

Multi-frequency SuperDARN interferometer calibration

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Key Points:

- Calibration of SuperDARN interferometer angle of arrival data remains an outstanding challenge
- Analysis of data at multiple frequencies can overcome the inherent 2π measurement ambiguity in the phase correction factor (t_{diff})
- Example applications are shown for historical and contemporary multi-frequency SuperDARN observations

Abstract

The ground-based, high-frequency radars of the Super Dual Auroral Radar Network (SuperDARN) observe backscatter from ionospheric field-aligned plasma irregularities and features on the Earth’s surface out to ranges of several thousand kilometers via over-the-horizon propagation of transmitted radio waves. Interferometric techniques can be applied to the received signals at the primary and secondary antenna arrays to measure the vertical angle of arrival, or elevation angle, for more accurate geolocation of SuperDARN observations. Calibration of SuperDARN interferometer measurements however remains challenging for several reasons, including a 2π ambiguity in the phase correction factor needed to account for differences in the electrical path lengths between signals received at the two antenna arrays. We present a new technique using multi-frequency ionospheric and ground backscatter observations for the calibration of SuperDARN interferometer data, and demonstrate its application to both historical and recent data.

1 Introduction

The Super Dual Auroral Radar Network (SuperDARN) is an international collaboration of ground-based, high-frequency (HF) radars which monitors space weather conditions in the Earth’s ionosphere and lower atmosphere (Greenwald et al., 1995; Chisham et al., 2007; Nishitani et al., 2019). Most SuperDARN radars have a smaller interferometer antenna array, displaced from the main antenna array by ~ 60 – 185 m, to measure the vertical angle of arrival, or elevation angle, of the received radar signals (Shepherd, 2017). These elevation angle measurements are important for the accurate geolocation of SuperDARN backscatter observations, both from field-aligned ionospheric irregularities for mapping two-dimensional plasma flow (e.g., Ruohoniemi & Baker, 1998; Britton et al., 2022) and from the Earth’s land and sea surfaces (Greenwald et al., 2017).

In SuperDARN radar systems, the physical separation between the main and interferometer antenna arrays introduces a phase delay in the signals that are received by each antenna that depends on the elevation angle. In addition, there are typically differences in the electrical path lengths from the two antenna arrays to the point at which the return signals are correlated. The corresponding difference in the signal travel time along each path, known as t_{diff} , causes a systematic offset in the measured phase difference between the signals from the two arrays, which must be accounted for (Chisham et al., 2021).

While it is possible to measure these electrical path differences, it can be challenging without specialized test equipment and cannot be performed for historical datasets. Several methods have therefore been developed to estimate t_{diff} using ionospheric scatter (IS) or ground scatter (GS) measurements assumed to follow known propagation characteristics, e.g., from meteor trails (Chisham & Freeman, 2013; Chisham, 2018), $\frac{1}{2}$ -hop E -region IS (Ponomarenko et al., 2018), 1-hop F -region GS (Ponomarenko et al., 2015), or targets with a known ground location such as artificially generated irregularities (Burrell et al., 2016). However, because SuperDARN radars typically operate at only one or two frequencies, these techniques are susceptible to a 2π ambiguity when estimating t_{diff} and therefore may not identify the true value which is valid across all possible operating frequencies (assuming one exists).

To address this issue, we have extended the “virtual height method” described by Chisham et al. (2021) to utilize multi-frequency observations of both IS and GS echoes for the automated calibration of t_{diff} . The methodology of this multi-frequency approach is described in section 2. In sections 3 and 4 we demonstrate the application of this technique to historical SuperDARN data, as well provide contemporary t_{diff} estimates for selected radars. In section 5 we discuss our results from applying the technique.

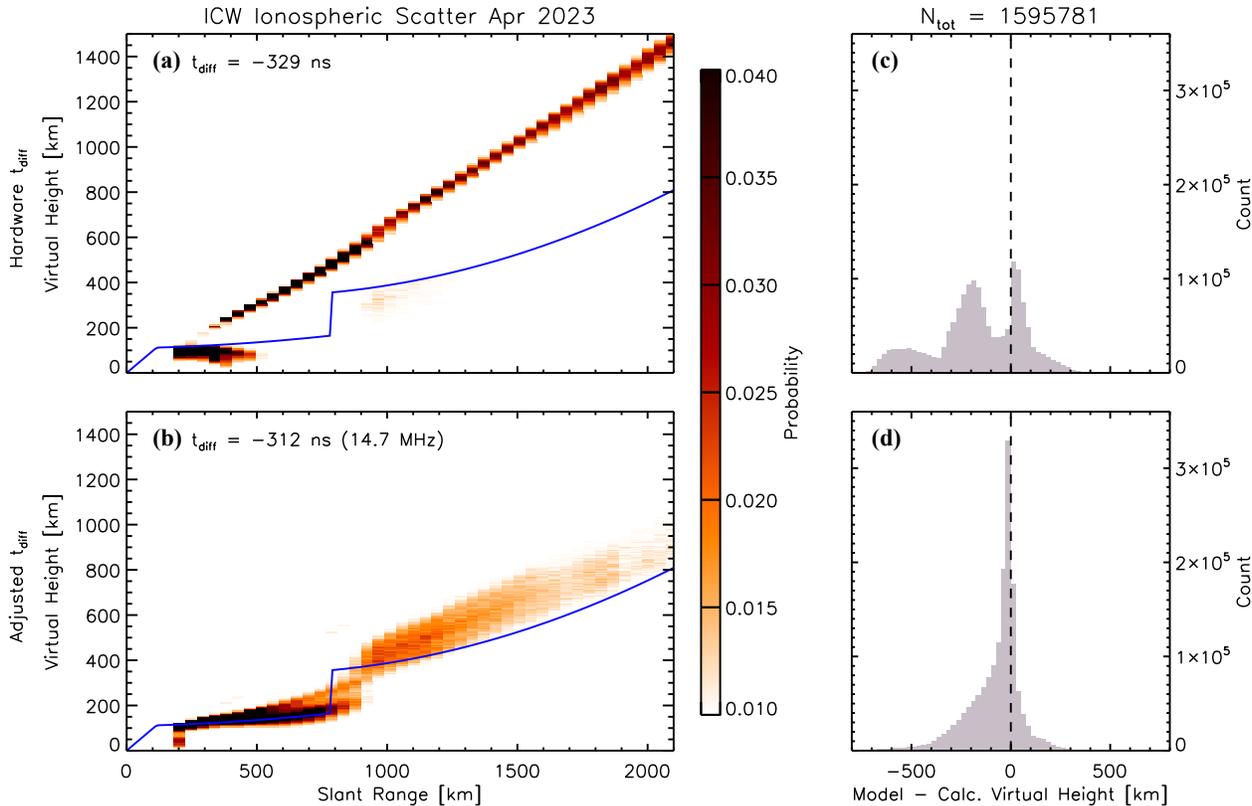


Figure 1. Virtual height comparison method for the Iceland West (ICW) SuperDARN radar when using the initial t_{diff} estimate (-329 ns) and the t_{diff} found by adjusting the $\frac{1}{2}$ -hop ionospheric backscatter to match the Chisham VHM (-312 ns at 14.7 MHz), after Chisham et al. (2021). Panels (a) and (b) show two-dimensional histograms of the slant range and virtual height distribution of the ICW data from the entire month of April 2023 for ranges 0–2100 km, with the Chisham VHM overlaid in blue. Panels (c) and (d) show histograms of the difference between the modeled and measured virtual heights at these same ranges.

60 **2 Methodology**

61 The virtual height comparison method introduced by Chisham et al. (2021) is based
 62 on the presumption that, for a correct t_{diff} value, the observed elevation angle (or vir-
 63 tual height) variations with range should nominally match the behavior of an empiri-
 64 cal virtual height model (VHM). These VHMs are the default method for geolocation
 65 of line-of-sight (LOS) SuperDARN observations and are required for use at radar sites
 66 which either do not have an interferometer antenna array or t_{diff} has not yet been cali-
 67 brated (e.g., Chisham et al., 2008; Thomas & Shepherd, 2022). Chisham et al. (2021)
 68 demonstrated how, for $\frac{1}{2}$ -hop IS, an obviously incorrect t_{diff} value can be adjusted un-
 69 til the observed distribution of slant range and virtual height align with the VHM of Chisham
 70 et al. (2008) (hereafter referred to as the Chisham VHM).

71 As an example, we apply this virtual height calibration technique to the recently
 72 constructed Iceland West (ICW) SuperDARN radar which, along with the co-located Ice-
 73 land East radar, saw first light in January 2023. An initial t_{diff} estimate of -329 ns was
 74 determined for ICW based on the difference in coaxial cable lengths from the main and
 75 interferometer antenna arrays to a central equipment shelter housing the radar electron-
 76 ics. Figure 1a shows the joint probability distribution of IS observed by ICW during the

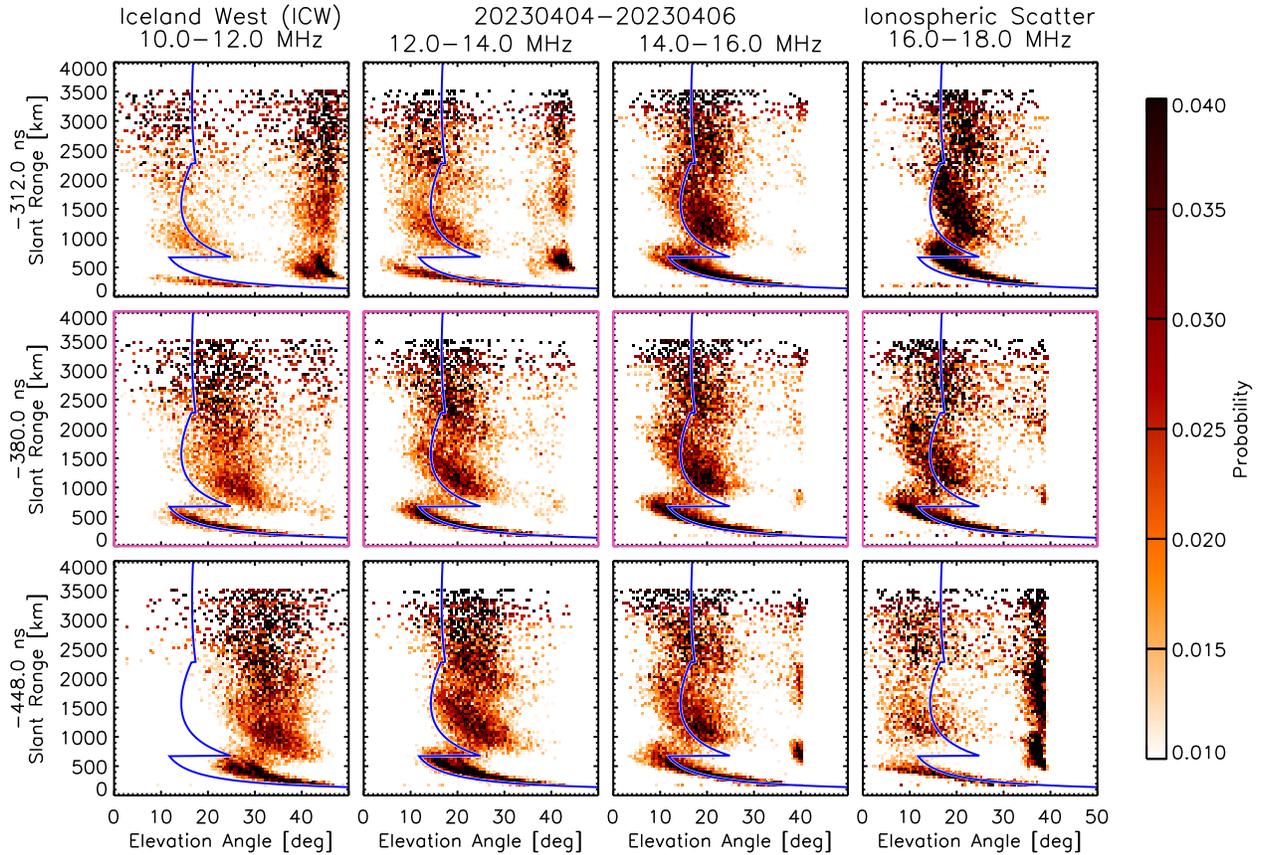


Figure 2. Joint probability distributions of elevation angle and slant range for ionospheric scatter observed by ICW on 4–6 April 2023 in 0.5° elevation and 45 km range bins, sorted by radar operating frequency into four bands between 10 MHz (left column) and 18 MHz (right column) and calculated using three values of t_{diff} : -312 ns (top row), -380 ns (middle), and -448 ns (bottom). The Chisham et al. (2008) empirical virtual height model is overlaid on each panel in blue.

entire month of April 2023 at a single frequency (14.7 MHz) divided into 45 km range and 5 km height bins, with the Chisham VHM overlaid in blue. Note the distributions are normalized by the maximum occurrence at each range bin, after Chisham et al. (2008, 2021).

Beyond ~ 500 km range, all of the data in Figure 1a are aliased to very large and linearly increasing virtual heights, suggesting the initial t_{diff} estimate is incorrect. The histogram in Figure 1c shows that the distribution of differences between the VHM and calculated virtual heights has multiple peaks: the peak near zero is likely due to the near-range *E*-region echoes where the virtual heights are small, while the other two peaks at larger negative values are likely attributable to the $\frac{1}{2}$ -hop *F*-region echoes at farther ranges where the calculated virtual heights are much greater than the model suggests. Figure 1b shows the same data using a revised t_{diff} value of -312 ns where the calculated virtual heights now align much more closely with the Chisham VHM predictions in both the $\frac{1}{2}$ -hop *E*- and *F*-region regimes. The corresponding histogram in Figure 1d confirms this improved agreement with a single-peaked distribution located maximizing near zero and a tail towards negative values. This tail can again be attributed to $\frac{1}{2}$ -hop *F*-region echoes with slightly greater virtual heights than the empirical Chisham VHM predicts.

In Figure 2 we consider multi-frequency data observed by ICW with a special sounding mode over three days in April 2023, with each of the four columns corresponding to a 2 MHz frequency band between 10–18 MHz. Note the ICW radar was actually operating at 8 equally-spaced frequencies from 9.6–16.6 MHz at 1 MHz increments; more details about this special sounding mode are provided in section 4. Each panel in Figure 2 shows joint probability distributions of the observed IS divided into 0.5° elevation and 45 km range bins. A different t_{diff} value is used in each row to calculate the elevation angles: -312 , -380 , and -448 ns. Note that -312 ns (top row) was the value previously identified with the virtual height comparison method in Figure 1. Overlaid on all of the panels in blue is the Chisham VHM; horizontal segments indicate transitions between the three propagation modes described by this model (i.e., $\frac{1}{2}$ -hop E -region, $\frac{1}{2}$ -hop F -region, and $1\frac{1}{2}$ -hop F -region backscatter).

In the third column of Figure 2 (14–16 MHz) the elevation-range distributions appear nearly identical regardless of which of the three t_{diff} values are used. Specifically, the near-range echoes in all three panels closely follow the empirical VHM curve corresponding to $\frac{1}{2}$ -hop backscatter from either meteor trails or E -region irregularities. However for the other frequency bands, the elevation-range distributions in each of the three rows no longer agree with one another. Only when calculating elevation angles using the t_{diff} value from the center row (-380 ns, pink outline) do the distributions behave similarly at each frequency band while also nominally aligning with the Chisham VHM. Note the t_{diff} values applied to each row of Figure 2 are a multiple of 68 ns, or one wavelength at 14.7 MHz (i.e., near the center of the 2 MHz frequency band in the third column), to illustrate the 2π ambiguity when estimating t_{diff} .

While this tuning procedure may be performed manually to find the optimal t_{diff} value for all frequency bands, in practice an automated procedure is desirable given the large dataset from the more than 30 continuously operating SuperDARN radars. Here we describe a simple iterative procedure which has been found to quickly converge to an appropriate t_{diff} , examples of which are shown in the following sections:

1. For each frequency bin, calculate elevation angles observed by a given radar using the generalized algorithm of Shepherd (2017) for $-1000 < t_{\text{diff}} < +1000$ ns, at steps of 5 ns
2. Find the median difference between an empirical VHM and the calculated elevation angles for each t_{diff} at each frequency
3. Identify the t_{diff} for which the median elevation angle difference is a minimum across all frequency bins
4. Repeat steps 1–3 using a new search range of ± 10 ns centered about the previously identified t_{diff} value, at steps of 1 ns

Note there are currently three available SuperDARN VHMs: the “standard” VHM, the Chisham VHM, and the Christmas Valley VHM (hereafter referred to as the CV VHM). Both the standard and Chisham VHMs are restricted to IS only, while the CV VHM has separate models for IS and GS propagation modes (Thomas & Shepherd, 2022). In the next sections we use the CV VHM rather than the Chisham VHM to allow inclusion of both IS and GS observations in our analysis.

3 Common Mode Data

To our knowledge, the earliest available multi-frequency SuperDARN observations with quality interferometer data were collected by the Goose Bay radar (GBR) during standard operations from Jan–Mar 1994 on four frequency bands: 11.5–12.0 MHz, 12.3–13.0 MHz, 14.4–14.9 MHz, and 16.4–16.9 MHz. Figure 3 shows joint probability distributions of (top) IS and (bottom) GS observations by GBR during this 3-month inter-

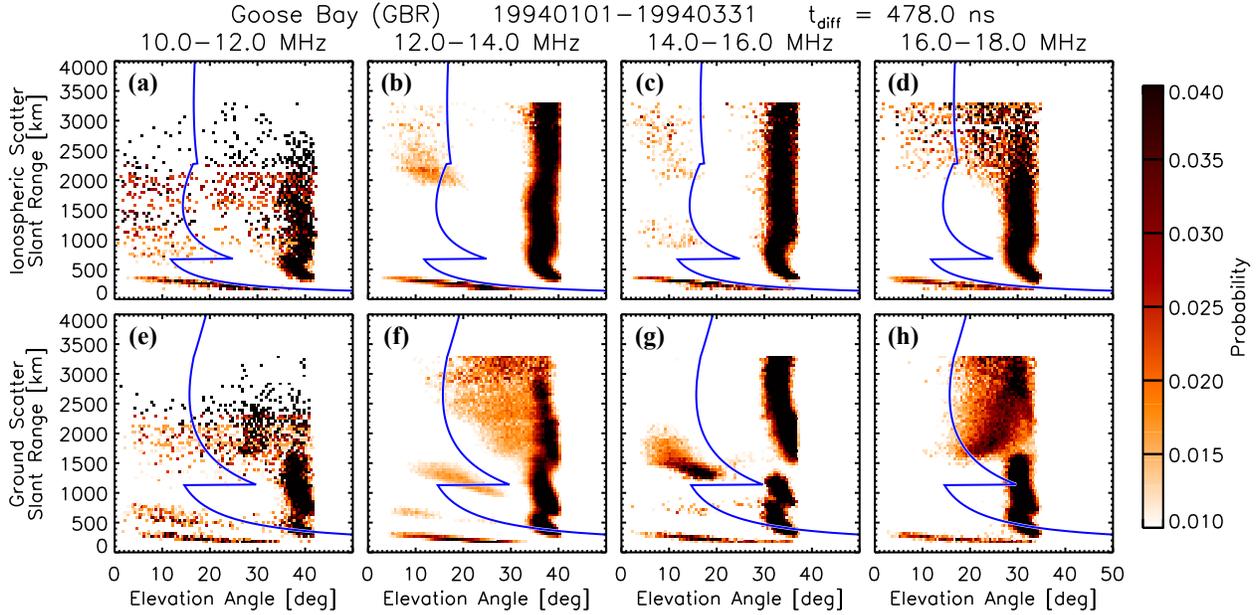


Figure 3. Joint probability distributions of elevation angle and slant range for (top) ionospheric scatter and (bottom) ground scatter observed by the Goose Bay (GBR) SuperDARN radar from Jan–Mar 1994 in 0.5° elevation and 45 km range bins, sorted by radar operating frequency into four bands between 10–18 MHz, using the current hardware t_{diff} value of 478 ns. The Christmas Valley empirical virtual height model is overlaid on each panel in blue (Thomas & Shepherd, 2022).

143 val divided into 0.5° elevation and 45 km range bins and normalized in the same fash-
 144 ion as Figure 2. Each column corresponds to a different frequency band, and the empir-
 145 ical CV VHM is overlaid on each panel in blue for reference.

146 The elevation angles in Figure 3 are calculated using the t_{diff} value of 487 ns listed
 147 for GBR in the SuperDARN analysis software for these dates (SuperDARN Data Anal-
 148 ysis Working Group et al., 2022). While observations from a high-latitude radar such
 149 as GBR are not expected to perfectly align with the empirical CV VHM (which was de-
 150 rived from mid-latitude observations), the data and model curves do not agree for any
 151 of the four frequency bands. In fact, all but the nearest range data are aliased to very
 152 high elevation angles, which Chisham et al. (2021) demonstrated to be a likely symp-
 153 tom of an incorrect t_{diff} value (e.g., Figure 1a).

154 Applying the multi-frequency calibration technique suggests a t_{diff} value of 457 ns,
 155 the result of which can be seen in Figure 4. Here we see the IS and GS measurements
 156 are now much better aligned with the empirical CV VHM at all four frequency bands,
 157 with a few notable exceptions. The GS distributions in both the 10–12 MHz (Figure 4e)
 158 and 16–18 MHz (Figure 4h) bands appear “mirrored” for ranges beyond 1500 km and
 159 elevation angles above 20° . These regions are likely associated with observations from
 160 the backlobe, which is a well-known characteristic of the log-periodic antenna design used
 161 by the GBR radar (e.g., Milan et al., 1997; André et al., 1998).

162 4 Sounding Mode Data

163 Hughes et al. (2002) developed a special sounding mode for SuperDARN radars
 164 which utilized the ~ 12 s of down-time available at the end of each standard 1 min az-

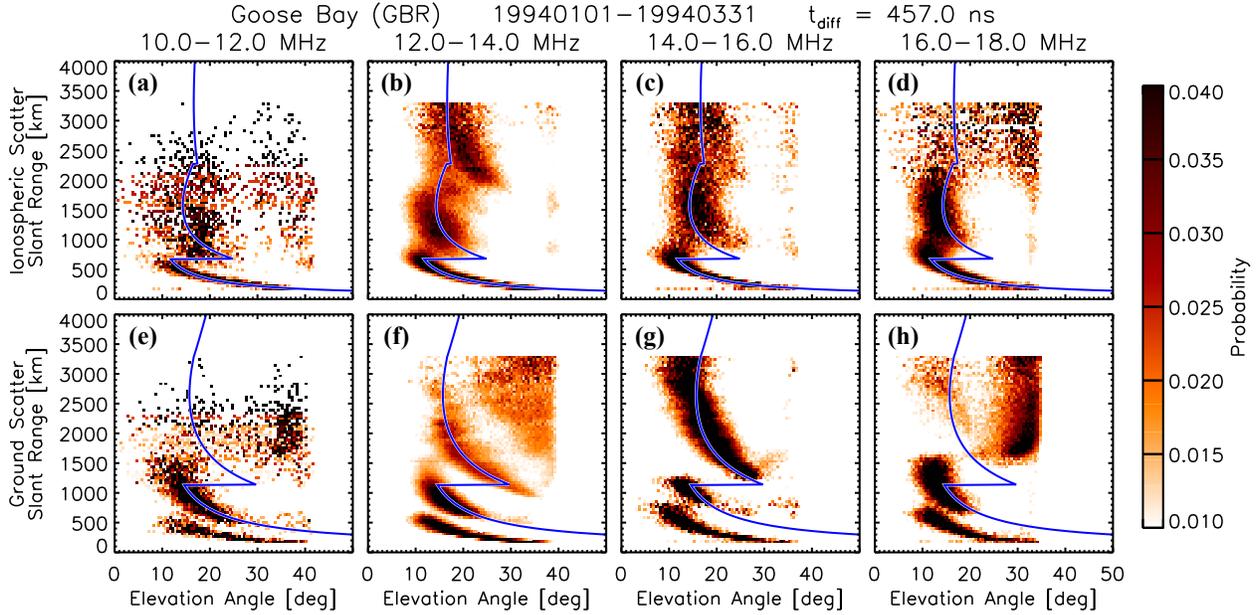


Figure 4. Joint probability distributions of elevation angle and slant range for (a-d) ionospheric scatter and (e-h) ground scatter observed by GBR in the same format as Figure 3, using a revised t_{diff} value of 457 ns.

165 imuthal scan to step through a range of ~ 8 equally-spaced frequencies in the 8–20 MHz
 166 band. A complete scan across all radar beams and sounding frequencies can be obtained
 167 every ~ 15 –30 min with this mode, and is therefore an ideal candidate for the multi-frequency
 168 calibration approach. Unfortunately, there are several challenges associated with the original
 169 sounding mode data files, including their minimally documented binary format and
 170 limited distribution. Furthermore, the elevation angles stored in the sounding data files
 171 were calculated using on-site processing software at the radars, seemingly preventing any
 172 calibration using post-processing techniques (e.g., Chisham et al., 2021).

173 Nevertheless, sounding mode files from a subset of radars have been recovered for
 174 various times over the years 2005–2018. By making an educated assumption about the
 175 original t_{diff} and interferometer offset values applied by the on-site FITACF software,
 176 one can convert the recorded elevation angles back to the observed phase differences (ψ_{obs})
 177 such that post-processing calibration techniques can be applied in pursuit of a different
 178 t_{diff} . Using the notation of Shepherd (2017), this conversion takes the form

$$179 \quad \psi_{\text{obs}} = 2\pi f_{\text{Tx}} \left[\frac{Y}{c} (\cos^2 \phi_0 - \sin^2 \alpha)^{\frac{1}{2}} - t_{\text{diff}} \right] \quad (1)$$

180 where f_{Tx} is the radar operating frequency, c is the speed of light in free space, Y is the
 181 geometric separation between the main and interferometer antenna arrays along the array
 182 normal direction (with $+Y$ in the direction of the array normal), α is the elevation
 183 angle, and ϕ_0 is the direction at $\alpha = 0^\circ$ (horizontal) set electronically by the radar hardware.
 184 This expression is equivalent to the original elevation angle equation (e.g., Milan
 185 et al., 1997; André et al., 1998) which did not account for interferometer offsets in the
 186 X - or Z -directions.

187 Another radar we applied this technique to is TIGER Bruny (TIG) which operated
 188 the sounding mode periodically from 2005–2011. Figure 5 shows joint probability
 189 distributions of elevation angle and slant range for GS echoes observed by TIG for the
 190 entire month of Dec 2005 as extracted from the sounding mode files. Each of the nine

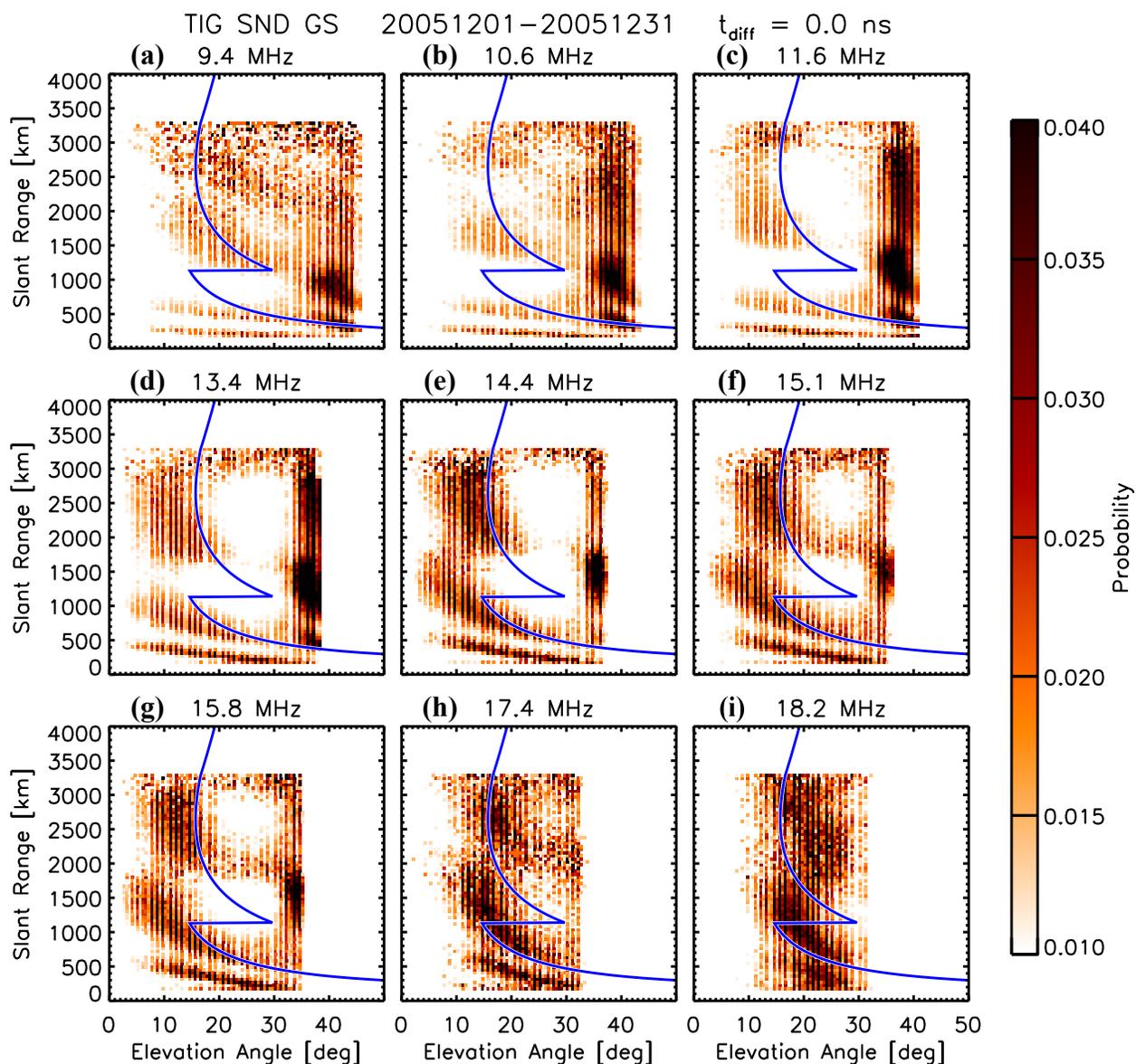


Figure 5. Joint probability distributions of elevation angle and slant range for ground scatter observed by the TIGER Bruny (TIG) radar during the entire month of Dec 2005 at 9 sounding mode frequencies, using the current hardware t_{diff} value of 0 ns. The Christmas Valley ground scatter virtual height model is overlaid on each panel in blue (Thomas & Shepherd, 2022). Note the vertical striping is due to the integer precision of the original sounding mode data format (Hughes et al., 2002).

191 panels corresponds to a different sounding frequency, and the GS component of the CV
 192 VHM is again overlaid in blue for reference. Based on current and historical versions of
 193 the TIG information in the SuperDARN analysis software, we have assumed a t_{diff} value
 194 of 0 ns and an interferometer offset solely in the Y-direction of 100 m were used in the
 195 original on-site calculation of these elevation angles. Note the vertical striping in Fig-
 196 ure 5 is due to the integer precision of the elevation angles (and all other parameters)
 197 stored in the sounding mode files.

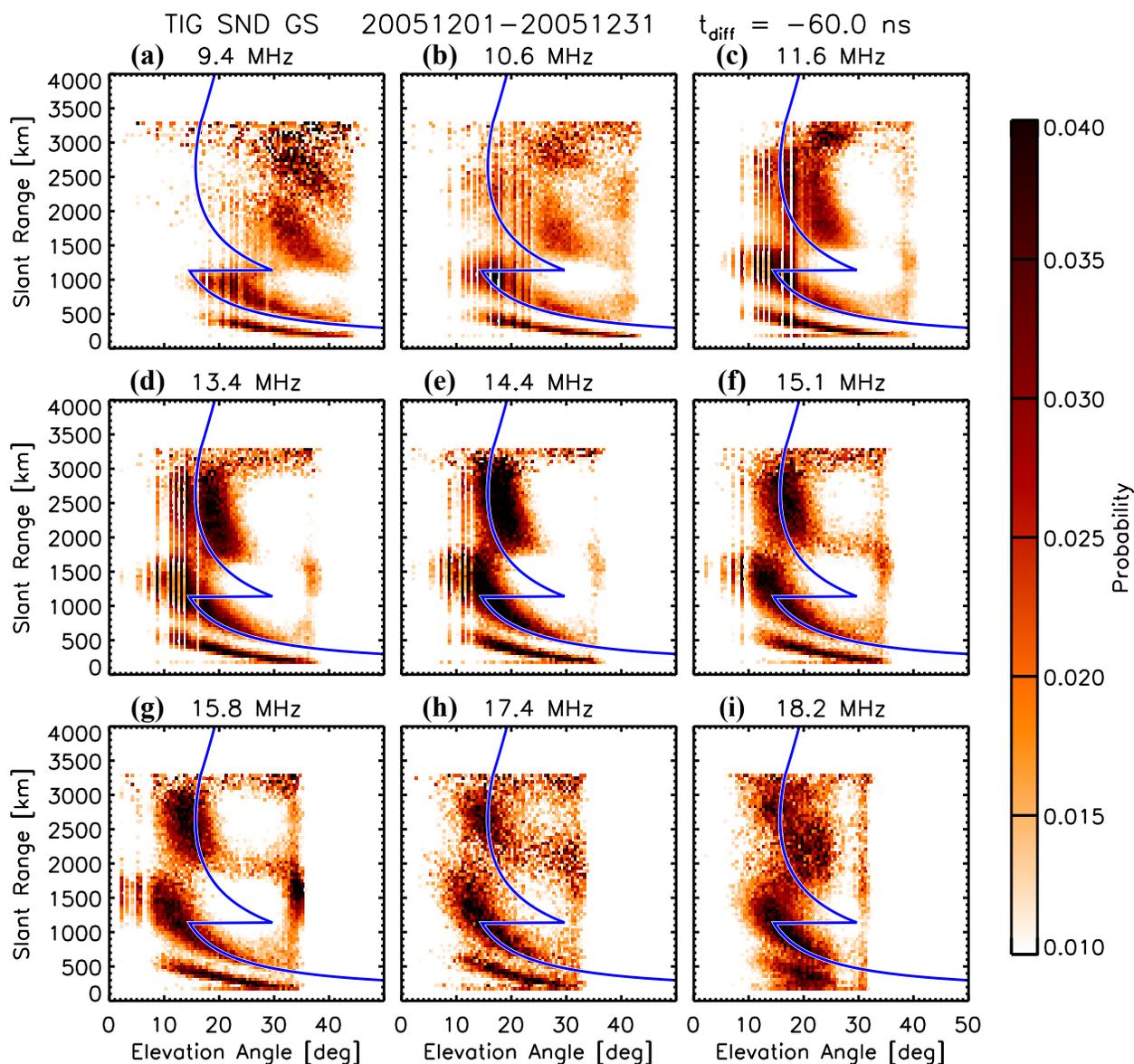


Figure 6. Joint probability distributions of elevation angle and slant range for ground scatter observed by the TIG radar during the entire month of Dec 2005 in the same format as Figure 5, using a revised t_{diff} value of -60 ns.

198 The TIG elevation-range distributions appear quite reasonable at frequencies above
 199 14.4 MHz (Figures 5e–i). Below 14.4 MHz however, the majority of the data at each range
 200 are aliased to very large elevation angles, similar to the GBR results in Figure 3. Ap-
 201 plying the multi-frequency calibration technique suggests a t_{diff} value of -60 ns, the re-
 202 sult of which can be seen in Figure 6. The GS distributions at higher frequencies remain
 203 largely unchanged from those in Figure 5, while at lower frequencies they are now in much
 204 better agreement with the CV VHM. This result (or agreement) can be understood in
 205 terms of the inherent 2π measurement ambiguity of SuperDARN interferometry, as 60 ns
 206 corresponds to the period of a 16.7 MHz wave, which lies near the middle of the upper
 207 TIG sounding frequencies (14.4 MHz / 69.4 ns and 18.2 MHz / 54.9 ns). An equivalent

Table 1. t_{diff} values found by applying multi-frequency calibration to SuperDARN radars participating in 2023 sounding mode experiments.

Radar Name	Code	t_{diff} (ns)
Clyde River	CLY	0
Christmas Valley East	CVE	-398
Christmas Valley West	CVW	-346
Goose Bay	GBR	-63
Hokkaido East	HOK	-40
Iceland East	ICE	-300
Iceland West	ICW	-380
Inuvik	INV	0
Kapuskasing	KAP	-34
Prince George	PGR	14
Rankin Inlet	RKN	45
Saskatoon	SAS	5

208 set of figures showing the pre- and post-calibrated TIG sounding mode results for IS echoes
 209 are available in the supplementary material.

210 Beginning in April 2023, network-wide sounding mode experiments have been con-
 211 ducted for three consecutive days each month using an updated version of the radar con-
 212 trol software originally developed by Hughes et al. (2002). The sounding mode data are
 213 now stored using the same format as the standard SuperDARN RawACF and FitACF
 214 files and can be processed using the open source Radar Software Toolkit (RST) (SuperDARN
 215 Data Analysis Working Group et al., 2022). The new data files also retain the observed
 216 phase differences (ψ_{obs}) for easier application of post-processing calibration techniques.
 217 Note the multi-frequency data from the ICW radar shown in section 2 were obtained us-
 218 ing this new sounding mode. In Table 1, we provide t_{diff} estimates for each of the par-
 219 ticipating SuperDARN radars found by applying the multi-frequency calibration approach
 220 to the 2023 sounding mode data. We hope the values in Table 1 will not only aid researchers
 221 attempting to use elevation data from these radars but also for future validation against
 222 other interferometer calibration techniques.

223 5 Discussion

224 The multi-frequency calibration approach improves upon the original virtual height
 225 comparison method (Chisham et al., 2021) by resolving the 2π measurement ambigu-
 226 ity when estimating t_{diff} . However there remain several limitations to this technique which
 227 must be addressed. First, it requires radar measurements obtained at different operat-
 228 ing frequencies which ideally span several MHz. Many SuperDARN radars operate in
 229 a fixed frequency band while others use separate frequency bands for daytime and night-
 230 time operation to try to optimize the amount of observed backscatter. It is therefore not
 231 always possible to apply this technique to historical observations for resolving the 2π am-
 232 biguity in t_{diff} if a radar did not collect data at a range of different operating frequen-
 233 cies.

234 Another source of uncertainty is associated with the use of an empirical VHM against
 235 which to compare the elevation angles calculated for different values of t_{diff} at different
 236 radar operating frequencies. For example, the Chisham VHM was derived using 5 years
 237 of data from a single high-latitude radar (Saskatoon) during solar cycle 23, while the more
 238 recent CV VHM was derived using 5 years of data during solar cycle 24 from two mid-
 239 latitude radars. HF propagation conditions are known to vary based on local time, sea-

240 son, and solar cycle conditions; therefore, use of the Chisham or CV VHM may not be
 241 appropriate for all radars under all geophysical conditions. More fundamentally, this tech-
 242 nique also assumes that t_{diff} was properly calibrated when deriving the VHM.

243 Thomas and Shepherd (2022) demonstrated that, in a climatological sense, HF prop-
 244 agation modes for backscatter from ionospheric irregularities are similar at both auro-
 245 ral and mid-latitudes. In the absence of more comprehensive empirical VHMs, which can
 246 describe a broader range of HF propagation conditions, we believe the CV VHM is cur-
 247 rently best suited for this multi-frequency calibration technique as it allows for the in-
 248 clusion of a significantly larger amount of input data as opposed to VHMs which only
 249 support ionospheric propagation modes. Improved performance could also be achieved
 250 by adjusting the ranges at which the VHMs change from one propagation mode to an-
 251 other as a function of frequency. This behavior can be clearly seen in Figure 6, where
 252 the transition between 1-hop *E*-region GS and 1-hop *F*-region GS varies from ~ 1000 km
 253 at 9.4 MHz to ~ 1700 km at 18.2 MHz, while the CV VHM transition remains fixed at
 254 1140 km.

255 Of course, the underlying principle behind the multi-frequency calibration approach
 256 (i.e., to account for the 2π measurement ambiguity in the observed phase difference) does
 257 not require the use of an empirical VHM at all. This technique can be generalized to any
 258 of the other calibration techniques using meteor echoes or other IS or GS targets sim-
 259 ply by comparing results at multiple (well-spaced) radar operating frequencies. Or, the
 260 multi-frequency method as described here could be applied first to provide a rough es-
 261 timate of t_{diff} for all frequencies, which could then be further refined using another tech-
 262 nique. We must also note that our approach assumes t_{diff} to be frequency independent,
 263 which has held true for all of the SuperDARN radars evaluated so far.

264 6 Summary

265 In this study we have presented a new approach for calibrating SuperDARN inter-
 266 ferometer measurements which can be applied to even the earliest available data. Us-
 267 ing both ionospheric and ground backscatter observations from a diverse set of radar op-
 268 erating frequencies, one can automatically identify the electrical path length difference
 269 (t_{diff}) which best agrees with empirical HF propagation characteristics at all frequen-
 270 cies. We have successfully applied this technique to obtain t_{diff} estimates for a dozen radars
 271 using multi-frequency observations from the past year. By regularly conducting multi-
 272 frequency sounding campaigns, this technique can allow for continuous calibration of el-
 273 evation angles and therefore improved geolocation of all SuperDARN backscatter obser-
 274 vations.

275 7 Open Research

276 The raw SuperDARN data used in this study are available from the British Antarc-
 277 tic Survey SuperDARN data mirror (<https://www.bas.ac.uk/project/superdarn/#data>).
 278 While data for this study were accessed from the British Antarctic Survey, they may also
 279 be obtained from the Canadian Federated Research Data Repository (FRDR) at
 280 <https://www.frdr-dfdr.ca/repo/collection/superdarn>. The Radar Software Toolkit to read
 281 and process the SuperDARN data can be downloaded from Zenodo (SuperDARN Data
 282 Analysis Working Group et al., 2022).

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Figure 1.

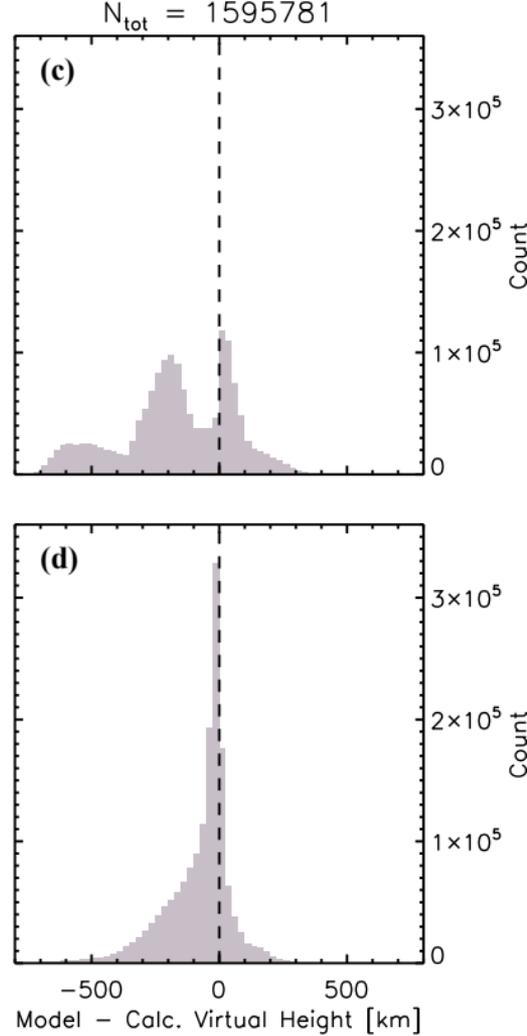
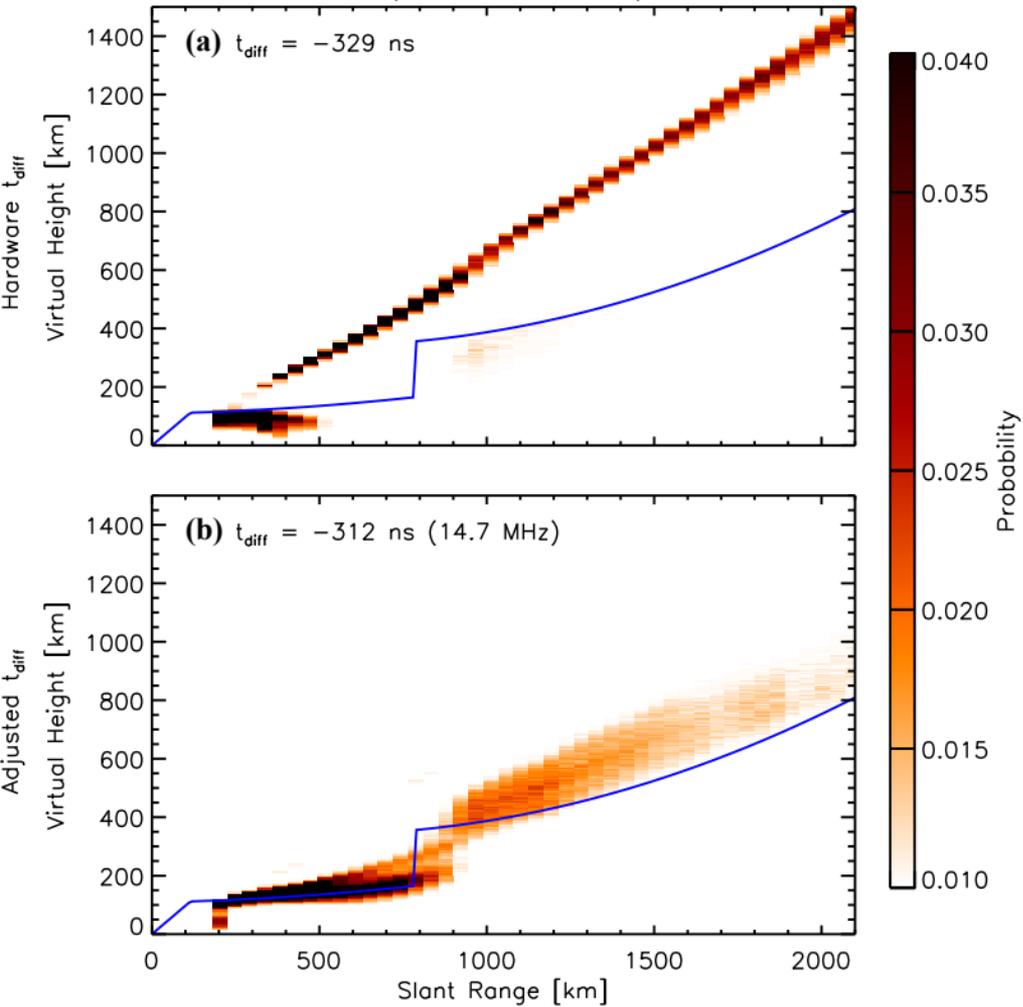


Figure 2.

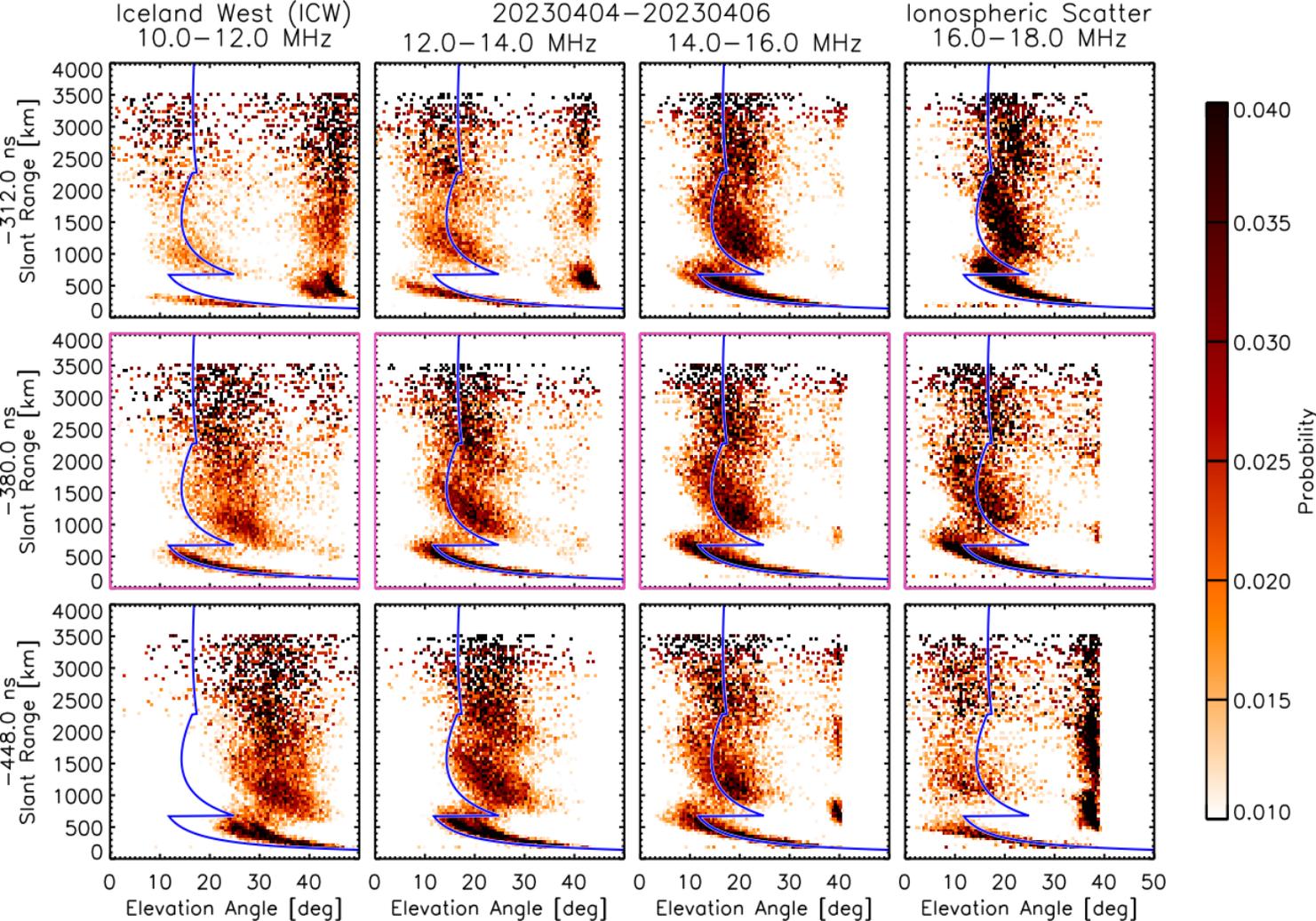


Figure 3.

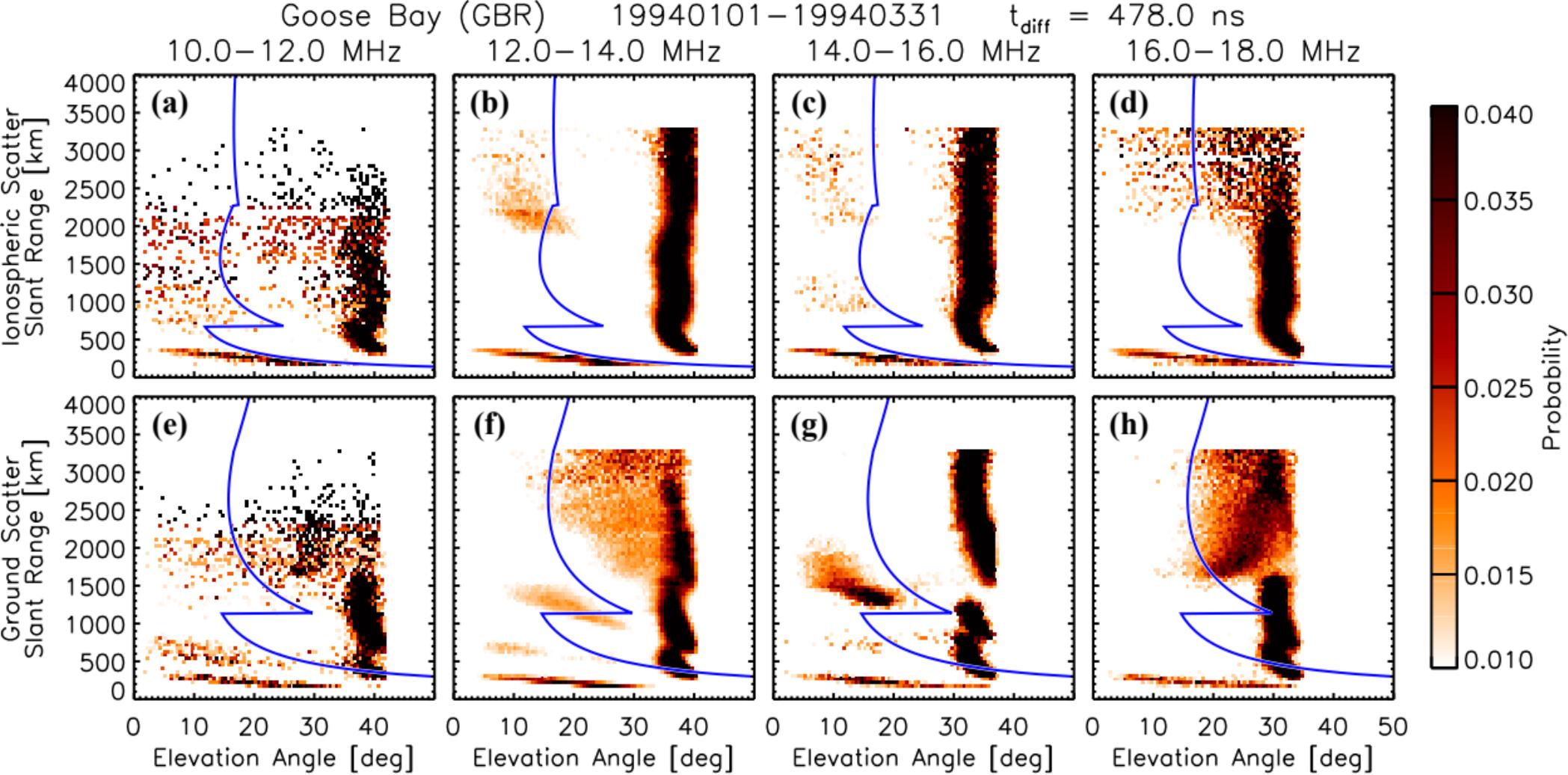


Figure 4.

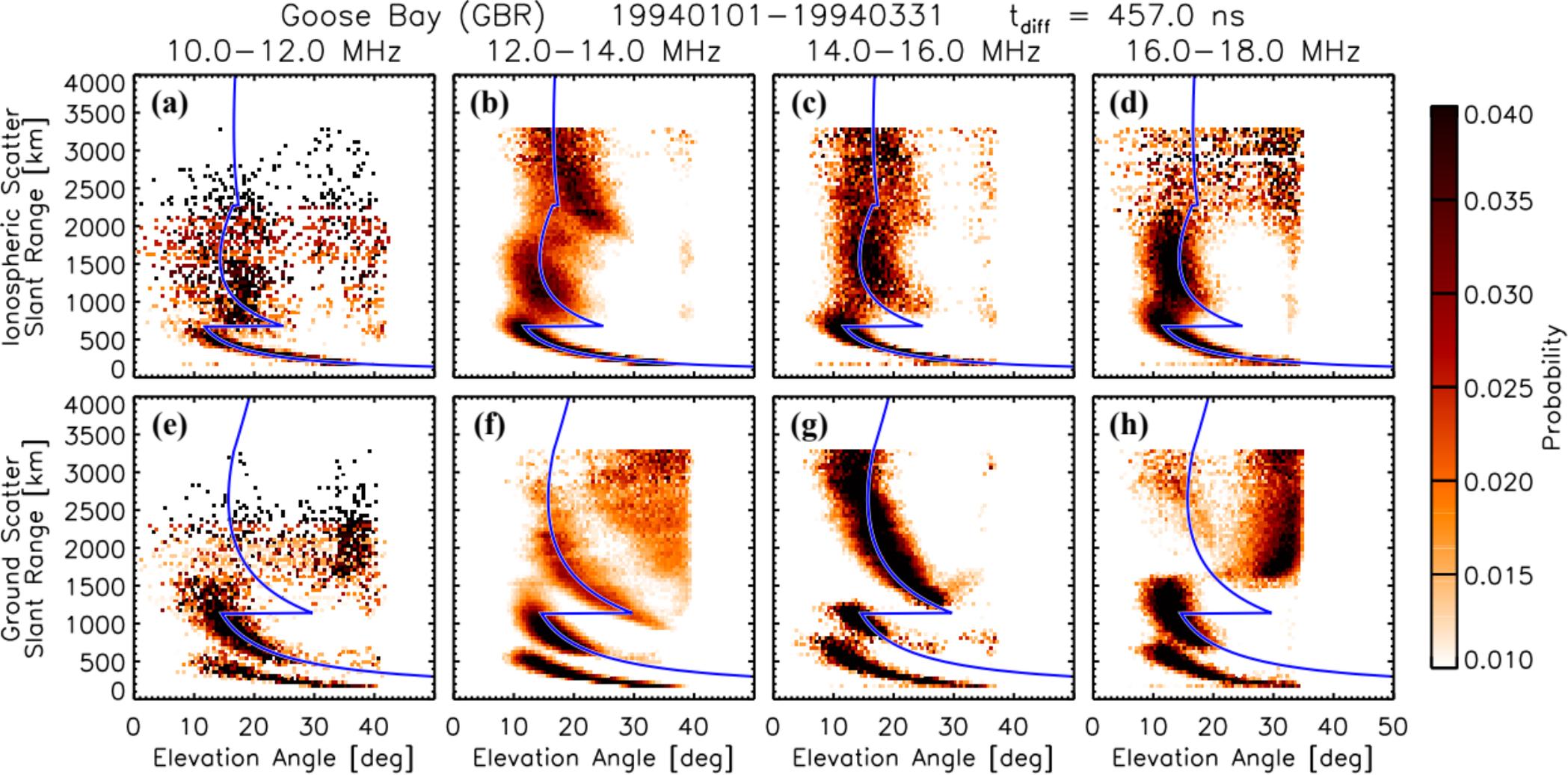


Figure 5.

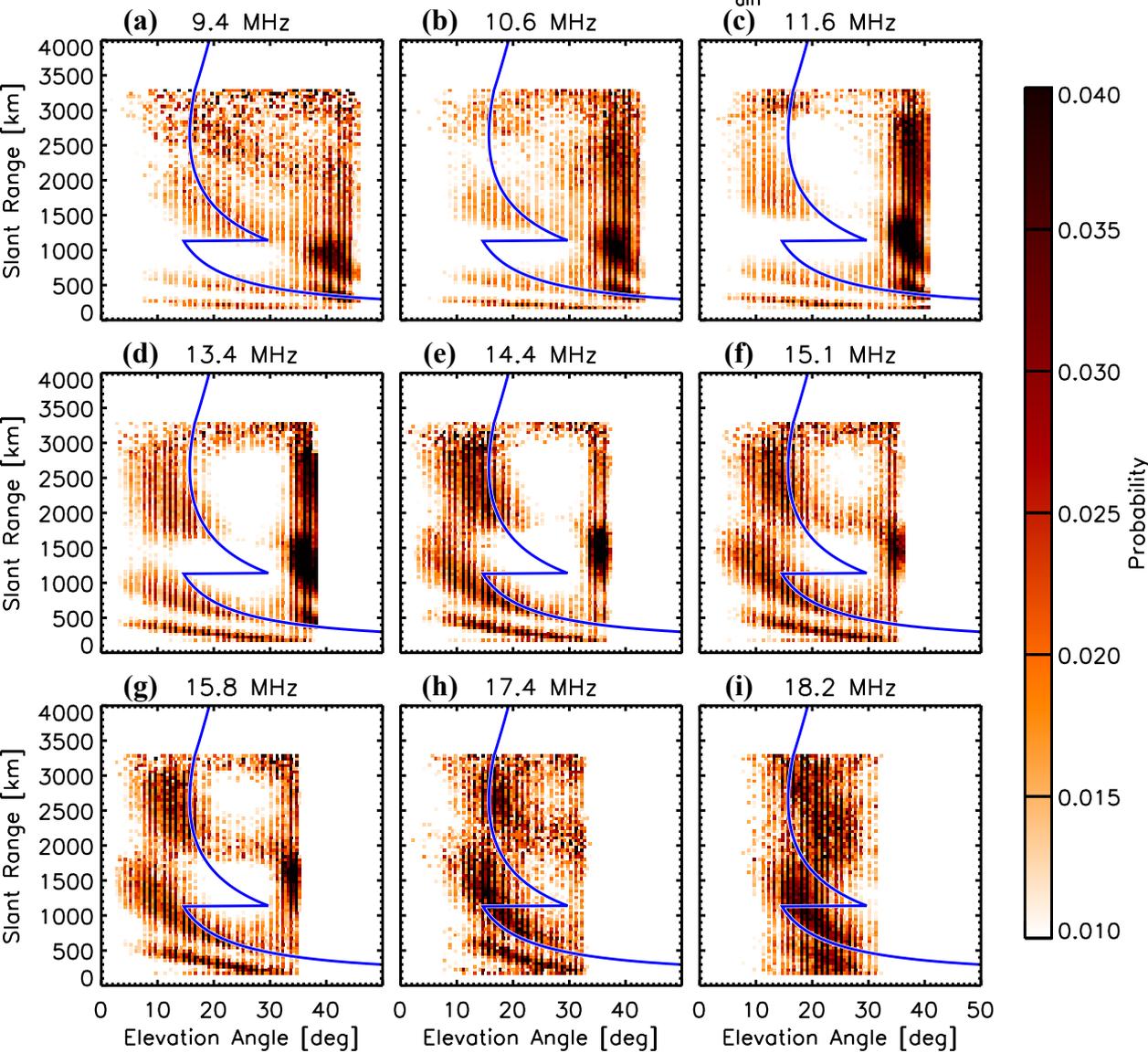


Figure 6.

