Application of Hausdorff fractal derivative to the determination of the vertical sediment concentration distribution

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5 Abstract:

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6 The Rouse formula and its variants have been widely used to calculate the steady-state vertical 7 concentration distribution for suspended sediment in steady sediment-laden flows, where the 8 diffusive flux is assumed to be Fickian. Turbulent flow, however, exhibits fractal properties, 9 leading to non-Fickian diffusive flux for sediment particles. To characterize non-Fickian dynamics 10 of suspended sediment, the current study proposes a Hausdorff fractal derivative based 11 advection-dispersion equation (HADE) model, where the Fickian diffusive flux in the Rouse 12 model is replaced by a fractal derivative re-scaled using a constant diffusivity. The order of the 13 Hausdorff fractal derivative is designed to characterize the influence of the multi-fractal 14 turbulence structure on sediment diffusion. Applications show that the HADE model, with the 15 analytical solution expressed using a stretched exponential function, can accurately describe the 16 observed vertical concentration profiles for suspended sediment with different sizes. This 17 improvement well captures the non-exponential decay of the vertical sediment concentration in 18 turbulent flow. Further analyses of measured sediment concentration profiles reveal that the 19 Hausdorff fractal order decreases with the Rouse parameter, which describes the stronger impact 20 of turbulent flow and a more uniform sediment concentration profile for smaller particles. Model 21 comparisons also show that the HADE model provides better performance in describing the 22 sediment concentration profiles than the improved Rouse formula and the standard fractional 23 derivative advection-dispersion equation (FADE), which either under- or over-estimates vertical 24 displacement of sediment particles, likely due to coherent turbulent structures.

25 Key words: Anomalous diffusion; Hausdorff fractal derivative; Vertical concentration distribution;

26 Suspended sediment; Metric transform

27 1. Introduction

Quantification of the vertical distribution of suspended sediment concentration underturbulent flow conditions, which is one of the essential parts of morphological computations, has

30 been a research focus for decades in the hydraulics and environmental science communities 31 (Cheng et al., 2018; Chien & Wan, 1999; Czuba et al., 2015; Nowacki et al., 2015; Park & 32 Latrubesse, 2015). Various theoretical or empirical formulas have been applied to model the 33 steady-state, vertical density/concentration profile of suspended sediment (Boudreau & Hill, 2020; 34 van Rijn, 1984; Rouse, 1937). For example, Umeyaina (1992) investigated the vertical distribution 35 of suspended sediment in uniform open-channel flow using a mixing length hypothesis, which can 36 fit the experimental data for both fine and coarse sediment particles. Wang and Fu (2004) 37 developed a kinetic model to quantify sediment suspension in solid/liquid two-phase flows, as 38 well as the particle velocity distribution function in steady flows (Fu et al., 2005; Wang & Fu, 39 2004; Wang & Ni, 1991). Mazumder and Ghoshal (2006) proposed a theoretical model based on 40 the Hunt equation to capture the profiles of velocity and sediment. More information on empirical 41 formulas and theoretical results on the vertical distribution of suspended sediment concentration 42 can be found in related references (Boudreau & Hill, 2020; Dey et al., 2018). The existing 43 formulas are mainly built upon the Fick's law based diffusive flux to describe the vertical 44 movement of suspended sediment (Chien & Wan, 1999; Coleman, 1986; Kumbhakar et al., 2017).

The standard models for the vertical distribution of suspended sediment concentration in
sediment-laden flows were derived from the advection-dispersion equation (ADE), which reduces
to the following form under steady-state conditions:

48
$$\omega S + \varepsilon_{sy} \frac{\partial S}{\partial y} = 0, \qquad (1)$$

where $S[ML^{-1}]$ denotes the sediment volumetric concentration, y[L] is the vertical 49 coordinate, $\omega [LT^{-1}]$ is the sediment settling velocity, and $\varepsilon_{sy} [L^2T^{-1}]$ is the sediment 50 51 turbulent diffusion coefficient along the y direction. Equation (1) assumes that, when reaching 52 equilibrium, the downward settling flux of sediment due to gravity is balanced by upward, Fickian diffusive flux due to turbulence. The diffusion coefficient of sediment, \mathcal{E}_{sy} , whose value cannot be 53 54 directly measured, was simplified as a parabolic function (Rouse, 1937), a linear function or 55 constant (Larras, 1969), or a parabolic function or constant (van Rijn, 1984). Assuming a parabolic function $\varepsilon_{sy} = \beta \kappa u_* (1 - y/h) y$ (where u_* is the shear velocity, h is the water depth, and κ is 56

the Von Karman constant), in the solution for the classical Rouse Eq. (1) takes the following form
(Rouse, 1937; Vanoni & Brooks, 1957):

ω

(2)

$$\frac{S}{S_a} = \left[\frac{\frac{h}{y}-1}{\frac{h}{a}-1}\right]^{\beta \kappa u_*},$$

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where S_a is the reference concentration at reference level a, $\beta = \varepsilon_{sy}/\varepsilon_m$ is the inverse of the 60 sediment Schmidt number, and ε_m is the fluid eddy viscosity. The determination of β is an 61 62 important topic in the investigation of the vertical distribution of the suspended sediment 63 concentration. Previous literature has revealed that β depends on the normalized settling 64 velocity (Jain et al., 2018), may also depend on the reference level and reference concentration 65 (Pal & Ghoshal, 2016). It may be larger than 1 or less than 1 in different cases (Cellino & Graf, 66 2002). In current study, $\beta = 1$ is applied when using Eq. (1) for simplicity. Although the Rouse Eq. 67 (1) has a clear physical meaning and has been widely used to study sediment transport, many 68 studies showed that its solution Eq. (2) cannot accurately capture the sediment distribution near 69 the river bottom and surface (Chien & Wan, 1999; Otsuka et al., 2017).

70 The obvious shortcoming of the Rouse formula motivated various improvements in the last 71 two decades. For example, Wang and Ni (1991) proposed a particle velocity distribution function 72 in two-phase flows, and then derived a theoretical model for the vertical distribution of suspended 73 sediment concentration using the kinetic theory. Cao et al. (1995) analyzed the velocity and 74 sediment concentration distribution in open channel flows using the fundamental two-phase flow 75 equation. Zheng et al. (2012) proposed a continuous distribution formula for suspended sediment 76 concentration by modifying the van Rijn formula (van Rijn, 1984). Chen et al. (2013) found that 77 anomalous turbulent diffusion plays an important role in suspended sediment transport, and 78 further presented a fractional derivative advection-dispersion equation (FADE) model to describe 79 the vertical distribution of suspended sediment concentration in steady turbulent flows. The 80 fractional-derivative model can successfully describe the vertical distribution of suspended 81 sediment concentration based on turbulent diffusion in steady flow (Chen et al., 2013; Kundu, 82 2018; Nie et al., 2017). However, Sun and Chen (2009) argued that anomalous diffusion in

83 turbulence might be better described by the Hausdorff fractal derivative model (which is 84 introduced in detail in the next section) than the FADE model, because the Hausdorff fractal 85 derivative captures different dynamics for material transport. This conclusion, however, has not 86 been checked against any real-world observations of sediment dynamics. Moreover, although the 87 fractional derivative model has been successfully applied to describe anomalous diffusion, its 88 nonlocal nature requires high computational cost in numerical simulations and is not 89 easy-to-implement for hydrologists and engineers. These challenges may be solved by the new 90 model using the Hausdorff fractal derivative, and, hence, it is an interesting topic to develop and 91 check the fractal model to quantify the complex turbulence structure effects on dynamics of 92 suspended sediment in turbulent flow.

93 The Hausdorff fractal derivative model has been applied to describe complex dynamics 94 observed in natural systems, including anomalous diffusion (Sun et al., 2017; Liang et al., 2019), 95 soil moisture movement (Sun et al., 2013), turbulent flow (Sun & Chen, 2009), hydrodynamics 96 (Balankin & Elizarraraz, 2012), and other fields (Balankin et al., 2013; Cai et al., 2016; 97 Reves-Marambio et al., 2016). The Hausdorff fractal derivative (representing a fractal time-space 98 metric transform) is introduced from physics to describe anomalous diffusion (or transport) in a 99 fractal structure (Kanno, 1998; Chen, 2006). Since the turbulent flow structure over rough beds is 100 known to have fractal features, the Hausdorff fractal derivative model may provide an improved 101 physically-based description of anomalous turbulent transport, relative to available models. In addition, the stretched exponential function, which forms the fundamental solution of the 102 103 Hausdorff fractal derivative model, has been widely confirmed to characterize statistical physical 104 quantities (e.g., velocity, acceleration, structure, and energy) of turbulent flow (La Porta et al., 105 2001; Kellay, 2017; Zhou et al., 2006). These successful applications and statistical properties of 106 turbulence motivated the current study to extend the Hausdorff fractal derivative to characterize 107 the physical dynamics of suspended sediment transport in steady turbulent flows.

This paper is structured as follows. Section 2 introduces the Hausdorff fractal derivative and the derivation of the Hausdorff fractal derivative model with an analytical solution for suspended sediment transport and the equilibrium concentration distribution in the vertical direction. Section 3 checks the applicability of the newly derived model by fitting two sets of experimental observations for sediment with different sizes. Section 4 discusses parameters of the Hausdorff fractal derivative model, and compares the new model with the improved Rouse model and the standard FADE model using more experimental data. Section 5 summarizes the main conclusions and discusses applications for steady-state vertical dynamics of suspended sediment.

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2. Hausdorff fractal derivative model

117 The Rouse model (1) has two main limitations: 1) a debatable assumption of the Fickian type 118 of turbulent flux and empirical assumptions for the eddy diffusivity, and 2) the inability to extend 119 the suspended sediment distribution all the way to the bed (which results because the near-bed 120 flow structure is more complex owing to wakes and other flow structures resulting from bed 121 roughness). Various studies have confirmed that turbulence structure is complex and often exhibits 122 fractal properties, which satisfy the conditions for the application of the Hausdorff fractal 123 derivative (Lanotte et al., 2015; Sreenivasan & Meneveau, 1986). Further, it has been shown that 124 anomalous turbulent diffusion can be well characterized by the Hausdorff fractal derivative model 125 (Chen, 2006; Sun & Chen, 2009). Therefore, here for the first time the Hausdorff fractal derivative 126 is introduced into the Rouse model, leading to a new formula for quantifying the steady-state 127 vertical distribution of the suspended sediment concentration under steady turbulent flow, which 128 will then be compared in detail with previous models.

The Hausdorff fractal space-time fabric is obtained by metric transform of the standard integer-order space-time fabric. The metric transformation is proposed based on two hypotheses: fractal invariance (meaning that physical laws are invariant under the fractal transformation) and fractal equivalence (where the influence of anomalous environmental fluctuations on physical behavior is assumed to be equal to the fractal time-space transform) (Chen, 2006). Under these hypotheses, the Hausdorff fractal derivative can be restated as the normal derivative using the following metric transform:

136
$$\begin{cases} \hat{t} = t^{\alpha}, \\ \hat{y} = y^{\gamma}, \end{cases}$$
 (3)

137 where α and γ ($0 < \alpha, \gamma \le 1$) represent the order of the Hausdorff fractal derivative in time 138 and space, respectively. The Hausdorff fractal derivative indices α and γ describe the 139 Hausdorff dimensions of fractal time and fractal space, respectively. Here $\alpha = d_f \times d_{sc}$ and 140 $\gamma = d_f$, where d_f is the fractal dimension of space and d_{sc} is the fractal dimension of regions 141 excluding the holes in the fractal structure, based on the theoretical analysis in the (Kanno, 1998). 142 The metric transform Eq. (3) leads to the following definition of the Hausdorff fractal derivative 143 (Chen 2006; Sun et al., 2013):

144

$$\frac{\partial f(t)}{\partial t^{\alpha}} = \lim_{t_1 \to t} \frac{f(t_1) - f(t)}{t_1^{\alpha} - t^{\alpha}},$$

$$\frac{\partial g(y)}{\partial y^{\gamma}} = \lim_{y_1 \to y} \frac{g(y_1) - g(y)}{y_1^{\gamma} - y^{\gamma}}.$$
(4)

145 where f(t) is a function of variable t^{α} and g(y) is a function of variable y^{γ} .

To accurately characterize suspended particle motion in the turbulent flow structure with fractal dimension, it proposed to use the corresponding fractal space-time (y^{γ}, t^{α}) . The underling hypothesis is that the diffusion of suspended sediment follows Fick's Law in the fractal dimension (H^{γ}) . Under the steady-state condition (where the fractal time t^{α} can be deleted from the equation), the following Hausdorff fractal derivative based advection-dispersion equation (HADE) is obtained to model anomalous diffusion of suspended sediment:

152
$$\omega S + \overline{\varepsilon}_{sy} \frac{\partial S}{\partial y^{\gamma}} = 0, \qquad (5)$$

where $\overline{\varepsilon}_{sy}$ represents a depth-averaged diffusivity which can upscale the depth-dependent ε_{sy} in Eq. (1), considering the fact that anomalous diffusion usually happens in the entire model region (which is also called "nonlocal" in space). This effective diffusivity can be calculated by integrating the depth-variable $\overline{\varepsilon}_{sy}$ in the Rouse model from the reference height *a* (where *a* = 0.05 *h* is widely accepted) to the water surface (*y* = *h*) (Chen et al., 2013), and $\omega [LT^{-1}]$ is the sediment settling velocity (Zhang, 1998) which are expressed by

159
$$\overline{\varepsilon}_{sy} = \frac{\int_{0.05h}^{h} \kappa u_* \left(1 - y/h\right) y dy}{0.95h} = \frac{209 \kappa u_* h}{1200}, \tag{6}$$

160
$$\omega = \sqrt{\left(13.95\frac{\nu}{d}\right)^2 + 1.09\frac{\gamma_s - \gamma_w}{\gamma_w}gd} - 13.95\frac{\nu}{d}.$$
 (7)

161 where $v[L^2T^{-1}]$ is the kinematic viscosity; $g[LT^{-2}]$ is the gravitational acceleration; d[L]

162 is the particle diameter; γ_s and γ_w are the specific weights of sediment and water, respectively. 163 Under the condition $S(y=a)=S_a$, the analytical solution of Eq. (5) can be obtained by using 164 the variable substitution $(\hat{y} = y^{\gamma})$

165
$$\frac{S}{S_a} = e^{-\frac{\omega}{\overline{c}_{sy}}\left(y^{\gamma} - a^{\gamma}\right)}, 0 < \gamma \le 1.$$
(8)

Note that the Hausdorff fractal derivative order γ is a parameter in the analytical solution Eq. (8) which cannot be directly measured in the field or using existing flume experiments. Hence, the HADE model Eq. (5) contains one more parameter (γ) compared with the classical Rouse model Eq. (1). Notably, when $\gamma = 1$, the Eq. (8) reduces to the exponential expression that has previously been proposed for suspended sediment distributions based on empirical observations (Larras, 1969).

Figure 1 shows the distribution of S/S_a is sensitive to the order of the Hausdorff fractal derivative γ . In general, a smaller γ leads to a larger gradient of the vertical distribution of suspended sediment concentration (representing stronger turbulent diffusion), which can facilitate model fitting for river engineers using the observed concentrations of suspended sediment.

176

177 Fig. 1. The normalized concentration S/S_a for suspended sediment (calculated using Eq. (8)) with different fractal 178 indices γ . The other parameters (dimensionless in this case) are the same, which are $\kappa = 0.4$, $\bar{\epsilon}_{sy} = 0.007 (L^2/T)$, 179 $\omega = 0.0047 (L/T)$, and $u_* = 0.05 (L/T)$.

180 **3. Model applications and experimental data analysis**

181 To test the applicability of the HADE model Eq. (5) and its solution Eq. (8) in describing the 182 vertical distribution of suspended sediment concentration, we consider two groups of experimental 183 data were considered for natural sands (with different sizes) in fully developed, steady 184 open-channel flows. To evaluate the difference between the experimental data and the model 185 predictions, the root mean square error (RMSE) was used. More experimental data is checked in 186 the next section. The experiments done by Einstein and Chien (1955) and Coleman (1986) are 187 used for comparative purposes. The flow and sediment characteristics of the experiments are listed 188 in Table 1.

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Run	Depth,	Particle size,	Shear velocity,	Settling velocity,	Von Karman
Number	<i>h</i> (cm)	<i>d</i> (mm)	u_* (cm/s)	(cm/s)	Constant, K
S-1	13.8	1.300	11.47	13.96	0.322
S-2	12.0	1.300	12.85	13.96	0.261
S-3	11.7	1.300	13.26	13.96	0.246
S-4	11.5	1.300	14.28	13.96	0.210
S-6	14.3	0.94	11.82	11.29	0.295
S-7	14.3	0.94	11.79	11.29	0.281
S-8	13.9	0.94	11.53	11.29	0.263
S-9	13.5	0.94	11.85	11.29	0.247
Coleman 2	17.1	0.105	4.1	0.673	0.403
Coleman 6	17.0	0.105	4.1	0.673	0.410
Coleman 15	17.1	0.105	4.1	0.673	0.414
Coleman 16	17.1	0.105	4.1	0.673	0.432
Coleman 22	17.0	0.210	4.1	2.35	0.457
Coleman 23	17.0	0.210	4.1	2.35	0.453
Coleman 26	17.1	0.210	4.1	2.35	0.466
Coleman 27	16.8	0.210	4.1	2.35	0.446

192	Table 1. Flow and	l sediment character	istics in the exper	iments of Einsteir	n and Chien ((1955) and	Coleman (1986	6)
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194 *3.1. Experimental Case 1*

195 This experiment was conducted in two-dimensional, fully developed, steady open-channel 196 flows (Einstein & Chien, 1955). The mean size of sediment particles was 1.3 mm for experimental 197 runs S-1 ~ S-4 and 0.94 mm for S-6 ~ S-9, representing coarse and medium sand, respectively. 198 Detailed information on the experiment can be found in Fu et al. (2005). Here the applicability of 199 the HADE model Eq. (8) is examined and the HADE model is compared with the classical Rouse 200 model Eq. (2). The reference height is selected at a = 0.02 h. The best-fit results of the HADE 201 model (obtained by minimizing the RMSE) for the experimental datasets S-1 \sim S-4 and S-6 \sim S-9 202 are depicted in Figs. 2 and 3, respectively. Figures 2 and 3 show that the HADE model generally 203 matches the experimental data better than the Rouse model, especially for the regions near the 204 river bed and surface. The improved performance of the HADE model is attributed to its improved 205 ability to capture anomalous diffusion of sediment particles, which is driven by turbulent bursting 206 and cannot be efficiently characterized by the use of Fick's law for diffusive fluxes in the Rouse 207 model.

208

Fig. 2. Case 1: Comparison of the vertical distribution of suspended sediment concentration fitted by the two models (the HADE model Eq. (8) and the classical Rouse model Eq. (2)) and the experimental data. The mean sediment size is d = 1.3 mm, and the Hausdorff fractal order is $\gamma = 0.93$ for dataset S-1, and $\gamma = 0.99$ for datasets S-2 ~ S-4, with the reference height at a = 0.02 h (experimental data from Einstein and Chien (1955)).

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Fig. 3. Case 1: Comparison of the vertical distribution of suspended sediment concentration fitted by the two models (the HADE model Eq. (8) and the classical Rouse model Eq. (2)) and the experimental data. The mean sediment size is d = 0.94 mm, and the Hausdorff fractal order is $\gamma = 0.80$ for datasets S-6 and S-7, and $\gamma = 0.88$ for datasets S-8 and S-9, with the reference height a = 0.02 h (experimental data from Einstein (1955)).

218 *3.2. Experimental Case 2*

219 This experiment was done in a uniform flume, where the mean sediment size is 0.105 mm for 220 Coleman Runs 2, 6, 15, and 16, and 0.210 mm for Coleman Runs 22, 23, 26, and 27, belonging to 221 fine particles and medium particles (Coleman, 1986). Figures 4 and 5 provide the best-fit results 222 of the HADE model for the four runs of each experiment. Similar to Case 1, here the fitting results 223 also show that the HADE model can fit well the experimental data. The best-fit Hausdorff fractal 224 derivative order is $\gamma = 0.73$ for Coleman 2, $\gamma = 0.65$ for Coleman 6, $\gamma = 0.53$ for Coleman 15, 225 $\gamma = 0.51$ for Coleman 16, and $\gamma = 0.99$ for Coleman 22, 23, 26, and 27. This finding implies that 226 fine particles experience stronger super-diffusion than medium sediment particles, because a fine 227 particle has a higher probability to travel a long distance in turbulent flows.

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Fig. 4. Case 2: Comparison of the vertical distribution of suspended sediment concentration fitted by the two models (the HADE model Eq. (8) and the classical Rouse model Eq. (2)) and the experimental data, where the sediment diameter is d = 0.105 mm, the best-fit fractal order: (a) $\gamma = 0.73$; (b) $\gamma = 0.65$; (c) $\gamma = 0.53$; (d) $\gamma = 0.51$ and the reference height is a = 0.02 h (experimental data from Coleman (1986)).

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Fig. 5. Case 2: Comparison of the vertical distribution of suspended sediment concentration fitted by the two models (the HADE model Eq. (8) and the classical Rouse model Eq. (2)) and the experimental data, where the sediment diameter is d = 0.210 mm, the best-fit fractal order is $\gamma = 0.99$, and the reference height is a = 0.02 h (experimental data from Coleman (1986)).

238 4. Discussion

4.1. Impact of parameters on sediment profiles and their correlation in the HADE model

240 In this section, a number of available data sets are used to explore the impact of the Hausdorff fractal derivative order γ and the Rouse parameter ($\omega/\kappa u_*$) on vertical distributions of 241 242 sediment, as well as the relation between these two major parameters in the HADE model. Four 243 data sets representing a wide range of turbulent flow conditions and sediment sizes (from medium 244 silt to very coarse sand) were selected, including the experimental data derived from Einstein and 245 Chien (1955) (denoted as Case A for description simplicity), Coleman (1986) (denoted as Case B), 246 Wang and Qian (1989, 1992) (denoted as Case C), and the authors measurements at Jianli Station 247 of the Jingjiang River Reach (i.e., the upstream section of the Yongtze River), China (denoted as 248 Case D).

The fractal index γ in the HADE model is the core parameter describing the influence of turbulent bursting on suspended sediment, making estimation of this parameter important for application. The best-fit parameter γ listed in Table 2 varies from 0.51 to 0.99, depending on the turbulence characteristics and the sediment dimension.

253 The measured Rouse parameter, $\omega/\kappa u_*$, listed in Table 2, which is a ratio characterizing the 254 gravity and turbulent diffusion effects, ranges from 0.3800 to 4.6558. It affects the shape of the 255 vertical profile for suspended sediment. An increase in the Rouse parameter, representing 256 decreased dispersion relative to particle advection, results in a profile of suspended sediment with 257 higher nonuniformity. 258

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262 Table 2. Values for the best-fit Hausdorff fractal derivative index γ in the HADE model, the measured sediment

263 diameter, d, and the measured Rouse parameter ($\omega/\kappa u_*$). The detailed fitting results of the HADE model using the

experimental and field measurement data are provided in Appendix A.

D 1	Hausdorff fractal		Rouse parameter
Run number	derivative index γ	Particle size <i>d</i> (mm)	<i>ω</i> / <i>к</i> и _*
Al	0.93	1.300	3.7803
A2	0.99	1.300	4.1629
A3	0.99	1.300	4.2802
A4	0.99	1.300	4.6558
A5	0.80	0.940	3.2375
A6	0.80	0.940	3.4074
A7	0.88	0.940	3.7227
A8	0.88	0.940	3.8568
A9	0.82	0.274	0.9425
A10	0.85	0.274	1.3547
A11	0.86	0.274	1.4652
A12	0.86	0.274	1.2295
A13	0.88	0.274	1.4484
B1	0.73	0.105	0.4073
B2	0.65	0.105	0.4004
B3	0.53	0.105	0.3965
B4	0.51	0.105	0.3800
B5	0.99	0.210	1.2542
B6	0.99	0.210	1.2563
B7	0.99	0.210	1.2300
B8	0.99	0.210	1.2851
C1	0.98	0.150	0.6411
C2	0.97	0.150	0.6377
C3	0.96	0.150	0.6576
C4	0.99	0.960	0.5427
D1	0.98	0.375	1.2885
D2	0.80	0.175	1.6268
D3	0.78	0.075	1.1969
D4	0.56	0.025	0 7353

265

The foregoing experimental data, grouped based on the sediment size (see Table 2), show that

267 the Hausdorff fractal order, γ , increases approximately linearly with the Rouse parameter ($\omega/\kappa u_*$)

268 (Fig. 6). This positive relation might be due to the fact that a smaller γ or a smaller $\omega/\kappa u_*$ can 269 generate a more uniform vertical distribution of suspended sediment, which represents strong 270 super-diffusion (upward jumps) due to turbulent coherent structures. But a universal regression 271 model cannot be proposed based on the existing experimental data and limited knowledge. On one 272 hand, as shown by Fig. 1, the fractal order γ controls the overall trend of the sediment vertical 273 profiles. When $\gamma \rightarrow 1$ (i.e., weak turbulence), the suspended sediment distributes as an 274 exponential function with more mass near the river-bed. A smaller γ indicates more anomalous 275 transport of suspended sediment caused by stronger upward jumps that rapidly move the bottom 276 sediment upward. When $\gamma \rightarrow 0$ (i.e., strong turbulence), sediment distributes uniformly along 277 the vertical direction with a constant concentration $S/S_a = 1$. On the other hand, the Rouse 278 parameter, $\omega/\kappa u_*$, acts as a scaling factor that controls the expansion of the sediment vertical 279 profile: a smaller $\omega/\kappa u_*$ indicates more anomalous transport of suspended sediment caused by 280 stronger diffusive jumps (compared to advective jumps), which can smooth the sediment profile by decreasing the vertical concentration gradient. The sensitivity of the sediment vertical profile 281 to $K = \omega/\overline{\varepsilon}_{sy}$ is shown in Fig. 7. Figure 7 shows that a smaller $\omega/\overline{\varepsilon}_{sy}$, likely resulting from 282 283 stronger turbulence, enhances anomalous transport (i.e., fast displacement) and results in a more 284 uniform vertical concentration distribution.

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Fig. 6. The relation between the Hausdorff fractal derivative index γ (fitted) and the Rouse parameter $\omega/\kappa u_*$ (measured). The line represents the best-fit linear relation for the various datasets (symbols).

288

289 Fig. 7. Effect of $K = \omega/\overline{\varepsilon}_{sy}$ on the vertical distribution of suspended sediment (for Run S-8, d = 0.94**290** mm, $\gamma = 0.88$, and $K = \omega/\overline{\varepsilon}_{sy} = 153.7739$, of the experimental data from Einstein and Chien (1955)).

It can be seen that both the fractal order, γ , and the Rouse parameter, $\omega/\kappa u_*$, define the smoothness of sediment distribution. Notably, Fig. 6 implies that the suspension index, $\omega/\kappa u_*$, is not the only factor that can affect the Hausdorff fractal order. Other factors including the mean flow velocity and Reynolds number also play an important role in the random motion of suspended particles in turbulent flows, generating, therefore, the sediment profile and determining the value of γ .

297 4.2. Model comparison

298 The sediment Schmidt number, β , in the Rouse model relates the sediment diffusivity to the 299 fluid eddy viscosity under steady-state and uniform flow conditions. Adjusting β in the classical 300 Rouse model may reconcile the model solution to the measured data (Nie et al., 2017). In open 301 channel flows, β may characterize the complex interactions between the fluid and the solid 302 sediment grains in suspension. It may also act as a correction factor to fix the intrinsic modeling 303 error due to the inability to directly simulate all relevant fluid-particle interactions based on 304 real-world physical information (e.g., turbulence damping, hindered settling, and mobile-bed 305 effects). Hence, an improved model for the Rouse formula can be obtained by adjusting β .

306 Here the following formula is used to calculate β (Cheng et al., 2013)

$$\beta = \frac{1 + \lambda Z}{1 + \xi/Z}$$
(9)

308 where $Z = \frac{\omega}{\kappa u_*}$ is the Rouse parameter, the factor $\lambda = 1$, and $\xi = 0.034$. Pal and Ghoshal (2016)

309 developed a new expression for β , which is denoted as:

310
$$\beta = 2.204 \left(\frac{\omega}{u_*}\right)^{0.667} a^{0.178} S_a^{0.017}.$$
(10)

For comparison, the improved Rouse model and the M1 model has been obtained by using Eq. (9)and Eq. (10) to modify the Rouse model, respectively.

313

314 Fig. 8. Solutions of three models (HADE, Improved Rouse, and M1 models) compared with the experimental data. **315** The mean particle size is d = 0.94 mm, the Hausdorff fractal order is $\gamma = 0.93$, $\beta = 1.3657$ in the improved Rouse **316** model, $\beta = 1.4371$ in the M1 model, and a = 0.02 h (experimental data from Einstein and Chien (1955)).

The results show that the improved Rouse model can only adjust part of the sediment profile, especially the intermediate section, while the overall trend of the profile cannot be adjusted significantly (Appendix A). One example is shown in Fig. 8, where the improved Rouse model and M1 model still cannot capture the overall sediment concentration profile, and the HADE 321 model provides a better fit to the measured data than the improved Rouse model.

322 The FADE model has been used to describe the vertical distribution of suspended sediment in323 steady turbulent flow (Chen et al., 2013) as follows:

324
$$\omega S + \overline{\varepsilon}_{sy} \frac{\partial^{\eta} S}{\partial y^{\eta}} = 0, \qquad (11)$$

325 where η ($0 < \eta \le 1$) (dimensionless) denotes the order of the Caputo-type fractional derivative. 326 The analytical solution of Eq. (11) is as follows (Chen et al., 2013):

327
$$\frac{S}{S_a} = E_{\eta} \left(-\frac{\omega}{\overline{\varepsilon}_{sy}} (y-a)^{\eta} \right), \tag{12}$$

Atangana and Baleanu (2016) developed the fractional derivative with Mittag-Lefflerfunction kernel, which is defined as follows (Yu et al., 2018):

330
$${}^{ABC}_{a}D^{\eta}_{y}f(y) = \frac{B(\eta)}{1-\eta}\int_{a}^{y}f'(\tau)E_{\eta}\left[-\eta\frac{(y-\tau)^{\eta}}{1-\eta}\right]d\tau.$$
(13)

331 where $B(\eta)$ is a normalization function, ${}^{ABC}_{a}D^{\eta}_{y}$ is the Caputo fractional derivative 332 which defined by the Atangana and Baleanu (2016). f'(y) is the first derivative of 333 function f(y), τ is the integral variable. So the corresponding Rouse equation can be 334 expressed as:

$$\omega S + \overline{\varepsilon}_{sy} \, {}^{ABC}_{a} D^{\eta}_{y} S = 0 \tag{14}$$

336 Then the analytical solution of Eq. (14) can be expressed as:

337
$$\frac{S}{S_a} = \frac{1}{1 + \frac{\omega}{\overline{\varepsilon}_{sy}} - \frac{\omega}{\overline{\varepsilon}_{sy}}\eta} E_{\eta} \left[\frac{-\frac{\omega}{\overline{\varepsilon}_{sy}}\eta}{1 + \frac{\omega}{\overline{\varepsilon}_{sy}} - \frac{\omega}{\overline{\varepsilon}_{sy}}\eta} (y - a)^{\eta} \right].$$
(15)

338 where $E_{\eta}(z)$ denotes the Mittag-Leffler function, which can be written as:

339
$$E_{\eta}(z) = \sum_{k=0}^{\infty} \frac{z^{k}}{\Gamma(\eta k+1)}$$
(16)

340 The comparison results of three models is shown in Fig. 9.

341

 Fig. 9. Solutions of three models (HADE, Fractional, and ABC Fractional models) compared with the experimental data. The mean particle size is d = 1.3 mm, $\gamma = 0.93$ in the HADE model, $\eta = 0.91$ in the Fractional model, $\eta = 0.9996$ in the ABC Fractional model, and a = 0.02 h (experimental data from Einstein **345** (1955)).

346 The FADE model Eq. (11) differs from the HADE model Eq. (5) in at least two ways. First, 347 the FADE model considers the nonlocality of particle motion in turbulence, while the HADE 348 model is built upon a metric transform between normal and fractal structures. Second, the FADE 349 model leads to a Mittag-Leffler (or power-law) type of sediment distribution, while the HADE 350 model produces a stretched exponential distribution. The power-law density function for random 351 motion yields much larger jumps, whereas the stretched exponential density function is weighted 352 more towards smaller motions and is not heavy-tailed. This discrepancy results in the calculated 353 concentration of the HADE model declining relatively faster than the standard FADE model. 354 Suspended particles may move upward with coherent flow structures such as Kolk-Boil vortices 355 while undergoing settling, resulting in a travel path distribution "lighter" than the power-law 356 distribution assumed typically for cascade eddies, which suggests that the HADE model better 357 describes the underlying motion than the standard FADE model. It is also worthwhile to 358 emphasize that the stretched exponential density function is more physically justified here, since 359 turbulence structures have defined scales (i.e., they break up or dissipate to the background 360 turbulence). It is expected that this limited extent of non-locality can be better represented by a 361 stretched-exponential model (i.e., the HADE model).

362 Comparison among the HADE model, the FADE model (fractional derivative order, η), and 363 the improved Rouse model is shown in Fig. 10. The results show that the HADE model and the 364 FADE model match the data better than the improved Rouse model, and the HADE model gives 365 better fitting results than the FADE model. Generally speaking, the FADE model describes a pure 366 power-law distribution of vertical particle jumps, which overpredicts the vertical transport 367 observed in the sediment profiles. The improved Rouse model with a adjustable Schmidt number 368 can only slightly modify the intermediate range of the sediment profiles. The HADE model 369 describes the particle's vertical travel path distribution in the form of a stretched exponential 370 function, which better matches the measured concentration distributions. Direct measurements of particle vertical jumps under different flow conditions are needed to confirm the best
mathematical description for sediment vertical diffusion in turbulent flows, as well as to better
predict transport parameters (e.g., the Rouse parameter) for practical field applications in the
future.

375

376

377

378 Fig. 10. Solutions of three models (HADE, Fractional, and ABC Fractional models) compared with the **379** experimental data. (a) d = 1.3 mm, $\gamma = 0.93$ in the HADE model, $\eta = 0.91$ in the FADE model, and **380** $\beta = 1.3657$ in the improved Rouse model; and (b) d = 0.94 mm, $\gamma = 0.88$ in the HADE model, $\eta = 0.88$ in the **381** FADE model, and $\beta = 1.3599$ in the improved Rouse model. The parameter a = 0.02 h (experimental data from **382** Einstein and Chien (1955)).

383 *4.3. Physical interpretation of particle jumps and feasibility of the HADE model*

384 The main assumption applied in the HADE model Eq. (5) is that the jump size distribution 385 (or the travel path distribution during a single jump event) for suspended sediment particles 386 follows a stretched exponential function, which is relatively lighter than the heavy-tailed (i.e., 387 power-law) distribution assumed by the FADE model Eq. (11). Coherent turbulent structures (such 388 as Kolk-Boil vortices, which are the underwater vortices created by rushing water passing an 389 obstacle at the river-bed which can pluck sediment) may initiate from bed roughness elements and 390 propagate upward in the water column, resulting in random vertical transport for suspended 391 sediment particles that is "heavier" than the exponential distribution described by the classical 392 Fickian diffusive flux. The random vertical jump of sediment particles, therefore, might be 393 affected by the competition between gravity (which generates the settling velocity) and the stable 394 vortices. While some of the vortices can reach the free surface of the water and result in a 395 heavy-tailed distribution of travel paths over the entire water column, sediment particles might not 396 completely follow these structures because of the concurrent downward settling. Hence, some of 397 the particles may "drop out" of the vortex before reaching the water surface, resulting in a lighter 398 than heavy-tail or power-law distribution (such as the stretched exponential function) of vertical 399 particle jumps.

400 Limitations from the system geometry also suggest that a stretched exponential function is 401 more appropriate than a heavy-tailed power law in capturing suspended particle movement in a 402 finite bounded domain. The power-law distributed large displacements (with an infinite variance) 403 of sediment particles would require large-scale, long-term coherent structures throughout the 404 water column, but these structures are limited by the vertical boundaries, resulting in a new 405 distribution with a finite variance such as the stretched exponential function. Such a truncation is 406 common in hydrological processes, such as the truncated jumps for particles moving in fluvial 407 systems where the ancient channel and floodplain deposits have a finite size (Zhang, 2010; Zhang 408 & Meerschaert, 2011).

In summary, the HADE model may better capture suspended sediment dynamics than either the standard FADE model (which assumes pure power-law jumps for particles driven by eddy cascades) or the Rouse model (which assumes that Fickian diffusion governs upward particle motion). These hypotheses need further validation, especially additional observations or simulation of the vertical jump distributions for suspended particle transport in rivers.

It is also noteworthy that none of the suspended sediment transport studies that were analyzed in the current study provided observed turbulence structures or turbulence velocity fluctuations, and, hence, here only the qualitative relationship between the Hausdorff index, γ , and turbulence properties can be proposed. Scale analysis of the Hausdorff model and its index γ based on turbulence intensity and energy and space-time scales of coherent structures remains an open research question.

420 5. Conclusions

Suspended sediment exerts an important control on geomorphological and biogeochemical processes in rivers. Efficient quantification of suspended sediment dynamics remains a challenge. A non-Fickian transport model is proposed as an alternative to the classical Rouse model to capture complex dynamics for suspended sediment in steady sediment-laden flows. Model development, numerical analysis, real-world validation, and model comparison reveal the following three main conclusions.

427 First, a HADE model, which is built upon the Hausdorff fractal derivative and can be solved428 analytically, is proposed to capture the steady-state vertical distribution of suspended sediment.

429 This novel stochastic model contains a single fitting parameter, the fractal order γ , to efficiently 430 characterize non-Fickian dynamics likely due to turbulent flows. The fractal order can also capture 431 the size-selective behavior of suspended sediment dynamics.

Second, the fractal order controls the overall shape of the suspended sediment concentration profile, and the Rouse parameter (which is a measurable parameter in the HADE model) controls the spatial expansion of this profile. These two parameters are strongly correlated, which may improve the prediction performance of the HADE model. The full predictability of the complex stochastic process for suspended sediment, however, requires much more direct information on the motion of suspended particles in turbulent flows.

438 Third, model comparisons show that the HADE model can fit the sediment concentration 439 profiles better than both the improved Rouse formula and the standard FADE model, which either 440 under- or over-estimates the vertical displacement of sediment particles in coherent turbulent 441 structures. The competition between gravity and large-scale, long-term vortices, or the finite river 442 depth, can limit large jumps of sediment particles and result in a jump size distribution "lighter" 443 than the power-law function and "heavier" than the exponential function, motivating the 444 application of the stretched exponential function in the HADE model. Further tests and 445 experimental data are needed to check the feasibility of the HADE model for different river flows 446 and morphologies.

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451 Appendix A. Additional model results and comparison

This appendix shows the best-fit results using the HADE model Eq. (5) and the other models for experimental and field measurement data listed in Table 2. Figures A1, A2, and A3 show the model results for the experimental sediment profiles measured by Einstein and Chien (1955), Wang and Qian (1989, 1992), and the authors' field test, respectively. The RMSEs for the HADE model Eq. (5), the FADE model Eq. (11), and the improved Rouse model are listed in Table A1. Both the figures and RMSEs indicate that, for most cases, the HADE model can fit the sediment 458 concentration profiles better than the improved Rouse formula and the FADE model, a conclusion459 consistent with that found in the main text.

460 To evaluate the fitting results of the models, the root mean square error (RMSE) is calculated 461 as follow:

462
$$RSME = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (S_{i0} - S_{ie})^2}$$
(17)

463 where S_{i0} represents the measured sediment concentration, S_{ie} represents the simulated 464 sediment concentration, and N is the number of measured concentration data at the observation 465 point.

466

467 Fig. A1. Comparison of the vertical distribution of suspended sediment concentration computed using three **468** models (HADE, Fractional (FADE), and Improved Rouse models), and the experimental data from Einstein and **469** Chien (1955) for d = 0.274 mm. The best-fit parameters are: (a) $\gamma = 0.82$ in the HADE model, $\eta = 0.81$ in the **470** fractional-derivative model, and $\beta = 1.0562$ in the improved Rouse model; (b) $\gamma = 0.85$, $\eta = 0.84$, **471** and $\beta = 1.1077$; (c) $\gamma = 0.86$, $\eta = 0.84$, and $\beta = 1.1205$; (d) $\gamma = 0.86$, $\eta = 0.84$, and $\beta = 1.0927$; and (e) **472** $\gamma = 0.88$, $\eta = 0.86$, and $\beta = 1.1186$.

473 Fig. A2. Comparison of the vertical distribution of suspended sediment concentration computed using three 474 models (HADE, Fractional (FADE), and Improved Rouse models), and the experimental data from Wang and Qian 475 (1989, 1992) for the sediment sizes d = 0.150 mm for SQ1 ~ SQ3, and d = 0.96 mm for SM1. The best-fit 476 parameters are: (a) $\gamma = 0.98$ in the HADE model, $\eta = 0.96$ in the fractional-derivative model, and $\beta = 1.0105$ in 477 the improved Rouse model; (b) $\gamma = 0.97$, $\eta = 0.95$, and $\beta = 1.0099$; (c) $\gamma = 0.96$, $\eta = 0.95$, and $\beta = 1.0134$; and 478 (d) $\gamma = 0.99$, $\eta = 0.97$, and $\beta = 0.9921$.

479

480 Fig. A3. Comparison of the vertical distribution of suspended sediment concentration computed using three 481 models (HADE, Fractional (FADE), and Improved Rouse models), and field data from Jingjiang River at Jianli. 482 (a) The mean sediment size d = 0.375 mm, $\gamma = 0.98$ in the HADE model, $\eta = 0.98$ in the fractional-derivative 483 model, and $\beta = 1.0998$ in the improved Rouse model; (b) d = 0.175 mm, $\gamma = 0.80$, $\eta = 0.90$, and $\beta = 1.1389$; (c) 484 d = 0.075 mm, $\gamma = 0.78$, $\eta = 0.85$, and $\beta = 1.0888$; and (d) d = 0.025 mm, $\gamma = 0.56$, $\eta = 0.63$, 485 and $\beta = 1.0621$.

495 Table A1. The root mean square errors of fitting results using the HADE model, FADE model and the improved **496 Provide and Chief (1955)** SOL \sim S

497	and SM1 from Wang and Qian (1989	, 1992), and Jianli1 ~	- Jianli4 from the Jianli Station	on the Jingjiang River).
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Run number	RMSE (HADE model)	RMSE (FADE model)	RMSE (Improved Rouse model)
S-1	0.1072	0.1324	0.4076
S-8	0.2745	0.3034	0.9391
S-11	0.0312	0.0310	0.0613
S-12	0.1605	0.2220	0.5689
S-13	0.4639	0.5167	0.7777
S-14	0.2238	0.6448	1.2038
S-15	0.9531	1.3673	2.3313
SQ1	0.2052	0.1816	0.1932
SQ2	0.5886	0.4736	1.3918
SQ3	0.4390	0.3892	2.1554
SM1	0.0070	0.0075	0.0197
Jilanli 1	0.0122	0.0124	0.0030
Jilanli2	0.0268	0.0203	0.0358
Jilanli3	0.0176	0.0142	0.0299
Jilanli4	0.0729	0.0568	0.1931

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