I. INTRODUCTION

Flexural waves traveling in the sea ice cover are dispersive, the higher the frequency the faster the propagation velocity, at least for periods lower than about 10 s. This dispersion is controlled by the elastic properties of the material, and also by the ice thickness (Ewing et al., 1934; Anderson, 1958). Comparing the observed dispersion of flexural waves with the theoretical dispersion curve could allow the determination of the ice thickness, provided all the elastic constants are known. In seismic studies, the latter can be constrained by estimating the non-dispersive velocities of longitudinal plate (LP) and horizontally polarized shear (SH) waves. Several field experiments have been conducted in the past to exploit these seismic methods (Yang and Giellis, 1994; Stein et al., 1998), see also Anderson (1958) for a review of earlier works. The typical experimental set-up involves the use of active source, e.g., hammer blows or shots; these methods therefore require human assistance and are thus non-autonomous.

For example, Stein et al. (1998) used networks of triaxial geophones to invert for ice characteristics, including ice thickness. They generated impulsive signals at a well-defined source location with a sledgehammer, and exploited the travel times of the three phases (P, SH and flexural). The frequency ranged from 5 to 50 Hz, and the estimates were area-averaged parameters. Performing their inversion on two very different datasets, they obtained good accuracy on the ice thickness for undeformed first-year ice and for more complex, irregular multi-year pack ice, although other characteristics (Young’s modulus, shear modulus, and Poisson coefficient) could not be estimated for the latter data.

We here describe how the inversion of ice thickness can be performed without active sources, by exploiting the dispersion of the ice swell. These ubiquitous flexural waves have a narrow spectral content, generally peaking at about 25 to 30 s, and dominate the seismic signal, at least far from the coast. Compared to active sources, the ice swell is a natural, permanent source. The disadvantages in using it are (1) that multiple reflections and scattering of the ice swell make the wavefield complex to study; (2) that the dominant period is too high to allow for a good resolution of the ice thickness.

To illustrate the second issue, we make use of the phase velocity $v_\phi$ of flexural waves with angular frequency $\omega$ propagating in an ice cover of thickness $h$ lying above an infinitely deep water column, as given by (Stein et al., 1998):

$$\frac{\rho_w}{D} \left( g - \frac{\omega^2}{\sqrt{\frac{\omega^2}{v_p^2} - \frac{\omega^2}{v_\phi^2}}} \right) = \frac{\hbar \omega^2 \rho}{D} - \frac{\omega^4}{v_\phi^2} \tag{1}$$

$\rho_w$ is the seawater density, $D$ is the water column depth, $g$ is the gravitational acceleration, $\hbar$ is the ice thickness, $\rho$ is the ice density, $v_p$ is the longitudinal wave velocity, and $v_\phi$ is the phase velocity of the flexural waves.
where \( v = 1440 \text{ m/s} \) is the sound speed in water, \( \rho_w = 1000 \text{ kg/m}^3 \) is the density of water, \( g = 9.81 \text{ m/s}^2 \) is the acceleration of gravity, \( \rho \) is the density of sea ice, and
\[
D = \frac{E h^3}{12(1 - \nu^2)}
\]
with \( E \) Young’s modulus and \( \nu \) the Poisson coefficient of sea ice. For periods much longer than \( 2\sqrt{h} \), i.e., 6 s to 14 s for a typical ice thickness of 1 to 5 m, the term in brackets in Eq. (1) becomes vanishingly small, and \( v_\phi \simeq g/\omega \), the phase velocity of gravity waves, which is independent of the ice thickness. The group velocity \( v_G \) is deduced from \( v_\phi \) as
\[
v_G = \frac{v_\phi}{1 - \frac{\omega}{\omega_0} \frac{dv_\phi}{d\omega}}. \tag{2}
\]

As an example, we show in Fig. 1 the group velocity \( v_G \) deduced from \( v_\phi \) for periods ranging between 1 s and 30 s, for various values of the ice thickness. Dependence on the elastic parameters is weak, when probing typical values proposed in the literature for Young’s modulus, Poisson coefficient and the density of sea ice. Similarly, the dependence on temperature fluctuations through changes in mechanical properties of the sea ice is even weaker than the effect of changing \( E \) from 7.2 to 9.7 GPa as shown in Fig. 1. For periods greater than about 20 s, hence including most of the energy of the ice swell, the curves collapse onto a single curve \( v_G = \frac{g}{2\pi T} \), which is the group velocity of gravity waves. Therefore, the dispersion cannot be used to constrain the ice thickness if we restrict the analysis to the period that are typical of the ice swell: shorter periods must be explored.

In this article, we detail how these two issues can be addressed and check the validity of this method by performing several tests and analyses of field data.

II. DATA

The seismic data analyzed in this work were already described in Marsan et al. (2011): a seismic network was deployed in April 2007 as part of the measurement campaign at the Tara drifting station operating in the framework of the DAMOCLES project (Gascard et al., 2008). From this network made of 16 short-period (1 Hz) vertical seismometers and 5 broad-band Guralp CMG-3ESPC seismometers, we here use data recorded by four of the latter instruments, named thereafter stations 1 to 4. Data from the remaining CMG-3ESPC could not be exploited, as it experienced acquisition problems. Station 4 was located on a different floe from the others. As such, its signal does not correlate as well with the ones recorded by stations 1 to 3. Since the method developed in this work requires a minimum of three stations, we will most of the time discard station 4 from the analysis. We however discuss in Sec. V the effect of adding station 4.

FIG. 1. Group velocity function of period, for flexural waves traveling in an uniform ice plate with varying thickness \( h \) as labelled on the graph. Elastic parameters were set to \( \rho = 910 \text{ kg/m}^3 \) (ice density), and \( \nu = 0.33 \) (Poisson coefficient). Two values of Young’s modulus are used: our preferred value of \( E = 7.2 \text{ GPa} \) (Stein et al., 1998) as shown in thick lines, and a larger value of \( E = 9.7 \text{ GPa} \), shown in thin line. This value was given by Pounder and Langleben (1964) for low porosity sea ice with less than 10% of brine content.

FIG. 2. (Color online) (Top) Relative positions, in meters, of the four seismic stations (labeled 1 to 4), along with approximate positions of the drill-hole (DH) and electromagnetic induction (EM) 800 m-long profile, and of the 2160 m-long EM profile, see text. The North direction varies with time as the ice floes drift along. (Bottom) Ice thickness from DH measurements along the ~800 m-long profile.
The stations were placed at distances of 620 m to 1580 m from each other: denoting by $X_i$ the position of station $i$ relative to the center of the array, we have that

$$X_1 = (-229, -558), X_2 = (386, -488), X_3 = (116, 96), \text{ and } X_4 = (-273, 949),$$

all distances being in meters, see Fig. 2.

The broad-band seismometers have a 60 s low frequency cut-off, and their signals were sampled at 100 Hz. We will analyze data acquired between the 27 April and the 25 May 2007, during which stations 1 to 3 operated together (663 hours overall). This duration reduces to 482 h if adding station 4. The network drifted along with the Tara base camp at roughly constant latitude (88°14′ to 88°32′) in the Amundsen basin, about 200 km east of the Lomonosov ridge. The water depth is ~4000 m there, significantly greater than the maximum wave lengths involved in this study (~500 m), justifying the assumption formulated in Sec. I of an infinitely deep water column. No detectable deformation occurred within this network during this time period. We here mostly exploit the vertical displacement rates recorded at these three stations, although results obtained with horizontal displacement rates are discussed in Sec. V.

During this experiment, an independent ice thickness dataset was collected at or close to the location of the array (Haas et al., 2011). On May 8, 2007, drill-hole (DH) measurements were performed every 5 m along a 800 m-long profile, see Fig. 2. An electromagnetic induction (EM) survey was conducted on the same day along the same line, for calibration purposes. Finally, and again at the same date, another EM survey was performed along a close-by 2160-m-long profile. The ice thickness was found to be largely variable along both lines, ranging between about 1.5 m (first-year undeformed ice) to 8 m (pressure ridges), with a mean of 2.70 m (DH) and 2.53 m (EM) for the first profile, and 2.75 m (EM) for the second profile. These mean values are obtained for two lines, and comparison with our estimated area-averaged thickness as detailed in Sec. V should therefore be taken cautiously. They are however likely to be characteristic of the ice floe the three stations 1, 2, and 3 were placed on. Moreover, it should be noted that very little change in ice thickness was observed between May 8 and June 26, as shown by a repeat of the survey conducted along the same 2160 m-long line (Haas et al., 2011).

III. ICE SWELL

The broad-band signal is strongly dominated by the ice swell. Computation of the amplitude spectrum gives a peak at periods ranging between 25 and 30 s, see Fig. 3. This spectrum shows some temporal evolution. Indeed, as already pointed out by Wadhams and Doble (2009), distant storms originating in the North Atlantic can generate periods of increased wave amplitude, that then propagate through the ice cover as flexural waves. Given the dispersion of these waves, the arrival time of the high amplitude wavetrains depends on their period.

A spectrogram at long periods, averaged over the three stations 1–3, is shown in Fig. 4. A clear dependence of the arrival time with period is observed at days 141–142 following the increase in significant wave height at latitudes greater than 60° in the North Atlantic Ocean. This dependence can be modeled by assuming a traveling distance of 1400 km in the open ocean and 1600 km through a sea ice cover with 2.7 m of mean ice thickness. A similar, but less remarkable feature is also observed at days 129–130. While an estimate of the ice thickness is feasible on the ground of

![Image](https://example.com/image.png)
IV. METHOD FOR ESTIMATING THE ICE THICKNESS

Here we detail the method developed for estimating the ice thickness from the cross-correlation functions of the vertical signals recorded at the stations. As explained in Sec. IV G, a minimum of 3 stations is required by this method. We therefore describe it for exactly 3 stations; generalization to a larger number of stations is straightforward, and the corresponding results are discussed in Sec. V A.

A. Imaging by correlating seismic noise

Correlating noise between two sensors allows the retrieval of the Green’s function that characterizes the propagation of elastic waves, even in the absence of a dominant source, like an earthquake in the case of crustal applications, to illuminate the medium (Weaver and Lobkis, 2001; Lobkis and Weaver, 2001; Campillo and Paul, 2003; Sabra et al., 2005). In the crust, noise with periods in the 10–20 s range has been shown to be primarily caused by ocean swell (Stehly et al., 2006). For Arctic studies, this is clearly the dominant source, as evidenced by the spectral content of the recorded waves, cf. Fig. 3, although the propagation of the swell in a cohesive ice-covered ocean shifts the peak periods to slightly larger values (25 to 30 s, in our case). The feasibility of using noise correlation methods to determine the dispersion of flexural waves, like those associated with the propagation of ice swell, has been evidenced in the case of lab-controlled experiments on plexiglass plates (Larose et al., 2007).

Further complexity arises in our study because of the anisotropy of the sources: the flexural waves come not from one privileged direction, nor from an homogeneous, i.e., isotropic, distribution of sources, but rather from a complex, time-varying distribution, partly controlled by the occurrence of distant storms in the North Atlantic, see Fig. 4. Anisotropic source distributions have been shown to cause asymmetric cross-correlation functions (Larose et al., 2005; Stehly et al., 2006; Froment et al., 2010), so that studying this asymmetry can help estimating the source distribution.

We here develop a method based on a model that reproduces both the temporal and the absolute features of the correlations between stations. It thus allows to invert the anisotropic source distribution, and, more importantly in our application, the flexural dispersion curve, hence the thickness of the ice cover.

B. Cross-correlation functions

We denote by $s_i(t)$ the vertical signal (displacement rate) recorded at station $i$. The mean Fourier spectrum for the three stations given by $\hat{S}(f) = (|\hat{s}_1(f)| + |\hat{s}_2(f)| + |\hat{s}_3(f)|)/3$, where the symbol ‘$\hat{}$’ denotes the Fourier transform, is shown in Fig. 3.

We first compute the normalized correlation functions $C_{ij}(t)$ between stations $i$ and $j$ with no pre-processing of the waveforms, by inverse Fourier transforming $\hat{C}_{ij}(f) = \frac{1}{N\sigma_i\sigma_j} \hat{s}_i(f)\hat{s}_j^*(f)$, with $N$ the number of samples ($N = 360000$ for our 1 h long records sampled at 100 Hz) and $\sigma_i$ the standard deviation of $s_i(t)$. The normalization by $N\sigma_i\sigma_j$ is done so that $C_{ij}(t)$ is effectively the linear correlation coefficient of $s_i$ and $s_j$, hence ranging between $-1$ and $+1$. Compared to previous works on the correlation of seismic noise that do not perform this normalization but instead normalize $C_{ij}(t)$ with $\max C_{ij}(t) = 1$, we here exploit the absolute rather than the relative values of $C_{ij}(t)$ as it contains important information that helps constraining the modeling of the complex wavefield.

For each pair $(i,j)$ of stations, we compute $C_{ij}(t)$ on 1 h long records, and average this function over all 663 records. Figure 5 shows the resulting cross-correlations, dominated by a 25 to 30 s period as expected given the Fourier spectrum of Fig. 3. Very similar $C_{ij}(t)$ functions were obtained for one-bit versions of the waveforms (Larose et al., 2004). A clear asymmetry is observed, at least for the 1–3 and 2–3 pairs.

In order to explore frequencies outside the 25–30 s peak, we define band-pass filters $G^{(T)}(f) = e^{-\Delta f^2 - (1/T)^2/2\Delta f^2}$ centered at frequency $1/T$ and with $\Delta f$ width. In the following, we will use six such Gaussian filters, with $T = \{4, 5, 7, 9, 12, 20\}$ s center periods, and $\Delta f = \{0.06, 0.05, 0.03, 0.03, 0.01, 0.01\}$ Hz, respectively. Band-pass signals are computed as $\hat{s}_i^{(T)} = G^{(T)}(f)\hat{s}_i(f)$: These filtered signals preserve the phase of the original signals, and their modulus are equal to $G^{(T)}(f)$. This whitening is done to force $s_i^{(T)}$ to have a frequency content effectively centered on $T$. 

FIG. 5. (Color online) Cross-correlation functions for the three pairs of stations, with no pre-processing. The envelopes are also shown.


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clearly observe that the correlation increases as the time-shifted version of the observed function as a whole, hence considering both the absolute level of correlation and its dependence with time, will allow us in Sec. IV D to estimate the ice thickness $h$.

Our model considers that the vertical signals at the stations are dominated by the ice swell, which consists in a complex wavefield made of a mixture of independent plane waves with varying incident angle $\theta$ relative to our station array. The fact that there is more than just one single incident angle is based on empirical evidence, as in this case the band-pass signals would be shifted in time from one station to the next with a shift depending on the angle between the azimuth of the plane wave and the axis defined by the two stations. The cross-correlation functions would then simply amount to a time-shifted version of $G^{(T)}(t)$, if no significant scattering occurs between the stations. Since this contradicts of the station pair relative to the incoming wavefield, $t_{\text{max,ij}}$ can be zero, yielding infinite apparent velocities. Figure 7 shows how the estimate of $v_G(T)$ varies with $T$, and compares this dependence with the theoretical dispersion curve for $h = 1$, $1.5$, and $2$ m. A mean thickness of $1.5$ m would correctly fit the observations. As we will see in later sections, this estimate is however of poor quality, and does not compare well with the DH and EM measurements. Very similar results are obtained when deconvolving the waveforms from one station by the waveforms recorded at another station, and band-pass filtering the deconvolved signal. This is likely due to the fact that this estimation assumes that the ice swell is a simple plane wave coming from a single direction. This would cause the cross-correlation functions to be strongly asymmetric, with a well-defined maximum close to $1$, time-shifted by the time it takes for this wave to propagate from the first to the second station. This model is here clearly not realistic. We therefore drop the assumption that the ice swell is a simple plane wave, and generalize it to the case of a mixture of plane waves with different incoming angles and amplitudes.

C. Model

We now describe how the observed cross-correlation functions $C^{(T)}_{ij}$ can be modeled. Finding the best fit between the observed and modeled functions as a whole, hence considering both the absolute level of correlation and its dependence with time, will allow us in Sec. IV D to estimate the ice thickness $h$.

FIG. 6. (Color online) Cross-correlation functions between stations 1 and 2, for three out of the six band-pass filters centered at $T = 4$ s, 7 s and 20 s. The envelopes are also shown.

The cross-correlations $C^{(T)}_{ij}(t)$ between stations $i$ and $j$ for the frequency band centered at $1/T$ are computed from $C^{(T)}_{ij}(f) = \frac{1}{N_0 \sigma_{ij}^{(T)}(f)} \sigma_{ij}^{(T)}(f)$, where $\sigma_{ij}^{(T)}$ is now the standard deviation of $s_{ij}^{(T)}(t)$. We therefore compute 18 cross-correlation functions (3 pairs of stations, 6 band-pass filters).

We show in Fig. 6 three of the cross-correlation functions between stations 1 and 2, at $T = 4$ s, 7 s and 20 s. We clearly observe that the correlation increases as $T$ increases and becomes closer to the peak period of the ice swell. This implies that, at too low a band-pass filter period $T$, the correlation becomes too weak, preventing us from using such periods in the estimation of $h$. Moreover, the maximum of the envelopes is seen to be restricted to short times at low $T$ values, and to spread to longer times at larger $T$. This is caused by the dispersive nature of the propagation of flexural waves, with propagation velocities decreasing as $T$ increases from 4 s to 7 s and to 20 s.

An estimate of $h$ can be simply determined using the time $t_{\text{max}}$ of this maximum of the envelope, as for example shown in Fig. 6. For the cross-correlation $C^{(T)}_{ij}$, we determine the time $t_{\text{max,ij}}$ such that

$$\max_{t} C^{(T)}_{ij}(t) = C^{(T)}_{ij}(t_{\text{max,ij}}).$$

(3)

The group velocity at period $T$ is then estimated as the minimum over the three pairs $\{i,j\}$ of $(|\vec{X}_i - \vec{X}_j|)/(t_{\text{max,ij}})$. We here keep the minimum, because, for favorable orientations...
the observations (Fig. 6), we instead assume the ice swell can come from all directions, with an amplitude $A(\theta)$ function of the incident angle, cf. Fig. 8. This mixing of plane waves can be explained by interactions (reflections) of an initial wave with changes in bathymetry or changes in ice thickness and sea-ice concentrations (Squire et al., 2009).

We denote the modeled cross-correlations using the tilde symbol $\tilde{C}_{ij}^{(T)}(t)$. Their Fourier transforms are

$$
\tilde{C}_{ij}^{(T)}(f) = \frac{|\tilde{G}_{ij}^{(T)}(f)|^2 |\tilde{A}_{ij}^{(T)}(f)|^2}{\int df |\tilde{G}_{ij}^{(T)}(f)|^2} d\theta A^2(\theta).
$$

with

$$
\tilde{A}_{ij}^{(T)}(f) = \int d\theta A^2(\theta) e^{2\pi i \Delta_{ij}^{(T)}(\theta)}.
$$

The time delay $\Delta_{ij}^{(T)}(\theta)$ between stations $i$ and $j$ at group velocity $v_{G}(T)$ for a harmonic plane wave of period $T$ with incident angle $\theta$ is simply

$$
\Delta_{ij}^{(T)}(\theta) = \frac{(X_i - X_j) \cdot \hat{u}_\theta}{v_{G}(T)},
$$

where $\hat{u}(\theta)$ is the unitary vector with incident angle $\theta$.

We emphasize here that direct inspection of the cross-correlation functions, as for example those shown in Fig. 6, is enough to conclude that $A(\theta)$ is neither constant with $\theta$ (uniform azimuthal distribution of the incoming ice swell), nor that it is peaked at a given value of $\theta$ (one dominant incoming direction). Indeed, for $A(\theta) = A_0 \delta(\theta)$, hence a uniformly distributed ice swell source, we have $\tilde{A}_{ij}^{(T)}(f) = 2\pi A_0^2 J_0 \left(2\pi f \frac{|X_i - X_j|}{v_{G}(T)} \right)$, where $J_0$ is the Bessel function of the first kind and order 0. Then $\tilde{C}_{ij}^{(T)}(f)$ is real, and consequently the cross-correlation function is symmetric in time: $\tilde{C}_{ij}^{(T)}(-t) = \tilde{C}_{ij}^{(T)}(t)$. This symmetry is not observed in Fig. 6, showing that the ice swell azimuthal distribution $A(\theta)$ is non-uniform. As an example, we display in Fig. 9 the cross-correlation $\tilde{C}_{12}^{(T)}(t)$ for $T = 20$ s that would be obtained if the distribution $A(\theta)$ were uniform.

Moreover, if $A(\theta) = A_1 \delta(\theta - \theta_1)$, i.e., the ice swell being a simple plane wave coming from direction $\theta_1$, then $\tilde{C}_{ij}^{(T)}(t)$ would be the inverse Fourier Transform of $\frac{|\tilde{G}_{ij}^{(T)}(f)|^2}{\int df |\tilde{G}_{ij}^{(T)}(f)|^2}$ shifted by the time it takes for the wave to propagate from station $i$ to station $j$, hence $\Delta_{ij}^{(T)}(\theta_1)$. In particular, the cross-correlation would then have a peak value equal to 1. This is not observed, cf. Fig. 9 in the case of $\tilde{C}_{12}^{(T)}(t)$ for $T = 20$ s obtained with a single plane wave coming with angle $\theta_1 = 20^\circ$, and a 2.5 m thick ice cover.

Finally, we note that the cross-correlation between two stations remains unchanged with varying $\theta_1$ in the case of $A(\theta) = A_1 \delta(\theta - \theta_1)$, as long as $v(T)/\cos \theta_1$ is constant. This trade-off between $v(T)$ and $\cos \theta_1$ thus requires to first determine $\theta_1$; as a consequence, the method requires at least 3 stations to correctly estimate both $A(\theta)$ and $h$.

**D. Inversion**

The modeled cross-correlation functions are parameterized by the amplitude density $A(\theta)$ and the ice thickness $h$ which controls the group velocity $v_{G}(T)$. The thickness is the key unknown quantity, and $A(\theta)$ only a by-product of our inversion. The quality of the fit provided by a model \{\hat{A}(\theta), h\} is measured by a quadratic cost function $J(A, h) = \sum_i \sum_{t} (\tilde{C}_{ij}^{(T)}(t) - \tilde{C}_{ij}^{(T)}(\theta_1))^2$ over the time interval $-150 < t < 150$ s. Note that the cost function is computed using all 18 cross-correlation functions at once.
We proceed as follows:

1. We test regularly spaced values of $h$, for example $h = 0.1$ m, $0.2$ m, ...
2. For each value of $h$, we invert $A(\theta)$ by minimizing $J$: $J(h) = \min_{A} J(A, h)$
3. We search for the minimum of $J(h)$ among all tested values of $h$.

### E. Test for uniform ice thickness

We test the method by inverting synthetic data. We construct the synthetic cross-correlation functions using the same station positions as the actual array, the same bandpass filters as with the real data analysis (see Sec. V for their characteristics), and impose a random azimuthal distribution $A(\theta)$ with the incident angle $\theta$ discretized in 9 windows $[0, 40^\circ], [40^\circ, 80^\circ])$, hence $\Delta \theta = 40^\circ$. The propagation velocity is computed using an ice thickness $h = 2.5$ m.

It is impossible to a priori know from the seismic data how the amplitude distribution $A(\theta)$ should be discretized. The discretization step $\Delta \theta$ must however be specified in our inversion scheme. It seems preferable to have a small $\Delta \theta$, but the smaller this value, the greater the number of amplitude $A$ values to be inverted, and the longer the computation time. Moreover, for any given $\Delta \theta$, spurious local or even global minima can be found, that are however not robust when changing $\Delta \theta$. We thus adopt the strategy of using several values of $\Delta \theta$; namely, we run 12 independent inversions using different discretizations, and average them all. We use $\Delta \theta = 20^\circ$, $40^\circ$, and $60^\circ$. For each value of $\Delta \theta$, the “first” window is $[\delta \theta, \delta \theta + \Delta \theta]$ with $\delta \theta/\Delta \theta = 0, \frac{1}{2}, \frac{1}{3}, \frac{1}{4}$.

Figure 10 shows the cost function $J(h)$ for these various discretizations, and for the overall average. The minimum of $J$ is indeed found for $h = 2.5$ m, whatever the discretization step $\Delta \theta$. The inversion is therefore well able to accurately estimate the ice thickness. We reproduce the same test for two other synthetics, after changing the input ice thickness to $h = 1.5$ m and $h = 4$ m. The averaged cost functions shown in Fig. 11 indicate that, in the first case $h = 1.5$ m, an ambiguity exists as two well-marked minima are found (at $1.5$ m and $2.2$ m), although the latter is only a local minimum. In the second case $h = 4$ m, we find a single minimum, giving an estimate of $3.8$ m.

Running 100 simulations with $h = 2.5$ m, and each time inverting for $h$, gives an estimate $2.5 \leq h \leq 2.8$ m, with a mean of $2.62 \pm 0.09$ m. We also tested our hypothesis that $A(\theta)$ is independent of $T$; given the limited range of 4–20 s used in this analysis, we do not expect strong changes in the source nor the reflector distributions with varying $T$. Nevertheless, we ran 100 simulations with $A(\theta)$ randomly drawn separately for all 6 analyzed periods, again with an imposed thickness of $2.5$ m. We obtain an estimate of $h$ ranging

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**Figure 10.** (Color online) Cost function $J$ vs ice thickness $h$, for the synthetic test with input $h = 2.5$ m. The angle distribution $\theta$ is discretized with $\Delta \theta$ equal to $20^\circ$, $40^\circ$, or $60^\circ$. The cost function obtained by averaging these three cost functions shows a clear minimum at the expected $2.5$ m thickness.

**Figure 11.** (Color online) Cost functions obtained by averaging the three cost functions with $\Delta \theta$ equal to $20^\circ$, $40^\circ$, or $60^\circ$, for three synthetic tests constructed with $h = 1.5$ m, $h = 4$ m, and a model with variable ice thickness, see text.
between 2.5 m to 3.1 m, with a mean of 2.81 ± 0.13 m. This unrealistically strong dependence of \( A(\theta) \) on \( T \) thus causes a slight over-estimation of \( h \).

**F. Test for variable ice thickness**

The previous test assumed that the ice thickness is uniform, both in the construction of the synthetics and in the inversion itself. This is a strong assumption, as pack ice typically exhibits an heterogeneous distribution, e.g., Wadhams et al. (2006). For the EM and DH profiles conducted close to our seismic stations, the ice thickness has been found by Haas et al. (2011) to vary from 1.5 to 8 m, with a standard deviation of 1.02 m (DH) and 1.11 m (EM, long profile) on May 8, 2007.

We therefore test the inversion, that assumes constant ice thickness, with synthetics constructed with variable ice thickness. We generate the synthetics with a very crude modelling, that consists in using ice thicknesses that are constant for each transect connecting two stations, but variable from one transect to the next. We here take \( h = 2.5 \) m for the transect between stations 1–2, \( h = 4 \) m for 1–3, and \( h = 3 \) m for 2–3.

Figure 11 shows that a global minimum is found at \( h = 3.0 \) m, instead of local minima at the three values \( h = 2.5, 3, \) and \( 4 \) m. This estimated thickness is not equal to the mean thickness of \( h = 3.17 \) m, although it is not far. Running a series of such tests, by changing the distribution \( A(\theta) \), shows that the minimum of \( J(h) \) is always within 0.2 m of \( h \) and is most of the time smaller than it.

These tests therefore indicate that the method is able to give a correct estimate of the mean \( h \), at least with this network geometry, with a typical error not greater than 0.2 m. The quality of the inversion is however likely to depend on the thickness variability: since the inversion assumes uniform thickness, increasing this variability makes the model less adequate to explain the data.

**G. Requirements on the network configuration**

We investigate how the spacing between the three stations can be optimized in order to provide a good resolution on \( h \). If the stations are too close to one another, then changes in the distribution of \( A(\theta) \) will cause very little changes in the cross-correlation functions. To avoid this lack of discrimination, a minimum spacing should therefore be imposed between the stations, that depends on the wavelengths of the ice swell at the peak frequencies of the band-pass filters, ranging from \( \sim 250 \) m to \( \sim 320 \) m for \( T \) between 4 s and 20 s, and \( h = 2.5 \) m. To find this spacing, we simulate the cross-correlation functions for an equilateral triangle of stations separated by distance \( \ell \), and check how these functions change when changing the distribution \( A(\theta) \). Specifically, we compute the variance of the cross-correlation function at each time sample \( t \) in the \( -150 \) to 150 s interval, for independent realizations of random \( A(\theta) \). The variance averaged over all time samples \( t \) defines the sensitivity of the network. Too low a sensitivity means that the same cross-correlation functions are always obtained whatever \( A(\theta) \), and therefore that too little information can be extracted from the data to constrain \( A(\theta) \), hence also \( h \).

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**Figure 12** shows this sensitivity, for \( \ell \) varying between 1 m and 10 km, and for \( h \) between 1 and 5 m. We here again use the same band-pass filters as defined above, hence with peak period ranging between 4 and 20 s. The sensitivity is maximum when the spacing is near the minimum wavelength (between \( \sim 100 \) m, for \( h = 1 \) m, to \( \sim 400 \) m, for \( h = 5 \) m) at the tested periods 4–20 s. A spacing of at least 100 m is thus required in order to invert the ice thickness using the ice swell, whatever \( h \). No strong gain in sensitivity is found as \( \ell \) increases past this 100 m value. Note that this analysis does not account for the scattering of the ice swell, that will result in a loss of correlation between stations that are far from one another, so that too large a spacing should be avoided.

**V. ANALYSES**

**A. Ice thickness**

We invert the cross-correlation functions as of Fig. 6, that were obtained for the 27 day-long period extending between April 27 and May 25, 2007. As with the test, we use three different angle discretization steps: \( \Delta \theta = 20^\circ, 40^\circ \) and \( 60^\circ \), each one with four different initial window (see above). We show in Fig. 13 the three averaged cost functions, and the overall cost function averaged over all 12 distinct angle discretizations. A clear global minimum is found for \( h = 2.5 \) m, which is coherent with the 2.70 m and 2.53 m mean ice thickness found by the drill-hole and the EM measurements, respectively.

The best fit for the estimated \( h = 2.5 \) m, and taking \( \Delta \theta = 20^\circ \), is shown in Fig. 14 for the 3 (out of 18) cross-correlation functions of Fig. 6. The quality of the fit is very good, especially at long periods \( T \). For this particular discretization of the angles \( \theta \), the minimum of \( J \) is obtained for \( h = 2.6 \) m, a value slightly different to the 2.5 m thickness that minimizes the cost function \( J \) averaged over all 12 angle discretizations.

The corresponding amplitude distribution \( A(\theta) \) is shown in Fig. 15. The distribution is complex, as already suggested by the cross-correlation functions (see Sec. IV B). It is roughly symmetric, with most of the ice swell energy coming from a direction aligned on the Transpolar Drift, hence
waves coming from the Siberian Sea and from the North Atlantic Ocean through the Fram Strait.

To test the robustness of the inversion, we use this azimuthal distribution $A(\theta)$ as shown in Fig. 15, and compute the cost function individually for the 3 pairs of stations 1–2, 1–3 and 2–3, with varying ice thickness $h$. If the inversion is indeed robust, then all 3 pairs should show a clear minimum at the same value of $h$, proving that this best model is indeed able to explain all cross-correlation functions. Figure 16 shows that the minimum at 2.6 m is effectively robust, with very little dispersion when considering the 3 pairs of stations individually.

Adding station 4 and reproducing the same analysis, still using $A(\theta)$ of Fig. 15, shows a rather different picture: the minimum of 2.6 m is not robust anymore, as evidenced by the large dispersion in the individual cost function values, see Fig. 16. The distribution $A(\theta)$ and the propagation velocities for $h = 2.6$ m therefore cannot explain all 6 cross-correlations obtained for the 4 stations. This is likely due to a significant change in mean ice thickness for the distinct floe station 4 is located on. We thus conclude that, while adding extra stations could potentially improve the ice thickness estimate, by providing better constraints on the dispersion curve, it can also degrade the quality of the estimate, as is the case here. It is thus important to favor stations located
on a unique floe when using this method. Use of small aperture broad-band arrays deployed on the same floe could help better constraining the azimuthal distribution $A(\theta)$.

Similar analyses on shorter durations, down to only 24 hours, give that $h = 2.5$ m without any resolvable changes (i.e., greater than ±0.2 m) over the May–June 2007 period. Haas et al. (2011) have shown that the ice thickness shrinks by only a few cm over this time period. This change is less than the error of 0.2 m on our estimate, so that it is best to run the analysis on the whole 663 h to maximize the robustness of the estimation.

B. Use of horizontal channels

A similar analysis was performed using the horizontal channels of the three stations. Those horizontal channels record horizontal displacement rates in two orthogonal directions. As is detailed in Marsan et al. (2011), these recordings show highly coherent episodic low-frequency wavetrains, that were named Low Frequency Bursts (LFB), and which are likely the seismic signature of remote, large-scale shear deformation episodes along major leads. Because of these LFBs, the cross-correlation of horizontal channels is greater than for the vertical displacement rates analyzed so far, see Fig. 17. Moreover, the maximum cross-correlation is obtained for time delays less than 1 s, cf. Fig. 18.

A gain in correlation is potentially interesting, as it gives a better signal to noise ratio, and should therefore reduce the uncertainty in the estimated $h$. However, LFBs are purely horizontally polarized, and travel at much faster velocity than the ice swell. Although it was not formally proved, several observations suggest that these wavetrains are SH waves, which in the sea-ice cover should propagate non-dispersively at 1600 to
1800 m/s (Marsan et al., 2011). The LFB occurrences being intermittent, the cross-correlations of the horizontal channels average over “quiet” periods, for which the ice swell is dominant, and more energetic periods dominated by LFB occurrences. This mixing of two very different wave phases and propagation velocities explain the general shape of the cost function $J(h)$ for the horizontal channels, see Fig. 19: a local minimum at about 2.8 m, that is possibly a signature of the ice swell dispersion curve, and a global minimum at large $h > 5$ m. For such large thickness, the flexural wave velocities are large (e.g., $v_G = 137$ m/s for $T = 4$ s and $h = 7$ m), and the model is forced to such values in order to explain the apparent very short time delays between the horizontal channels of the three stations.

In order to exploit horizontal channels for measuring the ice thickness, it is therefore necessary to remove the signal due to LFBs. We do this by following the approach described in Marsan et al. (2011), that is based on the observation that LFBs typically last several minutes and are clearly seen as anomalously high amplitude wavetrains on the horizontal channels. We thus consider that a 1 h long signal is LFB-free for all three stations if, for the 30 consecutive 2 mn-long windows that can be extracted from this hour, the ratio $R$ of the standard deviation of one horizontal channels with the standard deviation of the vertical channel is never greater than a threshold value $R_c$. We fix this threshold to $R_c = 3$, cf. Marsan et al. (2011). We find that, out of the 633 h used to compute the 6 cross-correlation functions, 281 contain LFBs, and thus 382 are LFB-free. Figure 18 shows that the cross-correlation obtained with horizontal channels during LFB-free periods is very similar to those found with vertical channels during all the 663 h (with and without LFBs), as expected for the flexural mode of propagation of the ice swell. It is therefore easy to remove the influence of the LFBs from the horizontal channels, although there is no particular gain obtained by using these channels, compared to only using vertical channels as we have done so far.

VI. CONCLUSIONS

The feasibility of estimating the area-averaged ice thickness at intermediate scale (100 m to several kms) using seismic instruments with no active sources has been demonstrated. The ice swell signal is exploited, providing a natural source for analyzing the dispersion of flexural waves with frequency. Compared to active source experiments, this however requires to study a complex wavefield, made of waves coming from different angles rather than from a single source location. This makes for a delicate analysis, and generate uncertainties in the estimated ice thickness. In the case of the Spring 2007 Tara dataset analyzed here, we estimate this uncertainty to be less than 0.2 m, as the minimum of the cost function $J(h)$ shows variations with such an amplitude when exploring the parameter space (azimuthal angle discretization) relevant to this inversion. This uncertainty is also coherent with the typical error found with synthetic tests.

The method requires a minimum of 3 stations, separated by at least 100 m of each other. Too large a spacing can, however, become problematic, as damping and scattering of the wavefield would lead to a lesser correlation between the stations. Moreover, it is necessary to avoid the deployment of stations on distinct floes, as shown by the loss in robustness in the inversion when introducing station 4. This requirement implies that rather compact networks should be favored if using this method. Broad-band seismometers sensing vertical displacements of the ice cover are well designed for this treatment, although horizontal displacements yield similar information if episodic, high amplitude wavetrains (LFBs) related to remote icequakes are removed from the analysis.

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