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Ice crystal drag model extension to snowflakes: experimental and numerical investigations.

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To ensure safe flight under snowy conditions, aircraft manufacturers must demonstrate that each engine and its air inlet system can operate throughout the flight power range of the engine (including idling) in both falling and blowing snow conditions. This study is part of an effort to develop models for snow accretion. The focus here is on the modeling of snowflake transport by the carrier airflow and more specifically on the modeling of the drag coefficient of snowflakes. Based on comparisons between numerical and experimental results, we show that, provided that the complex shape of the snowflakes is accounted for, the models initially developed for ice crystal icing can also be used to estimate the drag coefficient of a snowflake. We also propose a methodology to estimate the geometric parameters of the drag models from 2D images of the particles which in general are the only available experimental data.

Nomenclature

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m_p	= Particle mass
\mathbf{v}_p	= Particle velocity
\mathbf{v}_a	= Air velocity
\mathbf{w}	= Relative velocity between air and particle
f_{max}	= Maximum diameter of Feret
f_{max}^\perp	= Maximum crosswise diameter of Feret
A^\perp	= Particle shadow area
$V_{spheroid}$	= Volume of the reconstructed spheroid
$A_{spheroid}$	= Surface of the reconstructed spheroid
d_V	= Particle volume equivalent diameter
d_m	= Particle mass equivalent diameter
Φ	= Particle sphericity
Φ^\perp	= Particle crosswise sphericity
\mathbf{F}_D	= Drag force
\mathbf{g}	= Gravity
ρ_a	= Air density
A_{ref}	= Particle reference surface associated to the drag coefficient in the drag force definition.
d_{ref}	= Particle characteristic length scale
C_D	= Drag coefficient
Re_p	= Particle Reynolds number
ρ_p	= Particle density
ρ_w	= Water density
μ_a	= Air dynamic viscosity
A_r	= Area ratio
x_{px}	= Width of one pixel
PIP	= Precipitation Imaging Probe

Introduction

To ensure safe flight under snowy conditions, aircraft manufacturers must demonstrate that each engine and its air inlet system can operate throughout the flight power range of the engine (including idling) in both falling and blowing snow conditions, without adverse effect on engine operation (power or thrust loss, surge, stall or flameout).

Snow particles are precipitation sized ice particles that form in clouds and precipitate to the ground. Falling snow

formation has its origin in a complex interplay of microphysical processes (vapor deposition, riming and aggregation, fall of the hydrometeors at different speeds depending on their size and shape), which are governed by cloud dynamics (vertical and horizontal transport) and thermodynamics (temperature and humidity vertical profiles). Snow particles have interested scientists since a long time and the first systematic studies of snow crystals started with the pioneering work of Nakaya [1]. Numerous studies about snow particles have been published in contemporary literature since these early studies. In particular, in the framework of meteorological applications, the works of [2–7], dedicated to the prediction of the terminal velocity of naturally falling snowflakes, can be mentioned. The terminology of Pruppacher and Klett [8] is adopted here: ice particles grown by vapor deposition and/or riming are called *snow crystals* and aggregates of snow crystals are referred to as *snowflakes*.

Due to the variability of atmospheric conditions prevailing in snow clouds, the shape, size, density, and related properties such as drag and terminal fall speed of natural snow particles (snow crystal and snowflakes) are found to be highly variable. Regarding the shape for instance, Fig. 1 presents samples of ice crystals (columns, columns with plates, plates in the top row, planar dendrites in the third row, graupel or snow pellets in the fifth row) and snowflakes with different degrees of riming in the three last rows.

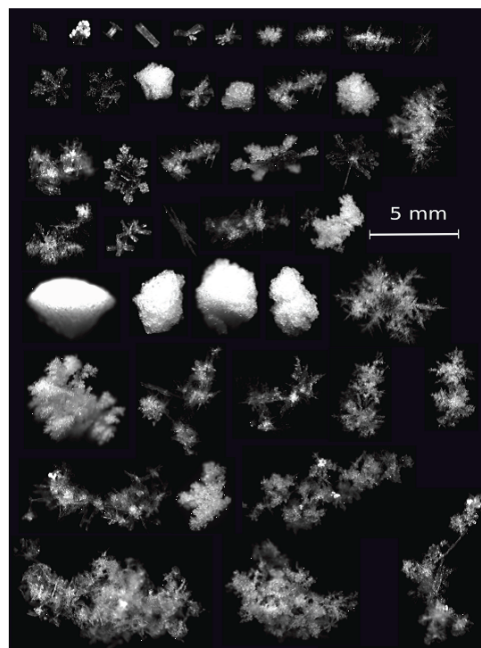


Fig. 1 MASC picture of hydrometeor illustrating the variability in the shape and size of snow particles. Reprinted with permission from [9]. Copyright 2012 authors (distributed under the Creative Commons Attribution 3.0 License).

In a context where aircraft manufacturers must comply with certifications requirements to ensure safe aircraft operation in snowy conditions, models and numerical simulation methodologies are an important asset to limit costly icing wind tunnel and flight tests. The main objective of this paper is to propose models for the transport of snowflakes by

the airflow, which is a first step in the computation of snow accretion on an aeronautical surface. This paper focuses on the trajectory of snowflakes, including the evaluation of the drag coefficient. Thermal aspects, in particular the melting of particles during transport, will be addressed in a future paper. In the particular case of snow, the main difficulty consists in developing models that take into account the geometrical complexity of the particle from experimental data that are often partial and that do not allow the complete geometrical characterization of the snowflake. This is the case when using a Precipitation Imaging Probe (PIP) from Droplet Measurement Technologies to characterize the geometry of snowflakes during a flight test campaign. Figure 2 shows the installation of the PIP among two other probes, the Cloud Imaging probe (CIP) and the High Speed Imager (HSI) on the left wing of an ATR 42. Only PIP and CIP probes were used for the ICEGENESIS [10] flight test campaign, as the HSI probe was ultimately not installed. The operating

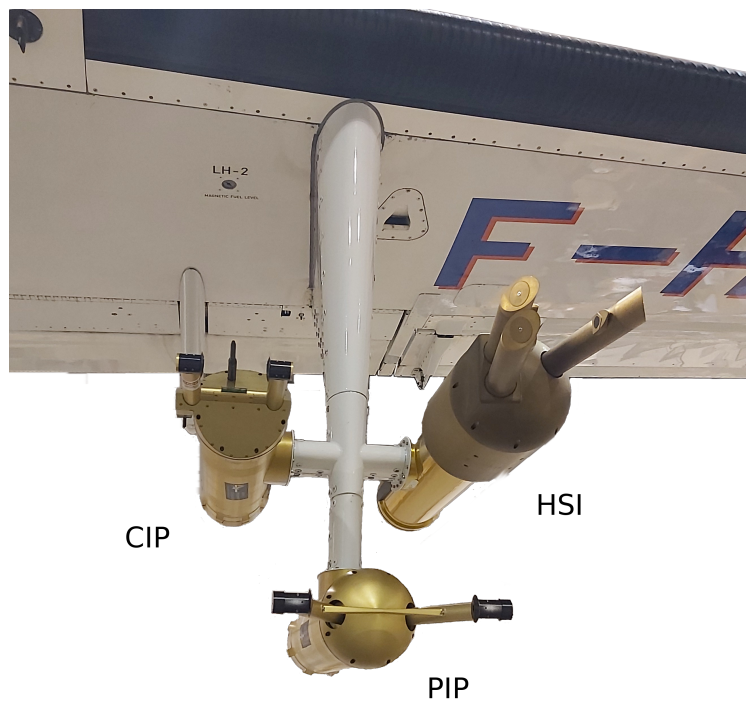
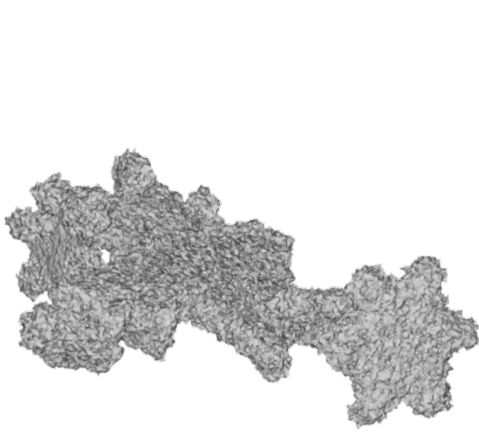


Fig. 2 CIP, HSI and PIP probes installed on the left wing of the ATR 42 managed by SAFIRE, the French facility for airborne research.

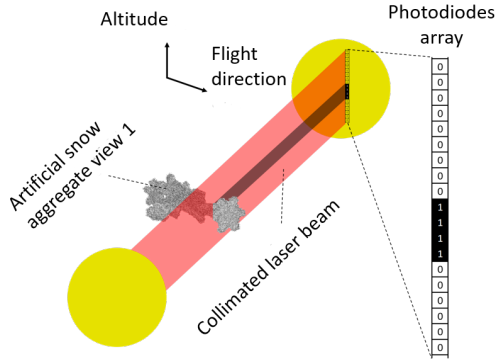
principle of the PIP is illustrated in Fig. 3. It is an imager used to measure the snowflake shape from shadow images of particles passing through a collimated laser beam that are projected onto a linear array of 64 photodetectors. Particle images are reconstructed from individual 1D slices, each slice corresponding to the part of the particle illuminated by the laser beam at a given time (Figs. 3c and 3d). The snowflake is then represented by a single projected 2D image (Figs. 3e and 3f), which limits the geometric description of the whole particle. The post-processed image obtained depends on the position of the PIP sensors with respect to the orientation of the particle. This is illustrated in Fig. 3 where the



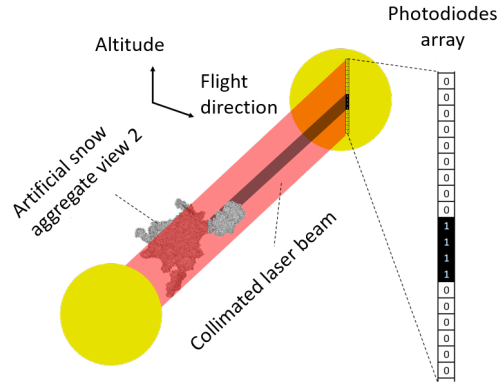
(a) Snowflake view #1.



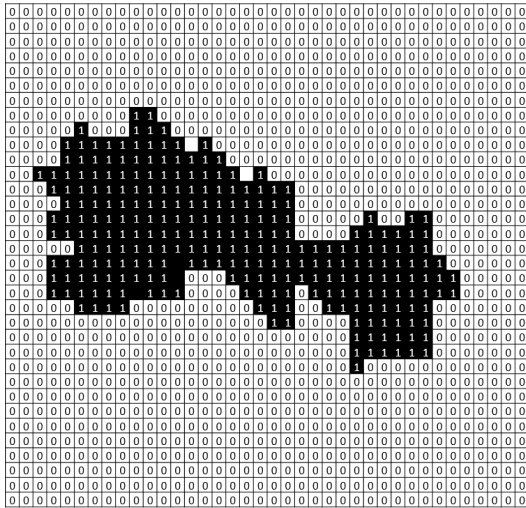
(b) Snowflake view #2.



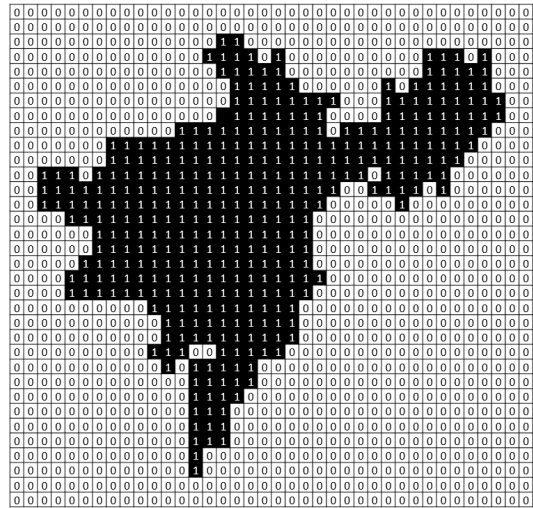
(c) Image acquisition from PIP (view #1).



(d) Image acquisition from PIP (view #2).



(e) Post processed image (view #1).



(f) Post processed image (view #2).

Fig. 3 Snowflake post processed image from PIP. Left: view #1. Right: view #2.

same snowflake is shown from two different orientations (Fig. 3a for view #1 and Fig. 3b for view #2) resulting in two different 2D images from the PIP. In addition, from the point of view of drag force calculation, the position of the PIP sensor relative to the particle is not necessarily the most relevant since the measured projected area should normally correspond to the cross section of the particle normal to the flow. Furthermore, the characterization of the snowflake shape by a single image does not account for the anisotropy of the particle.

The variability in particle shape and bulk density was not addressed by the European projects HAIC [11] and MUSIC-haic [12], where models for trajectory, impact and erosion [13–29] were derived for ice particles (referred to as *ice crystals* in the rest of the manuscript) with high bulk densities (typically of the order of 917 kg.m^{-3}) and fairly regular geometric shapes (sphericities larger than 0.8). Regarding the specific risk of snow accretion or accumulation, there are currently no validated engineering tools (test facilities and numerical tools) available for snowflakes. The extension and validation of the models from ice crystal icing to snowy conditions is the main objective of this paper. More specifically the focus is on the modeling of the drag coefficient of snowflakes. The influence of the accuracy of the input data for the models will be studied. To this end, emphasis will be placed on the geometric procedure which is used to reconstruct a simplified 3D shape of the particle from a single post-processed PIP image (typical data format from a flight test campaign). This work is thus part of an effort to meet the requirements of the aeronautical community, namely to propose models whose input data are compatible with those from wind tunnels and flight tests.

The rest of the paper is divided in four parts followed by some conclusions. In a first part, the experiments dedicated to the measurement of the drag coefficient of a free falling snowflake are described. The procedure for post-processing images generated by the experiments and allowing a 2D to 3D geometric reconstruction is also described. The second part presents the extension of the drag models for ice crystals to the case of snowflakes and the third part is dedicated to the comparison between model predictions and experimental results. A fourth part is dedicated to discussion.

I. Experiments: free fall of a snowflake

A. Experimental Apparatus

For the experiments, artificial snowflakes are used, generated using the SnowFALL device provided by the company IAG. These snowflakes have an irregular shape, close to natural snowflakes like those depicted in Fig. 1. The experimental setup used for the measurement of snowflake drag is schematically shown in Fig. 4. The entire setup and measurement equipment is mobile and can be operated at ambient temperatures down to -20°C . The setup consists of a large PMMA tube with a diameter and height of 150 mm and 300 mm, respectively. Through a baffle (opening of 10 mm) at its top, free falling snowflakes enter the PMMA tube, which has the purpose of shielding them from possible external gusts influencing their trajectory. Immediately after a snowflake falls out of the PMMA tube, its movement is captured using high-speed cameras in a front, side and bottom view using a recording speed of 3000 frames

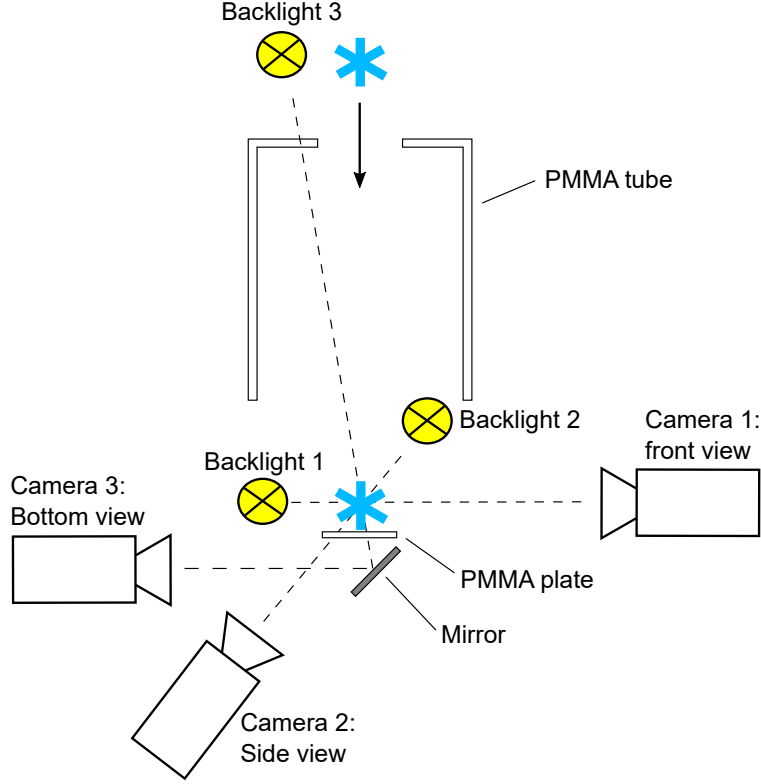


Fig. 4 Schematic view of the experimental setup used to measure snowflake drag.

per second. The resolutions of the front, bottom and side views are $24.1 \mu\text{m}/\text{pixel}$, $34.6 \mu\text{m}/\text{pixel}$ and $51.7 \mu\text{m}/\text{pixel}$, respectively. The side view has the largest field of view and is primarily used for measuring the free fall velocity. After the measurement of their terminal velocity, the snowflakes are caught and melted on a transparent, superhydrophobic coated PMMA plate, forming nearly spherical liquid drops. These drops are captured using the front view camera, which allows measurement of the drop diameter and therefore, the computation of the snowflake mass. Measurement of the snowflake projected area in three nearly orthogonal views, as well as measuring their mass, allows the calculation of the snowflake drag coefficient corresponding to its characteristic terminal velocity. Figure 5 shows example images of a snowflake in the front view, falling with a terminal velocity of 1.36 m/s and finally landing on the PMMA plate. The snowflake's maximum dimension is 3.5 mm . As shown in Fig. 5, a snowflake might be very fragile and shatter when it impinge the PMMA plate, even if it falls with a rather moderate terminal velocity. In the case of fragmentation, all fragments are collected and melted together, using a stream of hot air directed toward the bottom of the PMMA plate. As a result, an almost spherical drop is formed on top of the superhydrophobic PMMA plate, an example which can be seen in Fig. 6. It can be observed that the droplet surface is flat at its contact with the PMMA plate. Given that the drop can be depicted as a body of revolution, its geometry can be defined using only a single image in which the drop symmetry axis is aligned with the camera focal plane. The symmetry axis of the drop is assumed to be aligned normal to the PMMA surface, intersecting with the drop centroid and the distance of each pixel inside the drop to this axis in

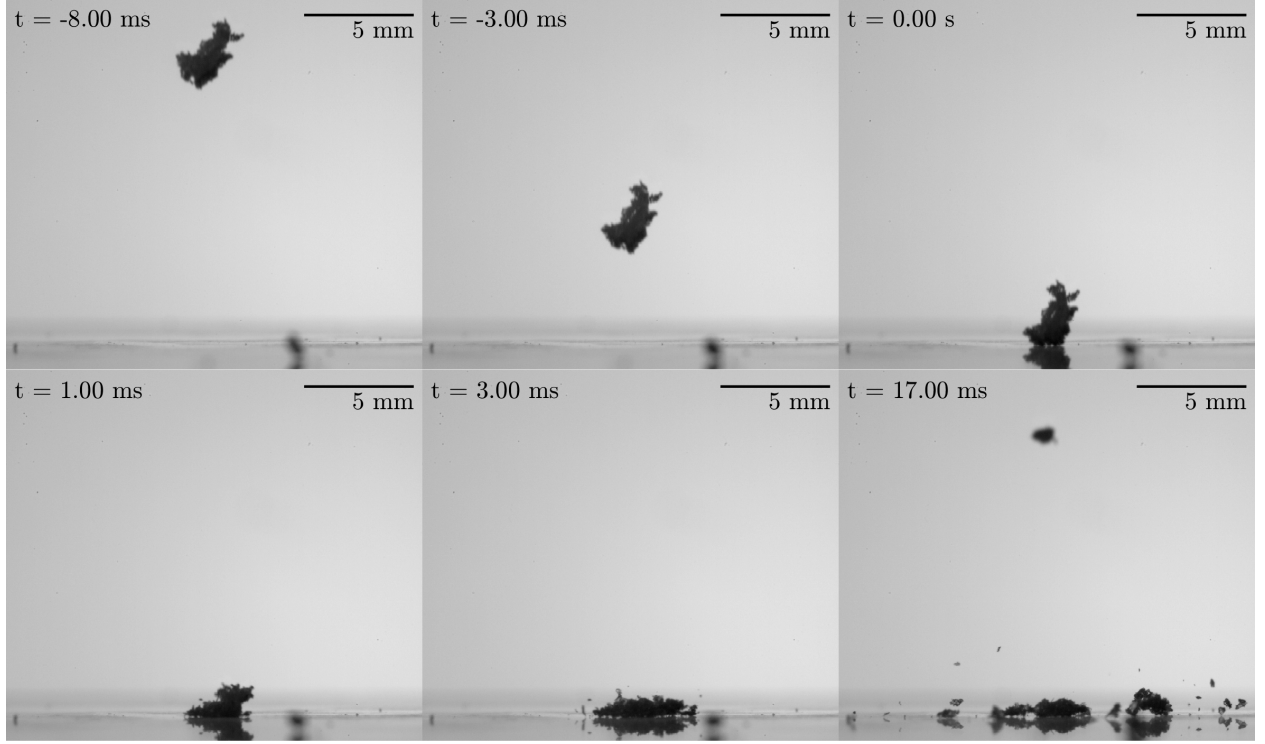


Fig. 5 Example images of a snowflake recorded with camera 1 (front view) having a resolution of $24.1 \mu\text{m}/\text{pixel}$.

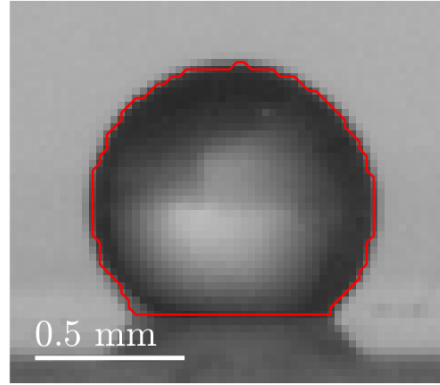


Fig. 6 Image of the melted snowflake, having a mass of 0.263 mg and a diameter of 0.795 mm .

meters, r_j , is determined. Then, the drop mass can be computed as:

$$m_p = \rho_w \sum_{j=1}^N \frac{\pi x_{px}}{2} \left[(r_j + x_{px}/2)^2 - (r_j - x_{px}/2)^2 \right], \quad (1)$$

where ρ_w is the water density, x_{px} is the width of one pixel in meters and N is the number of all pixels located inside of the drop. The diameter of the drop is approximately 0.8 mm and its mass is computed as 0.263 mg . Using the snowflake

images of the front, side and bottom view, characteristic size and shape parameters can be derived. Unlike a PIP where only a single image is available, the three views obtained from the three high-speed cameras will allow to evaluate the sensitivity of the computed terminal falling velocity with respect to the position of the camera used for the 2D to 3D geometric reconstruction of the particle (see Sec. I.B).

B. Image post-processing

In this part we briefly present the post-processing procedure of the images acquired during the experiments described in the previous section. The technique described here is the one that would be also used to process PIP images. Umbroscopic measurements provide grayscale or black and white images of the snowflake shadow area in the direction of the camera (Fig. 7a). A preliminary step of digitizing the grayscale image is performed. For each pixel of the image,

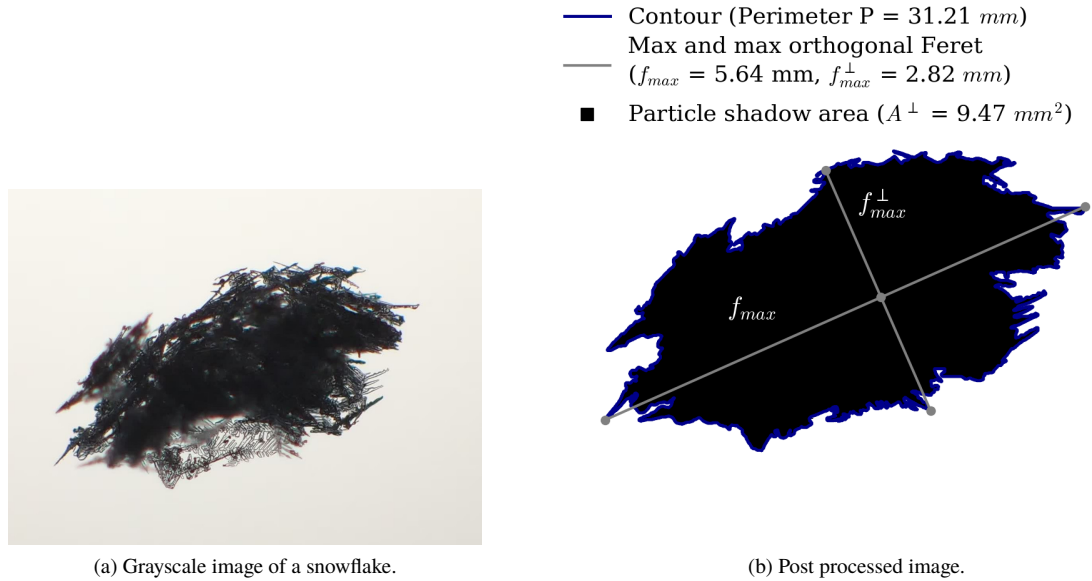


Fig. 7 Data post-processing illustration.

if the gray level is higher than that of the background of the image, the value of the pixel is imposed to 1. Otherwise, the value of the pixel is imposed to 0. An automatic image post-processing tool has been developed to provide the maximum f_{max} and maximum crosswise f_{max}^{\perp} diameters of Feret (in gray in Fig. 7b) from the digitized image. Reconstructed snowflake shadow area A^{\perp} is also derived (in black in Fig. 7b). f_{max}^{\perp} is defined as the largest diameter among all diameters orthogonal to f_{max} . Using Feret's diameters, it can be deduced an ellipse of major semi axis $a = f_{max}/2$ and minor semi axis $c = f_{max}^{\perp}/2$ (in gray in Fig. 8). From the axes of the ellipse it can be constructed either an oblate spheroid (generated by rotation of the ellipse around its z minor axis, Fig. 8a) or a prolate spheroid (generated by rotation of the ellipse around its y major axis, Fig. 8b) of volume $V_{spheroid} = \pi/6 A^2 C$. For an oblate spheroid $A = 2a$

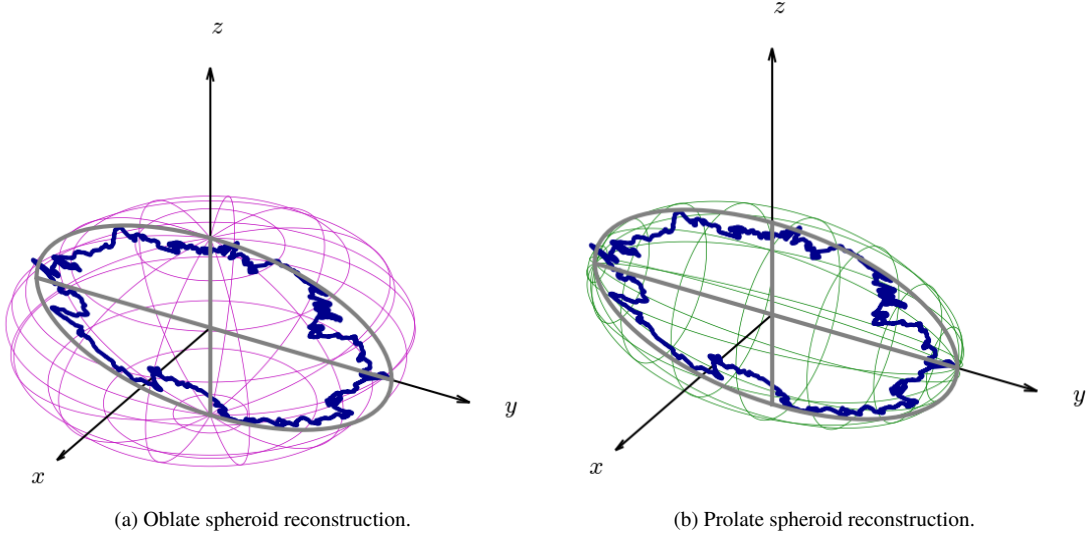


Fig. 8 Oblate and prolate spheroid reconstructions.

and $C = 2c$, while for a prolate spheroid $A = 2c$ and $C = 2a$. Using these parameters, the following characteristics can be deduced:

$$d_V = \left(\frac{6V_{spheroid}}{\pi} \right)^{1/3}, \quad \Phi = \frac{\pi d_V^2}{A_{spheroid}}, \quad \Phi^\perp = \frac{\frac{\pi}{4} d_V^2}{A^\perp} \quad (2)$$

where d_V is the volume equivalent diameter, A^\perp the snowflake shadow area (Fig. 7b) and $A_{spheroid}$ is the surface of the spheroid. $A_{spheroid} = 2\pi a^2 + \pi \frac{c^2}{e} \ln \left(\frac{1+e}{1-e} \right)$ for an oblate spheroid and $A_{spheroid} = 2\pi a^2 \left(1 + \frac{c}{ae} \arcsin e \right)$ for a prolate spheroid. Φ and Φ^\perp are respectively the particle sphericity and crosswise sphericity. The eccentricity of the ellipse is defined as $e = \sqrt{1 - \frac{c^2}{a^2}}$.

II. Models for snowflake trajectory

For the computation of snow particle trajectory, only the drag force is considered in the momentum equation, and the influences of the added mass force, the Basset force, and the lift force are neglected. Regarding the added mass and Basset forces, this is justified because the density of the air is much lower than the bulk density of a snowflake. Concerning the lift force, this comes from the fact that a snowflake rotates on itself during its motion so that the average value (but not the instantaneous value) of the lift force can be considered as zero. Hence, the particle motion equation simply reads:

$$m_p \frac{d\mathbf{v}_p}{dt} = \mathbf{F}_D + m_p \mathbf{g}, \quad \mathbf{F}_D = -\frac{1}{2} \rho_a \|\mathbf{w}\| \mathbf{w} A_{ref} C_D(Re_p) \quad (3)$$

where m_p is the snow particle mass, \mathbf{F}_D the drag force, \mathbf{g} the gravity, ρ_a the air density and $\mathbf{w} = \mathbf{v}_p - \mathbf{v}_a$ the relative velocity between the air (\mathbf{v}_a) and the particle (\mathbf{v}_p). In the following, $\|\mathbf{w}\|$ will simply be denoted w . The drag coefficient

C_D is defined with respect to the reference surface A_{ref} . It is a function of the particle Reynolds number $Re_p = \frac{\rho_a w d_{ref}}{\mu_a}$ where μ_a is the air dynamic viscosity, d_{ref} a characteristic length scale and ρ_a the air density. Note that in the present study the mass of the particle m_p is known experimentally. For free fall conditions, the airflow being at rest, the stationary final particle velocity v_{p_∞} is the solution of Eq. (3) for which the left-hand term has been cancelled:

$$v_{p_\infty} = \sqrt{\frac{2m_p g}{\rho_a A_{ref} C_D(Re_p)}} \quad (4)$$

At steady state, the particle Reynolds number Re_p being itself a function of v_{p_∞} , Eq. (4) is an implicit equation of the unknown v_{p_∞} which is solved numerically by a Newton algorithm. Note that, according to Eq. (4), if the mass of the particle is known (which is the case here), the velocity v_{p_∞} is a function of the only product $A_{ref} \cdot C_D$. The models described in the rest of the manuscript will assign different weights to the two contributions A_{ref} and C_D .

From a single post processed image of the snowflake (obtained for instance from the PIP), two sets of definitions will be compared to define d_{ref} (as a parameter of Re_p), A_{ref} and C_D . The first one, based only on a surface description of the particle, is the underlying assumption of models denoted as *surface-based geometric models* in the rest of the manuscript. This is the assumption made by the historical Heymsfield and Westbrook model for snowflakes [2] for which a short description is proposed in Sec. II.A. This model will be the baseline for evaluating the drag coefficient of complex-shaped snow particles. The second one is based on a volume description of the particle, leading to *volume-based geometric models* according to the denomination used in the rest of the paper. This is achieved by introducing the diameters d_V and d_m , defining respectively the volume and mass equivalent diameters corresponding to the diameter of the equivalent spherical particle having the same volume (resp. mass) as the snowflake. This is the approach adopted by the Haider *et al.* [30], Ganser [31] and Hölzer *et al.* [32] correlations for non-spherical particles and which has already been validated for ice crystal icing [20, 29]. The extension of these last three models to snowflakes is proposed in Sec. II.B.

A. Surface-based geometric model for C_D : the Heymsfield and Westbrook (H&W) approach

The Heymsfield and Westbrook model [2] is an improvement of the models based on the Abraham's drag coefficient of a sphere [33].

1. The Abraham model for spherical particles

In Abraham [33], the problem is considered as an assembly of a smooth sphere of radius a plus a thin boundary layer of thickness δ in a moving fluid of density ρ_a at a velocity v . The assembly is then assumed to evolve in an asymptotic low viscosity aerodynamic flow in a regime where the drag coefficient is independant of Re_p . These assumptions lead

to the following expression for the drag force of a sphere of radius $a + \delta$:

$$F_D = \frac{\rho_a v^2}{2} \pi (a + \delta)^2 C_{D_0} \quad (5)$$

where C_{D_0} is a constant parameter to be adjusted. Equation (5) may be re-written as:

$$F_D = \frac{\rho_a v^2}{2} \pi a^2 C_{D_0} \left(1 + \frac{\delta}{a}\right)^2 \quad (6)$$

which is the drag force for a particle of radius a with $C_{D_0} \left(1 + \frac{\delta}{a}\right)^2$ as drag coefficient. The boundary layer thickness is given by $\frac{\delta}{a} = \frac{\delta_0}{\sqrt{Re_p}}$ where δ_0 is a dimensionless coefficient. This leads to the following drag coefficient definition for the particle:

$$C_D = C_{D_0} \left[1 + \frac{\delta_0}{\sqrt{Re_p}}\right]^2 \quad (7)$$

where C_{D_0} and δ_0 are chosen to ensure $C_{D_0} \delta_0^2 = 24$ (Stokes regime) and $\delta_0 = 9.06$.

2. Extension to non-spherical particles

Many authors have used Eq. (7) for non-spherical particles with $C_{D_0} = 0.35$ and $\delta_0 = 8$. In the definition of Re_p , the diameter d_{ref} is then chosen as:

$$d_{ref} = f_{max} \quad (8)$$

where f_{max} is the maximum Feret diameter (Fig. 7b). The reference surface A_{ref} is given by:

$$A_{ref} = A^\perp \quad (9)$$

Mitchell, Heymsfield, Khvorostyanov and Curry (MHKC) [3–7] used this method to calculate the drag coefficient and compared the resulting fall velocities for a wide range of particles. Planar and columnar-type, graupel and aggregates with a size range from 250 μm to 8 mm were studied. More recently, Heymsfield and Westbrook found that fall velocities were overestimated for particles with a low area ratio. In some cases the relative error exceeded 100 % and this effect appears to be stronger at low Reynolds number. Needles, dendrites, stellars and aggregates are key particle types that are affected by this bias. To mitigate this high sensitivity to the area ratio, they considered the following modified drag coefficient [2]:

$$C_D^* = C_D A_r^{1/2} \quad (10)$$

where A_r is the area ratio defined by:

$$A_r = \frac{A^\perp}{\pi f_{max}^2 / 4} \quad (11)$$

Provided that m_p is known (which is the case here), all necessary geometric information for the computation of the drag coefficient C_D is extracted from the 2D image (f_{max} and A^\perp).

B. Volume-based geometric models for C_D : the Haider *et al.*, Ganser and Hölzer *et al.* approaches

Among all the general mathematical expressions that can be found in the literature for the drag coefficient of non-spherical particles, we can mention the models of Haider & Levenspiel (H&L) [30], Ganser (Ga) [31] and Hölzer & Sommerfeld (H&S) [32]. for which the drag coefficients, respectively denoted as $C_D^{H\&L}$, C_D^{Ga} and $C_D^{H\&S}$, can be written as a function of Re_p , Φ and Φ^\perp as follows (see [30–32] for a complete description and the appendix for a summary of the models):

$$C_D^{H\&L} = C_D^{H\&L}(Re_p, \Phi) \quad (12a)$$

$$C_D^{Ga} = C_D^{Ga}(Re_p, \Phi, \Phi^\perp) \quad (12b)$$

$$C_D^{H\&S} = C_D^{H\&S}(Re_p, \Phi, \Phi^\perp) \quad (12c)$$

For the three models H&L, Ga and H&S, several definitions are possible for d_{ref} . These will be given below. From d_{ref} , the reference surface is defined as:

$$A_{ref} = \pi d_{ref}^2 / 4 \quad (13)$$

while the particle Reynolds number is given by:

$$Re_p = \frac{\rho_a w d_{ref}}{\mu_a} \quad (14)$$

The main point is the definition of the particle reference diameter d_{ref} . Two approaches will be compared to define it, namely an *ice-core density-based* approach and an *ice-bulk density-based* approach.

1. Ice-core density-based approach

This is the approach proposed in the framework of ice crystal icing [20, 29] where ice crystals are assumed to be compact with a density close to the ice core density ρ_i^{core} :

$$\rho_p = \rho_i^{\text{core}} \quad (15)$$

where $\rho_i^{\text{core}} = 917 \text{ kg.m}^{-3}$. The reference diameter d_{ref} is then defined as the mass equivalent diameter d_m corresponding to the diameter of the equivalent spherical particle of density ρ_i^{core} having the same mass m_p as the snowflake:

$$d_{ref} = d_m = \left(\frac{6m_p}{\pi\rho_i^{\text{core}}} \right)^{1/3} \quad (16)$$

The volume of the equivalent spherical particle is given by m_p/ρ_i^{core} . The values of Φ and Φ^\perp are respectively defined in this case as:

$$\Phi = \frac{\pi d_m^2}{A_p} \quad (17a)$$

$$\Phi^\perp = \frac{(\pi/4) d_m^2}{A^{\perp low}} \quad (17b)$$

where A_p is the surface of the snowflake and $A^{\perp low}$ its projected surface normal to the airflow. Note that A_p and $A^{\perp low}$ are not known a priori.

In the spirit of the models derived for ice crystal icing [20, 29], the ice-core density-based approach is constructed from the single hypothesis $\rho_p = \rho_i^{\text{core}}$. This approach is therefore suitable when the other geometric parameters d_m (or equivalently m_p , Eq. (16)), Φ and Φ^\perp describing the snowflake are provided experimentally by respectively particle mass, sphericity and orthogonal sphericity distributions representing the variety of snowflakes encountered in the clouds.

2. Ice-bulk density-based approach

The reference diameter d_{ref} is defined as the volume equivalent diameter d_V corresponding to the diameter of the equivalent spherical particle having the same volume V_p as the snowflake:

$$d_{ref} = d_V = \left(\frac{6V_p}{\pi} \right)^{1/3} \quad (18)$$

V_p is defined here as $V_p = V_{spheroid}$, where $V_{spheroid}$ is the volume of the reconstructed oblate or prolate spheroid from planar images of the particle (Eq. (2), Sec. I.B). The density of the particle, then defined by:

$$\rho_p = \rho_i^{\text{bulk}} = \frac{m_p}{V_{spheroid}} \quad (19)$$

can be considered as an ice bulk density such that $\rho_i^{\text{bulk}} < \rho_i^{\text{core}}$. The values of Φ and Φ^\perp are computed from Eq. (2).

The ice-bulk density-based approach is therefore suitable when images are provided by the experiments allowing a spheroid geometrical reconstruction of the snowflake from a single planar view (Sec. I.B).

III. Model validation using experimental data

The volume-based geometric models (H&L, Ga and H&S, Sec. II.B) are compared with the surface-based reference geometric model (H&W Sec. II.A) for the computation of the drag coefficient for snowflakes. Moreover, regarding the volume-based geometric models, the description of the snowflake by a dense particle (ice-core density-based approach, Sec. II.B.1) is compared to that by a less dense particle (ice-bulk density-based approach, Sec. II.B.2). The influence of the 2D to 3D reconstruction (oblate or prolate spheroid) is also studied in the case of the ice-bulk density-based description of the particle. This is summarized in Tab. 1 where the abbreviations used in the following are specified.

Table 1 Abbreviations used for model comparison.

Type of geometric representation	Particle equivalent density (only for volume-based models)	Abbreviation	Ref.
Surface-based geometric model		H&W	Sec. II.A
Volume-based geometric models	ice-core density-based ($\rho_p = \rho_i^{\text{core}}$)	H&L ^{core}	Sec. II.B, Eqs. (16) and (15)
		Ga ^{core}	
		H&S ^{core}	
	ice-bulk density-based ($\rho_p = \rho_i^{\text{bulk}}$)	H&L ^{bulk}	Sec. II.B, Eqs. (18) and (19)
		Ga ^{bulk}	
		H&S ^{bulk}	

Terminal free fall velocities for 16 snowflakes are measured with the experimental method described in Sec. I.A. Conditions are described in row "Experiments" of Tab.2. For each of the studied drag models, the reconstructed geometrical parameters are also given for each of the three views (front, side and bottom). The averaged values of the parameters over the three views are noted *av*. The standard deviation is written *std*. Note that for each snowflake, the mass is measured experimentally (Sec. I.A). In this study, m_p is therefore a known input for the models.

A. Ice-bulk density-based description for volume-based geometric models

Figure 9 shows the comparison between the experimental ($\bar{v}_{p_\infty}^{exp}$, *X* axis) and numerical ($\bar{v}_{p_\infty}^{num}$, *Y* axis) results for the snowflake terminal velocity. The H&L^{bulk}, Ga^{bulk} and H&S^{bulk} models are compared with the H&W model (reference). The sensitivity to the 2D to 3D reconstruction is studied (oblate vs. prolate spheroid). Main diagonal (black solid line) means a perfect agreement. $\pm 30\%$ relative error regions are shown (dashed lines). For the numerical solutions, each symbol in Fig. 9 is obtained by the arithmetic average (noted $\bar{v}_{p_\infty}^{num}$) of the three velocities computed from each of the three geometric reconstructions, which are themselves obtained from each of the three views (front, side, bottom). The three volume-based geometric models H&L^{bulk}, Ga^{bulk} and H&S^{bulk} are able to reproduce the terminal free fall velocity with an average relative accuracy of 30%, which is similar to that of the reference model H&W, except for the models H&L^{bulk} and Ga^{bulk} which tend to overestimate the snowflake fall velocity for some test cases, either for the prolate or oblate spheroid reconstruction. Generally speaking, the prolate or oblate spheroid reconstruction has a minor influence

Table 2 Snowflake free fall study. 16 test cases. Experimental conditions and reconstructed geometric parameters

N		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
Experiments	Fall speed $v_{p_{\infty}}$ (m/s)	0.67	1.83	1.47	1.49	1.44	1.07	1.8	0.92	1.38	1.69	1.28	1.49	1.64	1.41	1.51	1.21
	Mass m_p (mg)	0.063	1.48	0.576	0.132	0.546	0.17	0.298	0.047	0.243	0.459	0.157	0.134	0.349	0.578	0.266	0.15
(H&L, Ga, H&S) ^{core}																	
	d_m (mm)	0.51	1.46	1.06	0.65	1.04	0.71	0.85	0.46	0.8	0.98	0.69	0.65	0.9	1.06	0.82	0.68
	ρ_p (kg/m ³)	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0	917.0
(H&L, Ga, H&S) ^{bulk} (oblate reconstruction)	d_V (mm)	front	1.12	2.98	2.41	1.11	3.16	1.63	1.77	1.02	2.19	1.99	1.33	1.73	1.58	1.57	3.39
		side	1.24	2.74	2.44	1.36	3.76	1.94	1.7	0.9	2.2	2.57	1.64	1.13	1.69	3.02	3.58
		bottom	1.39	3.58	2.95	1.32	3.19	2.07	1.54	1.31	2.6	2.39	1.8	1.36	1.66	2.57	4.21
		av.	1.25	3.1	2.6	1.26	3.37	1.88	1.67	1.08	2.33	2.32	1.59	1.41	1.64	2.39	3.73
		std.	0.11	0.35	0.25	0.11	0.27	0.18	0.09	0.17	0.19	0.24	0.19	0.25	0.05	0.61	0.35
	ρ_b (kg/m ³)	front	86.0	107.0	79.0	183.0	33.0	74.0	103.0	85.0	44.0	111.0	126.0	50.0	169.0	288.0	13.0
		side	64.0	138.0	76.0	101.0	20.0	44.0	115.0	125.0	44.0	52.0	68.0	177.0	139.0	40.0	11.0
		bottom	45.0	62.0	43.0	110.0	32.0	37.0	155.0	40.0	27.0	64.0	52.0	101.0	145.0	65.0	7.0
		av.	65.0	102.0	66.0	131.0	28.0	52.0	124.0	83.0	38.0	76.0	82.0	109.0	151.0	131.0	10.0
		std.	17.0	31.0	16.0	37.0	6.0	16.0	22.0	35.0	8.0	26.0	32.0	52.0	13.0	111.0	3.0
	Φ (-)	front	0.97	0.91	0.96	0.99	0.96	0.98	1.0	0.98	0.97	0.99	0.87	0.96	0.99	0.98	0.93
		side	0.86	0.92	0.96	0.97	0.99	0.96	0.95	0.81	0.96	0.94	0.91	0.92	0.98	0.98	0.89
		bottom	0.98	0.98	1.0	0.99	0.98	0.97	0.95	0.92	1.0	0.96	0.98	0.84	0.98	0.98	1.0
		av.	0.93	0.93	0.97	0.98	0.98	0.97	0.97	0.9	0.98	0.96	0.92	0.91	0.98	0.98	0.94
		std.	0.05	0.03	0.02	0.01	0.01	0.01	0.02	0.07	0.02	0.02	0.04	0.05	0.0	0.0	0.04
	Φ^\perp (-)	front	1.41	1.75	1.85	1.41	1.46	1.24	1.72	1.4	1.27	1.31	1.4	1.5	1.31	1.27	1.54
		side	1.6	1.4	1.57	1.34	1.53	1.5	1.39	0.96	1.33	1.46	1.48	1.14	1.38	1.41	1.64
		bottom	1.53	1.37	2.03	1.29	1.27	1.51	1.59	1.38	1.49	1.39	1.33	1.46	1.36	1.5	1.55
		av.	1.51	1.51	1.82	1.35	1.42	1.42	1.56	1.25	1.37	1.39	1.4	1.37	1.35	1.39	1.58
		std.	0.08	0.17	0.19	0.05	0.11	0.12	0.13	0.2	0.09	0.06	0.06	0.16	0.03	0.1	0.04
	d_V (mm)	front	0.97	2.35	2.06	1.03	2.72	1.48	1.73	0.9	1.9	1.83	1.0	1.48	1.44	1.42	2.78
		side	0.91	2.18	2.1	1.19	3.47	1.67	1.44	0.63	1.9	2.11	1.3	0.91	1.52	2.74	2.76
		bottom	1.24	3.18	2.86	1.21	2.86	1.81	1.31	1.06	2.52	2.04	1.59	0.99	1.5	2.29	4.05
		av.	1.04	2.57	2.34	1.14	3.02	1.65	1.49	0.86	2.11	1.99	1.3	1.13	1.49	2.15	3.19
		std.	0.14	0.44	0.37	0.08	0.33	0.14	0.18	0.29	0.12	0.24	0.25	0.03	0.55	0.6	0.25
	ρ_b (kg/m ³)	front	130.0	217.0	126.0	230.0	52.0	100.0	110.0	122.0	67.0	143.0	296.0	79.0	222.0	389.0	24.0
		side	157.0	272.0	119.0	149.0	25.0	70.0	189.0	365.0	68.0	93.0	135.0	337.0	190.0	54.0	24.0
		bottom	64.0	88.0	47.0	143.0	45.0	55.0	255.0	76.0	29.0	104.0	75.0	266.0	198.0	92.0	8.0
		av.	117.0	192.0	97.0	174.0	40.0	75.0	185.0	188.0	55.0	113.0	168.0	227.0	203.0	178.0	19.0
		std.	39.0	77.0	36.0	40.0	11.0	19.0	59.0	127.0	18.0	22.0	93.0	109.0	14.0	150.0	8.0
	Φ (-)	front	0.97	0.93	0.97	0.99	0.97	0.98	1.0	0.98	0.97	0.99	0.9	0.97	0.99	0.98	0.95
		side	0.89	0.93	0.97	0.97	0.99	0.97	0.96	0.85	0.97	0.95	0.93	0.94	0.98	0.99	0.91
		bottom	0.98	0.98	1.0	0.99	0.98	0.97	0.96	0.94	1.0	0.96	0.98	0.87	0.98	0.98	1.0
		av.	0.95	0.95	0.98	0.98	0.98	0.98	0.97	0.92	0.98	0.97	0.94	0.93	0.98	0.98	0.95
		std.	0.04	0.02	0.01	0.01	0.01	0.01	0.02	0.05	0.01	0.02	0.03	0.04	0.0	0.0	0.04
	Φ^\perp (-)	front	1.07	1.09	1.36	1.21	1.08	1.02	1.64	1.1	0.96	1.11	0.79	1.1	1.09	1.04	1.04
		side	0.88	0.89	1.17	1.03	1.31	1.1	1.0	0.47	0.99	0.99	0.93	0.74	1.12	1.16	0.97
		bottom	1.21	1.08	1.9	1.09	1.02	1.16	1.14	0.89	1.41	1.01	1.04	0.76	1.11	1.19	1.44
		av.	1.05	1.02	1.47	1.11	1.14	1.09	1.26	0.82	1.12	1.04	0.92	0.87	1.1	1.13	1.15
		std.	0.14	0.09	0.31	0.07	0.12	0.06	0.27	0.26	0.21	0.05	0.1	0.17	0.01	0.07	0.21
H&W	f_{max} (mm)	front	1.29	3.78	2.81	1.2	3.67	1.81	1.81	1.15	2.53	2.17	1.77	2.01	1.73	1.73	4.14
		side	1.67	3.43	2.83	1.55	4.06	2.26	2.01	1.28	2.55	3.11	2.07	1.4	1.87	3.32	4.65
		bottom	1.56	4.02	3.05	1.44	3.56	2.36	1.82	1.62	2.67	2.81	2.03	1.88	1.85	2.89	4.38
		av.	1.5	3.75	2.9	1.4	3.76	2.14	1.88	1.35	2.58	2.7	1.96	1.77	1.82	2.65	4.39
		std.	0.16	0.24	0.11	0.15	0.21	0.24	0.09	0.2	0.06	0.4	0.13	0.26	0.06	0.67	0.21
	A_r (-)	front	0.54	0.36	0.4	0.61	0.51	0.66	0.56	0.56	0.59	0.64	0.4	0.49	0.64	0.65	0.44
		side	0.34	0.45	0.47	0.57	0.56	0.49	0.52	0.51	0.56	0.47	0.42	0.57	0.59	0.59	0.36
		bottom	0.52	0.58	0.46	0.65	0.63	0.51	0.45	0.47	0.63	0.52	0.59	0.36	0.6	0.53	0.59
		av.	0.47	0.46	0.44	0.61	0.57	0.55	0.51	0.51	0.59	0.54	0.47	0.47	0.61	0.59	0.46
		std.	0.09	0.09	0.03	0.03	0.05	0.08	0.04	0.03	0.03	0.07	0.08	0.09	0.02	0.05	0.1

on the computation of the terminal velocity. The only point for which all models have a relative error greater than 30% is case 15 which is characterized by the lowest reconstructed bulk densities of the entire database.

For each of the models $H\&S^{\text{bulk}}$, Ga^{bulk} , $H\&L^{\text{bulk}}$ and $H\&W$, and in addition to Fig. 9, Figs. 10 and 11 show error bars accounting for the snowflakes being described by three views (front, side and bottom, see Sec. I.A). The legend is similar to that of Fig. 9. The symbols showing $\bar{v}_{p_\infty}^{\text{num}}$ are reminded. The lengths of the error bars correspond to $\bar{v}_{p_\infty}^{\text{num}} \pm \sigma$ where σ is the standard deviation calculated from $\bar{v}_{p_\infty}^{\text{num}}$ and the different velocities computed from the particles reconstructed from each of the three views. The error bar lengths indicate the dispersion of the computed free fall velocity with respect to the image chosen for the geometric reconstruction of the snowflake. Although the snowflakes were observed to have a privileged direction of fall in the experiments, the choice of the view direction for the oblate or prolate reconstruction has a limited effect on the model error, since the error bars are as a whole in the $\pm 30\%$ relative error regions. This seems to be confirmed by the overall good behavior of the $H\&L$ model which does not depend on Φ^\perp and is therefore less sensitive to the view chosen to reconstruct the spheroid. For the 16 test cases, 31% of the best estimates for $\bar{v}_{p_\infty}^{\text{num}}$ are obtained with the front view, 28% with the bottom view, and 41% with the side view. This indicates that the models can be applied to estimate drag coefficients of snowflakes recorded from a random direction in airborne measurements.

Table 2 shows that the reconstructed densities ρ_b vary from 5.0 kg.m^{-3} to 389 kg.m^{-3} , far from the value of 917 kg.m^{-3} used for ice particles in the framework of ice crystal icing. This is consistent with the observed complex shape of snowflakes as aggregates, which are poorly dense and contain air pockets.

Note that for the 16 test cases in Tab. 2, the range of values for the particle Reynolds number Re_p varies from 44 to 362 for all the particles with the ice-bulk density-based description. This indicates that the drag force is driven by the intermediate law corresponding to a transition region between the Stokes' and Newton's regimes.

B. Ice-core density-based description for volume-based geometric models

Figure 12 shows the comparison between the experimental $v_{p_\infty}^{\text{exp}}$ and numerical $v_{p_\infty}^{\text{num}}$ results for the snowflake terminal velocity. The $H\&L^{\text{core}}$, Ga^{core} and $H\&S^{\text{core}}$ models are compared with the $H\&W$ model (reference). Main diagonal (black solid line) means a perfect agreement. $\pm 30\%$ relative error regions are shown (dashed lines). Since there is no 2D to 3D reconstruction for the ice-core density-based description, the values of Φ and Φ^\perp for the models $H\&L^{\text{core}}$, Ga^{core} and $H\&S^{\text{core}}$ cannot be deduced from the spheroid reconstruction as in the case of ice-bulk density-based models. The lack of particle geometric reconstruction from images also accounts for the missing error bars in Fig. 12. To determine Φ and Φ^\perp , an error minimization procedure is performed on all 16 flakes and for each model between the experimental and numerically computed terminal free fall velocities. Indeed, contrary to ice crystals where $\Phi = 1$ and $\Phi^\perp = 1$ lead to acceptable predictions of the particle trajectories and in particular of the collection efficiency [19, 22], such an assumption is no longer acceptable for the case of aggregates where the geometry is more complex and the

inertia is lower. The mapping of the L_2 -norm error as a function of Φ and Φ^\perp is shown in Fig. 13 for each model. The vertical lines in Fig. 13a simply mean that the H&L model does not depend on Φ^\perp . The values of Φ and Φ^\perp that minimize the L_2 -error are those used in Fig. 12. The optimal values obtained for Φ and Φ^\perp are below 0.4, at levels much smaller than those used in the framework of ice crystal icing where Φ and Φ^\perp were chosen between 0.8 and 1 [19, 22]. Thus, provided that the geometric complexity of the snowflake is taken into account by adapting Φ and Φ^\perp , the description of the snowflake by a dense particle allows the simulation of terminal free fall velocity with an accuracy comparable to the previous method where the snowflake is modeled by a less dense spheroid particle.

For the 16 test cases in Tab. 2, the range of values for the particle Reynolds number Re_p varies from 22 to 170 for all the particles with the ice-core density-based description. As expected, the values for Re_p are lower than for the ice-bulk density-based approach. However, the observed regimes for the drag force are similar, namely the intermediate law between the Stokes' and Newton's regimes.

IV. Discussion

Surface-based and volume-based geometric models provide comparable accuracy for computing the terminal free fall velocity of snowflakes. This validates the extension to snowy conditions of the volume-based geometric models initially developed for ice crystal icing. Among the volume-based models, the ice-bulk and ice-core density-based descriptions are fairly similar in the sense that they assign a different relative weight to C_D and A_{ref} , keeping in mind that only the product $C_D \cdot A_{ref}$ is relevant for the computation of the drag force (Eq. (3)). For the ice-bulk density-based approach, the snowflake is represented by a particle of regular shape and low bulk density, taking into account air gaps inside the particle. This leads to a higher value for A_{ref} and in return a lower value for C_D . On the contrary, for the ice-core density-based approach, the characteristic size of the particle d_{ref} is arbitrarily defined as the equivalent mass diameter thus leading to a smaller value for A_{ref} and in turn to a higher value for C_D (resulting from smaller values of Φ and Φ^\perp). However, even if both approaches ultimately lead to results of comparable accuracy, the ice-bulk density-based approach is more physically satisfactory since it takes into account (in a simplified way) the geometry of the particle and thus does not require the use of adjustable parameters (Φ and Φ^\perp) whose optimal values obtained in the case of this study are not intended to be general.

The preferred use between the ice-bulk or the ice-core density-based geometric reconstruction for the snowflake depends on the type of input data available to describe the particle. If an image of the snowflake is provided, then the ice-bulk density-based approach is to be preferred since all the geometric characteristics of the particle can be deduced from the 3D spheroid reconstruction. If such an image is not available, then the ice-core density-based approach can provide an interesting alternative. All that is needed is then the mass distribution of the particles in the cloud and as well the corresponding distributions for the particle sphericities (and orthogonal sphericities for the Ga and H&S drag models). The need to provide sphericity is clearly the weak point of this second approach. A parametric study in the

spirit of the present study would be necessary to provide sphericity values for different types of particles that may be encountered in a cloud.

Regarding the accuracy of the geometric reconstruction of the snowflake, Leinonen *et al.* [34] proposed a 3D reconstruction from a triplet of images and an artificial intelligence algorithm based on Generative Adversarial Network (GAN) to reconstruct the mass and geometric descriptors of the particle. Future work should propose new drag models with more input data adapted to the improved accuracy of the geometric reconstruction.

Conclusions

In this paper, a comparison is presented between different empirical approaches to estimate the drag force of a snowflake when incomplete information on its shape is available. In this case, it is necessary to use simple geometrical descriptors in order to compute the reference surface A_{ref} of the particle and its drag coefficient C_D , knowing that only the product is relevant for the drag force. The conclusion of the present study, based on the analysis of an experimental database containing 16 snowflakes of various shapes and sizes, is that the different approaches are globally equivalent in terms of relative error on the terminal free fall velocity of the snowflakes. However, the ice-bulk density-based approach seems more satisfactory because there is a simple procedure to determine the geometric parameters (d_V , Φ and Φ^\perp) from a single 2D image of the particle. Determining Φ and Φ^\perp is more problematic in the case of the ice-core density-based approach since these parameters are difficult to be estimated *a priori* for very irregular particles and should rather be considered as adjustable parameters for the computation of the drag coefficient. It remains questionable whether the values of the order of 0.2 to 0.4 found in the present work can be applied to a large class of snowflakes. It is however confirmed [29] that they are far too low for more regular particles such as ice crystals.

The level of confidence in the snowflake mass estimate has not been discussed in this work since it is known experimentally. In reality, the estimation of the particle mass is often provided by correlative mass-diameter relations [8] which are sources of additional inaccuracies. Thus, to improve the prediction of snowflake trajectories, joint efforts will have to be made in the geometrical description of the particles, the estimation of their mass, and the accuracy of the drag coefficient model.

In the context of future developments of heat exchange models for snow particles, the validity of the volume-based geometric models to describe the dynamics of snowflakes is an argument in favor of extending the thermal models initially proposed for ice crystals to snow particles. The latter are actually derived from a Reynolds-type analogy between heat transfer and momentum where drag forces are computed with the H&S model for the drag coefficient C_D [29].

Appendix

A summary of the drag models $C_D^{H\&L}$, C_D^{Ga} and $C_D^{H\&S}$ is provided here.

- For the Haider & Levenspiel (H&L) model [30], the drag coefficient is given by:

$$C_D^{H\&L} = \frac{24}{Re_p} \left(1 + a Re_p^b \right) + \frac{c}{1 + d Re_p^{-1}} \quad (20)$$

where the coefficients a , b , c and d are given by:

$$a = \exp \left(2.3288 - 6.4581\Phi + 2.4486\Phi^2 \right) \quad (21a)$$

$$b = 0.0964 + 0.5565\Phi \quad (21b)$$

$$c = \exp \left(4.905 - 13.8944\Phi + 18.4222\Phi^2 - 10.2599\Phi^3 \right) \quad (21c)$$

$$d = \exp \left(1.4681 + 12.2584\Phi - 20.7322\Phi^2 + 15.8855\Phi^3 \right) \quad (21d)$$

- For the Ganser (Ga) model [31], the drag coefficient is given by:

$$C_D^{Ga} = C_{shape} \cdot C_D^{Ga,\star} \quad (22)$$

where $C_D^{Ga,\star}$ is defined as:

$$C_D^{Ga,\star} = \frac{24}{Re_p^\star} \left(1 + 0.1118 \left(Re_p^\star \right)^{0.6567} \right) + \frac{0.4305}{1 + 3305 \left(Re_p^\star \right)^{-1}} \quad (23)$$

Re_p^\star is written as a function of the particle Reynolds number Re_p as follows:

$$Re_p^\star = \frac{C_{shape} Re_p}{f_{shape}} \quad (24)$$

In Eqs. (22) and (24), C_{shape} and f_{shape} are defined as:

$$C_{shape} = 10^{a[-\log(\Phi)]^b} \quad (25a)$$

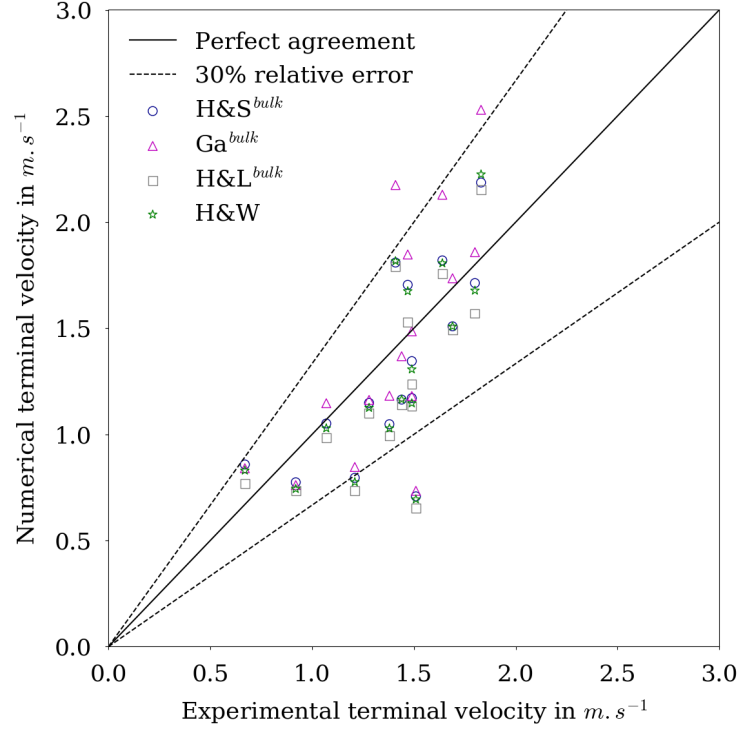
$$f_{shape} = \frac{2}{3}\Phi^{-1/2} + \frac{1}{3}(\Phi^\perp)^{-1/2} \quad (25b)$$

- For the Hölzer & Sommerfeld (H&S) model [32], the drag coefficient is given by:

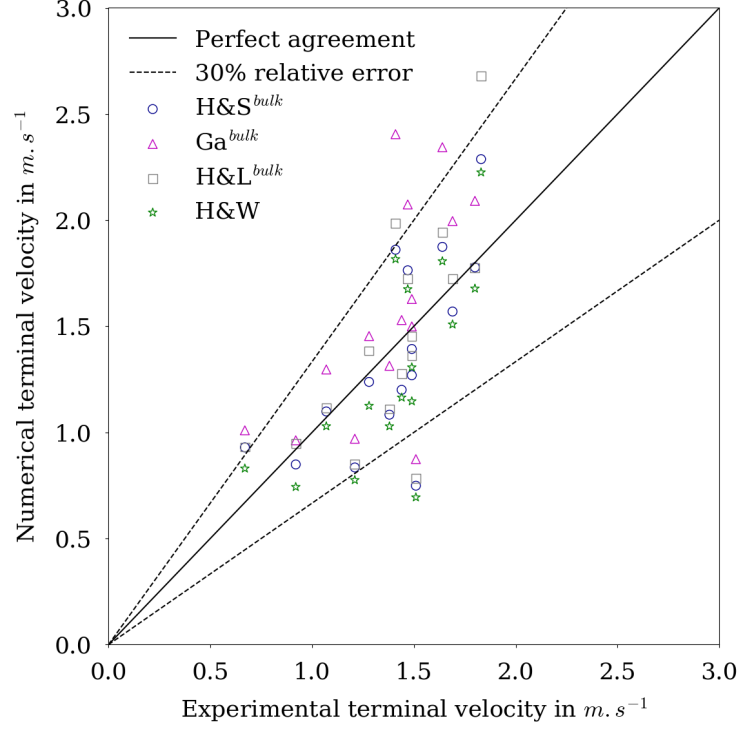
$$C_D^{H\&S} = \frac{8}{Re_p} (\Phi^\perp)^{-1/2} + \frac{16}{Re_p} \Phi^{-1/2} + \frac{3}{Re_p^{1/2}} \Phi^{-3/4} + \frac{0.42}{\Phi^\perp} 10^{0.4[-\log(\Phi)]^{0.2}} \quad (26)$$

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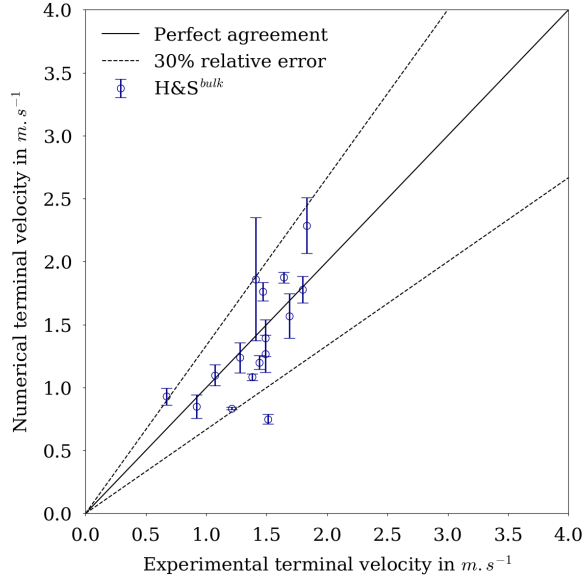


(a) Oblate spheroid reconstruction.

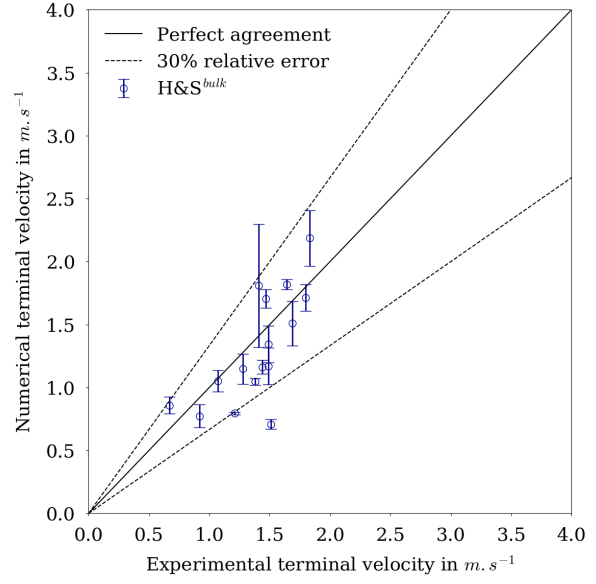


(b) Prolate spheroid reconstruction.

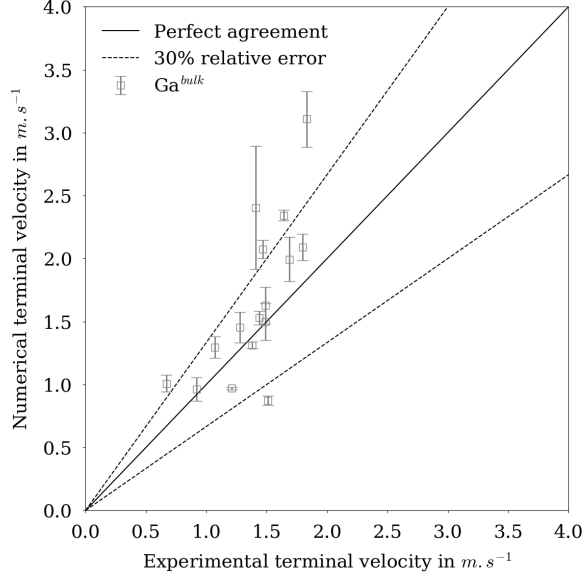
Fig. 9 Average numerical velocity ($\bar{v}_{p_\infty}^{num}$) vs. experimental velocity ($v_{p_\infty}^{exp}$). Ice-bulk density-based reconstructions (H&L^{bulk}, Ga^{bulk} and H&S^{bulk}) vs. H&W model (reference).



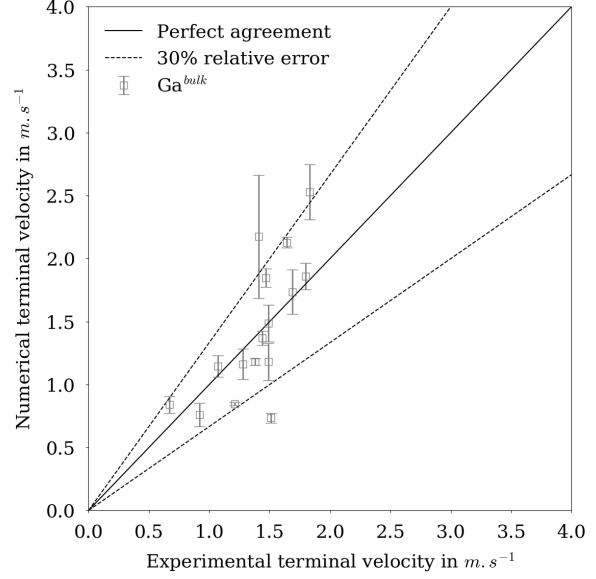
(a) $\text{H\&S}^{\text{bulk}}$, prolate.



(b) $\text{H\&S}^{\text{bulk}}$, oblate.



(c) Ga^{bulk} , prolate.



(d) Ga^{bulk} , oblate.

Fig. 10 $v_{p_\infty}^{\text{num}}$ vs. $v_{p_\infty}^{\text{exp}}$. Dispersion of $v_{p_\infty}^{\text{num}}$ around $\bar{v}_{p_\infty}^{\text{num}}$ and influence of the chosen view for the reconstructed geometry. $\text{H\&S}^{\text{bulk}}$ and Ga^{bulk} models.

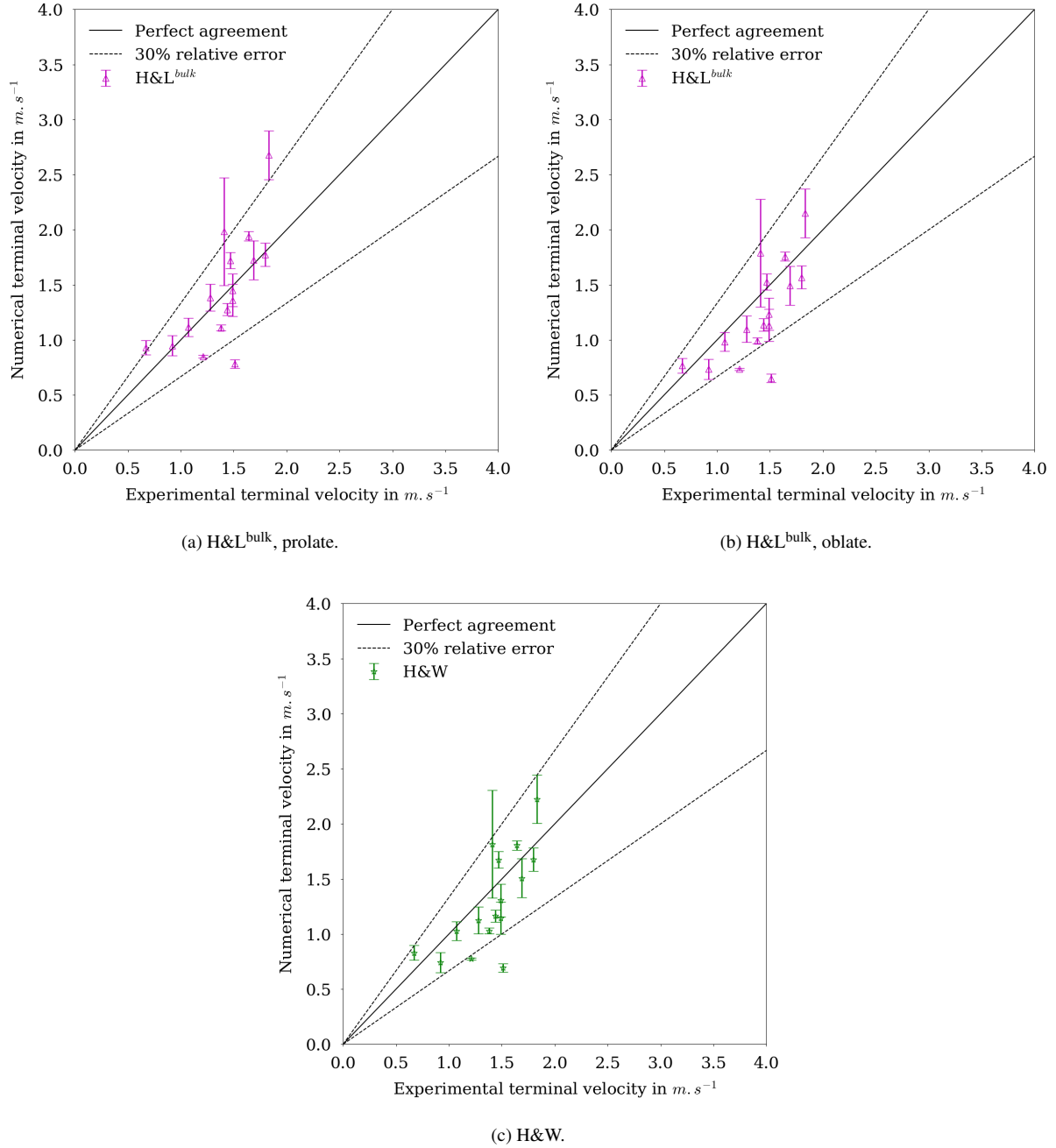


Fig. 11 $v_{p_\infty}^{num}$ vs. $v_{p_\infty}^{exp}$. Dispersion of $v_{p_\infty}^{num}$ around $\bar{v}_{p_\infty}^{num}$ and influence of the chosen view for the reconstructed geometry. H&L^{bulk} and H&W models.

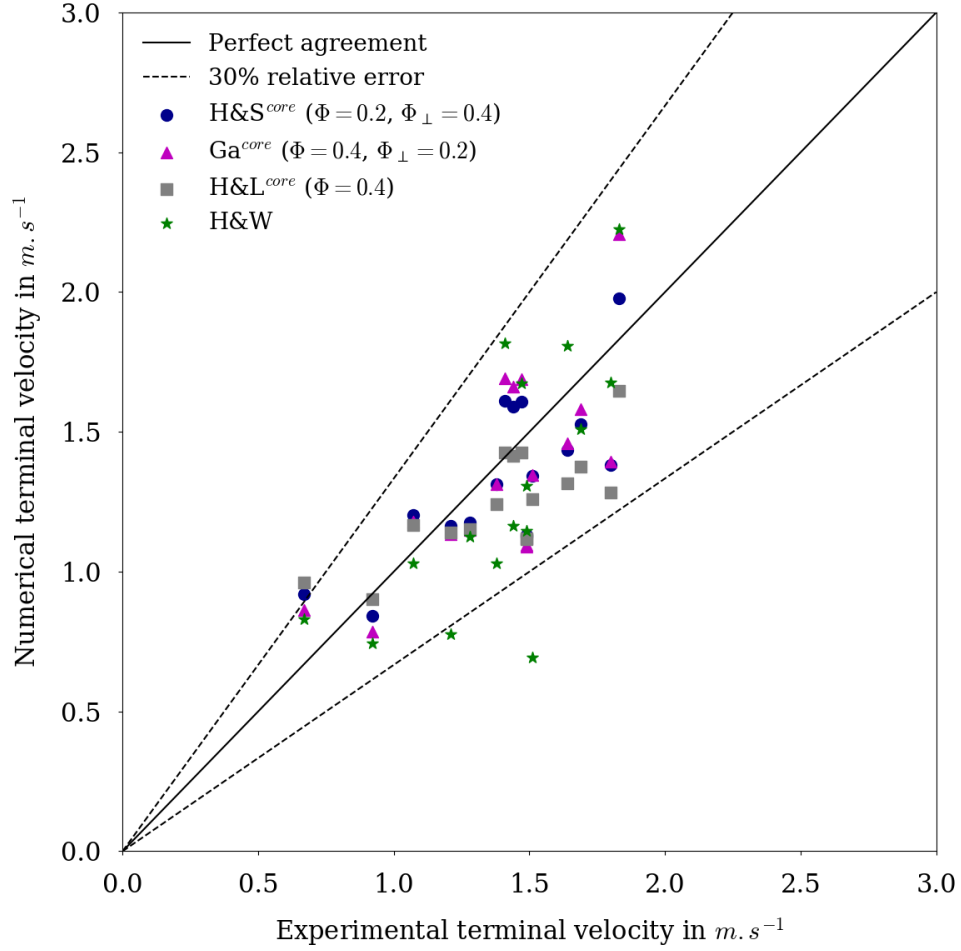
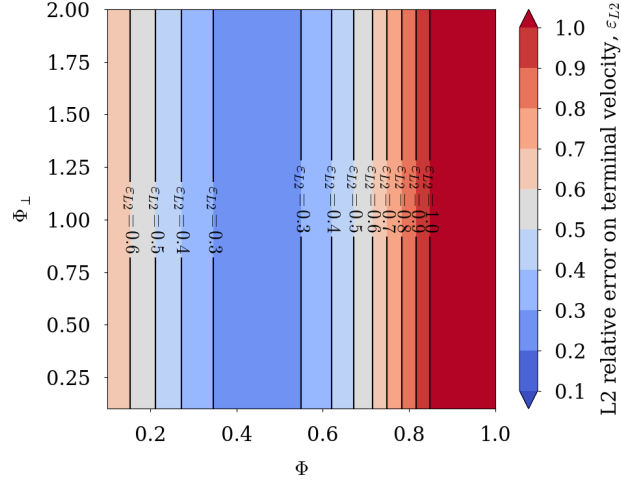
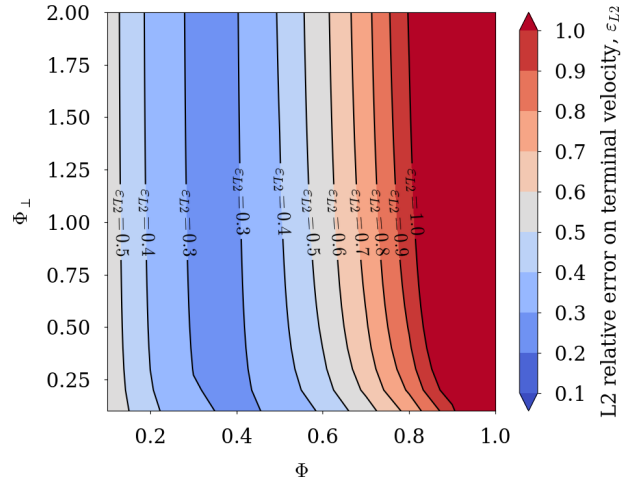


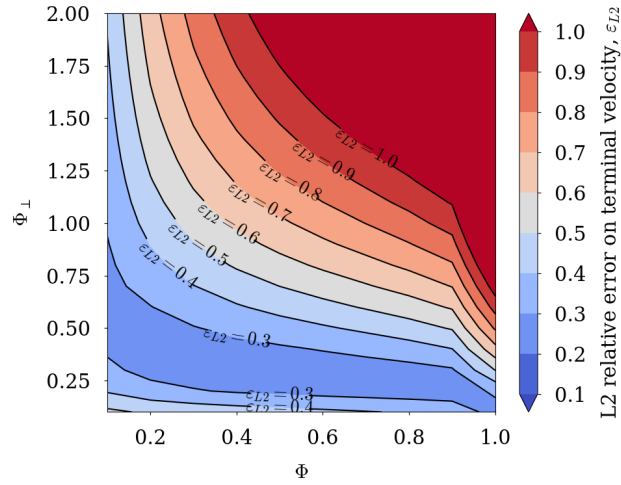
Fig. 12 Numerical velocity ($v_{p\infty}^{num}$) vs. experimental velocity ($v_{p\infty}^{exp}$). Ice-core density-based reconstructions (H&L^{core}, Ga^{core} and H&S^{core}) vs. H&W model (reference).



(a) $H\&L^{\text{core}}$



(b) Ga^{core}



(c) $H\&S^{\text{core}}$

Fig. 13 L_2 -error between $v_{p_\infty}^{\text{num}}$ and $v_{p_\infty}^{\text{exp}}$ (all snowflakes). Influence of Φ and Φ^\perp on the ice-core density-based description.

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