

1 **SubZero: A Sea Ice Model with an Explicit**
2 **Representation of the Floe Life Cycle**

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5 **Key Points:**

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Abstract

Sea ice dynamics exhibit granular behavior as individual floes and fracture networks become particularly evident at length scales $O(10\text{--}100)$ km and smaller. However, climate models do not resolve floes and represent sea ice as a continuum, while existing floe-scale sea ice models tend to oversimplify floes using discrete elements of predefined simple shapes. The idealized nature of climate and discrete element sea ice models presents a challenge of comparing the model output with floe-scale sea ice observations. Here we present SubZero, a conceptually new sea ice model geared to explicitly simulate the life cycles of individual floes by using complex discrete elements with time-evolving shapes. This unique model uses parameterizations of floe-scale processes, such as collisions, fractures, ridging, and welding, to simulate a wide range of evolving floe shapes and sizes. We demonstrate the novel capabilities of the SubZero model in idealized experiments, including uniaxial compression, the summer-time sea ice flow through the Nares Strait, and winter-time sea ice growth. The model naturally reproduces the statistical behavior of the observed sea ice, such as the power-law appearance of the floe size distribution and the long-tailed ice thickness distribution. The SubZero model could provide a valuable alternative to existing discrete element and continuous sea ice models for simulations of floe interactions.

Plain Language Summary

Sea ice is an inherent part of our climate system that responds rapidly to climate change. It is commonly conceptualized as a collection of many strongly interacting floes (sea ice fragments). However, climate models treat sea ice as a continuum, as resolving the complexity of floe-scale mechanical and thermodynamical processes is challenging. Here we present a conceptually new sea ice model that can explicitly simulate the life cycle of individual sea ice floes, including collisions, fractures, ridging and rafting, welding, and growth. We demonstrate the novel capabilities of SubZero in idealized experiments, including simulations of summer-time sea ice flow through a narrow strait and winter-time sea ice growth. Both experiments were successful in reproducing the statistical behavior of the observed sea ice, specifically the distribution of floe sizes and thicknesses. The unique SubZero capabilities may improve the realism of sea ice modeling.

1 Introduction

Sea ice motion at relatively large scales, $O(100)$ km, is commonly represented in climate models (Keen et al., 2021) using continuous rheological models (Hibler, 1979; Coon, 1980). However, at relatively small scales, $O(10\text{--}100)$ km and smaller, sea ice can be viewed as a granular material consisting of a collection of interacting floes (Rothrock & Thorndike, 1984; Toyota et al., 2006; Perovich & Jones, 2014; Zhang et al., 2015; Roach et al., 2018; Stern et al., 2018). The discrete floe dynamics are particularly pertinent in marginal ice zones where interacting floes are distinctly observed in satellite images, and sea ice resembles granular material (Figure 1). In consolidated pack ice, floes can be frozen to each other (welded) but externally forced large-scale sea ice motion can occur due to frequent anisotropic fractures and deformation (Hibler III & Schulson, 2000; Hutchings et al., 2011). Since specific floe configurations, their mechanical properties, and existing fracture networks are expected to affect the short-term evolution of sea ice, explicitly representing these features in some models is desirable.

Although it is technically possible to run continuum sea ice models at very high resolutions that approach floe scales, the model equations are formally applicable under the assumption that the grid box size is significantly larger than the characteristic floe size. Under this assumption, the floe interactions can be represented statistically (Hibler III, 1977; Feltham, 2008). Nonetheless, high-resolution numerical simulations can generate discontinuities that resemble observed linear kinematic features (Hutter & Losch, 2020;

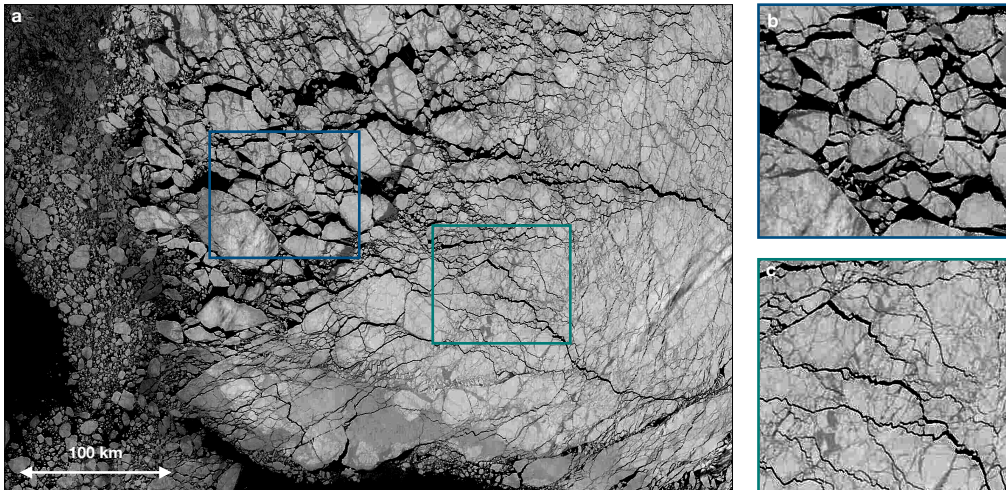


Figure 1. Example of the summertime sea ice in the Beaufort Sea, near Banks Island demonstrating its granular discontinuous nature and spatial heterogeneity. **(a)** A filtered reflectance image from the NASA WorldView website encompassing a region about 550 by 350 km in size bounded by 71–76°N in latitude and 126–137°W in longitude, taken on May 17th, 2021. The image filtering included making it gray scale and adjusting the level curves to highlight the fracture network and individual floes. **(b,c)** Zoomed-in view of the rectangular regions about 100 by 100 km in size as denoted in **(a)**.

60 Mohammadi-Aragh et al., 2020; Mehlmann et al., 2021; Hutter et al., 2022). But despite
 61 the major progress of continuous modeling of large-scale sea ice and the ongoing devel-
 62 opments in pushing their applicability limits by increasing the resolution, the rheolog-
 63 ical models are not meant to represent the scales of motion at which individual floes start
 64 to affect dynamics (Coon et al., 2007). Essentially, continuous models are not designed
 65 to generate the highly fragmented sea ice as shown in Figure 1. Hence, the validation
 66 against floe-scale observations for continuous models is only possible using statistical char-
 67 acteristics or large-scale sea ice motion because the rheological parameters parameter-
 68 ize the cumulative effects of floe interactions. Consequently, direct comparisons of con-
 69 tinuous models to remote sensing or field observations of individual floe behavior are chal-
 70 lenging, even considering that sea ice motion is inherently stochastic (Lemke et al., 1980;
 71 Percival et al., 2008; Rampal et al., 2009).

72 Alternatives to continuous rheology models are Discrete Element Models (DEMs),
 73 developed initially in the context of granular assemblies and rock dynamics (Cundall &
 74 Strack, 1979; Potyondy & Cundall, 2004). DEMs represent media as a collection of a large
 75 number of colliding bonded elements of specified shapes and contact laws and hence are
 76 typically computationally demanding. Since the continuous equations of motion are of-
 77 ten unknown, DEMs resort to specifying the interaction laws between its elements and
 78 strive to calibrate them using macro-scale observations or laboratory experiments (Grima
 79 & Wypych, 2011). Another way of simulating fluid motion with known rheology is the
 80 Smoothed Particle Hydrodynamics approach that also simulates particle motion but the
 81 laws of their interaction are derived from the continuous fluid rheology (Monaghan, 1992;
 82 Gutfraind & Savage, 1997; Lindsay & Stern, 2004; Marquis et al., 2022). As such, DEMs
 83 present a more general class of models that could simulate media for which correspond-
 84 ing macro-scale rheology might not exist, provided that the interaction laws between its
 85 particles could be constrained from observations.

86 With increasing computational capabilities and the emergence of comprehensive
 87 field and remote sensing observations at the floe-scale, the DEM approach (Cundall &
 88 Strack, 1979) has been adapted for modeling discontinuous sea ice dynamics (Hopkins
 89 et al., 2004; Wilchinsky et al., 2010; Herman, 2013, 2016; Kulchitsky et al., 2017; Dams-
 90 gaard et al., 2018; Liu & Ji, 2018; Tuhkuri & Polojärvi, 2018; West et al., 2021). At en-
 91 gineering scales, below about $O(10-100)$ m, sea ice DEMs have implemented a bonded
 92 particle model (Liu & Ji, 2018; Tuhkuri & Polojärvi, 2018). At these scales, the mod-
 93 els could be cross-validated with laboratory experiments, specialized field observations,
 94 and measurements of stress from structure-ice interactions, including ships. Sea ice DEMs
 95 have also been used for exploring idealized processes, including jamming and ice bridge
 96 formation in straits (Damsgaard et al., 2018) and wave-floe interactions (Herman et al.,
 97 2019). At larger regional scales, up to a few 100 km, the CRREL model (Hopkins et al.,
 98 2004; Wilchinsky et al., 2010) and its recent modification that utilizes level sets to com-
 99 pute collisions (Kawamoto et al., 2016) has been adapted for regional simulations of Nares
 100 Strait (West et al., 2021). The Siku model (Kulchitsky et al., 2017) is capable of sim-
 101 ulating the formation of basin-scale linear kinematic features in the Beaufort Gyre as-
 102 sociated with the coastal features. DEMs are computationally demanding due to requir-
 103 ing a large number of particles and small computational time steps. As such, their use
 104 in coupled Earth system models is challenging but can be done. One example of a pro-
 105 totype large-scale sea ice model within (global) Earth system models currently under de-
 106 velopment is DEMSI (Turner et al., 2022).

107 Existing sea ice DEMs [see Tuhkuri and Polojärvi (2018) for a review] follow a con-
 108 ventional approach of using simple pre-defined shapes for the elements, e.g., points or
 109 disks (Herman, 2013; Damsgaard et al., 2018; Chen et al., 2021), polygons (Kulchitsky
 110 et al., 2017) or tetrahedra (Liu & Ji, 2018). However, observations demonstrate that floes
 111 range dramatically in shapes and sizes (Figure 1) and evolve in time subject to a vari-
 112 ety of processes like fractures, rafting and ridging, lateral growth/melt, welding, etc. Hence,
 113 using pre-defined element shapes brings some ambiguity about what elements and bonds
 114 between them physically represent. Are elements supposed to approximate the behav-
 115 ior of aggregates of floes (similar to what continuous rheological models are assuming),
 116 or perhaps they are representing bonded constituents of floes or some other metric of a
 117 sea ice state? Without a robust understanding of what a DEM element represents, it is
 118 difficult to search for direct correspondence between the state variables of the DEM and
 119 the observed sea ice. These are challenging questions, and the answers depend on the
 120 modeling philosophy because sea ice is a multi-scale media where grains are not well de-
 121 fined.

122 This manuscript presents a prototype for a conceptually new discrete element ap-
 123 proach to sea ice modeling that relies fundamentally on using elements with evolving bound-
 124 aries to more realistically represent the floe life cycle by modeling the creation, growth/melt,
 125 welding, and break-up of individual pieces of sea ice. Our goal is to develop a model that
 126 could be used in conjunction with floe-scale satellite and in situ observations for floe-scale
 127 sea ice predictions and process studies. While the ice floe model consists of several me-
 128 chanical and thermodynamic components, our ultimate focus is on developing a set of
 129 floe interaction rules that could lead to realistic sea ice mechanics, including distribu-
 130 tions of floe sizes, thicknesses, and shapes. In contrast with existing sea ice DEMs that
 131 use prescribed simple shapes of elements (like disks), our approach is based on more re-
 132 alistic floes conceptualized as complex-in-shape time-evolving elements instead of spec-
 133 ifying a large number of stiffly-bonded simple elements to represent floes. We argue that
 134 the model capability of developing floe shapes naturally, due to specific physical processes
 135 at play, might bring us closer to direct model validation with floe-scale observations. The
 136 numerical implementation of our proposed method is publicly available as the SubZero
 137 sea ice model on GitHub (Manucharyan & Montemuro, 2022), and its releases are pub-
 138 lished on Zenodo (Montemuro & Manucharyan, 2022). Below, we provide the model for-
 139 mulation and present a few idealized simulations to showcase the novel capabilities.

Table 1. A list of prognostic and diagnostic variables that define the state of a floe in the SubZero model.

Floe variable	Description
area	Floe area
h	Floe thickness
mass	Floe mass
c_alpha	Rotated floe vertices relative to geometric center of area
c_0	Unrotated floe vertices relative to geometric center of area
inertia_moment	Floe moment of inertia
angles	Interior angles of floe corresponding to vertices of c_0/c_alpha
rmax	Maximum distance from geometric center of area to a floe vertex
StressH	History of instantaneous stress tensors on a floe
Stress	Average of instantaneous stress tensors on a floe
FxOA, FyOA	X & Y component of forces per unit area from Ocean and Atmosphere
torqueOA	Torque per unit area from Ocean and Atmosphere
X, Y	X & Y location of Monte-Carlo points in unrotated plane
A	Logical matrix saying if location [X,Y] is inside floe shape
alive	Logical value describing if floe is alive or will be discarded
X_i, Y_i	X & Y location of floe geometric center of area
alpha_i	Rotation value of floe from original unrotated position
U_i, V_i	Velocity of centroid of floe in X & Y direction
ksi_ice	Angular velocity of floe
d{X_i, Y_i, ...}_p	Time rate of change of X_i, Y_i, ... previous time step
interactions	List of interactions with other floes
potentialInteractions	List of all potential interactions with other floes
collision_force	Summation of all forces from interactions with other floes
collision_torque	Summation of all torques from interactions with other floes
OverlapArea	Summation of all overlapping area with other floes

140 2 SubZero model philosophy

141 In contrast with existing sea ice DEM methods, our sea ice DEM simulates the mo-
142 tion of elements that change their shapes, much like the observed sea ice floes do dur-
143 ing interactions with other floes or boundaries. SubZero keeps a data structure track-
144 ing a set of necessary state variables for each individual floe. The complete list of state
145 variables is included in Table 1. Crucially, the ability of model elements to change shape
146 is not simply an additional improvement over existing DEMs that use fixed element shapes
147 but something that leads to fundamentally different dynamics of floe interactions. Specif-
148 ically, closely-packed concave elements in our model can lead to interlocking behavior:
149 floes appearing like rigid puzzle pieces cannot substantially move relative to each other
150 except when they are allowed to fracture. For such interlocked floes, the relative motion
151 can only occur if floes undergo area-reducing processes such as deformations induced by
152 micro- and macro-scale fractures (for example, ridging/rafting). Consequently, bonds be-
153 tween the interlocked elements are not entirely necessary as their role is partially trans-
154 ferred towards parameterizations of floe fractures and other processes that change the
155 shape of individual floes. We hypothesize that a DEM formulation based on floe shape
156 evolution would make the model comparisons with observed floes less ambiguous.

157 The increased complexity of floe interaction physics is the trade-off for using ele-
158 ments with freely evolving shapes. Floes undergo many processes that affect their shapes,
159 including fracturing, ridging/rafting, and welding, making them concave. In addition,
160 the fracture process, which is essential to the model dynamics, rapidly increases the num-

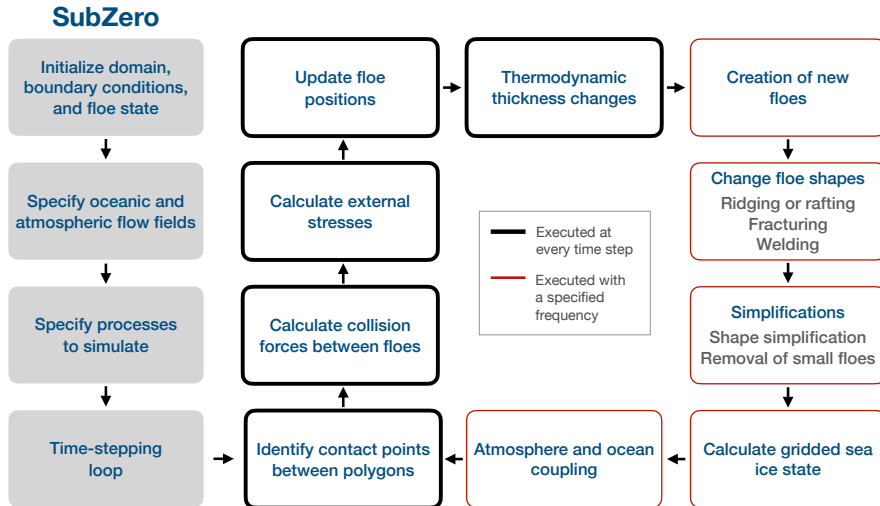


Figure 2. Operational flow chart for the SubZero sea ice model. The shaded gray boxes represent the different sections of the program, the red outlined boxes are processes that are executed every at specified intervals, and the black outlined boxes are processes that occur at every time-step.

161 number of floes. To avoid an explosion of the number of floes in a model, it is necessary to
 162 model only sufficiently large floes and treat sufficiently small floes as unresolved. This
 163 means that we remove any floe with an area below a designated minimum floe size from
 164 the model and put this mass into a separate array to track. Conventional DEMs can also
 165 generalize floes as a set of fixed-shaped elements that are bonded together, but the dif-
 166 ference with our SubZero model is that by representing the complex floe shapes by their
 167 polygonal boundaries, bonds are not needed to simulate the interactions of elements cover-
 168 ing its surface area. In other words, the trade-off in representing floes is between us-
 169 ing a large number of simple fixed-shape elements with simple interaction rules versus
 170 representing it with a single complex-shaped polygon and complex physics describing its
 171 shape changes upon interactions with other floes. While using concave shape-changing
 172 floes as elements in a sea-ice DEM may lead to improved realism of simulations, it also
 173 creates new challenges in numerical integration and parameterizations of floe-scale physics
 174 that we address below.

175 **3 Dynamical core of the SubZero model**

176 Below we describe the essential components of the model, providing a relatively basic
 177 representation of crucial sea ice processes acting at the floe scale (see Figure 2 for the
 178 simulation workflow). Our modeling philosophy envisions iterative improvements of its
 179 components upon input from a broad sea ice research community as the model is used
 180 in conjunction with observational, experimental, modeling, and theoretical studies.

181 **3.1 Floes as polygons with changing boundaries**

182 Motivated by observations of sea ice fracture networks and floe boundaries that ap-
 183 pear piece-wise linear (Figure 1), we choose to use the polygonal representation of floes.
 184 The model homogenizes sea ice properties, such as the thickness within the floe, such

185 that its polygonal shape defines the center of mass, total volume, and moment of iner-
 186 tia. The floes (i.e., their vertex coordinates) are translated following the velocity and an-
 187 gular velocity of the floe, which are calculated using the momentum and angular momen-
 188 tum equations written for individual floes (Section 3.3). The model has the capability
 189 of splitting floes into rigidly connected sub-floes to keep track of floes that were ridged
 190 and/or welded together, with each sub-floe carrying its own properties, like thickness.
 191 However, this configuration is computationally demanding, and so we expect it to be used
 192 only when high-resolution information about intra-floe variability and floe fractures is
 193 needed. The basic version of the model does not keep track of the sub-floes and homog-
 194 enizes floe characteristics after processes like welding.

195 While convex element shapes lead to dramatic simplifications in calculations of the
 196 collision forces, our model allows for concave floes for better realism. Such crucial pro-
 197 cesses as floe fractures, welding, and ridging are in no way restricted to preserving the
 198 convex nature of the floes. In addition, creating new floes in complex empty areas be-
 199 tween existing floes becomes a much simpler task when concave floes are used, allowing
 200 an arbitrarily-high concentration to be achieved without substantially modifying the floe-
 201 size distribution of existing floes. While the SubZero model can be reduced to a conven-
 202 tional DEM by using fixed-shape convex elements, its ability to simulate complex time-
 203 evolving floe shapes provides much more flexibility to enhance the realism of the model
 204 output.

205 In comparison with conventional sea ice DEMs, bonds between floes in the Sub-
 206 Zero model play a less critical role, especially in highly-packed winter simulations, as some
 207 of their functionality is transferred towards parameterizations of floe fractures and ridg-
 208 ing/rafting. In winter-time simulations, wherein our model element shapes are allowed
 209 to evolve in time, the model state is composed of highly complex and packed floes in-
 210 terlocked with each other. The interlocking behavior of complex-shaped polygons ensures
 211 that they are essentially bonded without having any explicitly prescribed bonds between
 212 them. The only plausible way to have relative motion in this system is to generate a set
 213 of fracture/ridging/rafting events that could split a sufficient amount of floes from each
 214 other, creating some open area to allow motion. As a result, having bonds between floes
 215 is not entirely necessary, as their role is transferred to such parameterizations as frac-
 216 tures/ridging/rafting that change the shapes of floes and reduce the sea ice area. Nonethe-
 217 less, the bonds are necessary for more complex configurations of our model that can re-
 218 solve dynamics within individual floes by splitting them into bonded sub-floes. Such con-
 219 figurations bring more detail to resolving the stress/strain within the floes, which may
 220 be relevant for predicting processes like fractures occurring at a subset of floes in the lo-
 221 cation of interest, like field camps or ship/submarine paths.

222 **3.2 Creation of new floes algorithm**

223 Two primary scenarios call for the creation of new sea ice floes (also referred to here
 224 as ‘packing’). First, at the beginning of a run, it is necessary to define the initial state
 225 of the floes corresponding to a designated sea ice concentration. Second, new floes will
 226 be created to fill the open space around existing floes if required by the thermodynamic
 227 criteria. New floes are created by the packing algorithm that requires specifying a tar-
 228 get concentration for the entire domain (see an example in Figure 3) or by inputting a
 229 2D matrix that specifies the desired spatially-varying concentration on a specified Eu-
 230 lerian grid.

231 The packing algorithm is designed in the following way. First, it identifies the space
 232 unoccupied by existing floes using polygonal operations like unions and differences. Then,
 233 the identified region is broken up into polygons using the Voronoi tessellation (Boots et
 234 al., 2009) and ensuring that each added new floe is not overlapped with existing ones by
 235 cutting the overlap region if it exists. The Voronoi tessellation partitions the domain into

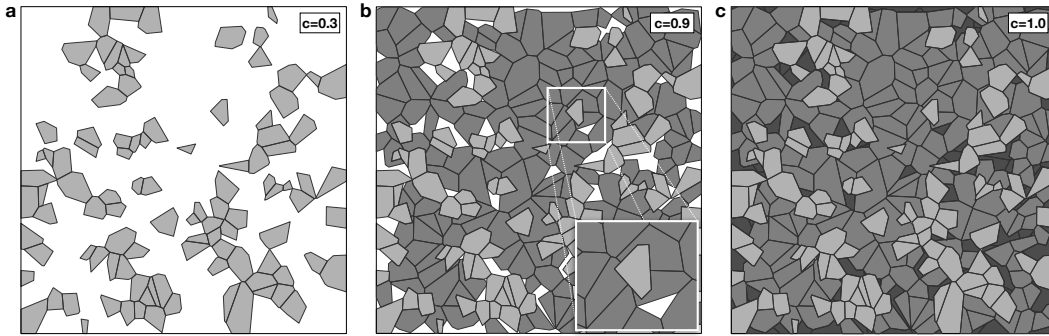


Figure 3. The model’s initial state was achieved using the floe packing algorithm that incrementally increases the number of floes to match the desired mean sea ice concentration. Panels (a-c) correspond to the initial state at 30% sea ice coverage, floes are added to reach 90% coverage, and then later, more floes are added to reach 100% sea ice coverage. White indicates open ocean, while the newer ice is thinner and is shaded as a darker color. Note that all floes are non-overlapping, and new floes are created in open areas without affecting the old floes. This creates concave floes that are interlocked with other floes, an example of which is shown in the panel b inset.

236 non-overlapping regions (cells) using a set of random seeds (points on a two-dimensional
 237 plane). These Voronoi cells become new floes, and are added until a targeted concen-
 238 tration is reached (Figure 3b and Figure 3c). Control on the characteristic sizes of the
 239 floes exists by prescribing the number of points used for Voronoi tessellation. The new
 240 floes are then added to the floe structure that carries all floe parameters. The initialized
 241 floe velocities match the ocean velocity. However, the new floe velocity could also be set
 242 to zero in most circumstances as the floe velocity has a relatively short adjustment timescale
 243 to the external forcing. The packing algorithm is time-consuming and hence is not used
 244 at every time step but with a specified frequency. The thickness of newly created floes
 245 follows Stefan’s law for ice growth (Leppäranta, 1993) and is related to the time separa-
 246 tion between packing events and on the heat fluxes received by the sea ice (Cox & Weeks,
 247 1988):

$$h_0 = \sqrt{\frac{2\kappa\Delta t N_{pack}(T_o - T_a)}{\rho_{ice}L}}, \quad (1)$$

248 where κ is the thermal conductivity of the surface ice layer, Δt is the time step during
 249 a model run, N_{pack} is the time steps between floe creation events, L is the latent heat
 250 of freezing and has units of Joules per kilogram, T_a is the temperature at the ice/air in-
 251 terface, and T_o and is the temperature at the ice/ocean interface. The values used in Sub-
 252 Zero are provided in Table 2 and Table 3. We note that snow has a different conductiv-
 253 ity from sea ice, but it isn’t present in the current version of SubZero. Nonetheless, adding
 254 a snow model to SubZero would be straightforward, and we envision doing so in the fu-
 255 ture, at a stage of implementing two-way coupling with an atmospheric and oceanic model.

256 3.3 Floe momentum and angular momentum evolution

257 Each floe in the model is treated as a rigid body with its center of mass \mathbf{X}_i accel-
 258 erating due to internal and external forces. At the same time, its angular velocity Ω_i re-
 259 sponds to the torques:

$$\begin{aligned}
 m_i \ddot{\mathbf{X}}_i &= \iint_{A_i} (\tau_{\text{ocn}} + \tau_{\text{atm}}) dA + \sum_{j \neq i; k} \mathbf{F}_{ij}^k + \bar{\mathbf{f}}_i, \\
 I_i \dot{\Omega}_i &= \iint_{A_i} (\mathbf{r} - \mathbf{X}_i) \times (\tau_{\text{ocn}} + \tau_{\text{atm}}) dA + \sum_{j \neq i; k} (\mathbf{R}_{ij}^k - \mathbf{X}_i) \times \mathbf{F}_{ij}^k + \bar{g}_i.
 \end{aligned} \tag{2}$$

Here, indices i and j denote different floes and k enumerates their contact points as several of those could exist for concave floes; m_i, I_i, A_i are floe mass, moment of inertia (Marin, 1984), and area; \mathbf{r} is the location of all points within a floe being integrated over; τ_{ocn} and τ_{atm} represent kinematic stresses from ocean and atmosphere; \mathbf{F}_{ij}^k and \mathbf{R}_{ij}^k are the interaction forces and coordinates of the k^{th} contact point for colliding i^{th} and j^{th} floes (land is conveniently treated as a stationary floe); and $\bar{\mathbf{f}}_i, \bar{g}_i$ are average forces and torques due to interactions with small-scale floes, that are unresolved owing to the floe-size truncation. An Adams-Bashforth two-step method is used to time-step the model when calculating the trajectories of each floe. A constant time-step is currently used in this prototype version of the model, but an adaptive time-stepping algorithm will be implemented in future versions. A description of the kinematic stresses from the ocean and atmosphere calculations is provided in section 6. Other body forces, such as the Coriolis and sea surface tilt forces, can be turned on. In addition to shape-conserving interactions, the model includes a criterion for floe mergers (welding), ridging, as well as fractures leading to the creation of new smaller floes. Like continuous sea-ice models, the floe model is also limited in its effective resolution by imposing the minimum floe size threshold parameter to bound the total number of elements. This minimum floe size is a parameter that could vary depending on the type of sea ice in a given region, the physical problem under consideration, and available computing resources. For simplicity, and given the lack of developed parameterizations, the basic version of the model does not include the forces and torques from unresolved floes (so $\bar{\mathbf{f}}_i = \bar{g}_i = 0$), but later versions would include stochastic representations of the impact of unresolved floes on the dynamics of the resolved floes.

3.4 Contact forces between the floes

3.4.1 Detection of contact points

Each floe has a bounding circle associated with it, with the radius corresponding to the distance from its center of mass to the furthest vertex of its boundary (Figure 4). The bounding floe radii are then used to identify pairs of floes that could be potentially interacting. The polygons of the potentially interacting floes are copied into the memory for each of the floes to enable parallel computation of more complex polygonal operations to determine if the floes are actually overlapping and calculate the collision forces. Note that the floes are considered rigid (non-deformable) bodies, but we allow a small numerical overlap between the floes to exist in order to compute collision forces that depend on the geometry of the overlap area, as common with soft-body discrete element methods (Cundall & Strack, 1979; Luding, 2008; Radjai & Dubois, 2011). Collision forces, \mathbf{F}_{ij} , consist of elastic (normal) and frictional (tangential) components which correspondingly are directed along and perpendicular to the line of contact between the two floes.

3.4.2 The normal direction at a contact point

The desired capability of simulating collisions between complex-shaped floe translates into some ambiguity in defining the normal and tangential directions at the contact points, which isn't present for simple convex shapes like circles. For concave polygons, two issues need to be addressed. First, there can be multiple contact points between two concave floes (see an example in Figure 4), and the forces associated with each need to be resolved separately. Second, when sufficiently large forces are driving the floes, the overlap area in some contact points can be of very complex shape such that it isn't

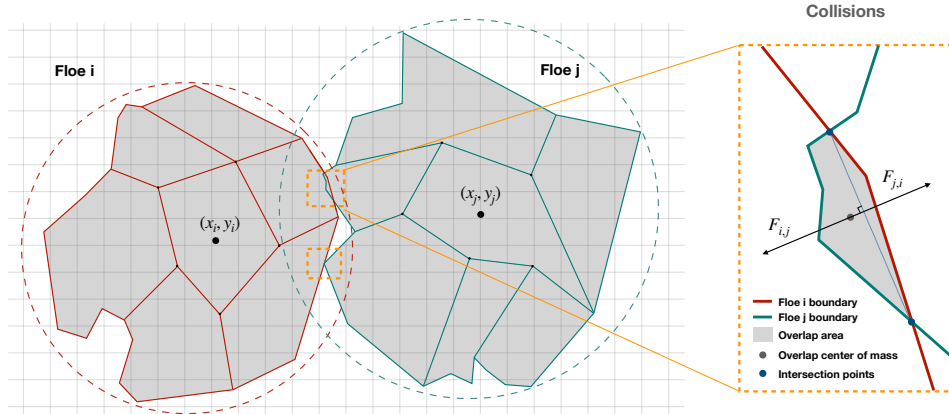


Figure 4. Example of two colliding floes outlining the corresponding normal collision forces appearing at the overlap areas. The bounding circles are shown for both floes and are used to determine if any two floes could be potentially colliding. Each floe could be composed of several rigidly-connected polygonal sub-floes if a more accurate floe-fracture model is needed. An example Eulerian grid that can be used for coupling with the oceanic and atmospheric model is shown with gray lines.

304 clear how to define the directions of the collision forces. Here, we define the normal direction
305 motivated by Feng et al. (2012). First, at each contact point, the floe polygons
306 intersect each other at two points, and we store the mid-point between them. Second,
307 we calculate the center-of-mass position of the overlap area. The normal force is defined
308 as pointing from the center of mass of the overlap area towards the mid-point of the poly-
309 gon intersections. Finally, a check is made to ensure the overlap area would be reduced
310 if the floes are displaced in the direction of the corresponding normal forces; the normal
311 direction is flipped if the check fails, which occurs in rare marginal cases with complex
312 shapes of the overlap areas.

313 3.4.3 Normal forces

314 For each pair of interacting floes, an algorithm is used to determine the geometry
315 of the overlapped area and the corresponding forces and torques. An energy-conserving
316 contact algorithm (Feng et al., 2012) is used. The floe collision rules are based on simple
317 physical laws for inelastic collisions of rigid bodies (Hopkins et al., 2004; Kulchitsky
318 et al., 2017; Wilchinsky et al., 2010). The normal forces depend on the relative location
319 of the floes, being proportional to the overlap area at each contact location and the pro-
320 portionality constant \mathcal{K} . For a given interaction force, increasing the parameter \mathcal{K} de-
321 creases the overlap area between the floes to the extent that they start to appear like
322 rigid bodies; however, this occurs at the expense of having to use a relatively small time
323 interval to accurately resolve the collision forces. The parameter \mathcal{K} could be taken to be
324 as large as possible depending on the computational capabilities and the desired accu-
325 racy of collisions. However, keeping it finite brings a physical meaning that the floes are
326 elastic and could have deformation expressed in the form of a finite overlap region be-
327 tween the flow and a general decrease of the overall area in the domain under compres-
328 sion. The equation for the normal force between the i^{th} and j^{th} floes at the k^{th} contact
329 point, $\mathbf{F}_{ij,n}^k$, is given as

$$\mathbf{F}_{ij,n}^k = \mathcal{K} \mathcal{A} \mathbf{n}_{ij}^k, \quad (3)$$

where \mathcal{A}^k is the overlap area and \mathbf{n}_{ij}^k is the normal direction at the k^{th} contact point between the i^{th} and j^{th} floes. Note that for concave elements, there could be multiple contact points, and hence k could be greater than one. The elastic component is similar to a simple linear spring so an effective stiffness \mathcal{K} can be found through the equation for springs in a series

$$\mathcal{K} = \frac{E_i h_i E_j h_j}{E_i h_i r_j + E_j h_j r_i}, \quad (4)$$

330 where E is an elasticity value, h is the thickness, and r quantifies the floe size. The floe
 331 size is defined to be r_i , which is the square root of the area of the i^{th} floe. Note that smaller
 332 floes have higher effective stiffness, requiring smaller time steps to resolve their collisions.
 333 The values used in SubZero are provided in Table 2 and Table 3. If there is an individ-
 334 ual floe interacting with a non-deformable boundary ($E_j \rightarrow \infty$) then the equation sim-
 335 plifies to

$$\mathcal{K} = \frac{E_i h_i}{r_i}. \quad (5)$$

3.4.4 Tangential forces

336
 337 Discrete element models with bonds commonly utilize force-displacement laws for
 338 viscous-frictional tangential forces (Cundall & Strack, 1979; Hopkins et al., 2004; Her-
 339 man, 2016; Damsgaard et al., 2018). For this model, which does not have bonds, the fric-
 340 tional tangential force is associated with the average tangential velocity difference be-
 341 tween the floes at the contact location (Chen et al., 2021). The basic frictional force model
 342 defines a coefficient of static friction and a smaller coefficient for kinetic friction, taking
 343 the force to be proportional to the normal force only.

344 When the floes are in motion, the adjustment for the frictional laws is proportional
 345 to the velocity difference between the two floes, the time step, and the chord length c_{ij}^k .
 346 The chord length is defined as the distance between the interaction points (Figure 4).
 347 The equation for the tangential force between the i^{th} and j^{th} floes at the k^{th} contact point,
 348 $\mathbf{F}_{ij,t}^k$, is given as

$$\mathbf{F}_{ij,t}^k = c_{ij}^k G v_{ij}^k (\Delta t) |\mathbf{F}_{ij,n}^k| \mathbf{t}_{ij}^k, \quad (6)$$

349 where G is the shear modulus, velocity v_{ij}^k gives the difference between the two floes, and
 350 \mathbf{t}_{ij}^k is the tangential direction at the k^{th} contact point between the i^{th} and j^{th} floes. The
 351 values used in SubZero are provided in Table 2 and Table 3. The velocity v_{ij}^k is given
 352 by

$$v_{ij}^k = [(\mathbf{v}_j + \omega_j \times \mathbf{r}_j^k) - (\mathbf{v}_i + \omega_i \times \mathbf{r}_i^k)] \cdot \mathbf{t}_{ij}^k, \quad (7)$$

353 where \mathbf{r}_{ij}^k is the position vector of that contact point from the center of mass of the i^{th}
 354 floe to the contact point at the k^{th} contact point; \mathbf{v}_i , ω_i are the linear and angular ve-
 355 locity of the i^{th} floe at center of mass. However, per friction laws the tangential force
 356 is limited by a proportionality to the coefficient of friction (μ) and magnitude of the nor-
 357 mal force (Cundall & Strack, 1979; Hopkins, 1996) such that

$$\mathbf{F}_{ij,t}^k \leq \mu |\mathbf{F}_{ij,n}^k| \mathbf{t}_{ij}^k. \quad (8)$$

358 The presence of tangential forces leads to energy dissipation upon collisions.

359 **3.5 Interactions with boundaries**

360 Coastal boundaries are naturally prescribed as stationary polygonal floes, and an
 361 arbitrary number of such boundaries (defined to be the value N_b in Table 2 and Table
 362 3) are possible if, for example, one is interested in simulating the sea ice in Fjords with
 363 many islands. The interaction forces with the coastal boundaries are calculated in a sim-
 364 ilar way as with other floes, but assuming that the elasticity of a boundary is infinite (i.e.
 365 all elastic deformation occurs within a floe). The frictional parameters with coastal bound-
 366 aries could be different, although they are kept the same by default. Periodic boundary
 367 conditions could be used in addition to coastal boundaries in channel-type configurations.
 368 Periodic (and double-periodic) boundary conditions are achieved by using ghost floes.
 369 The ghost floes are shifted copies of all floes that are close to one boundary and have the
 370 potential to overlap with the floes at the other boundary. The framework dealing with
 371 periodic boundary conditions is also directly applicable for parallel implementation as
 372 each processor could resolve its sub-domain in physical space and exchange the infor-
 373 mation about the location of ghost floes at its edges with neighboring processors. This
 374 capability will be implemented in future versions of the code, but in its current form, par-
 375 allel computing is utilized by cores within a single node with Matlab's "parfor" loops.

376 **4 Processes affecting floe shapes**

377 **4.1 Floe fractures**

378 *4.1.1 Defining the floe stress tensor*

379 Stress and strain rates are important for physical processes such as fracture and
 380 lead formation. The collision function keeps track of the location and forces associated
 381 with each collision. We treat the stress as being homogeneous across individual floes and
 382 calculate the stress tensor ($\underline{\underline{\sigma}}_i$) of individual floes (Rothenburg & Selvadurai, 1981; André
 383 et al., 2013) via

$$\underline{\underline{\sigma}}_i = \frac{1}{2\mathcal{V}_i} \sum_{j,k} f_{ij}^k \otimes r_{ij}^k + r_{ij}^k \otimes f_{ij}^k, \quad (9)$$

384 where \mathcal{V}_i is the volume of the i^{th} floe, f_{ij}^k is the force at the k^{th} contact point between
 385 the i^{th} and j^{th} floes and r_{ij}^k is the position vector of that contact point from the center
 386 of mass of the i^{th} floe to the contact point. The stress tensor is later used to define the
 387 floe fracture criteria. The continuous representation of the stress tensor over a coarse Eu-
 388 lerian grid (see section 6.2) is obtained by volume-weighted averaging of the stress ten-
 389 sors of the individual floes (Chang, 1988) within each grid box:

$$\underline{\underline{\sigma}}(x, y) = \frac{1}{\mathcal{V}_{tot}} \sum_i \underline{\underline{\sigma}}_i \mathcal{V}_i, \quad (10)$$

390 where the index i includes only floes with centers of mass located inside the coarse grid
 391 box at the location (x, y) and \mathcal{V}_{tot} is the total volume those floes excluding the floe ar-
 392 eas outside the grid box.

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4.1.2 Fracture criteria based on homogenized floe stress

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The homogenized stresses are used in the following way, depending on the configuration of model parameterizations. The main usage revolves around defining the appropriate criteria for fracturing individual floes based on local and/or non-local stress criteria. Specifically, it is straightforward to define fracture criteria based on, e.g., Mohr-Coulomb failure envelope (Figure 5) that is defined in the space of principal stresses of a floe stress tensor (Weiss & Schulson, 2009). The equation for the failure envelope boundaries are $\sigma_1 = q\sigma_2 + \sigma_c$ where $q = 5.2$ and $\sigma_c = 250\text{kPa}$. Here σ_1 is the associated maximum principal stress and σ_2 is the intermediate principal stress. Other options for floe-fracture criteria could be derived from yield curves that are used in continuous models (Hibler, 1979). The connection with the SubZero model, where floes are rigid (non-deformable) objects, is that the macro-scale strain rate appears when floes are fracturing (or ridging/rafting). Thus, satisfying criteria for individual floe fractures would lead to macro-scale sea ice motion, which in continuous formulations is described by the presence of a yield curve. For example, in viscous-plastic sea ice rheology, an elliptical yield curve is used with a strength parameter (P) where $P = P^*h$ that is proportional to sea ice thickness h for fully ice-covered regions (Hibler, 1979). The values of P^* is a fixed empirical constant and the value used in SubZero are provided in Table 2 and Table 3.

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The basic isotropic fracture mechanism is implemented based on the stress experienced by floes and fractures a floe into a number of smaller pieces (Figure 6) when the principal stress values satisfy the specified fracture criteria (Figure 5). When it is determined that a fracture should occur, a floe is split into the desired number of elements via Voronoi tessellation based on random x and y points coordinates (uniform distribution) acting as centers of the Voronoi cells. The mass, momentum, and angular momentum are conserved after the floe fractures into smaller pieces.

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The number of elements into which the floe splits can be determined via a probabilistic process based on the proximity of the floe stress to the boundaries of the failure criteria or simply preset at a fixed number (e.g., $N_{Pieces} = 3$) as we did in our idealized model configurations (Table 2, 3). The shattered pieces form new floes that could continue breaking until stresses are relieved. This is a simple procedure leading to an isotropic distribution of fractures regardless of the direction of the principal stresses. Note, without fracturing, the packed and interlocked floes would have no motion, and hence the movement occurs when the particle fracture criteria are satisfied. Therefore, one could draw connections between the concepts of the yield curve in continuum mechanics and the fracture criteria of the elements, but those would need to be constrained with floe-scale observations.

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The basic fracture criteria implemented in the model include the Mohr's cone and the elliptical yield curve used in viscous-plastic rheology (Figure 5). Any other breakage criteria could be easily implemented. For studies focusing on the analysis of linear kinematic features, it would be necessary to formulate more advanced floe fracture criteria or use bonds between floes to explicitly simulate fracture formation. This is an ongoing area of model development, and we envision enabling this capability in future versions of SubZero.

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4.1.3 Corner grinding

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Observations of older floe fields show a tendency to form rounder shapes through repeated interactions with other floes. The corner grinding process uses the contact overlap areas to determine whether a floe could have its corner fractured; the likelihood of this happening is proportional to $\text{OverlapArea}/\text{FloeArea}$. The model tracks the contact points during a collision with other floes, and if there is a contact point nearby, it is qualified to fracture. The properties of the new floes are calculated to satisfy mass, momentum, and angular momentum conservation laws. For a corner with interior angle α and

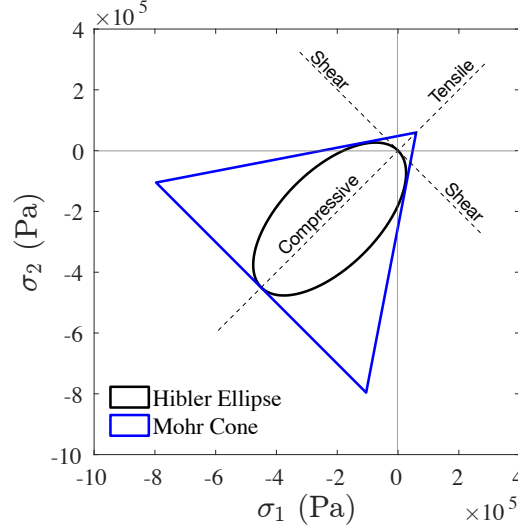


Figure 5. Examples of fracture criteria plotted as boundaries in the (σ_1, σ_2) principal stress space, including Mohr’s cone and Hibler’s ellipse. Floes for which homogenized stresses are large enough to reach (or temporarily exceed) the fracture criteria boundaries end up fracturing into several elements. Those boundaries could be interpreted as yield curves for individual floes because only upon reaching those boundaries can there be any motion within the floe by means of fracturing it into smaller pieces.

444 adjacent sides of length l_1 and l_2 , where $l = \min(l_1, l_2)$ (Figure 7a), at least one con-
 445 tact must be within the radius l of the corner. For each eligible corner of the polygon,
 446 a fracture probability is defined as $1-\alpha/\text{Anorm}$, where $\text{Anorm}=360-180/N$, and N is the
 447 total number of vertices. This way, the probability of fracture increases as α approaches
 448 0° . For all floe corners that fracture, a triangle is defined with the same angle α and ad-
 449 jacent edges five times smaller than l . Figure 7 shows a floe field going through the corner
 450 fracture process. It can be seen that some of the sharper corners are broken off from
 451 7a as the angles trend closer to that of a regular polygon. Figure 7b shows the rounded
 452 floes after many collisions, and the fractured pieces have been plotted with a dark gray
 453 color to distinguish them from the initial floes (colored with light gray).

454 **4.2 Welding**

455 It is common for two ice floes to weld together when the temperature dips below
 456 freezing over the winter in the arctic. We define welding as the freezing of neighboring
 457 ice floes to form a bigger consolidated floe (Figure 6). We model this process by using
 458 thermodynamic criteria to determine if two overlapping floes will weld together. When
 459 welding occurs, the properties of the newly created floe are determined by satisfying the
 460 mass, momentum, and angular momentum conservation laws. Our most straightforward
 461 parameterization defines the welding probability (P_{weld}^i) of a floe in contact with another
 462 floe as

$$P_{weld}^i = P_{F_{heat}^i} \frac{\delta A_{i,j}}{A^i} \tag{11}$$

463 where $\delta A_{i,j}$ is the overlap area between two floes, and the proportionality constant $P_{F_{heat}^i}$
 464 is non-zero only when the ice is freezing. Improvements to this simple process could spec-

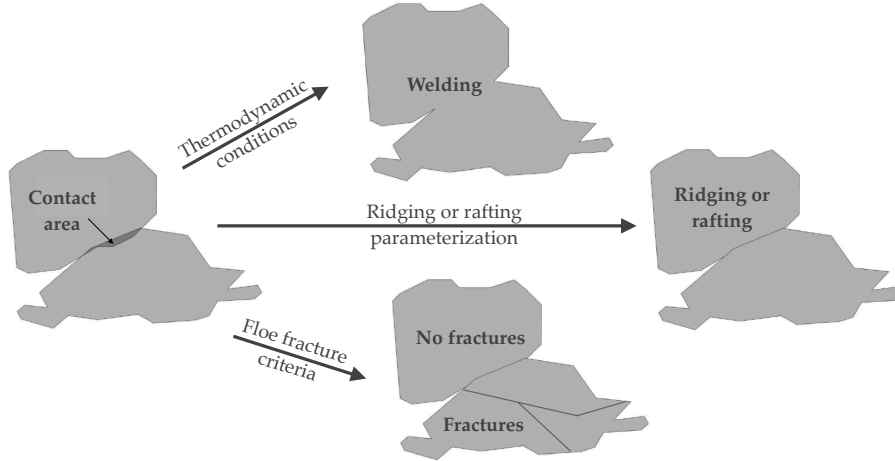


Figure 6. Example of two floes in contact leading to various possible outcomes, including welding, ridging/rafting, and fractures. The floe interaction forces are computed based on the geometry of the overlapping area. Collision forces define the homogenized floe stress tensor used in the fracture parameterization that splits the floe into several pieces. Welding occurs if a thermodynamic criterion is satisfied and leads to the merger of two floes into one. The ridging/rafting parameterization determines if the overlap area between the floes will be absorbed into increasing the thickness of one of the two floes in contact.

465 ify the probability to depend on the heat flux out of the ice floe or duration of the con-
 466 tact (Shen & Ackley, 1991).

467 4.3 Ridging and rafting

468 Upon contact with other floes, a sea ice floe can either become thicker or transfer
 469 some of its mass to another floe through the ridging process. For this model, we imple-
 470 mented a simple parameterization based on a critical thickness that is set to determine
 471 if ridging or rafting is possible for two floes in contact (Parmeter, 1975). Additionally,
 472 a probability for ridging (P_{ridge}) is defined so that only a subset of floes will undergo the
 473 ridging process. For the current version, it is set to a simple percentage value, and if at
 474 least one of the floes exceeds this threshold, then ridging will take place. However, more
 475 complex probabilities can depend upon compressive stress and thickness (Hibler III, 1980;
 476 Hopkins, 1998; Hopkins et al., 1999; Tuhkuri & Lensu, 2002; Damsgaard et al., 2021).
 477 When ridging occurs, the area of the floes is reduced as the mass is transferred toward
 478 increasing the thickness of one of the colliding floes. If both floes exceed the critical thick-
 479 ness ($h_c = 0.25$), a probability function (P_{Floe_i}) is set to determine the exchange of mass
 480 between the two floes, where the probability that the mass moves from floe i to floe j
 481 is

$$P_{Floe_i} = \frac{1}{1 + h_i/h_j}, \quad (12)$$

482 where h_i and h_j are the thicknesses of the two floes undergoing ridging. If only one floe
 483 exceeds the thickness, then the thin floe loses its mass to the thicker floe. Floe proper-
 484 ties are updated to ensure that the overall mass and momentum are conserved upon the
 485 adjustment of floe shapes (Figure 6). The ridging of sea ice can lead to complex sea ice

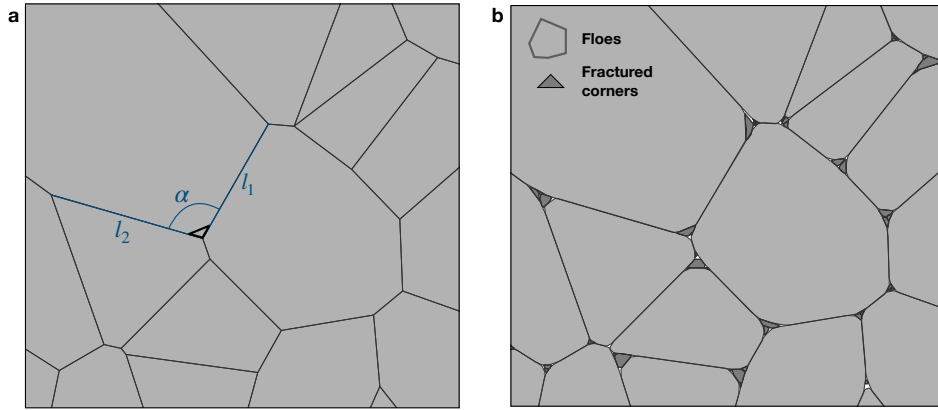


Figure 7. Example of floes where the sharp corners are breaking off upon tight contact with other floes. **(a)** The initial intact floe configuration with fully-packed interacting floes. Denoted are an interior angle, α , the lengths of adjacent edges, l_1 and l_2 . The black line denotes the corner that will be fractured (isosceles triangle with the same angle α). **(b)** The state of the floes after the occurrence of multiple corner fractures. Fractured corners are modeled the same as regular floes, but here they have been plotted with a dark gray color to distinguish from the initial floes that are colored with light gray.

486 shapes with a computationally prohibitive number of vertices. To reduce their complex-
 487 ity, we implement an algorithm that dynamically simplifies floe shapes (see Section 5.2).

488 When the two interacting floes are both below this critical thickness threshold, h_c ,
 489 they have a possibility of rafting where P_{raft} is a value set by the user. The numerical
 490 algorithm for the rafting process is similar to ridging, and mass will transfer from one
 491 floe to the other. After this rafting event, the floe that loses mass will also have its area
 492 updated. Floe properties are updated to ensure that mass, momentum, and angular mo-
 493 mentum are conserved throughout this operation. The updating of floe geometry is also
 494 similar to that shown in Figure 4.

495 4.4 Thermodynamic thickness changes

496 For existing floes a Semtner 0-layer approach is taken (Semtner Jr, 1976). The ba-
 497 sic version of the thermodynamic sea ice growth calculates the tendency of its thickness
 498 based on the net atmospheric and oceanic heat fluxes, and the tendency is inversely pro-
 499 portional to its thickness. This thickness growth assumes that the temperature inside
 500 the sea ice is always equilibrated to a linear profile, and the changing thickness is the only
 501 variable governing the heat flux. This basic version of the code is aimed at simulating
 502 sea ice mechanics, and hence the thermodynamic processes are simplified. Future ther-
 503 modynamic schemes will include the option of using multi-layer thermodynamics and
 504 include the treatment of snow cover. For small-scale floes (about 100 m and smaller),
 505 lateral growth and melting can be important, and this capability will also be implemented
 506 in future versions of the code.

507 In open-ocean regions where there are no ice floes, and freezing conditions are satis-
 508 fied such that the surface ocean temperature is maintained at the freezing point, the
 509 lost heat fluxes are partitioned into creating new floes with a prescribed minimum thick-
 510 ness. Thus, the total volume of new floes to be created in an open area, together with

511 the minimum floe thickness, defines the total area of the new floes that are then gener-
 512 ated using the packing algorithm.

513 5 Peculiarities of the numerical implementation

514 5.1 Tracking unresolved floes

515 Keeping track of all the small floes generated through the fracturing and ridging
 516 processes performed in the model becomes computationally expensive. This expense comes
 517 from both an increased particle count and shorter time steps associated with the higher
 518 elasticity in small floes. Thus, a lower limit is set, at which point any floe with a smaller
 519 area is removed from the simulation and kept track of in a separate variable. The mass
 520 of all unresolved floes is stored in a variable on a coarse Eulerian grid. Utilizing the Eu-
 521 lerian sea ice velocity (see section 6.2), the dissolved ice mass is advected around the do-
 522 main to preserve mass. Under proper thermodynamic conditions, this unresolved floe
 523 variable can act as a source for newly generated floes via section 3.2, conserving the mass
 524 of the system. In future versions of the model, parameterizations of the cumulative dy-
 525 namical impact of small-scale unresolved sea ice will be used in the calculation of forces
 526 and torques on the remaining floes.

527 5.2 Dynamic simplification of floe boundaries

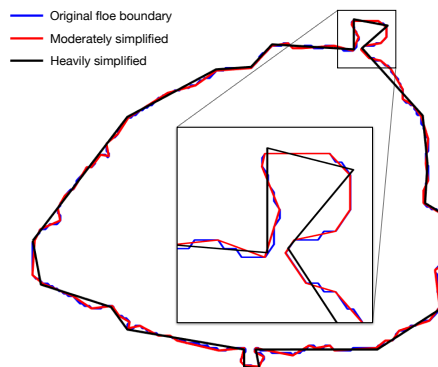


Figure 8. Example of a boundary simplification for a polygonal floe using the Douglas-Peucker algorithm. Initial floe boundary with 292 vertices (blue), its moderate simplification to 81 vertices (red), and heavy simplification to only 23 vertices (black). Inset shows a zoomed-in view of the protruding region at the top of the floe inside a black square box.

528 Repeated application of certain processes in the numerical implementation (such
 529 as ridging, welding, and floe creation) can lead to floes with a very large number of ver-
 530 tices, which is problematic for two reasons. First, running simulations with large num-
 531 bers of floes create excessively large data structures that need to be stored. Secondly,
 532 performing operations such as rotating, translating, or calculating overlaps with other
 533 floes becomes computationally cumbersome. To avoid this, we periodically check the num-
 534 ber of vertices and, when appropriate, apply a Douglas-Peucker simplification algorithm
 535 to reduce the complexity of the shape. The floes retain qualitatively similar shapes as
 536 shown in Figure 8. After its simplification, the thickness of the floe is updated to con-
 537 serve mass and momentum.

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5.3 Parallel for-loops for multi-core processors

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The SubZero program can run the collision algorithm, update floe trajectories, create new floe elements, weld floes, and fracture floes in parallel. To achieve this, we define for each given floe the potential interactions field that essentially copies all the necessary information about only those surrounding floes that have their bounding circles overlapping with a given floe. The potential interactions are found as described in section 3.4. The floe number, vertices, velocities, thickness, area, and centroid are all stored. This data is required to calculate the collisions between two floes and when two overlapping floes weld together independently of other rows in the floe structure. Updating floe trajectories and fracturing floes can be done in parallel and do not rely upon information from other floes in the structure. The creation of new elements and the welding algorithm divides the domain into smaller regions and bin the ice floes based on location. These subregions are then run in parallel.

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6 Coupling with ocean and atmosphere models on the Eulerian grid

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6.1 Atmosphere and ocean forcing of individual floes

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The atmospheric and oceanic equations of motion could be solved either within the Eulerian or Lagrangian frameworks, although typical climate models are Eulerian. We hence provide the coupling capability with the floe model based upon the gridded (Eulerian) representation of sea ice variables. For calculating the oceanic and atmospheric forces and torques acting on individual floes, a Monte-Carlo method (Cafisch, 1998) is used for the integration of stresses over the surface areas of the floes. The Monte-Carlo integration method uses random sampling of the desired function to numerically estimate the integral. The integral of the desired function is approximated by averaging samples of the function at random points over the surface, while typical algorithms evaluate the integral on a regularly spaced grid. For this model, random points in space are assigned, and ocean and atmosphere flows are interpolated onto these points, after which stresses are computed. Less than about 100 points are needed for an accurate estimation of stresses, resulting in about 5% accuracy (Oberle, 2015). The surface stresses as well as salt and heat fluxes that the ocean model receives from the sea ice model are computed by taking averages of the floe stresses and growth/melt rates over an Eulerian grid of the ocean model. This achieves a two-way coupling of both dynamic and thermodynamic components of the ocean and ice models. The same coupling can be arranged with the atmospheric model, and this capability would be implemented in the code as part of future developments.

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6.2 Mapping the state of the floe model to the Eulerian grid

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A coarse Eulerian grid is designated for the domain to diagnose the macroscale motion of the sea ice and couple it with Eulerian oceanic and atmospheric models. The domain is divided into smaller regions that align with this coarse spatial grid shown by the black lines (Figure 9). Floes that overlap with any piece of the subregion are identified, and the concentration is calculated first. Next, variables such as sea ice velocity and acceleration are calculated by scaling the contribution of individual floes by the mass of a floe present within the cell in question. Other variables, such as the total force exerted on a coarse grid cell, are not weighted by the mass of the floe experiencing the force.

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7 Examples of simulated sea ice behavior

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Here we present several test cases demonstrating the potential utility of the SubZero sea ice model. Specifically, we showcase simulations that highlight the specific physics of the model, including the role of floe fractures in a pure compression experiment, the

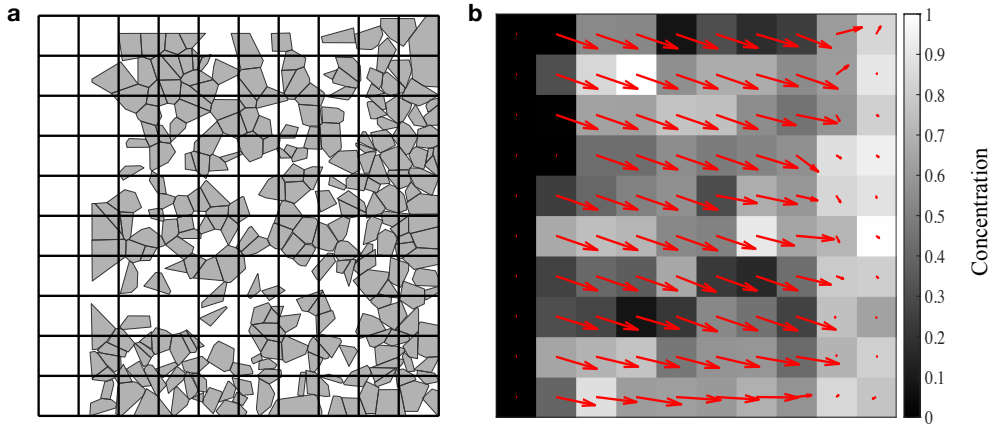


Figure 9. Example of a coarse graining including (a) a set of floes with total 50% concentration mapping the state of the floe model to the Eulerian grid where the domain is split into 10x10 grid with stationary solid boundaries. The ocean is stationary and the winds are blowing at 10m/s from left to right. (b) The homogenized values of floes plotted on the coarse Eulerian grid with shading indicating the concentration within the subregion and the arrows indicating the coarse sea ice velocity with a maximum velocity of 0.2m/s.

585 evolution of floe size distribution in a domain with a complex coastline, and the winter-
 586 time simulation that includes all model physics.

587 7.1 Evolution of sea ice floes under uniaxial compression

588 The behavior of granular-type materials, including sea ice, is commonly tested using
 589 idealized deformation experiments, e.g., subjecting the material to externally-imposed
 590 pure compression (Figure 10), tension, or shear. Here we demonstrate the behavior of
 591 sea ice floes subject to uniaxial compression in a confined domain, which is just one of
 592 the possible experiments that illustrates the non-standard formulation of the SubZero
 593 model. Each run is initialized with 200 floes in a fully-packed domain (Figure 10a), the
 594 North/South boundaries moving towards the center of the domain, and stationary East/West
 595 boundaries. A relatively small time step, $dt=5$ s, is used to resolve the elastic waves in
 596 response to external boundary motion and changes in the floe configuration due to fractures.
 597 The atmospheric and oceanic stresses are set to zero for this simplified test. The floes
 598 are subject to Mohr-Coulomb fracture criteria ($N_{frac} = 100$), but there is no floe
 599 simplification, corner grinding, welding, ridging, rafting, or creation of new floes in this
 600 scenario. The boundaries move with a constant prescribed velocity, $v_b = 0.1 \text{ m s}^{-1}$, and
 601 this leads to an initial increase in the numbers of floes (Figure 10c) and a reduction of
 602 the sea ice area when small floes are removed (minimum floe size allowed is 4 km^2) and
 603 ensures convergent sea ice motion (Figure 10b). The scenario is run for a range of three
 604 different Young's moduli, $E = (5 \times 10^7, 10^8, 1.5 \times 10^8) \text{ Pa}$, with the temporal evolution
 605 of the maximum normal stress averaged over an entire domain shown in Figure 10d. This
 606 experimental setup is included in the Zenodo repository (Montemuro & Manucharyan,
 607 2022).

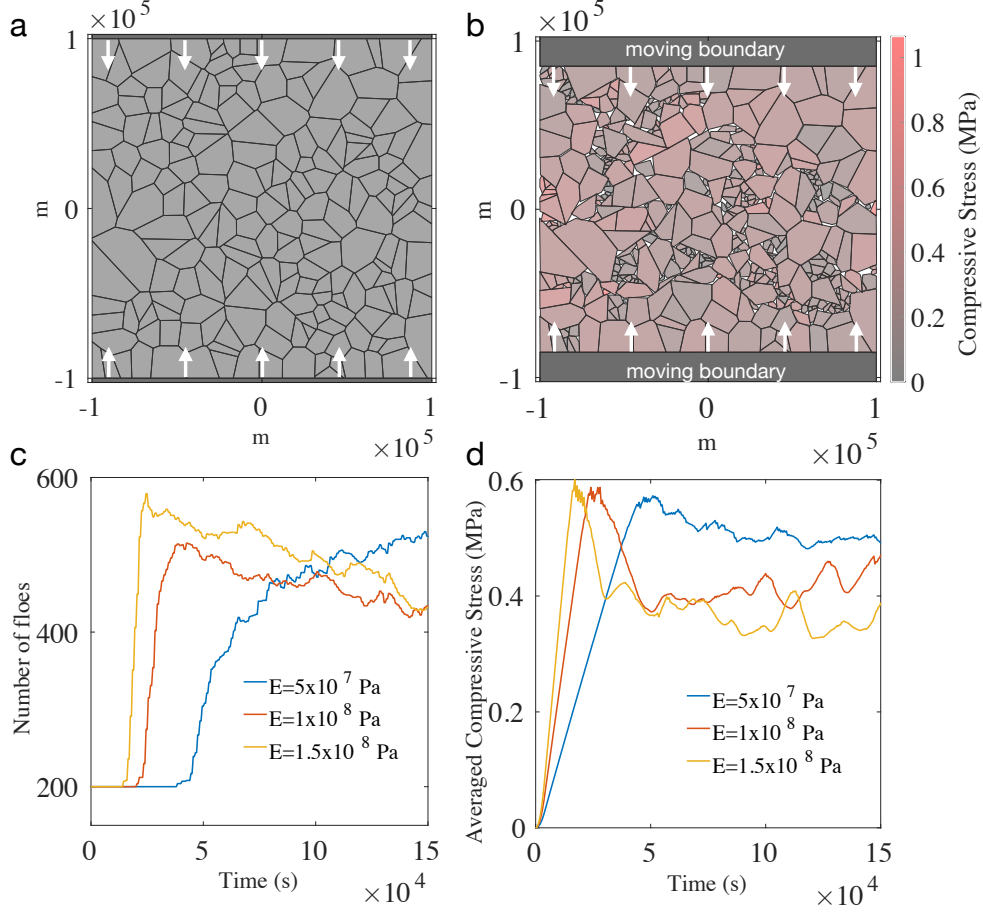


Figure 10. Evolution of sea ice under an idealized compression experiment. (a) The state of the floes at the beginning of the simulation for all three compression experiments. White arrows represent the imposed direction of motion of the top and bottom boundaries of the domain; the left and right boundaries remain stationary. (b) The state of the floes at the end of the simulation with a reference value of Young's modulus of $E=1.5 \times 10^8$ Pa, corresponding to the yellow curve in panel (c) and panel (d). Panel (c) shows the evolution of the number of floe elements in the simulation; the three curves represent runs with different Young's moduli, E , prescribed for the floes. (d) Temporal evolution of the maximum normal stress averaged over an entire domain; the three curves represent simulations with different Young's moduli, E , prescribed for the floes. The video of the simulation is available as supplementary material.

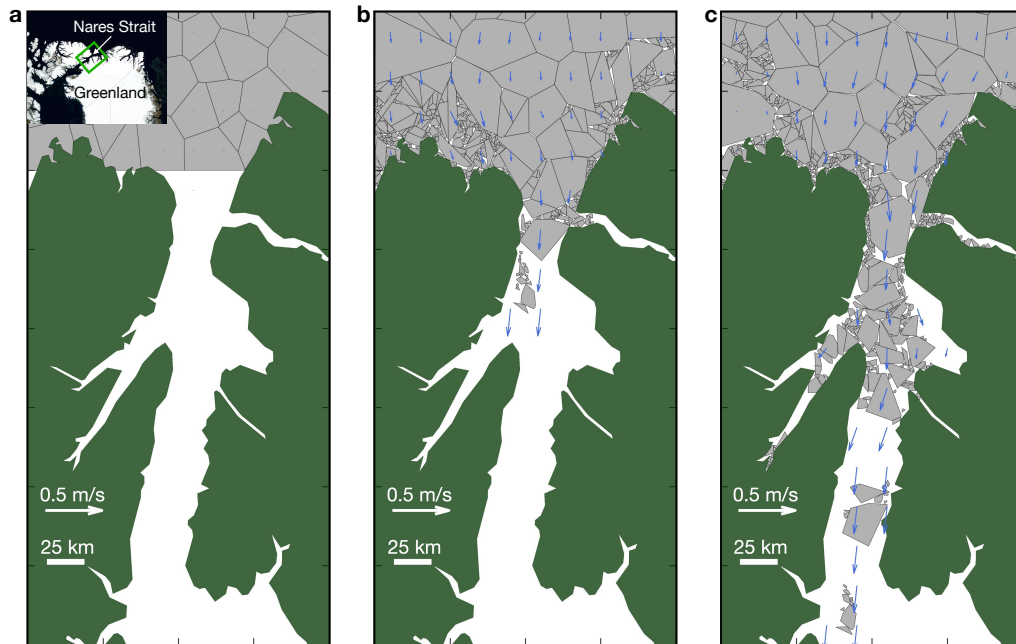


Figure 11. The evolution of sea ice floes as they pass through the Nares Strait, including (a) initial floe state with the inset showing the location of Nares Strait, (b) floes shortly after sea ice breakup that occurred after about three days, and (c) floe state after ten days when many floes have passed through the Nares Strait. The initial distribution of floes was generated using Voronoi tessellation, and the subsequent evolution of floe shapes is only subject to floe fractures. The green box in (a) shows the part of Nares Strait being simulated. The blue arrows represent sea ice velocity after averaging floe momentum on an Eulerian grid. The video of the simulation is available as supplementary material.

Table 2. A list of key parameters used in the SubZero model Nares Strait simulation, including their default numerical values, a brief description, and the processes that use these parameters.

Parameter	Description	Process
$E = 5 \times 10^7$ Pa	Young's Modulus	Floe Interactions
$G = \frac{E}{2(1+\nu)}$	Shear Modulus	
$\nu = 0.3$	Poisson's ratio	
$\mu = 0.25$	Coefficient of Friction	
$N_{Fract} = 150$	Time steps between fracturing	Floe Fractures
$N_{Pieces} = 3$	Number of pieces for fracturing	
$P^* = 1 \times 10^5$ N m ⁻¹	Floe strength-to-thickness ratio	
$\rho_i = 920$ kg m ⁻³	Density of ice	Floe mass and moment of inertia
$\rho_a = 1.2$ kg m ⁻³	Density of air	Surface stresses
$\rho_o = 1027$ kg m ⁻³	Density of ocean	
$C_{d_{atm}} = 10^{-3}$	Atmosphere-ice drag coefficient	
$C_{d_{ocn}} = 3 \times 10^{-3}$	Ocean-ice drag coefficient	
$N_{MC} = 100$	Number of sample points for Monte Carlo integration over floe surface	
$\Delta t = 10$ s	Integration time step	Time-stepping
$A_{min} = 2$ km ²	Minimum area of resolved floes	Floe state
$N_b = 18$	Number of floes creating the boundary	

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7.2 Summer sea ice motion through Nares Strait

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The Nares Strait simulation demonstrates the role of floe fractures in wind-driven sea ice transport through narrow straits. Nares Strait is a channel between Ellesmere Island (Canada) and Greenland (Figure 11a). The simulation aims to reflect spring or summer-like conditions of Arctic sea ice export through Nares Strait after the breakup of its winter arches (Figure 11). Due to floe jamming as they pass through the narrow constriction, the sea ice transport through the strait occurs in the form of episodic events (Kwok et al., 2010; Moore et al., 2021). Since the transport events are relatively short (order of days or less), the effects of thermodynamic sea ice melt could be considered secondary relative to mechanical floe processes such as collisions and fractures. We thus randomly initialize the model with relatively large floes of uniform thickness, covering only the area just north of the strait (see Table 2 for the list of parameters used in this simulation). The uniform 10 m/s southward winds generate stresses that push the floes through the strait, while the ocean is assumed to be stagnant. Coastal boundaries are prescribed using a series of $N_b = 18$ static floes. All physical processes except collisions and fractures are turned off to model the spring/summer breakup of floes. To suppress the rapid creation of tiny floes due to frequent fractures, we set up the simulation to resolve only floes with an area greater than 2 km². In this basic model formulation, we assume that the unresolved small floes do not significantly affect the dynamics of retained floes, and the model only tracks their mass density using the Eulerian grid to ensure mass conservation. Note that in more complex model formulations, the mass density could be used to parameterize the cumulative effect of small-scale floes on the dynamics of resolved floes. This experimental setup is included in the Zenodo repository (Montemuro & Manucharyan, 2022).

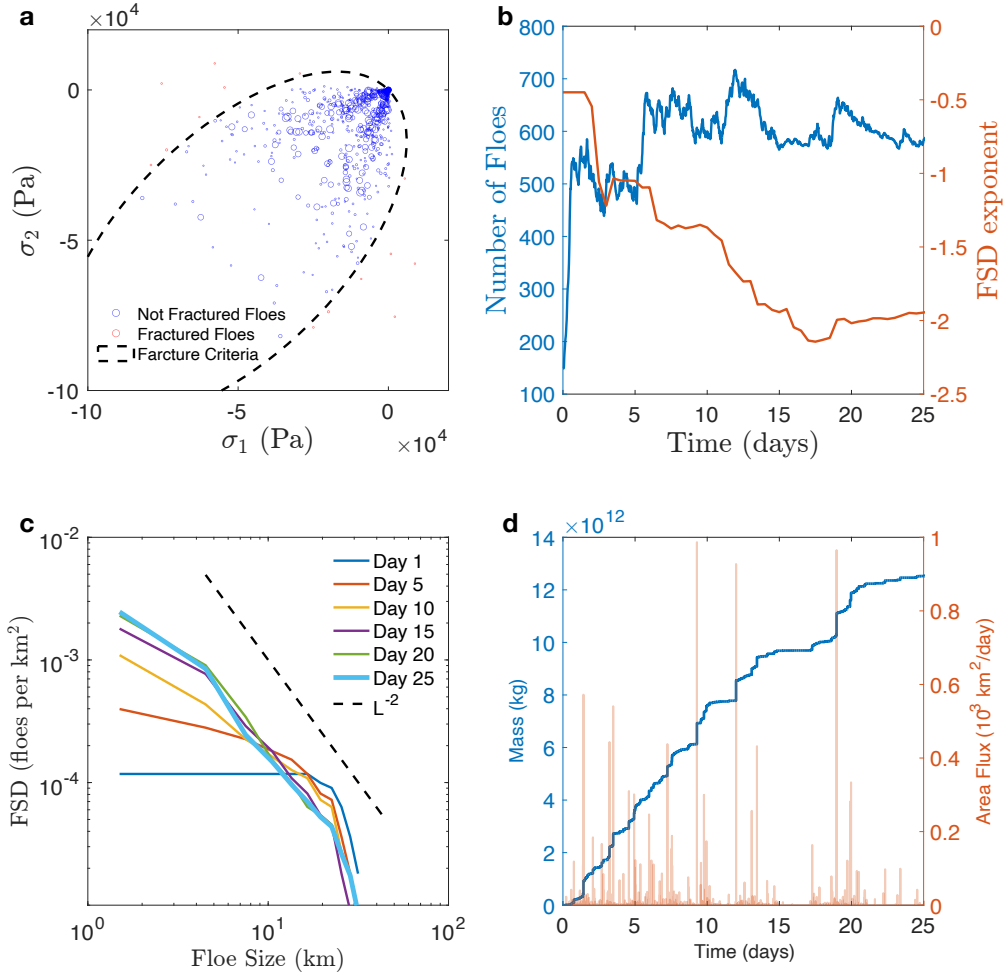


Figure 12. The evolution of the Nares Strait simulation. **(a)** Principal components of the individual floe stresses, with floes categorized by those that will experience fracture (red) and those that will not (blue). The black dashed curve represents a boundary for floe fracturing, an ellipse similar to a yield curve used in viscous-plastic sea ice rheology. **(b)** Temporal evolution of the number of resolved floes (blue) and the FSD exponent (red). **(c)** Floe size distributions for sea ice floes that are inside the Nares Strait. **(d)** The cumulative sea ice mass transport through the northern entrance to the Nares Strait (blue) and the corresponding area flux (red).

632 As winds push the initially large floes through the strait, the frequent floe fractures
 633 lead to an equilibrated floe size distribution (FSD) in just a few weeks (Figure 12b). The
 634 number of floes grows from dozens initially to several hundred (Figure 12b), but the FSD
 635 takes the form of a power-law distribution with an exponent close to -2 (Figure 12c). The
 636 FSD is free to equilibrate to a different power-law exponent (or not be a power law at
 637 all) depending on the forcing and floe interaction and fracture laws. In a winter-like sim-
 638 ulation described in the next section, the FSD also equilibrates to a power-law distri-
 639 ution but with a different exponent. Power-laws in FSDs have been commonly reported
 640 based on observations in various Arctic Ocean regions, with exponents ranging from about
 641 -3 to -1 (Rothrock & Thorndike, 1984; Holt & Martin, 2001; Horvat et al., 2019; Den-
 642 ton & Timmermans, 2021). A recent study using very high-resolution images demon-
 643 strates that within a wide range of floe sizes, the power-law exponent for the area-based
 644 FSD belongs to an approximate range from (-2, -1.65), which translates to a range of slopes
 645 (-3, -2.3) if size as the square root of the area is used to define FSD (Denton & Timmer-
 646 mans, 2021). The SubZero simulation with fractures only driven by mechanical floe in-
 647 teractions results in the FSD power law exponent of about -2, which compares reason-
 648 ably well with observations.

649 As the sea ice breaks into smaller floes, they can propagate through the relatively
 650 narrow strait. The sea ice mass flux through the strait is not smooth as floes often jam
 651 in narrow constrictions (Figure 11b). The jamming occurs when relatively large floes cluster
 652 in narrow parts of the strait, and sea ice can only move after some of those floes break
 653 into smaller pieces. The breaking of floes depends on the fracture criteria; an ellipse was
 654 used for this simulation to conceptually mimic Hibler’s elliptical yield curve used in con-
 655 tinuous viscous-plastic sea-ice models (Figure 12a). Floes with stresses lying inside the
 656 ellipse do not break, and those on the ellipse or just outside of it end up fracturing. These
 657 floe fractures lead to intermittent but large fluxes of sea ice area and transported mass
 658 (Figure 12d). The sea ice area fluxes in Nares Strait estimated using satellite and flux-
 659 gate observations are of the order $O(10^3)$ km²/day (Kwok et al., 2010; Moore et al., 2021)
 660 and generally agree with the idealized simulation with $O(10^3)$ km²/day for relatively rare
 661 high-transport events and about $O(10^2)$ km²/day for more frequent events. Thus, the
 662 idealized SubZero experiments can qualitatively simulate many aspects of sea ice dynam-
 663 ics relevant to flow through a narrow channel. However, the parameterization of certain
 664 physical processes still requires tuning using floe-scale observations. We expect that ob-
 665 servational estimates of FSD and mass fluxes inside Nares Strait and the driving forces,
 666 such as wind stress and boundary stresses, would be crucial for constraining floe colli-
 667 sion and fracture parameterizations. Winter-time sea ice dynamics in the Nares Strait
 668 also present a critical case study since sea ice can form arches that temporarily shut down
 669 its transport. This experiment is left for future studies, and we expect that it can be used
 670 to tune the balance between welding processes that bond floes together and fractures that
 671 break them apart.

672 7.3 Winter ITD and FSD equilibration

673 Here we demonstrate an essential case of model equilibration in winter-like con-
 674 ditions, where all parameterizations are active. For a model like SubZero that simulates
 675 time-evolving floe shapes and has a freely-evolving number of floes, it is of particular in-
 676 terest to explore if the FSD and ITD equilibrate to distributions resembling observations.
 677 We subject sea ice to strong mechanical and thermodynamic forcing over a five week pe-
 678 riod to facilitate an accelerated model evolution away from the initialized floe shapes,
 679 sizes, and thicknesses towards typical winter-like distributions. Specifically, we prescribe
 680 idealized ice-ocean stresses in the form of four equal-strength counter-rotating gyres (ar-
 681 ranged like mechanical gears, see Figure 13a) that create relative sea ice motion and fa-
 682 cilitate floe fractures and ridging. Alternatively, one could prescribe atmosphere-ocean
 683 stresses to achieve the same goal, but in this run the winds are set to 0. To make this
 684 a winter-like simulation, we ensured a continuous sea ice growth by specifying a fixed

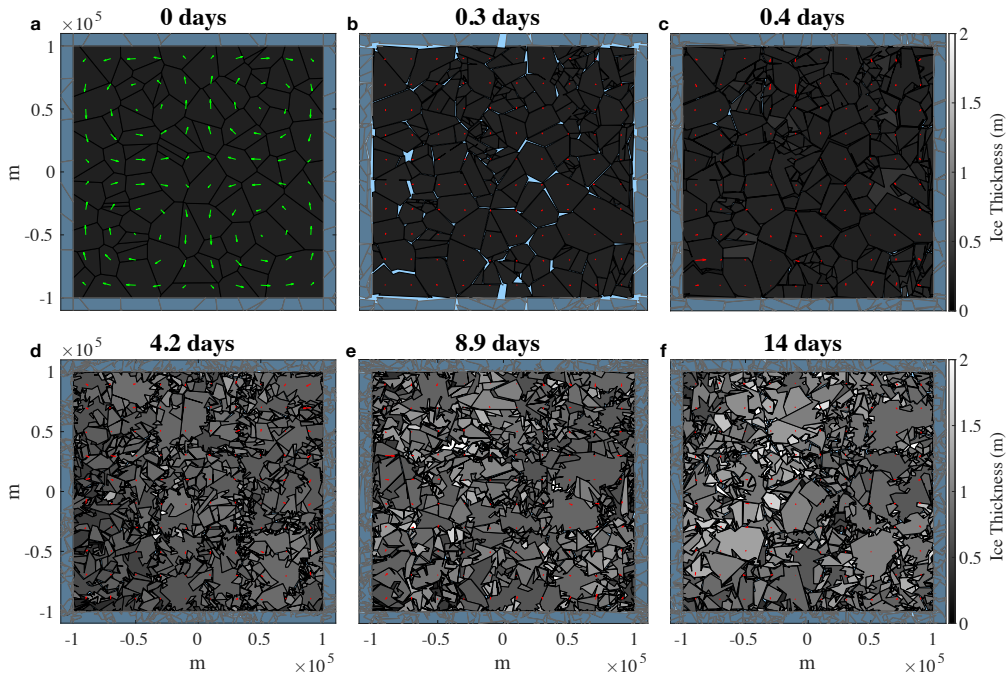


Figure 13. Evolution of sea ice during the winter-like simulation of sea ice growth with full physics of the model enabled. Panels **a-f** correspond to snapshots of floes and their thicknesses (shown with a grayscale color bar) at model times denoted in the panel titles. Panel **a** shows the underlying ocean forcing in green while panels with the maximum ocean velocity being 0.15m/s , and **b-f** show the sea ice velocity in a continuous sense after averaging floe momentum within grid boxes of an Eulerian grid with the red arrows. The video of the simulation is available as supplementary material.

685 negative heat flux that increases the thickness of existing ice floes, the formation of new
 686 ice floes in open ocean regions, and welding between floes (see Figure 2 for the simula-
 687 tion workflow). This idealized setup is aimed to demonstrate the evolution of floe shapes,
 688 sizes, and thickness under strong mechanical and thermodynamic forcing. We initialized
 689 the model with a fully-packed domain (100 floes) in which floes are cells of the Voronoi
 690 tessellation, all having the same thickness of 0.25 m and similar sizes (Figure 13). These
 691 initial floe thickness and size distributions are highly unrealistic. Below we describe how
 692 the dependence on these initial conditions is lost as the simulation progresses and how
 693 the emerging distributions start resembling the observed ones. This experimental setup
 694 is included in the Zenodo repository (Montemuro & Manucharyan, 2022).

695 In the early times of the simulation (within the first days), floe fractures and ridg-
 696 ing/rafting processes lead to rapid changes in ITD and FSD (Figure 14). The rates at
 697 which these processes occur are given in Table 3. The floe fractures form smaller floes,
 698 and this process establishes an approximate power-law distribution in the range of re-
 699 solved floes, which are larger than a few km. The floe fracture criteria used here again
 700 was an ellipse to conceptually mimic Hibler’s elliptical yield curve. The ice-free areas open
 701 up due to ridging/rafting, and new ice floes are formed there and consequently partici-
 702 pate in all processes. Note, the simulation is set to resolve floes with size above a cer-
 703 tain threshold, which we set to 2 km^2 for this simulation. After about a week, the power-
 704 law exponent of the FSD equilibrates to a value of about -3 , and the FSD starts resem-
 705 bling observations. Power laws in FSD are commonly found in various types of satellite
 706 sea ice observations, with the -3 exponent being well within the range of reported val-

Table 3. A list of key parameters used in the SubZero model, including their default numerical values, a brief description, and the processes that use these parameters.

Parameter	Description	Process
$E = 6 \times 10^6$ Pa	Young's Modulus	Floe Interactions
$G = \frac{E}{2(1+\nu)}$	Shear Modulus	
$\nu = 0.3$	Poisson's ratio	
$\mu = 0.3$	Coefficient of Friction	
$N_{Fract}=75$	Time steps between fracturing	Floe Fractures
$N_{Pieces} = 3$	Number of pieces for fracturing	
$P^* = 5 \times 10^3$ N m ⁻¹	Floe strength-to-thickness ratio	
$N_{cor}=10$	Time steps between corner grinding	Corner Grinding
$N_{Weld}=25$	Time steps between welding	Floe Welding
$P_{Heat} = 150$	Welding probability coefficient	
$P_{ridge} = 0.1$	Ridging probability coefficient	Floe Ridging & Floe Rafting
$P_{raft} = 0.1$	Rafting probability coefficient	
$h_c = 0.25$	Critical thickness for ridging to occur	
$N_{pack} = 5500$	Time steps between floe creation	Floe Creation
$\kappa = 2.14$ W m ⁻¹ K ⁻¹	Thermal conductivity of surface ice layer	
$L = 2.93 \times 10^5$ J kg ⁻¹	Latent heat of freezing	
$N_{simp} = 20$	Time steps between simplification of floe boundaries	Floe Simplification
$\rho_i = 920$ kg m ⁻³	Density of ice	Floe mass and moment of inertia
$\rho_a = 1.2$ kg m ⁻³	Density of air	Surface stresses
$\rho_o = 1027$ kg m ⁻³	Density of ocean	
$Cd_{atm} = 10^{-3}$	Atmosphere-ice drag coefficient	
$Cd_{ocn} = 3 \times 10^{-3}$	Ocean-ice drag coefficient	
$N_{MC} = 100$	Number of sample points for Monte Carlo integration over floe surface	
$\Delta t = 10$ s	Integration time step	Time-stepping
$A_{min} = 2$ km ²	Minimum area of resolved floes	Floe state
$N_b = 0$	Number of floes creating the boundary	

ues (Rothrock & Thorndike, 1984; Stern et al., 2018). Notably, our model simulation equilibrated to an approximate -3 power law, having only internal sea ice interactions as a cause of fractures. However, in marginal ice zones (regions where FSDs are often computed from observations), floes are also fractured by surface waves (Montiel & Squire, 2017) – a process that is not yet included in our model. Since the inclusion of waves would preferentially create smaller-scale floes, the FSD might have a steeper slope, making the power-law exponent closer to the observations. But before the wave fracture parameterization is included, our simulation can be considered applicable for conditions in pack ice, away from marginal ice zones.

The ITD also departs rapidly from the initial delta function distribution (all floes were initialized with the same thickness). By the end of the first week, the ITD takes the form of a double-peak distribution, with a second peak emerging at around 0.6 m due to ridging processes (Figure 14a). However, as time progresses, the second peak gets smeared out because many different ice thickness categories are ridged with each other. By the end of the month, the ITD takes a form of a smooth, single-peak distribution with a pronounced asymmetric tail for thick ice. The ITD continues to move towards thicker sea ice because of the thermodynamic growth, while the tail of the distribution and its asymmetry increase due to ridging (Figure 14c). At this stage, the dependence on the initially-prescribed ITD shape is lost, but the equilibrium is not reached as the ice continues to grow. The simulation would need to be run over multiple seasonal cycles, with winter-like sea ice growth followed by summer-like melt, to achieve an equilibrated ITDs. Nonetheless, we can still evaluate if these transient ITDs resemble winter-time observations, at least qualitatively.

The observed ITD is known to have an asymmetric shape that has been theoretically described using a gamma function distribution (Goff, 1995; Toppaladoddi & Wetlaufer, 2015) and the simulated ITD also resembles the gamma function distribution (Figure 14a, dashed line). While the shape of the ITD resembles observations, some of its quantitative metrics do not compare well. Specifically, Arctic-wide satellite-derived FSD for a winter month, like February, has a mean of 1.7 m and a standard deviation of 0.77 m (Kwok et al., 2020). The simulated ITD reaches a similar mean of about 1.5 m, but the standard deviation is only about 0.4 m, significantly lower than observations. Of course, our model simulation is highly idealized, and the resulting ITD would depend on the imposed mechanical and thermodynamic forcing and model parameters, all of which could be tuned for a better match with observations. However, an important reason for the mismatch is that the observed ITD is composed of sea ice that is a mixture of first-year ice and multiyear ice, with a ratio of about 1.4:1 in February, while our model simulation only has first-year ice as it is run for a short amount of time. Since multiyear ice is typically thicker than first-year ice, its presence skews the ITD towards higher thicknesses and contributes to its large standard deviation. Considering these factors, the simulated ITD can be considered to be in qualitative agreement with observations. With a more elaborate experimental design, it might be possible to reach a quantitative agreement. Since this paper aims to introduce general SubZero capabilities, we envision many crucial process studies performed by the broader sea ice modeling community.

8 Summary and Discussion

We constructed a sea ice model that treats ice floes as discrete polygonal elements. Its main advantage, and the key difference from existing sea ice DEMs, is that SubZero’s elements can change their shapes due to parameterized processes such as welding, fracturing, ridging, etc. Existing sea ice DEMs use fixed-shape elements (e.g., disks, rectangles, or tetrahedra), which can limit the interpretation of the model state when defining individual floes for comparison with data. Our model aims to bridge this gap and provide a framework that can be directly used to predict sea ice floe motion, either collectively in the form of floe size or thickness distributions or individually for each floe.

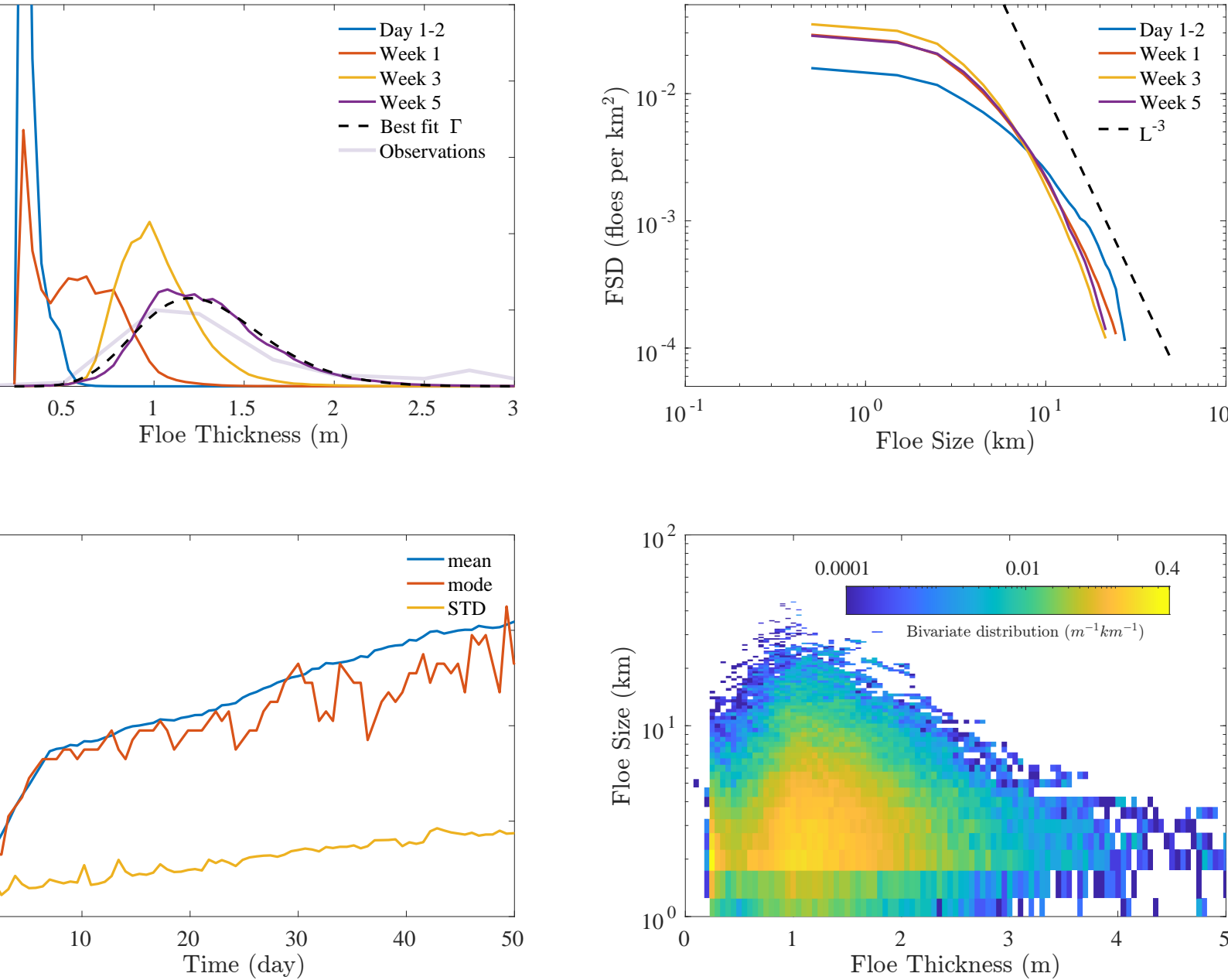


Figure 14. The evolution of floe size and thickness distributions for the winter simulation. **(a)** Ice thickness distribution (ITD) achieved in the early time after the initialization (blue), intermediate time (red), late time (orange) and at the end of the model simulation (purple); the best-fit gamma function is plotted for reference (dashed black line) and the lighter line shows observed thickness distribution via satellite. **(b)** Floe size distribution (FSD), plotted as the number of floes in a particular size bin per square kilometer; the L^{-3} power-law, L being the floe size, is shown for reference (dashed line). Note, floes smaller than 2 km^2 are not resolved in the simulation and only appear in the model as short-lived floes of recently fractured of larger floes. **(c)** Time evolution of the ITD mean, mode, and standard deviation. **(d)** Bivariate probability distribution of floe sizes and thicknesses, plotted for week 4 of the simulation.

759 We tested SubZero in several idealized scenarios to demonstrate its capabilities as
 760 a model of a granular and brittle material (the summer-time Nares Strait simulation)
 761 and a model with an active creation of new elements in addition to welding and fracture
 762 mechanics (the winter-time simulation). In both scenarios with idealized forcing and bound-
 763 ary conditions, the model-generated FSD had a power-law exponent ranging from about
 764 -2 (for pure fractures) to -3 for winter-like simulation. Both power-law exponents are well
 765 within the observed range. Similarly, during the winter-time sea ice growth simulations,
 766 the ice thickness distribution approached a qualitatively similar shape to the observed
 767 distribution, consisting of a single peak and an asymmetric tail for thicker sea ice. Since
 768 the model formulation specifies only the rules of floe interactions, one cannot guaran-
 769 tee that sensible equilibrated floe size and thickness distributions would emerge or that
 770 those would even remotely resemble the observed distributions. Yet, including only core
 771 processes with minimal parameter adjustment and using highly-idealized forcing and bound-
 772 ary conditions, the model approached a regime that resembles the observed sea ice be-
 773 havior. This qualitative, and for many metrics, quantitative consistency with observa-
 774 tions provides a substantial rationale for exploring various improvements to model physics.
 775 In particular, given its ability to explicitly simulate the floe life cycle, the philosophy be-
 776 hind SubZero strives to create a new generation of sea ice models.

777 We presented a proof of concept of a DEM with a varying number of elements that
 778 change their shapes subject to parameterized floe-scale physics. While the SubZero model
 779 already exhibits behavior consistent with sea ice observations, several improvements need
 780 to be made for it to become an operational sea ice model. Specifically, a more realistic
 781 formation of linear kinematic features could be achieved by developing more advanced
 782 floe fracture parameterizations, which would be an essential step toward mimicking floe-
 783 scale sea ice deformation. Another drawback of our model, and DEMs in general, is that
 784 its improved realism of floe dynamics is computationally demanding, and running such
 785 a model on basin scales presents a significant challenge. This issue could be addressed
 786 by improving the computational speed of the code using high-performance languages and
 787 GPU-enabled architectures. However, there will always be a limit to computing capa-
 788 bilities. Hence, to facilitate more accessible research and faster progress, developing com-
 789 putationally cheap basin-scale models would be necessary. One could envision theoret-
 790 ical studies attempting to formulate rescaled floe interaction rules (e.g., slightly mod-
 791 ified contact laws, fracture rules) such that floes in the model would effectively repre-
 792 sent clusters of floes of a particular scale. The problem of rescaling the floe interaction
 793 rules is tightly linked to the issue of representing the impact of unresolved floes and quan-
 794 titatively defining what a floe represents in physical space. Even in its prototype-like state,
 795 SubZero is an attractive new sea ice model that could be valuable for idealized process
 796 studies and regional simulations.

797 We now comment on key distinctions of SubZero from existing continuous and dis-
 798 crete element sea ice models. Continuous rheology models, like viscous-plastic models
 799 (Hibler, 1979), are meant to represent basin-scale sea ice motion and formulated for length
 800 scales larger than 10–100 km to describe characteristics averaged over a large number
 801 of floes. Unlike the SubZero sea ice model, continuous rheology models do not provide
 802 direct information about the positions, sizes, and shapes of individual floes, but they could
 803 provide statistical information such as FSD and ITD by solving their evolution equations
 804 subject to parameterized physics. SubZero’s output also can be presented in the form
 805 of Eulerian sea ice variables, like velocity or concentration. However, it is not a given
 806 that this discrete element model has equivalent continuous rheology describing the evo-
 807 lution of its Eulerian diagnostics. Hence, significant questions remain about using DEMs
 808 like SubZero to improve continuous sea ice models.

809 Comparing SubZero to existing sea ice DEMs, we can point out some key differ-
 810 ences. A general concept behind DEMs is to use pre-defined element shapes (such as points,
 811 disks, rectangles, or tetrahedra) to simplify calculations of collisions. More complex struc-

812 tures can be formed as clusters of simple elements that are bonded together. But this
 813 comes at the expense of computing forces for those bonds, which is typically a stiff prob-
 814 lem requiring small integration time steps. Consequently, it is challenging to use exist-
 815 ing sea ice DEMs for long-term simulations to study equilibrium sea ice distributions (such
 816 as FSDs and ITDs). Instead, such models are commonly used to address problems where
 817 the sea ice state does not dramatically evolve from initial conditions, i.e., initial-value
 818 problems. SubZero bypasses the issue of using a large number of stiffly-connected sim-
 819 ple elements by using complex floes with concave time-dependent shapes. Using com-
 820 plex floe shapes allows a straightforward creation of new elements in complex open-ocean
 821 regions between existing floes and simulating conditions with 100% ice cover using a mod-
 822 est number of floes. However, reducing the number of elements by transitioning to com-
 823 plex concave element shapes results in increased computational expense for resolving col-
 824 lisions and the need to parameterize floe-scale processes such as fractures and ridging.
 825 Parameterizations for the floe-scale processes could be derived by using the SubZero model
 826 by setting it up to resolve the sub-floe dynamics within individual floes; this approach
 827 is similar to nested runs used for resolving small-scale oceanic or atmospheric processes.
 828 The rationale behind SubZero’s formulation is that it might be sufficient to use param-
 829 eterized floe fractures and ridging (instead of explicitly resolving them) because these
 830 processes occur with high frequency and at a wide range of scales due to the highly vary-
 831 ing and strong wind forcing typical for the Arctic Ocean. When only the statistical be-
 832 havior of sea ice floes is of interest and exact details of individual fractures and ridging
 833 are not, a model like SubZero can effectively perform regional simulations of sea ice be-
 834 havior at seasonal scales. Thus, SubZero demonstrates a new approach to floe-resolving
 835 sea ice modeling, being distinct from existing continuous and discrete element sea ice mod-
 836 els. How the unique capabilities of the SubZero model could lead to our improved un-
 837 derstanding of sea ice dynamics remains to be demonstrated in future studies.

838 9 Data Availability Statement

839 The most up to date SubZero code (Manucharyan & Montemuro, 2022) is provided
 840 at the public GitHub repository <https://github.com/SeaIce-Math/SubZero>. SubZero
 841 v1.0.1 (Montemuro & Manucharyan, 2022) associated with this publication and test cases
 842 shown above can be found on Zenodo <https://doi.org/10.5281/zenodo.7222680>.

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