

1 **Interactive role of rolling friction and cohesion on the angle of repose**
2 **through a micro-scale assessment**

3
4 **Thao Doan**

5 *BEng (Hons)*

6 PhD Candidate, Transport Research Centre, University of Technology Sydney, NSW 2007. Australia.

7 ORCID ID: <https://orcid.org/0000-0003-0229-8376>

8 **Buddhima Indraratna¹**

9 *PhD (Alberta), MSc (Lond.), BSc (Hons., Lond.), DIC, FTSE, FIEAust., FGS, CEng, CPEng*

10 Distinguished Professor of Civil Engineering, Founding Director of Australian Research Council's Industrial
11 Transformation Training Centre for Advanced Technologies in Rail Track Infrastructure (ITTC-Rail), Director
12 of Transport Research Centre, School of Civil and Environmental Engineering, University of Technology
13 Sydney, NSW 2007, Australia.

14 ORCID ID: <https://orcid.org/0000-0002-9057-1514>

15 ¹Corresponding author e-mail: buddhima.indraratna@uts.edu.au

16 **Thanh Trung Nguyen**

17 *PhD, CPEng, M.ASCE, M.EAust*

18 Research Fellow, Transport Research Centre, School of Civil and Environmental Engineering, University of
19 Technology Sydney, NSW 2007, Australia.

20 ORCID ID: <https://orcid.org/0000-0001-6078-2559>

21 **Cholachat Rujikiatkamjorn**

22 *PhD, MEng (AIT), BEng (Hons)*

23 Professor, Transport Research Centre, School of Civil and Environmental Engineering, University of
24 Technology Sydney, NSW 2007, Australia.

25 ORCID ID: <https://orcid.org/0000-0001-8625-2839>

26 **Abstract**

27 Cohesion and rolling friction play key roles in governing the behaviour of soil, however, only
28 a limited number of studies have been able to assess the simultaneous contributions of these
29 two micro-parameters on the macro-properties of soil. In this respect, the innovation of the
30 current study includes an attempt to examine the interplay of these two primary parameters on
31 the angle of repose (AoR) based on the discrete element method (DEM). Lifting cylinder tests
32 on cohesive wet sand have been carried out in DEM, while the cohesion and rolling friction
33 are captured through proposed computational models. In this paper, macro-parameters such as
34 the geometry and developmental stages of sand piles obtained in DEM simulation are compared
35 to experimental data, while their micro-evolution is quantified in detail. The results show that
36 a large AoR can only be obtained when the cohesive and rotational frictional forces work in
37 tandem. Increasing the cohesion and rolling friction results in smaller contact numbers with
38 increasing chain-like connections between particles and larger pore spaces to account for a
39 larger AoR. For the first time, this study distinctly identifies three major stages that contribute
40 to the AoR, based on the development of contact numbers and the transformation of energy.
41 Accordingly, the linkage between macro-scale AoR and the micro-structural coordination
42 number is formulated with varying levels of cohesion and rolling friction. The DEM results
43 prove that the more cohesive the particles are, the greater the delay in the dissipation of kinetic
44 energy.

45

46 *Keywords:* cohesion, angle of repose, energy transformation, rolling friction, cohesive forces

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48

49 **1. Introduction**

50 Soils are inherently particulate, which means their macroscopic behaviour is governed by
51 collective microscopic interactions such as the frictional and cohesive contacts between
52 discrete particles. In essence, cohesion arises in various forms that mainly include van der
53 Waals forces between fine particles (Chen and Anandarajah 1996; Hamaker 1937), capillary
54 forces in wet particles (Roy et al. 2016; Schwarze et al. 2013), and electrostatic forces
55 (Matsusaka et al. 2010). Friction induced by sliding and rolling components is an essential
56 aspect of particulate geomaterials due to the irregular shapes and surfaces of particles in nature.
57 Obviously, the larger the angularity and roughness of particles, the larger the frictional
58 resistance. A great deal of effort through advanced experimental and numerical approaches has
59 been made to characterize the role that friction and cohesion has on the macro-behaviour of
60 geomaterials. Of those studies, the use of discrete element method (DEM) has proved to be an
61 effective and reliable approach of advancing our knowledge of various particle-scale
62 behaviours of soil; for example, non-linear deformation under applied loads (Yimsiri and Soga
63 2010); stress state evolution (O’Sullivan 2011; Potyondy and Cundall 2004); granular breakage
64 and settlement (Chen et al. 2022; Einav 2007; Huang and Tutumluer 2011; Nguyen et al. 2017);
65 filtration and internal erosion (El Shamy and Zeghal 2005; Indraratna et al. 2021; Nguyen and
66 Indraratna 2020; Tao and Tao 2017; Wu et al. 2021; Yin et al. 2021; Zhao and Shan 2013); and
67 among others. Nevertheless, most previous DEM studies have been carried out on cohesionless
68 granular soils even though cohesion has had a significant influence on the response of soil in a
69 wide range of contexts. This means that more effort is needed in DEM to simulate cohesive
70 soils more accurately and realistically.

71 Despite the clear physical mechanisms of particle cohesion, the effort to include
72 cohesive forces in DEM models remains limited due to its complexity and excessive
73 computation. Hence, it is common practice in DEM to utilize cohesion contact models such as

74 the JKR (Johnson-Kendall-Roberts) (Johnson et al. 1971) and the DMT (Derjaguin-Muller-
75 Toporov) (Derjaguin et al. 1975) models where different attractive forces are presented through
76 several simplified parameters. While JKR theory is considered to be more appropriate for soft
77 materials with large particles where the attractive forces are observed experimentally to only
78 appear inside the enlarged contact area, DMT theory is better suited to hard materials where
79 attraction only appears on the outside of the contact region (Barthel 2008; Prokopovich and
80 Perni 2011). Past studies (Johnson and Greenwood 1997; Li et al. 2011) have stated that the
81 JKR equation gives a decent prediction of contacting area with cohesive forces, compared to
82 the various experimental conditions.

83 In contrast to the above, the frictional behaviour of particles and its impact on the
84 macro-behaviour of soils have been investigated extensively; for example, the shear resistance
85 (Belheine et al. 2009; Muftah and Gutierrez 2010), granular flows (Goniva et al. 2012; Shan
86 and Zhao 2014), and rockfalls (Gentilini et al. 2012; Thoeni et al. 2014). Previous micro-
87 scanning studies (Nguyen and Indraratna 2020; Zhou et al. 2017) have shown that the
88 roundness and angularity indices of particles such as sand can vary from 0.5 to 0.8, a range that
89 can result in having a significant frictional resistance. The contribution made by rolling and
90 sliding friction on the overall response of geomaterials can change considerably depending on
91 the context, thus requiring specific justification when modelling. In highly confined systems
92 where soil shears under confinement, the sliding friction is significant, whereas rolling
93 resistance can dominate in uncompacted materials where particles have greater degrees of
94 freedom. For instance, when rock particles are falling down a slope, their rolling resistance
95 which depends mainly on the shape characteristics of the particles will govern how the particles
96 travel and transport kinetic energy in the system.

97 Although previous DEM studies have investigated the effect that rolling friction and
98 cohesion has on the behaviour of soil, their combined influence on the response of soil has

99 rarely been studied, especially in the case of fine or wet particles where cohesive forces become
100 increasingly more dominant in relation to the self-weight of particles. Therefore, this study
101 aims to close this gap by examining the importance of these parameters on the angle of repose
102 (AoR), which is one of the most fundamental parameters used to characterize the frictional
103 response of geomaterials in a naturally forming state, i.e., the steepest angle that soil can make
104 in its loosest state, without slumping (Terzaghi 1943). This parameter is also used to present
105 the morphology and internal friction of particulate materials, both of which have valuable
106 implications for practical designs such as slope stability and retaining walls. Previous
107 investigations (Grima 2011; Roessler and Katterfeld 2019) reveal that rolling friction and
108 cohesion play significant roles in AoR of geomaterials, however, these studies normally use
109 single values for these parameters and evaluate them individually, which gives an incomplete
110 understanding of their interactive role in governing soil behaviour. It is noteworthy that while
111 sliding friction can also affect the response of soil particles in forming the AoR, this influence
112 is usually insignificant compared to rolling friction and cohesion (Carstensen and Chan 1976;
113 El-Kassem et al. 2021; Meier et al. 2019; Zhou et al. 2001).

114 The current study employs DEM incorporated into the open-source software named
115 LIGGGHTS (Kloss and Goniva 2011) to investigate the interactive roles of cohesion and
116 rolling friction on the angle of repose. The macro-scale results are then validated against the
117 experimental data, while the variation of micro-scale parameters is captured and analyzed.
118 Based on these outcomes, this study attempts to link microscale elements to the conventional
119 macro-parameter AoR, while tracking the energy transformation during the formation of AoR,
120 which are key innovations.

121 2. Theoretical background

122 The governing equations of the discrete element method (DEM) used to model particle motion
 123 are given by:

$$m_i \frac{d\mathbf{v}_i}{dt} = \sum_{j=1}^{n_i^c} (\mathbf{F}_{ij}^n + \mathbf{F}_{ij}^t + \mathbf{F}_{ij}^{coh}) + \mathbf{F}_i^g \quad (1)$$

$$I_i \frac{d\boldsymbol{\omega}_i}{dt} = \sum_{j=1}^{n_i^c} (\mathbf{R}_i \times \mathbf{F}_{ij}^t + \mathbf{M}_{ij}^r) \quad (2)$$

124 where m_i and I_i are the mass and moment of inertial particle i respectively; \mathbf{v}_i and $\boldsymbol{\omega}_i$ are the
 125 translational and angular velocities of particle i ; \mathbf{F}_{ij}^n and \mathbf{F}_{ij}^t are the normal and tangential
 126 contact forces induced on particle i by particle j ; \mathbf{F}_{ij}^{coh} is the cohesive force between particle i
 127 and particle j while \mathbf{F}_i^g is the gravitational force acting on particle i ; \mathbf{R}_i is the radius vector
 128 connecting the centre of particle i and the point of contact where the tangential force is initiated;
 129 \mathbf{M}_{ij}^r is the rolling resistance torque induced to oppose the rotational interaction between
 130 particles i and j ; and n_i^c is the number of contact of particle i .

131 Previous studies (Li et al. 2005; Zhu et al. 2007) have successfully used the nonlinear
 132 Hertz-Mindlin contact model to simulate granular materials, so this contact model is adopted
 133 in this current study. The normal and tangential contact forces (i.e., \mathbf{F}_{ij}^n and \mathbf{F}_{ij}^t) derive from
 134 the corresponding contact stiffnesses (i.e., k_n and k_t) as follows:

$$k_n = \frac{4}{3} E^* \sqrt{R^* \delta_n} \quad (3)$$

$$k_t = 8G^* \sqrt{R^* \delta_n} \quad (4)$$

135 where E^* , G^* and R^* are the equivalent Young's modulus, shear modulus, and the radius of
 136 two particles respectively, and δ_n is the normal overlap or displacement vector. The
 137 determination of these parameters can be found in other studies (Li et al. 2005; Zhu et al. 2007).

138 In this study, to include the effect of rolling resistance, the elastic-plastic spring-dashpot
 139 (EPSD2) was used (Ai et al. 2011). Wensrich and Katterfeld (2012) stated that the EPSD2
 140 model can be modified to accommodate a higher value of rolling stiffness; this can overcome
 141 the oscillatory behaviour and eliminate the need to consider rolling damping. The rolling
 142 resistance torque proposed by Iwashita and Oda (1998) can be computed with respect to the
 143 rolling friction coefficient μ_r as follows:

$$|\mathbf{M}_{ij}^r| \leq \mu_r R^* |\mathbf{F}_{ij}^n| \quad (5)$$

144 One of the most common cohesive models is the JKR model (Johnson et al. 1971),
 145 which has been modified from the classic Hertz (1881) contact model to consider the tensile
 146 forces at the contact edges. This model uses the concept of surface energy to represent the
 147 collective action of all the cohesive forces (i.e., the van der Waals, electrostatic, and capillary
 148 forces). The JKR model has been used extensively across different disciplines to model fine
 149 and wet sticky materials (Barthel 2008; Deng and Davé 2013; Louati et al. 2019). The
 150 governing equations of the JKR model are described as follows (Barthel 2008):

$$\mathbf{F}_{ij}^{coh} = \frac{4E^* a^3}{3R^*} - 2\sqrt{2\pi E^* \gamma a^3} \quad (6)$$

$$\delta_{ij}^{coh} = \frac{a^2}{R^*} - \sqrt{\frac{2\pi a \gamma}{E^*}} \quad (7)$$

151 where \mathbf{F}_{ij}^{coh} and δ_{ij}^{coh} are the cohesive force and the deformation vector based on JKR theory
 152 between particles i and j ; a is the enlarged contact radius, and γ is the equivalent surface energy.
 153 For simplicity, the JKR model can be modified into a linear cohesion model called SJKR, in
 154 which the surface energies, elastic modulus, and contact radius are combined into a single
 155 numerical parameter called Cohesion Energy Density (CED). In this simplified approach, the
 156 cohesive force is a product of CED (J/m^3) and the contact area A which is given by:

$$\mathbf{F}_{ij}^{coh} = CED \cdot A \quad (8)$$

$$A = \pi a^2 = \frac{\pi}{4d^2} (4d^2 R_1^2 - (d^2 - R_2^2 + R_1^2)^2) \quad (9)$$

157 where R_1 and R_2 are the radii of two contacting particles, and d is the distance between the
158 centroids of the particles.

159 **3. Numerical procedure**

160 Due to its simplicity, the lifting cylinder test (or slump test) is the preferred method for
161 determining the AoR of geomaterials (Lajeunesse et al. 2004; Rackl et al. 2017). A series of
162 lifting cylinder tests were therefore carried out using DEM to investigate the combined effect
163 of cohesion and rolling friction on the AoR. Sandy soils in dry and wet states commonly used
164 in previous laboratory investigations were also used in this current study. By conducting lifting
165 cylinder tests on various particle and cylinder sizes, Roessler and Katterfeld (2019) concludes
166 that the use of up-scaled particles is a timesaving approach without a significant effect on the
167 AoR value. On this basis, up-scaled particles were also used in this current investigation (Fig.
168 1). The SJKR model was used to simulate the cohesive behaviour of wet sand where the main
169 underlying mechanism is governed by the presence of liquid bridges which leads to the
170 development of capillary and viscous forces. While the SJKR model can be used to capture the
171 capillary force, the effect of viscous forces can be compensated for by using the low restitution
172 coefficient (i.e., 0.2), as demonstrated in previous studies (Behjani et al. 2017).

173 Initially, a sample of 7000 particles was generated randomly to fill a cylinder and then
174 allowed to stabilize under gravity. The insertion phase was satisfied when the particles became
175 fully static (i.e., the average velocity of particles was less than 0.5 mm/s). After stabilization,
176 the cylinder was lifted at 0.008 m/s to maintain a quasi-static flow condition and prevent the
177 particles from plashing downwards (Hassanzadeh et al. 2020). The simulation was completed
178 when the kinetic energy of the pile reached a minimal value. A schematic of this lifting cylinder
179 test is shown in Fig. 2. When the soil particles are completely contained inside the cylinder,

180 the boundary provides confinement through the particle-wall (P-W) contact that helps to
181 sustain particle frictional contact (P-P contacts) and prevent them from moving (see Fig. 2a).
182 It is noteworthy that the particle-wall contact was assumed the same for both sidewall and floor.
183 When the cylinder is lifted upwards, the particles lose their contacts due to the loss of boundary
184 confinement (see Fig. 2b), and then migrate laterally under gravitational force. It is apparent in
185 this process that the particles roll over each other and become displaced radially. As the
186 cohesion and/or rolling friction increases, the P-P contact forces are enhanced and the AoR
187 changes accordingly. The following sections present the numerical results with varying
188 cohesion and rolling friction, and an extensive analysis and discussion.

189 Since this study mainly focuses on rolling friction and cohesion where their input
190 parameters were varied, other DEM parameters were assumed based on selected previous
191 studies. For example, experimental validation by Roessler and Katterfeld (2019) showed that
192 the particle-wall friction coefficient could vary from 0.35 for the contact between sand and
193 steel to 0.37 for the contact between sand and plastic. Accordingly, the current study adopted
194 an average value of 0.36. Moreover, the shear modulus value was reduced to 10×10^7 (Pa) to
195 enhance computational efficiency, whilst noting that this assumption would not affect the
196 contact behaviour of sand particles with respect to the unconfined state of soil during an AoR
197 test (Lommen et al. 2014). The DEM input parameters are summarised in Table 1.

198 To approximate the range of rolling friction, this study followed the empirical
199 formulation provided in Ucgul et al. (2014), which resulted in the coefficients of rolling friction
200 varying from 0.05 to 0.4 for sand, with a radius varying from 0.1 to 10 mm. Since in previous
201 studies (Derakhshani et al. 2015; Liu et al. 2018; Phan et al. 2021) where the coefficient of
202 rolling friction for granular materials varied from 0.1 (relatively rounded particles) to 0.5
203 (crushed angular particles), the same range was adopted for the current numerical simulations.
204 The selection of CED values was based on previous DEM studies which simulated wet and

205 sticky granular materials. For example, Grima (2011) correlated the value of CED based on an
206 approximation of tensile stress (Rumpf 1962), and then proposed a CED from 0 to 3.5×10^6
207 (J/m^3) for wet washed coal. Roessler and Katterfeld (2019) estimated a CED from 4.6×10^6 to
208 5.4×10^6 (J/m^3) for wet sands, with a moisture content of 9.5% and a bulk density of 2083
209 kg/m^3 . The current study used a CED ranging from 0 for non-cohesive materials to 5×10^6
210 (J/m^3) for highly cohesive assemblies. The range of rolling friction coefficients and CED are
211 shown in Table 2.

212 **4. Results and Discussion**

213 4.1 Comparison of DEM results with experimental observation

214 Figure 3 shows four phases of the simulated lifting cylinder test for a given lifting height (h),
215 which is the gap between the lower boundary of the cylinder and the horizontal base (i.e., $h =$
216 90 mm, 120 mm, 160 mm and 280 mm for a 375mm high cylinder). The result shows that the
217 flow characteristics of the simulated wet sand are comparable with the experimental outcomes
218 by Roessler and Katterfeld (2019), as highlighted in the enclosed areas marked with a red dash.
219 Generally, there are four different stages in forming the angle of repose (AoR) considering
220 experimental observations. The first stage ($h = 90\text{mm}$), shows a stable column of material in
221 the DEM simulation without any bend or materials breaking out. This observation agrees with
222 the experimental findings by Roessler and Katterfeld (2019) where repeated tests showed that
223 wet sand had a consistent shape for the given lifting height. As the gap increases to 120 mm, a
224 slight bending of wet sand could be seen numerically and experimentally (Stage 2). With a
225 160mm height gap, the simulations and experiments indicate that the materials spread outwards
226 to form a convex slope (Stage 3), and when the gap reaches 280mm (Stage 4), there is a stable
227 formation of cohesive sand as indicated experimentally and numerically. The four stages in the
228 development of AoR of wet sand (Fig. 3) corroborated well through both experimental and

229 DEM simulations. The only difference between the numerical and experimental results was
230 that the numerical method could not simulate the cracks occurring while the wet sand expanded
231 laterally due to the non-uniformity distribution of water content and liquid-bridging forces
232 between soil particles in the experiment.

233 To obtain a quantitative comparison, for a lifting height between 80 mm to 140 mm (i.e.,
234 $h = 80, 100, 120$ and 140 mm), the pile diameters (d_{pile} , defined by a yellow horizontal dash
235 line in Fig. 4a) corresponding to each 20 mm predefined slice height were extracted to evaluate
236 the spread of materials outside the column. The relative pile diameters (d_{rel}) were computed
237 relative to the diameter of the cylinder ($d_{cylinder}$) to demonstrate the percentage of convex
238 bending of the sand pile. The formulation of d_{rel} is determined as follows:

$$d_{rel} = \frac{d_{pile} - d_{cylinder}}{d_{cylinder}} \times 100 (\%) \quad (10)$$

239 Figure 4b shows that the relative diameters of the piles for a given lifting height, as
240 obtained by the current simulations and experiments, agree very well. For instance, with a
241 lifting height where $h=80$ mm, the simulated pile remains almost vertical with 2.3% convex
242 bending, compared to 2.1% bending in the laboratory measurements. As the lifting height
243 increases to 140mm, the percentage of bending in the simulation increases to 25.1%, compared
244 to 25% in the test case. Exceeding the gap height further led to inconsistent bending of the
245 material (Roessler and Katterfeld 2019), and therefore it was not considered. In essence, the
246 DEM simulations successfully captured the main features of the experimental outcomes (see
247 Fig. 3) and depicted the quantitative relationship between the lifting height and the percentage
248 of convex bending (Fig. 4b).

249 4.2 The effect of rolling friction coefficient and cohesion on the angle of repose

250 Based on the final configuration of the simulated piles, the final height and the average radius
251 of the pile were computed to determine the approximate values of the angle of repose (AoR)

252 with respect to the method given by Kermani et al. (2015). As summarised in Table 3 and Table
253 4, rolling friction coefficients and cohesion have a considerable effect on the value of AoR,
254 such that as the rolling friction and cohesion increase, the final pile is higher and slimmer,
255 implying that the angle of the slope is higher. Table 5 shows that without cohesion (i.e., CED
256 = 0), AoR varies between 27 to 35 degrees for a system of spherical particles where the
257 coefficient of rolling friction varies from 0.1 to 0.5, respectively. This range of AoR agrees
258 with the average angle of the slope of dry sand piles reported in previous experimental studies;
259 for example, 34⁰ in (Beakawi Al-Hashemi and Baghabra Al-Amoudi 2018; Derakhshani et al.
260 2015) and 29.7⁰ for Toyoura sand (Nakashima et al. 2011). This result attests to the dependency
261 of the macroscopic AoR on the microscopic rolling resistance of individual particles
262 (Derakhshani et al. 2015; He et al. 2021; Hoshishima et al. 2021). A higher rolling resistance
263 means particles consume more kinetic energy, which leads to the formation of heaps with a
264 high potential (Zhou et al. 2001). Moreover, the opposition to the rotational movement of
265 particles helps prevent them being rearranged due to gravity, which leads to a higher build-up
266 of particles.

267 Table 5 also shows that an increasing level of cohesion gives rise in the AoR for all
268 levels of rolling friction. Several past studies (Mason et al. 1999; Mitarai and Nori 2006)
269 indicate that for a given material, the slope angle of a wet assembly is always higher than a dry
270 system, however, they did not investigate how varying the degree of cohesion can affect the
271 response of this slope angle. The simulated AoR can reach from 40 to 45⁰, which agrees with
272 the values of wet cohesive sand piles reported previously (Beakawi Al-Hashemi and Baghabra
273 Al-Amoudi 2018). Nase et al. (2001) observed the effect of cohesion on the flowability of
274 materials and indicated that the more cohesive it is, the less “flowable” is the system because
275 the additional cohesive force induced by liquid prevents particles from sliding and rolling on
276 an inclined surface and helps to form a stable structure.

277 The contour graph in Fig. 5 shows the coupled influence of rolling friction and cohesion
278 on the AoR outcomes. Notably, each contour line has a similar tendency where a steep slope
279 initially occurs followed by a flattening behaviour when the rolling friction or cohesion
280 becomes dominant. This postulates that when the rolling friction is insignificant, the AoR
281 results do not vary as much with the levels of cohesion, and vice versa. Therefore, a large AoR
282 can only be accomplished when the rolling friction and cohesion work in tandem.

283 4.3 Variations of porous feature with degrees of cohesion and rolling friction

284 The overall porosity is an important macroscopic parameter which is governed by microscopic
285 parameters such as particle size and inter-particle forces. Given its significance, this study
286 applied a numerical algorithm based on the Monte-Carlo approximation to estimate the
287 porosity within a volumetric region. The theoretical background of a Monte-Carlo
288 approximation can be found elsewhere (Amberger et al. 2012; Campillo et al. 2021). To
289 implement Monte-Carlo algorithms, a Python program is developed to apply a ‘hit-and-run’
290 approach and arbitrary walks into the representative volume element (RVE) of the final pile of
291 sand. The RVE samples were chosen to be sufficiently large enough to contain at least 1000
292 particles, and they proved to yield macroscopically reasonable results (Wiącek et al. 2012).
293 With this technique, the porosity of a certain region can be computed without knowing its
294 specific geometric features and the boundary conditions which are usually required in
295 conventional analytical methods.

296 Figure 6 shows the variation of porosity which was initially 0.54 and changed with
297 different levels of rolling friction coefficient and cohesion. It illustrates that a more porous
298 structure is obtained when the rolling resistance and cohesive forces become larger. With a
299 higher coefficient of rolling friction, the porosity increases proportionally irrespective of the
300 level of cohesion. Specifically, without cohesion (i.e., the black solid line), porosity changes

301 from 0.586 to 0.602 when the rolling friction coefficient increases from 0.1 to 0.5. The same
302 influence also repeats across different levels of cohesion. For example, in the case of very small
303 rolling resistance (i.e., $\mu_r = 0.1$), increasing CED from 0 to 5×10^6 (J/m³) causes an increase in
304 porosity from 0.586 to 0.600. This is understandable because when cohesion increases, the
305 particulate structure of the material becomes progressively separated with larger voids (Deng
306 and Davé 2013; Parteli et al. 2014; Thakur et al. 2014; Yang et al. 2000; Yu et al. 2003).

307 To elucidate the insightful mechanism of particle rearrangement of sand pile under
308 increasing cohesion, cross-sections of the cohesionless and cohesive sand piles are shown in
309 Fig. 7. When observing the contact numbers marked with different colours (i.e., from blue to
310 red for smaller to larger contact numbers respectively), the number of touching particles
311 increases from the edge of the pile to the centre. This phenomenon occurs because the central
312 portion of the pile is mostly formed by vertically deposited particles, which undoubtedly have
313 less freedom to displace, and hence contact more frequently with their neighbours. The result
314 also indicates that when cohesion is present, the soil particles tend to bond to each other by
315 attraction which promotes increasing chain-like connections and a more heterogeneous porous
316 structure. For example, Fig. 7 shows that the particles and pores of the cohesionless sand pile
317 are distributed more homogeneously, whereas larger pores are formed more frequently by
318 chain-like connections (highlighted in red rectangles). This finding is corroborated with several
319 past studies (Rognon et al. 2007; Xu et al. 2007) where cohesion creates a more non-uniform
320 porous structure of particulate media. However, none of them showed quantitative changes in
321 porosity due to the combined effect of cohesion and rolling friction given in this current study.
322 The large voids observed in numerical cohesive systems can be correlated to a phenomenon
323 known as increasing open micro-fabrics with cohesion, as reported in past experiments with
324 mixtures of sand and clay (Goudarzy et al. 2021; Gratchev et al. 2006).

325 4.4 Micro-evolution of forming the angle of repose

326 In order to quantify changes in the micro-features of soil while forming the angle of repose
327 (AoR), the total contact number (n_c), which was computed and outputted through DEM contact
328 algorithm, represents how many contacts that a particle or group of particles have with each
329 other. The larger the contact number, the more robust the particulate structure. Figure 8 shows
330 how the total number of contacts in a sand pile change over time with different degrees of
331 cohesion and rolling friction, based on current DEM investigations. These results show that
332 despite different magnitudes of cohesion and rolling friction, there are three major stages in
333 forming AoR. First, n_c decreases with time and there is no significant impact between cohesion
334 and rolling friction during this initial stage, but then it experiences a state where the value
335 fluctuates with time (i.e., a chaotic state) and the effects of cohesion and rolling friction become
336 apparent. Finally, n_c becomes stable when the particles form stable slopes in accordance with
337 different input values of cohesion and rolling friction. It is interesting that with a higher rolling
338 friction, soil particles take longer to reach a chaotic stage; for example, when $\mu_r = 0.1$, it takes
339 about 6.5s whereas when $\mu_r = 0.4$, it takes 10s. Moreover, the variation in rolling resistance
340 leads to a change in the number of contacts. Figure 8 also shows that when μ_r increases from
341 0.1 to 0.4, the number of contacts decrease from 2.94×10^4 to 2.75×10^4 in the initial stage. In the
342 context of small rolling resistance (i.e., $\mu_r = 0.1$), the number of contacts decreases
343 significantly from 2.8×10^4 to 2.6×10^4 when the CED increases from zero (i.e. cohesionless) to
344 5×10^6 (i.e. cohesion-dominated systems) (Fig. 8a). This declining number of contacts with
345 cohesion also occurs when the rolling friction increases (i.e., $\mu_r = 0.4$), as shown in Fig. 8b.
346 This micro-scale development shows three major stages in forming AoR, unlike the four stages
347 in the macro-scale (experimental) observations shown in Fig. 3. In fact, stages 1 and 2, where
348 soil particles have not moved very much, as shown earlier in Fig. 3, can be combined into the
349 initial stage based on the evolution of contact number.

350 The coordination number CN (i.e., an average number of contacts per particle) is
351 generally computed based on the total contact number (n_c) and particle number (n_p), as follows
352 (Thornton 2000):

$$CN = \frac{2n_c}{n_p} \quad (11)$$

353 Figure 9 shows a two-dimensional contour map to represent the interactive roles that rolling
354 friction and cohesion can have on the coordination number CN. In the absence of cohesion,
355 increasing the rolling friction from 0.1 to 0.5 leads to a reduction in the mean CN from 8 to
356 7.3. This range of CN values corresponds to the results obtained from previous studies on
357 cohesionless materials (Yang et al. 2000; Yu et al. 2003). The coupled effect of rolling friction
358 and cohesion also causes the CN to experience a continuous decline to reach values between 6
359 and 7 when both factors increase.

360 To further explain the interaction between cohesive and rolling resistant forces, as they
361 affect the coordination number, the equilibrium of all external and internal forces that act on
362 particles is explored. Yu et al. (2003) indicated that it is common for cohesionless granules to
363 have an average of six contacts per particle to stabilize the gravitational influence. However,
364 with increased rolling friction and cohesion, the combined magnitude of cohesive and frictional
365 forces is sufficient to resist the effect of gravitational forces. Therefore, stability can be
366 achieved with fewer interparticle contacts, which can be as minimal as two contacts per particle
367 (Xu et al. 2007). Figure 10a & Figure 10b show the distribution of contact numbers under
368 variations of cohesion and rolling friction, respectively. For $\mu_r = 0.2$, highly cohesive material
369 (i.e., $CED = 5 \times 10^6$) has the largest frequency of distribution, i.e., around 3 contacts per particle,
370 whereas non-cohesive material (i.e., $CED = 0$) experiences a more distributed peak of around
371 4 contacts per particle (Fig. 10a). On the other hand, without cohesion, Fig. 10b shows that
372 with lower rolling friction (i.e., $\mu_r = 0.1$ shown as a solid black line), the frequency curve tends
373 to change to the right to account for the higher frequency of larger contact numbers (i.e., > 4

374 contact/particle). In general, the peak of the distribution curve shifts to the left with a lower
375 contact number when the cohesion and rolling friction of the system increase.

376 4.5 Energy transformation and particle displacement process

377 The process of energy transformation is theoretically recognized as an inherent characteristic
378 of dynamic particulate systems. Since the microscopic details of particles can be tracked over
379 space and time, DEM is now used to compute the energetic components of granular soils such
380 as the potential and kinetic energies (PE and KE) (Kermani et al. 2015; Nguyen et al. 2020;
381 Nguyen and Indraratna 2020). The potential and kinetic components are computed as follows:

$$E_p^t = \sum_{i=1}^{n_p} m_i g z_i^t \quad (12)$$

$$E_k^t = E_k^{translational} + E_k^{rotational} = \sum_{i=1}^n \frac{1}{2} m_i (v_i^t)^2 + \frac{1}{2} I_i (\omega_i^t)^2 \quad (13)$$

382 where z_i^t is the height of particle i with respect to the horizontal datum at time t .

383 When the simulated particles have settled inside a cylindrical column with a stabilized
384 kinetic energy, the total amount of energy is considered to be equal to the potential energy at
385 rest ($t=0$). The cylinder is then lifted, making the column of particles start to collapse and
386 resulting in an instantaneous exchange between the potential and kinetic energy of particles.
387 Figure 11a & Figure 11b reveal an apparent shift in the magnitude of potential energy over
388 time using different levels of cohesion (i.e., CED ranges from 0 to 5×10^6) for two representative
389 cases of rolling friction 0.1 and 0.4. When the rolling friction is small (i.e., 0.1), the variations
390 of potential energy remain almost constant for the first 6-8 seconds, they then decrease
391 significantly for 8-16 seconds and finally reach a constant value after 16 seconds (see Fig. 11a).
392 This consistent trend of potential energy makes it a reliable criterion for signifying the stability
393 of granular heaps, as also recognized elsewhere (Zhou et al. 1999). Moreover, the magnitude

394 of potential energy for different levels of cohesion and rolling friction can be used to quantify
395 the slope feature of granular piles. For instance, the potential energy of a highly cohesive pile
396 (i.e., $CED = 5 \times 10^6$) has a higher value than for non-cohesive ($CED = 0$) and moderately
397 cohesive cases (see Fig. 11a & Fig. 11b). The higher potential energy at the final stage means
398 the slope has a steeper angle.

399 Figure 12a & Figure 12b show the variations of total kinetic energy (KE) in
400 cohesionless and cohesive systems at two levels of rolling friction (i.e., $\mu_r = 0.2$ and 0.3). This
401 shows that the peak values of kinetic energy in the cohesionless system are higher than in cases
402 with cohesion. The larger the cohesion, the smaller the peak value of KE; this is because
403 increasing cohesive forces provides additional resistance against the movement of particles,
404 and hence reduces the kinetic energy. However, this variation cannot be generalized in every
405 case of cohesion and rolling friction due to difficulties in maintaining a quasi-static condition
406 and preventing the sticky particles from avalanching. Furthermore, the kinetic energy fluctuates
407 greatly with localised peaks when the wet sand column began to collapse. These fluctuations
408 are the result of a massive difference in magnitude between the computation of potential and
409 kinetic energies, and thus, with a slight change in particle elevation, the kinetic energy can be
410 varied significantly. It takes much longer to reach the peak and stabilize KE when cohesion is
411 added to the system, and this indicates that more time is needed for KE to dissipate in cohesive
412 material. For example, Fig. 12a shows that the cohesionless system takes about 17.5s to fully
413 stabilize, whereas cohesive systems take around 19 to 20s to reach the same state.

414 The total displacement of all particles in the sand piles is computed across different
415 levels of cohesion and rolling friction to further enhance our understanding of the evolution of
416 kinetic energy in the system. Figure 13 shows the results where $\mu_r = 0.2$ and $\mu_r = 0.3$, with
417 $CED = 0$ and 5×10^6 , both of which corroborate the behaviour of kinetic energy shown earlier.
418 Those particles without cohesion (black line) begin to migrate and reach the first peak earlier

419 than in the cohesive cases. For example, Fig. 13a shows that cohesionless particles begin to
420 displace at around 6s, whereas soil particles with cohesion ($CED = 5 \times 10^6$) begin to migrate at
421 approximately 7.5s. However, cohesionless particles stop travelling (stabilizing) at 18s, rather
422 than the 21s when cohesion is included. Total displacement is highly susceptible to fluctuating
423 peaks due to the sudden avalanching of cohesive and highly frictional particles. Figure 13b also
424 shows that the largest peak of displacement that cohesionless particles can make is much higher
425 than for the cases with cohesion; for example, 49.5m for non-cohesive material compared to
426 35 m for cohesive particles (i.e., $CED = 5 \times 10^6$). This indicates that the material becomes less
427 brittle, i.e., the material reaches to lower peak deformation and takes longer time to stabilize
428 the kinetic energy when cohesion is added.

429 4.6 Correlation between macro- and micro-parameters

430 This analysis has shown that the bulk-scale angle of repose (AoR) and particle-scale
431 coordination number (CN) of the sand piles are a function of the levels of cohesion and rolling
432 friction. Therefore, by appraising the overall dataset, the relationship between these two
433 parameters (i.e., AoR and CN) in the two typical cases (i.e., the lower ($\mu_r = 0.1$) and upper (μ_r
434 $= 0.5$) rolling resistance with varying cohesion) can be formulated, as shown in Fig. 14a & Fig.
435 14b. These figures show that the fitted regression lines describe the relationship between AoR
436 and CN very well, with a percentage of error less than 5%. These equations also illustrate the
437 negative dependency of AoR on CN, thus indicating that the higher the average number of
438 contacts per particle, the smaller the AoR values. It is generally understood that the average
439 connection between each particle and its neighbours rises with densification, and vice versa
440 (O'Sullivan 2011). This means that when CN increases, the gravity-induced densification and
441 rearrangement of particles are enhanced considerably. In fact, as the effect of gravity increases,
442 particles are more prone to moving, which leads to a smaller AoR. By contrast, as cohesion

443 increases, the governing effect of gravity and the process of densification are compromised,
444 leading to higher AoR and smaller CN values (i.e., when the CED increases, the AoR-CN
445 relationship follows an upward trend to the left, as shown in Fig. 14a & Fig. 14b). This
446 generalization is also applicable for all the other levels of rolling friction where μ_r ranges from
447 0.1 to 0.5.

448 **5. Conclusion**

449 This study adopted the discrete element method (DEM) to investigate how cohesion and rolling
450 resistance could simultaneously affect soil behaviour considering the angle of repose (AoR).
451 Macro-parameters such as the geometry of the sand piles predicted by DEM compared quite
452 well with the experimental data, while the microscopic parameters were computed to further
453 understand the micro- and energy transformation of the systems. The innovation of this study
454 was that the systemic variations of cohesion and rolling friction in DEM allowed their coupled
455 effects on the macroscopic and microscopic levels to be captured, which were then used to
456 provide a linkage between the conventional AoR and the micro-scale elements. The salient
457 findings of this study are summarised as follows:

- 458 • The simulation results proved that the AoR was a function of cohesion and rolling
459 resistance, as represented by the contour plot. Increasing cohesion and rolling friction
460 resulted in a larger AoR, which can, for instance, reach 44° when $CED = 5 \times 10^6$ and $\mu_r =$
461 0.5. When either of these parameters was insignificant, the dependency of AoR on another
462 was greatly reduced. Therefore, an angle with a steep slope can only be accomplished when
463 both cohesive and rotational frictional forces are high.
- 464 • The overall porosity of sand piles increased regardless of the level of cohesion when the
465 rolling friction increased, and vice versa. The cross-section of cohesive piles showed that
466 the soil became more heterogeneous with increasing chain-like bonds between particles

467 when cohesion increased, leading to larger porosity and a higher AoR. The coordination
468 number at the peak frequency tended to decrease as cohesion between the particles of soil
469 increased.

470 • The micro-analysis showed that the total and average contacts in the system (i.e., n_c and
471 CN) significantly decreased when cohesion and rolling resistance dominated. When
472 resistance was enhanced from cohesive and frictional actions, the particles needed less
473 inter-particle contacts to counteract the gravitational forces, and thus resulting in a larger
474 AoR. Based on the numerical data, empirical relationships ($R^2 > 0.98$) between the macro-
475 parameter AoR and the micro-parameter coordination number (CN) were established for
476 the first time by considering different degrees of cohesion and rolling friction.

477 • The variations in particle contact (via n_c) indicated that there were three major stages in
478 forming the angle of repose, AoR. They were, (i) the initial stage where n_c dropped over
479 time without any significant effects of cohesion and rolling friction, (ii) the chaotic stage
480 during which n_c fluctuated significantly and the influences of cohesion and rolling friction
481 became apparent, and (iii) the final stabilizing stage where n_c and the corresponding AoR
482 set totally different levels in accordance with different degrees of cohesion and rolling
483 friction.

484 • The analysis of energy transformation showed that the residual potential energy was larger
485 as cohesion and/or rolling friction increased, which corroborated with the higher value of
486 AoR. The kinetic energy and the corresponding displacements of particles became less
487 significant as cohesion and/or rolling friction increased. The more cohesive the particles,
488 the greater the delay in kinetic energy being triggered and dissipated (i.e., the less brittle
489 the particle response), and the corresponding displacement.

490

491 **Data Availability Statement**

492 Some or all data, models or code that support the findings of this study are available from the
493 corresponding author upon reasonable request.

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496 Government through the Australian Research Council's Linkage Projects funding scheme
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498 SMEC, Sydney Trains, ACRI and Coffey are greatly appreciated.

499 **Tables**

500 Table 1: DEM input parameters of simulated study

Parameters	Inputs	References
Particle density (kg/m ³)	2650	
Poisson's ratio	0.3	
Shear modulus (Pa)	10x10 ⁷	
Restitution coefficient	0.2	(Behjani et al. 2017)
Particle-wall friction coefficient	0.36	(Roessler and Katterfeld 2019)
Particle-particle sliding friction	0.5	(Nguyen and Indraratna 2020; Phan et al. 2021)
Timestep (s)	10 ⁻⁵	10% Rayleigh time step
Gravitational acceleration (m/s ²)	9.81	

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502

503

Table 2: DEM calibration parameters

Parameters	Value range	Value increment
Rolling friction coefficient (μ_r)	0.1 - 0.5	0.1
Cohesion Energy Density (CED) (J/m^3)	0 - 5×10^6	1×10^6

504

505

Table 3: The final height of the material formation with varying cohesion and rolling friction

		Height of the final pile (m)				
CED (J/m³)	Rolling friction	0.1	0.2	0.3	0.4	0.5
	0		0.1541	0.1644	0.1771	0.1910
1x10 ⁶		0.1542	0.1688	0.1788	0.2015	0.2078
2x10 ⁶		0.1553	0.1763	0.1850	0.2102	0.2126
3x10 ⁶		0.1594	0.1801	0.1911	0.2148	0.2196
4x10 ⁶		0.1702	0.1849	0.2030	0.2196	0.2251
5x10 ⁶		0.1703	0.1917	0.2107	0.2237	0.2273

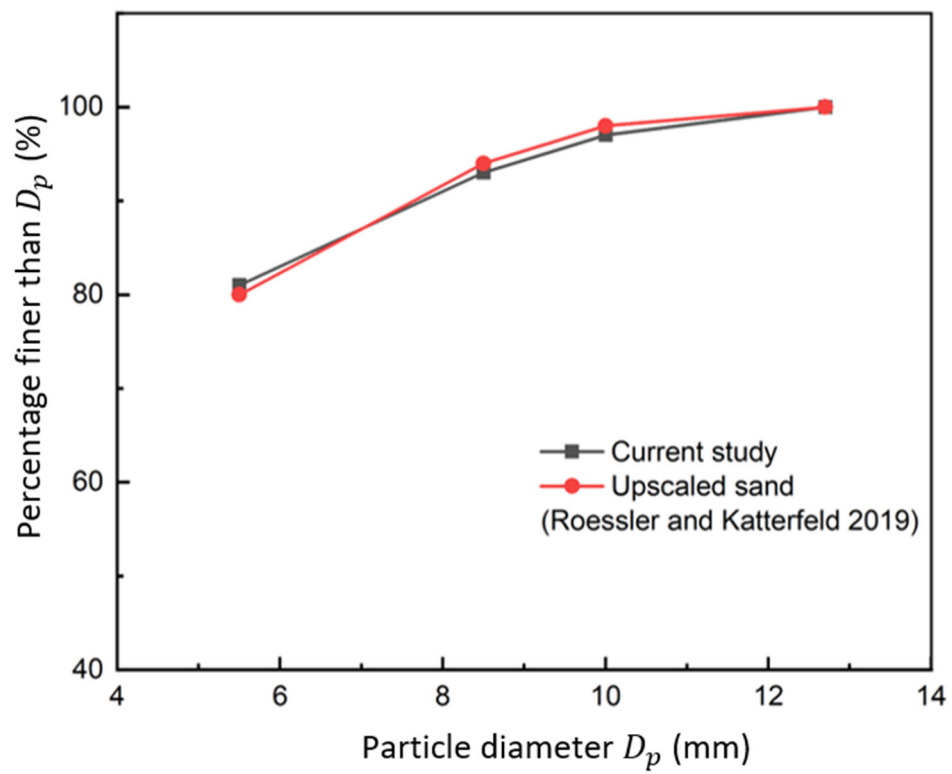
Table 4: The average radius of the material formation with varying cohesion and rolling friction

		Radius of the final pile (m)				
CED(J/m³)	Rolling friction	0.1	0.2	0.3	0.4	0.5
0		0.3019	0.2858	0.2843	0.2725	0.2704
1x10 ⁶		0.2883	0.2775	0.2826	0.2705	0.2651
2x10 ⁶		0.2880	0.2726	0.2800	0.2655	0.2620
3x10 ⁶		0.2883	0.2682	0.2730	0.2601	0.2584
4x10 ⁶		0.2829	0.2670	0.2672	0.2401	0.2421
5x10 ⁶		0.2674	0.2616	0.2567	0.2471	0.2217

Table 5: The variations of AoR with various levels of cohesion and rolling friction

		Angle of repose (AoR) (degrees)				
CED(J/m³)	Rolling friction	0.1	0.2	0.3	0.4	0.5
	0		27.04	29.91	31.92	35.03
1x10 ⁶		28.06	31.31	32.32	36.68	38.09
2x10 ⁶		28.34	32.89	33.45	38.37	39.06
3x10 ⁶		28.94	33.88	34.99	39.20	40.36
4x10 ⁶		31.03	34.70	37.23	41.82	42.92
5x10 ⁶		32.49	36.23	39.38	42.15	45.71

512 **Figures**

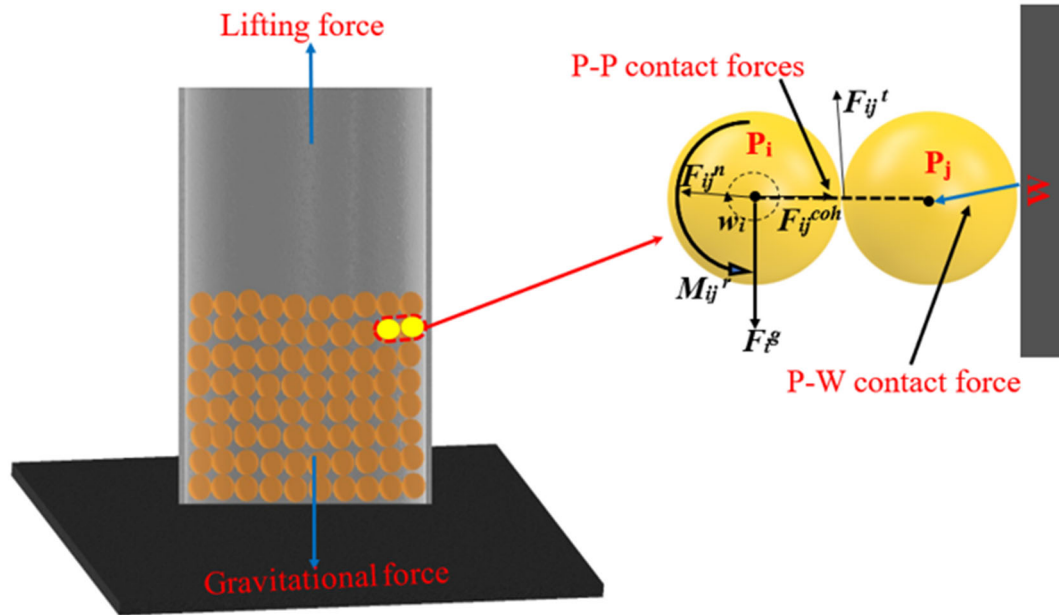


513

514

Fig. 1. Particle size distribution of simulated soil

a)



b)

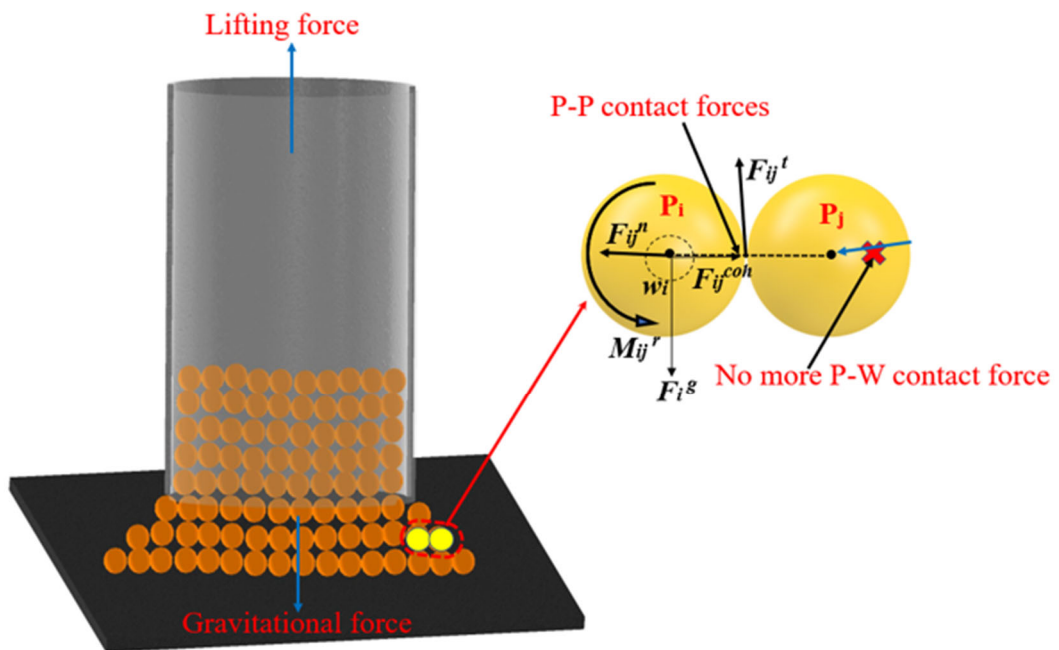


Fig. 2. The lifting cylinder test a) before lifting, b) during lifting

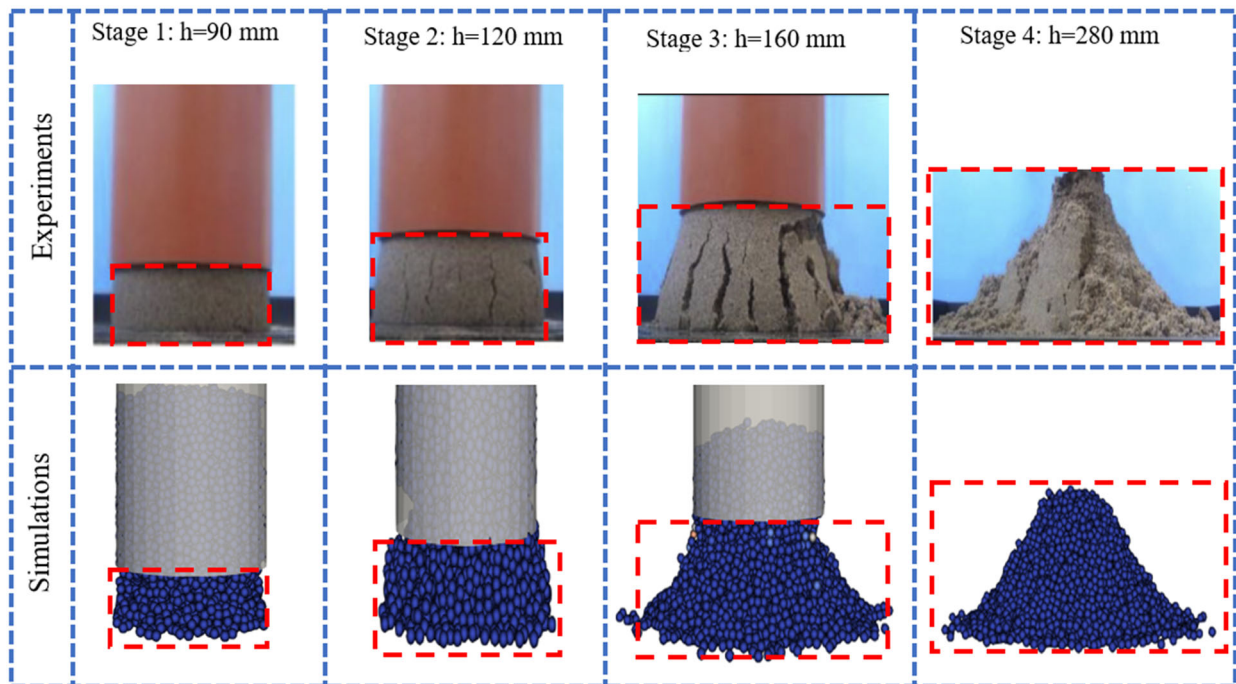


Fig. 3. Validation of the current simulated DEM (i.e., $\mu_r = 0.5$, $\mu_s = 0.5$ and $CED = 5 \times 10^6 \text{ J/m}^3$) with previous experiment results (Roessler and Katterfeld 2019)

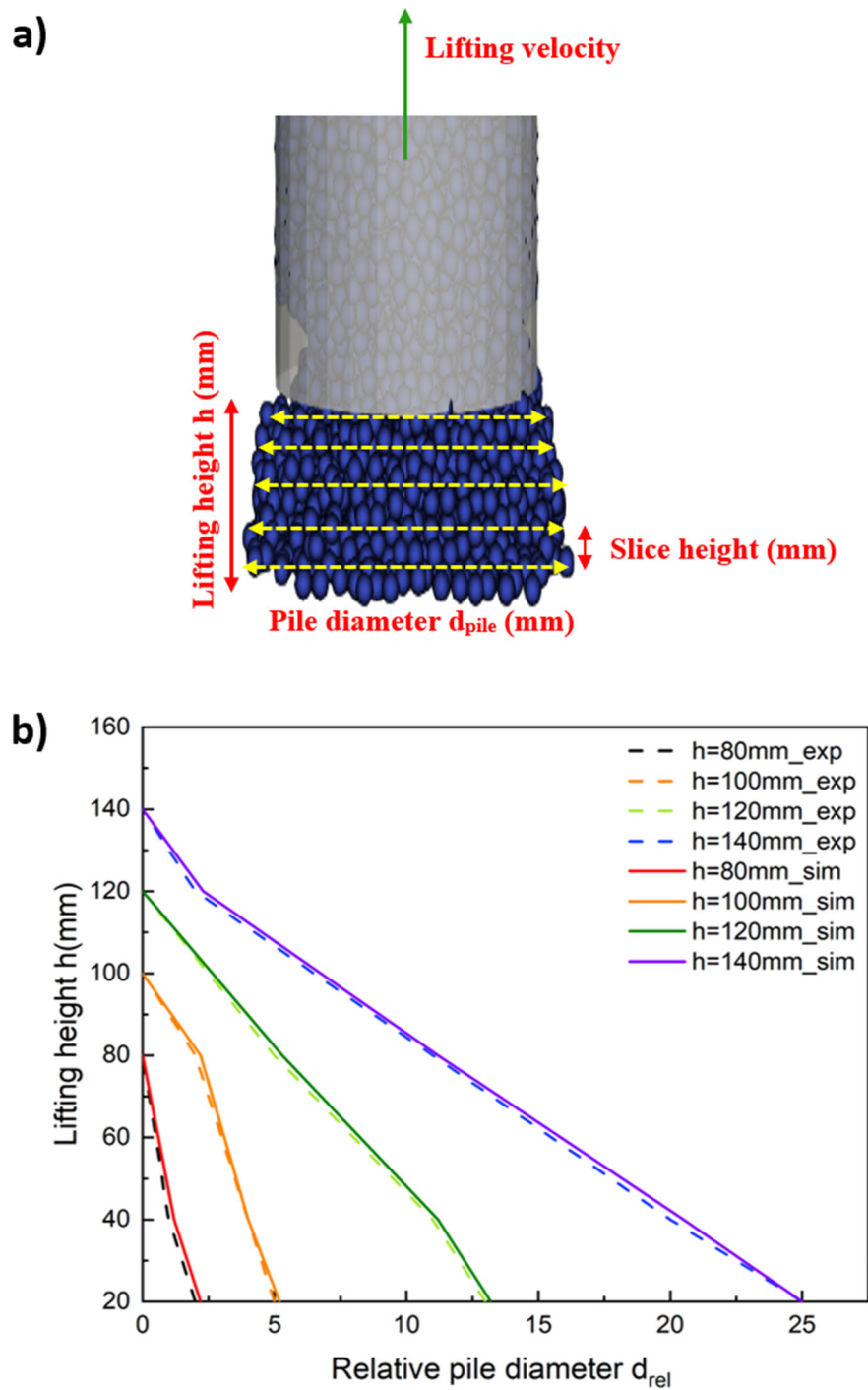


Fig. 4. a) Definition of lifting height and pile diameter, b) Lifting height versus relative diameter predicted by the current DEM in comparison with previous results by (Roessler and Katterfeld 2019)

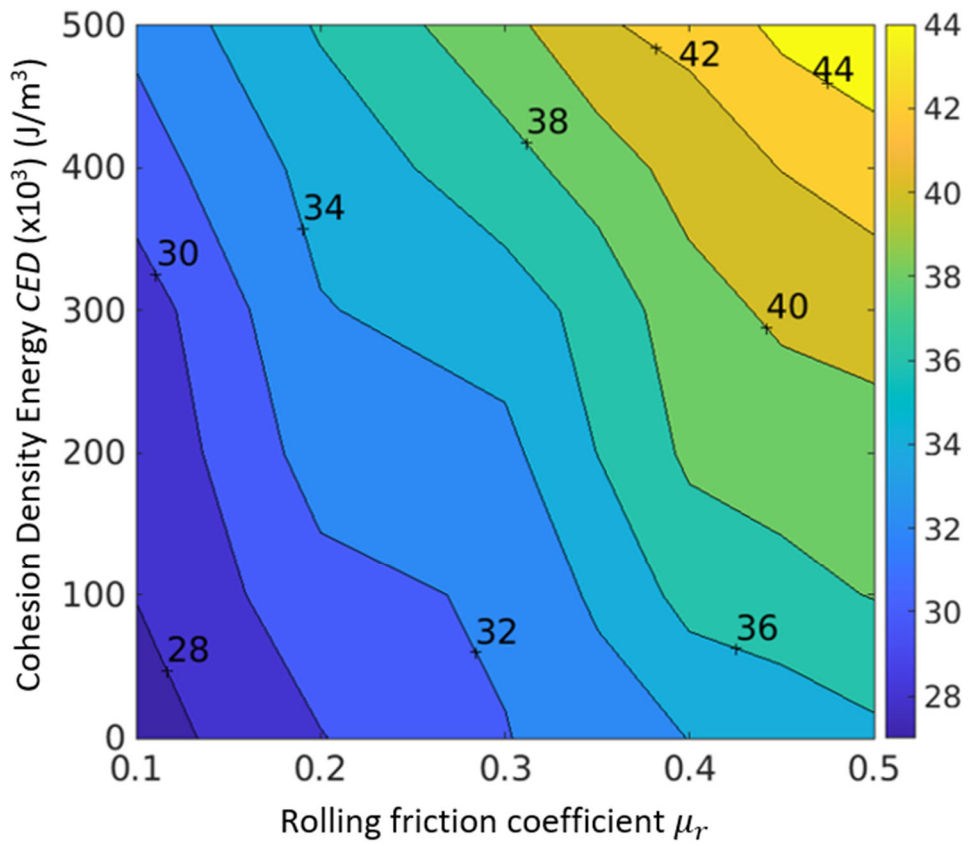


Fig. 5. Angle of repose as a function of cohesion (CED) and rolling friction

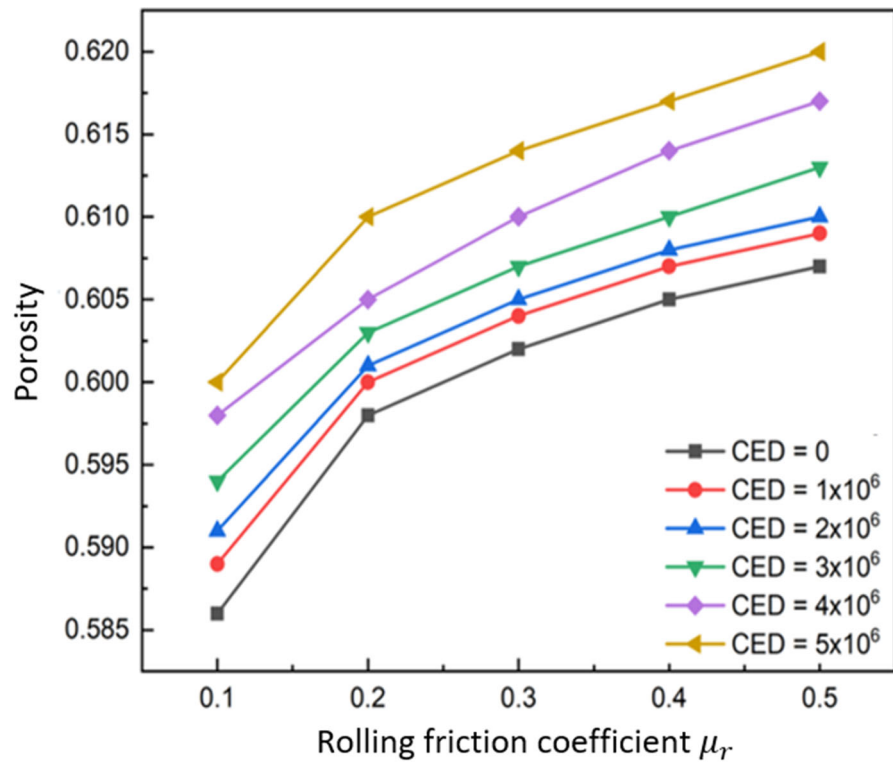


Fig. 6. Variations in porosity with cohesion and rolling friction

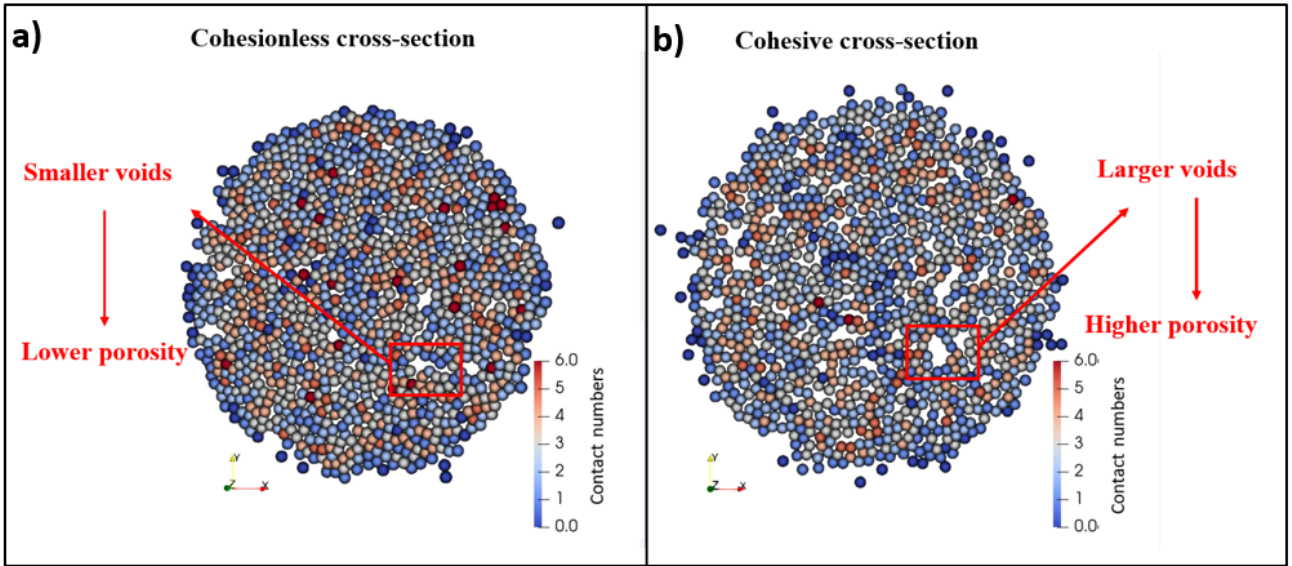


Fig. 7. Extracted cohesionless and cohesive cross-section in DEM (a) $\mu_r = 0.2$, $CED = 0 \text{ J/m}^3$ (b) $\mu_r = 0.2$, $CED = 5 \times 10^6 \text{ J/m}^3$

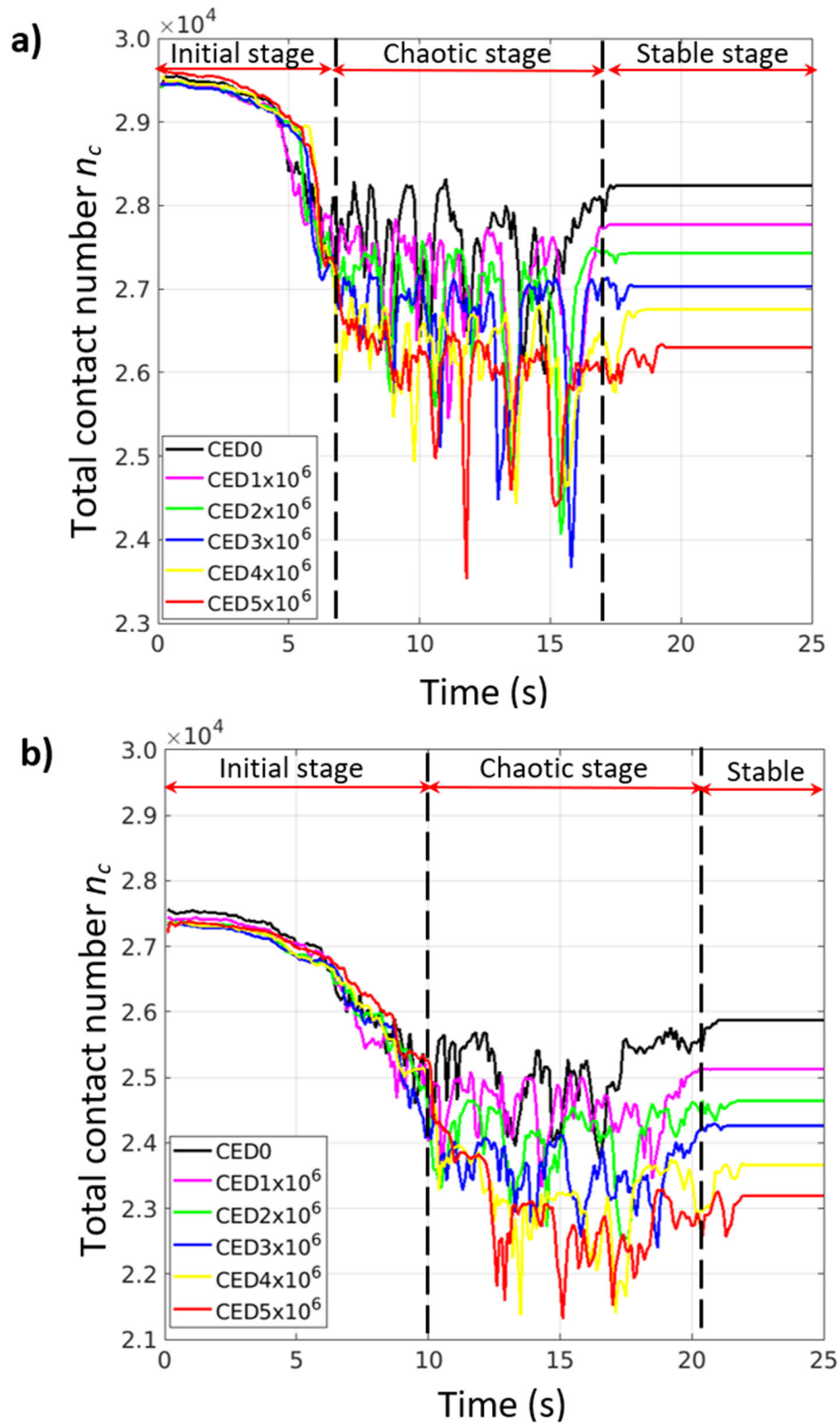


Fig. 8. Contact numbers with different cohesion levels: a) $\mu_r = 0.1$; and b) $\mu_r = 0.4$

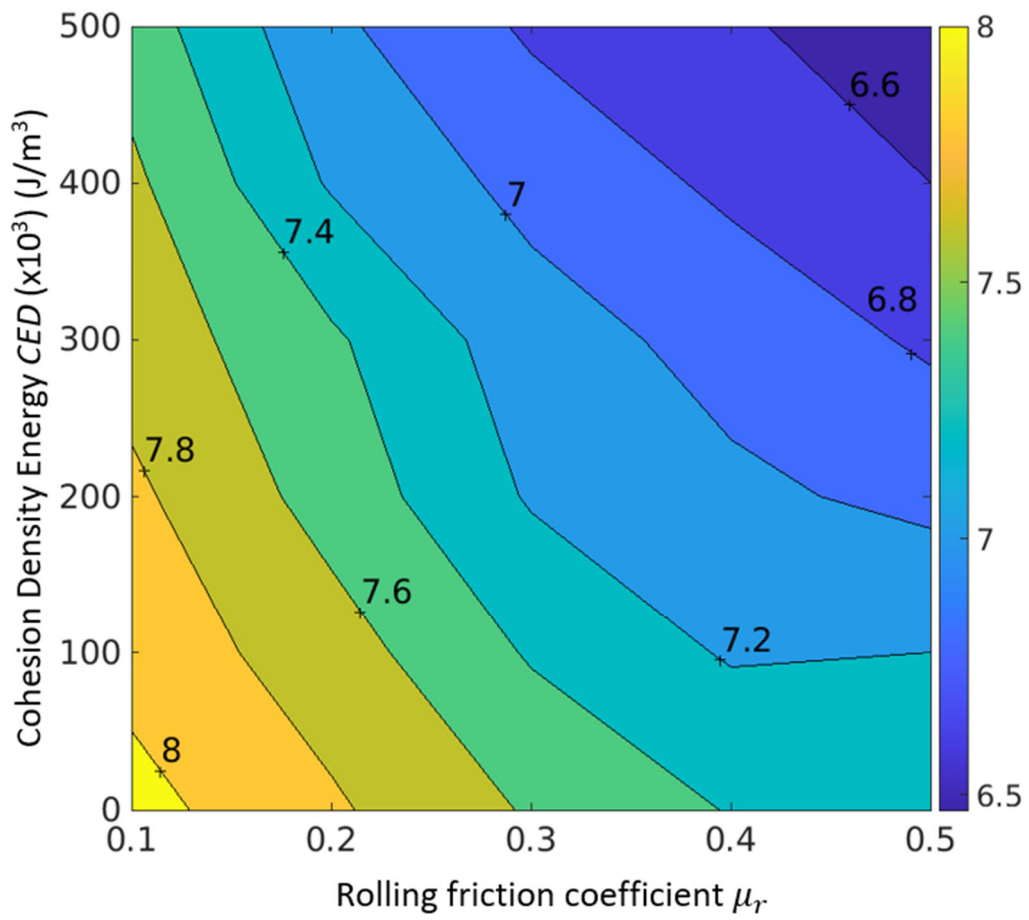


Fig. 9. Mean coordination number as a function of cohesion and rolling friction

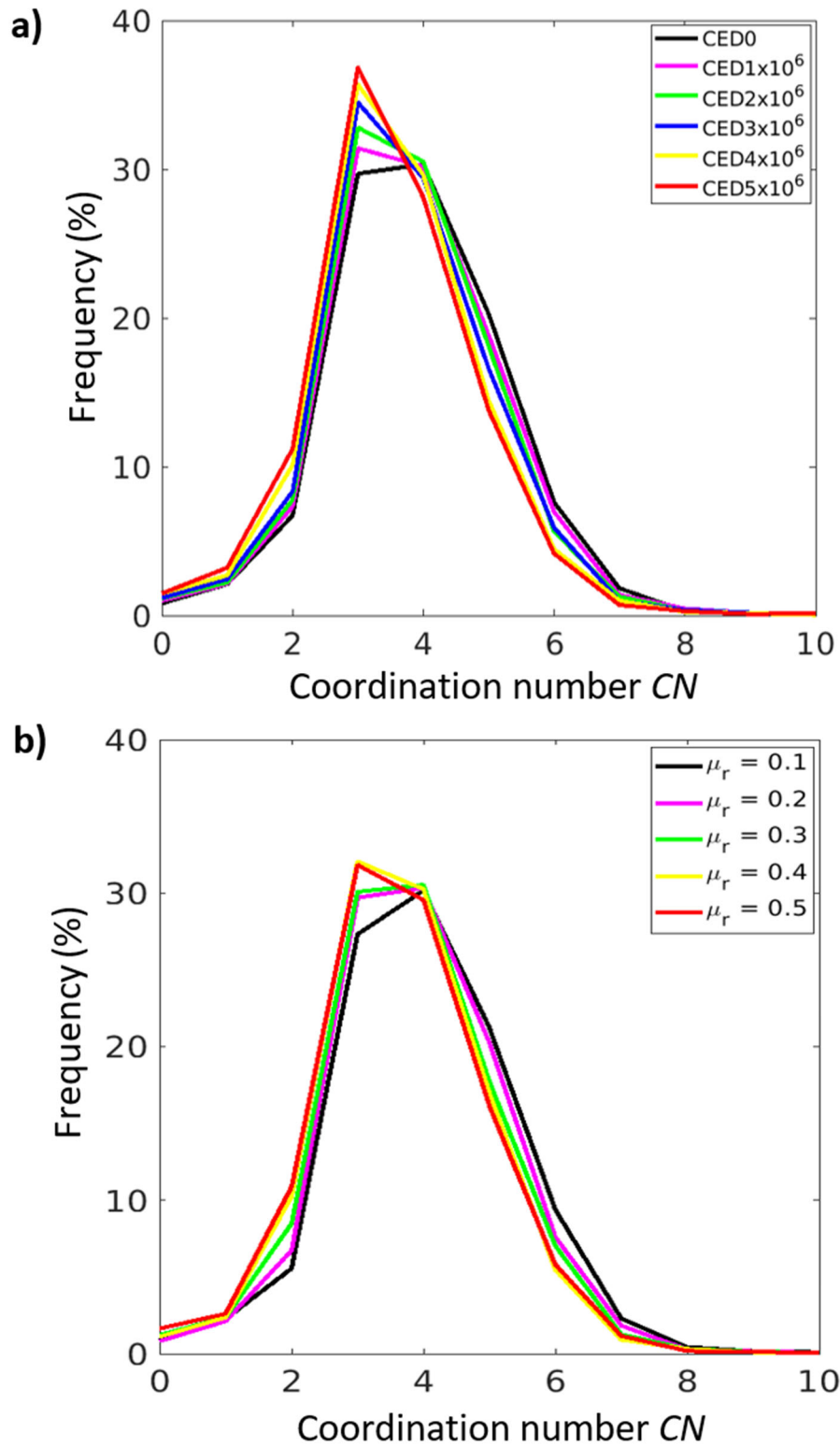


Fig. 10. Frequency of coordination number with (a) varying cohesion levels for $\mu_r = 0.2$; and (b) varying rolling friction for cohesionless case

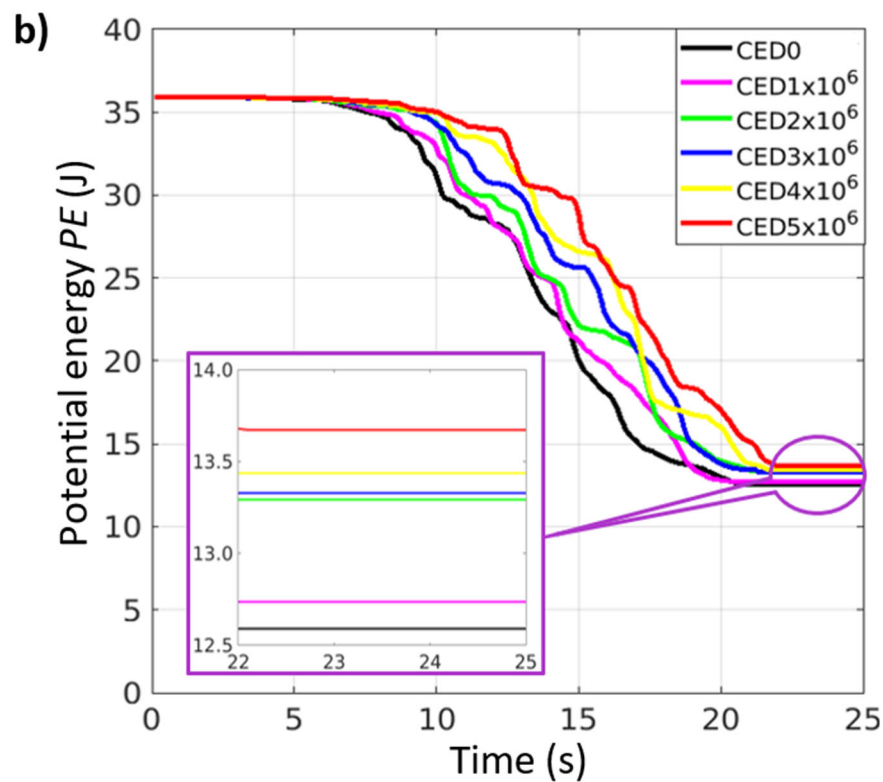
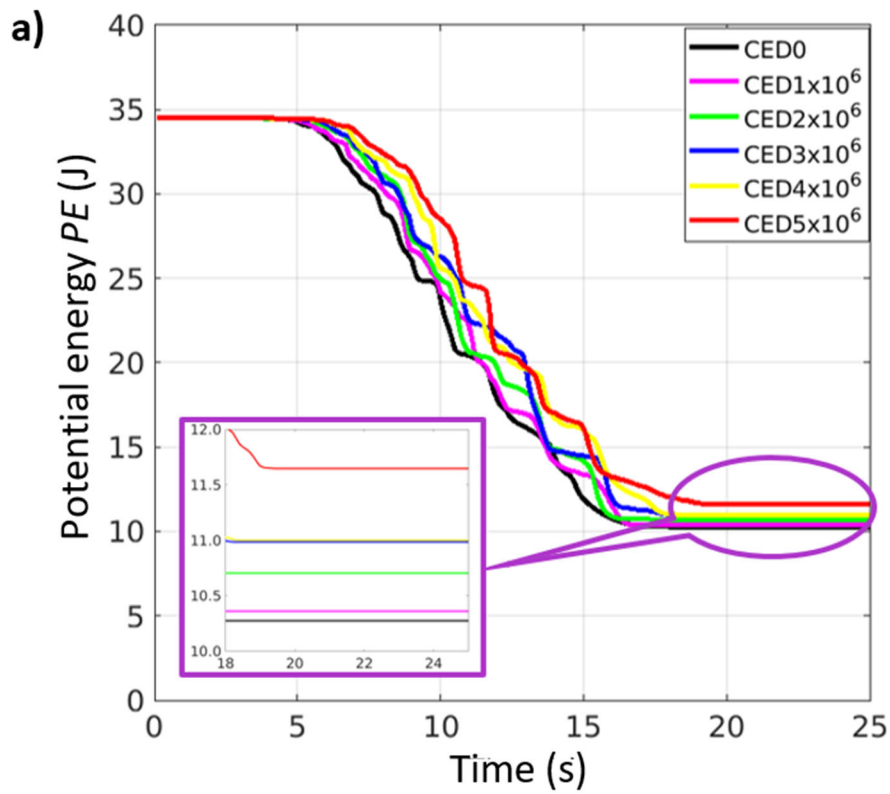


Fig. 11. Potential energy difference with time with varying cohesion levels: for (a) $\mu_r = 0.1$ and

(b) $\mu_r = 0.4$

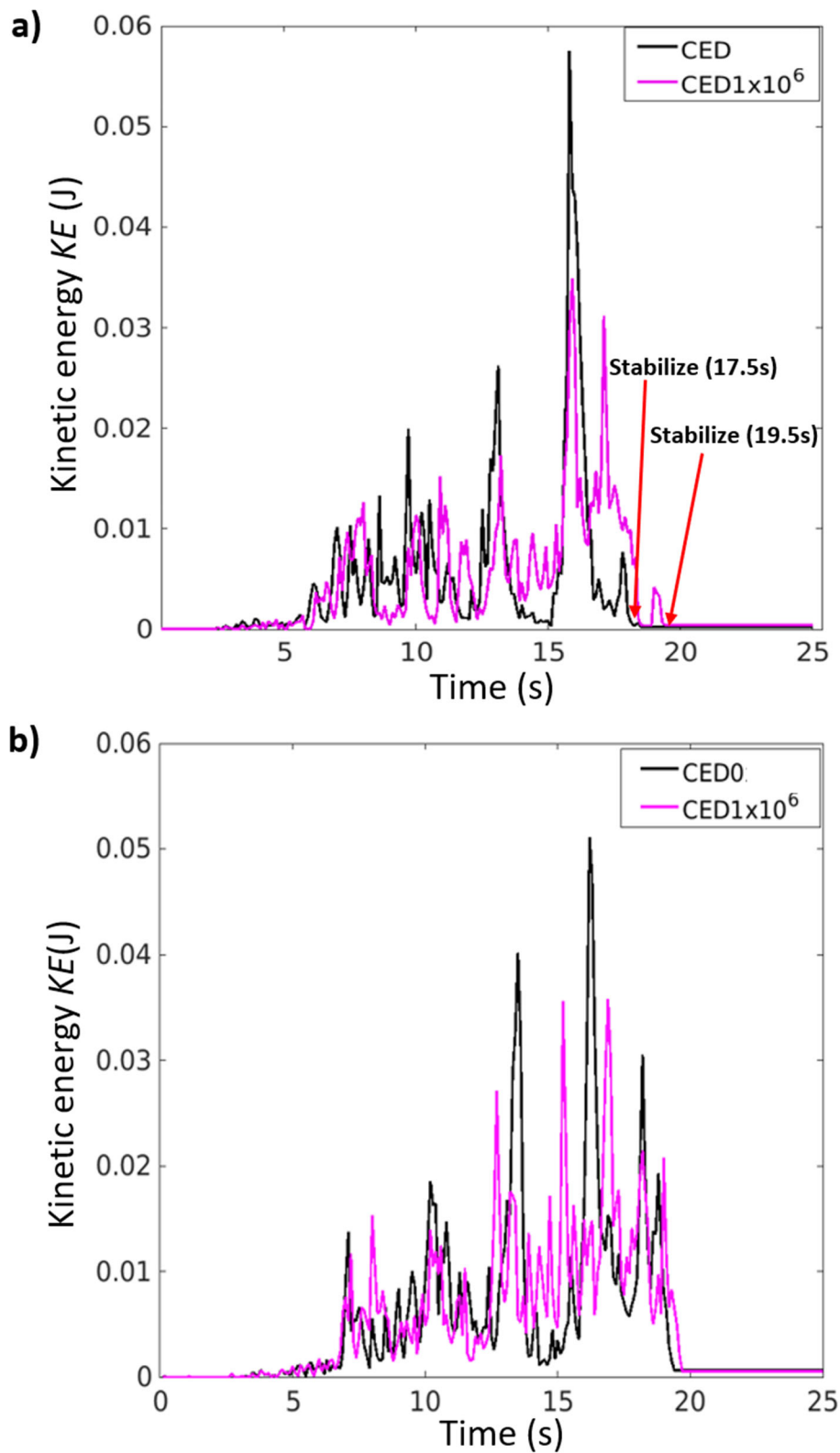


Fig. 12. Variations of average KE with varying cohesion levels: for (a) $\mu_r = 0.2$ and (b) $\mu_r = 0.3$

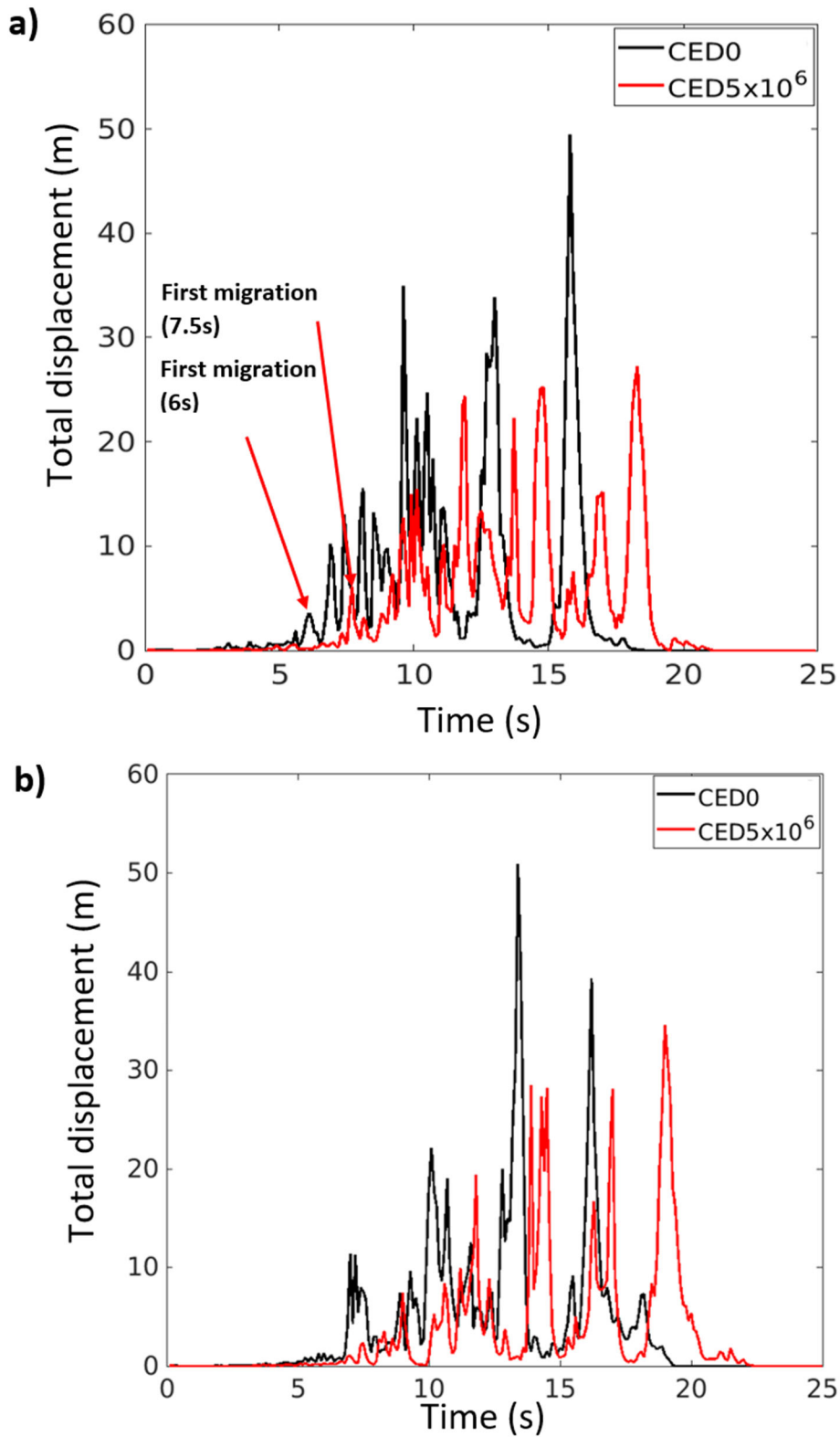


Fig. 13. Variation of the total displacements of particles considering different cohesion levels: for (a)

$\mu_r = 0.2$ and (b) $\mu_r = 0.3$

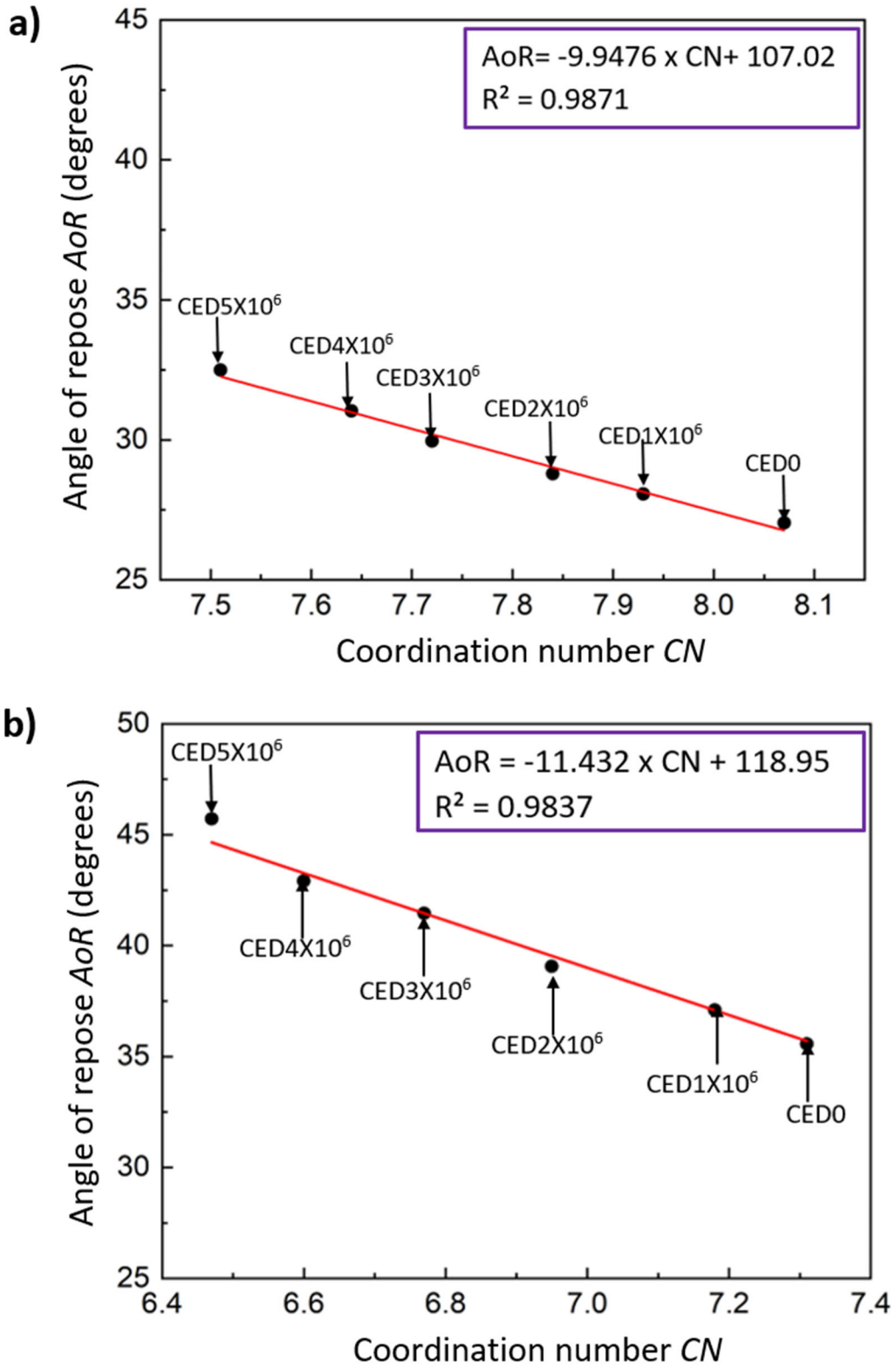


Fig. 14. Correlation between AoR and CN with different cohesion levels: for (a) $\mu_r = 0.1$ and (b) $\mu_r =$

0.5

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