Characterizing Sub-Glacial Hydrology Using Radar Simulations

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Abstract. The structure and distribution of sub-glacial water directly influences Antarctic ice mass loss by reducing or enhancing basal shear stress, and accelerating grounding line retreat. A common technique for detecting sub-glacial water involves analyzing the spatial variation in reflectivity from an airborne radar echo sounding (RES) survey. Basic RES analysis exploits the high dielectric contrast between water and most other substrate materials, where a reflectivity increase $\geq 15\,dB$ is frequently correlated with the presence of sub-glacial water. There are surprisingly few additional tools to further characterize the size, shape, or extent of hydrological systems beneath large ice masses.

We adapted an existing radar backscattering simulator to model RES reflections from sub-glacial water structures using the University of Texas Institute for Geophysics (UTIG) Multifrequency Airborne Radar Sounder with Full-phase Assessment (MARFA) instrument. Our series of hypothetical simulation cases modeled water structures from $5\,m$ to $50\,m$ wide, surrounded by bed materials of varying roughness. We compared the relative reflectivity from rounded Röthlisberger channels and specular flat canals, showing both types of channels exhibit a positive correlation between size and reflectivity. Large ($> 20\,m$), flat canals can increase reflectivity by more than $20\,dB$, while equivalent Röthlisberger channels show only modest reflectivity gains of $8 - 13\,dB$. Changes in substrate roughness may also alter observed reflectivity by $3 - 6\,dB$. All of these results indicate that a sophisticated approach to RES interpretation can be useful in constraining the size and shape of sub-glacial water features. However, a highly nuanced treatment of the geometric context is necessary.

Finally, we compared simulated outputs to actual reflectivity from a single RES flight line collected over Thwaites Glacier in 2022. The flight line crosses a previously proposed Röthlisberger channel route, with an obvious bright bed reflection in the radargram. Through multiple simulations comparing various water system geometries, such as canals and sub-glacial lakes, we demonstrated the important role that topography and water geometry can play in observed RES reflectivity. From the scenarios that we tested, we concluded the bright reflector from our RES flight line cannot be a Röthlisberger channel, but could be consistent with a series of flat canals or a sub-glacial lake. However, we note our simulations were not exhaustive of all possible sub-glacial water configurations.

The approach outlined here has broad applicability for studying the basal environment of large glaciers. We expect to apply this technique when constraining the geometry and extent of many sub-glacial hydrologic structures in the future. Further

25 research may also include comprehensive investigations of the impact of sub-glacial roughness, substrate heterogeneity, and computational efficiencies enabling more complex and complete simulations.

1 Introduction

The size, shape, and distribution of sub-glacial water is important to ice dynamics, and remains a significant uncertainty in projecting sea level rise due to ice mass loss. Hydrological structures directly influence basal shear stress distribution, which

30 defines the boundary condition for rheology at the ice/bed interface (Gilbert et al., 2022; Brinkerhoff et al., 2021; Hoffman et al., 2016). Widely distributed water has been shown to lubricate the base and weaken sediments (Dunse et al., 2015; Hoffman et al., 2016). Conversely, narrow water channels may have little impact on shear stress at the bed (Schroeder et al., 2013), but act as conduits for concentrating meltwater produced upglacier. The size and location of such channels can directly influence grounding line retreat by controlling the volume of water transported to the grounding line (Young et al., 2016; Wright et al.,

35 2012; Schroeder et al., 2013).

Airborne Radar Echo Sounding (RES) is an established technique for studying sub-glacial hydrology throughout Earth's cryosphere (Peters et al., 2005; Chu et al., 2016; Schroeder et al., 2013). Attenuation loss through ice at common radar wavelengths ($\sim 2-5m$) is relatively low, enabling reliable imaging of bed surfaces beneath ice masses several thousand meters thick. The dielectric contrast at an ice / water interface is much higher than at an ice / rock interface, resulting in higher

40 bed reflectivity when water is present. Additionally, some ice / water interfaces may be relatively smooth, depending on the geometry of the hydrological structure. These properties are commonly exploited in RES surveys to infer the location of sub-glacial water, based on reflected power $>\sim 15\,dB$ higher than the surrounding area (Schroeder et al., 2015; Peters et al., 2005; Rutishauser et al., 2018; Schroeder et al., 2013; Young et al., 2016).

Deducing the presence of water beneath large ice sheets using RES is relatively common, however methods for testing

45 further hypotheses regarding size, geometry, or hydrological structure remain challenging. Direct observation of the glacier bed over any significant spatial extent is infeasible with current methods (e.g. drilling (Priscu et al., 2021)), limiting our ability to calibrate radar returns to observed hydrological features.

This problem has precedent in inter-planetary science, where radar experiments are designed to test hypotheses with limited in situ evidence about surface or sub-surface characteristics. Backscattering simulators have proven especially useful in

50 modeling radar returns for celestial targets (Spagnuolo et al., 2011; Russo et al., 2008; Gerekos et al., 2018). Most simulators approximate a target surface as a series of flat facets acting as point backscatterers. The backscattered electric field strength at the radar antenna position is estimated with common mathematical approximations such as the Stratton-Chu integral. In such a point-scatterer formulation, facets much smaller than the radar wavelength (typically $< \lambda/10$) are required to approximate real world instrument results (Gerekos et al., 2018). This constraint necessitates access to high-end computing resources, often

55 making point-scattering radar simulators unrealistic for RES modeling.

However, as algorithms improve and computational costs decrease, it is increasingly attractive to attempt such a simulation method with ice penetrating RES problems. Gerekos et al. (2018) described a simulation technique that is particularly intriguing

for the study of sub-glacial hydrology. The methodology is unique in two distinct ways from other simulators, which are helpful in modeling ice penetrating RES. First, the simulator can estimate strength and direction of signals transmitted through multiple
60   layered material interfaces. This makes it conducive to targets such as the ice / bedrock system. Second, the algorithm allows phase to vary linearly across the facet (termed the Linear Phase Approximation or LPA). This feature enables modeling with significantly larger facets ($\sim \lambda$ or larger), drastically reducing computational resources needed for accurate simulations.

This paper demonstrates the radar simulator's application to geometric scattering from common hydrological targets: flat canals and Röthlisberger channels. We discuss the relevant parameters to achieve accurate model results, and illustrate its utility
65   in interpreting radar signatures from hydrological features beneath the ice. Finally, as an example, we demonstrate simulations of a hydrological target beneath Thwaites Glacier. In the future, we anticipate the technique will have broad applicability to interpreting RES data in the context of the sub-glacial environment.

2   Simulation Methodology

Figure 1 describes a generic conceptual model for a RES radar simulation. Parameters were chosen to emulate a typical
70   helicopter-based airborne radar survey with the University of Texas Institute for Geophysics (UTIG) Multifrequency Airborne Radar Sounder with Full-phase Assessment (MARFA) instrument (Castelletti et al., 2017; Lindzey et al., 2020). An ice surface and a bed surface are defined in a 3-dimensional Cartesian coordinate system, noted as S_{ice} and S_{bed} respectively. Figure 1a illustrates a 2-D representation of these surfaces, divided into N facets using a Delaunay triangulation algorithm, with characteristic length l_f. The linear phase approximation (LPA) employed in Gerekos et al. (2018) allows the phase of the
75   incident and reflected electric fields to vary linearly across each facet. LPA enables accurate simulations of coherent radar using relatively large facets ($\sim \lambda$ vs. $\lambda/10$ with other methods). The facet size should be constrained according to Eq. 1, where h is the aircraft height and λ is the free space radar wavelength (Gerekos et al., 2018). For simulations presented here, $h = 500m$ and $\lambda = 5m$, consistent with a typical UTIG MARFA helicopter experiment. $l_f = 5m$ was chosen, consistent with Eq. 1.

80   $$l_f \leq 0.2\sqrt{\lambda h/2} \tag{1}$$

The radar's spherical wavefront is simulated as a series of plane waves, with wavevectors $\bar{k}_i$ and field strength vector $\bar{E}_i$ directed at each ice surface facet within a radius R beneath the aircraft (Fig. 1b). A wider R will provide a more complete approximation of the radiated and returned electric field, but comes at a computational cost proportional to R^2. Therefore, R is chosen to balance these competing priorities, as we will discuss in detail.
85   Each ray's path is traced through transmission at S_{ice} using Snell's Law, then reflection from S_{bed} (Fig. 1c). In Fig. 1, the reflection from the ice surface is omitted for simplicity, since we are most concerned with reflections from the bed in this study. Reflected and transmitted field strengths ($\bar{E}_t$ and $\bar{E}_r$) are calculated from the real component of the dielectric constant at S_{ice} and S_{bed}, as discussed below. Attenuation loss between the surfaces is dependent on the complex component of the dielectric constant for ice (Gerekos et al., 2020), which is assumed constant over the short simulation distances presented here.
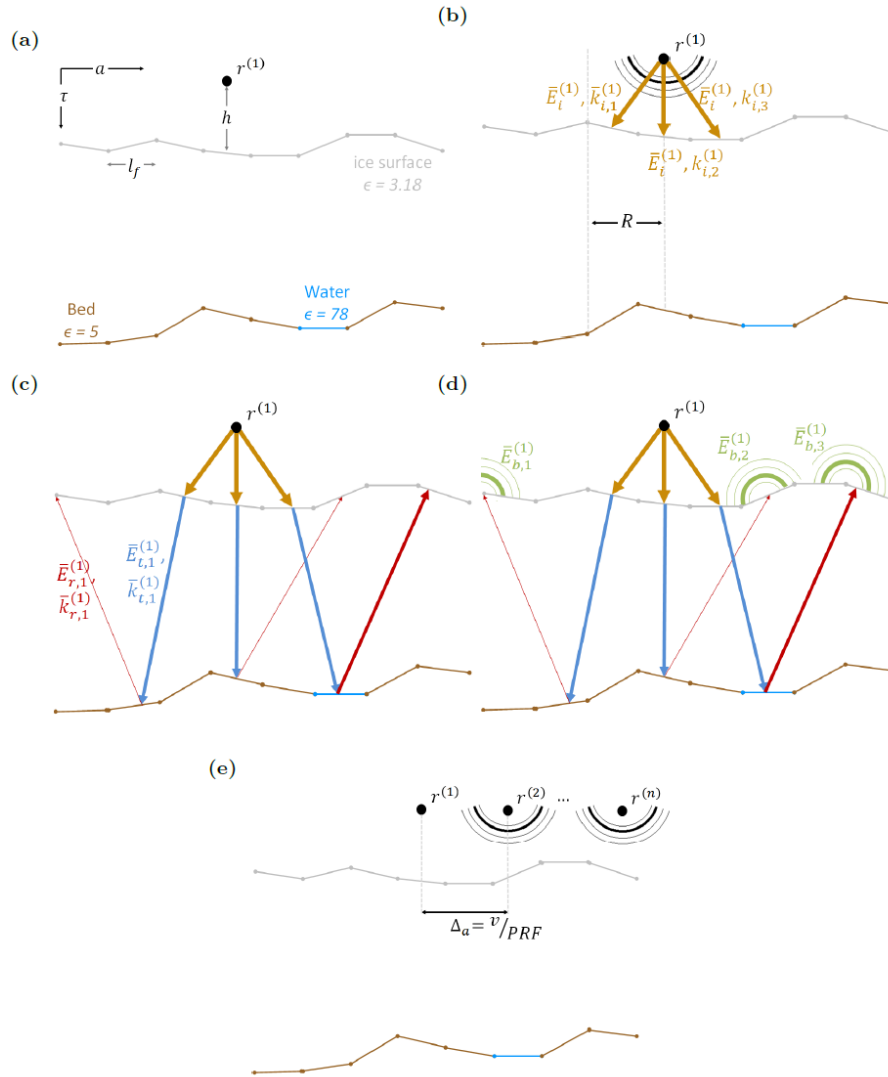
Figure 1. 2-D Conceptual model of a RES simulation, defined in the direction of flight: a) The material and geometric model for a radar observation. b) The radar's spherical wavefront is approximated as rays directed at facets within a footprint beneath the aircraft. The simulator calculates a reflected and transmitted field strength, based on the dielectric constant at the surface (ϵ_{ice}). We do not show the surface reflection for simplicity. c) The rays propagate through the ice, with a direction obeying Snell's Law. At the bed, reflected field strength and direction are controlled by dielectric contrast and incidence angle with the basal facet. d) The backscattered electric field strength (E_b) is approximated with the Stratton-Chu integral. e) The aircraft's position is incremented and the process is repeated.

90   The total field at the radar antenna is then approximated as a summation of the backscattered electric fields ($\bar{E}_b$) from individual propagated waves using the Stratton-Chu integral (Fig. 1d). A detailed treatment of the simulation mathematics is

Table 1. Summary of RES simulation parameters. Parameters were chosen to replicate helicopter-based MARFA ice sounding experiments.

Parameter	MARFA Value*
Geometric	
Aircraft height (h)	$500\,m$
Aircraft velocity (v)	30 m/s
Instrument	
Center frequency (f_c)	60 MHz
Bandwidth (B_w)	15 MHz
PRF*	6.4 kHz (30 Hz)
Power*	8 kW (1.71 MW)
Sampling frequency (f_s)	50 MHz
Pulse length (T_s)	1 μs
Receiving window (T_r)	50 μs

*For fields marked with an asterisk, the native MARFA radar value differs from the parameter used in the simulation. Simulation parameters are in parentheses, and represent values after coherent pre-summing of raw MARFA data.

presented in Gerekos et al. (2018). Because of higher ice/water dielectric contrast, rays reflecting from a facet within S_{bed} identified as water will have higher amplitude than facets comprised of rock.

The radar pulse repetition frequency (PRF) and aircraft velocity v determine the spatial resolution between radar observations in the azimuth direction, Δ_a, as shown in Fig. 1e. In the field, the MARFA instrument employs a native pulse repetition frequency (PRF) of 6400 Hz and peak power of 8 kW, then these observations are coherently stacked in post processing to achieve along track resolution of $1m$ before focusing (Peters et al., 2007). We chose to simulate the final (stacked) PRF and power instead of the native parameters for computational efficiency.

2.1 Dielectric Material Model

It is instructive to consider energy returned to the radar as the superposition of dielectric and geometric effects. Dielectric effects result from material property changes at an interface, while geometric effects result from the orientation of the interface's topography and the radar antenna. We will begin by discussing the implications of dielectric parameters then move on to geometric considerations.

A material model is applied to the simulation, where all facets on S_{ice} and S_{bed} are assigned a complex dielectric constant, ϵ. The refractive index η for each material is derived from ϵ (Eq. 2). For simplicity, we have chosen a three material model consisting of ice, rock, and water, with dielectric constants as presented in Table 2. Facets across S_{ice} are assigned ϵ_{ice}. Facets on S_{bed} are assigned ϵ_{rock} unless they are part of a water structure, which are assigned ϵ_{H_2O}.

Table 2. Complex dielectric constants for materials in radar simulations (Peters et al., 2005; Fujita et al., 2000; Midi et al., 2014; Glover, 2015). For each material, we chose a value within the range published in the literature. Further analysis of sensitivity to these material parameters is presented in the discussion.

Material	Literature Range ϵ	Simulated ϵ
Ice	$3.18 - 3.2$	$3.18 + .02i$
Water	$77 - 80$	$78 + .1i$
Bedrock	$4 - 6$	$5 + .15i$

For a facet at nadir, absolute reflectivity R_{abs} results from contrast between the refractive indices of ice and the bed material (η_{ice} and η_{rock}, respectively). A horizontal facet assigned ϵ_{rock} will have a lower reflectivity than one assigned ϵ_{H_2O} by about $15\,dB$ (Eq. 2). Real RES instruments are rarely calibrated to measure absolute reflectivity, and thus changes in received power, corrected for ice attenuation loss and geometric scattering, are assumed proportional to relative changes in reflectivity at the bed (R_{rel}). This forms the basis for the widely accepted assumption that $R_{rel} \geq 10 - 15\,dB$ over surrounding reflections implies the presence of liquid water (e.g. Young et al. (2016); Peters et al. (2007); Rutishauser et al. (2022)).

$$R_{abs,m} = \left(\frac{\eta_{ice} - \eta_m}{\eta_{ice} + \eta_m} \right)^2, \quad \eta_m = \sqrt{Re(\epsilon_m)}, \quad m = rock, H_2O \tag{2}$$

The simulations presented in this paper assume only the three materials described in Table 2, with well defined boundaries between hydrological and bedrock features. In real-world sub-glacial environments, additional material heterogeneity such as clays or hydrated tills exist, in addition to ambiguity in hydrological boundaries. We address the implications of this relatively simplistic material model in the *Discussion* section.

2.2 Geometric Model

We ultimately seek to develop the Stratton-Chu simulation method for constraining the extent and cross-sectional geometry of sub-glacial water features. Given this objective, it imperative to consider a menu of geometric constraints and how simulation parameters will emulate real-world RES returns from different targets. A basic simulation geometry for a hypothetical case is shown in Fig. 2a. These simulations consist of flat surfaces S_{ice} at elevation $0\,m$ and S_{bed} at $-d_{ice}$, where d_{ice} is the nominal ice thickness in m. The flight path is defined along the y-direction, with the radar's dipole antenna oriented along the x-direction. On the bed surface, a straight channel of width c_w is oriented perpendicular to the flight path.

2.2.1 Surface Roughness

Small-scale topography variation below the radar's detection limit, which we will refer to as roughness, can impact reflectivity by diffusely scattering incident radar energy. To account for this effect, random isotropic Gaussian variation is introduced to both S_{ice} and S_{bed} via Eq. 3. l_c is the correlation length in both the x and y directions. To capture scattering behavior due to
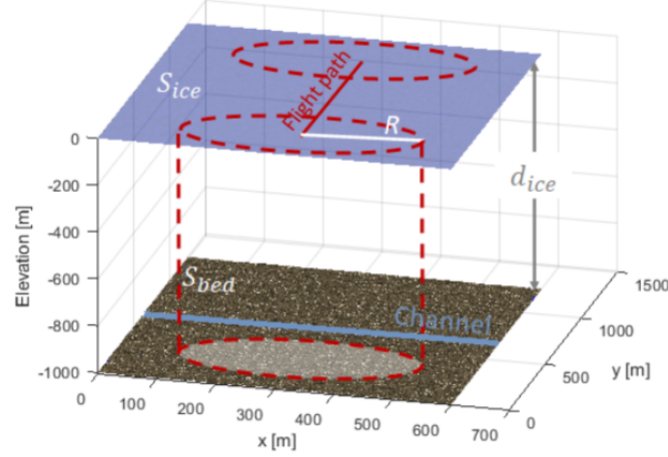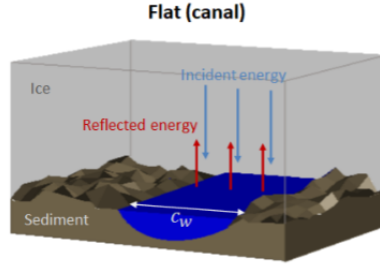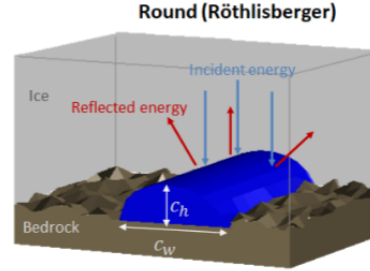
Figure 2. Radar simulation geometry a) Flight path (red), aircraft / channel orientation, simulation radius (R) across model footprint. b) A flat canal with width c_w. c) Geometry of a Röthlisberger or round channel with width c_w.

topography changes at the radar wavelength scale, l_c should be at least a few times the facet length l_f, while remaining near λ. In our simulations, both l_f and λ are $\sim 5m$, leading us to select $l_c = 15m$.

$$cov_{x,y} = \sigma_m e^{-(x^2+y^2)/l_c^2}, \qquad m = ice, bed \tag{3}$$

When we assume our surfaces are smooth relative to λ, we must add at least a negligible, non-zero roughness to avoid simulation artifacts, such as Bragg resonance (Gerekos et al., 2020). We always make this assumptions for S_{ice}, therefore $\sigma_{ice} = 0.2m$ in all simulations presented here. We also use $\sigma_{bed} = 0.2m$ in "smooth" bed simulations.

Hubbard et al. (2000) measured topography of recently deglaciated bedrock at high resolution in the Swiss Alps. They showed variations near $1m$ over horizontal distances of $15m$, which serves as our upper bound on σ_{bed}. We acknowledge this Alpine sub-glacial environment may differ considerably from common RES survey targets such as Thwaites Glacier in Antarctica. However, roughness studies from the Thwaites region integrate over long horizontal scales irrelevant to radar

140 scattering (Bingham and Siegert, 2009; Hoffman et al., 2022). We proceed with the understanding that our range for σ_{bed} from $0.2m$ to $1m$ may represent an imperfect but reasonable range of expected variation in small-scale topography.

2.2.2 Simulation Radius

An appropriate choice of simulation radius R is vital to accurately simulate radar geometric scattering from the sub-glacial environment. R defines the radius of a vertical cylinder, bounding the simulation scope at each aircraft position increment. If
145 R is sufficiently large to capture the entire antenna beam pattern cast on the bed surface, then the simulation will provide a complete representation of off-nadir clutter and target range migration. In many layered radar simulation experiments, R is chosen primarily to capture off-nadir clutter at the sub-surface target's apparent depth (Gerekos et al., 2018). Our experiments involve both a thick ice material layer and smooth S_{ice} relative to λ, limiting the impact of off-nadir clutter. Therefore, we base our choice for R on two alternative criteria:

150   – R must be greater than pulse limited radius R_{pl} at the glacier bed.

     – R captures adequate range migration to facilitate along track focusing.

The first criterion assumes the majority of returned energy from a nadir-directed radar will come from within the pulse-limited footprint (R_{pl}) beneath the aircraft. Eq. 4 approximates an upper limit on R_{pl} for a radar instrument with bandwidth B_w, where c is the speed of light in a vacuum. Actual R_{pl} will always be smaller than Eq. 4 predicts, due to refraction at
155 the air/ice interface. For our simulations, with $d_{ice} \sim 1000m$, $R_{pl} \approx 130m$ (Eq. 4). We consider this the minimum acceptable simulation radius, although clearly changes in aircraft height or simulated ice thickness will alter this limit.

$$R > R_{pl} \approx \sqrt{\frac{c\,(d_{ice} + h)}{B_w\,\eta_{ice}}} \tag{4}$$

An appropriate choice for R must also consider the desired range cell migration (RCM) at the bed surface. RCM is proportional to the change in physical distance a signal travels through air (r_{air}) and ice (r_{ice}) to reach a target as the radar
160 moves past (Eq. 5). In Synthetic Aperture Radar (SAR) processing, an aperture length L_a is chosen with sufficient range cell migration to optimize along-track focusing. This process improves signal-to-noise ratio and along-track resolution (Cumming and Wong, 2005). In order to facilitate simulated data focusing, R must be greater than the aperture required for desired range migration. Selection of L_a and the focusing process is described at length in *Simulated Data Processing*.

$$RCM(y) = \frac{2f_s}{c}\left[(r_{air}(y) - h) + \eta_{ice}(r_{ice}(y) - d_{ice})\right] \tag{5}$$

165 Our simulations target 3 cells of range migration at the bed surface ($RCM(L_a) \geq 3$). For $d_{ice} = 1000m$, and sampling frequency $f_s = 50MHz$, this translates to $L_a = 277m$. All simulations presented here use $R = 300m$, meeting both the pulse-limited and range migration criteria. Real airborne RES focusing typically includes more range cell migration ($RCM(L_a) = 5$

fast-time samples for MARFA), however such a simulation radius would be computationally unrealistic. Therefore, our choice of R represents a compromise, and the implications will be discussed in the *Discussion* section. Simulations in thicker ice may require significantly larger R or further compromises in range migration.

2.2.3 Channel Geometry

We seek to distinguish between two channel geometries common to sub-glacial hydrology. A canal-like structure (Fig. 2b) will have a flat cross-section and produce a specular reflection. This type of feature is common when the surrounding bed is comprised of sediment or other soft material, and the water pressure is high. Conversely, in a Röthlisberger channel (Fig. 2c), sub-glacial water carves a path through the ice above the bed surface (Röthlisberger (1972)). This type of channel is likely to form in areas where the substrate is impermeable bedrock with low water pressure (Walder and Fowler, 1994; Schroeder et al., 2013). We have confined ourselves to Röthlisberger channels and flat canals in this study as examples of common large-scale hydrological structures. Other known hydrologic features, such as Nye channels, may appear radiometrically similar to small canals, depending on their size and the radar wavelength. As algorithms and computational power enable higher resolution models, additional sub-glacial structures are a logical extension of this work.

A Röthlisberger channel has an elliptical cross section, with height c_h in addition to channel width c_w. We can infer from basic geometry it will reflect radar energy divergently (Fig. 2c), therefore the actual radar signature of a sub-glacial channel will be the superposition of its geometry and dielectric contrast. When representing the Röthlisberger curvature with flat facets, we must approximate the divergent scattering behavior by capturing reflections from multiple facets in the channel cross-section at nadir. In our simulations, we seek to capture reflections from at least the upper 6 facets within the simulation footprint as a reasonable estimate of Röthlisberger channel scattering (Fig. 3a).

Given our desired range of c_w and the constraint already imposed on R, it is clear that $l_f = 5m$ is too large to capture the scattering effect of a Röthlisberger channel. For our smallest channels ($c_w = 5m$), $l_f \leq .25m$ is required to capture the desired facet reflections (Fig. 3b). Setting such a high resolution over all of S_{ice} and S_{bed} is computationally unrealistic and negates many of the benefits of the Stratton-Chu simulation method (Gerekos et al., 2018). Therefore, we introduce a second facet length scale, l_{f2}, which defines the facet length on S_{ice} and S_{bed} only in a Röthlisberger channel location. l_{f2} has a maximum of $\sim c_w/20$, for $d_{ice} = 1000m$ and $c_h = c_w/4$ (Fig. 3b). When simulating flat canals, higher resolution channel facets are unnecessary due to the specular nature of the reflection, and therefore $l_{f2} = l_f$.

2.3 Simulated Data Processing

The simulator outputs rangelines, which represent the electric field strength vs. fast-time τ returned from a single pulse, at azimuth time a. τ is discretized into range bins (index j) with increments of $1/f_s$, where f_s is the sampling frequency. The rangelines are compiled sequentially in azimuth, or slow-time, into a 2-D raw radargram matrix (ξ_{raw}). Azimuth increments (index k) are equally spaced in time at $1/PRF$.

The rangelines are then focused using a version of the Range Doppler Algorithm (RDA) (Cumming and Wong, 2005; Hélière et al., 2007). We first perform range compression by convolving each raw rangeline with the complex conjugate of the

(a)

(b)
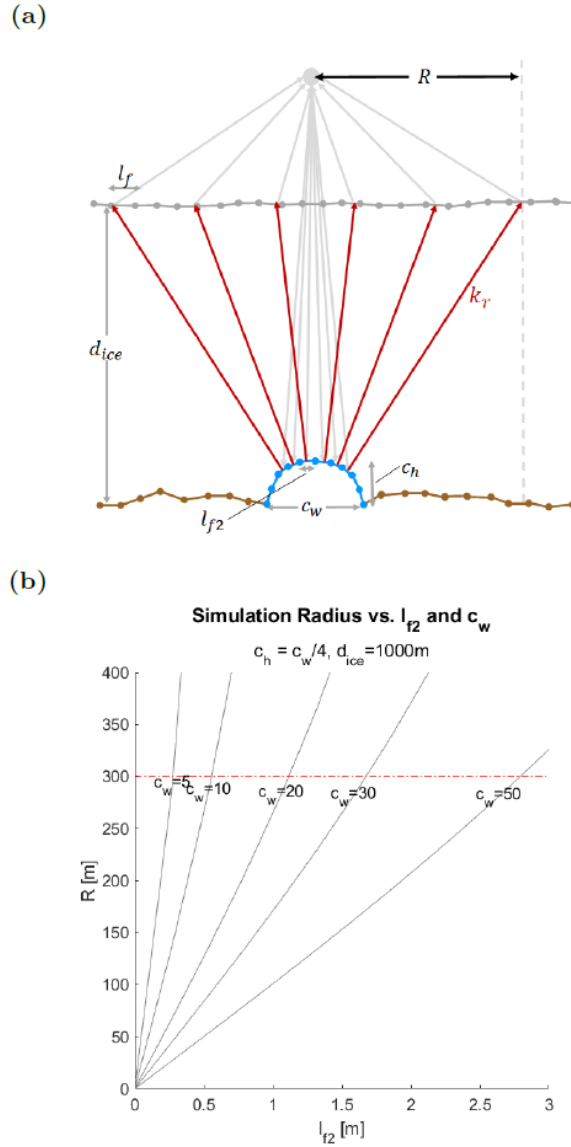
Figure 3. a) Schematic of simulation geometry for an elliptical cross-section of the Röthlisberger channel at nadir. A second, smaller facet length scale l_{f2} is introduced to facilitate accurate representation of curvature and divergent reflection character. b) Relationship between R and l_{f2} is defined for a range of c_w. We seek to limit $R \leq 300m$ for computational efficiency, imposing an upper limit on l_{f2} when simulating Röthlisberger channels

radar chirp, $g(\tau)$ over the radar receiving window T_r. The operation is performed by multiplication in the fast-time frequency domain.

$$g(\tau) = e^{-i\pi B_w \tau^2/T_s} \tag{6a}$$

205
$$\xi_{RC}(a,\tau) = \int_0^{T_r} \xi_{raw}(a, \tau + \tau')\, g(\tau)^* d\tau' \tag{6b}$$

We produce the focused radargram (ξ_f) by convolving the range compressed radargram (ξ_{RC}) with a 1-D reference function (ϕ) in along-track blocks (Eq. 7). The block size, L_a, for each fast-time value of τ is chosen such that range migration equals 3 fast-time samples. Thus it is important to note that L_a increases with depth, and simulation radius R must be greater than the maximum anticipated L_a.

210

$$\xi_f(a,\tau) = \int_{-L_a/2}^{L_a/2} \xi_{RC}(a + a', \tau)\, \phi(a', \tau)^* da' \tag{7}$$

A deeper discussion of mathematics and block processing required for along-track focusing is presented in the *Appendix A1*.

In all radargrams, we convert fast-time to physical depth (d) and slow time to along-track distance (y) via Eqs. 8 and 9, where $\tau_s = 2h/c$ is the fast-time value for the surface reflection, and PRF is the radar pulse repetition frequency.

215
$$d_j = \frac{c}{2\eta_{ice}}(\tau_j - \tau_s), \qquad \eta_{ice} = \sqrt{Re(\epsilon_{ice})} \tag{8}$$

$$y_k = k\frac{v}{PRF} \tag{9}$$

The focused field in ξ_f is converted to power in decibels. We define along-track bed reflectivity in absolute terms (R_{abs}) by taking the maximum reflected power beneath the ice surface (Eq. 10a). Relative reflectivity (R_{rel}) compares R_{abs} to the mean reflectivity for a simulation with the same σ_{bed} and σ_{ice}, but no channel present ($R_{abs,0}$). R_{rel} therefore measures the relative

220
reflectivity gain observed by the radar due to the channel's presence vs. surrounding frozen bed material.

$$R_{abs}(y_k) = max\{RG_{f,dB}(d, y_k)\}, \qquad d > 0 \tag{10a}$$

$$R_{rel}(y_k) = R_{abs}(y_k) - mean\{R_{abs,0}\} \tag{10b}$$

Table 3. Summary of RES simulation parameters for each channel geometry.

Parameter	All Simulations
Nominal ice thickness(d_{ice})	$1000\,m$
Simulation Radius (R)	$300\,m$
Facet length (l_f)	$5\,m$
Range migration for SAR Aperture (RCM)	$3\,cells$
Correlation length (l_c)	$15\,m$
Surface roughness (σ_{ice})	$0.2\,m$
Basal roughness (σ_{bed})	$0.2, 1\,m$
Channel width (c_w)	$0, 5, 10, 20, 30, 50\,m$

Parameter	Flat Canals	Round Channels
Channel facet size (l_{f2})	$5\,m$	$0.25 - 2.5\,m$
Channel height(c_h)	0	$c_w/4$

2.4 Hypothetical Simulation Cases

A basic simulation geometry is shown in Fig. 2. Our hypothetical simulation cases consisted of flat surfaces S_{ice} and S_{bed} with elevations of $0\,m$ and $-1000\,m$, respectively. Each surface had isotropic Gaussian roughness as defined above. On the bed surface a single channel of width c_w was oriented perpendicular to the flight path. Channel cross-sections were either flat canal-like structures (Fig. 2b), or round Röthlisberger channels with channel height of $c_w/4$ (Fig. 2c).

A series of simulation experiments were run for both types of channels. Channel width and basal roughness were varied according to Table 3. Simulations involving flat canals used a single facet length of $5\,m$, while Röthlisberger channels necessitated a smaller l_{f2} in the channel location to accurately represent geometric scattering. Rangelines from each simulation were processed as described above, and along track bed reflectivity from the focused radargram R_{rel} was compared for various scenarios.

3 Results - Hypothetical Simulation Cases

Basal roughness (σ_{bed}) impacted the absolute reflectivity of the solid bed material in our simulations ($R_{abs,0}$), which we use as a baseline for calculating the relative impact of channels on the radar echo (Eq. 10b). Simulations excluding liquid water, with σ_{bed} of $0.2\,m$ and $1\,m$ produced mean $R_{abs,0}$ equal to $-107.9 \pm 2.5\,dB$ and $-114.1 \pm 2.4\,dB$, respectively.

Simulations of hypothetical Röthlisberger channels and flat canals with the same c_w demonstrate the distinctly different radar signatures expected for the two geometries. Figure 4 shows along-track relative reflectivity for the two channel types,
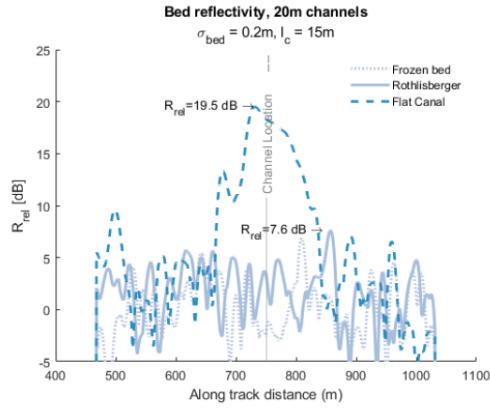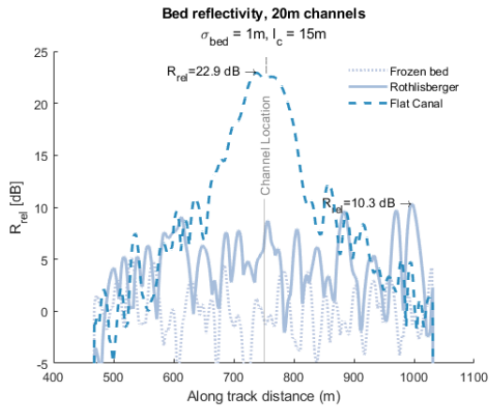
Figure 4. Simulated along track R_{rel} for $20m$ channels compared to a dry substrate with a) $\sigma_{bed} = 0.2m$. b) $\sigma_{bed} = 1m$. All simulations had the same Gaussian roughness correlation length $l_c = 15m$.

with $c_w = 20m$, compared to a frozen substrate with no liquid water. Flat canals have a distinct peak centered near the canal location, with maximum $R_{rel} = 19.5\,dB$ when $\sigma_{bed} = 0.2m$ and $22.9\,dB$ when $\sigma_{bed} = 1m$. Regardless of roughness, a $20m$ flat canal influences the radar echo for a few hundred meters along-track. The reduced gain of $3.4\,dB$ for a canal surrounded by a smoother substrate is consistent with the higher absolute reflectivity of the surrounding surface as discussed above.

245   When flowing through a rough bed (Fig. 4b), a $20m$ simulated Röthlisberger channel increased along-track R_{rel} for more than $500m$. The maximum peak in R_{rel} for this channel is $10.3\,dB$, and occurring when the channel is far from nadir. In our smoother bed simulations, the same $20m$ Röthlisberger cross-section produced peak R_{rel} of only $7.6\,dB$ (Fig. 4a), which could be nearly indistinguishable from fluctuations in reflectivity from the frozen substrate. Larger Röthlisberger channels produce only moderate gains in R_{rel}, with the largest ($50m$) channels having peak $R_{rel} = 13.2\,dB$ (Fig. 5).
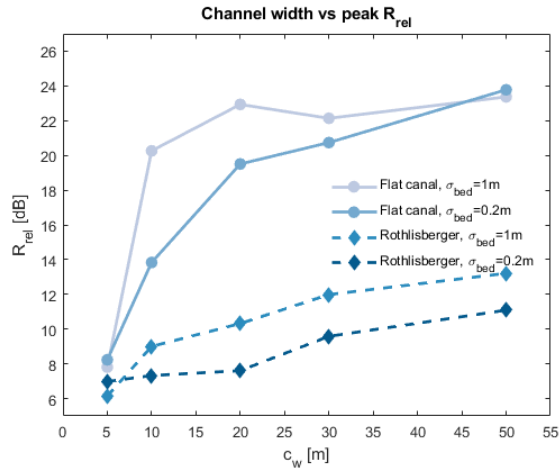
Figure 5. Positive correlation between c_w and R_{rel} for both Röthlisberger channels and flat canals. The increase in R_{rel} is much more pronounced for flat canals.

250   Figure 5 shows peak R_{rel} generally increases with width for both channels and canals, as an increasing proportion of the area within the radar's footprint contains liquid water. Flat canals exhibit a much stronger correlation between radar reflectivity response and c_w than Röthlisberger channels when $c_w < 20m$. At $c_w > 20m$, R_{rel} for flat canals increases more gradually as it appears to reach an asymptote of $\sim 24\,dB$.

4   Application to Thwaites Glacier

255 To demonstrate the simulator's utility in sub-glacial hypothesis testing, we compared simulated reflectivity to a single flight line collected with UTIG's 60 MHz MARFA instrument aboard an AS-350 B2 helicopter. The $16\,km$ line (THW2/ UBH0c/ X243a) was part of an airborne radar survey of Thwaites Glacier conducted between 2020 and 2022. The helicopter was flown at a nominal height of $500m$ above ground level with target velocity of $30m/s$. Precise aircraft positioning and orientation were recorded with an onboard Renishaw laser altimeter, Trimble Net-R9 dual frequency GNSS, and Novatel SPAN IGM-1A
260 inertial navigation, as described in Lindzey et al. (2020).

4.1   Study Area

Schroeder et al. (2013) first postulated a hydrological system dominated by channelized drainage within ~~$\sim 45\,km$~~ $\sim 45 - 70\,km$ of the Thwaites grounding line (Fig. 6)~~. This hypothesis is~~ based on data interpretation from the extensive 2003-2004 Airborne Geophysics of the Amundsen Sea Embayment, Antarctica (AGASEA) radar survey. The hypothesis was largely derived from
265 observations of high local bed reflectivity, combined with anisotropy in the radar specularity content. Specularity content measures the relative diffuse or specular nature of radar returned power. Schroeder et al. (2013) postulated that extensive Röthlisberger channels could produce such radiometric signatures due to their geometric orientation.
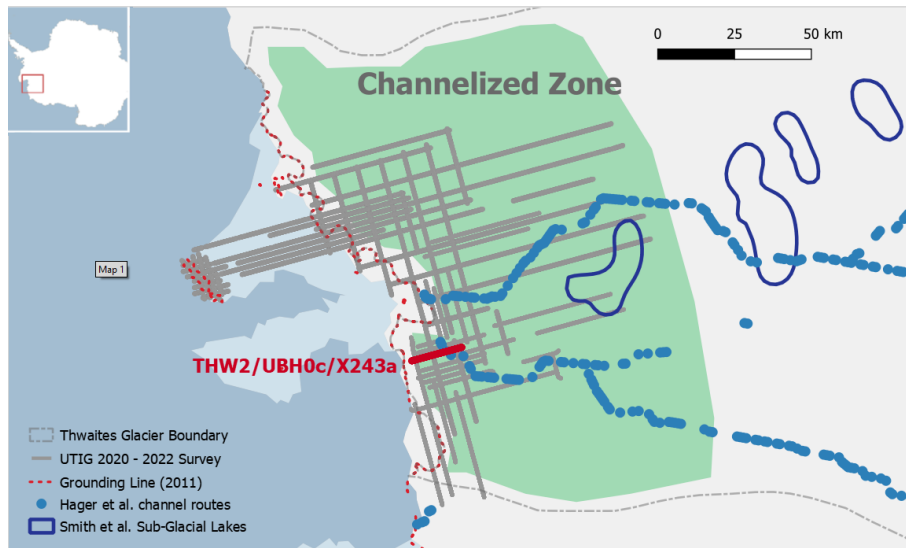
Figure 6. The grounding line region of Thwaites Glacier, with region of channelized hydrology proposed by Schroeder et al. (2013) highlighted. UTIG MARFA radar survey lines collected between 2020 and 2022 are shown in grey. Likely high volume channel routes suggested by Hager et al. (2022) are shown in blue, and the location of THW2/ UBH0c/ X243a is highlighted for reference.

Hager et al. (2022) ran a suite of sub-glacial hydrology simulations to evaluate the probability of persistent channelization routes beneath Thwaites. Their analysis concluded that Thwaites' near terminus hydrology is most likely comprised of a few persistent, high volume channels flowing toward the central grounding zone. Two probable routes were proposed, also shown in Fig. 6.

Although evidence for sub-glacial hydrology dominated by channelization in this region of Thwaites Glacier is strong, structures such as canals or lakes are not unprecedented. For example, Smith et al. (2017) identified a series of four active sub-glacial lakes that drained between June 2013 and January 2014. As seen in Fig. 6, at least one of these lakes is within the channelized region as defined in Schroeder et al. (2013).

Our chosen flight line, THW2/ UBH0c/ X243a, transects one of the proposed Hager et al. (2022) channel routes, and the radargram shows an isolated bright reflection coincident with this location, as shown in Fig. 7. Data from THW2/ UBH0c/ X243a were range compressed, corrected for geometric spreading loss and aircraft position, and focused in azimuth as described in Peters et al. (2007). This azimuth focusing is analogous to the procedure described in the *Simulated Data Processing* section, although a longer aperture sufficient for range migration of 5 cells is used.

The along-track surface and bed profiles were picked from the radargram. One-way attenuation loss of $13.8\,dB/km \pm 1.4\,dB/km$ was estimated using a spatially constrained linear regression model as outlined in Schroeder et al. (2016).
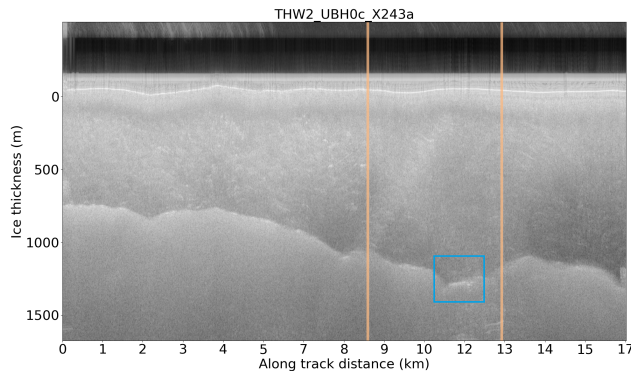
Figure 7. Focused radargram for THW2/ UBH0c/ X243a. The bright bed reflection corresponding to the proposed channel location is boxed. Vertical lines represent extent of the flight line re-produced in simulations.

4.2 THW2/ UBH0c/ X243a Simulation Methodology

We simulated a $4km$ segment of THW2/ UBH0c/ X243a containing the proposed channel, as depicted in Fig. 7. Along-track ice and surface elevations at nadir were calculated from the THW2/ UBH0c/ X243a focused radargram. From these data, we built elevation matrices for both the ice and bed surfaces ($S_{ice}^{rad}, S_{bed}^{rad}$) which vary in y according to the respective radar profiles, but have no x variation.

To build appropriate across-track (x) topography, separate 2-D topographic matrices were created from BedMachine V2 data ($S_{ice}^{BM}, S_{bed}^{BM}$) (Morlighem, 2020). The radar and BedMachine derived topography were superimposed to create simulation surfaces S_{ice} and S_{bed} via Eq. 11, where w is a quadratic weighting function varying between 1 at the surface edges and 0 at nadir. S^{rough} is an isotropic Gaussian surface, adding random roughness with $\sigma = 0.2m$ over correlation length $l_c = 15m$.

$$S = w S^{BM} + (1 - w) S^{rad} + S^{rough} - z_0(y) \tag{11}$$

In an ice penetrating RES survey, the aircraft attempts to "drape" the ice surface by flying at a constant height above ground level ($500m$ for UTIG helicopter based surveys). We simulate this with a polynomial interpolation of the radar-derived ice elevation along track, $z_0(y)$. For our simulations of THW2/ UBH0c/ X243a, z_0 is a 7th order polynomial, although in practice the polynomial order is somewhat subjective. The fitted function should approximate major terrain features in S_{ice}^{rad} with gentle elevation changes, consistent with actual aircraft operation.

$z_0(y)$ is subtracted from S_{ice} and S_{bed}, setting the average ice surface elevation to $\sim 0m$ (Eq. 11). The simulated aircraft elevation h is a constant $500m$. This approach preserves known ice geometry at nadir, and minor topographic features appear as variations in aircraft range to target.

The dielectric material model was applied to S_{ice} and S_{bed} as described previously. We ran individual simulations using the same surfaces, varying the width and geometry of across-track oriented channels in the location identified in Hager et al. (2022). Relative bed reflectivity from each simulation result was compared to the actual relative reflectivity from the focused

THW2/ UBH0c/ X243a radargram (Fig. 8c). The comparison between simulated and real data provides constraints on the extent and geometry of any real hydrological features at this location.

4.3   THW2/ UBH0c/ X243a Simulation Results

Figure 8 compares a $4\,km$ segment from THW2/ UBH0c/ X243a RES data with a simulation containing no water and $\sigma_{bed} = 0.2m$. There are two bright reflections in this section of the radargram (Fig. 8a). The first is centered around $1100m$ along-track with a maximum $R_{rel} = 11.6\,dB$. The second is a broad area of high reflectivity between $2300 - 2900m$, with R_{rel} peaks ranging from 13 to $15\,dB$ (Fig. 8c). This reflector coincides with the location of persistent Röthlisberger channelization proposed in Hager et al. (2022), and is the primary area of interest along THW2/ UBH0c/ X243a for this study.

The radargram from our frozen bed simulation (Fig. 8b) captures the basic bed topography well, but along-track reflectivity is not always aligned with the real THW2/ UBH0c/ X243a radargram. Simulated reflectivity near $1100m$ is consistent with the real data, indicating that this reflectivity peak could be the result of geometric effects from topography, rather than dielectric contrast from a water feature. However, for the region between $2300 - 2900m$, mean simulated R_{rel} is $22.8\,dB$ below the value observed in the real THW2/ UBH0c/ X243a data (Fig. 11). In our material model, a gain of this magnitude could only be consistent with a change in dielectric properties from rock to liquid water. It is also unlikely to be a Röthlisberger channel, since we have demonstrated such geometry is not conducive to increased reflectivity of more than $13.2\,dB$ for very large channels. We therefore propose that this basal reflector at $2300 - 2900m$ along-track could represent a specular hydrological structure, such as a flat canal, of unknown dimensions.

When a single flat canal with $c_w = 20m$ was added to the simulation, we observe a peak R_{rel} of $6.8\,dB$ over a narrow $150m$ range along-track (Fig. 9). This reflectivity gain is consistent with our findings from hypothetical simulation cases described above. However, the gain in reflectivity and along-track extent are insufficient to match the reflectivity profile observed in the real THW2/ UBH0c/ X243a RES data (Fig. 9b, Fig. 8c). Therefore, we conclude that the reflector cannot be a narrow, isolated flat canal.

The above results imply the reflector between $2300 - 2900m$ may be a broader hydrological feature, such as an area with multiple flat canals. This hypothesis is compatible with the topographic context, given that it is in a low-lying area with steep topography just down-glacier. This is an ideal location for till and liquid water to accumulate if the ice exists at its pressure melting point.

Figure 11 compares the difference between THW2/ UBH0c/ X243a and simulated R_{rel} for five simulations of such hydrological features. We compare both the mean and peak difference in R_{rel} between $2300 - 2900m$, and conclude that each more closely approximates THW2/ UBH0c/ X243a than our frozen bed or single $20m$ channel simulation. First, we consider the glacier sliding condition originally proposed by Weertman (1964). In this scenario, a thin water film (perhaps a few centimeters thick) coats the basal interface. We model this by maintaining the basal topography with $\sigma_{bed} = 0.2m$, but all facets from $2300m \leq y \leq 2900m$ are assigned ϵ_{H_2O}. This simulation has a broad increase in reflectivity in the area of interest (Fig. 10a), however the mean R_{rel} between $2300 - 2900m$ is $9.4\,dB$ below the actual THW2/ UBH0c/ X243a value (Fig. 11).
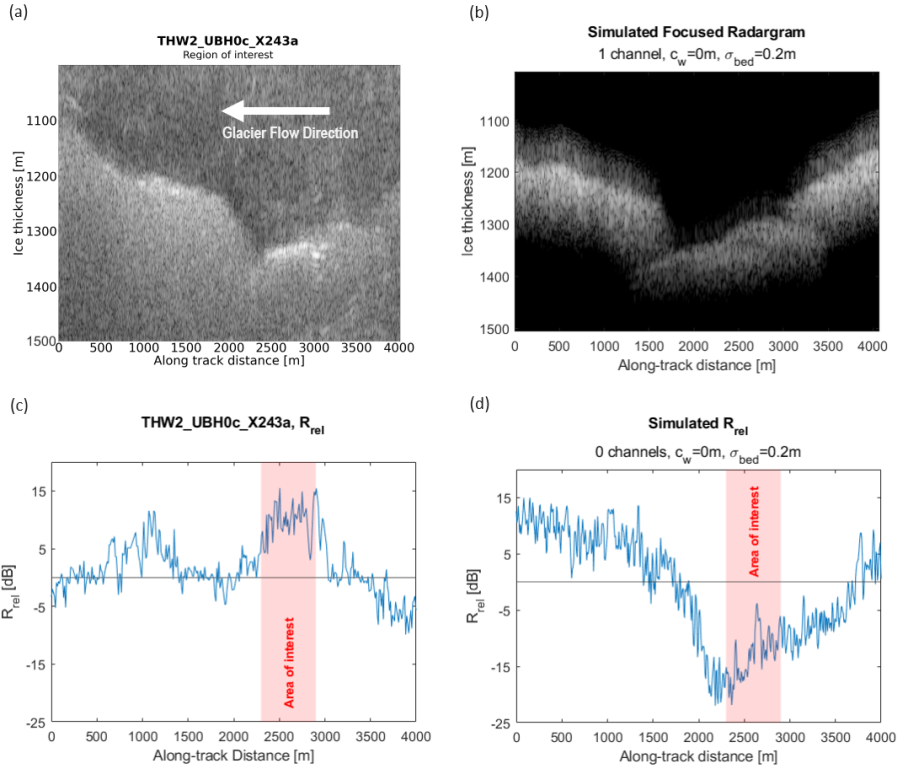
Figure 8. a) A $4km$ region from actual THW2/ UBH0c/ X243a radargram containing the Hager et al. (2022) channel location. b) Simulated THW2/ UBH0c/ X243a radargram with no water, $\sigma_{bed} = 0.2m$. c) Bed relative reflectivity from the actual THW2/ UBH0c/ X243a. The proposed channel location is highlighted in red as the area of interest. d) Simulated THW2/ UBH0c/ X243a relative reflectivity with no water, $\sigma_{bed} = 0.2m$. The along-track location corresponding to the area of interest is highlighted in red.

Figure 10b shows simulated R_{rel} when 30 narrow, flat canals ($c_w = 10m$) were placed over the same $600m$ region along-track. This creates a region where 50% of the area is covered with $10m$ flat canals between areas of bed material with smooth basal roughness ($\sigma_{bed} = 0.2m$). Mean R_{rel} at $2300 - 2900m$ for this simulation is slightly higher than the Weertman film scenario in Fig. 10a, but still $8.2\,dB$ below the target R_{rel} from THW2/ UBH0c/ X243a.

The third simulation in Fig. 10c has 8 larger flat canals ($c_w = 30m$) evenly spaced across the same $600m$ region. This simulation has less water coverage than the $30x10m$ flat canal simulation (40% vs. 50%), yet the wider channels increased mean R_{rel} by $3.5\,dB$ (Fig. 11). This result reinforces that the shape of hydrological feature, not just the extent, makes a significant difference to the resulting reflectivity profile. Mean R_{rel} for the $8x30m$ simulation was $4.8\,dB$ below actual. When the geometry is changed to 6 canals of $50m$ (Fig. 10d), mean R_{rel} improves to just $2.3\,dB$ below THW2/ UBH0c/ X243a.

The final simulation includes a very broad area of specular ($\sigma_{bed} = 0m$) water covering the bed between $2300 - 2900m$. Due to its size, we refer to this feature as the "$600m$ lake" in Figs. 10 and 11. Mean R_{rel} over $2300 - 2900m$ for this simulation
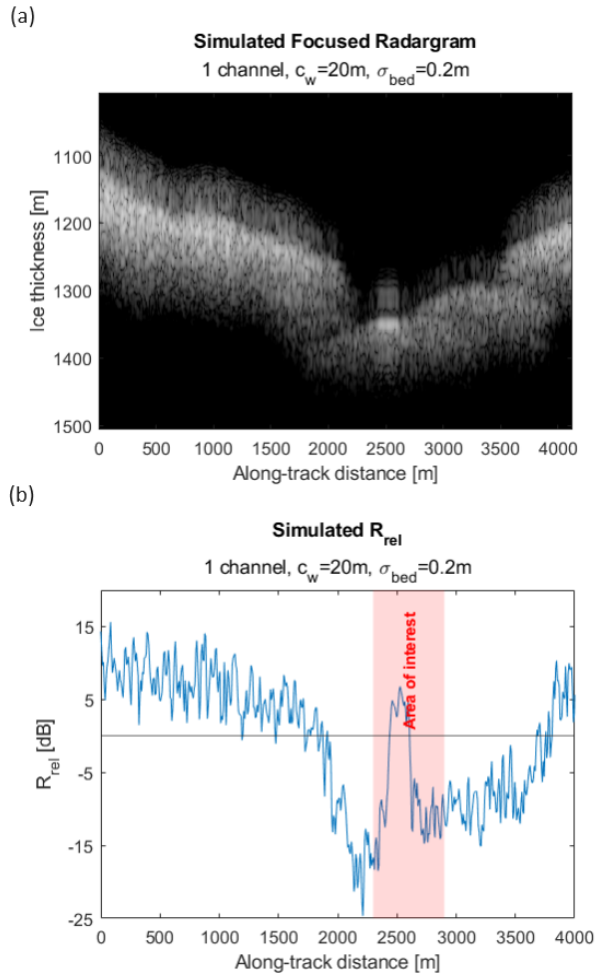
Figure 9. a) Simulated radargram with a single 20m canal at the bed. b) Relative reflectivity for simulation with 20m canal at the bed.

deviates by only $0.6\,dB$ from the actual THW2/ UBH0c/ X243a data. However, this $600m$ lake simulation includes several peaks as high as $18\,dB$, which is $3\,dB$ higher than the maximum peaks observed in THW2/ UBH0c/ X243a. The maximum R_{rel} from both the $8x30m$ and $6x50m$ simulations more closely matched the peaks observed in THW2/ UBH0c/ X243a.

We infer our feature at $2300-2900m$ could be a wide area of sub-glacial water. Of the scenarios we tested, simulated reflectivity was most consistent with canals averaging $\sim 30-50m$, with at least 50% water coverage, or a sub-glacial lake. Although THW2/ UBH0c/ X243a and our simulations demonstrate similar reflectivity profiles, we have not explored a comprehensive range of bed materials or water geometries. In the *Discussion* section, we address these omissions and the resulting limitations.
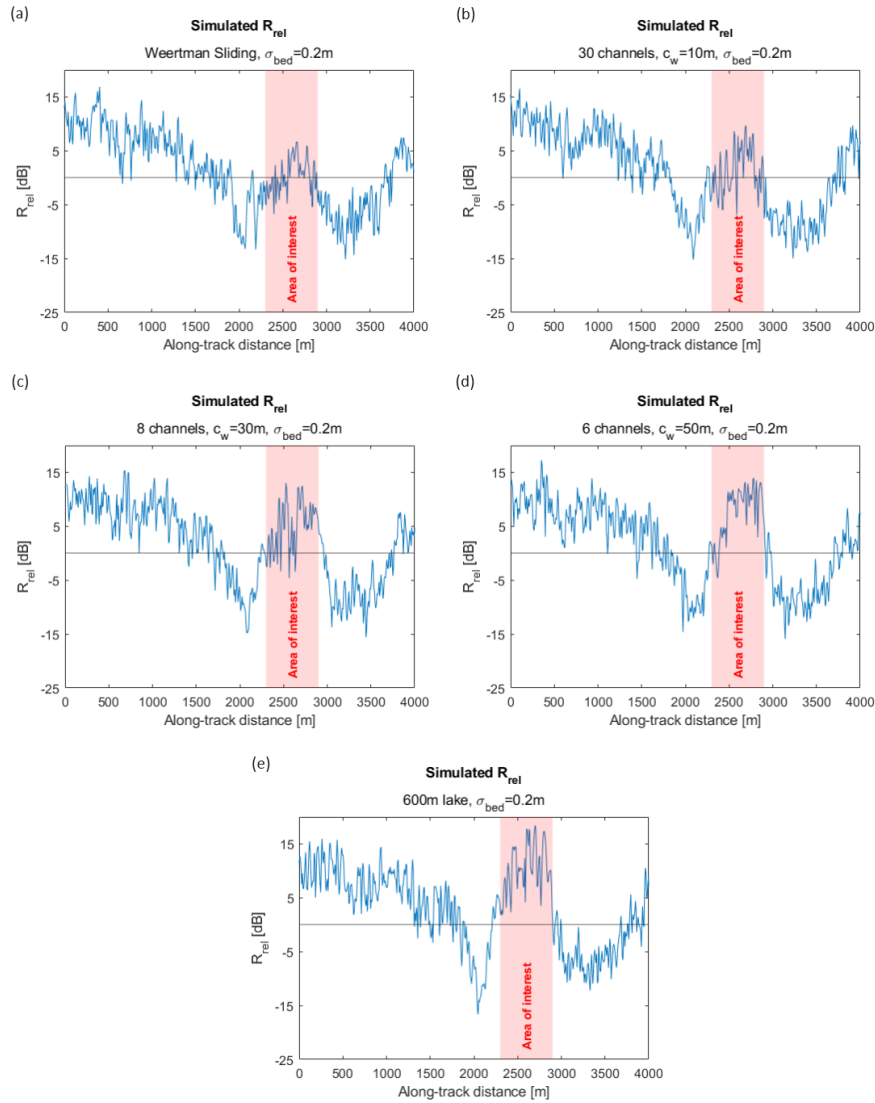
Figure 10. R_{rel} for distributed hydrology simulations at $2300 - 2900m$ along-track. a) a 600m wet bed surface (Weertman sliding), b) 30 flat canals with $c_w = 10m$, c) 8 flat canals with $c_w = 30m$, d) 6 flat canals with $c_w = 50m$, e) the entire $600m$ area covered with a flat, specular water body (e.g. sub-glacial lake).

5 Discussion

Our comparison of Röthlisberger vs. flat canal reflectivity implies that a real world Röthlisberger channel, even one of significant size, may not exhibit an obvious reflectivity increase. Röthlisberger channels will scatter energy divergently (Fig. 2c), therefore we see a smaller magnitude impact to R_{rel} over a greater distance than an equivalent flat canal. The simulator also demonstrates the radar's sensitivity to relatively narrow, flat canals. A single canal $\sim 10m$ can exhibit a reflectivity peak
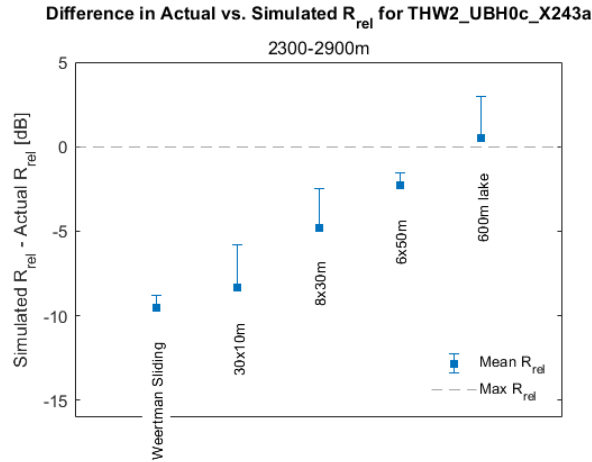
Figure 11. Mean difference between actual and simulated R_{rel} over the area of interest at $2300 - 2900m$, for all simulations of THW2/ UBH0c/ X243a. Positive error bars show the difference between mean and peak R_{rel} over $2300 - 2900m$.

360 $> 15\,dB$ (Fig. 5, *Flat canal*). Such a hydrological feature would cover $< 5\%$ of the radar's pulse-limited footprint under $1000m$ of ice.

~~Given the limited reflectivity response from Röthlisberger channels, more advanced analysis techniques may be required to detect and characterize them accurately. This could include examination of specularity content as described in Schroeder et al. (2015). Specularity content quantifies the diffuse or specular quality of radar reflections by comparing returned power over two~~
365 ~~different SAR apertures. Our simulation methodology produces a focused product, but requires a simulation radius $R > L_a$. Current computational constraints limit our choice of R to well below the aperture length of $2km$ recommended in Schroeder et al. (2015). Although simulating specularity content would be a valuable advance, we leave it to a future study.~~

In our simulations of hypothetical cases, increasing σ_{bed} from $0.2m$ to $1m$ reduced radar echo power by $6.2\,dB$. These results are intuitive. Scattering loss from a rougher surface will be larger and more variable, resulting in lower absolute reflectivity.
370 Increasing roughness also alters the geometric contrast between the hydrological feature and surrounding material. The effect is most pronounced for flat canals with $c_w = 10 - 20m$. The impact of the bed roughness near flat canals diminishes as c_w exceeds $30m$, when reflected energy from the water feature dominates the radar return.

Our examination of roughness in this study was deliberately limited, as the relationship between roughness and scattering is complex and dependent on correlation length l_c. Bingham and Siegert (2009) and Peters et al. (2005) each calculated wide
375 ranges in RES-derived roughness across Antarctica. Roughness in these studies was measured with $l_c > 1km$. Our method for building simulated ice and bed surfaces directly incorporates RES observations of topography with resolution $l_f = 5m$ (see *THW2/ UBH0c/ X243a Simulation Methodology*). Therefore, any relationship between scattering and large scale roughness, as defined in these previous studies, is captured explicitly in our simulations.

Smaller scale sub-glacial roughness ($l_c <$ a few wavelengths) is more challenging to measure directly, therefore actual values are poorly constrained. Our hypothetical case simulations demonstrate bed roughness over $l_c = 15m$ may alter radar echoes without a change in dielectric properties. This is consistent with Peters et al. (2005), who calculated theoretical scattering loss due to roughness could exceed $20\,dB$ when $l_c << 1^{st}$ Fresnel zone. A comprehensive examination of basal roughness, incorporating a broad range of both l_c and σ_{bed} is possible with our simulation method. Such an investigation is an active area of interest for future work. Our limited current examination of roughness demonstrates that RES detection of sub-glacial water must be nuanced. Large changes in R_{rel} may constitute a liquid water signature, but could also indicate spatial heterogeneity in roughness or substrate material.

Given the limited reflectivity response from Röthlisberger channels, more advanced analysis techniques may be required to detect and characterize them accurately. This could include examination of specularity content as described in Schroeder et al. (2015) . Specularity content quantifies the diffuse or specular quality of radar reflections by comparing returned power over two different SAR apertures. Schroeder et al. (2013) argued that specularity content from Röthlisberger channels will be orientation dependent; high values are expected if the radar path is aligned along the channel, and lower values if the path is transverse to it. By contrast, flat interfaces such as canals and lakes should exhibit high and isotropic specularity content. Although our simulation methodology produces a focused product, it requires a simulation radius R greater than the SAR aperture L_a. Current computational constraints limit our choice of R to well below the aperture length of $2km$ recommended in Schroeder et al. (2015). Therefore, although simulating specularity content would be a valuable advance, we have left it to a future study.

Our THW2/ UBH0c/ X243a simulations from Thwaites Glacier demonstrate ~~consistency~~ the reflectivity magnitude may be consistent with a wide, ~~specular~~ flat water feature surrounded by bedrock. ~~This configuration is likely, but~~ Although we did not simulate specularity content, we observe a prominent peak in actual specularity content from THW2/ UBH0c/ X243a (Fig. 12). This observation could support the hypothesis that the reflector at $2300 - 2900m$ is a series of flat canals or a lake, since THW2/ UBH0c/ X243a is transverse to the channelized flow hypothesized in Hager et al. (2022) (Fig. 6). An orthogonally oriented radar transect could enable a more definitive diagnosis of the hydrological feature.

Further, while our flat canal or lake configurations appear likely, we cannot preclude all competing hypotheses without exploring many additional scenarios beyond the scope of this work. Other water geometries, such as Nye channels or water sheets (Creyts and Schoof, 2009), combined with a broader range of roughness length scales, could fit the original THW2/ UBH0c/ X243a reflectivity profile. Future work should identify computational or process efficiencies to allow comprehensive coverage of all possible simulation variables.

The material model in this study was also deliberately simple. Many possible bed materials exist, ranging from frozen bedrock to clays with real dielectric constants between $2.7 - 36$ (Fig. 13) (Christianson et al., 2016; Tulaczyk and Foley, 2020; Peters et al., 2005). As we demonstrated in the frozen bed simulation, the feature at $2300 - 2900m$ requires an apparent reflectivity gain of $22.8\,dB$ over the surrounding substrate. Roughness may contribute some of this difference, however the observation is most consistent with a significant material transition. Figure 13 shows dielectric contrast between sub-glacial water and most clays or hydrated tills is less than $10\,dB$, making them unlikely bed materials. Conversely, substrates with the
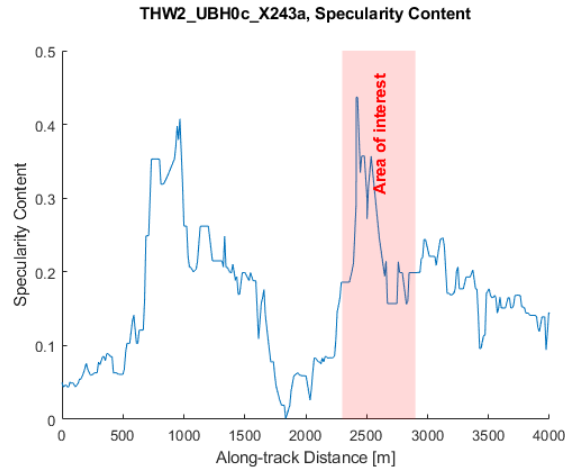
Figure 12. Actual specularity content from THW2/ UBH0c/ X243a.
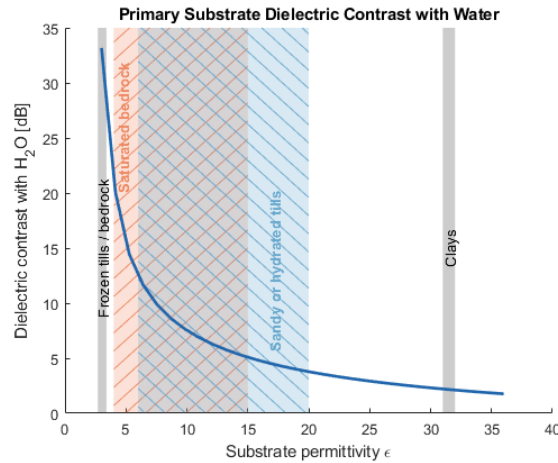
Figure 13. Dielectric contrast between liquid water ($\epsilon = 78$) and substrates with permittivity from $\epsilon = 2.7 - 36$. Example bed material permittivity ranges from Tulaczyk and Foley (2020) are provided as reference.

lowest dielectric constants are generally frozen, and therefore inconsistent with the presence of liquid water. These constraints limit the likely substrates to a narrow range of materials with $\epsilon \sim 5$, including saturated bedrock or hydrated tills with low permittivity.

We also assumed constant radar attenuation in the ice. This assumption is reasonable given the short physical distances of our radar simulations. Simulations over greater distances may require introducing variation in the imaginary component of ϵ_{ice} to account for heterogeneous attenuation loss.

420   Several notable inconsistencies between our best simulations and the original THW2/ UBH0c/ X243a data remain. First, R_{rel} over the beginning and final $500\,m$ of all the simulations are about $\sim 5-8\,dB$ higher than the THW2/ UBH0c/ X243a RES data. This may indicate a change in σ_{bed} near the edges of the simulated region. The low-lying area in the region is a perfect topographical feature to accumulate sediments, which likely have low roughness. The steeper and elevated topography near the edges of the simulated region may be exposed bedrock, with greater roughness, which we have shown in our hypothetical case

425   simulations can reduce R_{rel}. In future iterations of the simulator, we would enable heterogeneity in σ_{bed} in order to capture this type of variation explicitly.

R_{rel} near $2000\,m$ along track in all simulations was $\sim 10\,dB$ lower than actual reflectivity from THW2/ UBH0c/ X243a (Fig. 10). This is coincident with a steep topographical feature in the THW2/ UBH0c/ X243a radargram. Steep slopes such as this likely represent a limitation of our simulation approach. The surface representation using $5\,m$ flat facets will inherently direct

430   more reflected energy away from the antenna position than a real surface. Therefore, caution is imperative when interpreting results near steep topography.

There are several additional limitations of our simulation method for testing sub-glacial hypotheses. The simulation radius of $R = 300\,m$ explicitly limits the impact of range migration and clutter. A more complete simulation incorporating a larger R would be beneficial to simulations in thicker ice or focusing over longer apertures. Future work should combine additional

435   computing power with simulated facet roughness (Gerekos et al., 2023) for additional realism and efficiency.

Our 3 dimensional model for topography was derived from a single along-track RES flight line, which enables high confidence and sufficient resolution for ice geometry at nadir. However, this has the obvious limitation of requiring a previous flight line to build our simulation. We lack similar observation density across-track, leaving low-resolution ($500\,m$) open-source DEMs (such as BedMachine V2 (Morlighem, 2020)) as our best option for approximating off-nadir features. This asymmetry

440   is not problematic for replicating basic topography at nadir, but the lack of realistic off-nadir features reduces the sharpness of the focusing algorithm. We can see this effect by comparing the image quality in Fig. 8a to simulated results in Fig. 8b or Fig. 9a. This limitation also clearly reduces our ability to assess hypotheses involving any off-nadir targets.

It is also important to note that the simulation in Fig. 8d exhibits along-track R_{rel} variation $> 30\,dB$ without any change in dielectric properties at the bed. This demonstrates that significant changes in bed echo strength are possible due to topography

445   alone. When inferring the presence of sub-glacial hydrological features from RES data, care must be taken to consider reflectivity within the context of bed topography, hydraulic potential, attenuation, and other factors which may influence the radar echo strength. This observation also demonstrates the value of our simulation methodology for confirming the presence and extent of sub-glacial water.

6   Conclusions

450 In the exercise presented here, we optimized a radar simulation technique developed by Gerekos et al. (2018) to study the theoretical RES response from sub-glacial systems. The simulator incorporates the Stratton-Chu integral and Linear Phase Approximation to efficiently estimate backscattered radar signal from simulated targets. Through a series of hypothetical

simulation cases, we demonstrated the impact on relative reflectivity from rounded Röthlisberger channels or specular flat canals surrounded by bed materials of varying roughness. These simulations confirmed that reflectivity is highly dependent on both the size and cross-sectional shape of the sub-glacial water structure. Our results can be applied broadly to infer the presence, size, and structure of sub-glacial water bodies from RES surveys in a more robust and sophisticated way than previous methods.

In our example flight line THW2/ UBH0c/ X243a from Thwaites Glacier, we demonstrated the simulator's utility in testing relevant hypotheses in sub-glacial hydrology. A large water structure could produce the elevated reflectivity beneath Thwaites Glacier, in a region coinciding with a channel route proposed by Hager et al. (2022). Of the scenarios we tested, the radar signature was most consistent with canals averaging $> 30m$ and covering at least half the area for $600m$ along-track. Additional simulations including a broader range of hydrological structures and bed materials may be useful to conclusively characterize the feature. Further, although we do not see a Röthlisberger channel at this precise location, our findings do not preclude Röthlisberger channelization further upglacier. Simulations of new and existing RES survey data across the Thwaites catchment could characterize the extent of upglacier channelization.

The method we outline has broad applicability for studying the basal environment of large glaciers. As we have shown, the simulation methodology can offer useful constraints when testing sub-glacial hypotheses. Scientific intuition, additional data inputs, and more computational power will improve the promise of this technique. Future work may include additional computational efficiency to enable more extensive simulations, as well as comprehensive investigations on the impact of substrate roughness, material properties, and additional water geometries.

RES data collection is logistically challenging and expensive. A forward model capable of predicting optimal locations for sub-glacial survey targets, instead of modeling existing flight lines, is an area of interest for future work. In such a forward model, computing resources must be optimized by strategically constraining parameter sets, and performing sensitivity tests for many of the variables considered here.

Data availability. All referenced data in this paper are made available at: https://doi.org/10.5281/zenodo.8165256

Appendix A: Along-Track Focusing

We produce the focused radargram (ξ_f) by processing in along-track blocks. For a given fast-time range bin τ_j, the block size, L_a, is chosen such that range migration for a target equals 3 fast-time samples, as depicted in A1. Thus it is important to note that the block size increases with depth, and simulation radius R must be greater than the maximum anticipated L_a.

To process a block centered at slow-time a_0, with depth τ_j, we begin with a block of length $2L_a$ from the range compressed radargram (ξ_{RC}) as shown in A1. We calculated a 1-D reference function (ϕ), representing the Doppler phase modulation as the antenna travels across the aperture in slow-time a (A1) Peters et al. (2007); Legarsky et al. (2001); Hélière et al. (2007). The amplitude term, b, in A1 is used in real-world IPR processing to account for along-track variations in instrument gain, aircraft
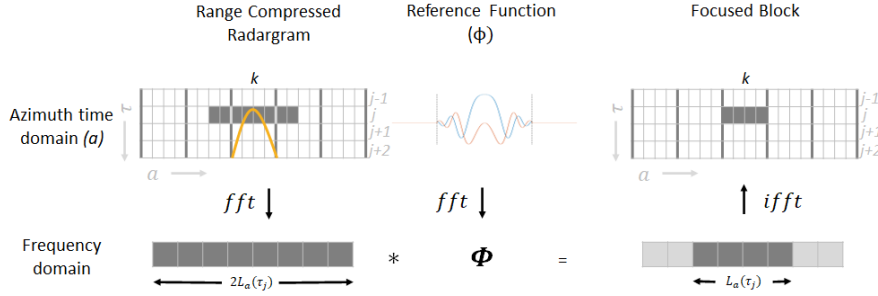
Figure A1. Schematic representation of along-track (azimuth) focusing for a single block at discrete fast-time increment τ_j and azimuth block k in a simulated radargram. The orange hyperbola superimposed on the range compressed radargram illustrates the theoretical range migration of a target with shortest fast-time range τ_j. We select an aperture length L_a such that range migration spans 3 sample cells (τ_j, τ_{j+1}, and τ_{j+2}).

motion, and to attenuate high Doppler frequency contributions at long apertures Legarsky et al. (2001); Peters et al. (2007). Our simulations do not contend with non-ideal flight or instrumentation variables, and we therefore use a simple Hamming window of width L_a for suppression of higher frequency sidelobes.

$$\phi_j(a) = b(a)e^{-i2f_c\tau(a)}\Big|_{a=a_0-L_a/2}^{a=a_0+L_a/2} \tag{A1}$$

$$\xi_f(a,\tau) = \int_{-L_a/2}^{L_a/2} \xi_{RC}(a+a',\tau)\,\Phi(a',\tau)^*da' \tag{A2}$$

The data block and reference function are Fourier transformed and convolved in the frequency domain (Fig. A2). The result is transformed back to the slow-time domain via inverse Fourier transform. Because the reference function ϕ is tuned at the center of the original block, the middle of the final block is better focused than the edges Hélière et al. (2007). For this reason, only half of the final block (length L_a) is written to the focused radargram ξ_f (A1). This process is repeated for each along-track block, at all fast-time range bins until a complete focused radargram, ξ_f, is formed.

Appendix B: List of Variables

Table B1. List of variables used in this manuscript

Symbol	Definition	Units
a	Time scale for radar observations along track (azimuth)	s
b	reference function amplitude scaling function	
B_w	Radar bandwidth	Hz
c	Speed of light in free space	$m\,s^{-1}$
c_h	Channel height	m
c_w	Channel width	m
d_{ice}	Nominal ice thickness	m
$\bar{E}_{b,n}$	Backscattered electric field strength from S_{ice} to radar antenna	$V\,m^{-1}$
$\bar{E}_{i,n}$	Incident electric field strength from radar antenna to $Sice$	$V\,m^{-1}$
$\bar{E}_{r,n}$	Reflected electric field strength from S_{bed} to $Sice$	$V\,m^{-1}$
$\bar{E}_{t,n}$	Transmitted electric field strength from S_{ice} to $Sbed$	$V\,m^{-1}$
f_c	Radar central frequency	Hz
f_s	Sampling frequency	Hz
g	Radar chirp signal	
h	Aircraft height	m
j	Fast-time (τ) incremental index	
k	Azimuth time (a) incremental index	
$\bar{k}_{i,n}$	Wavevector from radar antenna to facet n on $Sice$	
$\bar{k}_{r,n}$	Reflected wavevector from facet on S_{bed} to S_{ice}	
$\bar{k}_{t,n}$	Transmitted wavevector from facet n on S_{ice} to S_{bed}	
L_a	Aperture length for SAR focusing	m
l_c	Roughness correlation length	m
l_f	Facet length	m
l_{f2}	Secondary facet length for Röthlisberger channels	m
PRF	Pulse repetition frequency	Hz
R	Simulation radius	m
R_{abs}	Absolute reflectivity	dB
$R_{abs,0}$	Absolute reflectivity of frozen bed	dB
R_{pl}	Pulse limited radius	m
R_{rel}	Relative reflection coefficient	dB

Table B2. List of variables, continued

Symbol	Definition	Units
r_{air}	one-way travel distance through air from radar to ice surface	m
r_{ice}	one-way travel distance through ice from ice surface to target	m
RCM	range migration, number of discrete fast-time cells	
S_{bed}	Discretized bed surface	
S_{ice}	Discretized ice surface	
S^{BM}	2-D topography matrix derived from BedMachine V2 (ice or bed)	m
S^{rad}	2-D topography matrix derived directly from radar data	m
S^{rough}	2-D matrix representing random isotropic Gaussian roughness	m
T_r	Radar receiving window	s
T_s	Radar pulse length	s
v	Aircraft velocity	$m\,s^{-1}$
w	Across track quadratic weighting function (values [0,1])	
x	Cartesian spatial coordinate, across track	m
y	Cartesian spatial coordinate, along track	m
z	Cartesian spatial coordinate, elevation	m
z_0	Polynomial interpolation of ice elevation, approximating aircraft drape	m
ϵ_{H_2O}	Relative dielectric constant of water	
ϵ_{ice}	Relative dielectric constant of ice	
ϵ_{rock}	Relative dielectric constant of rock	
η_{H_2O}	Refractive index of water	
η_{ice}	Refractive index of ice	
η_{rock}	Refractive index of rock	
θ	Channel orientation angle relative to x-axis	rad
λ	Radar center wavelength	m
ξ_f	Focused radargram	$V\,m^{-1}$
ξ_{raw}	Raw radargram	$V\,m^{-1}$
ξ_{RC}	Range compressed radargram	$V\,m^{-1}$
σ	Roughness amplitude for surface i	m
τ	Fast-time scale for returned radar echoes	s
τ_s	Fast-time value for radargram ice surface reflection	s
ϕ	1-D reference function for radargram focusing	
Φ	Slow-time Fourier transform of reference function ϕ	

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