

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, especially in cold regions. Midwinter thaw events are often the cause of flooding due to the coupled impact of rain-on-snow, 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 soil without having to solve a discrete form of the coupled partial differential equations describing freeze-thaw and soil water content. The model tracks pore ice formation and soil cold content in terms of enthalpy. It is tested against data collected in Southern Saskatchewan and is shown to 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 over-winter streamflow and flooding 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). The prairie region is especially important agriculturally, and the fill-and-spill hydrology of prairie wetlands is highly sensitive to antecedent conditions, changes in precipitation timing and volume, and tends to have a memory extending beyond a single hydrologic year (Shaw et al., 2012; K. R. Shook & Pomeroy, 2011; K. Shook et al., 2013). This behaviour leads to hysteresis in the hydrologic response of these systems, and makes hydrologic prediction challenging (K. R. Shook & Pomeroy, 2011; Shanley & Chalmers, 1999). When coupled with an increase in climate variability and extreme climate events, this complex be-

49 behaviour can lead to significant challenges in predicting streamflow, flooding, groundwa-
50 ter supply and water quality. These challenges are not limited to Prairie and Boreal sys-
51 tems which are the focus of this paper, but other regions that undergo freeze/thaw pro-
52 cesses are affected by these and other modelling challenges.

53 The limitations of current hydrological modelling are, in part, due to a lack of rep-
54 resentation of soil ice content. In many hydrologic models, frozen soils are either treated
55 as strictly impermeable surfaces for the entire winter period (Niu & Yang, 2006) or em-
56 pirical models are used to address the changes in infiltrability due to ice content fluc-
57 tuations over the winter months (Luo et al., 2003). These approaches lead to an inabil-
58 ity to accurately report the soil moisture, thermodynamic state, hydraulic conductivity,
59 infiltrability, and water storage of the systems. In systems that are generally quiescent
60 over the winter months, empirical models of over-winter processes have been found to
61 be adequate (Luo et al., 2003). However, a recent increase in midwinter thaw events and
62 short duration freeze/thaw events in the shoulder seasons make these predictions less and
63 less accurate, to the point where they are insufficient to represent the hydrology of these
64 systems (Pavlovskii et al., 2019). This change is increasingly important as more extreme
65 precipitation, especially rain-on-snow events, can lead to severe flooding. Prediction of
66 flood timing and extent is dependent on infiltrability and hydraulic conductivity of par-
67 tially frozen soils (M. Seyfried & Murdock, 1997). To adequately simulate runoff in hy-
68 drologic models, it is crucial to have a sense of the infiltrability of soils (Luo et al., 2003).
69 The infiltrability is strongly controlled by the ice content of the soils, which in turn is
70 dependant on the freeze/thaw history of the soils. Midwinter melt events are known to
71 introduce ice lenses and layers which impede spring infiltration into frozen soils (Pavlovskii
72 et al., 2019). These melt events result in increased ice content in the near-surface soil
73 which, upon re-freezing, also affects the soil thaw rate in the spring.

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

82 al., 2019). Physically-based thermal models are notoriously demanding computationally,
83 especially when coupled to mass transport of water in soils, and the representation of
84 freezing and thawing often increases computational time more than ten-fold, and can also
85 lead to instabilities and non-convergence of models (Wang et al., 2017). This has mo-
86 tivated the use of empirical models such as that presented by Zhao and Gray (1999), which
87 improve model performance, but are not transferable to other study sites, nor are they
88 applicable in non-stationary systems such as those affected by changing climates.

89 The present work fills a gap in the existing spectrum of modelling tools (physically-
90 based discrete continuum models vs. purely empirical models) by proposing a semi-analytical
91 physical model that efficiently predicts freeze/thaw processes and ice content in soils, fo-
92 cusing on midwinter melt events and short-duration freeze/thaw events in the shoulder
93 seasons that are currently not captured well. Specifically, the objectives of this paper are
94 to (1) extend the method presented in Devoie and Craig (2020) to mineral soils with-
95 out permafrost (2) evaluate the new model against a continuum model benchmark, and
96 (3) apply the model intermittently frozen soil data collected at the Kenaston Field site,
97 with a focus on partial freezing in the near-surface soil.

98 **2 Methods**

99 A combination of two modelling techniques and field-based measurements are used
100 to establish the validity of the proposed interface model for the representation of freeze
101 and thaw events in seasonally frozen mineral soils, especially for short duration midwin-
102 ter melt events.

103 **2.1 Interface Model**

104 The interface model used here is a semi-analytical solution to the heat equation cou-
105 pled to an equilibrium solution to a mass balance relationship based on the van Genuchten
106 pressure-saturation relationship for the specified soil parameters. This interface-based
107 modelling approach, where the location of the frozen-unfrozen interface is treated as a
108 state variable, was first presented in Devoie and Craig (2020) in the context of active layer
109 modelling in discontinuous permafrost peatlands environments. The model was described,
110 benchmarked and validated in that paper, and applied to a specific case of thawing per-
111 mafrost. However, the approach was not applicable to lower porosity mineral soils with

112 low moisture content, did not readily support intermittent freeze/thaw cycles, and pre-
113 sumed the presence of permafrost at depth.

114 [Figure 1 about here.]

115 In this work, the interface model is extended to represent seasonally frozen min-
116 eral soils. The interface model reports the water table position as well as the freeze/thaw
117 fronts that exist in the subsurface (see figure 1). The model also includes a surface “buffer”
118 layer of fixed depth that is allowed to contain fractional ice content, accounting for the
119 near-surface soil behaviour and preventing the non-physical formation of many freeze/thaw
120 interfaces. In this work the buffer layer is taken to be 85 mm, which aligns with the depth
121 of the field measurements used in model validation. To adequately represent unsaturated,
122 seasonally frozen mineral soils, some elements of the original interface model were mod-
123 ified from Devoie and Craig (2020). Most importantly, the bottom boundary condition
124 was modified to reflect the thawed soil. The bottom boundary of the soil profile is fixed
125 at a constant temperature, either 0.1 °C, just above the freezing point for comparison
126 with the continuum model discussed in section 2.2, or a specified temperature based on
127 field measurements. Mineral soils with lower hydraulic conductivity challenged the orig-
128 inal assumption that the water table was in equilibrium, but some modifications were
129 made to water content representation, and this led to adequate results as the model did
130 not seem to be sensitive to small changes in water content (as discussed in section 4).
131 The numerical implementation details and derivation are included in appendix 5.1. The
132 model structure is otherwise unchanged from Devoie and Craig (2020), though the for-
133 mulation of soil layers was modified to accommodate seasonal freeze/thaw cycles con-
134 gruent with the system shown in figure 1.

135 For comparison with field data, the model domain was extended to a depth of 15
136 m and a fixed soil temperature of 5 °C was prescribed at the base of the profile, consis-
137 tent with data collected near Edmonton, Canada (Toogood, 1976). An initial water ta-
138 ble position was assigned at 1 m below the ground surface, based on soil moisture data
139 collected in the field. A no-flow boundary condition was assigned at the base of the soil
140 column to represent the near-impermeable bedrock underlying this system. The surface
141 temperature boundary condition was drawn from soil temperature collected at a depth
142 of 5 cm in the field sites near Kenaston, Canada, and forced with a seasonally cyclic mois-
143 ture boundary condition (reported in section 3) as direct application of the infiltration

144 flux data collected in the field precluded convergence of the continuum model used for
145 benchmarking. The soil column was initialized to a thawed uniform temperature of 5 °C,
146 and the freeze/thaw discriminant temperature was assigned based on the specific freez-
147 ing point depression determined from the field data, ranging between 0 and -0.4 °C (Pardo Lara
148 et al., 2020). Simulations were started in the summer of 2012, except sites 16 and 18 which
149 were started in summer 2013 due to lack of data. As described in Devoie and Craig (2020),
150 the surface layer of the interface model is a ‘buffer layer’ which may contain fractional
151 ice content. The depth of this buffer layer was assigned based on the zone of influence
152 of the soil moisture measurements made in the field. This allows the ice content of the
153 buffer layer to be compared to the measured ice content of the near-surface soil in the
154 field. Below the buffer layer the freeze/thaw front is a moving sharp interface and frac-
155 tional ice content is not permitted. Because of the moving interface, there is no spatial
156 discretization of the interface model, however there is temporal discretization, and the
157 simulations reported here are run with a 1 hour timestep for comparison with the finite
158 volume model in figure 9 and a 1 day timestep otherwise. Other soil parameters were
159 homogeneous and independent of depth, and are summarized in table 5.2 in Appendix
160 5.2 both for organic and mineral soils.

161 2.2 Continuum Model

162 The interface model above was directly compared to a coupled solution of the un-
163 saturated Richards’ equation and the energy balance equation solved via a finite volume
164 method with operator splitting, as discussed in Devoie et al. (2019). This solution al-
165 lows us to assess the impact of the simplifying assumptions made in the interface model
166 while being forced with identical initial and boundary conditions, as well as model rep-
167 resentations of soil properties, pressure saturation relations, soil freezing characteristic
168 curves and model domains. The comparison here is meant to ensure that the interface
169 model adequately represents the physics of the system. Identical initial conditions and
170 boundary conditions were used in this model, and a spatial discretization of 1 cm and
171 2 cm were compared, both for 1 hour time steps. The same soil parameters were used
172 for this model as were used in the interface model, with the addition of a linear soil freez-
173 ing characteristic curve for a freezing range of -0.005 to 0 °C.

174 **2.3 Kenaston Data-driven Model**

175 A field-based approach to determining the frozen or thawed state of the soil was
176 used to generate validation data for the interface model discussed above. This approach
177 uses soil moisture (permittivity) and temperature data to establish a site-specific freez-
178 ing point depression and uncertainty range for both freezing and thawing. The freezing
179 temperatures were estimated using a logistic growth model fit to the soil freezing curve,
180 as detailed in Pardo Lara et al. (2020). This allowed us to consistently estimate when
181 the soil is thawed, frozen, or undergoing phase change based upon the observation data.
182 These data were used to validate the predicted freeze/thaw status from the interface model
183 by specifying the field-data based freeze/thaw flag.

184 **2.3.1 Field Data**

185 Soil moisture, temperature and precipitation have been monitored at 22 stations
186 of the Kenaston Network located in the Brightwater Creek basin, east of Kenaston, SK,
187 Canada (Tetlock et al., 2019). This is predominantly an agricultural region, dominated
188 by annually cropped fields with some grazing land and without irrigation (Tetlock et al.,
189 2019). The instrumented monitoring network spans 40 km², with most of the instrumen-
190 tation within a flat 10 km² sub-region with slopes of less than 2%. The sites cover a soil
191 textural composition of 10.5 - 61.7 % sand, 31.2 - 72.4 % silt and 1.2 - 41.1 % clay, for
192 the base computational test, a representative soil (from Kenaston site 1) of 28 % sand,
193 53 % silt and 19 % clay was used (Pardo Lara et al., 2020). The mean annual air tem-
194 perature in this region is 8 °C, and in the last three decades the mean annual precip-
195 itation has been 400 mm of which approximately 30% falls as snow (Meteorological Ser-
196 vice of Canada, 2012). The catchment is semi-arid, and fluctuations in soil moisture fol-
197 low a seasonal pattern (Burns et al., 2016), though some fill-and-spill and non-contributing
198 areas are documented where water ponds in sloughs instead of contributing to the basin
199 outflow (K. Shook et al., 2013).

200 Soil moisture was measured using “HydraProbes”, commercially available electro-
201 magnetic sensors which report permittivity (M. S. Seyfried & Murdock, 2004). The sen-
202 sors have 4 metal tines which are 3 mm in diameter and 57 mm long. The zone of in-
203 fluence of the probe ranges approximately from 4.0×10^4 mm³ to 3.5×10^5 mm³, with
204 a radial range of approximately 13 to 35 mm (Pardo Lara et al., 2021). Given the in-

205 stallation of these probes, it is assumed that they are sensitive to water and ice content
206 in the top 50 ± 35 mm of soil, and this near-surface layer is used to report the frozen,
207 thawed, or transitioning state of the soil. A depth of 85 mm was assigned to the buffer
208 in the interface model to correspond to this near-surface layer for model validation. Soil
209 temperature was measured alongside soil moisture (as part of the soil moisture measure-
210 ment) at three depths: 5, 20 and 50 cm below the ground surface (Burns et al., 2016).
211 Precipitation was also measured at each site using tipping bucket rain gauges. All data
212 was collected at 30-minute intervals (Tetlock et al., 2019).

213 **3 Results**

214 The interface model presented in (Devoie & Craig, 2020) is extended to treat the
215 case of seasonal ground ice, enabling it to represent seasonal freeze thaw in mid-latitude
216 continental climates. Here, the simulation of seasonal freeze-thaw is first verified via a
217 numerical benchmarking study in a near-saturated peat soil column, followed by another
218 purely numerical comparison in an unsaturated system more typical of mineral soils in
219 the semi-arid climate of Southern Saskatchewan. In both cases, boundary conditions and
220 soil parameters were obtained from field data, but no direct measurements of soil mois-
221 ture or freeze/thaw status are available; these tests are purely to demonstrate numer-
222 ical accuracy of the method. Finally, the interface model predictions are compared to
223 the data-derived freeze/thaw status at sites in the Canadian prairies in order to eval-
224 uate the practical efficacy of the method.

225 **3.1 Near-Saturated peat soils**

226 This first test of the accuracy of the interface model was a comparison of the in-
227 terface model to the benchmarked continuum model simulates a near-saturated peat soil
228 column with a soil porosity of 0.8. The bottom boundary was fixed at a temperature of
229 0.1 °C at a depth of 3 m, and the profile was initially treated as thawed below the base
230 of the active layer. The surface boundary condition was drawn from field data collected
231 in the Scotty Creek Research Basin (Quinton et al., 2019), and is consistent with soil sur-
232 face temperatures of peat soils in a cold region. This data and details on model param-
233 eterization can be found in Devoie and Craig (2020), and specific peat soil characteris-
234 tics are included in table 5.2 in the Appendix. As seen in figure 2, the model agreement
235 is good with respect to water saturation, total integrated ice content (RMSE of 0.14 be-

236 tween ice content of interface and continuum models) and location of the freeze/thaw
237 interface. The largest discernible discrepancy between the models being that the inter-
238 face model freezes slightly deeper than the continuum model at the end of the freezing
239 season (see figure 6), indicating that perhaps the assumption of a linear (equilibrium)
240 temperature profile in the bottom-most layer may not be adequate to fully represent this
241 model domain.

242 [Figure 2 about here.]

243 **3.2 Unsaturated conditions**

244 Many temperate mineral soils maintain a water table between 0.25 and 5 m from
245 the ground surface (Fan et al., 2013). The interface model was therefore compared to
246 the continuum model for unsaturated mineral soil conditions where the water table was
247 initially at 2 m below the ground surface as seen in Figure 3. Specific mineral soil char-
248 acteristics were drawn from Kurylyk et al. (2014), and are included in table 5.2 in the
249 Appendix. Again, the agreement between continuum and interface model is generally
250 of high quality, without the deeper freezing anomaly seen in the saturated peat soil test
251 case, improving the RMSE between the ice content of the interface model and contin-
252 uum model to 0.004. This improved agreement is likely due to the increased diffusivity
253 of the soil matrix.

254 [Figure 3 about here.]

255 **3.3 Kenaston**

256 The model was evaluated for a five-year simulation based on field data collected
257 at one field site of the Kenaston Soil Moisture Network, with a 15 m vertical domain and
258 realistic thermal initial and boundary conditions as detailed in section 2.1. Figure 4 shows
259 the comparison between the continuum model, interface model, as well as field data in-
260 dicated the ‘frozen period’ (shaded in grey). The shaded grey areas in figure 4 (b) in-
261 dicate the period over which the near-surface soil (approximately 40 - 85 mm) at the field
262 site was frozen. This data is drawn directly from field measurements using the method-
263 ology outlined in section 2.3.1, and compares favourably with the reported freeze/thaw
264 timing. The use of field data resulted in an increase in RMSE to 0.04, which is still ex-
265 cellent agreement. The simulation was re-run using finer discretization in the continuum
266 model (shown in figure 5) to capture the exact timing of a specific freeze/thaw event.

267 [Figure 4 about here.]

268 [Figure 5 about here.]

269 The comparison of the interface and continuum model for the short-duration event
270 in figure 5 was generated using the same model configuration as figure 4, but with finer
271 spatial and temporal discretization of both models. The comparison of computational

272 efficiency can also be established in figure 5 as the continuum model run took 2 hours
273 and 22 minutes (in blue) while the interface model (red) only took 4.5 seconds for the
274 same size timestep and simulation setup. The model performance of the interface model
275 is arguably better than the continuum model: when the spatial step of the continuum
276 model is refined, it tends toward the interface model solution. Larger spatial steps lead
277 to a lack of convergence in the continuum model, and smaller temporal discretization
278 was computationally impractical. The interface model also shows better timing and more
279 gradual and physical response to freeze/thaw events. Neither model captures the initial
280 freezing event near day 7, likely due to the choice of freezing point depression (-0.005°
281 C) and the freezing range between 0 and -0.01° C for the interface and continuum mod-
282 els respectively. Subsequent figures generated using only the interface model without con-
283 tinuum model comparison use the freezing point depression determined from field mea-
284 surements at the given field sites in order to better capture such events.

285 **3.4 Midwinter Melt**

286 The benchmarked interface model (but not the continuum model) was then applied
287 to simulate all of the available data for similar mineral soil sites. A total of 22 sites were
288 considered in which subsurface temperature and soil moisture were recorded for a du-
289 ration of 4 - 6 years between 2014 and 2020. In 10 of these 22 sites clear mid-winter thaw
290 events were identified. The interface model was run using near-surface soil temperature
291 data available at these sites, and compared to the freeze/thaw flag extrapolated from the
292 field data. Here a second “transition” flag was added to the field data representing soils
293 undergoing phase change; if the surface layer of soil contained fractional ice content based
294 on its permitivity this flag was activated. This flag is shaded in dark grey in the sub-
295 sequent figures, while entirely frozen near-surface soils (with only residual water content)
296 were assigned a “frozen” flag, depicted in light grey and thawed near-surface soils were
297 left as white bands. The interface model was compared to the two field-data based flags
298 using the near-surface “buffer” layer in the model. The depth of this surface soil layer
299 is 85 mm in accordance with the (maximal) sensitivity of the soil moisture probes used
300 to collect the field data (Pardo Lara et al., 2021). Two separate flags were also imple-
301 mented in the model - the first “transition” flag representing fractional ice content in the
302 near-surface, and the second “frozen” flag indicating residual water content only, these
303 are assigned the same colours as the field data. Example results for the entire 5 year sim-

304 ulation at Kenaston site 3 are shown in figure 6, showing agreement between the inter-
305 face model and data-extrapolated freeze/thaw timing. Two error metrics are used to com-
306 pare the simulated and observed near-surface ice content. The first indicates the over-
307 all agreement between the modelled and measured data including frozen, thawed and tran-
308 sitioning states. For the data in figure 6, the agreement is 92%, indicating that the soils
309 did not have the same freeze/thaw state only 8% of the time. The second metric was con-
310 ceived to identify the effectiveness of the interface model at identifying frozen soils, and
311 so it compares the soil state only when the measured field data is frozen, and does not
312 take into account partially vs. completely frozen soils. For this study case, there is 91%
313 agreement, indicating that the interface model incorrectly identified frozen soil as thawed
314 9% of the time.

315 [Figure 6 about here.]

316 4 Discussion

317 Figure 6 demonstrates agreement in the timing of broad seasonal events between
318 the modelled data and data collected in the field, and figure 7 shows a more detailed view
319 that distinguishes the typical seasonal freeze/thaw (i.e. freeze in the fall/early winter and
320 thaw in the spring) from midwinter melt events. The interface model is highly effective
321 in detecting the timing of freeze/thaw initiation, however the freeze/thaw transitions of
322 the near-surface layer tend to occur sooner than in the measured data (Figure 7 & 8),
323 perhaps due to an under-estimate of the water content (and hence effective heat capac-
324 ity) of the soil, alternatively an under-estimate or mismatch of the depth of influence of
325 soil water content on the measurements made in the field. These explanations are also
326 supported by the tendency of the interface model to exit and enter the phase change state
327 more rapidly. An under-estimate or mismatch of the volume which must undergo phase
328 change due to an under-estimate of the near-surface layer would result in more rapid freeze/thaw.
329 It is also noteworthy that the total ice content in the soil column changes very little due
330 to these short-duration freeze/thaw events. Generally we see a flattening of the slope dur-
331 ing a midwinter melt (e.g. figures 7 and 8), where ice accumulation does not occur, how-
332 ever there is no clear evidence for significant ice loss during these events. It is difficult
333 to establish the measured extent of thaw from the available field data, but there is no
334 evidence that thaw extends beyond the first soil moisture and temperature sensor at a

335 depth of 50 mm, limiting the anticipated ice loss to less than 25 mm given unsaturated
336 soil conditions and a soil porosity not exceeding 0.5.

337 [Figure 7 about here.]

338 Figures 7 and 8 also demonstrate that thaw occurs sooner in the interface model
339 than in the extrapolated field data, though the interface model does accurately capture
340 96% of the frozen data. It is thought that this is due to the single freezing point depres-
341 sion that is assigned to the data. It is known that there is hysteresis in the freeze-thaw
342 process, and that the freezing point is generally lower than the thawing point (Saberi
343 & Meschke, 2021). This leads to more rapid modelled thaw as it is initiated at a colder
344 temperature than would realistically be observed in the field. More work including hys-
345 teretic behaviour in freeze/thaw modelling is needed.

346 [Figure 8 about here.]

347 The small difference in freeze/thaw timing may also be driven by a mismatch in
348 near-surface soil water content. The error in estimated soil water content may arise be-
349 cause an equilibrium soil moisture profile is implemented in the interface model, as de-
350 tailed in Appendix 5.1. The equilibrium assumption was first established for near-saturated
351 peat soils in which infiltration events were expected to be rapid due to the high hydraulic
352 conductivity of these soils. The transition to mineral soils was expected to require a more
353 complex representation of infiltration events including plug flow and moisture redistri-
354 bution, but these were not found to be necessary in the reproduction of the freeze/thaw
355 conditions in field observations of near-surface soils. The equilibrium assumption seems
356 to be adequate for two reasons; first, the surface mass balance used is based on seasonal
357 trends and is very smooth. This results in near-equilibrium moisture conditions in the
358 soil column over most of the freeze/thaw season. This boundary condition was chosen
359 for convergence reasons in the continuum model. Secondly, the quantity of interest is the
360 frozen state of the near-surface soil. When freezing is occurring, the impedance of ice
361 in the soil pores is such that infiltration and evapotranspiration are negligible, and there-
362 fore these processes have little effect on the model results.

363 Measurements of spring thaw (and some midwinter events) lead to small and rapid
364 fluctuations in ice content in the surface layer. Spring temperatures in the Kenaston re-

365 gion have strong diurnal fluctuations, where the daytime temperature is well above the
366 freezing point, but the overnight low is around $-1\text{ }^{\circ}\text{C}$. In the interface model, the near-
367 surface ice content is estimated in the top 85 mm of soil, deemed equivalent to the depth
368 of soil characterized by the field based freeze-thaw flag. This layer was included in the
369 model as a mathematical construct that would prevent the formation of very thin, non-
370 physical frozen and thawed layers at the soil surface. Even with this layer, the interface
371 model fails to capture many diurnal-fluctuation driven spring freeze/thaw events. How-
372 ever, these occur when the underlying soil is frozen, and so the inability to track frac-
373 tional ice content in the near-surface soil (especially when the ice content never freezes
374 the pore water completely) likely has very little effect on the hydrology of the system.
375 Water movement in the landscape is expected to be much more strongly affected by the
376 fully frozen near-saturated layer at a depth of 10 - 15 cm below the soil surface. The rel-
377 atively thin surface layer cannot store significant thermal energy, and the surface topog-
378 raphy generally exceeds the scale of this layer, restricting the formation of flow pathways
379 beyond the plot scale. The buffer layer may however still be meaningful physically speak-
380 ing, as there is evidence for the development of a surface layer symmetric to the buffer
381 layer concept in a soil subject to midwinter thaw events. As noted by the temperature
382 sensors in the soil profile, short thaw events do not extend beyond the top 100 mm of
383 soil, though this surface layer experiences temperature cycling and freeze/thaw through-
384 out the winter as well as the shoulder seasons when strong diurnal temperature cycles
385 are common. The increased freeze-thaw cycling can lead to changes in soil structure (Alkire
386 & Morrison, 1983) and changes in decomposition of soil organic matter (Yanai et al., 2004).
387 Further investigation is required to establish if this layer is physically significant across
388 landscapes experiencing freeze-thaw.

389 [Figure 9 about here.]

390 The interface model is notably better at representing early fall freezing events (Fig-
391 ure 9) which are of much higher hydrological importance as the underlying soil is ice-
392 free and the surface ice layer has the greatest impact on runoff partitioning. These re-
393 sults are promising for their potential improvement to hydrologic modelling.

5 Conclusion

An interface model is presented that simulates the ice content of variably saturated soils undergoing freeze/thaw processes. This model has been demonstrated to efficiently and stably reproduce the timing and magnitude of freeze/thaw events both on the inter-annual scale as well as on the sub-daily scale when compared to both a high-resolution finite volume model and to data collected at a site in Southern Saskatchewan. The interface model fills a utility gap between computationally intensive physically-based continuum models and low-fidelity empirical expressions for ground freeze-thaw, and its computational expediency lends itself towards integration into practical forecasting tools. Such a contribution is especially relevant in areas such as the Canadian prairies where an increase in midwinter freeze/thaw events of short duration is limiting the predictive ability of current hydrologic models.

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The data that support the findings of this study are openly available in the Federated Research Repository at <https://doi.org/10.20383/101.0116>.

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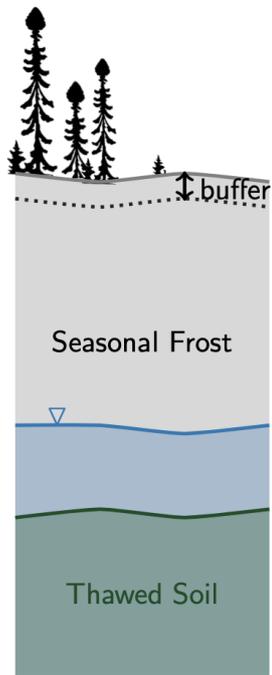


Figure 1. Schematic diagram showing model domain. Interface model tracks the buffer layer (where fractional ice content is permitted) and the interface between frozen and thawed soil. The water table is also computed separately and updated through an equilibrium mass balance.

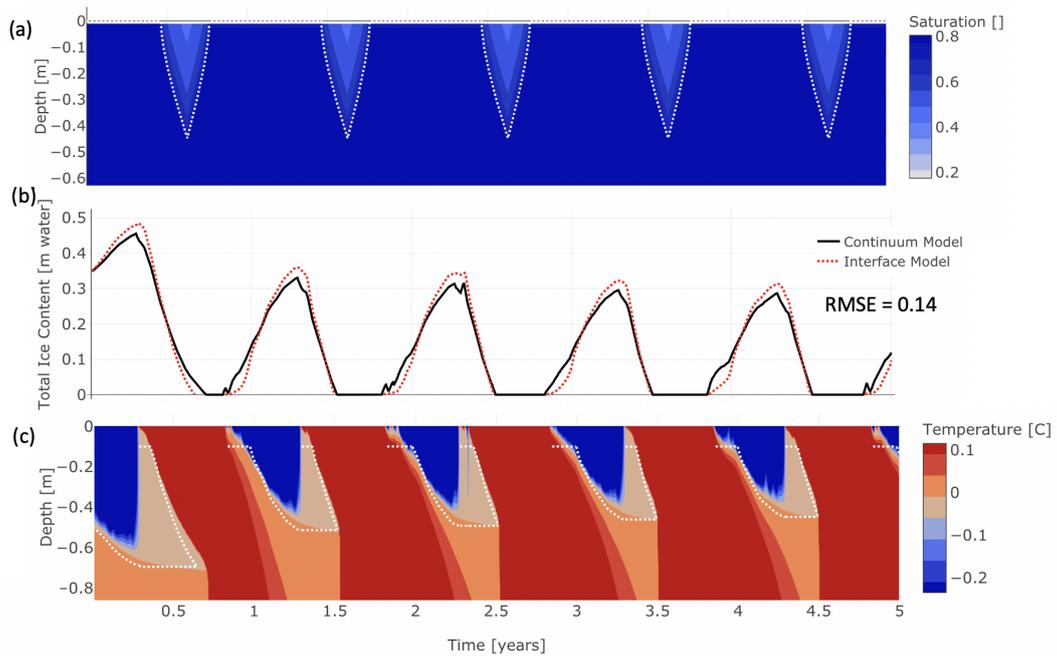


Figure 2. (a) Contour plot of continuum model saturation with water table position of interface model superimposed; (b) comparison of total ice content for each model and (c) contour plot of continuum model temperature with interface position (dashed white line) superimposed.

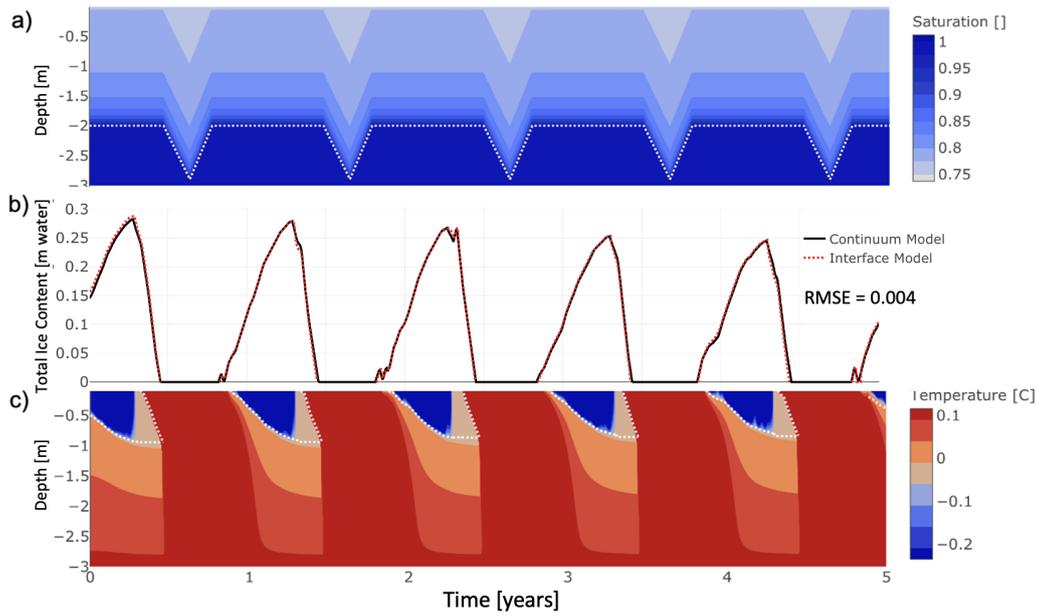


Figure 3. (a) Contour plot of continuum model saturation with water table position of interface model superimposed for mineral soils (b) comparison of total ice content for each model and (c) contour plot of continuum model temperature with interface position superimposed all for mineral soils. Depths are relative to the ground surface.

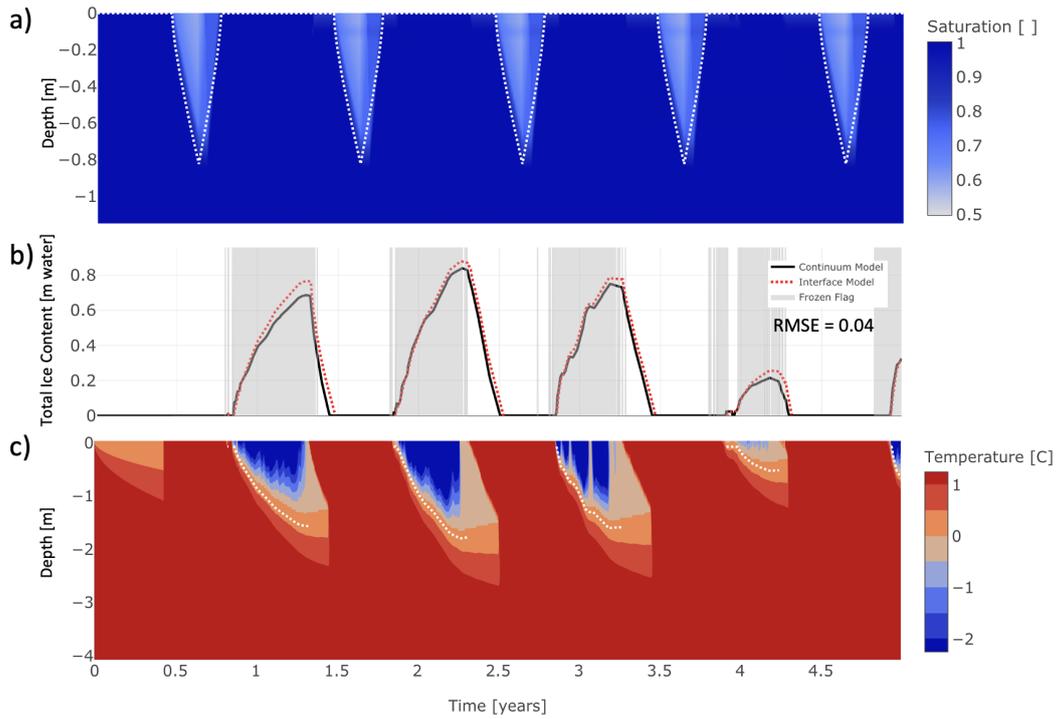


Figure 4. (a) Comparison of water content for interface and continuum model (b) comparison of total ice content for continuum model (black), interface model (red) and field-based near-surface frozen flag (shaded grey) and (c) contour plot of continuum model temperature with freeze/thaw interface position from interface model superimposed. 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 5.2 of Appendix

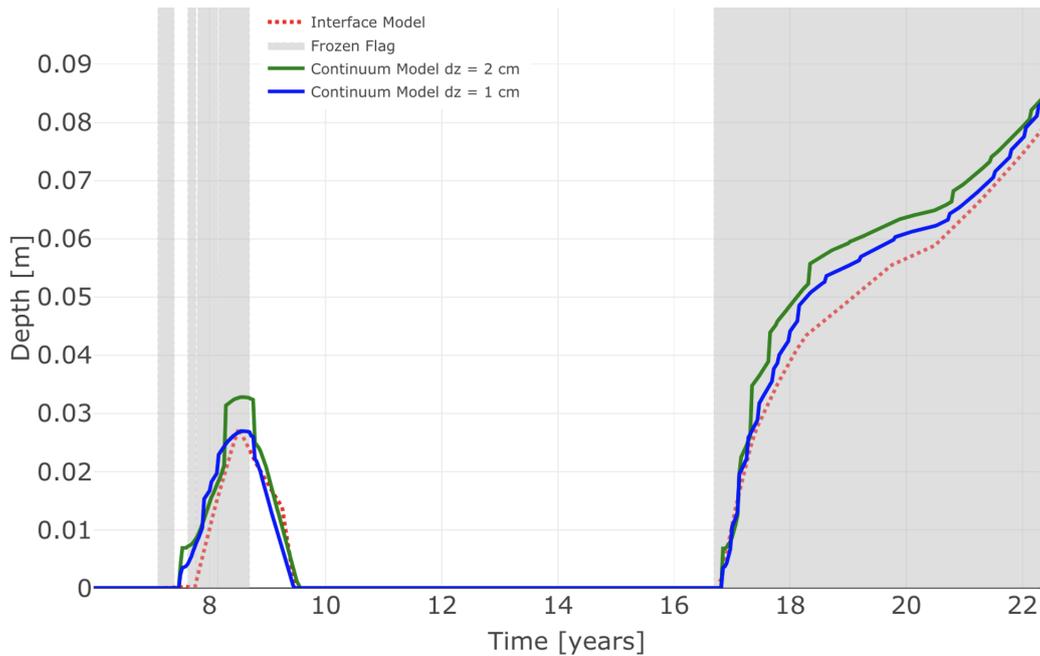


Figure 5. Short duration freeze/thaw initiation. Comparison of interface model (with 6 hour timesteps) to continuum model with varying timesteps. This short duration initiation of freezing results in a small quantity of near-surface ice, hence the small total ice content. Simulation duration for continuum model with 1 cm spatial discretization was 2 hours and 22 minutes, while the interface model ran in 4.5s. Continuum model does not converge for larger spatial or temporal steps than those shown. Grey shaded region indicates soil freezing according to the field-data based flag. Soil texture data drawn from Kenaston Site 1 in table 5.2 of Appendix

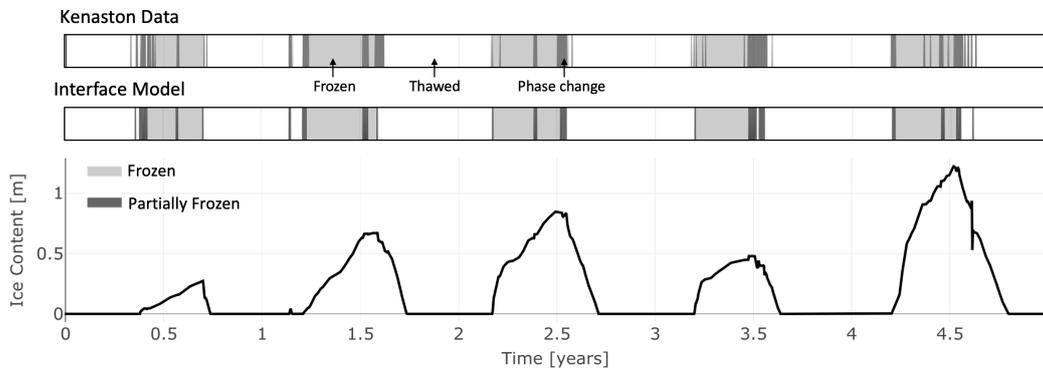


Figure 6. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). Total ice content from interface model shown along bottom axis. Soil texture data drawn from Kenaston Site 3 in table 5.2 of Appendix

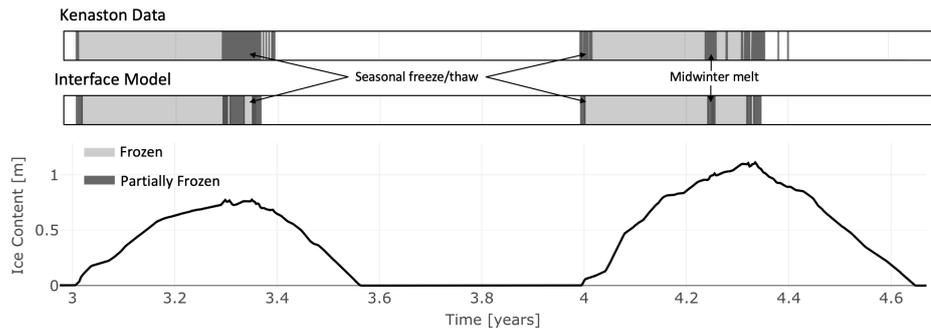


Figure 7. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). This 2 year subset from 5 year simulation drawn from Kenaston site 15. Seasonal freeze thaw at the near surface occurs in fall and early 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.

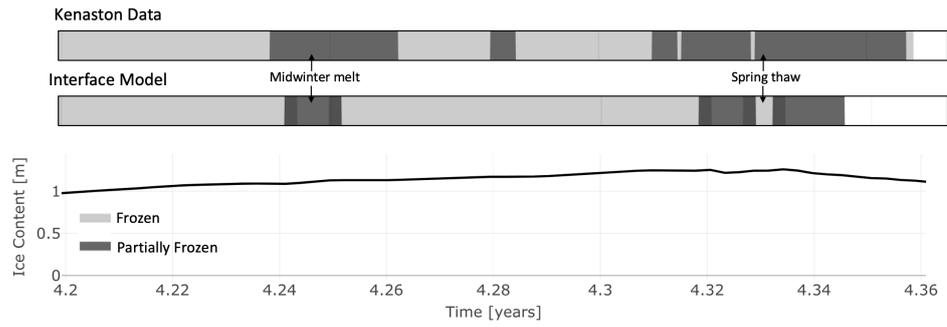


Figure 8. Comparison of Kenaston field-data and interface model generated near-surface ice content indicating phase change (dark grey) and frozen soil (light grey). Interface model phase change takes less time, perhaps because of an under-estimate of the freezing point. Overall agreement between the 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.

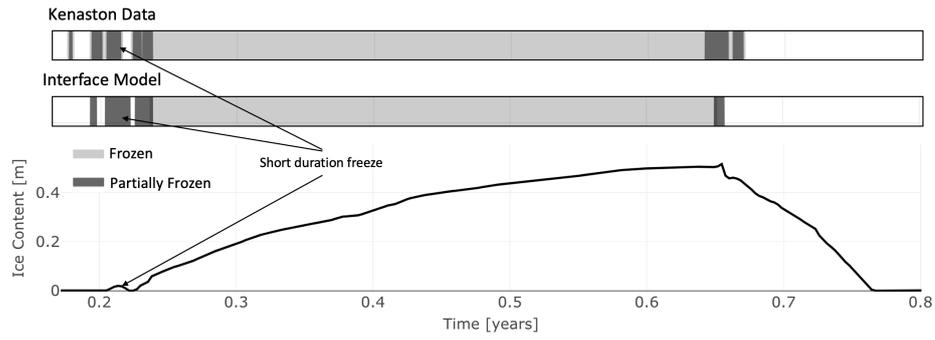


Figure 9. Early season short-duration freeze/thaw event comparison between field-data and interface-model generated freeze/thaw. The overall agreement between freeze/thaw states was 95%, while the interface model correctly identified 96% of the frozen period. Single year of data drawn from 5 year simulation of Kenaston site 10.