

Modelling near-surface ice content and midwinter melt events in mineral soils

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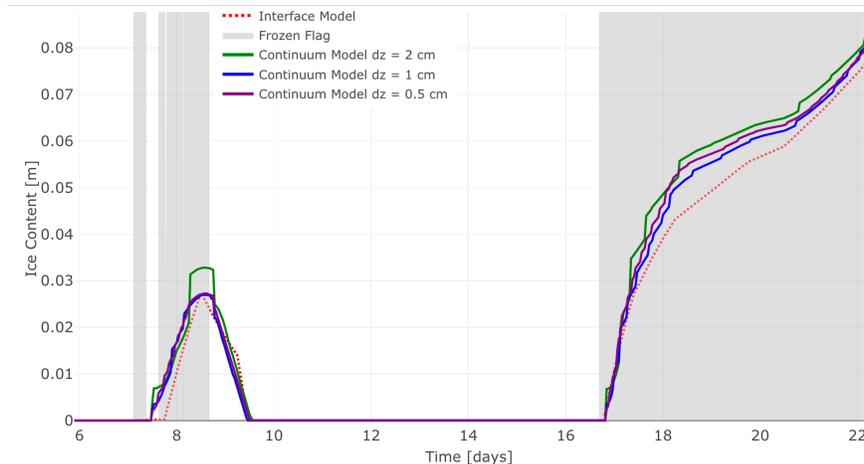
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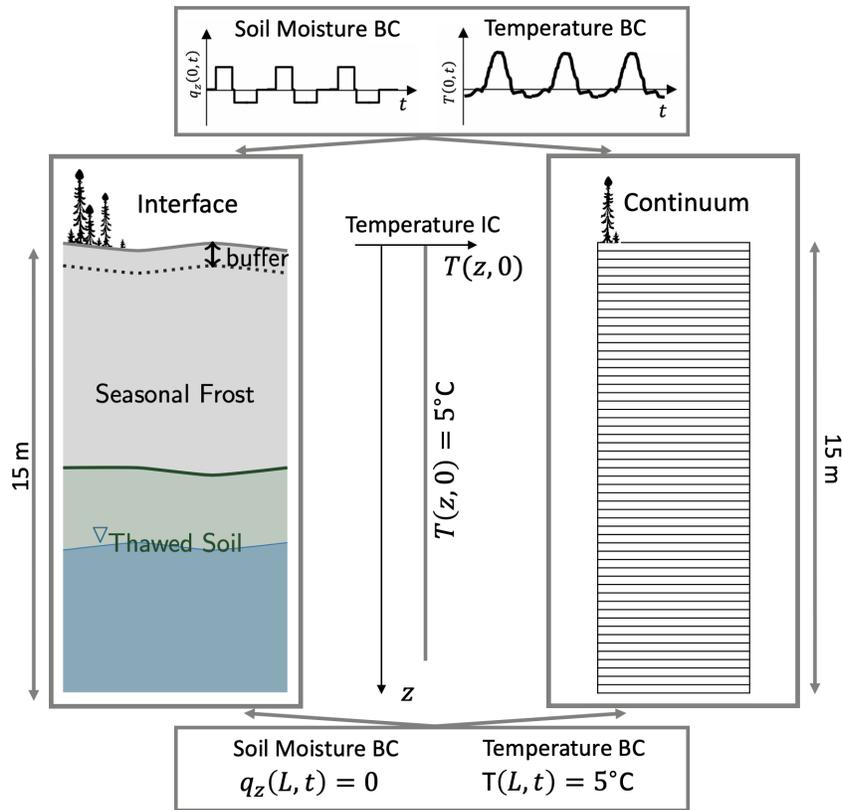
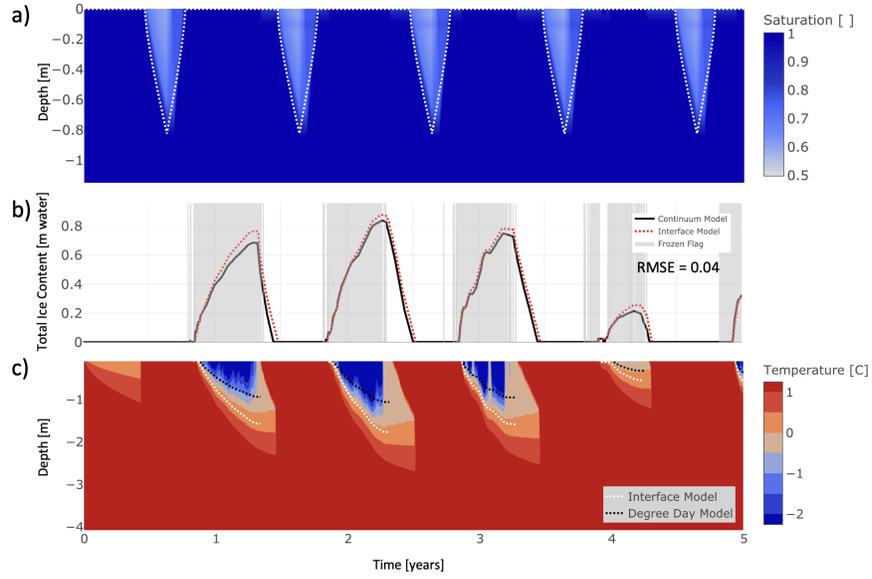
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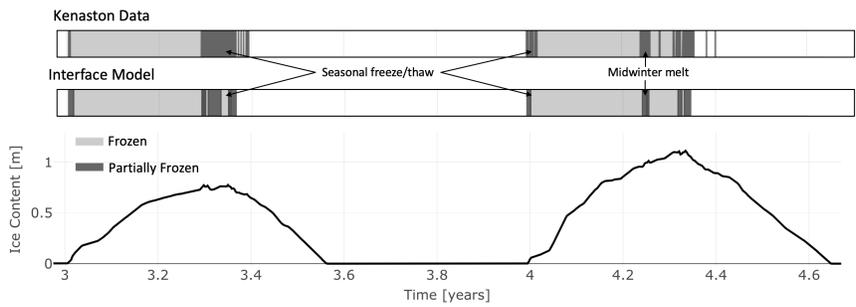
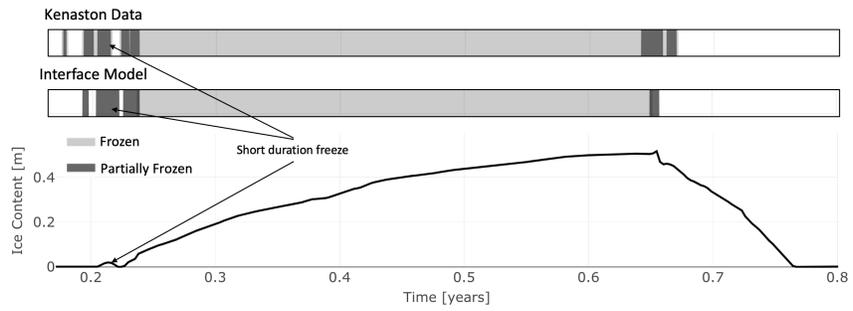
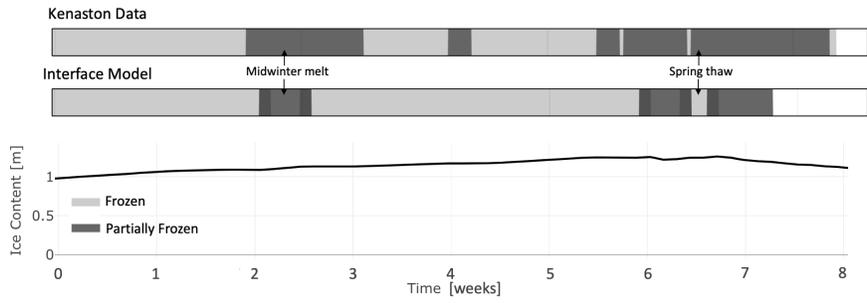
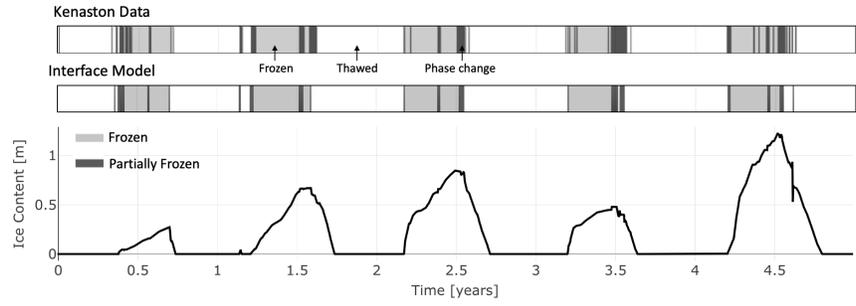
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Abstract

Over winter freeze-thaw events are notoriously difficult to represent in hydrologic models and have serious implications for the hydrologic function of intermittently freezing regions. With changing climate leading to higher variability in observed weather patterns, it is anticipated that mid-winter thaw events may become more numerous at locales where intermittent thaw was previously rare. Midwinter thaw events are often the cause of flooding due to the combined impacts of snowmelt, precipitation, and limited soil infiltrability. A numerically efficient, semi-analytical coupled thermal and mass transport model is presented that is capable of representing the ice content of near-surface soil. This model allows for rapid and stable prediction of the ice content of frozen or partially frozen near-surface soil without having to solve a discrete form of the coupled partial differential equations describing the soil water and energy balance. The model tracks pore ice formation and mean soil temperature in terms of enthalpy. It is tested against data collected in Southern Saskatchewan and is shown to capably reproduce field observations. This model is efficient enough to be incorporated as a module into existing regional hydrologic models and is expected to improve predictions of soil ice content, which can later lead to improved estimates of over-winter streamflow and flood potential.







Modelling near-surface ice content and midwinter melt events in mineral soils

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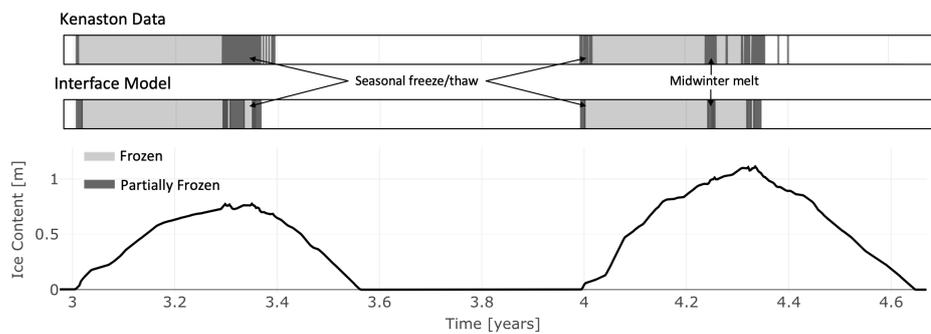
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Key Points:

- An interface model is presented to predict ice content of near-surface soil
- Model is benchmarked and validated against midwinter freeze-thaw event data collected in Southern Saskatchewan
- Algorithm performs efficiently and accurately and is recommended for use in hydrologic models



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Abstract

Over winter freeze-thaw events are notoriously difficult to represent in hydrologic models and have serious implications for the hydrologic function of intermittently freezing regions. With changing climate leading to higher variability in observed weather patterns, it is anticipated that mid-winter thaw events may become more numerous at locales where intermittent thaw was previously rare. Midwinter thaw events are often the cause of flooding due to the combined impacts of snowmelt, precipitation, and limited soil infiltrability. A numerically efficient, semi-analytical coupled thermal and mass transport model is presented that is capable of representing the ice content of near-surface soil. This model allows for rapid and stable prediction of the ice content of frozen or partially frozen near-surface soil without having to solve a discrete form of the coupled partial differential equations describing the soil water and energy balance. The model tracks pore ice formation and mean soil temperature in terms of enthalpy. It is tested against data collected in Southern Saskatchewan and is shown to capably reproduce field observations. This model is efficient enough to be incorporated as a module into existing regional hydrologic models and is expected to improve predictions of soil ice content, which can later lead to improved estimates of over-winter streamflow and flood potential.

Key Words: Seasonal Freeze/Thaw, Freeze/Thaw Modelling, Cold Region Hydrology, Midwinter Melt, Semi-Analytical Modelling

1 Introduction

It is well established that anthropogenic climate change is leading to increased variability in climate and more frequent and severe weather events (Pörtner et al., 2019). The Prairie and Boreal climate regions of Canada are characterized by seasonally frozen soils, with significant snow accumulation over winter (accounting for more than one third of the seasonal precipitation), an annual hydrograph dominated by spring freshet, and complete thaw of frozen soils by early to mid-summer (Fang et al., 2007). Intermittent frozen soils are ubiquitous in the temperate regions of the northern United States and Southern Canada. With a changing climate, areas that were previously frozen through the entire winter have been observed to be affected by more midwinter melt events (Williams et al., 2015). The ability to simulate midwinter melt events can be important, especially in sensitive Prairie and Boreal Plains systems.

48 In many hydrologic models, frozen soils are either treated as strictly impermeable
49 surfaces for the entire winter period (Niu & Yang, 2006) or empirical models are used
50 to address the changes in infiltrability due to ice content fluctuations over the winter months
51 (Luo et al., 2003). These approaches lead to an inability to accurately report the soil mois-
52 ture, thermodynamic state, hydraulic conductivity, infiltrability, and water storage of the
53 systems. Alternatively, many land surface schemes (e.g., Verseghy (2000)) explicitly sim-
54 ulate the full soil energy balance with freezing, which is typically accompanied by sig-
55 nificant computational cost. In systems that are generally quiescent over the winter months,
56 empirical models of over-winter processes have been found to be adequate (Luo et al.,
57 2003). This has motivated the use of empirical models such as that presented by Zhao
58 and Gray (1999), which improve model performance, but are not transferable to other
59 study sites, nor are they applicable in non-stationary systems such as those affected by
60 changing climates. A recent increase in midwinter thaw events and short duration freeze/thaw
61 events in the shoulder seasons make empirical predictions less and less accurate, to the
62 point where they may be insufficient to represent the hydrology of these systems (Pavlovskii
63 et al., 2019). This change is increasingly important as more extreme precipitation, es-
64 specially rain or rain-on-snow events over frozen ground, can lead to severe flooding. Pre-
65 diction of flood timing and extent is sensitive to estimates of infiltrability and hydraulic
66 conductivity of partially frozen soils (Seyfried & Murdock, 1997). To adequately sim-
67 ulate runoff in hydrologic models, it is crucial to understand infiltrability rates and pat-
68 terns (Luo et al., 2003). The infiltrability is strongly controlled by the ice content of the
69 soils, which in turn is dependant on the freeze/thaw history of the soils. Midwinter melt
70 events are known to introduce ice lenses and layers which impede spring infiltration into
71 froze soils (Pavlovskii et al., 2019). These melt events result in increased ice content in
72 the near-surface soil which, upon re-freezing, also affects the soil thaw rate in the spring.

73 The representation of soil ice content is included in some hydrologic models, espe-
74 cially those applied in permafrost regions (e.g. Wang et al. (2010); Luo et al. (2003); Wang
75 et al. (2017); Pomeroy et al. (2007)). It is shown that the accurate representation of frozen
76 soils, including the coexistence of frozen and liquid water, improves hydrologic predic-
77 tion in these regions (Niu & Yang, 2006), both for empirical and even more so for physically-
78 based models (Wang et al., 2010). Representing soil freeze/thaw processes directly is a
79 significant improvement over the null hypothesis that frozen soils are impermeable (Pomeroy
80 et al., 2007; Qi et al., 2019). However, physically-based thermal models are notoriously

81 demanding computationally, especially when coupled to mass transport of water in soils,
82 and the representation of freezing and thawing often increases computational time more
83 than ten-fold, and can also lead to instabilities and non-convergence of models (Wang
84 et al., 2017).

85 We here propose a semi-analytical physical model that efficiently predicts freeze/thaw
86 processes and ice contents in soils during midwinter melt and other short-duration freeze/thaw
87 events that are currently not well captured by empirical models. As such, this model will
88 fill the substantial gap separating physically-based, discrete continuum models from mod-
89 els that are purely empirical. The objectives of this paper are to (1) extend the meth-
90 ods developed for organic soils with permafrost by Devoie and Craig (2020) to mineral
91 soils without permafrost, (2) evaluate the extended model against a continuum model
92 benchmark, and (3) apply the model to intermittently frozen soil data collected at the
93 Kenaston Field site in Saskatchewan, Canada, with a focus on short-duration freezing
94 in the near-surface soil. Though the interface model is a front tracking model, the aim
95 of this study is to evaluate its ability to efficiently predict freeze/thaw events to inform
96 hydrologic models.

97 **2 Methods**

98 A combination of two modelling techniques and field-based measurements are used
99 to establish the validity of the proposed interface model for the representation of freeze
100 and thaw events in seasonally frozen mineral soils, especially for short duration midwin-
101 ter melt events. Model governing equations are described in Appendix A, while model
102 parameter definitions and values are summarized in Appendix D.

103 **2.1 Field**

104 ***2.1.1 Field Data***

105 Soil moisture, temperature and precipitation have been monitored at 22 stations
106 of the Kenaston Network located in the Brightwater Creek basin, east of Kenaston, SK,
107 Canada (Tetlock et al., 2019). This is predominantly an agricultural region, dominated
108 by annually cropped fields with some grazing land and without irrigation (Tetlock et al.,
109 2019). The instrumented monitoring network spans 40 km², with most of the instrumen-
110 tation within a flat 10 km² sub-region with slopes of less than 2%. The sites cover a soil

111 textural composition of 10.5 - 61.7 % sand, 31.2 - 72.4 % silt and 1.2 - 41.1 % clay, for
112 the base computational test, a representative soil (from Kenaston site 1) of 28 % sand,
113 53 % silt and 19 % clay was used (Pardo Lara et al., 2020, 2021). The mean annual air
114 temperature in this region is 8 °C, and in the last three decades the mean annual pre-
115 cipitation has been 400 mm of which approximately 30% falls as snow (Meteorological
116 Service of Canada, 2012). The catchment is semi-arid, and fluctuations in soil moisture
117 follow a seasonal pattern (Burns et al., 2016), though some fill-and-spill and non-contributing
118 areas are documented where water ponds in sloughs instead of contributing to the basin
119 outflow (Shook et al., 2013).

120 Soil moisture was measured using “HydraProbes,” commercially available electro-
121 magnetic sensors that report liquid water content from permittivity and temperature mea-
122 surements (Seyfried & Murdock, 2004). The sensors have 4 metal tines which are 3 mm
123 in diameter and 57 mm long. The zone of influence of the probe ranges approximately
124 from $4.0 \times 10^4 \text{ mm}^3$ to $3.5 \times 10^5 \text{ mm}^3$, with a radial range of approximately 13 to 35 mm
125 (Pardo Lara et al., 2021). Given the measurements from these probes installed at depths
126 of 5, 20, and 50 cm below the ground (Pardo Lara et al., 2020), it is assumed that the
127 near-surface probe is sensitive to water content in the top 50 ± 35 mm of soil, and this
128 near-surface layer is used to report the frozen, thawed, or transitioning state of the soil.
129 Soil temperature was measured alongside soil moisture and permittivity (as part of the
130 soil moisture measurement) at three depths: 5, 20 and 50 cm below the ground surface
131 (Burns et al., 2016). The mean annual soil surface temperature is approximately 5 °C.
132 Precipitation was also measured at each site using tipping bucket rain gauges. All data
133 was collected at 30-minute intervals (Tetlock et al., 2019).

134 ***2.1.2 Kenaston Data-driven Estimate***

135 A field-based approach to determining the frozen or thawed state of the soil was
136 used to generate validation data for the interface model discussed above. This approach
137 uses soil permittivity and temperature data to establish a site-specific freezing point. The
138 freezing temperatures were estimated using a logistic growth model fit to the soil freez-
139 ing curve, as detailed in Pardo Lara et al. (2020). This allowed consistent estimates of
140 when the soil is thawed, frozen, or undergoing phase change based upon the observation
141 data. These data were used to validate the predicted freeze/thaw status from the inter-
142 face model by specifying the field-data based freeze/thaw/transition flag. Though tem-

143 perature and soil moisture data are available, the sensors are only proven to indicate if
144 the soil is frozen, thawed or undergoing phase change (Pardo Lara et al., 2020). Extract-
145 ing the exact ice content from the HydraProbe data is unfortunately not yet proven for
146 these soils, and merits further investigation. This data was also used to identify midwin-
147 ter melt events, in which the freeze/thaw flag transitioned from frozen to partially frozen
148 and then returned to frozen.

149 **2.2 Model**

150 ***2.2.1 Interface Model***

151 The interface model described here is a semi-analytical solution to the heat equa-
152 tion coupled to an equilibrium solution to a mass balance relationship based on the van
153 Genuchten pressure-saturation relationship. This interface-based modelling approach,
154 where the depth below the ground surface of the frozen-unfrozen interface is treated as
155 a state variable, was first presented in Devoie and Craig (2020) in the context of active
156 layer modelling in discontinuous permafrost peatlands environments. The model was de-
157 scribed, benchmarked and validated in that paper, and applied to a specific case of thaw-
158 ing permafrost.

159 In this work, the interface model of Devoie and Craig (2020) is extended to rep-
160 resent seasonally frozen mineral soils. In this study, the bottom boundary condition of
161 the soil profile was fixed at a constant temperature based on field measurements. In each
162 of the simulations, the freezing point was specified based on field data. Mineral soils with
163 lower hydraulic conductivity challenge the original assumption that the water table was
164 in equilibrium, but modifications were made to water content representation (summa-
165 rized in Appendix C), and this led to adequate results as the model was not sensitive
166 to small changes in water content. The numerical implementation details and derivation
167 are included in Appendix C. Finally, the modelled soil layers were modified to accom-
168 modate seasonal freeze/thaw cycles congruent with the system shown in figure 1.

169 Given these modifications, the interface model reports the water table position as
170 well as the freeze/thaw fronts that exist in the subsurface (see figure 1). The updated
171 model also includes a surface “buffer” layer of fixed thickness that may contain a frac-
172 tional ice content (liquid water in excess of the residual unfrozen water content and solid
173 water co-existing in soil pores). This enables a better approximation of the near-surface

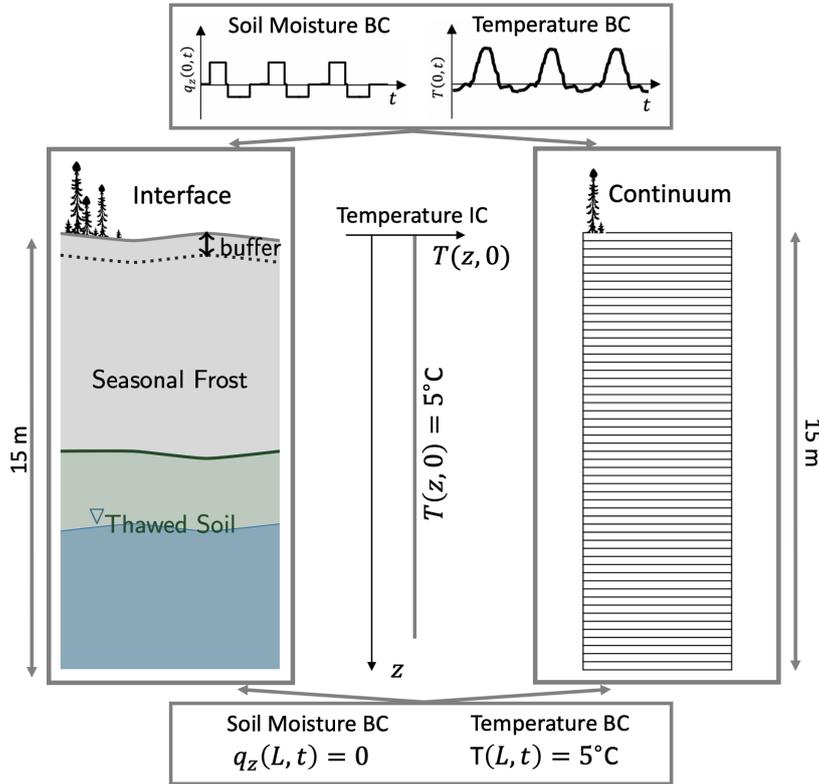


Figure 1. Schematic diagram showing model domain, boundary conditions (BCs) and initial conditions (ICs). The interface model (left) tracks the buffer layer (where fractional ice content is permitted) and the interface between frozen and thawed soil. The water table depth is also computed separately and updated through an equilibrium mass balance. The finite volume continuum model (right) used here for comparison uses operator splitting to solve the coupled PDEs describing heat and mass transport in 1D.

174 soil behaviour, and prevents the non-physical formation of many thin freeze/thaw inter-
175 faces that would have been required in the previous implementation. In this work the
176 buffer layer is specified to be 85 mm, which aligns with the depth of the field measure-
177 ments (section 2.1.1) of the near surface unfrozen water content used in model valida-
178 tion. This layer limits the fractional ice content to the near-surface, and is otherwise a
179 purely front-tracking model with user-specified residual unfrozen water content. The lim-
180 itations of this front-tracking approach are discussed in section 4.2. The interface model
181 is appropriate for representing the total ice content in the soil column (without its ex-
182 act spatial distribution), and estimating the freeze/thaw state of the near surface soil
183 as is needed for predictions of soil infiltrability.

184 *2.2.2 Continuum Model*

185 To validate the details of its formulation, the interface model was directly compared
186 to a coupled solution of the unsaturated Richards' equation and the energy balance equa-
187 tion solved via a finite volume method with operator splitting, as done previously in Devoie
188 et al. (2019). This detailed numerical solution allows us to assess the impact of the sim-
189 plifying assumptions made in the interface model while being forced with identical ini-
190 tial and boundary conditions, as well as test model representations of soil properties, pres-
191 sure saturation relations, soil freezing characteristic curves and model domains. The com-
192 parison here is meant to ensure that the interface model adequately represents the most
193 important physics. Identical initial conditions and boundary conditions were used in both
194 models (described in section 2.2.3), and a spatial discretization of 1 cm and 2 cm were
195 compared, both using 1 hour time steps in the continuum model. The same soil param-
196 eters were used for this model as were used in the interface model, with the addition of
197 a linear soil freezing characteristic curve (SFCC) for a (theoretical) freezing range of -
198 0.005 to 0 °C. This narrow range was chosen to match the interface model which does
199 not include an SFCC as it tracks a sharp interface without a slushy region.

200 *2.2.3 Initial and Boundary Conditions*

201 For comparison with field data, the model domains of both models were extended
202 to a depth of 15 m, using a fixed soil temperature of 5 °C at the base of the profile, con-
203 sistent with the mean annual soil surface temperature of 5 °C, and the negligible geother-
204 mal gradient of 0.002 °C/m (Majorowicz & Grasby, 2021). This temperature also aligns

205 with data collected at a depth of 15m near Edmonton, Canada (Toogood, 1976). This
206 was the nearest geothermal data available to the study site, and at the depth of 15 m,
207 the spatial variability of temperature is very low. This temperature is in agreement with
208 data taken in Saskatoon, where the average soil temperature at 3.0 m was 6 °C, decreasing
209 with depth, though there was still evidence of seasonal variation (Wittrock & Dunn,
210 2016). An initial water table position was assigned at 1 m below the ground surface, based
211 on the water table data collected in the field in early spring. Mass flux at the surface was
212 applied seasonally, with an average ET rate (-2.24×10^{-3} mm/d) applied in spring/summer
213 and an average recharge rate (2.88×10^{-3} mm/d) in the fall. The 15m depth was specified
214 to be well below both the expected minimum water table depth and the extinction
215 depth, but the precise choice of 15m is arbitrary: one of the benefits of the semi-analytical
216 model is that a large vertical extent does not increase computation time. The net mass
217 flux was zero annually. A no-flow boundary condition was assigned at the base of the
218 soil column to represent the near-impermeable unweathered till underlying this system,
219 and the bedrock beneath that (Shaw & Hendry, 1998). The surface temperature boundary
220 condition was drawn from soil temperature collected at a depth of 5 cm in the field
221 sites near Kenaston, and forced with a seasonally cyclic moisture boundary condition
222 (reported in section 3) as direct application of the infiltration flux data collected in the
223 field precluded convergence of the continuum model used for benchmarking. The soil column
224 was initialized to a thawed uniform temperature of 5 °C, and the freeze/thaw discriminant
225 temperature was assigned based on the specific freezing point depression determined from the
226 field data, ranging between 0 and -0.4 °C (Pardo Lara et al., 2020).
227 Simulations were started in the summer of 2012, except at sites 16 and 18 which were
228 started in summer 2013 due to lack of data. All simulations were run for a duration of
229 5 years, with associated computational time of 11s for each simulation. As described in
230 Devoie and Craig (2020), the surface layer of the interface model is a ‘buffer layer’ which
231 may contain fractional ice content. The depth of this buffer layer was assigned based on
232 the zone of influence of the soil moisture measurements made in the field. This allows
233 the ice content of the buffer layer to be compared to the measured ice presence of the
234 near-surface soil in the field. Below the buffer layer the freeze/thaw front is a moving
235 sharp interface and fractional ice content is not permitted. Because of the moving interface,
236 there is no spatial discretization of the interface model, however there is temporal
237 discretization, and the simulations reported here are run with a 1 hour timestep

238 for comparison with the finite volume model in figure 7 and a 1 day timestep otherwise.
239 Other soil parameters were homogeneous and independent of depth, and are summarized
240 in table 1 in Appendix D both for organic and mineral soils.

241 **3 Results**

242 The interface model presented in (Devoie & Craig, 2020) is extended to treat the
243 case of seasonal ground ice, enabling it to represent seasonal freeze thaw in mid-latitude
244 continental climates. Here, the simulation of seasonal freeze-thaw is first verified via a
245 numerical benchmarking study in an unsaturated system typical of mineral soils in the
246 semi-arid climate of Southern Saskatchewan. Boundary conditions and soil parameters
247 were obtained from field data, but no direct measurements of freeze/thaw are available
248 for the benchmark; these tests are purely to demonstrate numerical accuracy of the method.
249 Finally, the interface model predictions are directly compared to the data-derived freeze/thaw
250 status at sites in the Canadian prairies to evaluate the practical efficacy of the method.
251 An additional comparison between the interface model and the continuum model in a
252 near-saturated peat soil column is included in Appendix E for saturated and unsaturated
253 organic soils.

254 **3.1 Kenaston**

255 The model was evaluated for a five-year simulation based on field data collected
256 at one field site of the Kenaston Soil Moisture Network, with a 15 m vertical domain and
257 realistic thermal initial and boundary conditions as detailed in section 2.2.1. Figure 2
258 shows the comparison between the continuum model, interface model, as well as field data
259 indicating the ‘frozen period’ (shaded in grey). The shaded grey areas in figure 2 (b) in-
260 dicate the period over which the near-surface soil (approximately 40 - 85 mm) at the field
261 site was frozen. This data is drawn directly from field measurements using the method-
262 ology outlined in section 2.1.1, and compares favourably with the reported freeze/thaw
263 timing. The use of field data resulted in an increase in RMSE to 0.04 between the to-
264 tal ice content simulated by the interface model and continuum model, which is still ex-
265 cellent agreement. The interface position in figure 2(c) tracks the zero degree isotherm
266 relatively well, though the interface position is slightly deeper when compared to the dis-
267 cretized model. The performance of the interface model is however significantly better
268 than a simple degree-day method from Fox (1992) which significantly under-estimates

269 the thawing front, shown in black in figure 2(c). The simulation was re-run using finer
 270 spatial and temporal discretization in the continuum model (shown in figure 3) to cap-
 271 ture the exact timing of a specific freeze/thaw event, which was not captured using the
 272 model setup used to simulate the entire period (5 cm spatial discretization and 3 minute
 273 timestep).

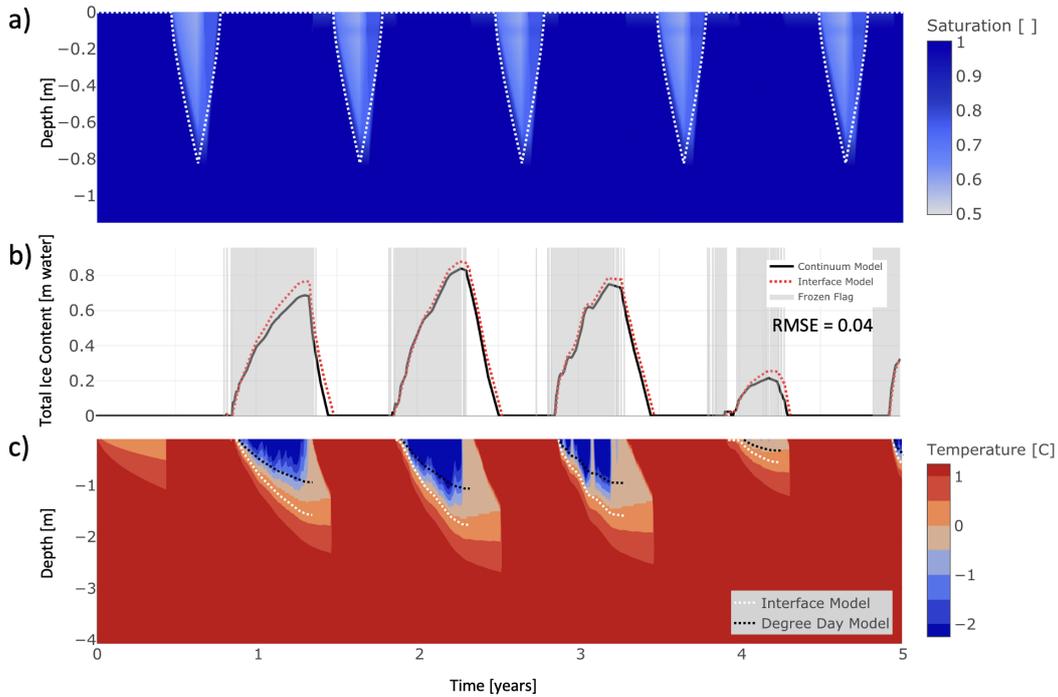


Figure 2. (a) Comparison of water content for interface and continuum model (b) comparison of total ice content for continuum model (black), interface model (red) and field-based near-surface frozen flag (shaded grey) and (c) contour plot of continuum model temperature with freeze/thaw interface position from interface model superimposed in white dashed line. Degree day model from Fox (1992) in black dashed line. Field-data driven with surface water flux approximated as seasonally uniform due to stability constraints for continuum model, soil texture data drawn from Kenaston Site 1 in Table 1 of Appendix D

274 The comparison of the interface and continuum model for the short-duration event
 275 in figure 3 was generated using the same model configuration as figure 2, but with finer
 276 spatial and temporal discretization of both models. The comparison of computational
 277 efficiency can also be established in figure 3 as the continuum model run took 2 hours
 278 and 22 minutes (in blue) while the interface model (red) only took 4.5 seconds for the

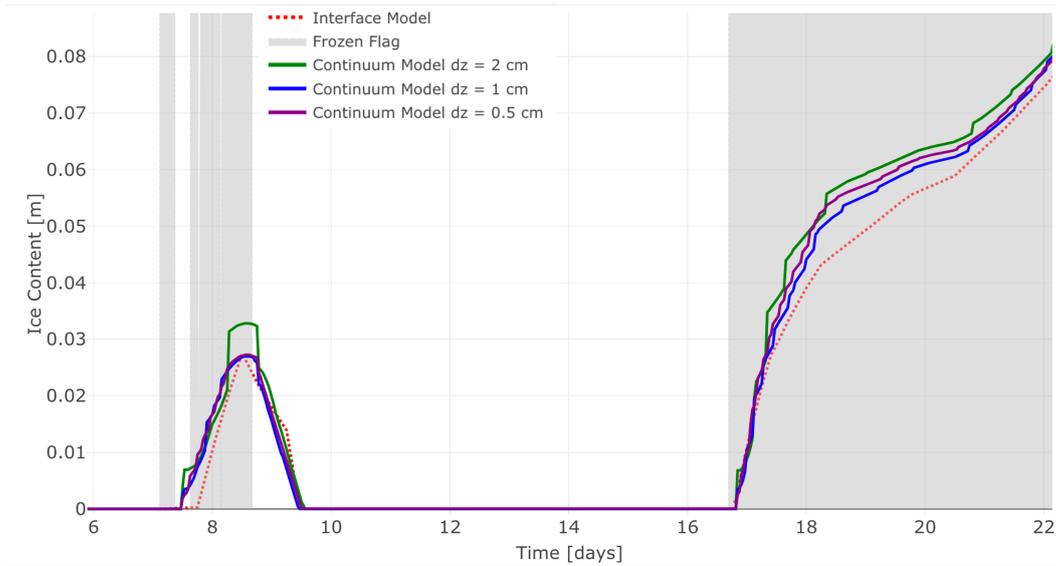


Figure 3. Short duration freeze/thaw initiation. Comparison of interface model (with 6 hour timesteps) to continuum model with timesteps chosen to satisfy convergence criteria given spatial discretization. This short duration initiation of freezing results in a small quantity of near-surface ice, hence the small total ice content. Spatial steps larger than 2 cm do not capture the near-surface freezing event in the continuum model. Model convergence is assumed based on the similarity between the 1 cm and 0.5 cm simulations. Grey shaded regions indicate soil freezing according to the field-data. Soil texture data drawn from Kenaston Site 1 in table Appendix D

279 same size time step and simulation setup. The performance of the interface model is ar-
280 guably better than the continuum model: when the spatial discretization of the contin-
281 uum model is refined, it tends toward the interface model solution. Larger spatial steps
282 lead to a lack of identification of the freezing event in the continuum model, and smaller
283 spatial (and associated temporal) discretization was computationally impractical. The
284 interface model also shows better timing and more gradual response to freeze/thaw events.
285 Neither model captures the initial freezing event near day 7, likely due to the choice of
286 (theoretical) freezing point depression ($-0.005\text{ }^{\circ}\text{C}$) and the freezing range between 0 and
287 $-0.01\text{ }^{\circ}\text{C}$ for the interface and continuum models respectively. Subsequent figures gen-
288 erated using only the interface model without continuum model comparison use the freez-
289 ing point depression determined from field measurements at the given field sites in or-
290 der to better capture such events.

291 **3.2 Midwinter Melt**

292 The benchmarked interface model (but not the continuum model) was then applied
293 to simulate all of the available data for similar mineral soil sites. A total of 22 sites were
294 considered in which subsurface temperature and soil moisture were recorded for a du-
295 ration of 4 - 6 years between 2014 and 2020. In 10 of these 22 sites clear mid-winter thaw
296 events were identified, in which the soil temperature warmed above 0°C . The interface
297 model was run using near-surface soil temperature data available at these sites, and com-
298 pared to the freeze/thaw flag extrapolated from the field data. Here a second “transi-
299 tion” flag was added to the field data representing soils undergoing phase change; if the
300 surface layer of soil contained fractional ice content based on its permittivity this flag
301 was activated. This flag is shaded in dark grey in the subsequent figures, while entirely
302 frozen near-surface soils (with only residual water content) were assigned a “frozen” flag,
303 depicted in light grey and thawed near-surface soils were left as white bands. The inter-
304 face model was compared to the two field-data based flags using the near-surface “buffer”
305 layer in the model. The depth of this surface soil layer is 85 mm in accordance with the
306 (maximal) depth of influence of the soil moisture probes used to collect the field data
307 (Pardo Lara et al., 2021). Note that once this buffer layer is completely frozen, the to-
308 tal ice content is allowed to continue to increase as the freezing front moves downward
309 beyond 85 mm. Two separate flags were also implemented in the model - the first “tran-
310 sition” flag representing fractional ice content in the near-surface, and the second “frozen”

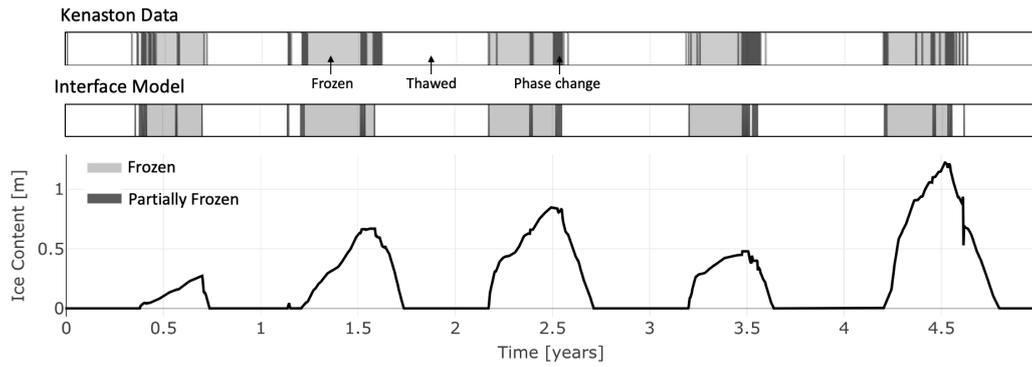


Figure 4. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). Total ice content from interface model shown along bottom axis. Soil texture data drawn from Kenaston Site 3 in table 1 Appendix D

311 flag indicating residual water content only, these are assigned the same colours as the
 312 field data. Sample results for the entire 5 year simulation at Kenaston site 3 are shown
 313 in figure 4, showing agreement between the interface model and data-extrapolated freeze/thaw
 314 timing. Two error metrics are used to compare the simulated and observed near-surface
 315 ice content. The first indicates the overall agreement between the modelled and mea-
 316 sured data including frozen, thawed and transitioning states. For the data in figure 4,
 317 the agreement is 92%, indicating that the measured and modelled soils did not have the
 318 same freeze/thaw state only 8% of the time. The second metric was conceived to iden-
 319 tify the effectiveness of the interface model at identifying frozen soils, and so it compares
 320 the soil state only when the measured field data is frozen, and does not take into account
 321 partially vs. completely frozen soils. For this study case, there is 91% agreement, indi-
 322 cating that the interface model incorrectly identified frozen soil as thawed 9% of the time.

323 4 Discussion

324 4.1 Interface Model Limitations

325 The interface model used in this study is a front tracking model, and its greatest
 326 limitation is therefore that it does not have the capacity to represent a slushy zone be-
 327 yond the buffer layer. It does not use a soil freezing characteristic curve (SFCC), and
 328 is therefore will not be as robust as a continuum model when detailed information on

329 the fractional ice content is needed. This is especially true for soils with SFCCs having
330 a wide temperature range such as clay-rich materials. The authors also caution against
331 the use of this model in small-scale systems with significant groundwater recharge or dis-
332 charge, as these processes depend on the detailed knowledge of distributed soil ice con-
333 tent to calculate fluxes. The interface model is however a good approximation of real-
334 ity in sandy, coarse-grained soils where the SFCC is quite steep and the slushy zone is
335 limited. It is also a valuable tool in the case of large-scale hydrologic simulations, which
336 are limited by computational efficiency. In these cases, the approximation of freeze/thaw
337 state in the near-surface provided by the interface model is superior to the current low-
338 fidelity empirical models, as seen in figure 2(c) (Fox, 1992).

339 **4.2 Model Evaluation: Near-surface buffer layer**

340 Figure 4 demonstrates agreement in the timing of broad seasonal events between
341 the modelled data and data collected in the field, and figure 5 shows a more detailed view
342 that distinguishes the typical seasonal freeze/thaw (i.e. freeze in the fall/early winter and
343 thaw in the spring) from midwinter melt events. The interface model is highly effective
344 in detecting the timing of freeze/thaw initiation, however the freeze/thaw transitions of
345 the near-surface buffer layer tend to occur sooner than in the measured data (Figure 5
346 & 6), in part due to the changing volume integrated in the HydraProbe’s measurements
347 and perhaps due to an under-estimate of the water content (and hence effective heat ca-
348 pacity) of the soil, alternatively an under-estimate of the volume integrated in the field
349 measurements. As the soil freezes, its permittivity decreases, and the integrated volume
350 of the HydraProbe measurement increases, delaying the observation of the frozen con-
351 dition by the sensor. These explanations are also supported by the tendency of the in-
352 terface model to begin to change phase more rapidly. An under-estimate or mismatch
353 of the volume which must undergo phase change due to an under-estimate of the near-
354 surface layer would result in more rapid freeze/thaw. It is also noteworthy that the to-
355 tal ice content in the soil column changes very little due to these short-duration freeze/thaw
356 events. Generally we see a flattening of the slope during a midwinter melt (e.g. figures
357 5 and 6), where ice accumulation does not occur, however there is no clear evidence for
358 significant ice loss during these events. It is difficult to establish the measured depth of
359 thaw from the available field data, but there is no evidence that thaw extends beyond
360 the first soil moisture and temperature sensor at a depth of 50 mm, limiting the antic-

361 ipated ice loss to less than 25 mm given unsaturated soil conditions and a soil porosity
 362 not exceeding 0.5.

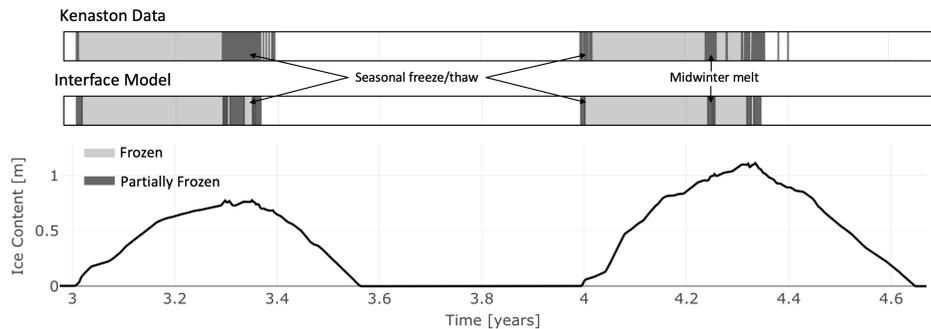


Figure 5. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). This 2 year subset (2016 - 2017) from 5 year simulation (2013 - 2018) drawn from Kenaston site 15. Seasonal freeze thaw at the near surface occurs in fall and early spring, while a mid-winter melt event is highlighted in year 4. For this simulation, the overall agreement between freeze/thaw states was 94%, while the interface model correctly identified 95% of the frozen period.

363 4.3 Model Evaluation: Freezing point

364 Figures 5 and 6 also demonstrate that thaw occurs sooner in the interface model
 365 than in the extrapolated field data, though the interface model does accurately capture
 366 96% of the frozen/thawed data. It is thought that this is due to the single freezing point
 367 depression used to interpret the data. It is known that there is hysteresis in the freeze-
 368 thaw process, and that the freezing point temperature is generally lower than the thaw-
 369 ing point (Saberri & Meschke, 2021). This leads to more rapid initiation of modelled thaw
 370 as it is initiated at a colder temperature than would realistically be observed in the field.
 371 More work, including investigation of hysteretic behaviour in freeze/thaw modelling is
 372 needed, such as (Amiri & Craig, 2019) or the physical analysis of hysteresis, such as (Pardo Lara
 373 et al., 2021).

374 4.4 Model Evaluation: Near-surface water content

375 The small difference in freeze/thaw timing may also be driven by a mismatch in
 376 near-surface soil water content between simulated and observed. The error in estimated

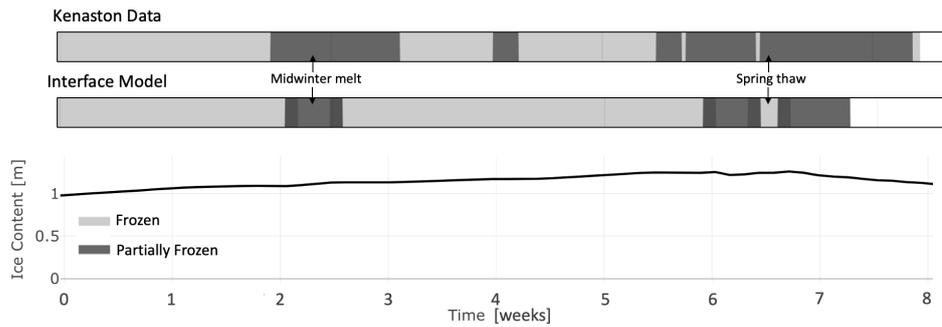


Figure 6. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). Interface model phase change takes less time, perhaps because of an under-estimate of the freezing point. Overall agreement between the freeze/thaw states is 94%, while the interface model correctly identifies 96% of the frozen period. Detail view from 5 year simulation drawn from Kenaston site 20.

377 soil water content may arise because an equilibrium soil moisture profile is implemented
 378 in the interface model, as detailed in Appendix C. The application of the model to min-
 379 eral soils was expected to require a more complex representation of infiltration events
 380 including plug flow and moisture redistribution, but these were not found to be neces-
 381 sary in the reproduction of the freeze/thaw conditions in field observations of near-surface
 382 soils. The equilibrium assumption seems to be adequate for two reasons; first, the sur-
 383 face mass balance used is based on seasonal trends and is very smooth. This results in
 384 near-equilibrium moisture conditions in the soil column over most of the freeze/thaw sea-
 385 son. Secondly, the quantity of interest is the frozen state of the near-surface soil. When
 386 mineral soils freeze, the impedance of ice in the soil pores is such that infiltration and
 387 evapotranspiration are negligible, and therefore these processes have little effect on the
 388 model results.

389 4.5 Spring thaw

390 Measurements of spring thaw (and some midwinter events) lead to small and rapid
 391 fluctuations in ice content in the surface layer. Spring temperatures in the Kenaston re-
 392 gion have strong diurnal fluctuations, where the daytime temperature is well above the
 393 freezing point, but the overnight low is around $-1\text{ }^{\circ}\text{C}$. In the interface model, the near-
 394 surface ice content is estimated in the top 85 mm of soil, deemed equivalent to the depth

395 of soil characterized by the field based freeze-thaw flag. This layer was included in the
 396 model as a mathematical construct that would prevent the formation of very thin, non-
 397 physical frozen and thawed layers at the soil surface. Even with this layer, the interface
 398 model fails to capture many diurnal-fluctuation driven spring freeze/thaw events. How-
 399 ever, these primarily occur when the underlying soil is frozen, and so the inability to track
 400 fractional ice content in the near-surface soil (especially when the ice content never freezes
 401 the pore water completely) likely has very little effect on the infiltration capacity and
 402 subsurface water movement. Water movement in the landscape is expected to be much
 403 more strongly affected by the fully frozen (less the residual water content) near-saturated
 404 layer at a depth of 10 - 15 cm below the soil surface. The relatively thin surface layer
 405 cannot store significant thermal energy, and the surface topography generally exceeds
 406 the scale of this layer, restricting the formation of flow pathways beyond the plot scale.
 407 The buffer layer may however still be physically meaningful, as there is evidence for the
 408 development of surface layer which undergoes freeze/thaw in a soil subject to midwin-
 409 ter thaw events. As noted by the temperature sensors in the soil profile, short thaw events
 410 do not extend beyond the top 100 mm of soil, though this surface layer experiences tem-
 411 perature cycling and freeze/thaw throughout the winter as well as the shoulder seasons
 412 when strong diurnal temperature cycles are common. The increased freeze-thaw cycling
 413 can lead to changes in soil structure (Alkire & Morrison, 1983) and changes in decom-
 414 position of soil organic matter (Yanai et al., 2004). Further investigation is required to
 415 establish if this layer is physically significant across landscapes experiencing freeze-thaw.

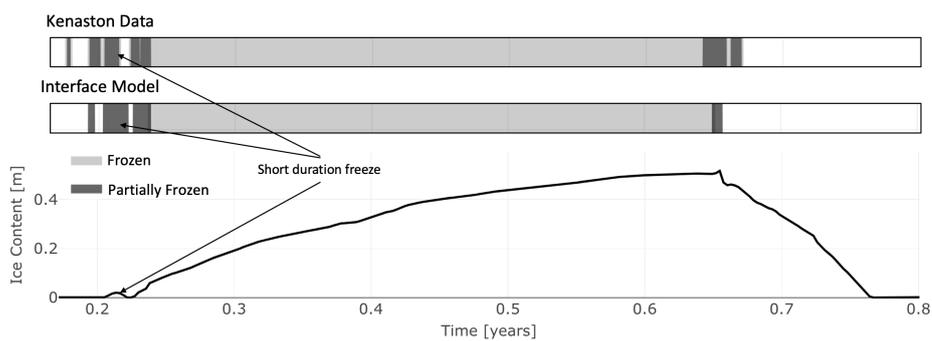


Figure 7. Early season short-duration freeze/thaw event comparison between field-data and interface-model generated freeze/thaw. The overall agreement between freeze/thaw states was 95%, while the interface model correctly identified 96% of the frozen period. Single year of data drawn from 5 year simulation of Kenaston site 10.

416 The interface model is notably better at representing early fall freezing events (Fig-
417 ure 7) which are of much higher hydrological importance as the underlying soil is ice-
418 free and the surface (buffer) layer has the greatest impact on runoff partitioning. These
419 results are promising for their potential improvement to runoff modelling.

420 **5 Conclusion**

421 An interface model was presented to simulate the ice content of variably saturated
422 soils undergoing freeze/thaw processes. This model has been demonstrated to efficiently
423 and stably reproduce the timing and magnitude of freeze/thaw events both on the inter-
424 annual scale as well as on the sub-daily scale when compared to both a high-resolution
425 finite volume model and to data collected at a site in Southern Saskatchewan. The in-
426 terface model fills a utility gap between computationally intensive physically-based con-
427 tinuum models and low-fidelity empirical expressions for ground freeze-thaw, and its com-
428 putational expediency lends itself towards integration into practical forecasting tools. Such
429 a contribution is especially relevant in areas such as the Canadian prairies where an in-
430 crease in midwinter freeze/thaw events of short duration is limiting the predictive abil-
431 ity of current hydrologic models.

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