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Anomalous transport and upscaling in critically-connected fracture networks under stress conditions

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Abstract

Anomalous transport in fractured rocks is of high importance for numerous research fields and applications in hydrogeology and has been widely studied in the last decades. This phenomenon is due to the structural heterogeneities of fractured rocks and the high contrast between the fractures and matrix properties. At the same time, the fracture properties in terms of fracture density and geometrical characteristics are related to in-situ stress conditions that impact the connectivity of the system and the distributions of fracture aperture and length, including the creation of new cracks when the differential stress conditions are strong enough. In order to understand how all these features impact the observed anomalous transport, we study the role of in-situ stress in mass transport through a two-dimensional fractured rock. The fracture network is based on a real outcrop with a connectivity state around the percolation threshold, over which we simulate stress-dependent fracture deformation and propagation, and perform hydrodynamic transport through the deformed rock mass. The impact of the changes in aperture and the creation of new cracks on transport behavior is evaluated for various stress scenarios and reproduced with upscaled representations of transport processes. We also consider matrix diffusion as a source of anomalous transport and demonstrate that this process can be incorporated into the proposed upscaled

models. Results showed that traditional upscaling methods with space-Lagrangian velocities description capture well the Gaussian-like breakthrough curves (BTCs) under low-stress ratio conditions, but fail under high-stress ratio conditions with multiple-peak early-times BTCs. To characterize the emergence of strong anomalous transport, we extend the Random Walk model Directed by a Markov Process with space-Lagrangian velocity sampling by incorporating multiple transition matrices conditioned by different initial velocity states. This new transport upscaling model shows high accuracy in predicting complex transport behaviors in critically connected fracture networks.

Keywords: anomalous transport, natural fracture network, in-situ stress, fracture network topology, particle tracking, upscaling

1 Highlights

- High-stress conditions lead to multiple-peak early-times anomalous transport behavior
- Traditional upscaling methods fail to reproduce this anomalous transport behavior
- A new upscaling method succeeds in reproducing fracture networks anomalous transport
- This method reproduces high-stress and matrix diffusion anomalous transport behavior

1. Introduction

Flow and solute transport in heterogeneous fracture systems have become an important research subject in the past decades as the accurate characterization and prediction of flow and transport behaviors are involved in various engineering applications, such as groundwater remediation, geological storage of carbon dioxide, geothermal development,

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12 and nuclear waste disposal [e.g., Tran et al., 2021, Yoo et al., 2021, Wang et al., 2023, Liu
13 et al., 2022]. However, due to the fact that fractures exist at all scales, it seems impossible to
14 accurately capture the transport behaviors based on the fine description of discrete fracture
15 network (DFN) modeling for all scenarios of fracture systems, especially at a large scale.
16 The potential uncertainty in the characterization of fractured media greatly challenges the
17 practical value of DFN modeling [e.g., Demirel et al., 2019, Dorn et al., 2012, Zhou et al.,
18 2021, Gattinoni and Scesi, 2018]. Therefore, developing upscaling methods, that are able
19 to parameterize the transport model with a few characteristic parameters and be used for
20 multi-scale prediction, has attracted increasing interest.

21 For the non-reactive solute (i.e., tracers) at the Darcy scale, the advection-dispersion
22 equation (ADE) is the traditional approach to characterize the flow and transport behaviors
23 when the breakthrough curves (BTC) exhibit a diffusion-like (Fickian) process [Bear, 2013].
24 However, the traditional approach is not always valid due to the emergence of anomalous
25 (non-Fickian) transport with the BTCs characterized by early arrivals and long tails. The
26 anomalous transport behaviors have been observed at various scales, from pore to single
27 fracture and rock core column, and finally to field scale [e.g., Cortis and Berkowitz, 2004,
28 Kang et al., 2015b, 2016, Gouze et al., 2020, 2023]. Therefore, some alternative models
29 have been developed to address this issue [e.g., Berkowitz, 2002, Neuman, 2005, Noetinger
30 et al., 2016]. Methods relying on the displacement of particles are particularly well suited
31 for this matter. It includes random walk (RW) simulations that sample from fitted velocity
32 distributions to reproduce the particle spreading pattern in two-dimensional DFNs, contin-
33 uous time-random walk (CTRW) formalism that provides semi-analytical results for plume
34 evolution, and Markov-directed random walk (MDRW) method that considers the necessity
35 of sequential correlation along the trajectory [Berkowitz and Scher, 1997, 1998, Painter and
36 Cvetkovic, 2005, Smith and Schwartz, 1984]. More recently, a spatial Markov model has also
37 been used to develop a correlated CTRW model, which successfully characterizes the impact

38 of velocity distribution and correlation on transport in theoretical and field-scale fractured
39 media [Kang et al., 2011, 2015a,b]. In addition, some alternative models are developed to fit
40 BTCs by modifying the traditional advection-dispersion equation such as Mobile-Immobile
41 (MI) model, Multiple Rate (MR) model and Multiple Region Advection-Dispersion (MRAD)
42 model, which have shown a good ability to reproduce breakthrough curves obtained from the
43 laboratory experiments [Majdalani et al., 2015, Wang et al., 2020] or numerical simulations
44 [Davit et al., 2010, Cherblanc et al., 2007] via a proper parameter tuning. In addition to
45 these alternative transport models, the importance of heterogeneities and uncertainties in
46 stochastic quantification and modeling has been underlined in several studies [Riva et al.,
47 2008, Schiavo, 2023].

48 Although the existing transport models successfully predict anomalous transport through
49 DFNs, the rock heterogeneity in most of those works is generated randomly, both in terms
50 of the geometric topology of the fracture network and the aperture distribution [Kang et al.,
51 2015a, 2017, Painter and Cvetkovic, 2005]. In reality, subsurface fractures are inevitably
52 subject to in-situ ground stress which is the essential factor in determining fracture aperture.
53 Previous studies regarding the mechanical simulations of fractured rocks have indicated that
54 the presence of fracture networks may lead to a non-uniform distribution of local stresses,
55 which in turn produces various geomechanical responses to fractures, such as varying degrees
56 of normal closure, shear sliding, dilatancy and crack propagation [Baghbanan and Jing, 2008,
57 Lei and Gao, 2018, Liu et al., 2018, Min et al., 2004]. Consequently, hydraulic properties (i.e.,
58 flow distribution and bulk permeability) of the stress-controlled fractured rocks are affected,
59 where the shear dilation under an anisotropic stress condition may lead to highly localized
60 flows and increase the bulk permeability [Lei et al., 2017a, Min et al., 2004]. Recently,
61 our study further shows that crack propagation induced by shearing may expand the flow
62 backbone network and reduce the flow localization, and then significantly enhance the bulk
63 permeability [Jiang et al., 2019]. In other studies, the impact of in-situ stress on anomalous

64 transport is investigated for single in-situ stress scenarios through studies that either focus
65 on the importance of fracture growth [Nick et al., 2011] or consider dense fracture networks
66 at the representative elementary volume scale [Rutqvist et al., 2013, Zhao et al., 2013].
67 Systematic analysis of various in-situ stress regimes is conducted in 2D and 3D fracture
68 networks [Kang et al., 2019, Sweeney and Hyman, 2020, Zhao et al., 2011], whose transport
69 properties are upscaled with a new dual-correlation-length CTRW approach in Kang et al.
70 [2019] without considering the impact of matrix diffusion. The latter studies are based
71 on the assumption of no stress-driven crack growth that holds in well-connected fracture
72 networks, for which in-situ stress loading mainly causes the deformation of pre-existing
73 fractures. However, this assumption is questionable when working with fracture networks
74 whose connectivity is close to the percolation threshold, which is often observed in realistic
75 domains [Lei and Wang, 2016, Renshaw, 1997].

76 In order to fully explore the impact of in-situ stress on anomalous transport in realistic
77 fracture networks, we apply various stress conditions on critically-connected fracture net-
78 works that are based on a real outcrop. The role of stress-driven crack growth is considered
79 and the effects of the stress orientation and magnitude on anomalous transport behavior are
80 analyzed. The results presented in this work show the need for a new upscaling method that
81 is provided to represent the observed anomalous transport behavior compared to existing
82 models. The impact of matrix diffusion on anomalous transport in the considered stress
83 scenarios is also illustrated and added to the proposed upscaling method.

84 This paper is organized as follows. The methodology of this work, including the consid-
85 ered fracture network, the geomechanical, fluid flow and transport models, and the upscal-
86 ing method, is described in Section 2. The results of the numerical simulations in terms of
87 anomalous transport with and without the effect of stress-driven new cracks are presented
88 with the comparison of different upscaling models in Section 3. Finally, discussion and
89 conclusions are provided in Section 4.

90 2. Methodology

91 2.1. Fracture network and geomechanical models

92 We consider the well-known outcrop collected at the Hornelen Basin in Norway [Odling,
93 1997] that corresponds to a critically connected network [Bour and Davy, 1997, Lei and
94 Gao, 2018] characterized by three major fracture sets of orientation 5° , 50° , and 120° (Figure
95 1). The geometric connectivity of a fracture network can be characterized by a statistical
96 parameter, i.e, the percolation parameter, $p = \frac{1}{L^2} \sum l^2$, where L is the modeling domain
97 length and l is the length of individual fractures. A higher p means a more connected fracture
98 network and the network is statistically connected once p exceeds the percolation threshold
99 p_c , which may have a scale-independent value around 5.8 [Bour and Davy, 1997]. Here,
100 the studied fracture network has a p of 5.6, which is critically connected. The fractured
101 rock is a Devonian age sandstone consisting of more than 2000 joints, which are mostly
102 perpendicular to the bedding plane. The readers can refer to numerous existing studies
103 to gather information about the geometrical distribution of the fractures [Berkowitz et al.,
104 2000, Bour, 2002] and the system behaviors in terms of stress variability [Lei and Gao, 2019],
105 fluid flow [Jiang et al., 2019, Odling, 2001], and thermal transport [Geiger and Emmanuel,
106 2010].

107 For the geomechanical modeling, we use the same methodology as in geomechanical work
108 previously conducted on this 2D fracture network [Lei and Gao, 2019, Jiang et al., 2019]
109 where the relevant equations and settings of material properties, as well as the implemen-
110 tation, can be found in detail. Here, we consider the stress loading directions $\theta = 50^\circ$,
111 80° , 120° , and 150° (Figure 1a), which are denoted *Config1*, *Config2*, *Config3*, and *Config4*,
112 respectively. The stress loading directions are aligned with one of the major fracture sets
113 for *Config1* and *Config3*, and in between the main fracture set orientations for *Config2* and
114 *Config4*. We also consider various stress conditions that have been defined in Jiang et al.
115 [2019]. These conditions are reported in Table 1 and applied to each configuration.

	<i>Iso1</i>	<i>Iso2</i>	<i>Anis1</i>	<i>Anis2</i>	<i>Anis3</i>
S_{\min}	0	5	5	5	5
S_{\max}	0	5	10	15	20

Table 1: Isotropic (*Iso1* and *Iso2*) and anisotropic (*Anis1*, *Anis2*, and *Anis3*) stress scenarios considered with S_{\max} the strength applied in the loading direction and S_{\min} in the orthogonal direction, both expressed in MPa.

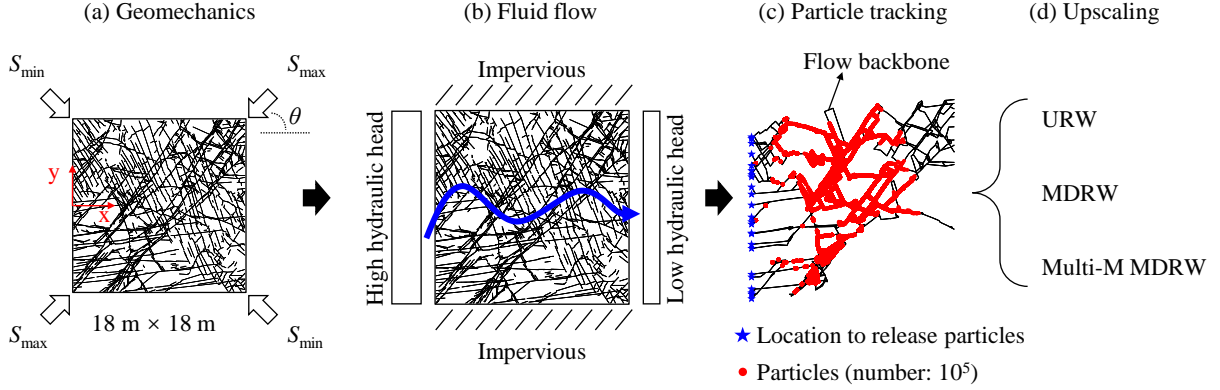


Figure 1: Modeling steps and boundary conditions for the (a) geomechanical, (b) fluid flow, (c) particle tracking, and (d) upscaling models, the latter being named Upscaled Random Walk (URW), Markov-Directed Random Walk (MDRW) and Multiple Matrices MDRW (Multi-M MDRW).

116 2.2. Hydrodynamic transport models

117 As done in numerous existing studies [e.g., Demirel et al., 2019, de Dreuzy et al., 2001,
118 Roubinet et al., 2020], the fluid flow simulation is conducted by solving laminar and incom-
119 pressible Darcy-type flow in fractured media with the fracture permeability governed by the
120 cubic law [Witherspoon et al., 1980]. Here, we assume that the rock matrix is impervious
121 to fluid flow, such that flow and advective transport only occur within the fractures while
122 diffusion is considered in the rock matrix [Zhao et al., 2011]. A constant macroscopic hy-
123 draulic gradient is imposed by setting the hydraulic head on the left ($x = 0$ m) and right
124 ($x = L$, L being the domain size set to 18 m) boundaries to 0 m and 18 m, respectively, and
125 by prescribing no-flow conditions to the top and bottom boundaries (Figure 1b). Both fluid
126 flow and particle tracking simulations are based on the flow backbone fractures where the
127 dead ends are removed (Figure 1c). The fracture flow rate between nodes i and j is defined
128 as $Q_{ij} = \delta \rho g a_{ij}^3 (h_j - h_i) / (12 \mu l_{ij})$ where δ is the thickness of the fractured rock (assumed to

129 be 1 m for convenience), a_{ij} the local aperture, h_k the hydraulic head at node k ($k = i, j$), l_{ij}
 130 the length between nodes i and j , ρ the fluid density, μ the fluid dynamic viscosity, and g the
 131 gravitational acceleration. The hydraulic head at each node is computed by imposing mass
 132 conservation and the flow velocity between nodes i and j is computed as $u_{ij} = Q_{ij} / (a_{ij} \times \delta)$.

133 Solute transport in the fracture network is modeled by injecting 10^5 particles on the left
 134 border of the domain, that migrate through the fracture network by advection until reaching
 135 the right border of the domain (Figure 1c). The effect of matrix diffusion is considered by
 136 introducing a retention time t_{diff} added to the advection time based on the assumption of
 137 infinite matrix [Painter and Cvetkovic, 2005, Roubinet et al., 2010]:

$$t_{\text{diff}} = \left(\frac{\phi_m \sqrt{D_m}}{2\alpha b} t_a \right)^2, \quad (1)$$

138 where ϕ_m is the matrix porosity, D_m the molecular diffusion coefficient, b the half aperture
 139 of the fracture, and $\alpha = \text{erfc}^{-1}(U[0, 1])$ with $U[0, 1]$ being a uniform random number drawn
 140 between 0 and 1. At fracture intersections, we apply a complete mixing rule where the
 141 probability for a particle exiting an outgoing link is proportional to the flow rate through
 142 that link. The solute spreading behavior across the entire domain is quantified by the
 143 breakthrough curve recorded at the outlet of the domain and defined as the probability
 144 density function (PDF) of particle arrival time τ_b on the right side of the domain. The
 145 results presented in this work are obtained by considering $\phi_m = 0.05$ and $D_m = 10^{-8}$ m²/s,
 146 and we verified that the same results are obtained for 10^5 and 10^6 particles, ensuring that
 147 the number of particles is high enough to correctly sample the considered domains.

148 2.3. Upscaled transport models

149 We consider existing upscaled transport models that are based on a set of spatial and
 150 temporal jumps, which are recorded during the fracture-network scale simulations. The 1D
 151 upscaled simulations are run by sampling into this database of spatial jumps until the sum of

152 these jumps exceeds the domain size L , and the predicted breakthrough time is expressed as
 153 the sum of the corresponding time steps. Among the existing upscaled transport models, we
 154 consider the Upscaled Random Walk (URW) and Markov-Directed Random Walk (MDRW)
 155 methods that are based on space-Lagrangian velocity statistics [Kang et al., 2017, 2019] and
 156 sequential velocity correlation [Painter and Cvetkovic, 2005], respectively. We also consider
 157 that the URW method can rely on a standard database (URW-std), with sampling from
 158 the segments discretized for flow and transport simulations [Painter and Cvetkovic, 2005],
 159 and on a modified database (URW-mod), which is defined to record the data by space-
 160 Lagrangian velocities with a fixed space lag $\Delta x'$ [Kang et al., 2019]. These methods are
 161 used in Section 3.3 to reproduce the transport behavior described in Sections 3.1 and 3.2
 162 considering the average length 0.05 m for the standard database sampling and $\Delta x' = 0.5$ m
 163 for the modified database.

164 As these methods fail to reproduce some of the anomalous transport behavior provided
 165 in Section 3, we suggest two improvements of MDRW method. (i) The modified database
 166 usually considered with URW method is used with the MDRW method, leading then to
 167 the distinction between MDRW-std and MDRW-mod methods using standard and modified
 168 database, respectively. Note that in both cases, the number of velocity states is $N_m = 10$
 169 for the results provided in Section 3.3. (ii) We consider that the first velocity experienced
 170 by the particles, i.e., the velocity of the fracture segment in which the particle is injected, is
 171 a determinant information for the particle trajectories. From this statement, we propose an
 172 alternative method that is called Multi-M MDRW and rely on multiple transition matrices
 173 according to the initial space-Lagrangian velocities. To do this, the initial velocity probabil-
 174 ity V defining the first velocity that is associated with the particles is recorded with V_k the
 175 probability of traveling with velocity v_k . When running the Multi-M MDRW method, the
 176 initial state of the particles is determined from V and the corresponding transition matrix
 177 A_k conditioned by the initial state v_k is used to determine the next flow velocities. After

178 assigning the initial velocity v_k from probability V , the particles move along a 1D line with
 179 the constant spatial step $\Delta x'$ and time step $\Delta t' = \Delta x'/v_k$, and the next velocities v are
 180 defined from the correlation matrix A_k at each jump, as in the existing MDRW method.
 181 Following the formulation used for MDRW method, the successive motion distribution of
 182 Multi-M MDRW is expressed as

$$f(\Delta\tau'_1, \Delta\tau'_2, \dots, \Delta\tau'_n) = \sum_{S_1 \in S} f(\Delta\tau'_1 | S_1) P_{S_1} \sum_{S_2 \in S} f(\Delta\tau'_2 | S_2) A_{S_1 S_2} \Big|_{S_1} \cdots \sum_{S_n \in S} f(\Delta\tau'_n | S_n) A_{S_{n-1} S_n} \Big|_{S_1} \quad (2)$$

183 where $f(\Delta\tau' | S)$ is the probability density function of $\Delta\tau'$ for a particle being at state S
 184 and P_{S_1} the probability of being at state S_1 .

185 The impact of diffusion in the matrix domain is added to the above framework by sam-
 186 pling the retention time from expression (1) at every sampling of space-Lagrangian velocity
 187 v . Note that the computation of t_{diff} is simplified by considering the mean aperture of the
 188 fracture network.

189 3. Results

190 3.1. Impact of stress-induced aperture changes

191 Figure 2 shows the BTCs that are obtained by running the particle-tracking method
 192 described in Section 2.2 at the fracture-network scale on the domains defined in Section 2.1.
 193 Matrix diffusion is not considered yet and will be analyzed separately in Section 3.4. The
 194 fracture aperture is initially set to the constant value of 0.1 mm and heterogeneous stress-
 195 induced aperture fields are obtained for the stress scenarios described in Table 1 by applying
 196 the geomechanical model described in Section 2.1. In order to focus on the impact of the
 197 changes in aperture, the growth of new cracks is not considered in the present section and is
 198 studied in Section 3.2. The results shown in Figure 2 are normalized by the peak arrival time

199 of the reference zero-stress configuration, denoted *Iso1* (Table 1). Furthermore, additional
 200 results are shown in Figures 3 and 4. Figure 3 shows the changes in aperture and particle
 201 trajectories due to the stress scenarios applied to the system with (i) the mean aperture
 202 of the three major fracture sets for the different stress scenarios normalized by that of the
 203 reference case with no stress (Figures 3a-d), and (ii) contribution of the three major fracture
 204 sets to transport defined as the time spent by the particles in each fracture set normalized
 205 by the time spent in all fracture sets. The contribution to transport is also normalized by
 206 that of the reference case (Figures 3e-h). Finally, Figure 4 shows the spatial distribution
 207 along the outlet border (right border) of the proportion of particles exiting the system for
 208 different anisotropic-stress cases. Note that we discuss here the impact of the different stress
 209 scenarios and orientation on the transport behavior. The reader can refer to Jiang et al.
 210 [2019] for an analysis of the fracture length and orientation and their impact on the BTCs.

211
 212 Applying the isotropic stress condition *Iso2* results in a simple delay in the arrival times of
 213 particles in comparison with the reference case *Iso1* without stress condition (Figure 2). The
 214 shapes of the curves are similar (from orange to green curves), which is consistent with the
 215 observations made in previous studies [Zhao et al., 2011, Kang et al., 2019]. This is because
 216 the closure in the fracture aperture under isotropic stress is controlled by normal compression
 217 while the shearing effect is negligible [Jiang et al., 2019]. This is also shown in Figure 3 where
 218 we observe that the mean aperture \bar{b} uniformly decreases over the three major fracture sets
 219 when applying isotropic stress (*Iso2*), since we observe the same value of \bar{b} , that is smaller
 220 than 1, for the fracture sets 5° , 50° and 120° for all the considered configurations of stress
 221 orientation (green curves in Figures 3a-d). This results in no changes in the contribution
 222 of the main fracture sets to transport (the same value of ~ 1 for all fracture sets, green
 223 curves in Figures 3e-h) and explains the overall delay observed in the BTCs for case *Iso2*.
 224 Note that the shape of BTCs exhibits a Gaussian-like distribution, which is similar to the

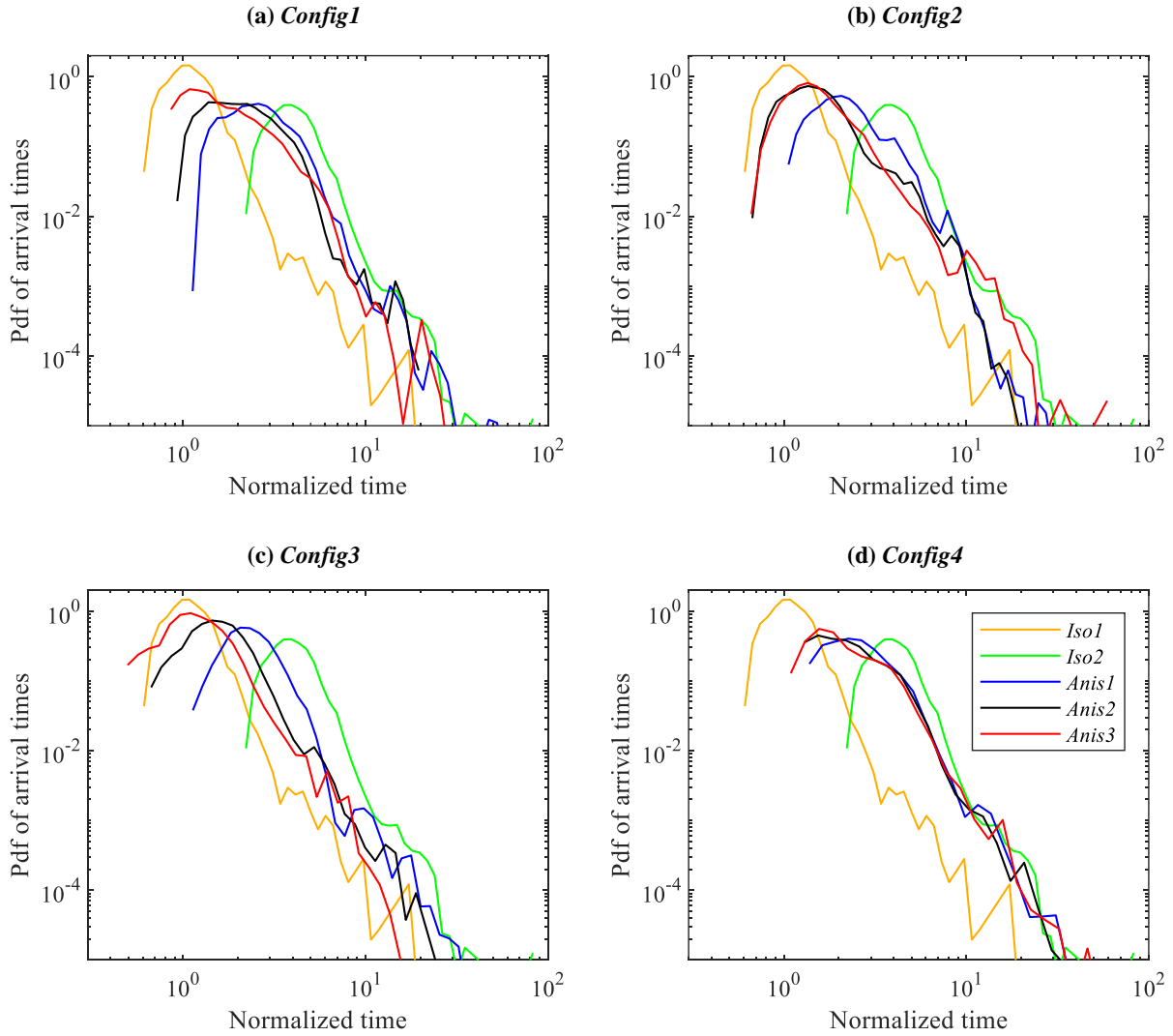


Figure 2: Breakthrough curves for the stress scenarios described in Table 1 without stress-driven crack growth and for the stress loading directions considered in (a) *Config1* ($\theta = 50^\circ$), (b) *Config2* ($\theta = 80^\circ$), (c) *Config3* ($\theta = 120^\circ$) and (d) *Config4* ($\theta = 150^\circ$).

225 observation in Kang et al. [2019]. This indicates that the geometrical heterogeneity of the
 226 fracture network is not sufficient to produce anomalous transport as the aperture is uniform
 227 for the no-stress case (also similar for the isotropic-stress case). It is also consistent with the
 228 distribution of emitted particles shown in Figure 4 where the preferential outlet ($Y = -2.79$ m)
 229 is not very pronounced (below 0.2) and secondary outlets emit an important proportion of
 230 particles (black bars in Figure 4).

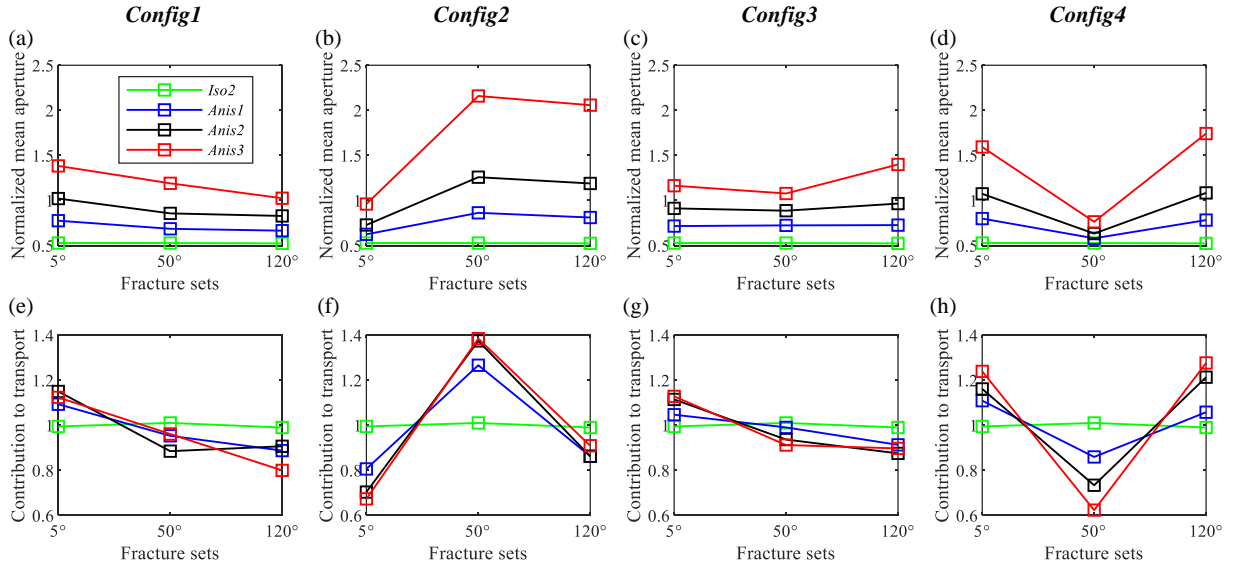


Figure 3: First row: mean aperture \bar{b} of the three major fracture sets (5° , 50° , and 120°) under the stress conditions described in Table 1 and normalized by that of the reference case. Second row: contribution of the three major fracture sets to transport defined as the time spent by the particles in each fracture set normalized by the time spent in all fracture sets. The contribution to transport is also normalized by that of the reference case.

231 Applying anisotropic stress conditions (*Anis1*, *Anis2*, and *Anis3*) results in broader
 232 arrival time distributions with earlier arrival times (from green to blue, black and red curves
 233 in Figure 2), which is attributed to local permeability enhancement by shearing [Zhao et al.,
 234 2011, Kang et al., 2019]. Shearing impact is visible on the changes in aperture and transport
 235 trajectories shown in Figure 3 where we observe non-uniform changes for all the anisotropic
 236 cases, except for *Anis1* and *Anis2* in *Config1* and *Config3*. In the latter cases, shearing on
 237 the fracture sets tends to be suppressed since \bar{b} varies from 0.6 to about 1 (from blue to
 238 black curves in Figures 3a and c). This is explained by the stress loading directions that are
 239 aligned with one of the major fracture sets in *Config1* and *Config3* while the three fracture
 240 sets tend to have large normal stress or small shear stress, which makes difficult to activate
 241 shearing and dilatancy. On the contrary, increasing the loading stress from *Anis2* and *Anis3*,
 242 or considering other configuration scenarios (*Config2* and *Config4*), results in heterogeneous
 243 changes in fracture apertures and transport behavior depending on the fracture orientation.

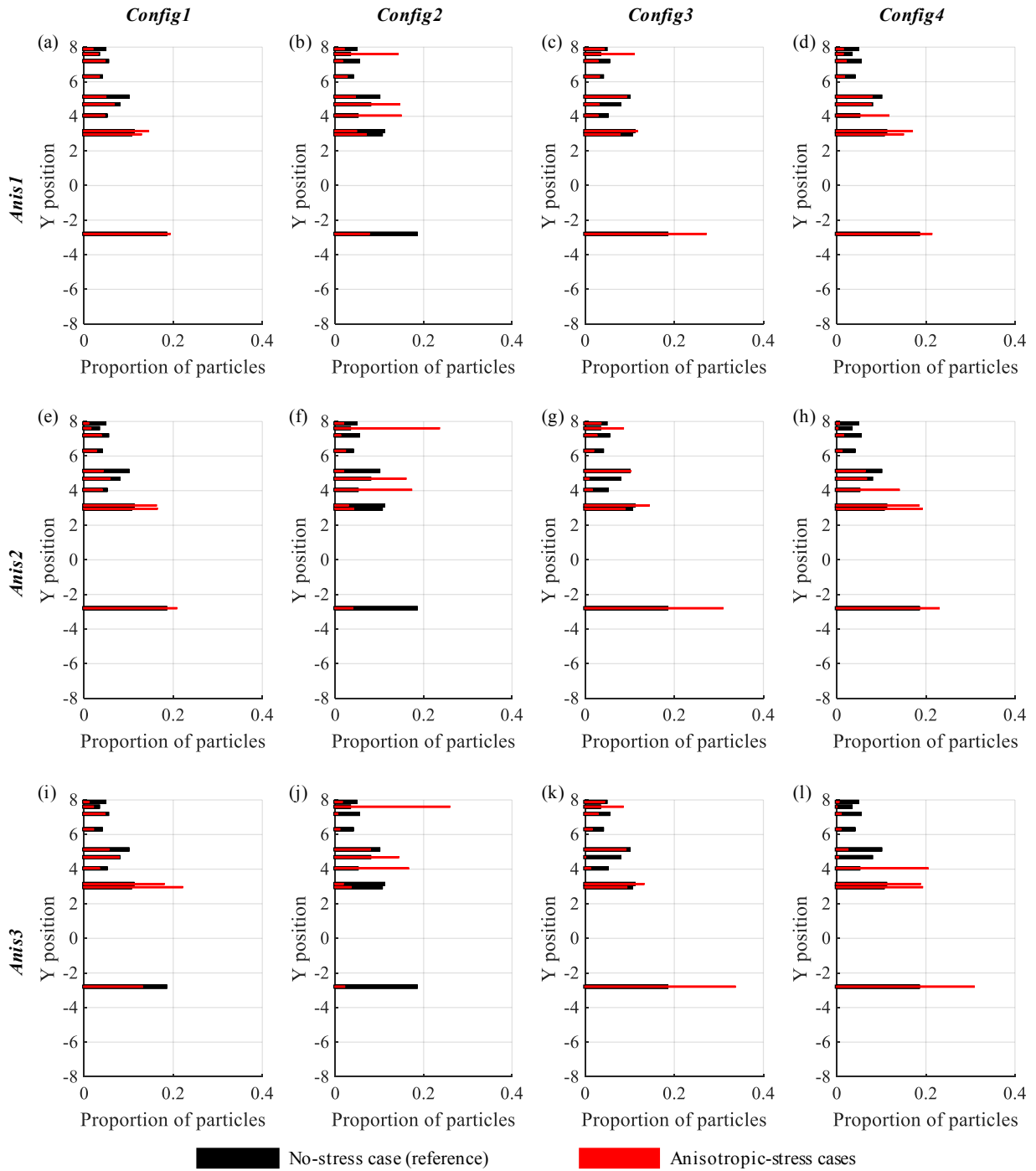


Figure 4: Spatial distribution along the outlet border (Y position from -8 to 8) of the proportion of particles exiting the system for different anisotropic-stress cases without stress-driven crack growth.

244 Different transport behaviors are also shown by the distribution of emitted particles at the
245 outlet boundary (Figure 4). We observe for example in *Config2* that there is not a main
246 preferential outlet for small stress ratio (*Anis1*, Figure 4b), while a new preferential outlet is
247 created when increasing the stress ratio from *Anis1* to *Anis2* (Figure 4f) with no significant
248 changes from *Anis2* to *Anis3* (Figure 4j). This explains the different behavior observed in
249 the BTCs in Figure 2b between *Anis1* and *Anis2*, and the similar behavior between *Anis2*
250 and *Anis3*. We also observe that in some cases, the preferential outlet observed for the
251 no-stress reference case ($Y=-2.79$ m) is greatly enhanced (*Config3* in Figures 4c, g and k),
252 and in other cases some secondary channels become stronger (*Config1* in Figures 4a, e and
253 i), resulting in different behavior in the corresponding BTCs where anomalous transport
254 behavior emerges with early arrival times (Figures 2a and c). Finally, for the cases related
255 to *Config4*, both the no-stress preferential outlet and the secondary outlets are enhanced
256 when increasing the stress ratio (Figures 4d, h and l), explaining that no significant changes
257 are observed in the corresponding BTCs (Figure 2d).

258 *3.2. Effect of stress-driven crack growth*

259 As the anisotropic stress loadings tend to drive the formation of new cracks due to
260 stress concentration at the tips of pre-existing fractures [Jiang et al., 2019], we further
261 examine the effect of stress-driven crack growth on transport behaviors. As confirmed in
262 Figure 5 and shown in Jiang et al. [2019], isotropic stress loading do not generate new crack
263 growth, implying that we study only the cases of anisotropic stress loadings by comparing
264 the breakthrough curves between the cases without and with new cracks (Figure 6). We
265 also compare the changes in flow velocities and particle trajectories (Figure 7), as well as
266 in outlet spatial distribution (Figure 8). Note that there are no changes in the normalized
267 mean aperture between the configurations without and with new cracks, implying that these
268 results are not shown.

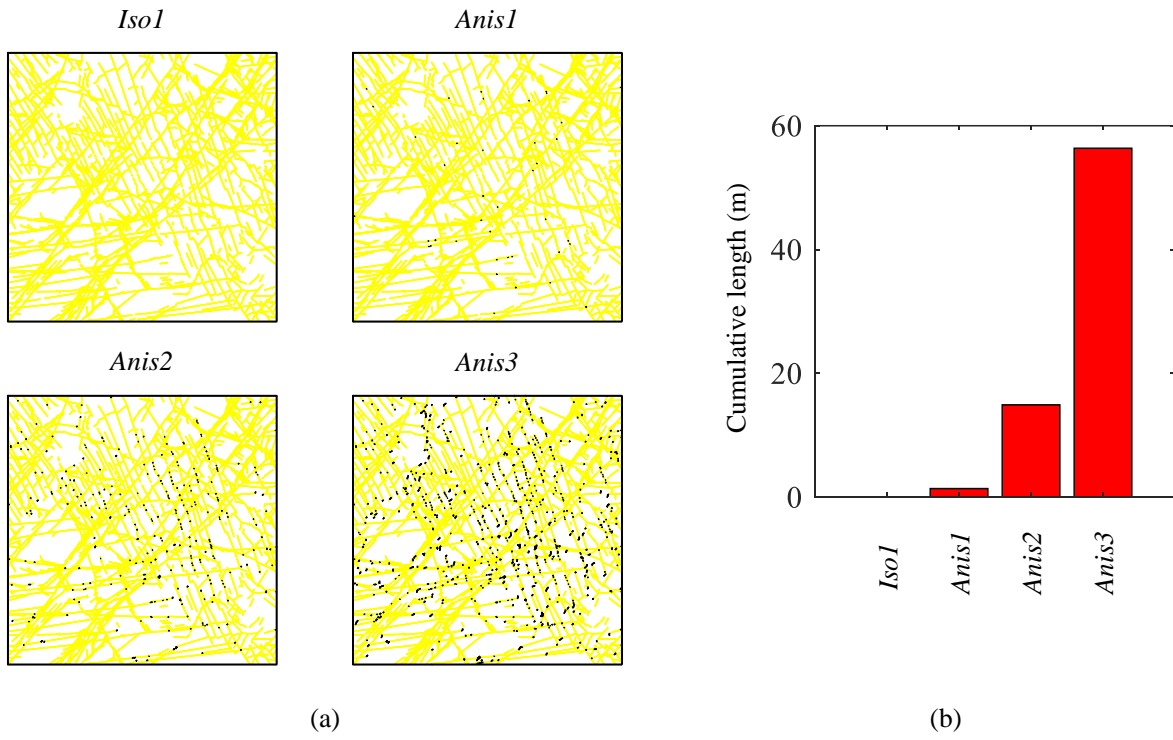


Figure 5: (a) Resulting fracture networks composed of the pre-existing fractures and new cracks (yellow and black lines, respectively) and (b) cumulative length of new cracks for the stress configuration *Config1* and the stress scenarios presented in Table 1.

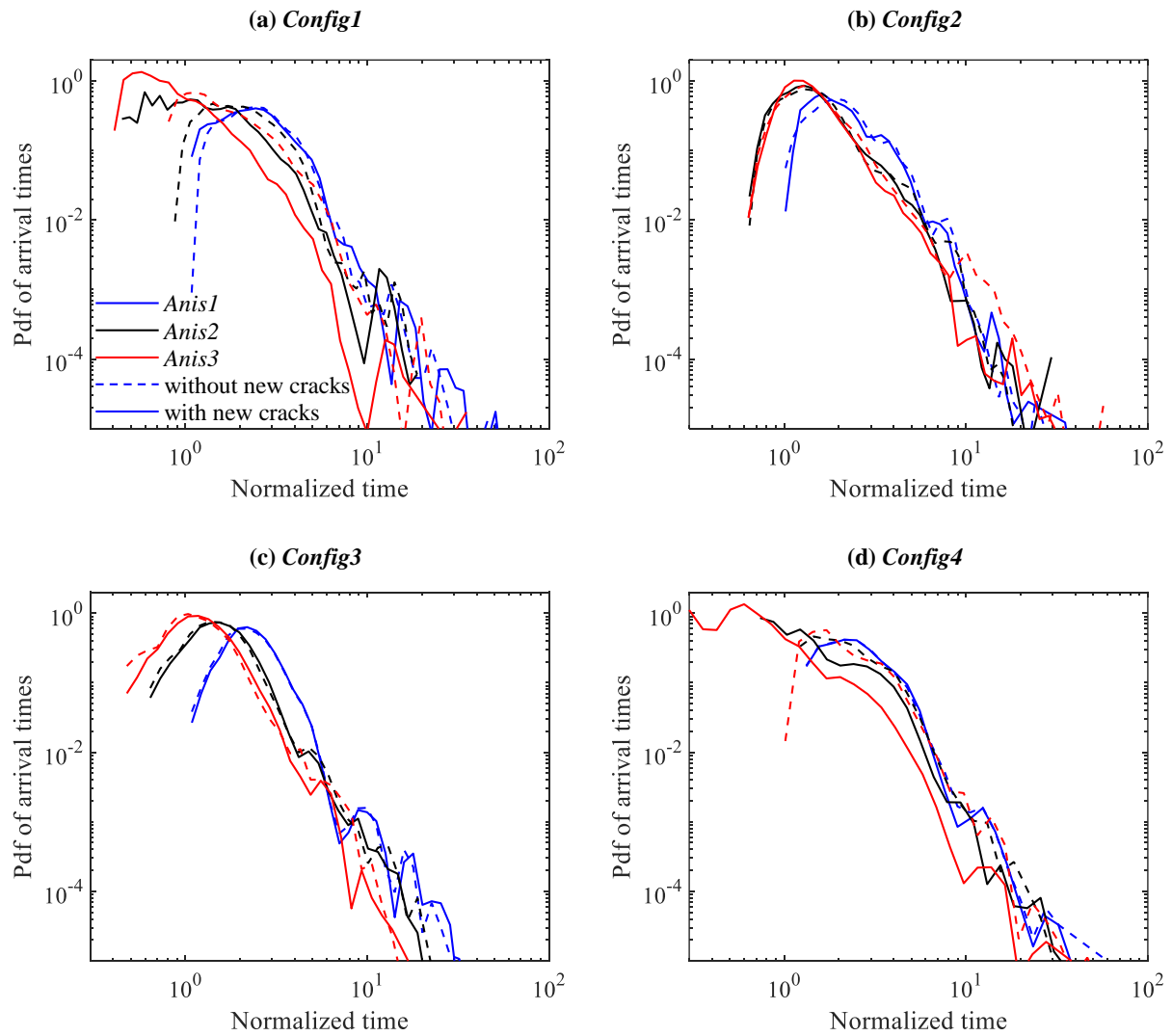


Figure 6: Comparison of breakthrough curves without and with (dashed and solid lines, respectively) new cracks for various anisotropic-stress cases.

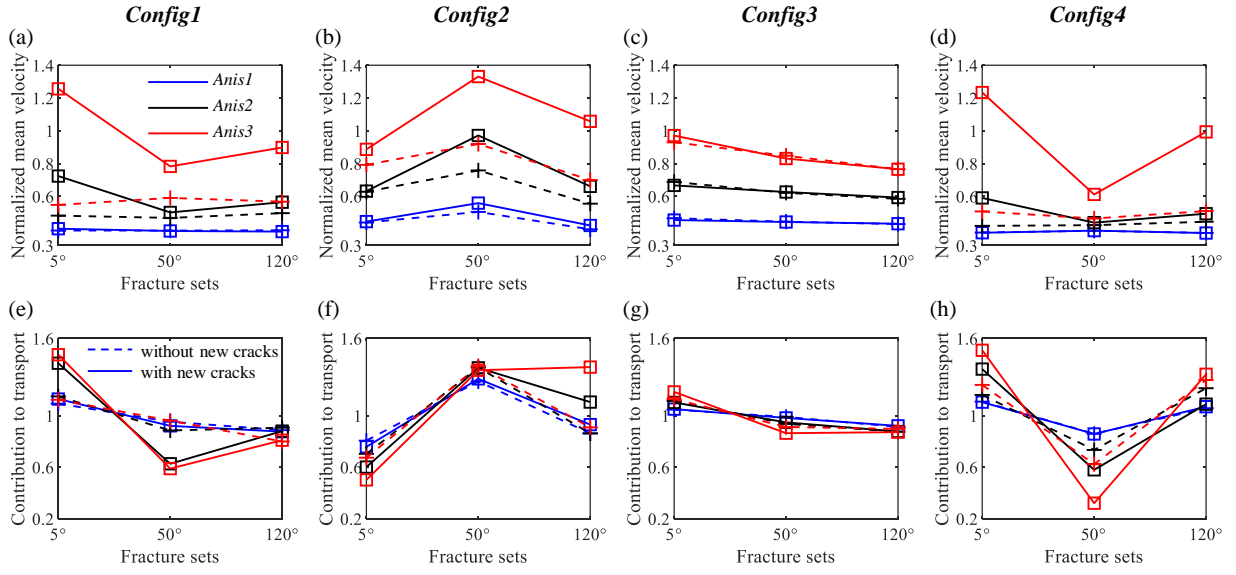


Figure 7: First row: mean velocity \bar{v} of the three major fracture sets (5° , 50° , and 120°) under the differential stress conditions *Anis1*, *Anis2*, and *Anis3* that are normalized by that of the reference case. Second row: contribution of the three major fracture sets to transport defined as the time spent by the particles in each fracture set normalized by the time spent in all fracture sets. The contribution to transport is also normalized by that of the reference case.

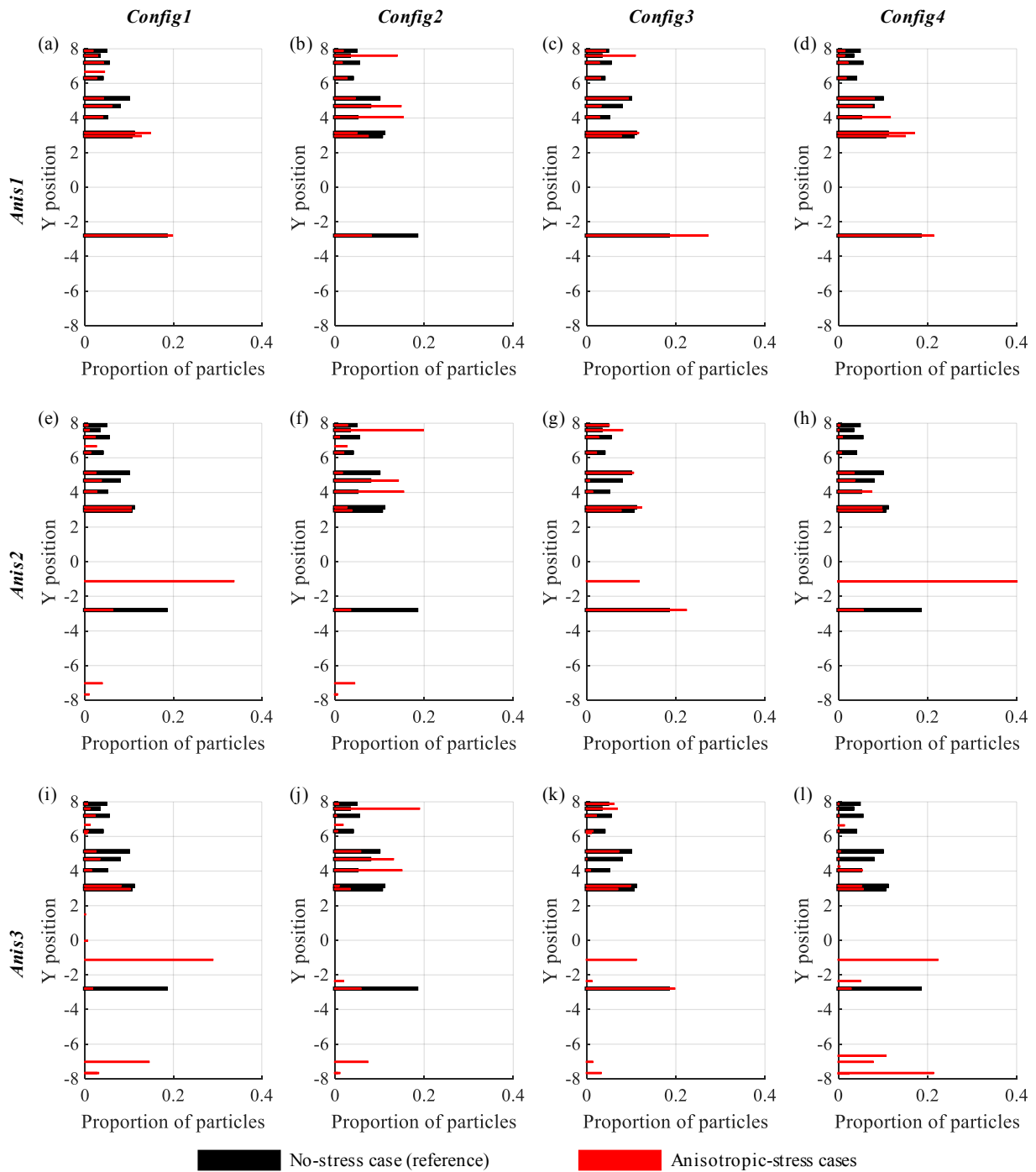


Figure 8: Spatial distribution along the outlet border (Y position from -8 to 8) of the proportion of particles exiting the system for different anisotropic-stress cases with stress-driven crack growth.

270 scenario *Anis1* is too small to impact the transport behavior since the results without and
271 with new cracks are similar for this case (blue curves in Figure 6). This is confirmed by
272 the normalized mean velocity and contribution to transport that are similar for the config-
273 urations without and with new cracks with scenario *Anis1* (blue curves in Figure 7). For
274 this scenario, we also observe no significant changes in the spatial distributions of particles
275 exiting the system between the configurations without new cracks (Figures 4a-d) and that
276 considering the effect of new crack growth (Figures 8a-d).

277 For scenarios *Anis2* and *Anis3*, the impact of new cracks on transport depends on the
278 loading configurations. For *Config3*, similar results are observed between the systems with-
279 out and with new cracks (Figure 6c), while changes in late arrival times are observed for
280 *Config2* and *Anis3* (red curves in Figure 6b). For *Config1* and *Config4*, changes in early
281 arrival times are observed with *Anis2* and over all the arrival times with *Anis3* (black and
282 red curves in Figures 6a and d, respectively). These observations are well explained by
283 the results shown in Figures 7 and 8. For *Config3*, we observe no significant changes in
284 the normalized mean velocity and contribution to transport of the three main fracture sets
285 (Figures 7c and g), and the changes in the proportion of particles along the outlet border
286 are restricted to either slightly enhancement of existing main paths (e.g., $Y=-2.79$ m in
287 Figures 8g and k) or new paths that do not have a significant impact with a proportion of
288 particles smaller than 0.2 (e.g., $Y=-1.12$ m in Figures 8g and k). Similar changes in the
289 spatial distribution of the particles exiting the systems are observed for *Config2* (Figures 8f
290 and j), with larger values of mean velocity when considering new cracks (Figure 7b) and a
291 larger contribution to transport of the fracture set 120° (Figures 7f), resulting in the larger
292 late arrival times observed in the BTCs for *Anis3* (red curve vs. blue curve in Figure 6b).
293 Finally, the smaller early arrival times observed with *Anis2* for *Config1* and *Config4* are
294 related to the creation of new main preferential paths that have a strong impact on the
295 transport organization with a proportion of particles exiting the system from these paths

296 larger than 0.3, these paths being located at position $Y=-1.12$ m for both configurations,
 297 respectively (Figures 8e and h, respectively). However, the changes in mean velocity and
 298 contribution to transport over the whole system are not significant (black lines in Figures 7a
 299 and e for *Config1* and Figures 7d and h for *Config4*), implying that only early arrival times
 300 are impacted by this stress scenario. Alternatively, when applying stress scenario *Anis3*, the
 301 changes in flow velocity and contribution to transport are significant (red lines in Figures 7a
 302 and e for *Config1* and Figures 7d and h for *Config4*), and the new outlets at the bottom are
 303 strengthened while the new preferential outlets in the middle are weakened (Figures 8i and
 304 l). Smaller arrival times over the full range of particle arrival times are then observed (red
 305 curves in Figures 6a and d) with a less uniform distribution of emitted particles at the outlet
 306 boundary compared to the cases without new cracks. This results in anomalous transport
 307 behavior where the propagation of new cracks generates BTCs with multiple peaks.

308 3.3. Upscaling of transport properties

309 In this section, we use a series of upscaling methods based on the random walk theory to
 310 capture the transport behavior under various stress conditions. Traditional methods (i.e.,
 311 URW and MDRW) are first applied to examine their predictive performance. To this end,
 312 we compare the three existing methods URW-std, URW-mod, and MDRW-std that are
 313 described in Section 2.3. Figures 9 and 10 show the results of upscaling of the small-scale
 314 transport behavior shown in Figures 2 and 6 without and with new cracks, respectively. All
 315 the configurations and stress scenarios presented in Section 2.1 are considered. Note that
 316 the results obtained for scenarios *Iso1* and *Iso2* are only shown once (Figures 9a-b) since the
 317 results are the same for the four considered stress configurations and the cases without and
 318 with new cracks. Furthermore, to quantify the performance of different upscaling methods,
 319 the statistics of mean, standard deviation, and skewness of each breakthrough curve [Schiavo
 320 et al., 2022] are shown in Figure 11. The skewness is defined as $\gamma = E \left[\left(\frac{x-\mu}{\sigma} \right)^3 \right]$ where E

321 is mathematical expectation, μ and σ are respectively the mean and standard deviation of
322 variable x . Here $x = \log_{10}(t)$ where t is the normalized time in Figures 9 and 10.

323 The results presented in Figures 9, 10 and 11 show that the method URW-std (dashed
324 blue curves) never succeeds in reproducing the transport behavior observed with the direct
325 small-scale simulations (solid black lines), because the intrinsic correlation between succes-
326 sive velocity samplings is not considered with this method. Changing the database used
327 by URW (URW-mod, dashed green lines) or introducing a velocity correlation with MDRW
328 method (MDRW-std, dashed red lines) help to reproduce well the reference simulations for
329 isotropic cases (Figures 9a-b, Figures 11a-c). Consequently, if the sampling database is
330 modified by space-Lagrangian velocities, the performance of the URW method is greatly
331 improved, with results that are similar to that of the MDRW method with the standard
332 database (Figure 11). This is because the modified database with URW includes the cor-
333 relation information within the distance of the large sampling lag (0.5 m) compared to the
334 standard database with a small segment discretization (0.05 m). However, for anisotropic-
335 stress cases, both URW-mod and MDRW-std do not reproduce well the small-scale results
336 due to the shearing-induced anomalous transport phenomenon that is hard to reproduce
337 (Figures 9 and 11a-c), this issue being even worse when considering cases with new cracks
338 for which the anomalous transport behavior is more pronounced and thus harder to repro-
339 duce (Figures 10 and 11d-f). This is consistent with previous studies showing that traditional
340 upscaling framework that only considers the average correlation property is likely to fail in
341 fracture systems driven by shear dilation-induced preferential flow paths [Kang et al., 2019].

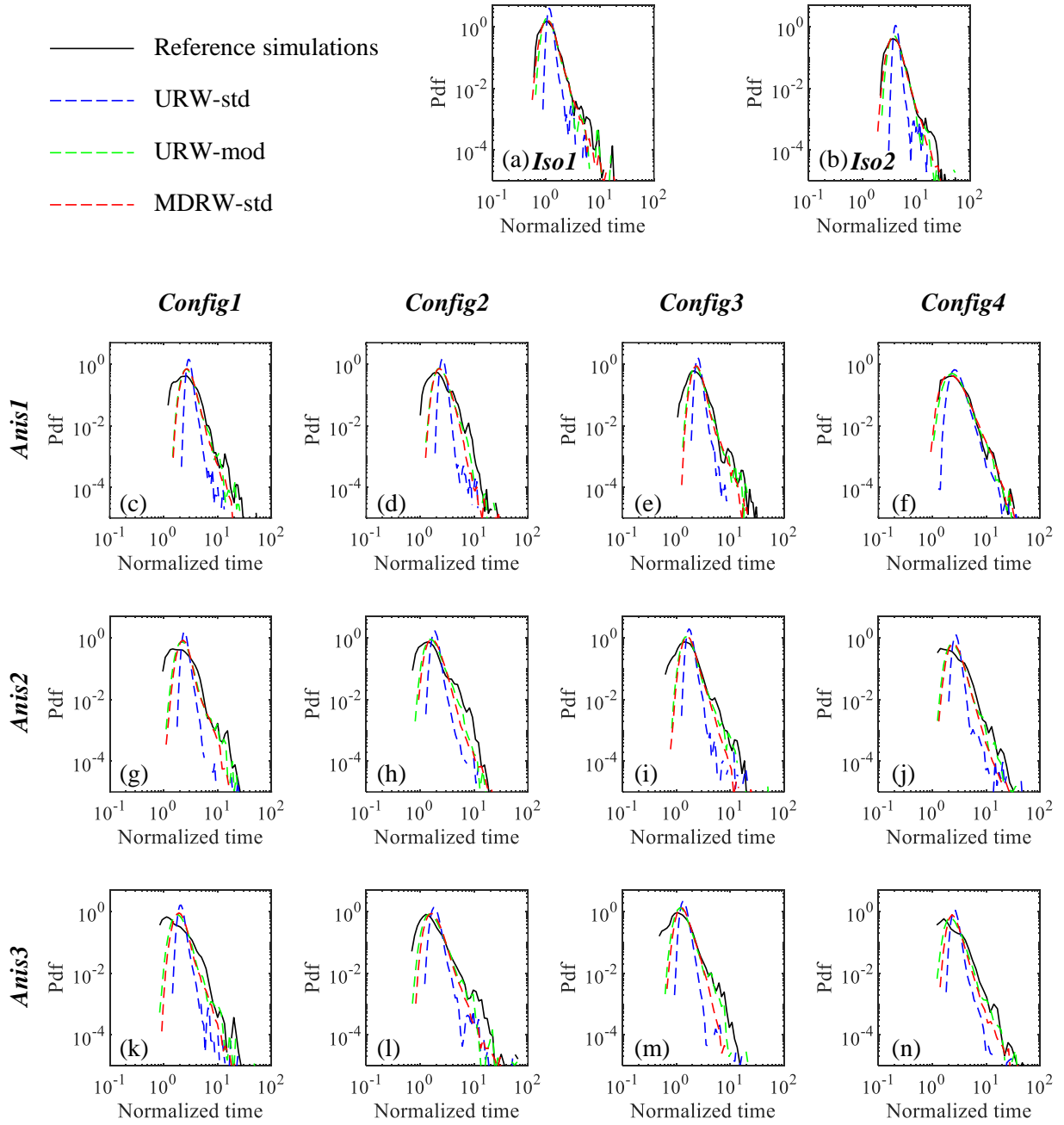


Figure 9: Breakthrough curves from reference small-scale simulations (solid black line) and upscaled model predictions (dashed lines) using URW-std, URW-mod, and MDRW-std methods (dashed blue, green, and red curves, respectively) without considering the effect of new crack propagation.

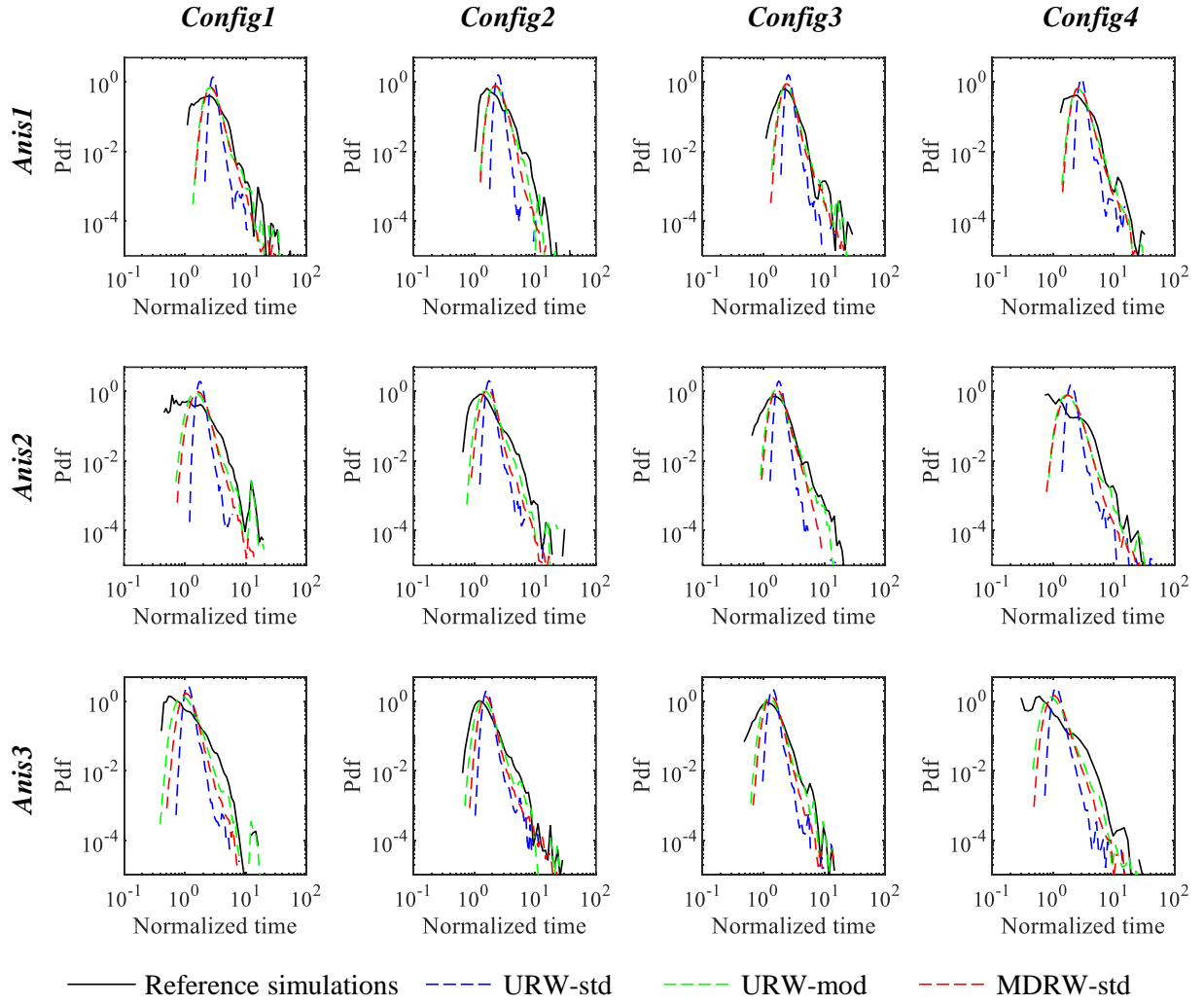


Figure 10: Breakthrough curves from reference small-scale simulations (solid black line) and upscaled model predictions (dashed lines) using URW-std, URW-mod, and MDRW-std methods (dashed blue, green, and red curves, respectively) considering the effect of new crack propagation.

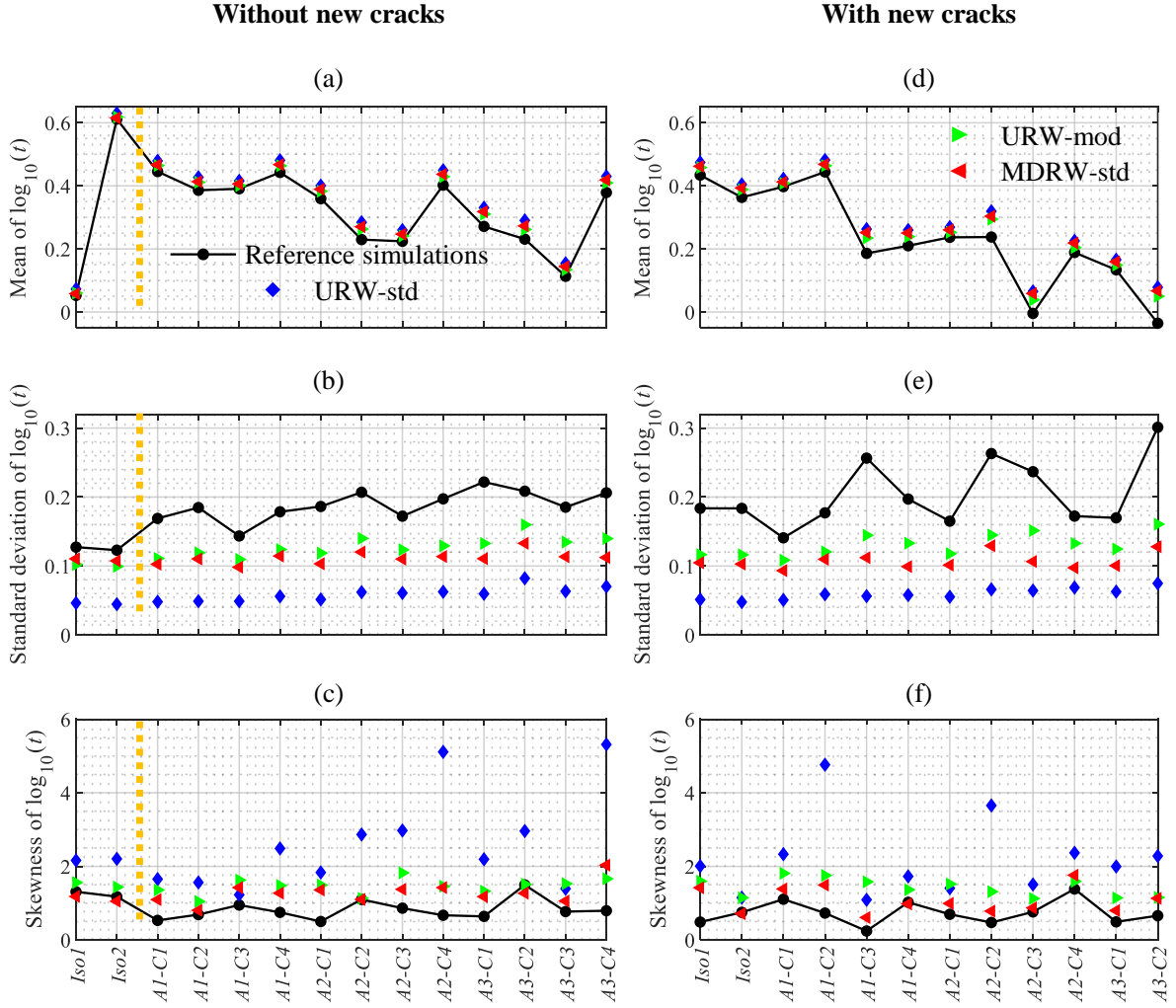


Figure 11: Statistics of mean, standard deviation and skewness of $\log_{10}(t)$ where t is the normalized time in Figures 9 and 10.

342 To overcome this issue, we attempt to improve the predictive performance of tradi-
343 tional upscaling methods with the modified methods MDRW-mod and Multi-M MDRW.
344 The corresponding results are shown in Figures 12 and 13 with the quantitative statistics
345 in Figure 14. We observe that, after adding the velocity correlation between the samplings
346 of space-Lagrangian velocities, the modified MDRW method is able to well capture the ma-
347 jority of the cases in this study. Compared to the MDRW with the standard database, the
348 performance of MDRW with the modified database is greatly improved, which highlights

349 the importance of space-Lagrangian velocities. Both MDRW-mod and Multi-M MDRW
350 provide satisfying results to reproduce the BTCs obtained at the fracture-network scale. In
351 some cases, the Multi-M MDRW method is better suited to reproduce the complex behavior
352 observed at the beginning of the curves. This is the case for example in the highly het-
353 erogeneous cases *Config4-Anis3* without and with new cracks (Figures 12n and 13n), for
354 which the information of the first velocity experienced by the particles (that is included
355 in the Multi-M MDRW method) is an information for the particle trajectories. From the
356 quantitative statistics in Figure 14, it is more pronounced that results from Multi-M MDRW
357 almost overlap the reference cases, especially for the standard deviation of $\log_{10}(t)$, which
358 indicates a better performance than MDRW.

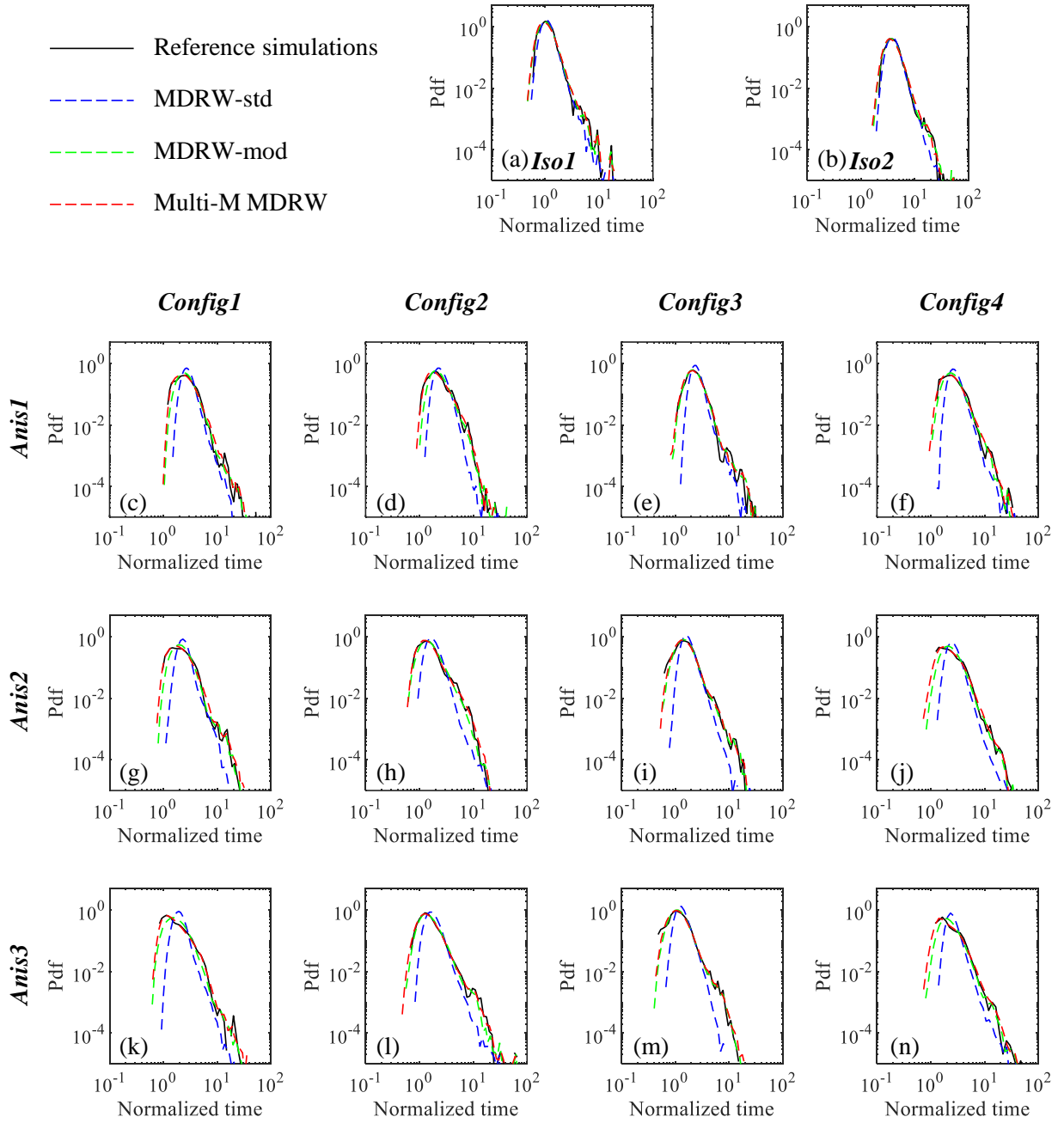


Figure 12: Breakthrough curves from reference small-scale simulations (solid black line) and upscaled model predictions (dashed lines) using MDRW-std, MDRW-mod, and Multi-M MDRW methods (blue, green, and red, respectively) without considering the effect of new crack propagation.

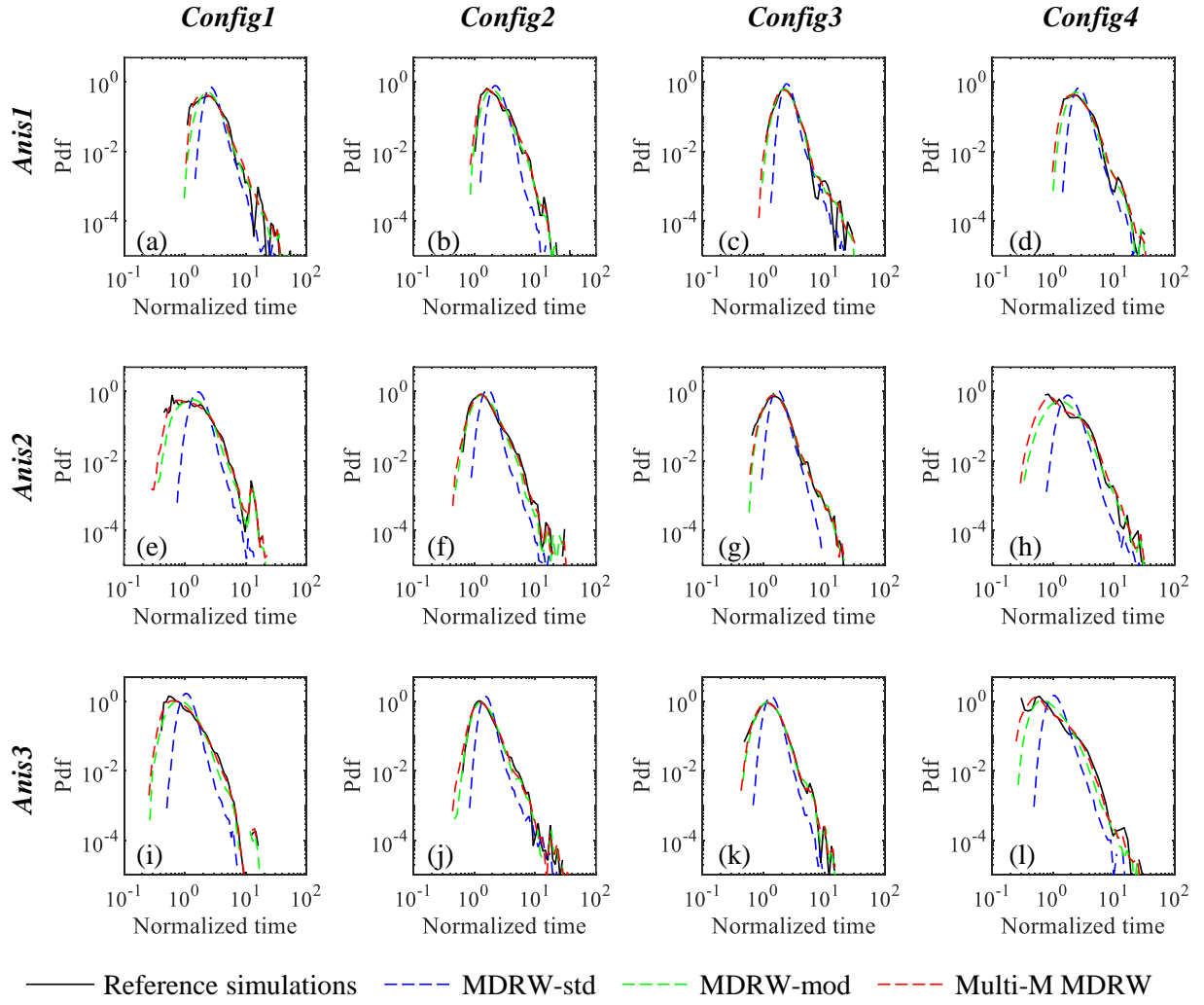


Figure 13: Breakthrough curves from reference small-scale simulations (solid black line) and upscaled model predictions (dashed lines) using MDRW-std, MDRW-mod, and Multi-M MDRW methods (blue, green, and red, respectively) considering the effect of new crack propagation.

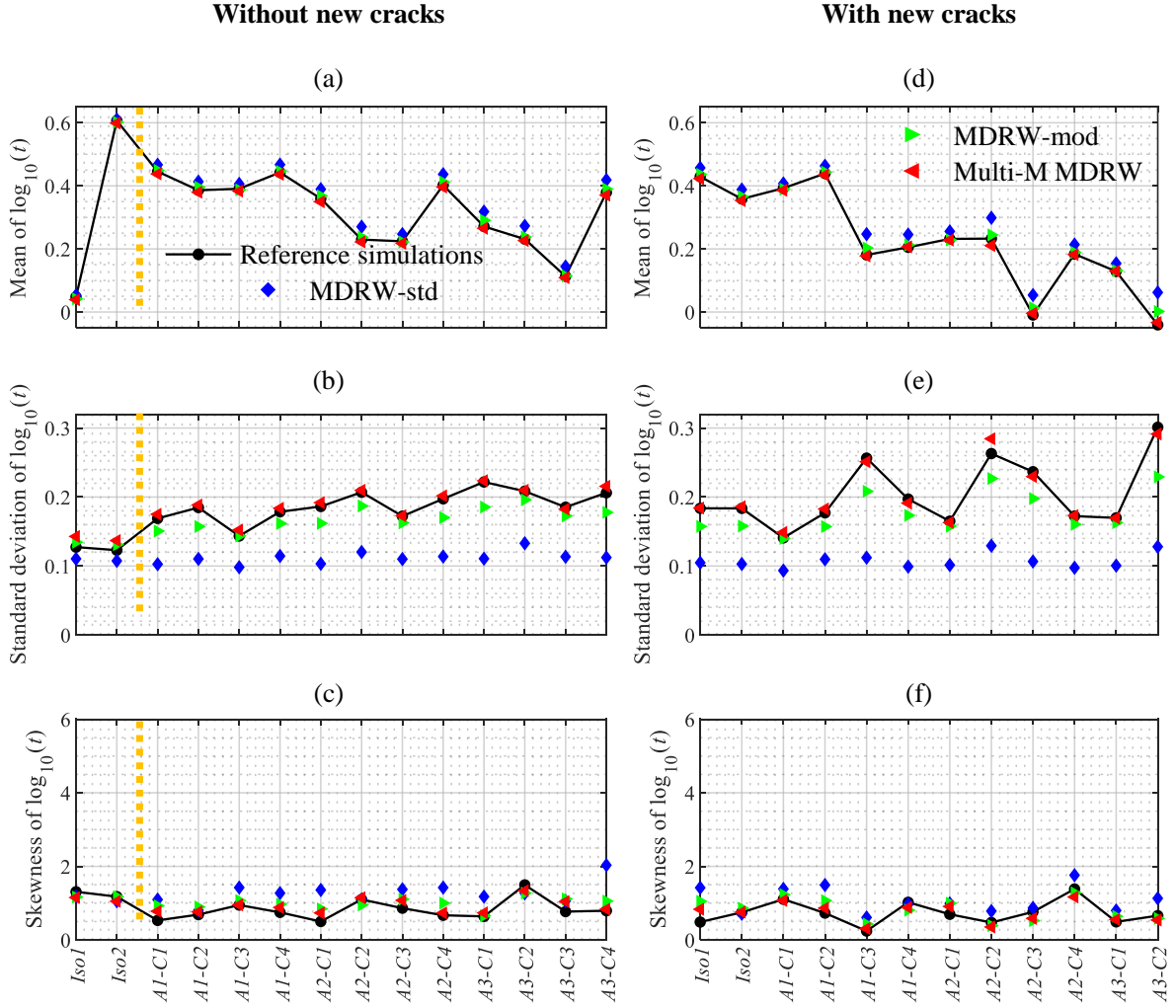


Figure 14: Statistics of mean, standard deviation and skewness of $\log_{10}(t)$ where t are the normalized time distributions corresponding to Figures 12 and 13.

359 3.4. Effect of diffusion transport in rock matrix

360 We wish now to test the upscaled models previously presented in the context of both
 361 advective and diffusive processes in the fractures and matrix, respectively. To this end,
 362 we consider equation (1) to define the matrix diffusion time in both direct and upscaled
 363 simulations. Figures 15 and 16 show the breakthrough curves obtained for the stress config-
 364 urations and scenarios previously defined with the direct small-scale simulations (solid black
 365 curves) and the MDRW-mod and Multi-M MDRW upscaled models (dashed blue and red

366 lines, respectively) without (Figure 15) and with (Figure 16) considering the effect of new
367 crack propagation. In Figure 17, the quantitative statistics show that, as expected, matrix
368 diffusion results in long tails of the breakthrough curves with a fixed slope of -1.5 that cor-
369 responds to pure diffusion process. Considering matrix diffusion tends to homogenize the
370 results with similar curves for the different considered stress scenarios. Compared to the
371 cases without matrix diffusion (Figures 2 and 6), the early part of the curves is smoother
372 and do not present multi-peak behavior as before for the most heterogeneous cases (*Config4-*
373 *Anis2* and *Config4-Anis3*). The two upscaling methods reproduce well the considered BTCs
374 with small differences for the early arrival times in some cases where the Multi-M MDRW
375 method provides a better fit than the MDRW method. For these cases, the better repre-
376 sentation of the initial velocities investigated by the particles plays an important role in
377 simulating the early arrival times, since the matrix diffusion time directly depends on the
378 advective times (expression (1)). This confirms the effectiveness of the Multi-M MDRW
379 method in upscaling stress-induced anomalous transport behavior for various configurations
380 in terms of stress scenarios and considered processes.

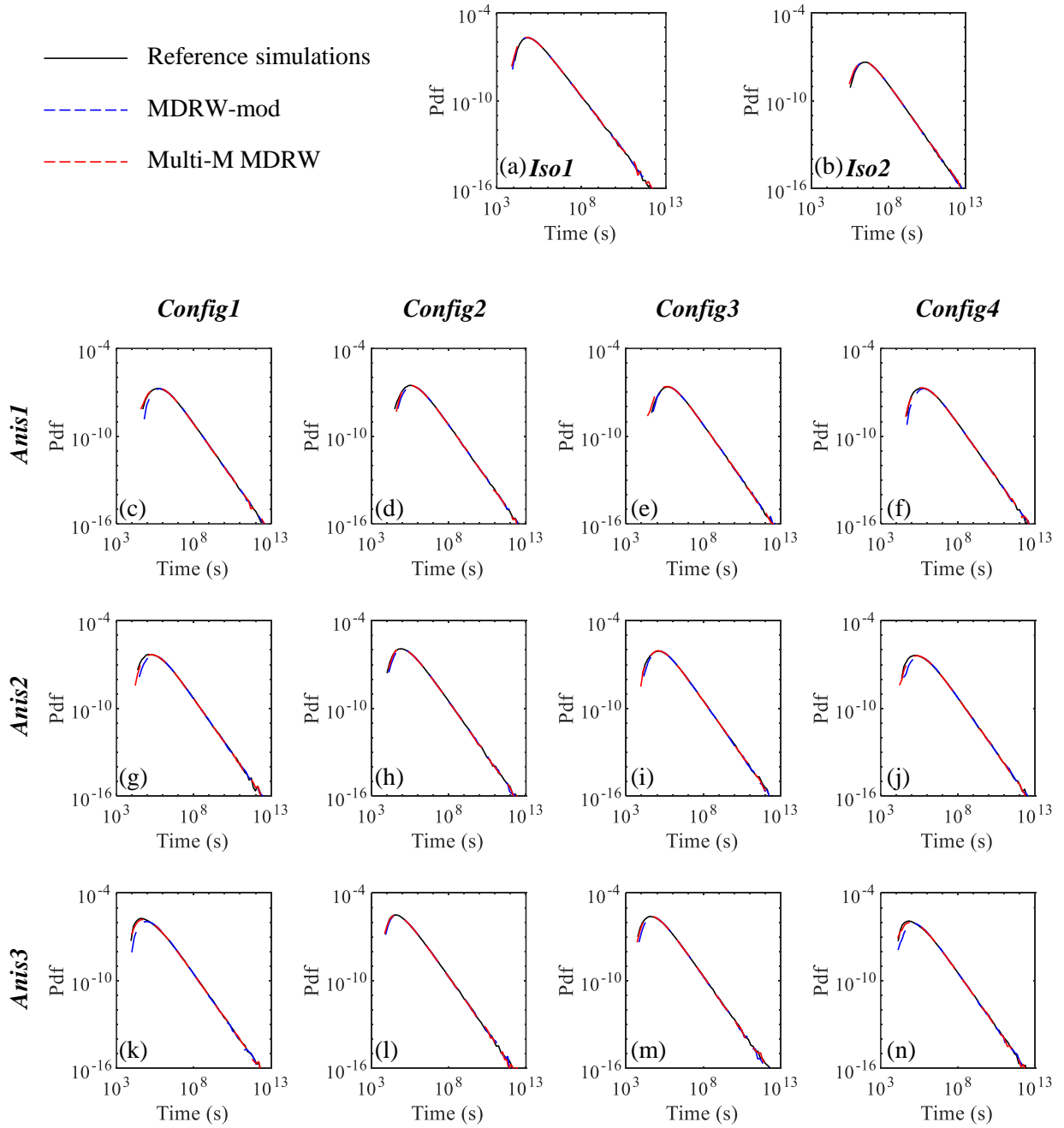


Figure 15: Breakthrough curves from the direct numerical simulations with matrix diffusion (black solid line) and upscaled model predictions using MDRW-mod and Multi-M MDRW methods with matrix diffusion (dashed blue and red lines, respectively) without considering the effect of new crack propagation.

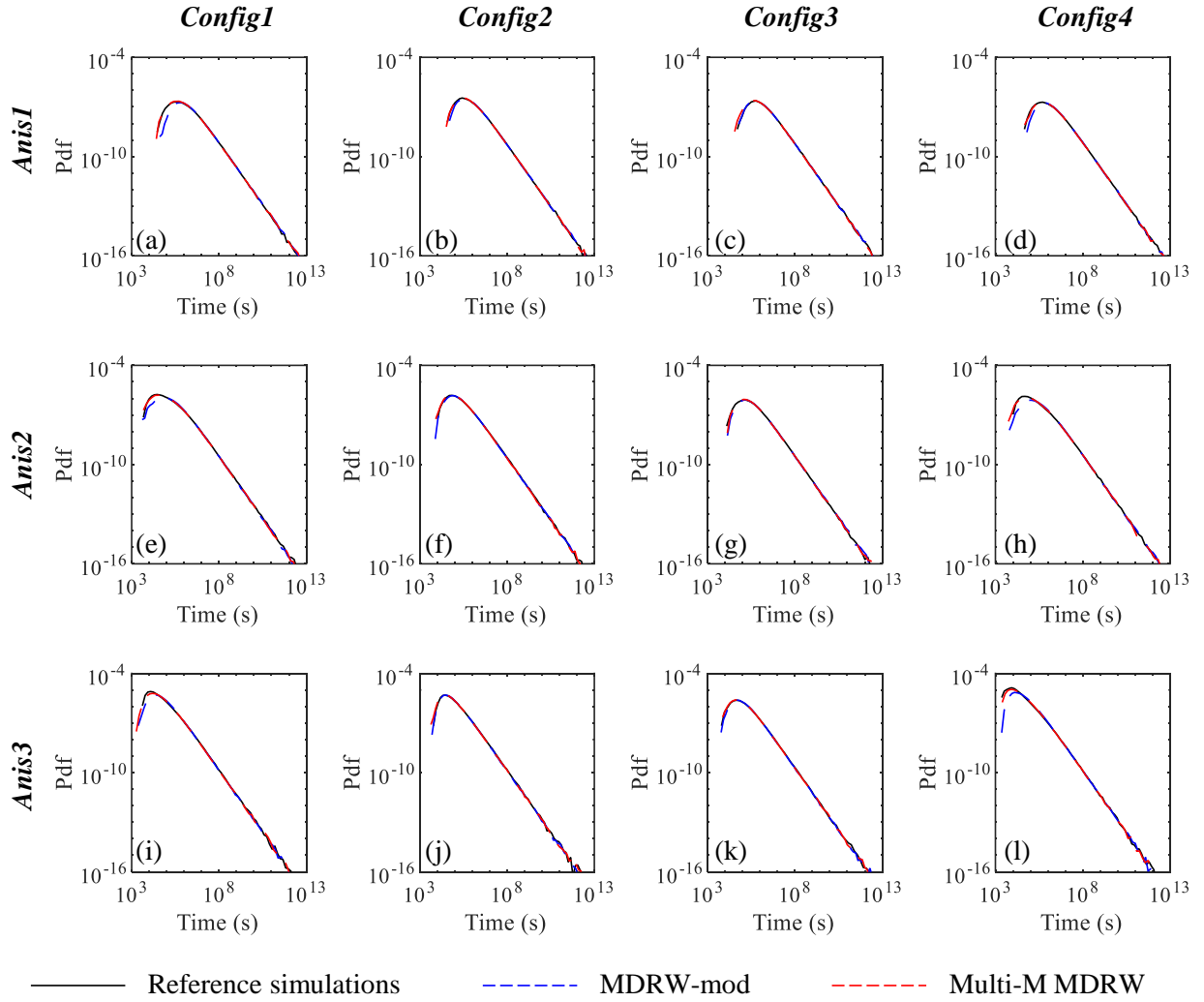


Figure 16: Breakthrough curves from the direct numerical simulations with matrix diffusion (black solid line) and upscaled model predictions using MDRW-mod and Multi-M MDRW methods with matrix diffusion (dashed blue and red lines, respectively) considering the effect of new crack propagation.

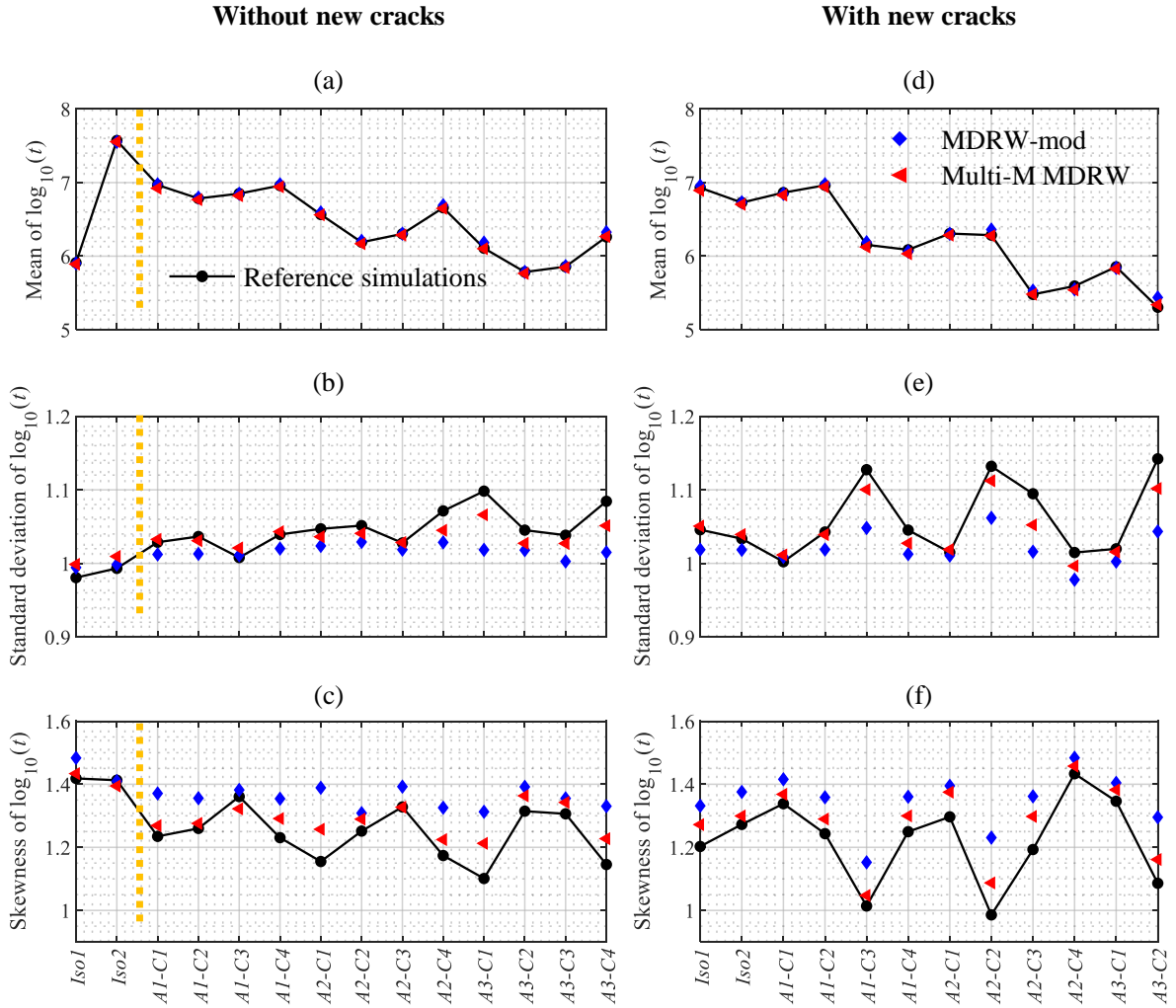


Figure 17: Statistics of mean, standard deviation and skewness of $\log_{10}(t)$ where t are the normalized time distributions corresponding to Figures 15 and 16.

381 **4. Discussion and conclusions**

382 In this paper, we evaluate the impact of various stress scenarios on transport behaviors
 383 in critically-connected fractured rocks with three main fracture sets. Various stress orienta-
 384 tions and a large range of stress ratios between the maximum and minimum principal stress
 385 strengths are studied, and both traditional and improved upscaling methods are considered
 386 to determine their applicability to such stress-induced transport behaviors. Under isotropic

387 stress conditions, the changes over the whole system tend to be uniformly distributed, and
388 the breakthrough curves are simply shifted from the reference results without stress. In
389 such conditions, transport behaviors usually exhibit a Gaussian-like distribution, that is
390 well captured by traditional upscaling methods with space-Lagrangian velocities descrip-
391 tion. Under anisotropic stress conditions, stress-induced transport exhibits broader arrival
392 time distributions with different behavior depending on the stress orientation and strength.
393 This is consistent with existing studies showing that the anomalous behavior featured by
394 the early arrival times may be not pronounced under some stress orientations because the
395 major compressive principal stress is perpendicular to the flow direction, suppressing and
396 altering the preferential flow paths [Sweeney and Hyman, 2020]. This differs from the case
397 of a ladder-shape fracture network topology, where the anisotropic stress orientation is along
398 the through-going long fractures and exhibits a breakthrough behavior similar to the no-
399 stress and isotropic-stress cases [Kang et al., 2019]. The behaviors observed in our study are
400 enhanced when considering the effect of new crack growth that is observed under anisotropic
401 stress loading and for high-stress ratio and some stress orientations, leading to strong anoma-
402 lous phenomenon with multi-peak BTCs.

403 In terms of upscaling representation, it is necessary to integrate the space-Lagrangian ve-
404 locities description with a large sampling distance to improve the performance of traditional
405 methods. The proposed modified MDRW method accurately predicts the transport behav-
406 ior under low-stress ratio conditions while improvement might be required for some cases
407 of high-stress ratio due to the strong anomalous transport phenomenon characterized by
408 multi-peak early-times BTCs. The single transition matrix representation, which describes
409 the average features of velocity correlation over the entire system, is extended to multiple
410 transition matrix representations that are conditioned by the initial velocity states of the
411 particles. (i.e., Multi-M MDRW). The resulting Multi-M MDRW successfully captures the
412 solute transport behaviors for all the considered cases including complex combined effects of

413 shear dilation and crack propagation in the system, as well as the impact of diffusive trans-
414 port in the rock matrix. Incorporating information related to the first velocities experienced
415 by the particles is critical to well reproduce the early arrival times, and might be critical for
416 future work on upscaling methods applied to reactive transport [Ding et al., 2013, Painter
417 et al., 2008, Roubinet et al., 2022].

418 It is important to emphasize that the insights provided in this work are restricted to
419 the deterministic geometry that was determined from a real outcrop. Since an accurate
420 representation of the fracture network is not feasible in most subsurface cases, future work
421 should focus on several realizations of networks with identical statistical properties in order
422 to evaluate the uncertainty of our results in relation with the uncertainty of the system
423 properties. Our recent study based on statistically generated fracture networks also found
424 that the critically connected fracture networks have a significant alternation of heat recov-
425 ery efficiency with the increase of stress load difference [Sun et al., 2020]. This may also
426 highlight the important role of crack growth on transport behaviors in terms of statisti-
427 cally generated fracture networks. Another limitation of our work is the consideration of
428 two-dimensional scenarios. Although our study regarding the geomechanical deformation
429 of three-dimensional fractured rocks has suggested that 2D analysis may provide some in-
430 dicative approximations [Lei et al., 2017b], 2D modeling certainly cannot fully represent the
431 polyaxial geomechanical behaviors of 3D fractured rocks. Future work should focus on 3D
432 fractured rocks and on evaluating the error induced by considering 2D domains. However,
433 working on 3D fractured networks requires to solve computational cost issues which limit
434 the number of fractures that can be considered so far.

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