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## The Structure of Agglomerates consisting of Polydisperse Particles

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### Abstract

Agglomeration is encountered in many natural or industrial processes, like growth of aerosol particles in the atmosphere and during material synthesis or even flocculation of suspensions, granulation, crystallization and with colloidal particle processing. These particles collide by different mechanisms and stick together forming irregular or fractal-like agglomerates. Typically, the structure of these agglomerates is characterized with the fractal dimension,  $D_f$ , and pre-exponential factor,  $k_n$ , of simulated agglomerates of monodisperse primary particles (PP) for ballistic or diffusion-limited particle-cluster and cluster-cluster collision mechanisms. Here, the effect of PP polydispersity on  $D_f$  and  $k_n$  is investigated with agglomerates consisting of 16 – 1024 PP with closely controlled size distribution (geometric standard deviation,  $\sigma_g = 1-3$ ). These simulations are in excellent agreement with the classic structure ( $D_f$  and  $k_n$ ) of agglomerates consisting of monodisperse PPs made by four different collision mechanisms as well as with agglomerates of bi-, tri-disperse and normally distributed PPs. Broadening the PP size distribution of agglomerates decreases monotonically their  $D_f$  and for sufficiently broad PP distributions ( $\sigma_g > 2.5$ ) the  $D_f$  reaches about 1.5 and  $k_n$  about 1 regardless of collision mechanism. Furthermore with increasing PP polydispersity, the corresponding projected area exponent,  $D_a$ , and pre-exponential factor,  $k_a$ , decrease monotonically from their standard values for agglomerates with monodisperse PPs. So  $D_f$  as well as  $D_a$  and  $k_a$  can be an indication for PP polydispersity in mass-mobility and light scattering measurements, if the dominant agglomeration mechanism is known, like diffusion-limited and/or ballistic cluster-cluster coagulation in aerosols.

### Keywords

aerosol physics; agglomerate; fractal dimension; polydisperse aerosol; morphology

### Introduction

Significant advances have been made in agglomerate characterization by employing the so-called fractal theory and relating agglomerate structure to its generation pattern through the fractal dimension,  $D_f$ . For example, agglomerates made by diffusion-limited cluster-cluster agglomeration (DLCA) have  $D_f = 1.78$  (Jullien et al. 1984) or by diffusion-limited particle-cluster agglomeration (DLA) have  $D_f = 2.5$  (Tolman and Meakin 1989), ballistic cluster-cluster agglomeration (BCCA) have  $D_f = 1.90$  (Meakin and Jullien 1988) and ballistic particle-cluster agglomeration (BPCA) have  $D_f = 3.0$  (Ball and Witten 1984) as were summarized for aerosols by Schaefer and Hurd (1990). These  $D_f$ -values have become the standard in agglomerate (physically-bound particles) and even aggregate (chemically- or

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sinter-bound particles) characterization despite the fact that such particles may not fully obey fractal theory but are sufficiently close to it to be called fractal-like.

This concept has served well process design of a wide spectrum of filamentary particles (carbon black, nickel and fumed silica), made mostly by aerosol coagulation. In fact, a number of characterization techniques and process design concepts have been developed capitalizing on these  $D_f$ -values to extract other particle properties (e.g. collision diameter, primary particle size) and design reactors for manufacturing such particles. They are characterized by a power law relating the number of primary particles in an agglomerate to primary particle radius and radius of gyration or collision radius. For example, in aerosol reactor design, the coagulation rate of spherical particles is corrected to that of fractal-like ones in the free molecular (Matsoukas and Friedlander 1991) and continuum (Oh and Sorensen 1997) regimes by introducing the radius of gyration and  $D_f$  in their collision area term (Mountain et al. 1986).

What might have been overlooked in characterization and simulations of fractal-like particles is that the above  $D_f$ -values have been developed for agglomerates of monodisperse primary particles. Notable exceptions come from Tence et al. (1986) and Bushell & Amal (1998) who had examined the effect of primary particle polydispersity on agglomerate structure and scattering behavior. They numerically generated BCCA and DLCA agglomerates having Gaussian-distributed (Tence et al. 1986) and mono-, bi- and tri-disperse (Bushell and Amal 1998) primary particles and found no effect on  $D_f$  for their employed polydispersities.

For coagulating aerosols, however, this needs to be carefully examined as Brownian coagulation-driven particle formation leads to polydisperse particles. Their distribution, even at a single streamline is, at best, as narrow as that given by self-preserving theory, e.g. having geometric standard deviation,  $\sigma_g$ , of about 1.45 (Landgrebe and Pratsinis 1989). When agglomerates are made in aerosol reactors at different (e.g. radial) residence time distributions and collected, typically, by filtration, much broader size distributions can be obtained. Agglomerates with quite polydisperse primary particle size distribution are obtained during gas-to-particle conversion or when primary particles with different residence time histories are mixed. So the common extraction of  $D_f$  by microscopic counting of such agglomerates and primary particles may not necessarily lead to an accurate assessment of their formation pathway as has been proposed by Schaefer and Hurd (1990). Similarly the use of  $D_f$  in designing aerosol synthesis of such particles needs to account for agglomerates of polydisperse primary particles.

Here fractal-like agglomerates consisting of 16, 64, 256, 512 and 1024 primary particles having closely controlled size distributions are generated by DLA, DLCA, BCCA and BPCA. So the employed primary particle radii are log-normally distributed with geometric standard deviation  $\sigma_g$  ranging from 1 (monodisperse) to 3 and the structure characteristics ( $D_f$  and pre-exponential factor  $k_n$ ) of such agglomerates are calculated by fractal theory for these collision mechanisms and compared to the literature at limiting cases. Furthermore the projected area exponent and pre-exponential factor that are used in aerosol characterization by mass-mobility measurements are obtained as a function of primary particle polydispersity.

## Theory

Agglomerates grown by particle collisions exhibit a power law scaling between number of primary particles,  $n_p$ , and radius of gyration,  $r_g$  (Mandelbrot 1982):

$$n_p = k_n \left( \frac{r_g}{r_p} \right)^{D_f}, \quad (1)$$

where the exponent  $D_f$  is the fractal dimension,  $r_p$  the monodisperse or geometric mean primary particle radius,  $k_n$  a prefactor and the radius of gyration is:

$$r_g^2 = \frac{\sum_i x_i^2 m_i}{\sum_i m_i}, \quad (2)$$

with  $m_i$  being the mass and  $x_i$  the distance of primary particle  $i$  to the agglomerate center of mass. For agglomerates of polydisperse primary particles, the equivalent monodisperse or surface area mean primary particle radius  $r_{va}$  and number  $n_{va}$  is commonly determined by the (total) agglomerate volume  $v$  and surface area  $a$ :

$$r_{va} = \frac{3v}{a}, \quad (3)$$

$$n_{va} = \frac{v}{4\pi r_{va}^3 / 3}, \quad (4)$$

This corresponds to the average primary particle radius determined experimentally by nitrogen adsorption,  $r_{BET}$ ; and small angle X-ray scattering (SAXS),  $r_{VS}$  (Hyeon-Lee et al. 1998). The  $r_{va}$  is equal to  $r_p$  for agglomerates of monodisperse spherical primary particles.

The projected agglomerate area  $a_a$  onto a plane is also related to  $n_p$  by a power law (Medalia 1967):

$$n_p = k_a \left( \frac{a_a}{a_p} \right)^{D_a}, \quad (5)$$

where  $a_p$  is the projected area of a primary particle and  $D_a$  and  $k_a$  are the power law exponent and prefactor, respectively. The projected area is an important agglomerate property as it can be used to define the agglomerate mobility diameter,  $d_m$ , in the free molecular regime (Meakin 1988):

$$d_m = \sqrt{\frac{4a_a}{\pi}}. \quad (6)$$

This holds well even in the transition regime up to Knudsen number  $Kn = 0.28$  (Rogak et al. 1993). Equations 5 and 6 are used to characterize agglomerates by mass-mobility measurements (Park et al. 2003).

The  $a_a$  is calculated by Monte-Carlo integration (Meakin 1988) and averaged over 50 homogeneously distributed random angles for each agglomerate. The  $D_f$  and  $k_n$  as well as  $D_a$  and  $k_a$  are obtained by ensemble averaging over agglomerates of different  $n_p$  (Eggersdorfer et al. 2010) and similarly to the evaluation of experiments (Sorensen et al. 1992). The  $D_f$  or  $D_a$  is determined from the slope of  $r_g$  or  $a_a$  versus the  $n_p$  or  $n_{va}$  in a double logarithmic plot, while the intersection with the y-axis is the logarithm of the prefactor  $k_n$  or  $k_a$ , respectively (Medalia 1967; Forrest and Witten 1979).

## Results and Discussion

### Validation

One hundred agglomerates of each number ( $n_p = 16-1024$ ) and polydispersity ( $\sigma_g = 1-3$ ) of primary particles (PP) are generated by DLA (Witten and Sander 1981), DLCA (Botet et al. 1984), BCCA (Tence et al. 1986) and BPCA (Sutherland 1966) to determine  $D_f$  and  $k_n$  by eq. 1 and  $D_a$  and  $k_a$  by eq. 5. Figure 1 shows typical images of agglomerates made with monodisperse primary particles ( $\sigma_g = 1$ ) for DLA, DLCA, BCCA and BPCA similar to Schaefer and Hurd (1990). These agglomerates have  $D_f = 1.79 \pm 0.03$  (DLCA),  $1.89 \pm 0.03$  (BCCA),  $2.25 \pm 0.02$  (DLA) and  $2.81 \pm 0.03$  (BPCA). The DLCA and BCCA values are in excellent agreement with Jullien et al. (1984) and Tence et al. (1986). The  $D_f$  of DLA and BPCA are slightly smaller than their asymptotic limits of 2.5 and 3 by Tolman and Meakin (1989) and Ball and Witten (1984), respectively. This is a finite size effect, as the  $D_f$  of small agglomerates is always lower than their asymptotic  $D_f$  (Meakin 1999). The  $k_n = 1.40 \pm 0.12$  for monodisperse DLCA agglomerates (Table 1) is consistent with Sorensen & Roberts (1997). The BCCA, DLA and BPCA agglomerates of monodisperse primary particles have  $k_n = 1.36 \pm 0.1$ ,  $0.86 \pm 0.04$  and  $0.46 \pm 0.03$  (Table 1, actual  $n_p$  &  $r_p$ ).

Agglomerates with bi- and tri-disperse primary particles were made also by DLCA to compare to Bushell and Amal (1998). These agglomerates consist of primary particles of two and three distinct diameters, respectively. Their bi-disperse agglomerates with a PP radius ratio of 3:1 and about 4% of the large particles were calculated here to have  $D_f = 1.79 \pm 0.03$ . Similarly their tri-disperse PP radius ratio of 5:3:1 with distributions of 1:18:81% and PP radius ratio of 5:2:1 with distributions of 10:60:30% were calculated here to have  $D_f = 1.77 \pm 0.03$  and  $1.79 \pm 0.03$ , respectively. All were in excellent agreement with Bushell and Amal (1998) who had reported  $D_f = 1.78 \pm 0.03$ . Tence et al. (1986) obtained by BCCA agglomerates of primary particles having a Gaussian-like distribution ( $n = 2$  and  $\beta = 5$ ) that corresponds to a  $\sigma_g = 1.35$ . The present simulations with primary particles having this  $\sigma_g$  resulted in agglomerates that had a  $D_f = 1.91 \pm 0.05$ , which is quite close to that of  $1.89 \pm 0.03$  (Tence et al. 1986). These results validate the employed algorithms to address the structure of agglomerates having arbitrarily broad distributions of primary particles.

### Effect of Primary Particle Polydispersity on Fractal Dimension & Prefactor

The primary particles are log-normally distributed with a geometric standard deviation  $\sigma_g = 1-3$ . Figure 2 shows a TEM-like image of 100 DLCA agglomerates consisting of 512 PPs with  $\sigma_g = 2$ .

Figure 3 shows close-up snapshots of randomly chosen agglomerates consisting of 1024 PPs of  $\sigma_g = 2$  and 3 made by DLCA, BCCA, DLA and BPCA along with the corresponding  $D_f$ . The broader the PP size distribution and thus the larger the difference between PPs, the stringier and more open are the agglomerates having a lower  $D_f$  than the benchmark agglomerates of monodisperse PPs (Fig. 1). At  $\sigma_g = 3$  all agglomerates look similar somehow, regardless of collision mechanism.

Figure 4a quantifies these pictures by showing the  $D_f$  of agglomerates as a function of their constituent PP size distribution from  $\sigma_g = 1$  (monodisperse) to 3 for DLCA (circles), BCCA (squares), DLA (triangles) and BPCA (diamonds). The  $D_f$  decreases with increasing  $\sigma_g$  and for  $\sigma_g = 3$  the different collision mechanisms converge to nearly the same  $D_f \approx 1.5$ . Clearly the presence of various primary particle sizes results in more open agglomerate structures. The more polydisperse are the PPs, the more open space leave between them within an agglomerate.

Polydispersity adds another element of disorder when uniform primary particles are replaced with non-uniform ones. As the motion of particles or clusters is independent from each other, their arrival or sticking point upon collision is fixed. Take for example particle-cluster agglomeration. The trajectory of the particle till its collision with a cluster is determined by its mobility diameter. As individual polydisperse PP arrive at the cluster, more irregular and open structures are developed than with uniform PPs. As a result, the compact agglomerate structures made by DLA and BPCA of monodisperse PPs with  $D_f$  of 2.25 and 2.81, respectively, are affected most dramatically by the PP polydispersity (Fig. 4a). Similar concepts apply to cluster-cluster agglomeration. The larger particles contribute much more to  $r_g$  than the smaller ones since  $r_g$  is calculated with the mass-weighted mean square displacement (eq. 2). So the  $D_f$  is very sensitive to PP polydispersity. The differences between particle masses increase for increasing  $\sigma_g$  so that at high polydispersities fewer particles effectively contribute to  $D_f$ . This results in  $D_f = 1$  in the limit of two effective particles as  $r_g$  represents the high order moments of the distribution (square root of ratio between 2<sup>nd</sup> and 0<sup>th</sup> moment of mass-based distribution).

Figure 4a shows also the DLCA simulations (filled circles) of bi- and tri-disperse PPs by Bushell and Amal (1998) as well as the BCCA ones (filled square) of Tence et al. (1986). Clearly both studies had been carried out with relatively narrow PP size distributions to observe any reduction of  $D_f$  by the polydispersity of primary particles seen here. This monotonic reduction of  $D_f$  with increasing  $\sigma_g$  can be used to determine the effect of PP polydispersity on the structure of agglomerates by all four collision mechanisms.

Figure 4b shows the prefactor  $k_n$  for the agglomerates made here as a function of PP  $\sigma_g$  also. While the  $D_f$  decreases monotonically for increasing  $\sigma_g$ , regardless of collision mechanism, not all  $k_n$  exhibit such a monotonic behaviour. Nevertheless all reach a  $k_n \approx 1$  for very broad PP polydispersities ( $\sigma_g > 2.5$ ). For DLCA and BCCA particles, the  $k_n$  decreases continuously with increasing PP polydispersity, while for both DLA and BPCA  $k_n$  reaches a minimum here at  $\sigma_g = 1.6$  and 1.5, respectively.

For  $\sigma_g = 1.45$ , the standard deviation of the self-preserving size distribution (SPSD) obtained by Brownian coagulation that corresponds best to DLCA, both  $D_f$  and  $k_n$  are within the error bars of agglomerates with monodisperse PPs and thus indistinguishable from each other. This indicates that PP polydispersity may not affect the structure of DLCA agglomerates made by aerosol coagulation on a single process streamline or residence time history. When, however, such particles are collected from a filter where particles of various residence time (and even temperature) histories are arriving, there would be a significant effect depending on their actual polydispersity. In practical systems, the primary particle size distribution is often unknown and for simplicity assumed to be monodisperse.

Table 1 compares the  $D_f$  and  $k_n$  determined with the actual  $r_p$  and  $n_p$  to that determined by the assumption of an average PP size of polydisperse ( $\sigma_g = 1-3$ ) PPs using  $r_{va}$  and  $n_{va}$  as might be determined by microscopic counting, nitrogen adsorption or SAXS. For all agglomeration mechanisms, the  $D_f$  is not affected by more than 10%. In contrast, the  $k_n$  is overestimated significantly when assuming all primaries to have the same size for all collision mechanisms (Table 1, last column).

### Effect of Primary Particle Polydispersity on Projected Surface Area & Mass-Mobility Characterization

The projected surface area  $a_a$  of agglomerates controls the mass and heat transfer, e.g. the collision frequency with particles much smaller than the gas mean free path and thus the mobility of agglomerates in the free molecular and transition regime (Meakin 1988; Rogak et al. 1993). For DLCA agglomerates of monodisperse PPs,  $D_a = 1.08 \pm 0.002$  in Eq. 5,

consistent with Pierce et al. (2006). Similar to  $D_f$  the  $D_a$  decreases with increasing  $\sigma_g$  (Fig. 5a). The  $a_a$  depends on the square of PP radius. So the largest PPs dominate while the smallest PPs hardly contribute to  $a_a$ . Agglomerates of the same mean PP radius but broader  $\sigma_g$  have also a larger  $a_a$  and in turn a reduced  $D_a$  regardless of collision mechanism.

Figure 5b shows the prefactor  $k_a$  of eq. 5 as a function of  $\sigma_g$ . The  $k_a$  decreases until  $\sigma_g = 2.0$  and levels off for higher polydispersities. Though even for  $\sigma_g = 3.0$ , where  $D_f$  and  $k_n$  are nearly independent of collision mechanism indicating a similar mass distribution of the agglomerates,  $D_a$  and  $k_a$  differ for the four collision types. The DLCA particles have always the largest  $k_a$  and smallest  $D_a$ , since their most open structure results in primary particles that are less shielded by neighbors compared to BCCA, DLA and BPCA agglomerates (Fig. 1 & 3). For all collision mechanisms, the monotonic dependence of  $D_a$  and  $k_a$  on  $\sigma_g$  can be used to extract the effect of PP polydispersity of agglomerates by mass-mobility measurements as with  $D_f$  (Fig. 4a).

As two measures of fractal dimension are available for these agglomerates of various PP polydispersities and collision mechanisms, it is worth comparing them. So Figure 6 shows the  $D_f$  vs.  $2D_a$  by combining Fig. 4a and 5a. For both diffusion-limited and ballistic cluster-cluster agglomeration, the  $r_g$ -based  $D_f$  is smaller than the  $r_m$ -based  $2D_a$  regardless of PP polydispersity, with  $r_m = d_m/2$ . This is not surprising as DLCA and BPCA agglomerates have a  $D_f < 2$  and very open chain-like structures even for a monodisperse PP size distribution, so always  $r_m < r_g$  (Sorensen 2011) and  $D_f < 2D_a$ . Particle-cluster agglomerates (triangles & diamonds), however, are rather compact at  $\sigma_g < 2$  with  $D_f > 2$  (Fig. 4a) so  $D_f > 2D_a$ . Both BPCA and DLA agglomerates of monodisperse PPs have  $r_m > r_g$  similar to that of a sphere with  $r_m/r_g = \sqrt{5/3}$  (Hiemenz 1986). With increasing PP polydispersity, however, the particle-cluster agglomerates become more open so  $r_m/r_g < \sqrt{5/3}$  and exhibit a transition at  $D_f \approx 2.1$  from  $D_f > 2D_a$  to  $D_f < 2D_a$  (between  $\sigma_g = 1.5$  and 2 in Fig. 4a & 5a) as their  $r_m < r_g$ . For  $D_f > 2$ , the projected areas are compact objects with screening of individual PP while for  $D_f < 2$  almost all PPs are accessible (Mulholland et al. 1988). So the polydispersity of PPs affects most dramatically the structure of particle-cluster agglomeration made by both diffusion-limited and ballistic collisions.

## Conclusions

Agglomerates consisting of polydisperse primary particles were generated by DLCA, BCCA, DLA and BPCA. The effect of primary particle size distribution on resulting agglomerate fractal dimension  $D_f$  and prefactor  $k_n$ , projected area exponent  $D_a$  and prefactor  $k_a$  was investigated. Asymptotic  $D_f$  and  $k_n$  values were obtained for agglomerates of primary particles having geometric standard deviation  $\sigma_g$  ranging from 1.0 to 3.0. These simulations were in agreement with the literature for fractal agglomerates of monodisperse primary particles and even with limited studies of fractal-like particles containing Gaussian-like, bi- and tri-disperse primary particles.

It was discovered that polydisperse primary particles result in more open agglomerate structures with lower  $D_f$ ,  $D_a$  and  $k_a$  than the classic with monodisperse PPs. This is most notable for particle-cluster (up to 50% reduction) and to a lesser extent for cluster-cluster (up to 20% reduction) ballistic and diffusion-limited collision-generated agglomerates. Furthermore it was shown that the assumption of an average primary particle size, which is commonly applied in particle characterization, hardly affects  $D_f$  but significantly overestimates  $k_n$  for polydisperse PPs for all collision mechanisms. Most remarkably, for processes generating agglomerates by a known collision mechanism, a  $D_f$ ,  $D_a$  or  $k_a$  smaller

than the benchmark values of agglomerates consisting of monodisperse PPs might be an indication for agglomerates of polydisperse primary particles.

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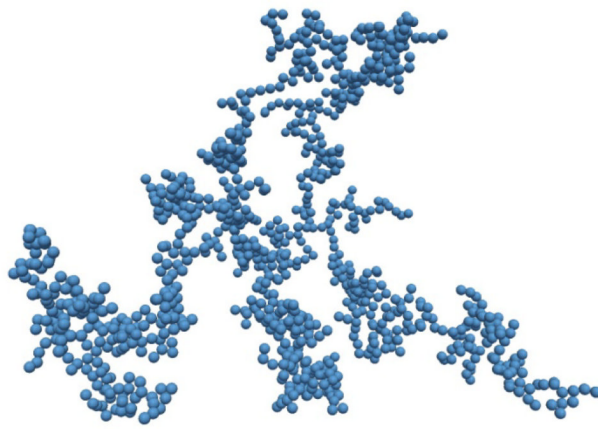
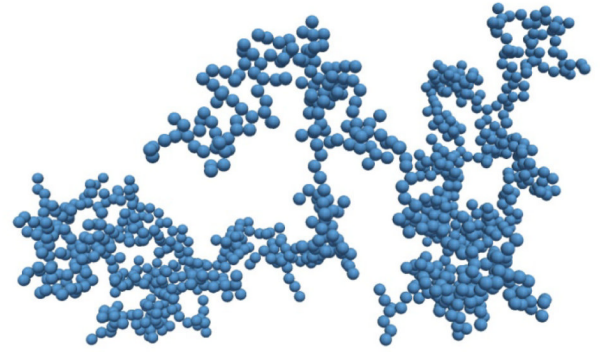
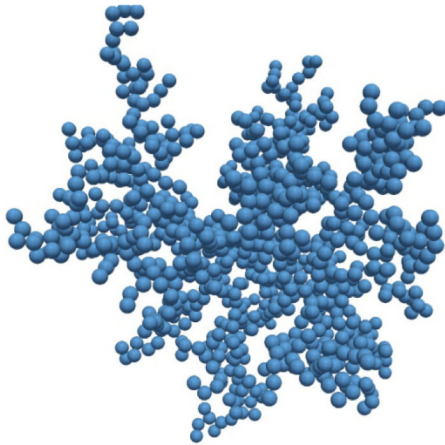
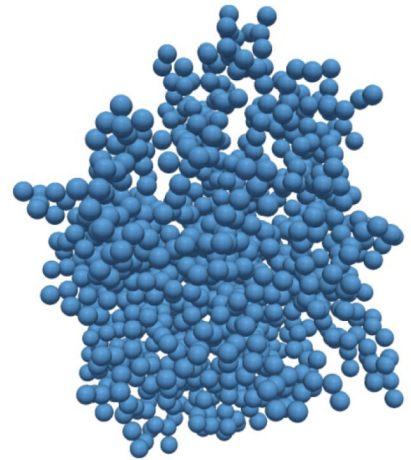
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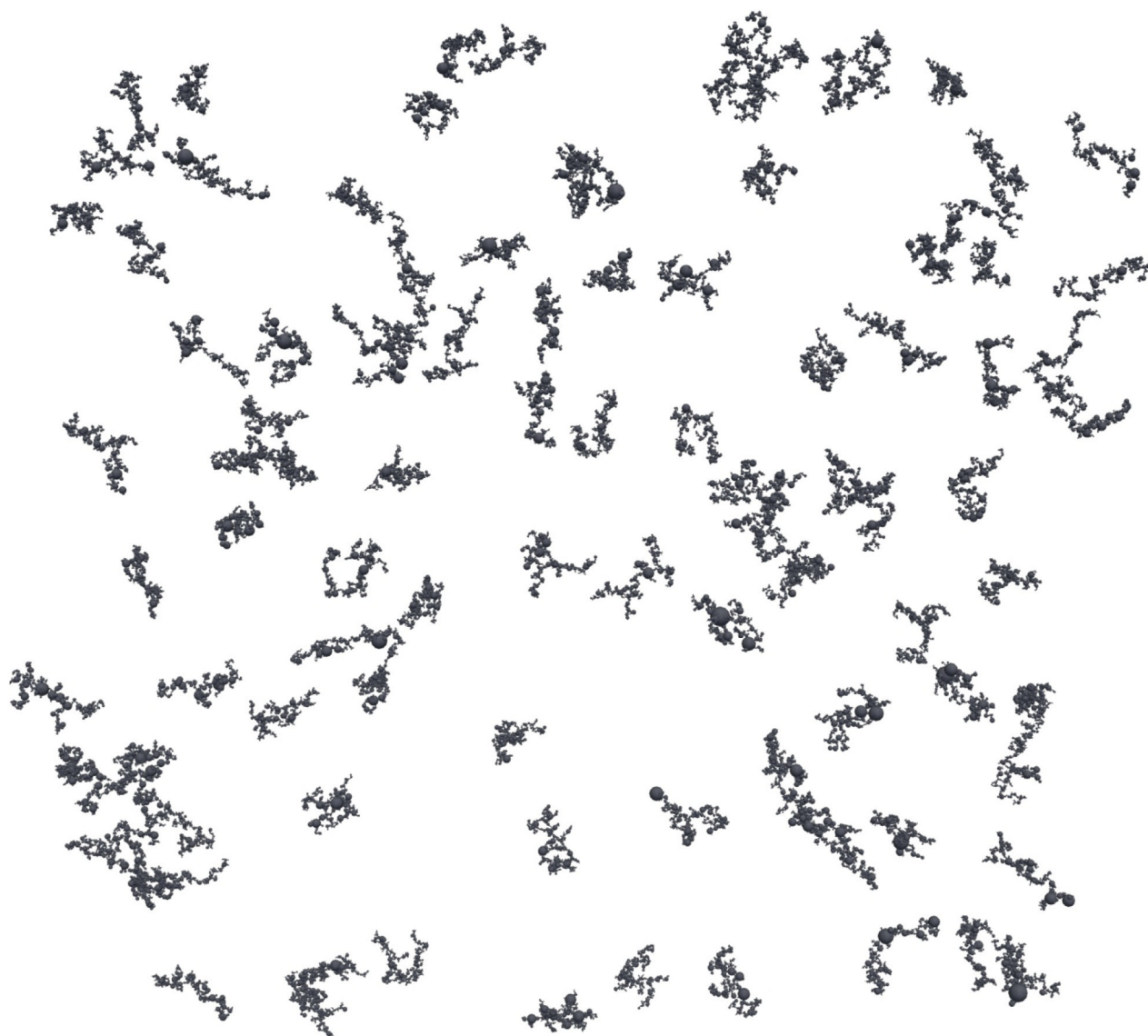
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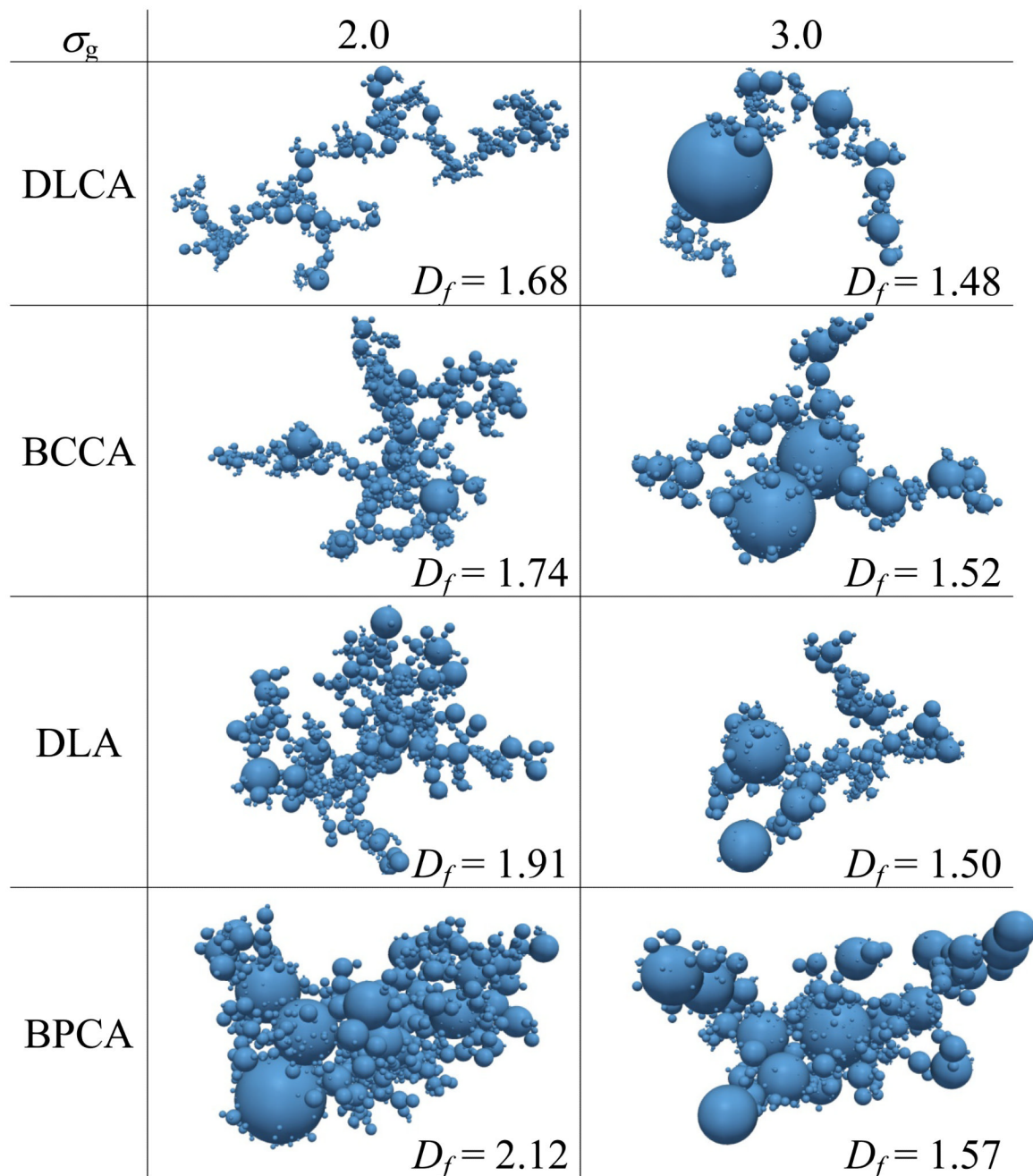


a) DLCA,  $D_f = 1.79$ b) BCCA,  $D_f = 1.89$ c) DLA,  $D_f = 2.25$ d) BPCA,  $D_f = 2.81$ **Figure 1.**

Agglomerates consisting of 1024 monodisperse primary particles made by a) diffusion-limited (DLCA) and b) ballistic cluster-cluster (BCCA) agglomeration as well as by c) diffusion-limited (DLA) and d) ballistic particle-cluster (BPCA) agglomeration. These agglomerates have  $D_f = 1.79, 1.89, 2.25$  and  $2.81$  identical to those calculated by Botet et al. (1982), Tence et al. (1986), Witten and Sander (1981) and Sutherland (1966), respectively.



**Figure 2.** Snapshot of 100 agglomerates made by diffusion-limited cluster-cluster agglomeration (DLCA) consisting of 512 primary particles with a log-normal size distribution with geometric standard deviation  $\sigma_g = 2$ .

**Figure 3.**

Agglomerates of polydisperse primary particles having geometric standard deviation  $\sigma_g = 2.0$  (left column) and  $3.0$  (right column) made by diffusion-limited (DLCA) and ballistic cluster-cluster agglomeration (BCCA) as well as diffusion-limited (DLA) and ballistic particle-cluster agglomeration (BPCA). The structure of agglomerates consisting of very polydisperse ( $\sigma_g = 3$ ) particles is quite similar ( $D_f = 1.48 - 1.57$ ) indicating that their polydispersity essentially “washes out” differences in agglomeration mechanisms.

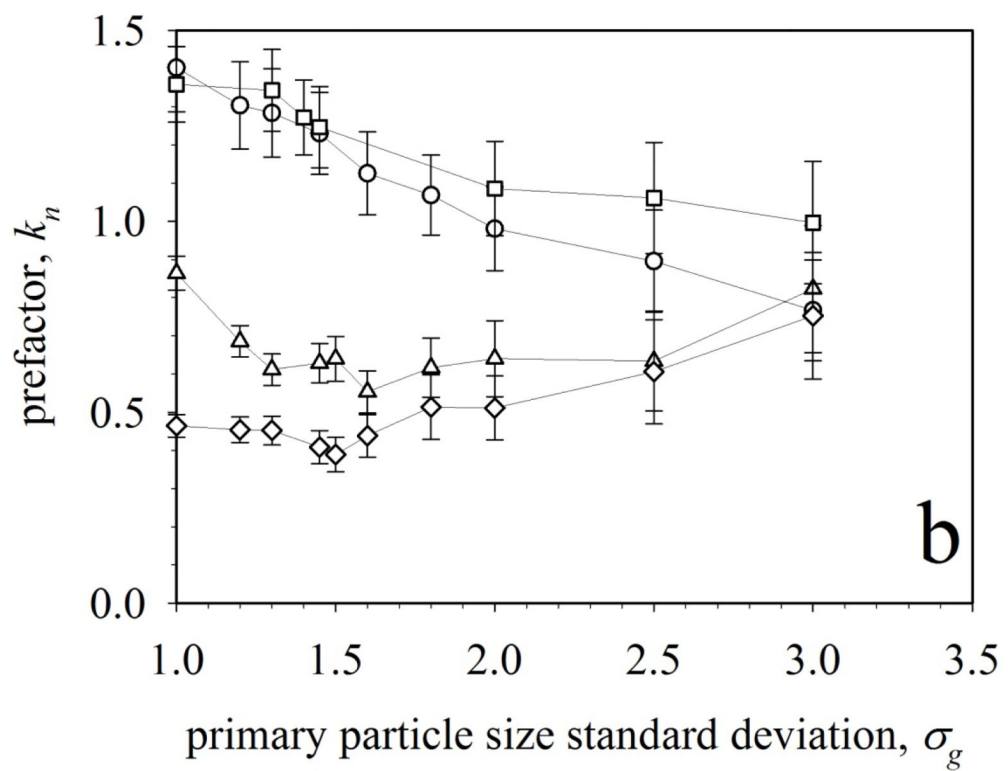
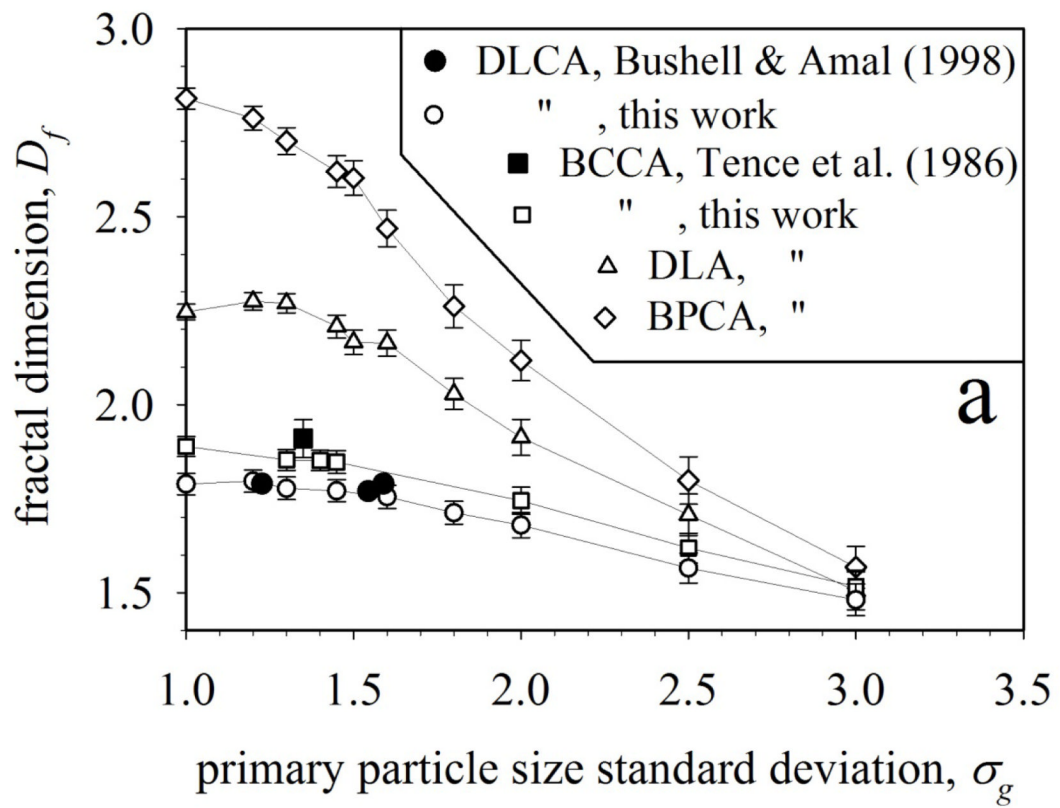
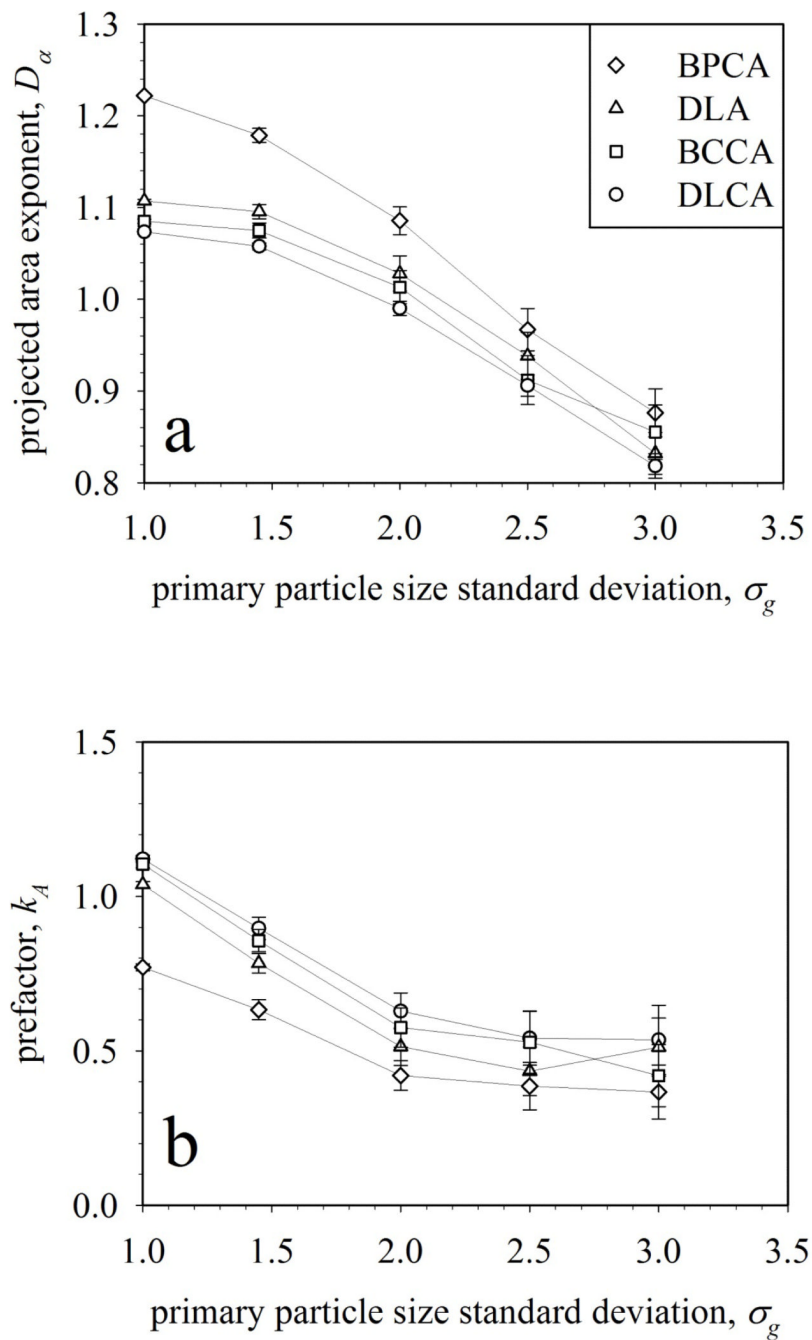
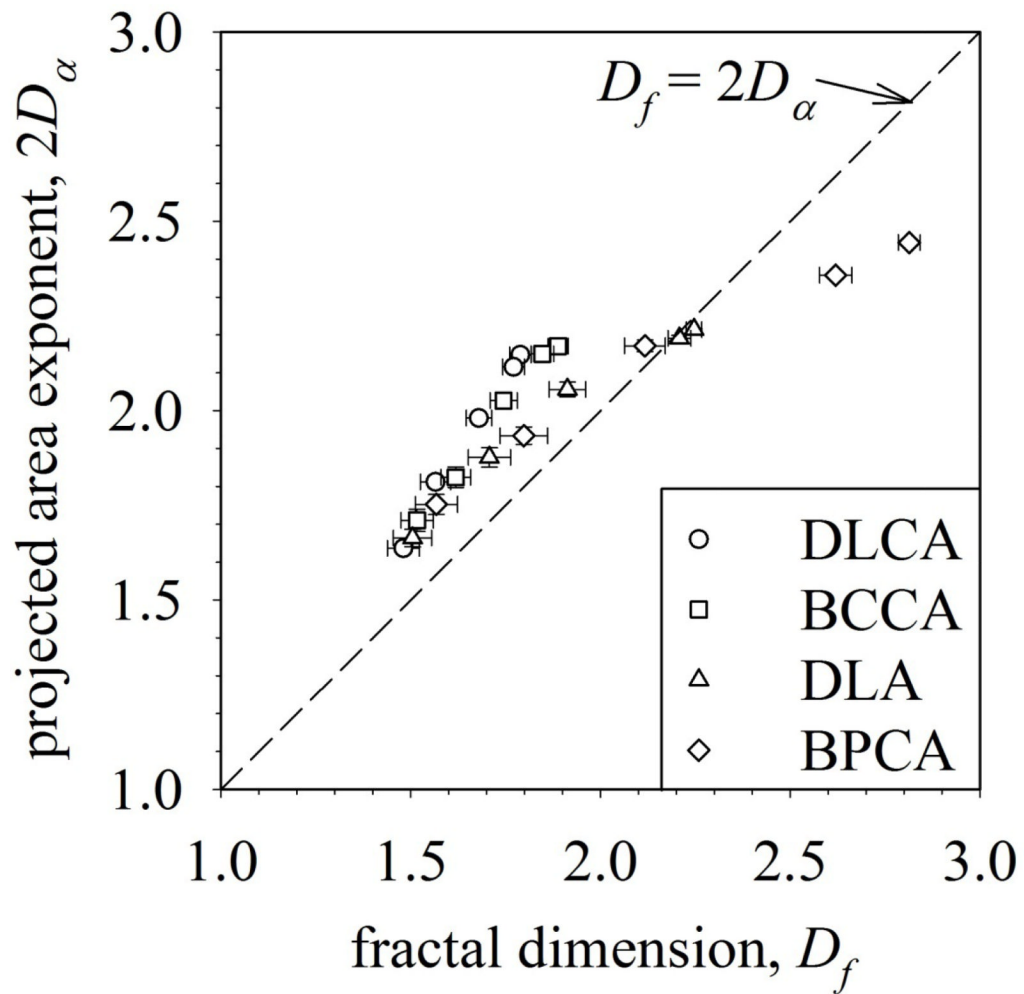


Figure 4.

The asymptotic (a) fractal dimension  $D_f$  and (b) prefactor  $k_n$  of 500 agglomerates as a function of the geometric standard deviation,  $\sigma_g$ , of their primary particles (PP) for different collision or agglomeration mechanisms. At  $\sigma_g = 1$  (monodisperse PP) the classic  $D_f$  of the corresponding structures are obtained. With increasing  $\sigma_g$ , the  $D_f$  decreases gradually reaching asymptotically  $\approx 1.5$  (a) and  $k_n \approx 1$  (b) at  $\sigma_g = 3$ , nearly independent of agglomeration mechanism.



**Figure 5.** The projected area exponent  $D_\alpha$  (a) and prefactor  $k_a$  (b) of DLCA (circles), BCCA (squares), DLA (triangles) and BPCA (diamonds) agglomerates as a function of the geometric standard deviation  $\sigma_g$  of their constituent primary particles. The  $D_\alpha$  and  $k_a$  decrease monotonically for increasing  $\sigma_g$  regardless of agglomerate generation mechanism.



**Figure 6.**

The fractal dimension  $D_f$  versus projected area exponent  $2D_\alpha$  for DLCA (circles), BCCA (squares), DLA (triangles) and BPCA (diamonds) agglomerates and  $\sigma_g = 1-3$ . The broken line corresponds to  $D_f = 2D_\alpha$ . For cluster-cluster agglomerates  $2D_\alpha$  is always larger than  $D_f$  while particle-cluster agglomerates have a transition at  $D_f \approx 2.1$  between  $\sigma_g = 1.5$  and 2.

**Table 1**

The prefactor  $k_n$  and fractal dimension  $D_f$  of agglomerates made by DLCA, BCCA, DLA and BPCA as obtained using the actual PP number  $n_p$  and radius  $r_p$  are compared to  $k_n$  and  $D_f$  obtained assuming monodisperse PPs of equivalent average number  $n_{va}$  and radius  $r_{va}$  (eq. 3 & 4).

$\sigma_g$	$D_f$	$k_n$	$D_f$	$k_n$
	Actual $n_p$ and $r_p$		Monodisperse assumption, $n_{va}$ & $r_{va}$ (eq. 3 & 4)	
	<b>DLCA</b>			
1.00	1.79 ± 0.03	1.40 ± 0.12	1.79 ± 0.03	1.40 ± 0.12
1.45	1.77 ± 0.03	1.23 ± 0.11	1.79 ± 0.03	1.42 ± 0.11
2.00	1.68 ± 0.03	0.98 ± 0.11	1.74 ± 0.03	1.53 ± 0.11
2.50	1.57 ± 0.04	0.90 ± 0.13	1.68 ± 0.04	1.66 ± 0.11
3.00	1.48 ± 0.04	0.77 ± 0.13	1.60 ± 0.04	1.80 ± 0.11
	<b>BCCA</b>			
1.00	1.89 ± 0.03	1.36 ± 0.10	1.89 ± 0.03	1.36 ± 0.10
1.45	1.85 ± 0.03	1.25 ± 0.11	1.86 ± 0.03	1.49 ± 0.11
2.00	1.74 ± 0.04	1.09 ± 0.12	1.75 ± 0.03	2.04 ± 0.14
2.50	1.62 ± 0.04	1.06 ± 0.15	1.53 ± 0.04	3.01 ± 0.19
3.00	1.52 ± 0.04	1.00 ± 0.16	1.36 ± 0.04	3.62 ± 0.20
	<b>DLA</b>			
1.00	2.25 ± 0.02	0.86 ± 0.04	2.25 ± 0.02	0.86 ± 0.04
1.45	2.21 ± 0.03	0.63 ± 0.05	2.21 ± 0.03	0.83 ± 0.05
2.00	1.91 ± 0.05	0.64 ± 0.10	1.91 ± 0.05	1.23 ± 0.10
2.50	1.71 ± 0.06	0.63 ± 0.13	1.71 ± 0.06	1.50 ± 0.13
3.00	1.51 ± 0.05	0.82 ± 0.17	1.51 ± 0.05	2.31 ± 0.2
	<b>BPCA</b>			
1.00	2.81 ± 0.03	0.46 ± 0.03	2.81 ± 0.03	0.46 ± 0.03
1.45	2.62 ± 0.04	0.41 ± 0.04	2.70 ± 0.04	0.55 ± 0.04
2.00	2.12 ± 0.05	0.51 ± 0.08	2.26 ± 0.05	1.17 ± 0.11
2.50	1.80 ± 0.06	0.61 ± 0.14	1.93 ± 0.06	1.76 ± 0.17
3.00	1.57 ± 0.06	0.75 ± 0.17	1.58 ± 0.06	2.56 ± 0.21