

1 **Meta-analysis of the effects of soil properties, site factors** 2 **and experimental conditions on solute transport**

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8 9 **Abstract**

10 Preferential flow is a widespread phenomenon that is known to strongly affect solute transport in
11 soil, but our understanding and knowledge is still poor of the site factors and soil properties that
12 promote it. To investigate these relationships, we assembled a database from the peer-reviewed
13 literature containing information on 733 breakthrough curve experiments under steady-state flow
14 conditions. Most of the collected experiments (585 of the 733 datasets) had been conducted on
15 undisturbed soil columns, although some experiments on repacked soil, clean sands, and glass
16 beads were also included. In addition to the apparent dispersivity, we focused attention on three
17 indicators of preferential solute transport, namely the 5%-arrival time, the holdback factor, and
18 the ratio of piston-flow and average transport velocities. Our results suggest that in contrast to
19 the 5%-arrival time and the holdback factor, the piston-flow to transport velocity ratio is not
20 related to preferential macropore transport but rather to the exclusion or retardation of the
21 applied tracer. Confirming that the apparent longitudinal dispersivity is positively correlated with
22 the travel distance of the tracer, our results also illustrate that this correlation is refined if the
23 normalized 5%-tracer arrival time is also taken into account. In particular, we found that the
24 degree of preferential solute transport increases with apparent dispersivity and decreases with
25 travel distance. A similar but weaker relationship was observed between apparent dispersivity,
26 5%-tracer arrival time, and lateral observation scale, such that the degree of preferential transport
27 increases with lateral observation scale. However, we also found that the travel distance and the
28 lateral observation scale in the investigated dataset are correlated which makes it difficult to

29 distinguish their influence on these transport characteristics. We observed that anionic tracers
30 exhibited larger apparent dispersivities than electrically neutral tracers under comparable
31 experimental conditions. We also found that the strength of preferential transport increased at
32 larger flow rates and water saturations, which suggests that macropore flow was a more
33 important flow mechanism than heterogeneous flow in the soil matrix. Nevertheless, our data
34 shows that heterogeneous flow in the soil matrix also occasionally leads to strong preferential
35 transport. Furthermore, we show that preferential solute transport under steady-state flow
36 depends on soil texture in a threshold-like manner: moderate to strong preferential transport was
37 found to occur only for undisturbed soils which contain more than 8% clay. Preferential flow
38 characteristics were also absent for columns filled with glass beads, clean sands, or sieved soil.
39 No clear effect of land use on the pattern of solute transport could be discerned, probably
40 because the available dataset was too small and too much affected by cross-correlations with
41 experimental conditions. Our results suggest that in developing pedotransfer functions for solute
42 transport properties of soils it is critically important to account for travel distance, lateral
43 observation scale, and water flow rate and saturation.

44 **1 Introduction**

45 During recent decades the number and quantity of man-made substances that are released onto
46 the soil has been increasing exponentially. Therefore it is becoming more and more important to
47 be able to quantify and predict water and solute fluxes through soil as knowledge of the latter is
48 fundamental to deciding on appropriate prevention or remediation strategies. Quantitatively
49 accurate estimation of water and solute fluxes in soils requires knowledge of hydraulic and solute
50 transport properties. However, their direct measurement is labour-intensive and costly. As they
51 are in most cases also spatially highly variable, it is not possible to measure them directly at a
52 sufficiently high spatial resolution at the relevant scales for management, such as the field,
53 region or landscape scale. Pedotransfer functions (PTFs) offer a way out of this dilemma
54 (Wösten et al., 2001). PTFs denote an approach in which soil properties that are difficult to
55 measure, e.g. the water retention properties, are estimated using other soil properties that are
56 easier to measure, e.g. the bulk density or texture, as proxy variables. Most work so far has
57 focused on soil hydraulic properties, and very little effort has been devoted to developing PTF's
58 for solute transport characteristics. Some approaches for identifying 'local' PTFs for parameters
59 of the convection-dispersion equation (CDE) or the mobile-immobile model (MIM) have been

60 published based on relatively small datasets (less than 25 samples in all cases) that had been
61 collected explicitly for the purpose (e.g. Goncalves et al. 2001; Perfect et al. 2002; Shaw et al.,
62 2000; Vervoort et al., 1999). In other studies, data from peer-reviewed literature was assembled
63 to construct larger databases of solute breakthrough curve (BTC) experiments (e.g. Rose, 1977;
64 Beven et al. 1993; Griffioen et al. 1998; Oliver and Smettem, 2003). In these studies, the authors
65 investigated correlations among CDE and MIM model parameters of between 50 and 359 BTC
66 experiments, but links to soil properties and experimental conditions were hardly discussed. In
67 contrast, such links were explicitly established in the study by Bromly et al. (2007), who focused
68 on the relationship of a CDE model parameter, the (longitudinal) dispersivity, to properties of
69 saturated repacked soil columns. Their database comprised 291 entries. Another large database
70 of BTC data was published by Vanderborght and Vereecken (2007). It contains 635 datasets of
71 flux and resident concentration BTC experiments with conservative tracers on undisturbed soil
72 and covers all scales between the small column-scale and the field-scale. Vanderborght and
73 Vereecken (2007) used the dataset to investigate how the longitudinal dispersivity is related to
74 scale, boundary conditions, soil texture, and measurement method. They confirmed that the
75 transport distance and the longitudinal dispersivity are generally positively correlated in soils.
76 The same observation had been previously reported for tracer experiments in groundwater
77 (Gelhar et al., 1992; Neuman, 1990).

78 All of the above discussed studies have ‘a priori’ assumed the validity of one solute transport
79 model, usually the CDE or the MIM. However, it seems likely that no single model is able to
80 properly characterize all of the contrasting flow regimes found in soils, including convective-
81 dispersive transport, heterogeneous flow (funnel flow), non-equilibrium flow in soil macropores
82 or unstable finger flow (Jury and Flühler, 1992). Indeed, it is commonly found that the flow or
83 mixing regime may change one or more times along the travel path (e.g. Vanderborght et al.,
84 2001), as soils are predominantly layered in the horizontal direction and solute transport
85 normally takes place in the vertical direction. In effect, a simple generally applicable model for
86 solute transport in soils that is at the same time consistent with the underlying physics is
87 presently not available. Therefore, model-independent (non-parametric) PTFs for solute transport
88 properties should be preferred to model-dependent ones. Some indicator of the strength of
89 preferential transport is then required in place of the model parameters. Several candidates for
90 such an indicator have been proposed during recent years. Among them are the skewness of the

91 BTC (e.g. Stagnitti et al., 2000), the pore volumes drained at the arrival of the peak concentration
92 (Ren et al., 1996; Comegna et al., 1999), the ‘holdback factor’, defined as the amount of original
93 water remaining in a column when one pore volume of displacing water has entered
94 (Danckwerts, 1953; Rose, 1973) and early quantiles of solute arrival times (Knudby and Carrera,
95 2005).

96 In this study, we expand and broaden earlier efforts (e.g. Vanderborght and Vereecken, 2007) to
97 develop a database of solute transport experiments derived from the published literature, which
98 comprises a larger number of BTCs (n=733) with accompanying information on soil properties,
99 site factors (e.g. land use and soil management) and experimental conditions. In contrast to
100 Vanderborght and Vereecken (2007) we only included BTC experiments with direct flux
101 concentration measurements to improve comparability of the collected data. Our main
102 motivation for this work was to create a dataset of transport experiments to enable the future
103 development of non-parametric PTFs for inert solute transport. In this paper, we present the
104 database and the results of initial analyses that relate derived BTC-shape measures to
105 experimental boundary conditions, soil properties and site factors.

106 **2 Material and methods**

107 We collected information on 733 BTCs for inert tracers in steady-state flow experiments on
108 undisturbed soil samples and from a smaller number of columns filled with glass beads, clean
109 sands, or sieved and repacked soil. The data was taken from 76 articles published in the peer-
110 reviewed literature. Details on the data sources are given in Table 1. We deliberately excluded
111 BTCs consisting of resident concentration data (e.g. sampled by time-domain reflectometry) or
112 data from local sampling methods (e.g. suction samplers). Thus, all the considered BTCs were
113 obtained from measurements of flux concentrations in column or tile-drain effluents. Alongside
114 the BTCs, additional information on corresponding soil properties, site factors and experimental
115 conditions was gathered and stored in a relational MySQL database. Table 2 gives an overview
116 on soil properties, site factors and experimental conditions collected in the database as well as
117 information on their completeness.

118 One difficulty in comparing experimental data is that several different soil texture classification
119 systems were used in the 76 articles. All the classification systems have in common that they
120 assign all particles with an equivalent diameter of less than two micrometers to the clay fraction,

121 but the boundary between the silt and sand fraction varies. We standardized all texture data to the
122 USDA classification system, which sets the silt/sand boundary at 50 μm . We did this by log-
123 linear interpolation (Nemes et al., 1999). For soil columns containing two or more soil layers, we
124 derived an effective soil textural composition by calculating the layer-thickness-weighted
125 average of the sand, silt and clay fractions, respectively. In addition, we computed the geometric
126 mean grain diameter using the approach published in Shirazi et al. (2001).

127 Another difficulty in comparing the shapes of different BTCs arises from the fact that the pulse
128 length during which the tracer was applied varies with the corresponding source publication. It is
129 therefore necessary to normalize the BTCs to a standard tracer application. We chose a Dirac-
130 like input as our standard. For this type of tracer application the travel-time probability density
131 function (PDF) of the tracer at the measurement location can be derived by simple scaling. This
132 process is denoted as BTC-deconvolution in the following. For the BTC-deconvolution, a
133 pseudo-transfer-function $f(d^{-1})$ is sought which describes the BTC, here denoted as $C_{\text{out}}(-)$, for a
134 given tracer application function $C_{\text{in}}(-)$:

$$135 \quad C_{\text{out}} = \int_0^{\infty} C_{\text{in}}(t - \tau) f(\tau) d\tau. \quad (1)$$

137 The solute concentrations C_{out} and C_{in} were normalized to a reference concentration. They are
138 therefore dimensionless. We also standardized all time variables including t (d) and τ (d) in eq. 1
139 to days. We denoted f as the “pseudo-transfer-function” because we do not attach any physical
140 meaning to it. It is important to note that f does not (necessarily) describe the evolution of the
141 BTC along the travel trajectory. Our study only requires that f fits eq. 1 at the location of the
142 measurement, namely at the outlet of the soil columns. This allows us to use arbitrary transfer
143 function types to estimate the PDF of the BTC, as long as it is able to fit the BTC data well
144 enough.

145 One advantage of this is that we can use CDE and MIM parameters-sets to reconstruct the
146 pseudo-transfer-function, f . By using CDE and MIM parameter-sets, we were able to also include
147 studies in which only MIM or CDE model parameters were reported rather than raw data of the
148 actual BTCs. We only considered BTCs for which the corresponding model could be fitted with
149 a coefficient of determination $R^2 > 0.95$. Note that for some BTCs, no measure of goodness of fit

150 is given. In these cases we assumed that the fit was sufficiently well if the MIM was used alone
151 or alongside with the CDE (as e.g. in Seyfried et al. 1987). Otherwise, we decided by visual
152 inspection whether the CDE fitted the BTC well enough to be included in our study. As a result
153 733 BTCs were investigated in the following.

154 The 733 BTCs in our database consist of 146 BTCs scanned from raw data, 399 BTCs for which
155 only MIM parameters were available and 188 BTCs for which CDE parameters were published.
156 For the 146 datasets for which the BTC raw data was available, MIM parameters were inversely
157 determined by fitting CXTFit 2.1 (Toride et al., 1999, command-line version published as part of
158 the STANMOD package, version 2.07). We included this step to make the 146 datasets with
159 BTC raw data more comparable to the remaining 587 BTCs for which only model parameters
160 were available. A drawback to this approach is that some PDFs are then only reconstructed in an
161 approximate manner due to the limited degrees of freedom of the MIM transfer-function and its
162 inability to fit some of the BTCs. Nevertheless, the MIM and CDE fitted the BTC very well in
163 most cases, with a geometric mean coefficient of determination, R^2 , of 0.99. Alternative methods
164 for PDF-reconstruction could be preferable in those few cases where the CDE or MIM did not fit
165 well. For example, the BTCs could be deconvoluted using a mixture of standard-type transfer
166 functions (see e.g. Koestel et al., 2011) or by imposing a smoothness constraint (Skaggs et al.,
167 1998).

168 We used analytical solutions of the CDE and MIM for Dirac-pulse input, flux concentrations in
169 input and effluent and a semi-infinite domain (Valocchi, 1985) to forward-model the pseudo-
170 transfer-functions which were then normalized to PDF's. We then derived four non-parametric
171 shape-measures from the reconstructed pseudo-transfer-functions and PDFs (Koestel et al., 2011)
172 to evaluate the respective solute transport properties. We especially focused on indicators of
173 preferential solute transport.

174 According to (Hendrickx and Flury, 2001), preferential flow and transport processes comprise
175 “all phenomena where water and solutes move along certain pathways while bypassing a fraction
176 of the porous matrix”. This is a rather vague definition as it remains unclear how the “porous
177 matrix” is defined or how large the “bypassed fraction” has to be. A more operational definition
178 of preferential transport is a mixing regime that is not convective-dispersive which assumes
179 complete mixing in the directions transverse to the flow (Flühler et al., 1996). For a convective-

180 dispersive mixing regime, the transport is described by the CDE. However, it is not possible to
 181 test the validity of the CDE with the type of data collated in our study, comprising breakthrough
 182 curves measured at one only travel distance (Jury and Roth, 1990). It is, therefore, more
 183 applicable for us to define the strength of preferential transport as the deviation of a BTC-shape
 184 from “piston-flow”-transport. The latter refers to the case of complete absence of any
 185 heterogeneity in the transport process. This implies also that all the water in the porous medium
 186 contributes equally to the solute transport. The shape of a BTC for piston-flow-transport is
 187 clearly defined. Its shape is identical to the one of the tracer-input time-series at the upper
 188 boundary of the soil column. The first, average and last tracer arrival times are identical and the
 189 average transport velocity equals the piston-flow velocity. In the following we use the term
 190 “preferential transport” to address BTCs with shape-measures indicating a large deviation from
 191 piston-flow.

192 The first indicator we investigated is the ratio of the piston-flow velocity, v_q (cm d⁻¹), to the
 193 average transport velocity, v_n (cm d⁻¹), denoted as η (-) and defined by

$$194 \quad \eta = \frac{v_q}{v_n} \quad (2)$$

196 where

$$197 \quad v_q = \frac{q}{\theta} \quad (3)$$

199 and

$$200 \quad v_n = \frac{L}{\mu'_1} \quad (4)$$

202 where q (cm/d) is the water flux, θ is the (total) volumetric water content (-), L is the column
 203 length (cm) and μ'_1 is the normalized first moment of the PDF,

204
$$\mu_1' = \frac{m_1}{m_0}$$

205 (5)

206 where m_0 and m_1 are the zeroth and first moments of the pseudo-transfer-function, f , respectively,
 207 defined as

208
$$m_0 = \int_0^{\infty} f dt$$

209 (6)

210 and

211
$$m_1 = \int_0^{\infty} t f dt .$$

212 (7)

213 The piston-flow to transport velocity ratio, η , is smaller than one if the solute is transported
 214 faster than the water and it is larger than one if the solute is retarded relative to the water. It is a
 215 non-parametric analogue to the retardation coefficient in the CDE and MIM. Vanderborght and
 216 Vereecken (2007) used the reciprocal of η , i.e. $1/\eta$, to investigate preferential transport. They
 217 suggested that $\eta < 1$ indicates bypass flow.

218 The second shape-measure used in this study is the normalized arrival-time of the first five
 219 percent of the tracer, $p_{0.05}$ (-). It can be derived from the normalized arrival times, T (-),

220
$$T = \frac{t}{\mu_1'}$$

221 (8)

222 and the PDF, f_n (-),

223
$$f_n = f \mu_1'$$

224 (9)

225 It is more easily obtained from the dimensionless cumulative distribution function (CDF), F_n (-),
 226 which is calculated by integrating f_n ,

227

$$F_n = \int_0^T f_n dT$$

228

(10)

229 Figure 1 illustrates how $p_{0.05}$ is derived for a BTC taken from Garré et al. (2010). $p_{0.05}$ is bounded
 230 by zero and one, where a value of one indicates piston flow. According to the numerical studies
 231 carried out by Knudby and Carrera (2005), $p_{0.05}$ is negatively correlated with the degree of
 232 preferential transport, since it indicates an early tracer arrival. The results of Koestel et al. (2011)
 233 indicate that early tracer arrivals are correlated with a long tailing. Note that these two BTC
 234 shape-features, early tracer arrival and a long tailing, are generally associated with preferential
 235 transport (see Brusseau and Rao, 1990).

236 We also investigated the holdback factor, H (-), as another indicator of early tracer arrival. This
 237 was introduced by Danckwerts (1953) to characterize the degree of mixing of two solutes in a
 238 vessel:

239

$$H = \int_0^1 F_n dT$$

240

(11)

241 It corresponds to the ‘amount of original fluid remaining in the column when one (water-filled)
 242 pore volume of displacing fluid has entered’ (Rose, 1973). It follows that a large H should
 243 indicate preferential characteristics in a transport process. H is calculated as the integral of the
 244 dimensionless CDF between zero and one. The holdback factor, H , is also illustrated in Figure 1.
 245 H has the advantage over $p_{0.05}$ that it samples a larger part of the CDF, but has the disadvantage
 246 that it is less robust to the type of pseudo-transfer-function chosen for the BTC-deconvolution
 247 (Koestel et al., 2011).

248 Finally, we also investigated the apparent dispersivity, λ_{app} (cm), which is defined as

249

$$\lambda_{app} = \frac{\mu_2 L}{2}$$

250

(12)

251 where μ_2 (-) is the second central moment of the PDF,

252

$$\mu_2 = \int_0^{\infty} (T-1)^2 f_n dT$$

253

(13)

254 Note that μ_2 , as it is defined here, is identical to the squared coefficient of variation. The apparent
 255 dispersivity, λ_{app} , is generally thought to be an indicator of heterogeneity of the solute transport
 256 process (Vanderborght and Vereecken, 2007). Koestel et al., (2011) found that λ_{app} is correlated
 257 to $p_{0.05}$ and H , but also carries additional information on the transport process and thus may
 258 complement the above discussed shape-measures. Because the additional information contained
 259 in λ_{app} stems from the late-arriving tracer it has the disadvantage that it is less robust to the type
 260 of pseudo-transfer-function chosen for the BTC-deconvolution than $p_{0.05}$ (Koestel et al., 2011),
 261 i.e. λ_{app} is less well defined by the BTC-data than $p_{0.05}$ and H . One advantage of λ_{app} as a shape
 262 measure is that it has already been intensively investigated in the literature (Bromly et al., 2007;
 263 Vanderborght and Vereecken, 2007; Hunt and Skinner, 2010).

264 3 Results and discussion

265 302 of the 733 experiments available in the database correspond to undisturbed soil samples
 266 from arable land (Table 3). 219 of them are from conventionally-tilled fields, 6 from fields with
 267 reduced or conservation tillage and 31 from fields with no tillage at all. For the remaining 46
 268 samples, the soil management practices were not specified. Managed or natural grassland is the
 269 second most common land use type represented in the database (n=104). Samples with arable
 270 and grassland land use are distributed over most of the texture triangle with no apparent bias
 271 towards any textural class (see Figure 2). In contrast, the 79 BTCs from samples from forest sites
 272 are restricted to soil samples with less than 25% clay (Figure 2). Other land uses, like orchard
 273 (n=19), grass ley (n=7) or heathland (n=2), are rare. 98 BTCs were measured on samples with
 274 unspecified land use. Finally, the 733 datasets also contain 116 experiments on sieved and
 275 repacked columns, 32 experiments on columns filled with clean sands or glass beads and 60
 276 experiments on undisturbed samples taken from more than 1 m below the land surface (Table 3).
 277 All studies were conducted on soil columns. Figure 2 illustrates that the majority of the solute
 278 transport experiments had been performed on undisturbed but rather short soil columns which
 279 had been sampled from one single soil horizon (see also Table 3).

280 An overview of Spearman rank correlations among the investigated soil properties, experimental
281 conditions, and BTC shape measures is given in Figure 3. The asterisks indicate p-values of less
282 than 0.001. Some correlations are unsurprising, such as the positive correlations between the
283 flux, q , the average transport velocity, v , the average pressure head, h , and the water content, θ .
284 Other similar examples are the correlations between geometric mean grain diameter, d_g , bulk
285 density, ρ , and clay, silt, and sand fractions. Also, the positive correlation between average
286 sampling depth and the soil sample length (which is identical to the travel distance), L , is easily
287 explained, as sampling pits for larger soil columns must necessarily extend deeper into the
288 ground. Likewise, the column cross-section, A , is positively correlated with L (and the sampling
289 depth).

290 We found a positive correlation of the apparent dispersivity, λ_{app} , with travel distance, L , and
291 lateral observation scale, A . This confirms what has been in general found in already published
292 reviews on dispersivity (e.g. Gelhar et al., 1992; Vanderborght and Vereecken, 2007), although it
293 is hardly possible to separate the effects of L and A on λ_{app} due to their large mutual correlation.
294 Also consistent with previous studies, Figure 3 shows a positive correlation between the apparent
295 dispersivity, λ_{app} , and the water flux, q , as well as the pressure head, h . Furthermore, the
296 correlation coefficients with texture data show that λ_{app} was in general larger for finer textured
297 soil and smaller for coarse textures which also is in accordance with empirical knowledge and
298 has also been reported by Vanderborght and Vereecken (2007). Finally, we observed no
299 correlation between organic carbon content, OC , and apparent dispersivity, λ_{app} .

300 Two of the three investigated indicators of early tracer arrival, namely the normalized 5%-arrival
301 time, $p_{0.05}$, and the holdback factor, H , were strongly negatively correlated. This confirms the
302 findings of Koestel et al. (2011) on a smaller dataset. According to these two shape-measures,
303 the degree of preferential transport increased with flux, q , pressure head, h , and water content, θ .
304 This is consistent with empirical findings that show that preferential flow and transport are more
305 likely to be observed under saturated and near-saturated conditions (Langner et al, 1999;
306 Seyfried and Rao, 1987). The correlation matrix indicates that the degree of preferential transport
307 was positively correlated with the lateral observation scale, A , but not with the transport distance,
308 L . An intuitive explanation for this is that increasing the lateral observation scale also increases
309 the probability of sampling preferential flow paths, whereas an increase in transport distance

310 decreases the probability of connected preferential flow paths in the transport direction. We
311 consider it likely that a negative correlation between transport distance and preferential transport
312 characteristics was masked by the strong mutual correlation between L and A . Both shape-
313 measures, $p_{0.05}$ and H , indicate a positive correlation between the degree of preferential transport
314 and the clay and silt fraction, and a negative correlation to the geometric mean grain diameter
315 and the sand fraction. Also, a weak negative correlation between the strength of preferential
316 transport and bulk density, ρ , was found, but no correlation to the organic carbon content, OC .

317 The fourth shape-measure, the piston-flow to transport velocity ratio, η , was not significantly
318 correlated to the normalized 5%-arrival time, $p_{0.05}$. A very weak positive correlation was found
319 between η and the holdback factor H and to the apparent dispersivity, λ_{app} . Moreover, we
320 observed that solute transport was increasingly retarded ($\eta > 1$) with increasing water flow rate,
321 q , and pressure heads, h . We found no significant correlations between η and any of the
322 investigated soil properties (i.e. geometric mean grain diameter, d_g , bulk density, ρ , texture
323 fractions and organic carbon content, OC). It follows that the piston-flow to transport velocity
324 ratio, η , reflects different information on solute transport characteristics as compared to the other
325 indicators for early tracer arrival, $p_{0.05}$ and H .

326 Figure 4a shows that strong correlation between the 5%-arrival time, $p_{0.05}$, and the holdback
327 factor, H , was weaker for small $p_{0.05}$ (large H), i.e. for BTCs displaying strong preferential
328 transport. Figure 4a suggests that H offers a better discrimination between soils showing strong
329 preferential transport whereas $p_{0.05}$ better resolves differences among soils with weaker
330 preferential transport characteristics. In Figure 4b and c, the piston-flow to transport velocity
331 ratio, η , is compared to $p_{0.05}$ and H . Note that no value for η was available if no independent
332 water content measurement was published for the respective BTC (see Eq. 2). Therefore, the
333 range of $p_{0.05}$ in Figure 4b appears to be different to the one in Figure 4a. Besides depicting the
334 minimal correlation of η to the other two indicators of early tracer arrival, these two figures also
335 illustrate that η was, in contrast to $p_{0.05}$ and H , sensitive to the choice of tracer in the BTC
336 experiments. Anionic tracers like chloride and bromide were generally transported faster than the
337 water flux whereas the electrically neutral tracers deuterium and tritium only occasionally
338 exhibited accelerated transport, namely when small $p_{0.05}$ and medium H indicated preferential
339 characteristics. As we only considered experiments where the anionic tracers were applied on

340 soils with electrically neutral or predominantly negatively charged media, the generally
341 accelerated solute transport for anionic tracers is well explained by anion exclusion (Rose et al.,
342 2009; Thomas and Swoboda, 1970). Notably, for very strong preferential transport ($p_{0.05} < 0.1$
343 and $H > 0.4$), the anionic tracers were retarded.

344 Figure 5a and b illustrate the impact of the choice of tracer on BTCs. The non-ionic tracers
345 tritium and deuterium were generally used on longer columns than chloride and bromide and
346 under similar water fluxes. Although longer columns should lead to larger apparent
347 dispersivities, λ_{app} (Figure 3), this was not observed for the BTCs obtained with tritium and
348 deuterium. This supports the validity of model approaches in which the solute dispersivity is not
349 only dependent on the pore-space geometry but also on the adsorptive properties of tracer and
350 soil matrix (Wels et al., 1997; Pot and Genty, 2007). In addition, the strength of preferential
351 transport, as expressed by $p_{0.05}$, was smaller for the non-ionic tracers than for the anions.

352 Figure 6a illustrates that for a given value of λ_{app} , $p_{0.05}$ increases with the column length, L . This
353 suggests that the strength of preferential transport decreases with travel distance. No significant
354 correlation was found between L and $p_{0.05}$ (Fig. 2), probably because it was masked by the non-
355 linearity of the ternary relationship between L , $p_{0.05}$ and λ_{app} , especially for strong preferential
356 transport ($p_{0.05} < 0.1$). Thus, including $p_{0.05}$ into a scaling-scheme for the apparent dispersivity,
357 λ_{app} , with travel distance, L , strongly increases the amount of explained variance. A principal
358 component analysis revealed that the first two principal components for the three measures \log_{10}
359 L , $\log_{10} \lambda_{app}$ and $p_{0.05}$ (normalized to a mean of zero and a standard deviation of one) explain
360 91.9 % of the variance between the three shape-measures. In contrast, the first principal
361 component of just $\log_{10} \lambda_{app}$ and $\log_{10} L$ explains only 66.2 % of the variance, exhibiting a
362 Spearman rank correlation coefficient of 0.369 (p-value < 0.001). A very similar ternary
363 relationship was found between $\log_{10} \lambda_{app}$, $p_{0.05}$, and the logarithm of the area of the breakthrough
364 plane, $\log_{10} A$ (Figure 6b), which explained 88.7 % of the inherent variance. The first principal
365 component between only λ_{app} and A explains 70.3% of the variance. The corresponding
366 Spearman rank correlation coefficient is 0.5 (p-value < 0.001).

367 Figure 7a-d show the dependency of v , λ_{app} , $p_{0.05}$, and η on water flow rates. Only undisturbed
368 samples were considered. Figure 7a-c show that not only the medians of v and λ_{app} monotonously
369 increase with the respective water flux class but also the strength of preferential transport (there

370 is negative relationship between $p_{0.05}$ and q). Note that correlation effects between water flow
371 rate, q , and travel distance, L , and lateral observation scale, expressed by A , are ruled out since
372 these quantities were not correlated (Figure 3). For undisturbed samples only, we found a
373 significant but very weak positive correlation between the water flow rate, q , and the clay
374 content (Spearman rank correlation coefficient is 0.15, not shown). Therefore we conclude that
375 the water flow rate was the most important factor for the relationships shown in Figure 7a-c. This
376 suggests that, for this dataset, macropore transport overshadows preferential transport caused by
377 heterogeneities in matrix hydraulic properties. Nevertheless, Figure 7c also illustrates that
378 preferential transport cannot be completely ruled out for small water fluxes. Little dependence of
379 the piston-flow to transport velocity ratios, η , on the water flux, q , is observed (Figure 7c). This
380 suggests that η is not strictly related to preferential transport in soil macropores. Indeed, η is
381 smallest for the experiments with the lowest water fluxes. As most of the experiments included
382 in this analysis were conducted with anionic tracers, a possible explanation for this behavior is
383 that anion exclusion was amplified for experiments under small water flow rates which by trend
384 correspond to experiments under far from saturated conditions when only meso- and micropores
385 are water-filled.

386 Figure 8 depicts how the soil horizon from which the sample had been taken is related to λ_{app} and
387 $p_{0.05}$. Firstly, Figure 8 illustrates that samples that contain both topsoil and subsoil exhibit larger
388 apparent dispersivities, λ_{app} , than samples from only topsoil or only subsoil. One obvious
389 explanation for this is that samples containing both topsoil and subsoil are generally longer, so
390 that λ_{app} is also larger due to its positive correlation with travel distance (see Figure 6a).
391 However, it is also plausible that features at the interfaces between topsoil and subsoil in these
392 columns, e.g. plow pans, enhance the spreading of a solute plume, such as observed for example
393 by Öhrström et al. (2002) and Koestel et al. (2009b). As samples taken from only the topsoil are
394 always restricted to lengths between 20 and 40 cm and because longer samples taken from only
395 the subsoil have seldom been investigated, it is not possible to appraise to what degree interfaces
396 between topsoil and subsoil add to the scaling effect of the apparent dispersivity, λ_{app} , with travel
397 distance. Furthermore, soil columns filled with clean sands or glass beads, which are tagged as
398 ‘irrelevant’ in Figure 8, generated strictly non-preferential BTCs.

399 The relationship between λ_{app} and $p_{0.05}$ and soil texture, characterized by the geometric mean
400 grain diameter, d_g , is somewhat more complicated (see Figure 9). Coarser-textured soils with
401 large d_g are not at all restricted to a specific range of apparent dispersivities or 5%-arrival times,
402 or specific combinations of the two. In contrast, for fine-grained soils, $p_{0.05}$ is always less than
403 0.6 and the apparent dispersivity always exceeds ca. 2 cm. Finally, the samples with an
404 intermediate d_g show low λ_{app} -to- $p_{0.05}$ ratios upon visual inspection (Figure 9). Such a ratio is also
405 typical for short transport distances (Figure 6a). A possible explanation may be that in our
406 dataset, experiments on soils with intermediate d_g were only carried out on short columns. In
407 summary, there are no smooth transitions apparent in Figure 9 and the geometric mean grain
408 diameter appears not to be a strong predictor for λ_{app} and $p_{0.05}$.

409 A clearer picture emerges if λ_{app} and $p_{0.05}$ are plotted in relation to USDA texture classes. Figure
410 10a shows that BTCs showing strong preferential transport characteristics ($p_{0.05} < 0.2$) are
411 restricted to samples containing at least 8 to 9% clay. This is similar to the clay content needed
412 for the formation of stable soil aggregates (Horn et al., 1994) and may also reflect an absence of
413 biopores in such soils, since both roots and earthworms avoid coarse single-grain soils. Also,
414 small $p_{0.05}$ values are less common for samples with more than 50% silt. However, the latter may
415 possibly be an artifact caused by the scarcity of experiments on short columns sampled from just
416 one single soil horizon in silty soils (see Figure 10d). The apparent dispersivity, λ_{app} , roughly
417 follows the distribution of $p_{0.05}$ on the texture triangle diagram (Figure 10b) which is not
418 surprising given the strong correlation between the two (see Figure 6). However, extreme λ_{app}
419 values were less clearly constrained to specific regions on the texture triangle diagram. They
420 mostly occurred for undisturbed samples containing more than one soil horizon. Finally, Figure
421 10c shows the distribution of the piston-flow to transport velocity ratio, η on the texture triangle.
422 Small piston-flow to transport velocity ratios ($\eta \ll 1$), were predominantly found for loamy soils
423 and were absent for soils in which one of the three fractions (silt, sand or clay) dominates. The
424 complete absence of $\eta < 1$ for soils of clayey texture may be related to anion exclusion as all
425 these experiments were conducted with anionic tracers (see Figure 4b and discussion above).
426 Small η occur exclusively in loamy soils which are characterized by a broader particle (and thus
427 pore) size distribution than soils from other texture classes. As a broader pore size spectrum
428 should enhance heterogeneous transport in the soil matrix, it is possible that, in addition to anion
429 exclusion, η reflects heterogeneous transport in the matrix rather than macropore flow.

430 Finally, we also investigated the relationships of the BTC shape-measures λ_{app} and $p_{0.05}$ with land
431 use and soil management practices. Figure 11a and b illustrate that the 585 undisturbed soil
432 samples exhibited a median apparent dispersivity of 6.72 cm and a median normalized 5%-
433 arrival time of 0.3 corresponding to steady state flow conditions with a median flux of 12.7 cm/d
434 and a median travel distance of 20 cm. Much smaller $p_{0.05}$ values were only found for samples
435 from arable sites with reduced tillage and grass leys (Figure 11a). However, the number of
436 samples for these land use classes was very small, while Figure 11b reveals that the experiments
437 were conducted on relatively short columns and large water fluxes, both of which promote low
438 $p_{0.05}$. Similarly, the experimental conditions were also not representative for the bulk of the
439 experiments on undisturbed samples for the ‘forest’ sites. For these samples, the experimental
440 conditions promoted larger $p_{0.05}$ values (Figure 11b). Figure 11a and b show that sieved and
441 repacked soil samples resulted in clearly larger $p_{0.05}$ values than samples of undisturbed soil,
442 even though the experimental conditions favored small values. A lack of preferential transport
443 for the disturbed samples is consistent with the destruction of natural well-connected pore-
444 structures by sieving. This furthermore underlines the importance of conducting leaching studies
445 on undisturbed samples (see also Elrick and French, 1966; Cassel et al., 1974; McMahon and
446 Thomas, 1974). Furthermore, no sign of preferential transport was found for the BTCs collected
447 from artificial porous media like clean sand or glass beads. They exhibited extremely large $p_{0.05}$
448 and extremely small λ_{app} , although the experimental conditions should have acted in the opposite
449 direction. Of the natural soils, only the two samples from heathland sites consisting almost of
450 pure sand (Seuntjens et al. 2001) show similar features (Figure 11a). We conclude that, with a
451 few exceptions, a complete absence of preferential characteristics in solute transport is only
452 observed in artificial homogeneous porous media. Apart from this, our data does not show any
453 clear relationship between land use and degree of preferential transport and solute dispersion.
454 However, such relationships cannot be ruled out, since in our dataset they may have been
455 obscured by a lack of comparable experimental conditions.

456 **4 Conclusions**

457 We investigated the controls on inert solute transport based on 733 breakthrough curve
458 experiments collected from the peer-reviewed literature, mostly conducted on undisturbed soil
459 columns. We focused especially on four breakthrough curve shape-measures, namely the
460 normalized 5%-arrival time, the holdback factor, the apparent longitudinal dispersivity and the

461 ratio of piston-flow and average transport velocities. The normalized 5%-arrival time, the
462 apparent dispersivity and the holdback factor were strongly correlated, while only weak
463 correlations were found between these shape-measures and the piston-flow to transport velocity
464 ratio, suggesting that the latter contains complementary information on solute transport. In
465 particular, our results suggest that the piston-flow to transport velocity ratio is more strongly
466 related to exclusion or retardation of the applied tracer and preferential transport in the soil
467 matrix, rather than to the degree of preferential solute transport in macropores.

468 Our results indicate that not only the transport velocity but also the apparent dispersivity is
469 dependent on the choice of tracer. Anionic tracers exhibited larger apparent dispersivities than
470 electrically neutral ones. Moreover, our results confirm the findings of previous studies that the
471 apparent longitudinal dispersivity is positively correlated with the travel distance of the tracer.
472 We found that this relationship is refined if the normalized 5% tracer arrival time is also taken
473 into account as a measure of the degree of preferential solute transport. In particular, we found
474 that the degree of preferential solute transport increases with apparent dispersivity and decreases
475 with travel distance. A similar relationship was found between the apparent dispersivity and the
476 lateral observation scale. However, the effects of travel distance and lateral observation scale on
477 these two measures are difficult to separate as travel distance and breakthrough plane cross-
478 sectional area were positively correlated.

479 The strength of preferential transport increased at larger flow rates and water saturations, which
480 suggests that macropore flow was a dominant cause of non-equilibrium conditions for the
481 experiments in our database. Nevertheless, our data shows that heterogeneous flow in the soil
482 matrix also occasionally leads to strong preferential transport characteristics, especially in loamy
483 soils. It should also be noted here that most of the studies included in the database were
484 conducted under relatively high intensity and steady-state irrigation boundary conditions and
485 saturated or near-saturated initial conditions. Therefore, the general relevance of transport
486 processes that are triggered under different initial and/or boundary conditions cannot be
487 investigated with our database. Examples are unstable finger flow (Scheidegger, 1960; Raats,
488 1973; Hendrickx et al., 1993) and preferential transport due to soil hydrophobicity (Thomas et
489 al., 1973; Ritsema and Dekker, 1996) or air-entrapment (Debacker, 1967; Sněhota et al., 2008).
490 These flow and transport phenomena have been frequently investigated, but mostly with aid of
491 dye tracers and only occasionally by means of BTC experiments. The lack of appropriate studies

492 to quantify the importance of these preferential transport processes as compared to the here
493 investigated BTC experiments should be addressed in the future.

494 Preferential solute transport was shown to depend on soil texture in a threshold-like manner:
495 moderate to strong preferential transport was only found in soils with a texture consisting of
496 more than 8 to 9% clay. As expected, columns filled with glass beads, clean sands, or sieved soil
497 exhibited no preferential transport. No clear effect of land use on the pattern of solute transport
498 could be discerned. However, we suspect that the dataset was too small and also too strongly
499 influenced by cross-correlations with soil type and experimental conditions to allow any firm
500 conclusions to be drawn on this.

501 The database opens up the possibility to develop pedotransfer functions for solute transport
502 properties in soil. Whilst they are generally encouraging, the results of the initial analyses
503 presented in this paper suggest that this will be a challenging task. In particular, it will be
504 critically important to distinguish the effects of experimental conditions (column dimensions,
505 initial and boundary conditions) from the effects of soil and site characteristics. Some initial
506 attempts in this direction are underway.

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Table 1: Primary source publication and other information on the BTC experiments collected in the meta-database.

primary reference	# of BTCs	tracer	PDF estimated from	median R ²	type of soil or porous medium	USDA texture class	undist. sample?	land use
Akhtar <i>et al.</i> , 2003	9	chloride	MIM and CDE param.	0.98	lamellic hapludalf [†] , glossaquic hapludalf [†] , fluventic eutrudept [†] , glossic hapludalf [†] , typic fragiudept [†]	loamy sand, loam, silt loam	yes	unknown
Anamosa <i>et al.</i> , 1990	6	tritium	MIM param.	unknown	typic gibbsiorthox [†]	unknown	yes	arable
Bedmar <i>et al.</i> , 2008	6	bromide	MIM param.	0.97	unknown	silt loam, silty clay loam	yes	arable
Bromly and Hinz, 2004	14	lissamine FF	MIM param.	unknown	clean sand	sand	no	irrelevant
Candela <i>et al.</i> , 2007	7	bromide	CDE param.	unknown	typic xerorthent [†]	silt loam	no	unknown
Coats and Smith, 1964	2	calcium	MIM param.	unknown	alundum	unknown	no	irrelevant
Comegna <i>et al.</i> , 1999	3	chloride	CDE param. and raw data	unknown	entisol [†] , vertisol [†] , andosol [†]	sand, clay loam, sandy loam	yes	arable
Comegna <i>et al.</i> , 2001	17	chloride	CDE param.	0.996	orchard, arable	silt loam, silty clay loam	yes	unknown
de Smedt and Wierenga, 1984	13	chloride	MIM and CDE param.	unknown	glassbeads	sand	no	irrelevant
Doussset <i>et al.</i> , 2004	6	bromide	raw data	0.99	gleyic luvisol [†]	silty clay loam	yes, no	grass ley
Dufey <i>et al.</i> , 1982	10	chloride	CDE param.	unknown	unknown	sandy loam	no	unknown
Dyson and White, 1987	1	chloride	raw data	0.999	calcaric cambisol [†]	sandy clay loam	yes	managed grassland
Dyson and White, 1989	17	chloride	raw data	0.999	calcaric cambisol [†]	sandy clay loam	yes	managed grassland
Elrick and French, 1966	2	chloride	CDE param.	unknown	unknown	loam, silt loam	yes, no	unknown
Ersahin <i>et al.</i> , 2002	12	bromide	MIM param.	0.988	mollic planosol [†]	silt loam	yes	natural grassland
Gaber <i>et al.</i> , 1995	4	tritium	MIM param.	0.98	typic haploboroll [†]	silty clay loam	yes	unknown
Garré <i>et al.</i> , 2010	2	chloride	raw data	0.996	orthic luvisol [†]	silt loam	yes	arable
Gaston <i>et al.</i> , 2007	4	bromide	MIM param.	unknown	thermic ochraqualf [†]	clay loam	yes	arable
Gaston and Locke, 1996	4	bromide	MIM param.	unknown	thermic ochraqualf [†]	clay loam	yes	arable
Gaston and Locke, 2000	4	bromide	MIM param.	unknown	thermic ochraqualf [†]	clay loam	yes	arable
Goncalves <i>et al.</i> , 2001	16	chloride	MIM param.	0.992	dystric fluvisol [†] , calcic vertisol [†] , calcaric cambisol [†] , vertic luvisol [†]	loam, clay, clay loam, sandy clay loam, sandy loam, sandy clay	yes	arable, orchard
Gwo <i>et al.</i> , 1995	3	bromide	MIM and CDE param.	unknown	unknown	unknown	yes	forest
Haws <i>et al.</i> , 2004	5	bromide	raw data	0.999	mesic typic endoquoll [†]	silt loam	yes	arable
Helmke <i>et al.</i> , 2005	24	bromide PFBA, PIPES	MIM and CDE param.	unknown	typic hapludoll [†] , typic hapludalf [†]	loam, clay loam, sandy loam	yes	irrelevant
Jacobsen <i>et al.</i> , 1992	10	tritium, chloride	MIM param.	0.99	orthic haplohumod [†]	loamy sand	yes	unknown
Javaux and Vanclooster, 2003	9	chloride	CDE param.	unknown	unconsolidated bedrock	sand	yes	irrelevant
Jensen <i>et al.</i> , 1996	19	chloride	MIM param.	0.998	unknown	sandy loam	yes	arable
Jensen <i>et al.</i> , 1998	2	tritium	raw data	0.995	aeric glossaqualf [†]	sandy loam	yes	arable
Jorgensen <i>et al.</i> , 2004	4	bromide	MIM param.	unknown	unknown	sandy loam, sandy clay loam	yes	arable
Kamra <i>et al.</i> , 2001	45	bromide	MIM and CDE param.	unknown	unknown	sandy loam,	yes	arable, forest
Kasteel <i>et al.</i> , 2000	1	bromide	MIM param.	unknown	orthic luvisol [†]	silt loam	yes	arable
Kim <i>et al.</i> , 2007	7	bromide	MIM and	0.999	unknown	unknown	no	unknown

			CDE param.					
<i>Kjaergaard et al., 2004</i>	33	tritium	raw data	0.992	stagnic luvisol [†]	sandy loam, sandy clay loam, clay	yes	arable
<i>Koestel et al., 2009a</i>	4	chloride	CDE param.	0.99	gleyic cambisol [†]	loamy sand	yes	Arable
<i>Krupp and Elrick, 1968</i>	5	chloride	MIM and CDE param.	<i>unknown</i>	glassbeads	sand	no	irrelevant
<i>Langner et al., 1999</i>	18	PFBA	MIM and CDE param.	0.988	typic haploboroll [‡]	<i>unknown</i>	yes	managed grassland
<i>Lee et al., 2000</i>	3	chloride	MIM param.	0.999	stagnosol [†]	<i>unknown</i>	yes	arable
<i>Lee et al., 2001</i>	4	bromide	raw data	0.998	stagnosol [†]	loam	yes	arable
<i>Lennartz et al., 2008</i>	3	bromide	raw data	0.972	rendzik leptosol [†]	silt loam	yes	arable
<i>Luo et al., 2010</i>	8	bromide	raw data	0.998	mesic typic hapludalf [‡]	silt loam	yes	arable, managed grassland
<i>Maraqa et al., 1997</i>	33	tritium	MIM and CDE param.	<i>unknown</i>	typic udipsamment [‡] , entic haplaquod [‡]	sand	no	<i>unknown</i>
<i>Mayes et al., 2003</i>	28	bromide PFBA, PIPES	MIM and CDE param.	0.97	acidic Inceptisol [‡] , unconsolidated bedrock	silt loam, sandy loam	yes	<i>unknown</i> , forest
<i>McIntosh et al., 1999</i>	4	bromide, chloride	raw data	0.993	thermic typic dystrochrept [‡] , thermic kanhapludult [‡]	sandy clay loam sandy loam	yes	forest
<i>Montoya et al., 2006</i>	23	bromide	raw data	0.992	typic argiudoll [‡]	clay loam, loam	yes	arable
<i>Mooney and Morris, 2008</i>	3	chloride	raw data	0.989	gleyic luvisol [†] , cambisol [†] , gleysol [†]	sandy loam, clay loam, clay	yes	arable
<i>Nkedi-Kizza et al., 1983</i>	34	tritium, chloride	MIM param.	<i>unknown</i>	oxisol [†] (sieved aggregates)	sandy loam	no	irrelevant
<i>Oliver and Smettem, 2003</i>	13	bromide	MIM and CDE param.	<i>unknown</i>	typic xeric psamment [‡]	<i>unknown</i>	no	<i>unknown</i>
<i>Pang et al., 2008</i>	16	bromide	MIM param.	0.975	typic dystrodept [‡] , aeric fragiaquept [‡] , fluventic eutrodept [‡] , typic udipsamment [‡] , typic udivitrand [‡] , typic hapludand [‡]	clay, silty clay, silt loam, sand, sandy loam, loam	yes	managed grassland
<i>Perfect et al., 2002</i>	2	chloride	raw data	0.998	typic udifluvent [‡] , vertic endoaquept [‡]	<i>unknown</i>	yes	managed grassland
<i>Pot et al., 2005</i>	4	bromide	MIM param.	0.988	stagnosol [†]	silt loam	yes	managed grassland
<i>Poulsen et al., 2006</i>	33	tritium	MIM param.	<i>unknown</i>	typic hapludalf [‡]	sandy loam	yes	arable
<i>Prado et al., 2006</i>	3	deuterium	CDE param.	<i>unknown</i>	pachic andosol [†]	silt loam	no	arable
<i>Prado et al., 2009</i>	9	deuterium	MIM and CDE param.	<i>unknown</i>	pachic andosol [†]	silt loam	yes	arable
<i>Raturi et al., 2001</i>	6	bromide	MIM param.	0.99	antroposol [†]	loamy sand	yes	managed grassland
<i>Ren et al., 1996</i>	20	bromide	MIM param.	0.99	durixerollic calciorthid [‡]	silt loam	yes	arable
<i>Reungsang et al., 2001</i>	12	bromide	MIM param.	<i>unknown</i>	typic haploaquoll [‡] , cumulic haploaquoll [‡]	sandy loam	yes	managed grassland, arable
<i>OScherr, 2009</i>	2	bromide	MIM param.	0.983	<i>unknown</i>	silty clay loam	yes	managed grassland
<i>Schoen et al., 1999</i>	3	bromide, chloride, deuterium	MIM param.	<i>unknown</i>	<i>unknown</i>	silt loam	yes	arable
<i>Schulin et al., 1987</i>	23	tritium, bromide	MIM param.	<i>unknown</i>	rendzik leptosol [†]	loam	yes	forest
<i>Segal et al., 2009</i>	1	bromide	MIM param.	<i>unknown</i>	<i>unknown</i>	<i>unknown</i>	yes	arable
<i>Selim and Amacher, 1988</i>	3	tritium	MIM param.	<i>unknown</i>	arguic fragiudalf [‡] , typic hapludalf [‡] , typic udipsamment [‡]	<i>unknown</i>	no	<i>unknown</i>

<i>Seo and Lee, 2005</i>	3	chloride	MIM param.	<i>unknown</i>	typic hapludult [†]	sandy loam	yes	<i>unknown</i>
<i>Seuntjens et al., 2001</i>	2	chloride	MIM param.	0.99	podsol [†]	sand	yes	heathland
<i>Seyfried and Rao, 1987</i>	14	tritium	MIM and CDE param.	<i>unknown</i>	typic distropept [†]	<i>unknown</i>	yes	arable, orchard
<i>Shaw et al., 2000</i>	13	bromide	MIM param.	<i>unknown</i>	kandiudult [†]	sand, sandy loam, loamy sand, sandy clay loam	yes	arable
<i>Singh and Kanwar, 1991</i>	6	chloride	raw data	0.997	mesic hapludoll [†]	<i>unknown</i>	yes	arable
<i>Smettem et al., 1983</i>	3	tritium	raw data	0.973	<i>unknown</i>	clay loam	yes	arable
<i>Smettem, 1984</i>	12	tritium	MIM and CDE param.	<i>unknown</i>	'well structured brown calcareous earth'	silt loam	yes	forest
<i>Stagnitti et al., 2000</i>	1	chloride	MIM param.	<i>unknown</i>	<i>unknown</i>	<i>unknown</i>	yes	managed grassland
<i>Tyler and Thomas, 1981</i>	1	chloride	raw data	0.981	fluventic haplodoll [†] , typic udifluent [†] , vertic haplaquept [†]	silt loam, silty clay loam, sandy loam	yes	arable
<i>Unold et al., 2009</i>	4	chloride	raw data	0.996	orthic luvisol [†] , glycic cambisol [†]	silt loam, sandy loam	yes	arable
<i>Vanderborcht et al., 2002</i>	2	chloride	MIM param.	<i>unknown</i>	stagnic cambisol [†]	clay loam	yes	forest
<i>Vervoort et al., 1999</i>	7	bromide, chloride	MIM param.	<i>unknown</i>	typic kandiudult [†]	sandy loam, sandy clay loam, clay, sandy clay	yes	managed grassland
<i>Vincent et al., 2007</i>	8	bromide	raw data	0.994	stagnosol [†]	loam, silt loam	yes	arable, managed grassland, forest
<i>Vogeler et al., 2006</i>	12	bromide, chloride	CDE param.	<i>unknown</i>	stagnic luvisol [†]	sandy loam	yes	arable
<i>Wilson et al., 1998</i>	2	bromide	raw data	0.972	typic paleudalf [†]	silt loam	yes	arable
<i>Zurmühl, 1998</i>	2	tritium	MIM param.	<i>unknown</i>	<i>unknown</i>	sand	yes	forest

773 [†] Classification according to the World Reference Base (WRB).

774 [‡] Classification according to the system of the United States Department of Agriculture (USDA).

775 **Table 2: Inventory of the data available in the database.**

Data	Available	Missing
Explicit information on water content, θ (cm ³ cm ⁻³)	487	246
Explicit information on water flux, q (cm d ⁻¹)	551	182
Travel distance, L (cm)	733	0
Area of breakthrough plane, A (cm ²)	733	0
Information on tracer detection method	733	0
Information on initial conditions	731	2
Pressure head at upper boundary, h_{UB} (cm)	333	400
Pressure head at lower boundary, h_{LB} (cm)	429	304
Average pressure head, h_{ave} (cm)	466	267
Hydraulic gradient, dH/L (-)	406	327
Information on Irrigation device	708	25
Information on outlet construction	694	39

Information on tracer	733	0	776
Information on tracer application method	733	0	
BTC raw data	146	587	
Information on land use	635	98	
Information on cropping	454	279	
Information on soil management practices	388	345	
Depth from which soil sample was collected (cm)	508	225	
Texture data	618	115	
Bulk density, ρ (g cm^{-3})	605	128	
Organic carbon content, OC (-)	488	245	
Porosity, ϕ ($\text{cm}^3 \text{cm}^{-3}$)	611	122	

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Table 3: Land use and soil management for the 733 datasets in the database.

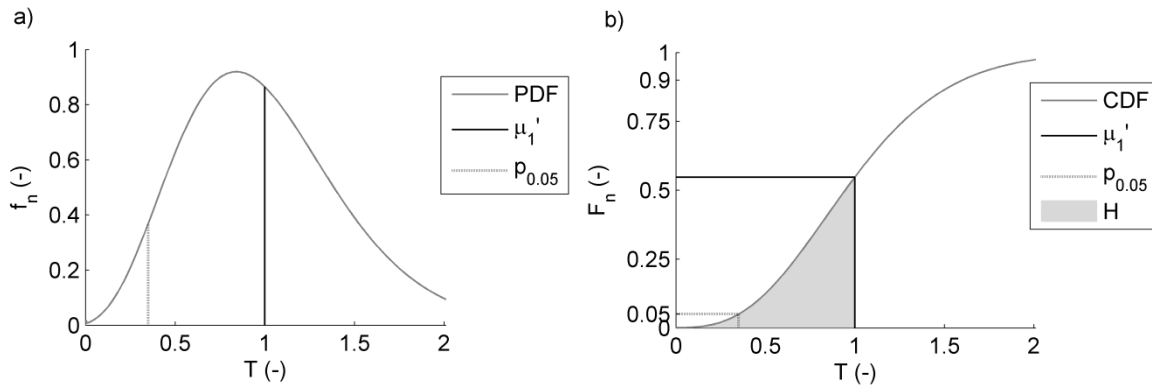
Land use	# of entries in the database
arable (all)	302
arable (conventional tillage)	219
arable (reduced tillage)	6
arable (no tillage)	31
arable (no further information)	46
forest	79
managed grassland	92
natural grassland	12
grass ley	7
heathland	2
orchard	19
unknown land use	98
sieved and repacked samples [†]	116
unconsolidated bedrock	60
clean sand or glass beads	32

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[†]note that for some of the sieved samples the land use was known

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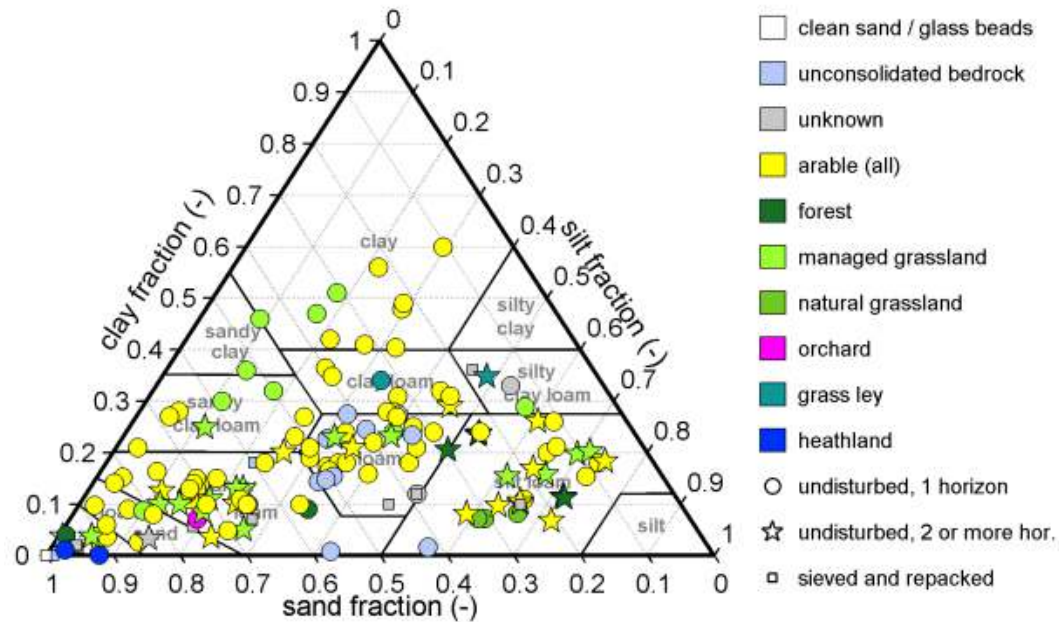


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Figure 1: The PDF (a) and CDF (b) of an example BTC taken from *Garré et al. (2010)* illustrating how the normalized first temporal moment, μ_1' , the normalized 5%-arrival time, $p_{0.05}$, and the holdback, H , are derived.



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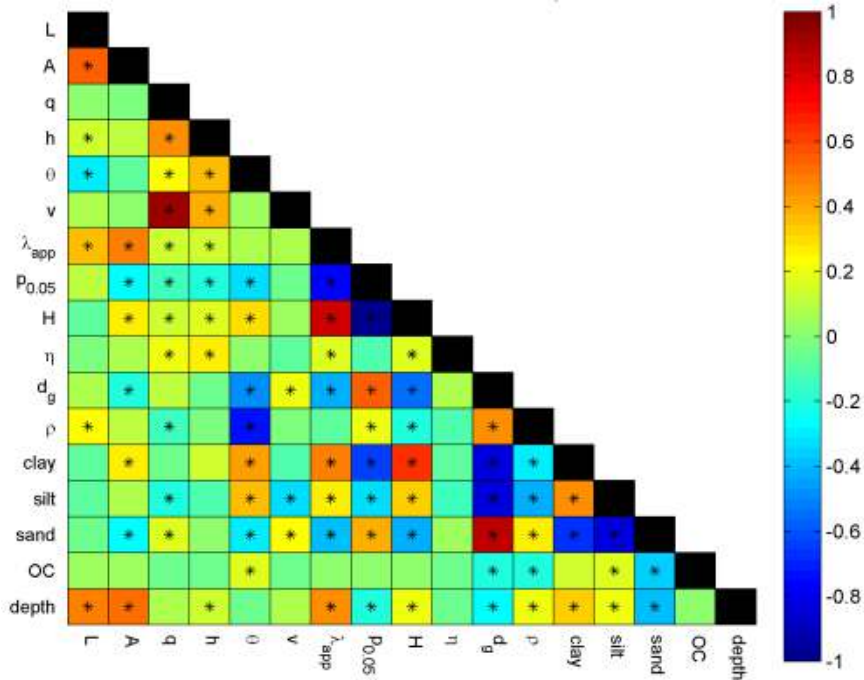
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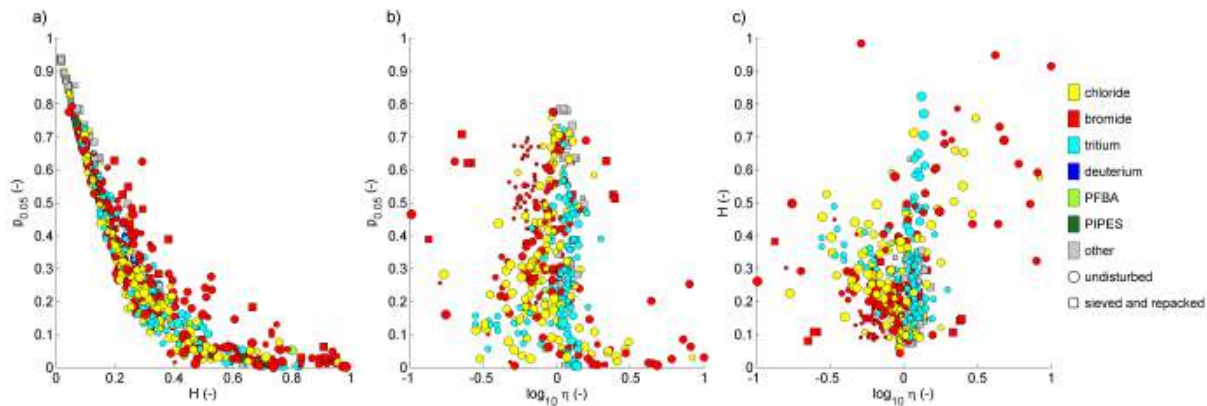
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Figure 2: Land uses corresponding to the soil samples on which the 733 considered BTC experiment had been carried out. Note that in most publications only average values are published for several soil samples and that several experiments are often conducted on one and the same soil sample under different hydraulic boundary conditions. Therefore, the number of datasets visible in the texture triangle is less than 733.



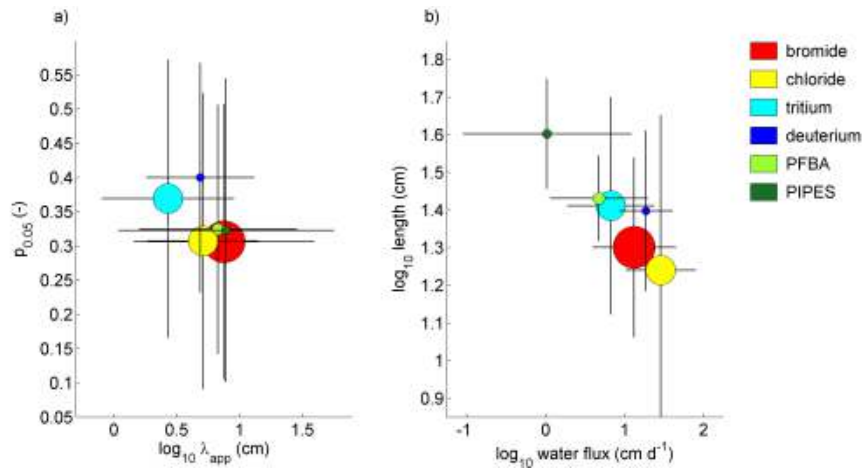
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Figure 3: Spearman rank correlation coefficients between various BTC-shape measures and soil and site as well as experimental properties. The boxes marked by an asterisk indicate significant correlations with p-values of smaller than 0.001. The correlations were carried out for the travel distance, L , the area of the breakthrough plane, A , the water flux, q , the suction head, h , the water content, θ , the transport velocity, v , the apparent dispersivity, λ_{app} , the normalized 5%-arrival time, $p_{0.05}$, the holdback, H , the piston-flow to transport velocity ratio, η , the geometric mean grain diameter, d_g , the soil bulk density, ρ , the clay fraction, *clay*, the silt fraction, *silt*, the sand fraction, *sand*, the organic carbon content, *OC*, the average sampling depth, and *depth*.



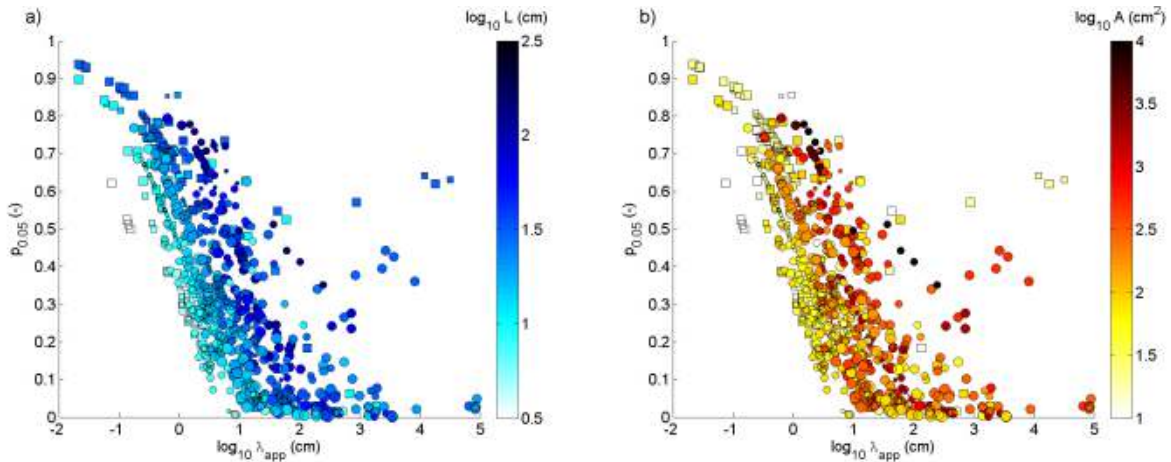
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Figure 4: Comparison between the shape-measures related to early tracer arrival: a) comparison between the holdback, H , and the normalized 5%-arrival time, $p_{0.05}$; b) comparison of the piston-flow to transport velocity ratio, η , and the normalized 5%-arrival time, $p_{0.05}$; c) comparison of the piston-flow to transport velocity ratio, η , and the holdback, H . In addition, the type of applied tracer is depicted. The symbol size corresponds to the water fluxes, q , under which the experiment was conducted, small symbols indicating small water fluxes, large symbols denoting large water fluxes.



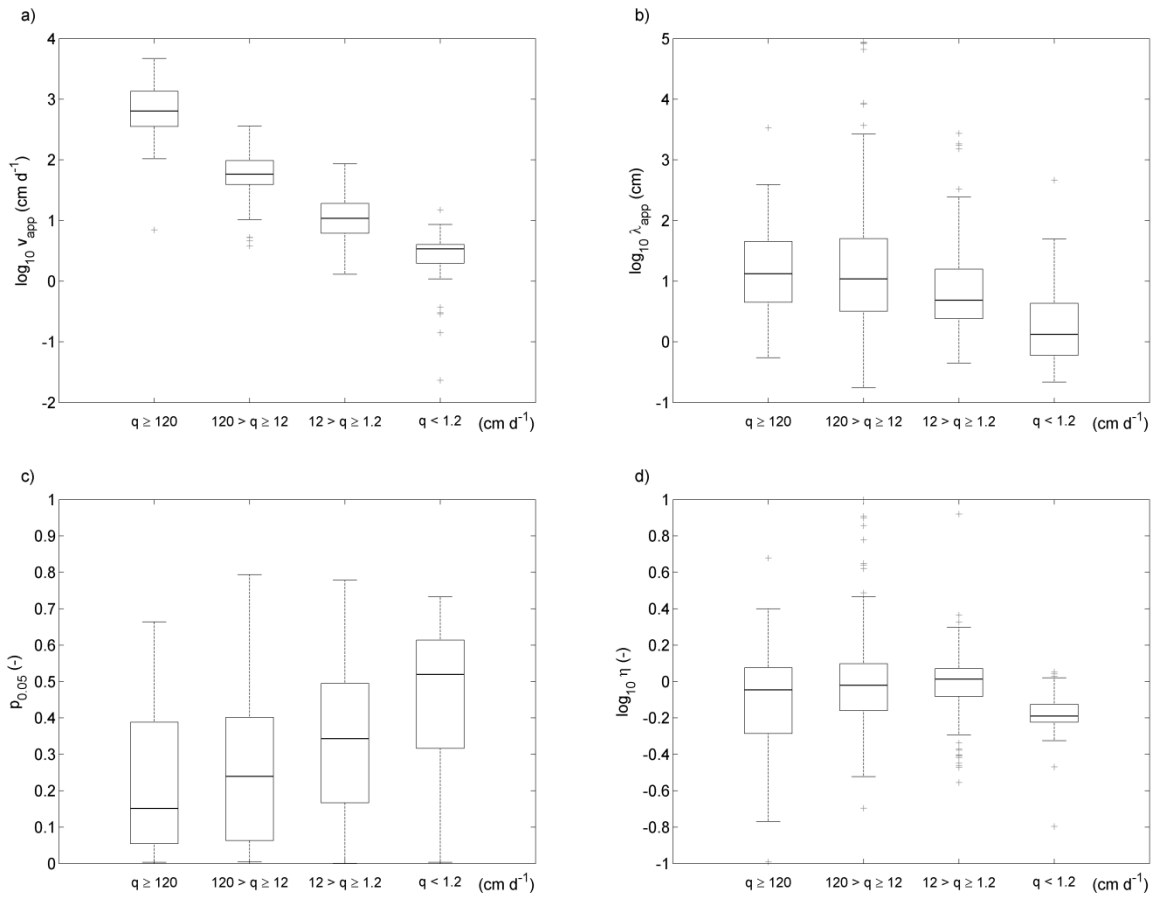
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Figure 5: The median apparent dispersivity, λ_{app} , and normalized 5%-arrival time, $p_{0.05}$, in dependence of the applied tracer (a) and the corresponding median experimental conditions (b). The center of each circle depicts the respective median value and the error bounds indicate the corresponding interquartile range. The size of each circle corresponds to the number of BTC conducted with the respective tracer.



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Figure 6: Comparison of the apparent dispersivity, λ_{app} , and normalized 5%-arrival time, $p_{0.05}$, with (a) the travel distance, L , and (b) the area of the breakthrough plane, A . The symbol size corresponds to the water fluxes, q , under which the respective experiment was conducted, small symbols indicating small water fluxes, large symbols denoting large water fluxes. The meaning of the symbol shape is explained in Figure 4.



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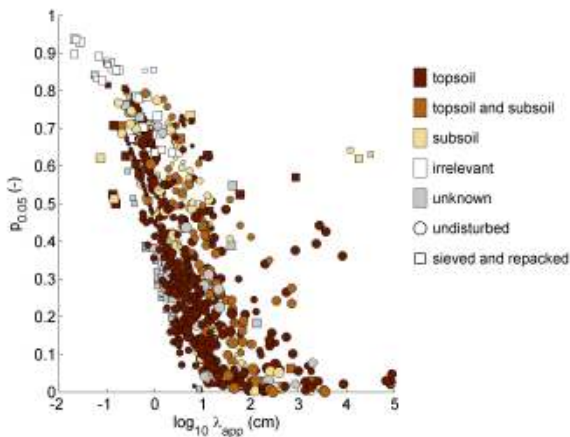
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Figure 7: Boxplots (a) transport velocity, v , (b) apparent dispersivity, λ_{app} , (c) normalized 5%-arrival time, $p_{0.05}$, and (d) piston-flow to transport velocity ratio, η according to the respective water flux class. Note that this figure is based on BTCs from undisturbed soil samples, only.



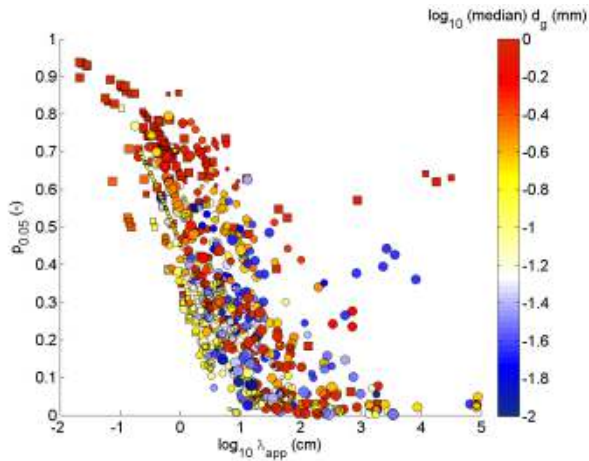
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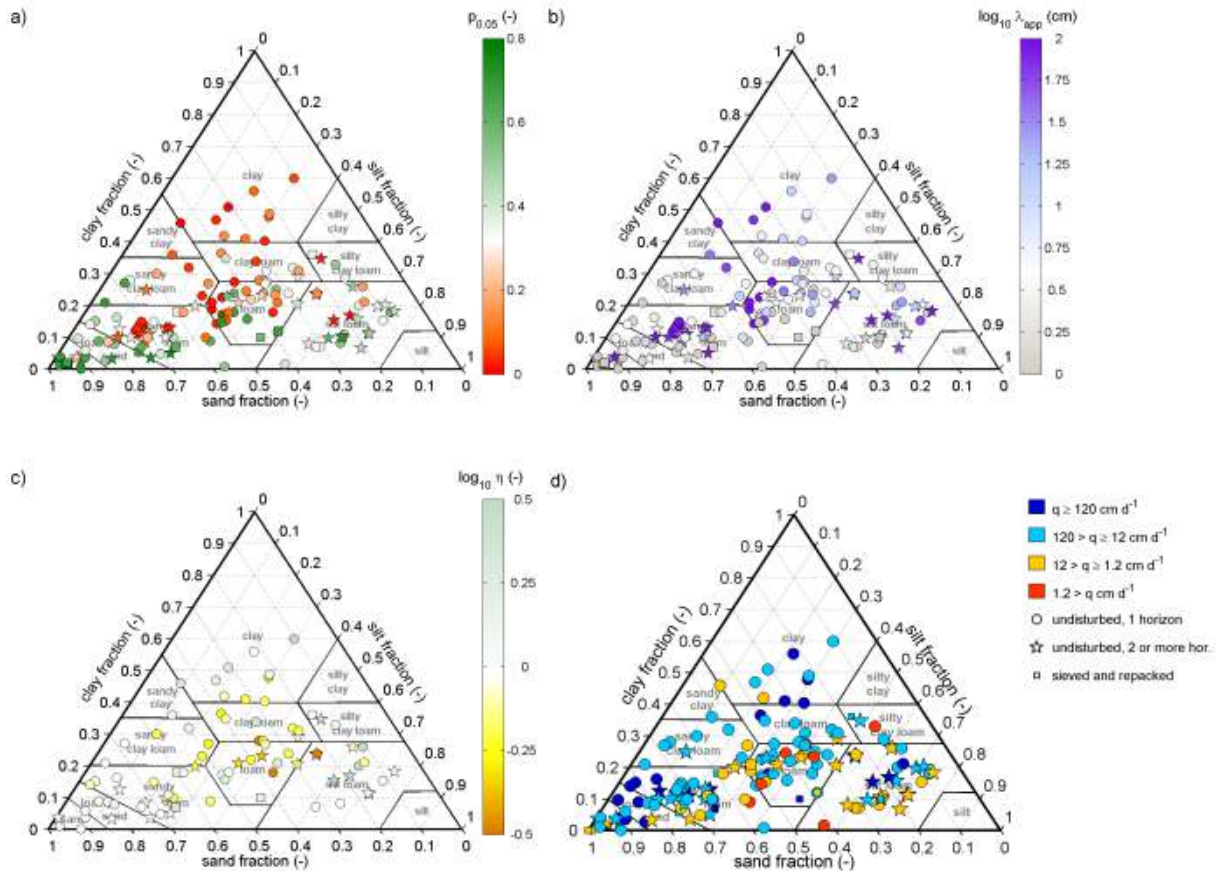
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Figure 8: Comparison of the apparent dispersivity, λ_{app} , and normalized 5%-arrival time, $p_{0.05}$, with sampling location of the respective soil sample. The symbol size corresponds to the water fluxes, q , under which the respective experiment was conducted, small symbols indicating small water fluxes, large symbols denoting large water fluxes.



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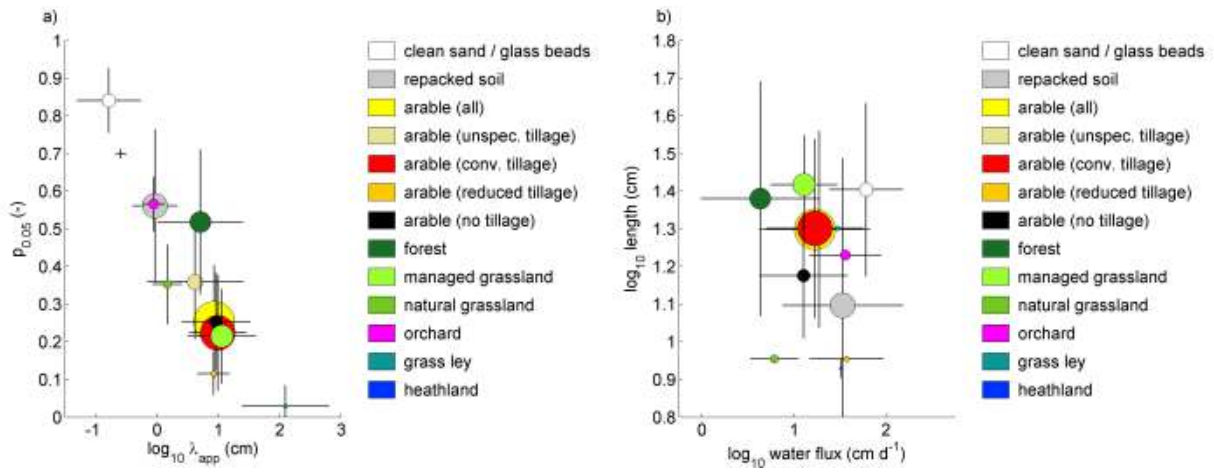
824 Figure 9: Comparison of the apparent dispersivity, λ_{app} , and normalized 5%-arrival time, $p_{0.05}$, with the geometric mean grain
 825 diameter, d_g , of the respective soil sample. The symbol size corresponds to the water fluxes, q , under which the respective
 826 experiment was conducted, small symbols indicating small water fluxes, large symbols denoting large water fluxes. The
 827 meaning of the symbol shape is explained in Figure 4.



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830 Figure 10: The (a) normalized 5%-arrival time, $p_{0.05}$; (b) apparent dispersivity, λ_{app} ; (c) the piston-flow to transport velocity
 831 ratio, η , and (d) the water flux classes corresponding to the considered BTC experiments.



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Figure 11: a) Comparison of the apparent dispersivity, λ_{app} , and normalized 5%-arrival time, $p_{0.05}$, with the respective land use; b) comparison of the water flux, q , and column length, L , with the respective land use. The center of each circle depicts the respective median value and the error bounds indicate the corresponding interquartile range. The size of each circle corresponds to the number of samples within each land use class.