchange, [-]
34	Γ_w	Fracture-matrix exchange term, [d^{-1}]
35	θ	volumetric moisture content, [m^3/m^3]
36	θ_s	saturated water content, [m^3/m^3]
37	θ_r	residual water content, [-]
38	σ	Kosugi parameter, [m]
39	Ψ	matric potential, [m]
40	Ψ_0	Kosugi parameter, [m]
41	$\Psi_{1,\infty}$	modified Kosugi model parameter, [m]
42	$\Psi_{1,0}$	modified Kosugi model parameter, [m]
43	Ψ_2	modified Kosugi model parameter, [m]
44	Ψ_{an}	matric potential threshold for anaerobiosis , [m]
45	Ψ_d	matric potential below which plant water stress begins, [m]
46	Ψ_w	wilting point, expressed as a matric potential, [m]

47

48 **1 Introduction**

49 In unconfined aquifers, recharge is here defined as the time varying flux of water
50 which passes from the base of the unsaturated zone into the saturated zone, with
51 the water table marking the boundary between the two (Rushton, 1997; Scanlon
52 et al., 2002). Over sufficiently long periods of time, the recharge volume will equal
53 the volume of infiltrated rainfall minus evapotranspiration, termed here as effective
54 rainfall. On shorter time scales (sub-annual) it is harder to quantify recharge, due
55 to the attenuation of effective rainfall by storage in the unsaturated zone, which
56 becomes more significant for increasingly shorter time scales. For example, on a
57 daily time scale, the volume of recharge on a particular day is likely, especially
58 under non-extreme rainfall conditions, to bear no relation to the volume of rain
59 that fell that day. Therefore, as well as difficulties associated with the accurate spa-
60 tiotemporal measurement of rainfall and evapotranspiration which can affect the
61 total volume of recharge, quantifying the timing, of recharge also presents a sig-
62 nificant challenge. Furthermore, in fractured porous media, such as the Chalk, the
63 timing and volume of recharge in response to effective rainfall can be highly non-
64 linear due to the activation of fracture pathways (Lee et al., 2006; Ireson et al.,
65 2009a). Groundwater resource assessment, which is a time integrated function of
66 recharge, is generally less sensitive to recharge timing. Hence, in groundwater mod-
67 els used for this purpose, simple recharge models which are unable to resolve the
68 timing of recharge, may still be suitable as long as they predict the long term vol-
69 ume of recharge with some degree of accuracy (note, there is some doubt that these
70 model are able to do even this under drought conditions, discussed by Ireson et al.,
71 2009b). However, two examples of where recharge timing is important are contam-

72 inant transport (e.g. Brouyère, 2006; Jackson et al., 2007; Gooddy et al., 2007), and
73 groundwater flooding (reviewed by Hughes et al., 2010). Habets et al. (2010) com-
74 pared four different types of model for reproducing the groundwater flooding in the
75 Somme catchment in 2000/1. They found the models were able to reproduce the
76 spatial extent of flooding reasonably well. However, none of the models were able
77 to reproduce the piezometric heads during and after the flooding. They attributed
78 this to the overly simplistic representation of the unsaturated zone flow processes,
79 in particular, the changing depth of the unsaturated zone, and the activation of pref-
80 erential recharge through the fractures.

81 Understanding and quantification of recharge processes in Chalk aquifers in the
82 UK, focussing in particular on the relative roles of the fractures and the porous ma-
83 trix, have developed over 40 years (reviewed by Ireson et al., 2009b). Various con-
84 ceptual models for how water moves within and between the matrix and fractures in
85 the Chalk unsaturated zone have been proposed, with perhaps the most significant
86 contributions from Wellings and Bell (1980), Price et al. (2000) and Haria et al.
87 (2003). Recent work has focussed on the development of physically based models
88 (Mathias, 2005; Mathias et al., 2006) and combining these with field observations
89 (Brouyère, 2006; Van den Daele et al., 2007; Ireson et al., 2009b). In addition,
90 workers have tried to infer preferential recharge mechanisms from rainfall-water
91 table response data (Lee et al., 2006; Ireson et al., 2009a). Using interpretations
92 from field data, Ireson et al. (2009a) have suggested that the activation of the frac-
93 tures does not necessarily mean that there will be a rapid (<1 d) recharge response
94 (see their Figures 2 and 4). Partial wetting of the fractures can occur by the mecha-
95 nisms described by Price et al. (2000) and Haria et al. (2003), and result in recharge

196 responses which lag effective rainfall by tens of days (Ireson et al., 2009a). In this
197 paper we concentrate on preferential recharge, by which we mean flow through
198 the fractures which results in a rapid (≤ 1 d), highly non-linear recharge response,
199 and not simply fracture flow (which may or may not be rapid). We develop work
200 presented in two earlier papers, Ireson et al. (2009a) and Ireson et al. (2009b).
201 Extended data sets have become available from the instrumented field sites in the
202 Pang and Lambourn catchments, Berkshire, UK (described in Section 2), cover-
203 ing periods of extreme high and low rainfall conditions. We address limitations
204 in the previously developed CUZ model (Ireson et al., 2009b) in the context of
205 rapid preferential recharge under extreme high intensity rainfall. Empirical insights
206 into controls on preferential recharge in the Chalk (Section 3) are combined with
207 insights from an improved physically-based model (Section 4). In the discussion
208 (Section 5), we draw together the findings from these two approaches to provide
209 insights into controls on preferential recharge to Chalk aquifers.

209 **2 Field sites studied**

210 This study makes use of updated data sets from the Pang and Lambourn catchments,
211 (Berkshire, UK, Fig. 1) collected partly under the NERC LOCAR programme,
212 as well as additional instrumentation installed in the catchments, supplied by the
213 FLOOD1 project (run jointly by BRGM, Orleans, the BGS and the University of
214 Brighton, and partly funded by the EU INTERREG IIIA initiative). Previous stud-
215 ies (Ireson et al., 2006, 2009b) looked at recharge sites at Warren Farm (SU 3655
216 8092, depth to water table ≈ 40 m) and West Ilsley (SU 484 836 depth to wa-
217 ter table ≈ 70 m), located on the Seaford Chalk formation, with thin soils and
218

119 deep unsaturated zones. These sites were instrumented to measure water content,
120 θ [-], and matric potential, ψ [m], over a series of depths down to 4 m, with read-
121 ings logged every 15 minutes. Unfortunately the West Ilsley site was discontinued
122 beyond 2004. However, in 2005, under FLOOD1, deep jacking tensiometers and
123 piezometers were installed in a borehole at East Ilsley (SU 4996 8114, depth to
124 water table \approx 20 m), located lower down the catchment, in the Pang Valley. In
125 this paper we therefore focus on data from Warren Farm (WF) and East Ilsley (EI)
126 (Figure 1).

127 Figure 1.

128 A complete summary of instrumentation used in this study are given in Table 1.

129 Table 1.

130 **3 Insights into preferential recharge from field data**

131 Previously, Ireson et al. (2009a) used data obtained from the deep jacking tensiome-
132 ters and piezometer at East Ilsley, combined with sub-hourly tipping bucket rain-
133 gauge data, to gain insights into Chalk recharge processes. It was suggested that
134 three modes of recharge are active in the Chalk, under different effective rainfall
135 conditions, as summarised in Table 2. Figure 2 shows the cumulative effective rain-
136 fall versus water table elevation at East Ilsley. During the low rainfall conditions in
137 2005/6 (as shown by the relatively shallow slope in effective rainfall, ER) recharge
138 is via the matrix, with lags between peaks and troughs of > 100 days (Ireson et al.,
139 2009a). In the winter of 2006/7 the water table rises more markedly, in response to

140 around 500 mm of effective rainfall over about 6 months, and lags of the order of
141 tens of days were reported (Ireson et al., 2009a). On 20th July 2007, a large rainfall
142 event (about 90 mm in 12 hours) caused a rapid (within 13 hours) and significant
143 (> 1 m water table rise) response. Following this event, the water table responded
144 before the matric potential in the unsaturated zone immediately above the water
145 table, shown in Figure 3. Therefore, this was interpreted as a preferential recharge
146 event, with flow transmitted through the fractures, bypassing the matrix (Ireson
147 et al., 2009a). As well as the immediate response, the water table continued to rise
148 for one month following the event (we return to this observation in Section 4.3).
149 This led to a high antecedent water level at the beginning of the recharge period
150 for 2007/8. Sustained high rainfall during the summer of 2008 (CEH/Met Office,
151 2008) resulted in a continual rising trend in effective rainfall throughout this year,
152 but despite this the water table dropped fairly steadily from March to September.

153 Table 1.

154 Figure 2.

155 In this paper, we focus on preferential recharge responses, such as the 20th July
156 2007 event. By close visual inspection of the water table data, a number of prefer-
157 ential responses to rainfall were identified by Ireson et al. (2009a). Here, we extend
158 this analysis to cover a three year period up to September 2008. We define a rain-
159 fall event as a cluster of non-zero rainfall measurements (on an hourly time step)
160 containing no gaps of longer than 6 hours, as shown in Figure 3. 536 events were
161 identified in the three year record, with durations from 1 hour to > 2 days and
162 mean intensities from 1.2 to 180 mm/d. 18 of these rainfall events were followed
163 within 24 hours by a perturbation of the water table, as determined by visual in-

164 spection (referred to as “events perturbing the water table”). The duration, D , and
165 mean intensity, I , of each rainfall event is shown in Figure 4, and events perturbing
166 the water table are highlighted. As before, it was possible to partition the parameter
167 space in this plot into three regions: events for which preferential recharge is highly
168 unlikely to occur (region A); events for which preferential recharge may or may not
169 occur (region B); and events for which preferential recharge is highly likely to oc-
170 cur (region C). The boundary between regions B and C was subtly moved from the
171 previous location (Ireson et al., 2009a) to maximise the number of events perturb-
172 ing the water table in region C, but the gradient of this line was kept the same, i.e.
173 $-2/3$ on a log-log plot. In addition, following the updated analysis, one anomalous
174 point was found in region C (i.e. a point for which no water table response was ob-
175 served). Nonetheless, this appears to be a reasonably robust method for predicting
176 the onset of preferential recharge. It must be noted, however, that this is a site spe-
177 cific and subjective analysis, and it has not been demonstrated whether or not the
178 method can be applied elsewhere, or the results can be generalised. In particular,
179 this analysis has identified events that give rise to an observed water table response
180 at a depth of around 20 m at this EI site. It is certainly possible that events that did
181 not produce a water table response within a day at this site could have caused rapid
182 preferential flow at smaller unsaturated depths, that was subsequently attenuated by
183 storage in the fracture domain. Likewise, events that did cause a response at 20 m
184 depth might not have caused a response at greater unsaturated depths.

185 Figure 3.

186 Figure 4.

187 The information in the left hand plot in Figure 4 can be presented as a single para-

188 metric measure, which we call the “event characteristic”, EC . Every point on this
189 log-log axis plot is translated along a path of gradient $-2/3$ onto the intercept (i.e.
190 where the event duration is $10^0 = 1$ day), and then the exponential is taken, to give
191 the EC , that is,

$$\log(I) = -2/3.\log(D) + \log(EC)$$

192 and hence

$$193 \quad EC = I.D^{2/3} = V.D^{-1/3} \quad (1)$$

194 where V is the event volume ($I \times D$). It can thus be seen that the event characteris-
195 tic is a non-linear combination of intensity and duration (or volume and duration),
196 with units of $\text{mm.d}^{-1/3}$. EC for each event is plotted on the right hand side in Fig-
197 ure 4. Note, the event on 20th July 2007, which caused by far the largest water
198 table response, has the largest EC , by a factor of about 2. These results suggest
199 that for East Ilsley an EC of greater than $26.3 \text{ mm.d}^{-1/3}$ will cause a preferential
200 recharge response and an EC of less than $8.3 \text{ mm.d}^{-1/3}$ will not, irrespective of
201 any other factors. For events within region B, that is with an EC between 8.3 and
202 $26.3 \text{ mm.d}^{-1/3}$, some other explanatory variable is required to predict whether or
203 not preferential recharge will occur. In Figure 5, observed soil moisture storage
204 in the top 60 cm of the unsaturated zone is plotted, with region B and C rainfall
205 events highlighted. Unfortunately, soil moisture measurements were only available
206 up to March 2007. It can be seen that there is no apparent relationship between soil
207 moisture storage and occurrence of region C events. However, region B events only
208 occur when the soil moisture storage is high, suggesting that under these rainfall

209 conditions, the onset of preferential recharge also depends on the wetness of the
210 soil/Chalk. We also looked at the antecedent water table depth, but no such rela-
211 tionship with region B events was apparent. We therefore suggest that the *EC* and
212 antecedent soil moisture can be used to determine the onset of preferential recharge,
213 but it is highly likely that the thresholds in each will be site specific.

214 Figure 5.

215 **4 Modelling of recharge in the Chalk**

216 *4.1 Performance of the previous CUZ model*

217 Previously, a physically-based model for the CUZ (Ireson, 2008; Ireson et al.,
218 2009b) was developed, which treated the matrix and fractures as an equivalent
219 continuum (i.e. assuming instantaneous exchange between the domains), as first
220 proposed by Peters and Klavetter (1988), and consistent with other flow models for
221 the Chalk (Chapter 4 of Mathias, 2005; Brouyère, 2006). Hereafter, we refer to this
222 model as the ECM model. This model included a novel means of representing near
223 surface progressive weathering of the Chalk by relating hydraulic properties to pore
224 size distributions (after Kosugi, 1996) and matrix/fracture domain fractions both
225 of which evolve with depth. The model was successfully applied to reproduce near
226 surface measurements of water content and matric potential at Warren Farm, cali-
227 brated using data from 2004 and validated using data from 2005. Since this work
228 was published, more data have become available, both from the instruments used
229 in that study at the Warren Farm recharge site and from additional instruments in-

230 stalled in the Pang catchment. The previous period studied (2004/5) was a period
231 of significant drought (Ireson et al., 2009a), whereas subsequent years were signif-
232 icantly wetter, and include an extreme high intensity rainfall event in the summer
233 of 2007 (described above in Section 3).

234 The model was re-applied to an extended data set (covering 2004-7) at the same
235 site, with no modifications to the parameters or model structure. The model was
236 driven with rainfall data from a tipping rain gauge, and two separate estimates of
237 evapotranspiration, all of which were measured at Warren Farm (approximately
238 300 m from the soil moisture instrumentation). As in the previous study, hourly ac-
239 tual evapotranspiration, AE, measured by eddy flux correlation was available (de-
240 scribed in Ireson et al., 2009b). This direct measurement accounts for the effects
241 of atmospheric demand, plant resistances (notably aerodynamic canopy resistance
242 and stomatal resistance) and soil water stress on evapotranspiration. Also available
243 from the automatic weather station were hourly meteorological variables (atmo-
244 spheric pressure, humidity, temperature, net short and net long wave radiation) and
245 soil heat flux, required to calculate potential evapotranspiration, PE, for a reference
246 grass crop (Allen et al., 1994). Land use at the site was grass throughout this period.
247 These two local estimates of evapotranspiration differed somewhat in certain peri-
248 ods, most notably 2004/5, which allows us to explore the impact of uncertainty in
249 the driving data. In both cases, following Ireson et al. (2009b), a modified version
250 of the Feddes et al. (1976) model for distributing root water uptake was applied,
251 such that the effect of soil water stress is to re-distribute root water uptake, but not
252 reduce it. The locations of these instruments are given in Table 1.

253 The results of the updated model runs are shown in Fig. 6 for matric potential and

254 water content change at 1 m depth, where the full range of conditions in each was
255 measured comprehensively (Ireson et al., 2006). In addition, the figure shows sim-
256 ulated matrix and fracture fluxes at 15 m below ground level, assuming these to
257 be indicative of the recharge to a water table just below this depth, such as at East
258 Ilsley. For reference, the occurrences of observed water table response attributed to
259 rapid preferential flow events at East Ilsley are also shown. For the period 2004/5,
260 the model driven with AE is identical to the model in Ireson et al. (2009b), and re-
261 produces water content and matric potential well during this period. Subsequently,
262 however, this model fails to reproduce the wetting up in early 2006, and hence
263 continues to underestimate the soil moisture state until early 2007. Attempts to
264 recalibrate the model for the 2006 period all failed - no parameter set was found
265 which could encompass the observations during this period, suggesting that either
266 the driving data or the model structure were erroneous. Driving rainfall data were
267 found to be consistent with surrounding gauges and Met Office radar data (NIM-
268 ROD), and using alternative gauges was found to have only a minor effect on the
269 model output. However, the driving evapotranspiration data were found to have a
270 significant impact, and by using calculated PE, a quite different response was ob-
271 tained. The PE driven model tended to perform better at reproducing the change
272 in water content over the entire period where observations were available (2004-
273 2007), but the matric potential was now overestimated in 2004/5. We are unable to
274 comment upon which evapotranspiration data set is more accurate, but the finding
275 that the model is highly sensitive to these differences is important. Evapotranspi-
276 ration is in general difficult to measure, and even more difficult to validate, and,
277 especially in areas where evapotranspiration makes up a large portion of the water
278 balance, uncertainties in evapotranspiration associated both with climatic variables

279 (e.g. Chun et al., 2009) and vegetation characteristics (Beven, 1979), will have a
280 significant impact on hydrological predictions. In the context of using inverse mod-
281 elling to identify hydraulic properties, as in this study, these uncertainties alone
282 mean that meaningful identification of an optimal parameter set is not possible (we
283 return to this issue below).

284 There was also a significant difference in the simulated recharge fluxes with each
285 model. For the PE driven model, unlike the AE driven model, recharge through
286 the fractures was simulated in the winters of 2004/5, 2006/7 and 2007/8. Neither
287 model simulated fracture flow when we expected it to occur during the summer of
288 2007, following an extreme high intensity rainfall event, but both simulate fracture
289 flow during the summer of 2008 (when sustained rainfall totals were high). In both
290 cases, whenever fracture recharge is simulated, it persists for months. Therefore,
291 the ECM is not able to reproduce the discrete preferential recharge responses to
292 high intensity rainfall that we expect to occur on the basis of the previous analysis.

293 Figure 6.

294 At 1.0 m depth, the pressure transducer tensiometer and equitensiometer provide
295 a continuous record of matric potential over the entire range of field conditions.
296 Combined with the profile probe data, it is possible to investigate the soil moisture
297 characteristic (SMC) relationship at this depth, as shown in Figure 7. The elongated
298 and inverted s-shape of this curve demonstrates the role of the fractures (wetting up
299 at high matric potentials) and matrix (remaining saturated down to around -15 m,
300 and draining at matric potentials below this). A notable feature of the SMC is that
301 it appears to exhibit significant hysteresis. The SMC representation in the ECM
302 model fits the primary drying curve of this observed data. To demonstrate further

303 how significant this hysteresis was likely to be, quantile mapping (Hashino et al.,
304 2007) was used to generate a time series of water content from matric potentials.
305 The result in Figure 8 shows inconsistencies between the two data sets, most no-
306 tably in the summer of 2004 and winter of 2006. Hence any model assuming a
307 constant relationship between θ and ψ , such as in the ECM model shown in Fig-
308 ure 7, would be unable to reconcile the observations in these two periods. This is
309 also when the two models driven with different evapotranspiration data performed
310 differently.

311 Figure 7.

312 Figure 8.

313 We therefore conclude that the structure of the ECM model is inadequate, especially
314 in the context of predicting preferential responses. In the following sections the
315 development and assessment of an improved model is described.

316 4.2 *Development of an improved CUZ model*

317 The classic cause of hysteresis in wetting and drying soils is the “ink-bottle” ef-
318 fect (Hillel, 1998). Modelling flow in hysteretic single porous media is extremely
319 challenging, and whilst various methods have been proposed (e.g. Mualem, 1974;
320 Pham et al., 2005), the authors are unaware of any models having been successfully
321 applied to reproduce field observations. This is probably because of the challenge
322 of parameterising the hysteretic relationships, in particular the $K(\psi)$ relationship,
323 which cannot be observed. However, the Chalk is not a single porous media, and we
324 postulate an alternative cause of the apparent hysteresis, which is that it is caused

325 by: *i*) pressure disequilibrium between the fracture and the matrix domain, espe-
326 cially following infiltration from a rainfall event which would wet the fractures
327 before the matrix; and *ii*) the fact that the instruments for measuring water con-
328 tent and matric potential sample different volumes of rock. Both neutron probes
329 and profile probes take an integrated reading of water content over some volume of
330 rock, with a minimum radius of about 0.1 m. It is therefore likely that these read-
331 ings are representative of the bulk fracture-matrix water content (especially in the
332 weathered zone), as discussed by Ireson et al. (2006). Tensiometers and equiten-
333 siometers, on the other hand, sample at a point scale (or more accurately, over the
334 contact area between the instrument tip and the rock). The tensiometer tip is likely
335 to be located in a fracture (either natural or caused by the installation), but also in
336 contact with the face of the matrix block. Therefore, we might expect a tensiometer
337 to respond to a rapid increase in pressure in the fracture domain, but not to dry out
338 below the pressure in the matrix. Thus, when the Chalk is dry, following a rainfall
339 event the “fractures” (which in the near surface are enhanced by weathering) may
340 wet up, causing a small increase in the bulk water content, but a large increase in
341 the fracture pressure, thereby giving rise to the apparent scanning curves present in
342 Fig. 7.

343 To model this effect, it is necessary to relax the assumption of instantaneous equi-
344 librium between the fracture and matrix domains, central to the ECM approach.
345 The simplest way to do this is to use a Dual Continua Modelling, DCM, approach
346 (Doughty, 1999). In DCM models flows in the fracture and matrix domain are mod-
347 elled separately, and exchange between these domains is governed by a first order
348 transfer function. This adds at least one additional model parameter, but the major

349 cost is that the numerical model is significantly more computationally expensive
 350 (having effectively doubled the number of nodes). However, the benefit of such
 351 a model is that, unlike the ECM, it allows us to simulate preferential flow in the
 352 fractures, which bypasses the matrix.

353 We adopt the DCM structure proposed by Gerke and van Genuchten (1993a), (GVG
 354 hereafter). In this model the dependent variables are the matrix potential in the
 355 fractures, ψ_f , and matrix, ψ_m , and flow in each domain, governed by Richards'
 356 equation, is modelled separately

$$357 \quad (1 - w_f)(S_{e,m}S_{s,m} + C_m) \frac{\partial \psi_m}{\partial t} = \frac{\partial}{\partial z} \left((1 - w_f)K_m \left[\frac{\partial \psi_m}{\partial z} - 1 \right] \right) + \Gamma_w - U_m \quad (2)$$

$$358 \quad w_f(S_{e,f}S_{s,f} + C_f) \frac{\partial \psi_f}{\partial t} = \frac{\partial}{\partial z} \left(w_f K_f \left[\frac{\partial \psi_f}{\partial z} - 1 \right] \right) - \Gamma_w - U_f \quad (3)$$

359 Note here that as in Ireson et al. (2009b), specific capacity, C , and hydraulic con-
 360 ductivity, K , for each domain are a function of depth (using the relationships given
 361 in the Appendix), to account for changes in properties in the soil and weathered
 362 Chalk layers. The local exchange of water between the domains is governed by

$$363 \quad \Gamma_w = \frac{\beta \gamma_w K_a}{a^2} (\psi_f - \psi_m) \quad (4)$$

364 where $\beta = 3$ for rectangular matrix blocks, γ_w was empirically determined to be
 365 0.4 (GVG), and a (the matrix half block width) and K_a (the hydraulic conductiv-
 366 ity governing exchange between the fractures and matrix) are to be determined.
 367 K_a was defined as a function of matrix potential using the same relationship as K_m
 368 (see Appendix), but with a modified saturated hydraulic conductivity, K_s^a , to be
 369 determined. We would expect the matrix block size to be smaller in the shallow,
 370 weathered Chalk than in the deep consolidated Chalk (see Figure 1 in Ireson et al.,

371 2009b). The progressive weathering of the Chalk was characterised in the previous
 372 study using the relationships given in the Appendix that scale the pore size distri-
 373 bution of the fracture domain, and the domain fractions of the fractures and matrix,
 374 as a function of depth. In the DCM model the same scaling relationship is applied
 375 to the matrix block half width, a :

$$376 \quad a = a_\infty + \frac{a_0 - a_\infty}{1 + \exp(z_\alpha(z - z_\beta))} \quad (5)$$

377 where a_0 is the matrix block half width at the ground surface, and a_∞ is the matrix
 378 half width at depth. Hence, according to Equation 4, the rate of exchange between
 379 the fractures and the matrix would be larger in the near surface than at depth.

380 Infiltration of precipitation forms the upper boundary condition, which is prescribed
 381 as a flux into the fractures, $Q_{T,f}$, and matrix, $Q_{T,m}$. As in the GVG model, infiltra-
 382 tion is assumed to occur into the Chalk matrix until its infiltration capacity, $Q_{m,0}$, is
 383 exceeded. The infiltration capacity is determined using Darcy's law, where the hy-
 384 draulic conductivity, K_m^* , and hydraulic gradient are found assuming that the matric
 385 potential in the matrix at the soil surface, $\Psi_{m,0}^*$, is zero. Hence

$$386 \quad Q_{m,0} = (1 - w_f)K_m^* \left(\frac{\Psi_{m,1} - \Psi_{m,0}^*}{\Delta z} - 1 \right) \quad (6)$$

387 where subscripts 0 and 1 refer to the soil surface and first node below the soil sur-
 388 face, respectively. This is equivalent to allowing infiltration to bring the matrix up to
 389 the point of saturation, but not beyond it to cause ponding. If the infiltration capac-
 390 ity is exceeded, the excess infiltrates into the fractures, whose infiltration capacity
 391 is sufficiently high that ponding or overland flow cannot occur. This is a reasonable

392 assumption, as overland flow has not been observed at the WF field site. Therefore

$$393 \quad Q_{T,m} = \begin{cases} P, & P \leq Q_{m,0} \\ Q_{m,0}, & P > Q_{m,0} \end{cases} \quad (7)$$

$$394 \quad Q_{T,f} = \begin{cases} 0, & P \leq Q_{m,0} \\ P - Q_{m,0}, & P > Q_{m,0} \end{cases} \quad (8)$$

395 The root uptake model, which distributes uptake over depth according to the soil
 396 moisture stress (based on Feddes et al., 1976), was adapted to additionally distribute
 397 root water uptake between the fracture and matrix domains. Plant root uptake was
 398 therefore distributed over depth and between domains, as a function of root density,
 399 r_d (a function of depth), soil-water stress r_s (a function of soil wetness) and the
 400 domain fraction, w_f (also a function of depth).

$$401 \quad U_m = \frac{r_s(\Psi_m)r_d(z)(1-w_f)}{\int_0^L r_s(\Psi_m)r_d(z)(1-w_f(z))dz + \int_0^L r_s(\Psi_f)r_d(z)w_f(z)dz}AE \quad (9)$$

$$402 \quad U_f = \frac{r_s(\Psi_f)r_d(z)w_f}{\int_0^L r_s(\Psi_m)r_d(z)(1-w_f(z))dz + \int_0^L r_s(\Psi_f)r_d(z)w_f(z)dz}AE \quad (10)$$

403 where

$$404 \quad r_s(\Psi) = \begin{cases} 0, & \Psi > \Psi_{an} \\ 1, & \Psi_{an} \geq \Psi > \Psi_d \\ 1 - \frac{\Psi - \Psi_d}{\Psi_w - \Psi_d}, & \Psi_d \geq \Psi > \Psi_w \\ 0, & \Psi_w \geq \Psi \end{cases} \quad (11)$$

405 and

$$406 \quad r_d(z) = \frac{\exp(-z/L_{rd})}{L_{rd}} \quad (12)$$

407 Ψ_{an} , Ψ_d and Ψ_w are water stress thresholds, assumed to have values of -0.5 m,
408 -4 m and -150 m respectively (Feddes et al., 1976). L_{rd} , the depth above which
409 approximately two-thirds of plant root density is located, was kept at 0.2 m.

410 As before, a fixed head water table at 40 m depth was used for the lower boundary
411 condition of the model. This was, again, because actual water table fluctuations can-
412 not be reproduced with a 1D model, and imposing an observed water table response
413 on the system would be expected to bias deep simulated fluxes. We previously (Ire-
414 son, 2008; Ireson et al., 2009b) demonstrated that this approach is reasonable if the
415 boundary is sufficiently deep compared with the depths where we are interested in
416 reproducing the observed states and fluxes (in this case states in the top 1 m and
417 fluxes at 15 m depth). 15 m depth was chosen because between September 2005
418 and September 2008 the water table at East Ilsley fluctuates within the range of \approx
419 15 to 28 m BGL.

420 The coupled system of equations is solved numerically in MATLAB using the
421 method of lines. Standard finite difference approximations are used to assess the
422 spatial derivatives, using a node centred grid. The hydraulic conductivity is esti-
423 mated at block boundaries using the arithmetic mean (Parissopoulos and Wheater,
424 2006). The temporal derivatives are integrated using the MATLAB ordinary dif-
425 ferential equation solver ode15s (Shampine and Reichelt, 1997). This employs an
426 adaptive time grid to minimise numerical errors, and boundary conditions are ap-
427 plied on an hourly time step.

428 In summary, the new model includes one additional state variable, required at every
429 node and an additional 3 parameters governing the exchange of water between the
430 two domains, namely a_0 , a_∞ and K_s^a .

431 4.3 Performance of the improved CUZ model

432 Initially, parameters from the original model were kept, physically realistic values
433 of a_0 and a_∞ were adopted and, following Gerke and van Genuchten (1993b) K_s^a
434 was set to $K_s^m/100$. However, we found it was necessary to modify some parameters
435 in order to achieve good model performance. Since the model was computationally
436 more demanding than the ECM model (taking about 30 minutes to run a 5 year sim-
437 ulation on an hourly time step with 50 nodes on an Intel X9650 3 GHz processor),
438 the number of parameters had increased, and the uncertainty in driving data could
439 not be resolved, it was not feasible to optimise the model parameters. Rather, we
440 focused on obtaining a “refined” parameter set to demonstrate the potential of the
441 model to reproduce observed system behaviour. This was achieved through man-
442 ual calibration. In doing this, we used calculated PE data to drive the model, since

443 this gave a better model fit to the observed change in water content using the ECM
444 model (Figure 6). We also performed a model sensitivity study, described later,
445 which looked at both parametric sensitivity and the effect of using the observed AE
446 to run the model. The ECM model parameters from Ireson et al. (2009b), and the
447 refined DCM model parameters, are shown in Table 3.

448 Table 3

449 *Near surface changes in soil moisture state*

450 The refined model did a reasonable job of reproducing the observed change in
451 water content throughout the top 1.0 m, as shown in Figure 9. At 1.0 m depth,
452 the performance was subtly better than the ECM model driven with PE (the RMSE
453 was 0.0143 as compared with 0.0146). If the observed soil moisture characteristic
454 curve is interpreted as bulk water content against fracture matrix potential then
455 consistent behaviour is reproduced by the simulation, as shown in Figure 10. The
456 precise form of this simulated hysteretic relationship was sensitive to parametric
457 changes. We did not attempt to actually fit the observed scanning curves (since
458 this was not the central focus of this study, but would merit further study). The
459 simulated behaviour appears to support our hypothesis that the matric potential
460 measured by tensiometers is not representative of the bulk fracture-matrix system.
461 This also implies that the assumption that the fractures and matrix will always be
462 in pressure equilibrium (inherent in all previous ECM modelling approaches) is not
463 strictly valid.

464 Figure 9

465 Figure 10

466 *Preferential recharge events*

467 The simulated recharge fluxes (that is, fluxes at 15 m depth in the profile) are shown
468 in Figure 11. Overall, matrix recharge dominates, but there are four discrete preferential fracture recharge events in the period shown. Three of these coincide with
469 observed water table responses at East Ilsley, which are also indicated in Figure
470 11. The largest observed water table response in July 2007, which had the largest
471 *EC*, was also simulated as the largest preferential recharge response, with a peak
472 intensity of almost 100 mm/d. The event with the second largest *EC* was on the
473 27th May 2007, but no preferential recharge was simulated on this date. Further-
474 more, preferential recharge was simulated on 19th January 2007 in response to an
475 event with an *EC* of 16, when no response in the water table at East Ilsley was ob-
476 served. This is a region B event characteristic, i.e. one that we would expect to give
477 a response only if the antecedent soil moisture was wet. These limitations are not
478 surprising given that a model conditioned on data at one field site (Warren Farm) is
479 being applied to try to reproduce observed responses at another (East Ilsley), that
480 no rigorous model calibration was possible, and that, as already discussed, there
481 are significant uncertainties in the driving data.
482

483 Figure 11

484 *Consistency of recharge-water table response*

485 The recharge fluxes transmitted through the matrix (Figure 11) follow a pattern
486 which appears reasonably consistent with the water table response at East Ilsley
487 (Figure 2), and is certainly an improvement over the ECM fluxes (Figure 6). The
488 recharge fluxes slightly lag the water table response, but this is not necessarily an

489 inconsistency. The water table response may be caused by the lateral propagation
490 through the saturated zone of recharge reaching the water table earlier in the valleys
491 where the unsaturated zone is thinner. Likewise, the continual rise of the water
492 table at East Ilsley following the 20th July 2007 preferential recharge event might
493 be caused by the delayed impacts of this recharge event reaching the water table
494 under the interfluves (where the unsaturated zone is thicker) later, again propagated
495 laterally through the saturated zone. This is speculation at this stage, and to make
496 further insights it will be necessary to perform 2D or 3D modelling of the couple
497 unsaturated/saturated zone.

498 In summary, the DCM modelling approach, whilst computationally demanding and
499 hard to calibrate, is able to simulate the observed behaviour of the CUZ: specifically
500 the near surface changes in water content, apparently hysteretic near surface soil
501 moisture characteristic curves, and deep preferential recharge fluxes consistent with
502 the types of water table responses that have been observed.

503 *4.4 Model sensitivity*

504 The DCM model has 22 hydraulic parameters, all of which have some physical
505 meaning, and can therefore be placed into three categories: Conductivity parame-
506 ters; Storage parameters; and Exchange parameters (governing the exchange of wa-
507 ter between the fracture and matrix domains). A sensitivity study was performed,
508 considering each of these three separately. An important finding during the man-
509 ual calibration exercise was that the exchange parameters and the fracture storage
510 parameters had a relatively small impact on the near surface changes in water con-
511 tent, but a highly significant impact on the deep fluxes. In this sensitivity study,

512 where the focus is on controls on preferential recharge, we concentrate on the sen-
513 sitivity of the deep fluxes. The previously refined model driven with PE is taken as
514 the benchmark (parameters are given in Table 3), and the impact of modifications
515 to certain parameters, or groups of parameters, on the relative amount of bypass
516 recharge are summarised in Table 4, and discussed below.

517 Table 4.

518 *Conductivity parameters*

519 As for the ECM model, in our judgement the most sensitive parameter in the DCM
520 model was the matrix saturated hydraulic conductivity, K_s^m . The effect of changes
521 in this parameter on fracture and matrix fluxes is shown in Figure 12. As K_s^m in-
522 creases then a larger proportion of the infiltrating flux can be transmitted through
523 the matrix, hence the matrix flux increases, and the fracture fluxes decrease in both
524 magnitude and occurrence. Moreover, in the DCM model K_s^m plays a key role in
525 the partitioning of infiltration between the matrix and the fractures at the surface
526 (see Equation 6). Particularly if the rate of exchange between the fractures and ma-
527 trix is small, this might be the dominant control in the model on deep preferential
528 recharge. These findings are reflected in Table 4, showing that when K_s^m was dou-
529 bled to 2 mm/d, no preferential recharge was simulated, whilst when it was halved
530 to 0.5 mm/d the volume of preferential recharge increased significantly.

531 Figure 12

532 The previous study found that the fracture saturated hydraulic conductivity, K_s^f ,
533 was only moderately sensitive over 3 orders of magnitude. In this study, the re-
534 fined value for K_s^f was increased to give a bulk saturated hydraulic conductiv-

535 ity ($w_f K_s^f$) consistent with values of Chalk saturated hydraulic conductivity (e.g.
536 Williams et al., 2006). A value of 27,000 m/d, combined with an increase in the
537 Mualem conductivity exponent parameter, L to 14.3, provided a reasonable perfor-
538 mance, which for a fracture porosity of 0.1 % (discussed below) is equivalent to a
539 bulk Chalk saturated hydraulic conductivity of 27 m/d. The impact of increasing or
540 decreasing K_s^f by one order of magnitude was to increase or decrease, respectively,
541 the volume of bypass flow, without significantly affecting the timing/onset. This
542 would therefore be an important parameter in a coupled unsaturated zone/saturated
543 zone model.

544 *Storage parameters*

545 Storage in the dual permeability, vertically heterogeneous, partially saturated soil/weathered
546 Chalk/consolidated Chalk, is complex, being described by 13 parameters. These de-
547 termine the volume of saturated storage and the rate at which storage reduces with
548 reducing pore water pressure in each domain, and how these change with depth.
549 Rather than performing a univariate sensitivity study, we looked separately at sen-
550 sitivity to dynamic storage in the near surface and sensitivity to storage in the deep
551 fractures. Storage in the matrix is less dynamic, and well characterised by the ob-
552 served soil moisture characteristic data.

553 To explore sensitivity to changes in the soil at the surface, as well as the benchmark
554 model (soil a), we considered four alternative soil/weathered Chalk layer configu-
555 rations, denoted as soils b), c), d) and e), depicted in Figure 13. Each was achieved
556 by parametric modifications shown in Table 5. Soils a), b) and c) all have the same
557 volume of dynamic storage (that is, storage associated with the fracture/soil do-

558 main), but this storage is distributed differently over depth. For a high porosity,
559 shallow soil (b) and a low porosity deep soil (c) no bypass recharge was generated,
560 in both cases because more water was able to pass from the fractures into the ma-
561 trix. In soil b) this was because of an increased gradient between the domains as the
562 shallow soil filled with water following infiltration. In soil c this was because of an
563 increased depth over which exchange between the domains was possible. It should
564 be noted that where the soil porosity is high, the matrix half block width is low,
565 and hence exchange between the fracture and matrix domains is higher. More pre-
566 dictably, when the volume of dynamic near surface storage is reduced (d) there is
567 less attenuation of infiltration in the near surface, and the incidence and magnitude
568 of preferential recharge increases, and vice versa when the volume is increased (e).
569 This demonstrates that there is a high sensitivity to both the volume and distribution
570 of near surface storage. In fact, the profile used in the benchmark model was not
571 modified from the previous ECM model, and was thus based on fitting the scaled
572 Kosugi (1996) model for $\theta(\psi)$ to observed drying curves at 0.2, 0.4, 0.6 and 1.0 m
573 depth, as described in Ireson et al. (2009b).

574 Table 5.

575 Figure 13.

576 The second most sensitive parameter in these experiments (after K_s^m) was the poros-
577 ity of the deep fractures, $w_{f,\infty}$. A typical value from the literature (Price et al., 1993;
578 Mathias et al., 2006) is 1%, which was used in the previous study. This was found
579 to be too large to generate deep preferential recharge (Table 4). Significantly im-
580 proved results were obtained by reducing this by an order to magnitude to 0.1%, as
581 used in the benchmark model. Reducing this further to 0.01 % led to an increase in

582 bypass recharge, but might be harder to justify physically. The fact that such a low
583 value was required in the model may in fact reflect the fact that not all of the deep
584 fractures are activated (so called flow focussing, Bodvarsson et al., 2003). For a
585 rock with an actual fracture porosity of 1 %, if only one in ten of the deep fractures
586 is actually connected to the active infiltration pathways, this would be equivalent
587 to an effective fracture porosity of 0.1 % in a continuum representation of the
588 system.

589 *Exchange parameters*

590 The exchange between fracture and matrix domains is governed by the head gradi-
591 ent between them, and a coefficient given by $\beta\gamma_w K_a(\psi)/a(z)^2$ (Equation 4). Here
592 we explore how variations in this bulk coefficient affect the model by changing K_s^a
593 which, unlike the other parameters, might vary over orders of magnitude (hence
594 uncertainties in the empirical β and γ_w parameters are negligible). We also explore
595 how variations in the depth distribution of this bulk coefficient affect the model by
596 changing a_0 and a_∞ .

597 When the coefficient is increased by an order of magnitude (by setting $K_s^a = K_s^m/10$,
598 see Table 4) there is more exchange between the domains, meaning that infiltra-
599 tion in the fracture domain is able to pass into the matrix domain, and preferential
600 recharge is reduced. Likewise, if the coefficient is reduced by an order of magni-
601 tude (by setting $K_s^a = K_s^m/1000$, see Table 4), less exchange between the domains
602 led to an increase in preferential recharge. Therefore, K_s^a is both a highly sensitive
603 and highly uncertain parameter. Good results were obtained using $K_s^a = K_s^m/100$, as
604 used by Gerke and van Genuchten (1993b) in their experiments. However, but this
605 was purely empirical and they noted that K_a is a critical parameter, for which little

606 is known about the physical or chemical properties. Therefore, accurately charac-
607 terising K_a for the Chalk is an outstanding and daunting challenge.

608 Changing the rate of exchange in the near surface, by modifying a_0 (see Table 4),
609 has a relatively small impact on simulated preferential recharge. Changing the rate
610 of exchange in the consolidated Chalk, by modifying a_∞ (see Table 4), has a more
611 significant impact, especially on the more moderate preferential recharge responses
612 (i.e. not 20th July 2007), since this affects the exchange over a much larger depth.
613 However, in general the sensitivity to modifications to the matrix block size, within
614 physically realistic bounds, is small compared with changes to the bulk exchange
615 coefficient associated with uncertainties in the K_a parameter.

616 *Driving data*

617 As can be seen in Table 4 (benchmark versus AE model run) the impact of uncer-
618 tainty in evapotranspiration driving data has a significant impact on the total pro-
619 portion of preferential recharge, but a negligible impact on the recharge response to
620 the extreme event on 20th July 2007. This is consistent with the finding in Section
621 3 that the response to this rainfall event was independent of antecedent soil mois-
622 ture which would have been affected by differences in evapotranspiration. This is
623 also consistent with the findings from the ECM model, and again highlights the
624 importance of uncertainty in driving data when trying to model field observations.

625 **5 Discussion**

626 In this paper we present two complementary approaches to assessing controls on
627 preferential recharge to Chalk aquifers. In the first approach (Section 3), inferences
628 are drawn from observations of rapid water table responses which coincide with
629 particular rainfall events. We propose that a measure of the magnitude of the rain-
630 fall event, the event characteristic, is a good predictor of when large preferential
631 recharge responses might occur. To predict responses to more moderate rainfall
632 events, it is also necessary to take the antecedent soil moisture into account. Due
633 to its simplicity, this is an attractive method for partitioning recharge, and could
634 easily be implemented in any soil water balance based recharge model (e.g. Pen-
635 man Grindley, Catchmod, QR Heathcote et al., 2004). However, this method has
636 not yet been demonstrated for other sites, and it is likely that the thresholds asso-
637 ciated with the onset of preferential recharge, in both the event characteristic and
638 antecedent soil moisture, will be site specific.

639 This analysis requires hourly rainfall data in order to be able to characterise ef-
640 fectively the event characteristic. This demonstrates that the system is sensitive to
641 sub-daily rainfall, which needs to be accounted for irrespective of the modelling
642 approach adopted. Hourly rainfall observations from tipping bucket rain gauges
643 are widely available in the UK. For assessing future climate impacts on recharge,
644 projections of downscaled daily rainfall are widely available, but hourly data less
645 so. For example UKCP09 does provide hourly rainfall generated using a weather
646 generator, but this is based on downscaled daily data which has been temporally dis-
647 aggregated (Jones et al., 2009) and it is unclear whether this has been adequately

648 validated for the types of sub-daily extreme rainfall that are important for gener-
649 ating preferential recharge. Therefore, quantifying the future impacts of climate
650 change on preferential recharge still presents a significant challenge.

651 The second approach in this paper (Section 4) focussed on the use of physically
652 based models of the Chalk unsaturated zone to predict preferential recharge. A
653 number of limitations in existing models developed for the CUZ were apparent.
654 Most significant was the finding that an equivalent continuum representation of
655 the matrix and fractures, which assumes instantaneous exchange of water between
656 these two domains, is unsuitable for predicting deep fracture flow responses. The
657 dual continua approach of Gerke and van Genuchten (1993a) appears better suited.
658 This model can reproduce observed soil moisture states and apparently hysteretic
659 soil moisture characteristics in the near surface, as well as the occurrence of prefer-
660 ential recharge responses at depth. The advantage of using such a physically based
661 model over a simple recharge model, is that all of the parameters have a physical
662 meaning, and whilst it is hard (perhaps not currently possible) to optimise these pa-
663 rameters, the impact of individual parametric modifications has a predictable effect
664 on the model performance (as demonstrated in Section 4.4). As such, it is possible,
665 using manual methods, to tailor the model to match observed system behaviour.
666 For example, this could be useful in a situation where water table observations are
667 available for two different boreholes with different unsaturated depths, where only
668 the shallower one responds to a particular rainfall event. In this case, the K_s^f and/or
669 K_s^a parameters could, in principle, be adjusted such that the fracture response is
670 propagated as far as the first water table, but not as far as the second.

671 One inherent limitation with the one-dimensional model is that the water table re-

672 sponse cannot be reproduced, since it also depends on lateral flow processes within
673 the saturated zone. In Section 4.3, we speculate as to how the simulated recharge
674 signal might be consistent with the water table response at East Ilsley, as a result
675 of the earlier and later impacts of the recharge signal down slope and up slope, re-
676 spectively. This appears a coherent interpretation, but to make further insights, 2 or
677 3 dimensional, coupled saturated-unsaturated flow modelling is required.

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793 **Appendix: Hydraulic properties**

794 The hydraulic properties ($K(\psi, z)$ and $C(\psi, z)$) for the matrix and fracture domains,
 795 and the $K(\psi, z)$ relationship for the exchange coefficient (K_a in Equation 4) were
 796 described by the modified Kosugi (1996) relationship, as given in Ireson et al.
 797 (2009b). For completeness, these relationships are included here, but a more thor-
 798 ough description is provided in Ireson et al. (2009b) and Ireson (2008). For defini-
 799 tions of symbols refer to the notation.

$$800 \quad K = K_s S_e^L \left[0.5 + 0.5 \operatorname{erf} \left(-\frac{\ln(\Psi/\Psi_0)}{\sigma\sqrt{2}} \right) \right] \quad (13)$$

$$801 \quad C = \frac{\theta_s - \theta_r}{(2\pi)^{1/2} \sigma(-\Psi)} \exp \left(-\frac{[\ln(\Psi/\Psi_0) - \sigma]^2}{2\sigma^2} \right) \quad (14)$$

802 where

$$803 \quad S_e = 0.5 + 0.5 \operatorname{erf} \left(-\frac{[\ln(\Psi/\Psi_0)/\sigma - \sigma]}{\sqrt{2}} \right) \quad (15)$$

804 and

$$805 \quad \sigma = \frac{\ln \left(\frac{\Psi_2}{\Psi_1} \right)}{x_2 - x_1} \quad (16)$$

$$806 \quad \Psi_0 = \frac{\Psi_1}{e^{(x_1 + \sigma)\sigma}} \quad (17)$$

807 where the constants x_1 and x_2 are given by

$$808 \quad x_1 = -\sqrt{2}(\operatorname{erf}^{-1}[2 \times 0.05 - 1]) \quad (18)$$

$$x_2 = -\sqrt{2}(\text{erf}^{-1}[2 \times 0.95 - 1]) \quad (19)$$

810 For the fracture domain only, the pore size distribution is modified as a function of
 811 depth, using the relationship

$$\psi_1^f = \psi_{1,\infty}^f + \frac{\psi_{1,0}^f - \psi_{1,\infty}^f}{1 + \exp(z_\alpha(z - z_\beta))} \quad (20)$$

813 The fracture domain fraction is also modified with depth using the relationship

$$w_f = w_{f,\infty} + \frac{w_{f,0} - w_{f,\infty}}{1 + \exp(z_\alpha(z - z_\beta))} \quad (21)$$

815 and the matrix domain fraction is given by

$$w_m = 1 - w_f \quad (22)$$

817 All symbols appearing here that are not given by one of these relationships are
 818 parameters, listed in Table 3.

Variable	Site	Instrument	Frequency	Period
Precipitation	WF	Tipping bucket raingauge	hourly	Sep 2003 - Sep. 2008
Actual Evap.	WF	Eddy flux correlation	hourly	Sep 2003 - Sep 2007
Potential Evap.	WF	Automatic weather station	hourly	Sep 2003 - Sep 2008
Water content	WF	Profile probes ¹ (≤ 1.0 m BGL)	15 min.	Jan 2004 - Mar 2007
Matric potential	WF	Pressure transducer tensiometers (≤ 1.2 m BGL)	15 min.	Jan 2004 - Jan 2008
Matric potential	WF	Equitensiometers (1.0 - 4.0 m BGL)	15 min.	Jan 2004 - Jan 2008
Matric potential	EI	Deep jacking tensiometers (10 - 24 m BGL)	hourly	Sep 2005 - Jan 2008
Water table	EI	Piezometer	hourly	Sep 2005 - Jan 2008

1. Profile probes were calibrated against 2 weekly neutron probe readings, as described in

Ireson et al., 2006

Table 1

Instrumentation used in this study

Extreme low intensity rainfall	Non-extreme rainfall/Extreme long duration rainfall	Extreme high intensity rainfall
Continuous slow drainage from the matrix; recharge persists throughout the summer (drought resilience)	Recharge via matrix and partially saturated fractures, lags of 10s of days. Cause of historic GW flooding.	Rapid bypass recharge, through fractures, lags of < 1 day. Potential to contribute to GW flooding.

Table 2

Modes of recharge in the Chalk (after Ireson et al., 2009)

Parameter	Parameter value	
	original ECM	refined DCM
θ_r^m	0	0
θ_s^m	0.35	0.35
θ_r^f	0	0
θ_s^f	1	1
$w_{f,0}$	0.12	0.08
$w_{f,\infty}$	0.01	0.001
a_0	-	0.03 m
a_∞	-	1.0 m
Ψ_1^m	-95.2 m	-95.2 m
Ψ_2^m	-14.1 m	-14.1 m
$\Psi_{1,0}^f$	-40.1 m	-40.1 m
$\Psi_{1,\infty}^f$	-1.29 m	-1.29 m
Ψ_2^f	-0.1 m	-0.1 m
K_s^m	0.53 mm/day	1.0 mm/day
K_s^f	2.83 m/day	27000 m/day
K_s^a	-	$K_s^m/100$
L^m	0.5	0.5
L^f	4.08	14.3
z_α	1.4 m^{-1}	1.4 m^{-1}
z_β	0.89 m	0.89 m
S_s^m	10^{-6} m^{-1}	10^{-6} m^{-1}
S_s^f	10^{-6} m^{-1}	10^{-6} m^{-1}

Table 3

Summary of all model parameters

Model run	Extreme event bypass	2007 total bypass
Benchmark	9.6 %	7.0 %
AE model run	9.2 %	14.4 %
<i>Sensitivity to hydraulic conductivity</i>		
$K_s^m=0.002$	0 %	0 %
$K_s^m=0.0005$	23.9 %	61.6 %
$K_s^f=2700$	3 %	1.4 %
$K_s^f=270000$	15.5 %	10.1 %
<i>Sensitivity to storage</i>		
Soil b)	0 %	0 %
Soil c)	0 %	0 %
Soil d)	43.9 %	21 %
Soil e)	0 %	0 %
$w_{f,\infty}=0.01$	0 %	0 %
$w_{f,\infty}=0.0001$	20.5 %	15.4 %
<i>Sensitivity to fracture-matrix exchange</i>		
$K_s^A = K_s^m / 10$	1.60 %	0.70 %
$K_s^A = K_s^m / 1000$	25.3 %	26.9 %
$a_0=0.5$	8.7 %	7.3 %
$a_0=0.003$	11.1 %	7.2 %
$a_\infty=0.5$	2.5 %	1.2 %
$a_\infty=2$	12.8 %	17.8 %

Table 4

Sensitivity of preferential recharge to different model configurations. Extreme event bypass is calculated as the volume of fracture flow on 20th July 2007, divided by the volume of rainfall on that day. 2007 total bypass is calculated as the volume of fracture flow in 2007 divided by the volume of fracture and matrix flow in 2007.

Soil	$w_{f,\infty}$	$w_{f,0}$	z_α	z_β
a)	0.001	0.08	1.4	0.89
b)	0.001	0.113	30	0.8
c)	0.001	0.02	3	4.6
d)	0.001	0.01	1.4	0.89
e)	0.001	0.12	1.4	0.89

Table 5

Parameters describing the different soil/weathered Chalk profiles used in Figure 13

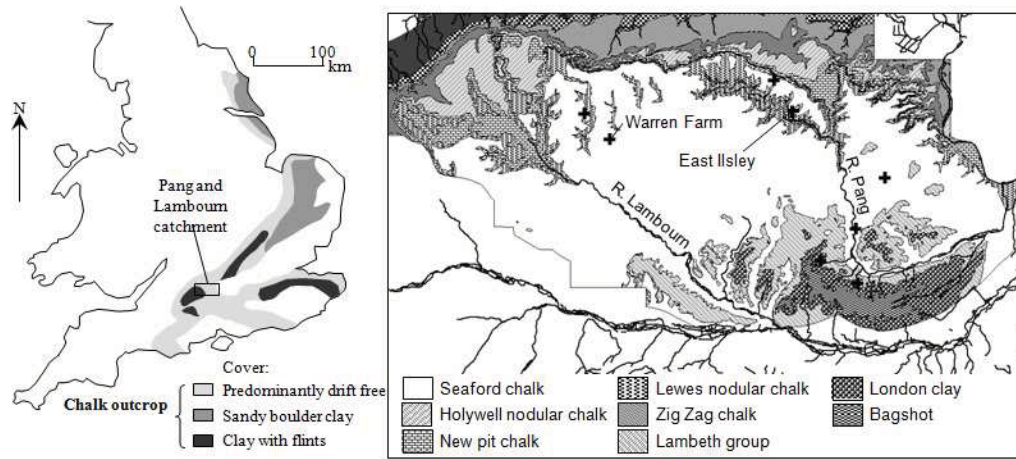


Fig. 1. Location of the catchments and field sites

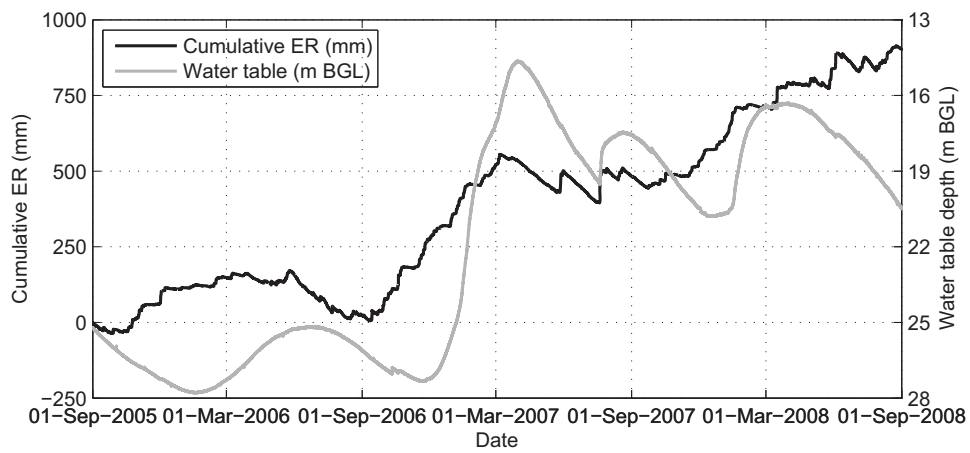


Fig. 2. Water table response to cumulative effective rainfall at East Ilsley

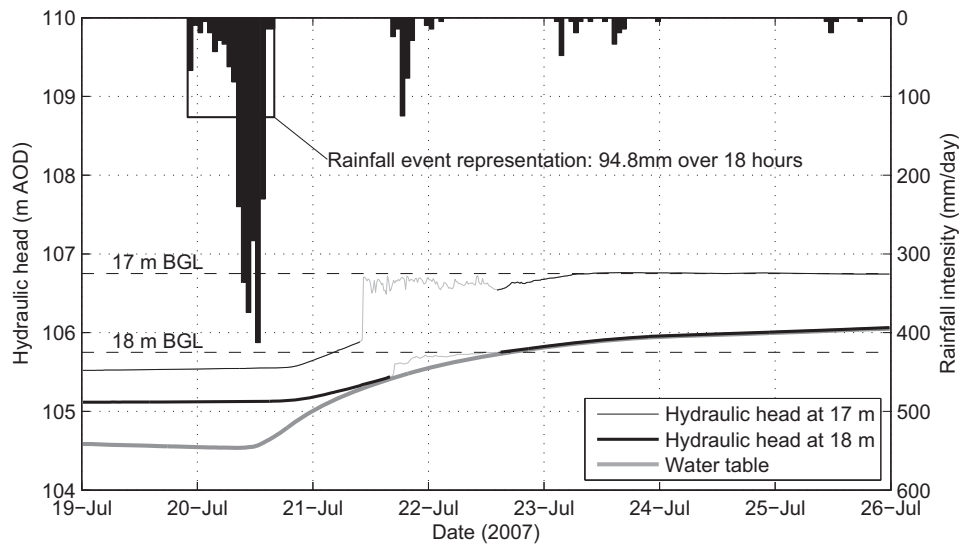


Fig. 3. Response to the rainfall event on 20th July 2007

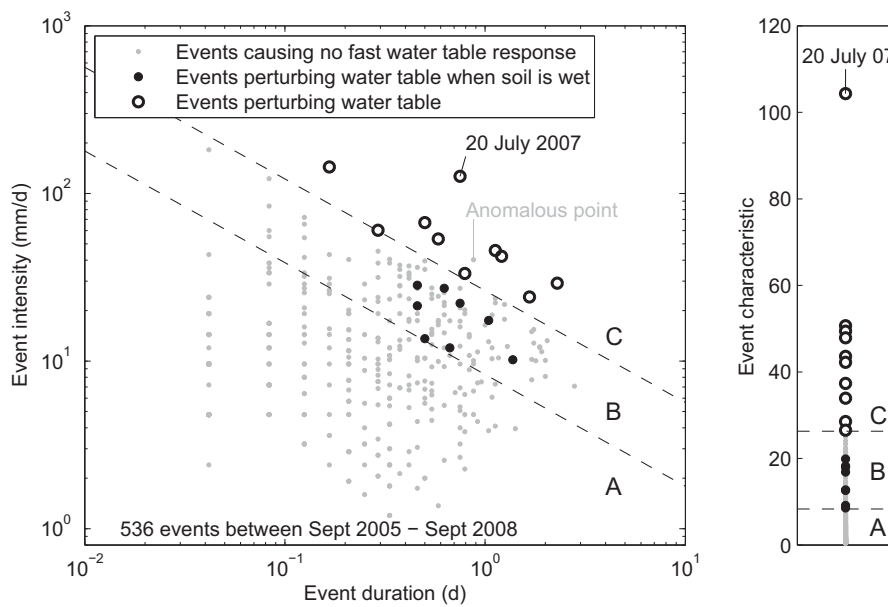


Fig. 4. Rainfall mean intensity versus duration (left) and Event Characteristic, EC, (right)

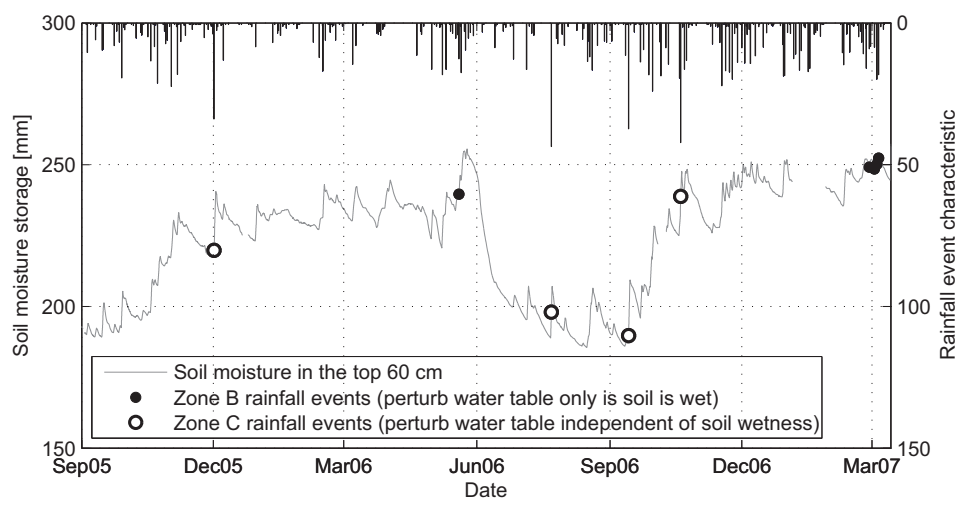


Fig. 5. Observed soil moisture storage in the top 60 cm, showing occurrence of EC region B and C rainfall events

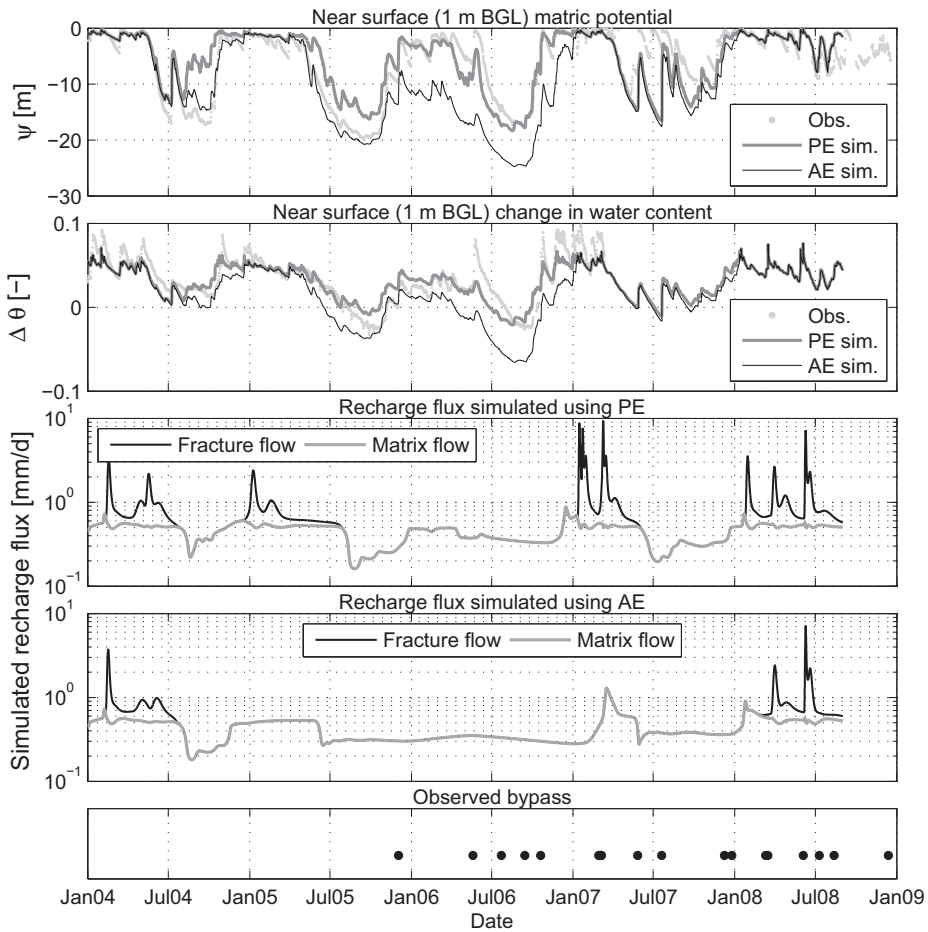


Fig. 6. Observed soil moisture data from Warren Farm compared with the model of Ireson et al. (2009b), driven using actual evapotranspiration, AE, estimated from eddy flux correlation and Penman Monteith reference crop evapotranspiration, PE

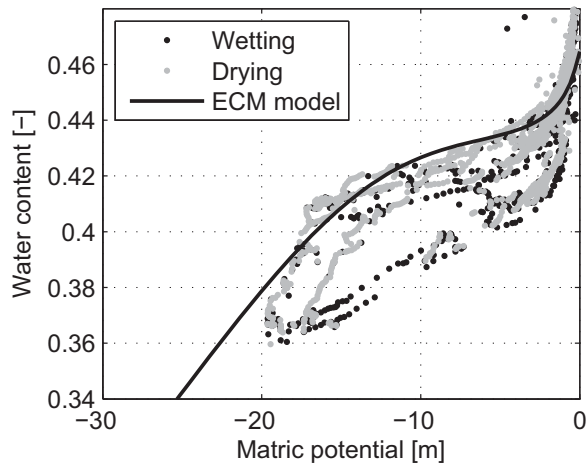


Fig. 7. Observed hysteretic soil moisture characteristic

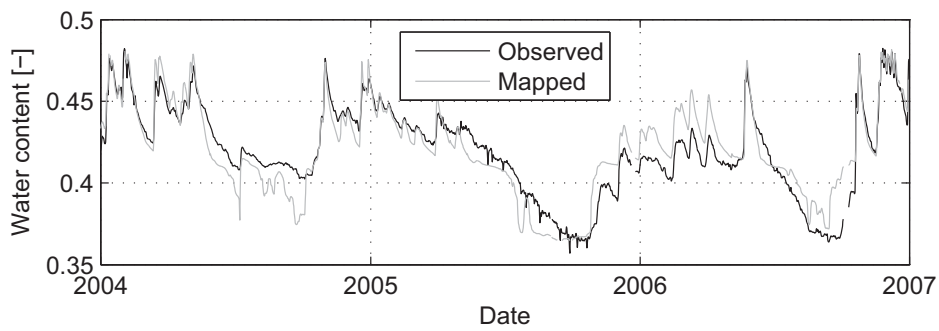


Fig. 8. Consistency of water content and matric potential observations: Observed water content vs water content obtained from matric potential using quantile mapping

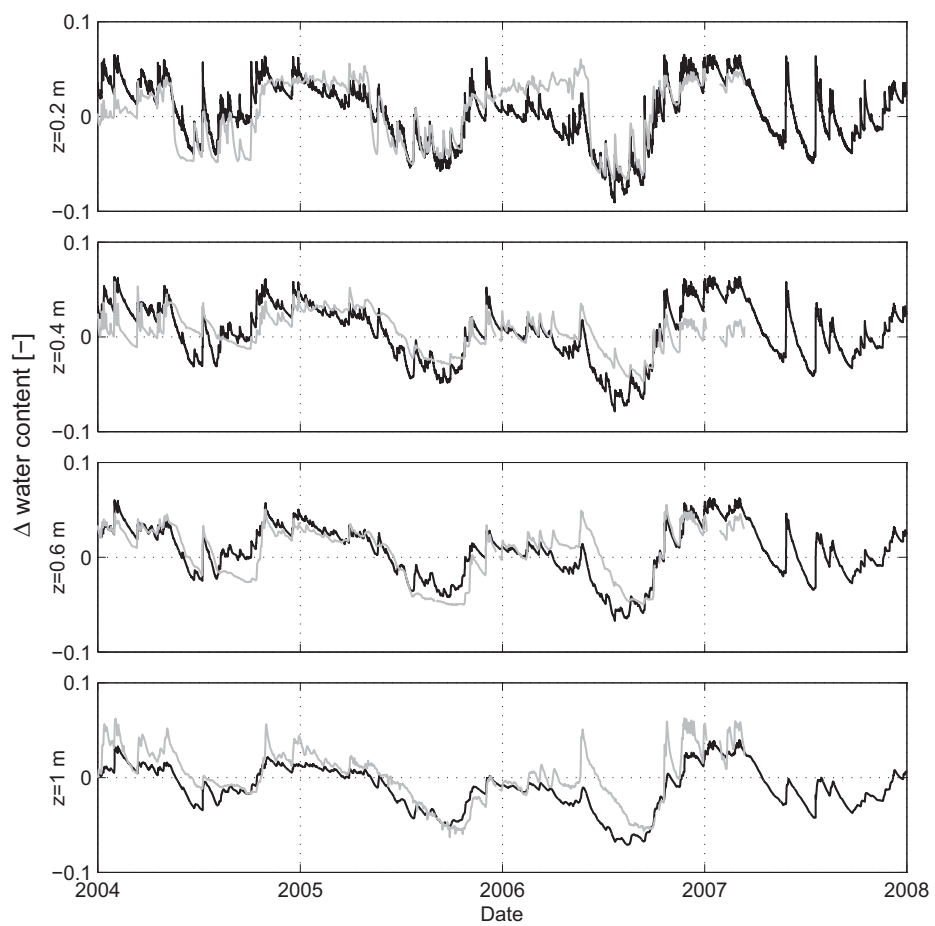


Fig. 9. Performance of the refined DCM model to reproduce observed changes soil moisture content at Warren Farm

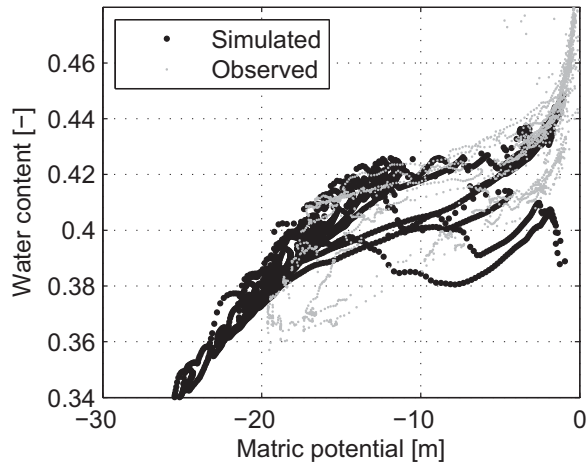


Fig. 10. Observed and simulated hysteretic soil moisture characteristic at 1.0 m depth

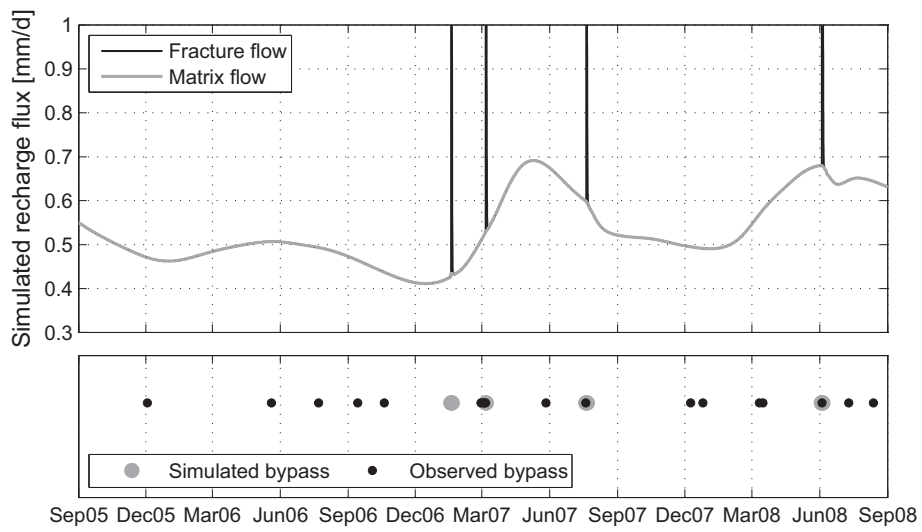


Fig. 11. Performance of the refined DCM model to reproduce preferential recharge responses

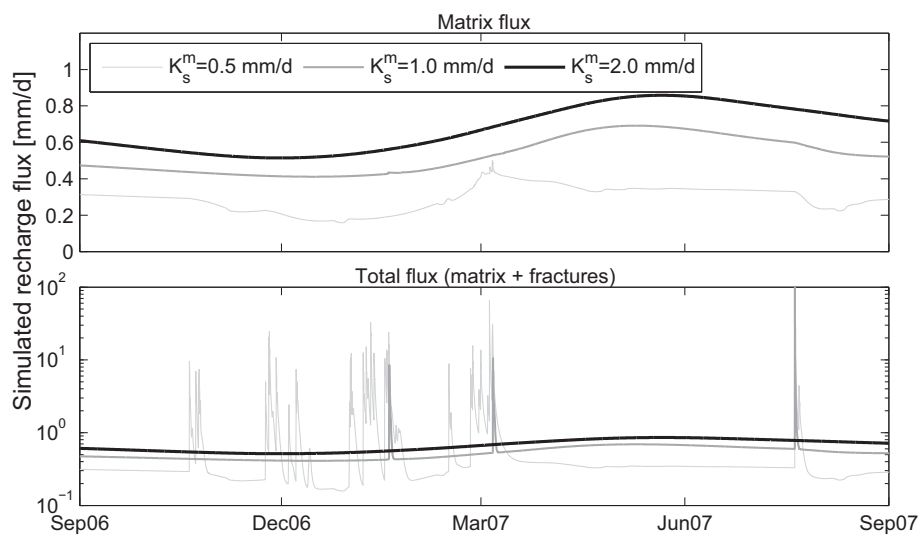


Fig. 12. Sensitivity of recharge fluxes to changes in the Chalk matrix saturated hydraulic conductivity, K_s^m

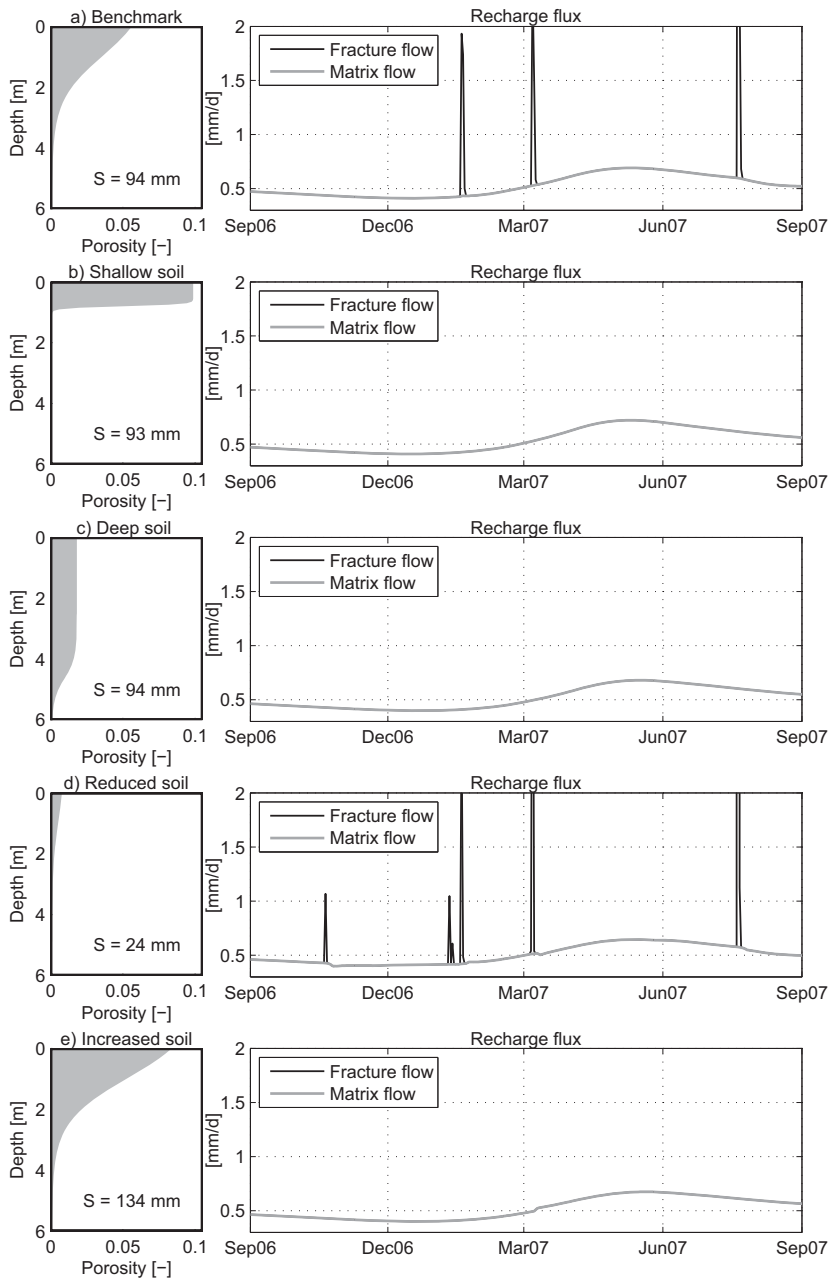


Fig. 13. Sensitivity of recharge fluxes to changes in the dynamic near surface storage in the soil/weathered Chalk