

# **FMCW Radar Signal Processing for Antarctic Ice Shelf Profiling and Imaging**

**Samiur Rahman**

Thesis submitted for the degree of

**Doctor of Philosophy**

**University College London**

Sensors, Systems and Circuits Group

Department of Electronic and Electrical Engineering

University College London

April 2016

I, Samiur Rahman, confirm that the work presented in this thesis is my own. Where information has been derived from other sources, I confirm that this has been indicated in the thesis.



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29/04/2016

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## Abstract

This thesis contains details of all the signal processing work being done on FMCW Radar (operating at VHF-UHF band) for the Antarctic Ice Shelf monitoring project that has been carried out at UCL. The system developed at UCL was based on a novel concept of phase-sensitive FMCW radar with low power consumption, thus allowing data collection for long period of time with millimetre range precision. Development of new signal processing method was required in order to process the large amount of data, along with the signal processing technique for obtaining the high precision range values. This was achieved during the first stage of the thesis, providing accurate ice shelf basal layer melt rate values. Properties of the FMCW radar system and experimental scenarios posed further signal processing challenges. Those challenges were met by developing number of novel algorithms. A novel shape matching algorithm was developed to detect internal layers underneath the ice shelf. Range migration correction method was developed to compensate for the defocusing of the image in large angles due to high fractional bandwidth of the radar system. Vertical error correction method was developed to compensate for any vertical displacement of the radar antenna during field experiment. Finally, a novel 3-D MIMO imaging algorithm for the Antarctic ice shelf base study was developed. This was done to process the 8x8 MIMO radar (developed at UCL) data. The radars have been deployed in the Antarctica during the Austral summer of each year from 2011-2014. The field experiments were done in the Ronne, Larsen-C, Larsen North, George VI and Ross ice shelves. The novel signal processing techniques have been successfully applied on the real data, allowing better understanding of the Antarctic ice shelf features.

## Acknowledgements

First and foremost, I would like to thank and express my gratitude to my supervisor Professor Paul Brennan, for accepting me as his PhD student. Throughout my PhD, he has always been very helpful. His guidance was the main factor that enabled me to complete my research work. He made it very easy for me to structure my whole thesis work, by always clearly explaining and updating me about the research goal. He has always been very easily accessible. Whenever I went to his room to discuss a problem even without an appointment, he has always given me time to provide suggestion to solve the problem. For this professional and friendly attitude towards me, I am grateful to him. Along with him, I am also very grateful to Dr. Lai Bun Lok. He is an immensely talented scientist and engineer. All the signal processing work I did during my PhD, was based on the system parameters and the data obtained from the radars that he built. Special thanks to Dr. Keith Nicholls and Dr. Hugh Corr from British Antarctic Survey, for letting me to work on their project of the Antarctic Ice Shelf.

A very special thanks to the Ministry of Science and Information Technology, Bangladesh, for funding my entire PhD. I would also like to thank Dr. Kenneth Tong and Professor Hugh Griffiths, for their valuable feedback during my transfer viva.

I would like to thank my former and current colleagues of the UCL radar group; namely Shashi, Arvind, Ajmal, Jian Ling, Allann, Hashir and Mandana. I consider myself very lucky to be part of this tightly knit research group.

I am forever grateful to my parents and siblings (Rusmila, Salman, Safwan) for providing the constant support in my life. Finally, I would like to thank my wife Tina, for all her encouragement and compassion. She has always made sure that I am in proper physical and mental state to complete my PhD smoothly.

## List of Publications

[1] Samiur Rahman, Paul V. Brennan, Lai Bun Lok, 'Millimetre-precision range profiling and cross-sectional imaging of ice shelves in Polar regions using phase-sensitive FMCW radar', Radar Conference, 2013 IEEE, April 29- 3 May 2013, Ottawa, Canada, DOI: 10.1109/RADAR.2013.6586055.

[2] Samiur Rahman, Paul V. Brennan, Lai Bun Lok, 'Shape matching algorithm for detecting radar hidden targets to determine internal layers in Antarctic ice shelves', Radar Conference, 2014 IEEE, 19-23 May 2014, Cincinnati, Ohio, USA, DOI: 10.1109/RADAR.2014.6875824.

[3] Paul V. Brennan, Samiur Rahman, Lai Bun Lok, 'Range migration compensation in static digital-beamforming-on-receive radar', IET Radar, Sonar and Navigation Journal, DOI: 10.1049/iet-rsn.2014.0355.

[4] Samiur Rahman, Paul V. Brennan, '3-D FMCW MIMO Radar imaging of Antarctic Ice Shelf Base', Cold Regions Science and Technology Journal (accepted).

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## Symbols

$A$	<i>Physical area of the receiving antenna</i>
$A_e$	<i>Effective area of the receiving antenna</i>
$B$	<i>Bandwidth</i>
$B(\Theta)$	<i>Beamforming function</i>
$c$	<i>Speed of light</i>
$d$	<i>Element spacing within the antenna array</i>
$dB$	<i>Decibel</i>
$\Delta\phi$	<i>Phase delay</i>
$\Delta L$	<i>Path length variation across antenna array</i>
$\Delta r, \Delta R$	<i>Range resolution</i>
$\Delta t$	<i>Time delay</i>
$\epsilon_r$	<i>Dielectric constant</i>
$f_c, f_o$	<i>Centre frequency</i>
$f_d$	<i>Deramped frequency</i>
$f_{d,down}$	<i>Deramped frequency for down chirp</i>
$f_{d,up}$	<i>Deramped frequency for up chirp</i>
$f_{step}$	<i>Step frequency</i>
$\phi$	<i>Elevation angle for 3-D MIMO beamforming</i>
$\varphi_d(t)$	<i>Deramped signal phase</i>
$\varphi_r(t)$	<i>Instantaneous phase of linear chirp (receive side)</i>
$\varphi_t(t)$	<i>Instantaneous phase of linear chirp (transmit side)</i>
$G$	<i>Antenna gain</i>
$K$	<i>Chirp rate</i>
$k_x, k_r$	<i>Spatial frequency</i>



$\lambda$	<i>Signal wavelength</i>
$N$	<i>Noise figure</i>
$p$	<i>Pad factor</i>
$p_r$	<i>Radar receiver power</i>
$p_t$	<i>Radar transmitter power</i>
$\rho_a$	<i>Efficiency of the antenna aperture</i>
$R$	<i>Range of the target</i>
$R_{coarse}$	<i>Coarse range</i>
$R_f$	<i>Fine range</i>
$R_o$	<i>Reference range</i>
$R_{max}$	<i>Maximum range of the radar</i>
$S_{IF}(t)$	<i>Deramped signal of the FMCW radar</i>
$S_r(t)$	<i>Receive signal of the FMCW radar</i>
$S_{tr}(t)$	<i>Transmitted signal of the FMCW radar</i>
$S_{min}$	<i>Minimum detectable signal</i>
$sc$	<i>Scaling factor for the shape matching algorithm</i>
$\sigma$	<i>Radar cross section</i>
$T$	<i>Chirp duration</i>
$T_p$	<i>Pulse width</i>
$\tau$	<i>Pulse duration</i>
$\Theta$	<i>Azimuth frequency for beamforming</i>
$v, v_{pl}$	<i>Velocity</i>
$w_n$	<i>Weighting coefficient for the beamforming</i>
$\omega_t(t)$	<i>Instantaneous transmitted frequency of the linear chirp</i>
$z$	<i>Vertical displacement of the array antenna element</i>

## Abbreviations

<i>ADC</i>	<i>Analog to Digital Converter</i>
<i>CPG</i>	<i>Coherent processing gain</i>
<i>CSA</i>	<i>Chirp Scaling Algorithm</i>
<i>CW</i>	<i>Continuous Wave</i>
<i>DDS</i>	<i>Direct Digital Synthesizer</i>
<i>FBP</i>	<i>Fast Backprojection Algorithm</i>
<i>FFBP</i>	<i>Fast Factorized Backprojection Algorithm</i>
<i>FFT</i>	<i>Fast Fourier Transform</i>
<i>FM</i>	<i>Frequency Modulation</i>
<i>FMCW</i>	<i>Frequency Modulated Continuous Wave</i>
<i>GBP</i>	<i>Global Backprojection Algorithm</i>
<i>GPR</i>	<i>Ground penetrating Radar</i>
<i>IF</i>	<i>Intermediate Frequency</i>
<i>InSAR</i>	<i>Interferometric SAR</i>
<i>MIMO</i>	<i>Multiple Input Multiple Output</i>
<i>PN</i>	<i>Pseudo Random Noise</i>
<i>pRES</i>	<i>Phase-sensitive Radio Echo Sounder</i>
<i>PRF</i>	<i>Pulse Repetition Frequency</i>
<i>PSLR</i>	<i>Peak to Sidelobe Ratio</i>
<i>Radar</i>	<i>Radio Detection and Ranging</i>
<i>RCS</i>	<i>Radar Cross Section</i>
<i>RDA</i>	<i>Range Doppler Algorithm</i>
<i>RMA</i>	<i>Range Migration Algorithm</i>
<i>SAR</i>	<i>Synthetic Aperture Radar</i>
<i>SCR</i>	<i>Signal to Clutter Ratio</i>
<i>SNR</i>	<i>Signal to Noise Ratio</i>

<i>UHF</i>	<i>Ultra High Frequency</i>
<i>VHF</i>	<i>Very High Frequency</i>

## **Chapter 1**

### **Introduction**

Accurate prediction of sea level rise is very important in the field of climate change research. Ice shelves in the Polar Regions have direct correlation with sea level rise. An ice shelf is a floating extension of ice sheet. Presence of an ice shelf limits the ice flow into the ocean. As ice shelves are already floating in the ocean, a major collapse in ice shelf can enhance the level of fresh water into the sea indirectly by speeding up the flow of land ice into the ocean [1]. A lot of ice shelves collapsed in the Antarctica in the last few decades [1]. Scientists reasoned that this phenomena is due to warmer climate [1]. It is therefore of great importance to analyze the pattern of the thickness change of the ice shelf over the year with respect to oceanographic data. This would give geoscientists an accurate idea about whether the melting of ice shelves from the bottom is due to a natural cycle or because of the climate change.

#### **1.1 Aim of the Thesis**

A purpose-built phase-sensitive FMCW radar has been developed at UCL. The radar uses phase sensitivity to determine range to high precision. This high precision is achieved by using a Vernier-like process that accounts for the fine range measurement of the target along with its coarse range. This fine range is attained by measuring the phase of the signal in the particular range bin where the coarse range is obtained. In order to obtain this, challenges have to be met for both hardware and signal processing. A similar phase-sensitive radar system (pRES) had been developed

before which also produced satisfactory results [2] [3]. The motivation for this project work was to develop an upgraded radar system capable of year-long operation with low power consumption while providing high precision data. Link budget modelling of the FMCW radar built for this project shows that with very low transmit power (0.1 W), 3 mm RMS precision at 1800 m range is achievable. The main features of this radar are low power consumption, low noise figure (6 dB), lightweight and suitable for low temperature operation. For these advantages, it can be used for long term continuous data collection in contrast to the pRES system (ice-penetrating radar built by British Antarctic Survey, predecessor of the current FMCW radar system), which is restricted to operation only during Austral summers and to provide a series of snapshots (because the pRES system is based on a general purpose network analyser, which has a high power consumption and requires a heavy petrol-generator, thus making it impractical for long term data collection).

The Antarctic Ice Shelf monitoring project involved novel radar hardware development as well as novel signal processing algorithm development for Ice Shelf profiling and imaging. This thesis work has dealt with the signal processing algorithm development part of the project. The main objective is to produce ice shelf range profiles of millimetre precision as well as imaging the features underneath the ice shelf. The signal processing challenges were to develop suitable algorithms for phase sensitive range profiling as well as imaging techniques. As is often the case, specific applications require customized techniques for data analysis. This has been the case during this thesis as well, as novel radar hardware system required newly developed signal processing algorithms. So, along with developing processing algorithms for data analysis by using conventional methods (for phased array, SAR, MIMO), new techniques have to be developed as well for accurate data processing. The aim of the

thesis is to obtain data from both Polar Regions and produce significant results (images and profiles) that would help geoscientists to better understand the melting pattern and other properties of ice shelves.

## **1.2 Thesis Outline**

**Chapter 1:** This chapter gives a brief outline of the thesis by giving an overview of the project aim.

**Chapter 2:** Theoretical background of the whole research work has been described in this chapter along with reviewing the research work done so far for remote sensing of ice.

**Chapter 3:** This chapter describes the specifics of the thesis work; the main signal processing algorithms that are required for polar data analysis. The theory of phased array radar, SAR and MIMO signal processing has been discussed in this chapter along with theoretical description of the phase sensitive range profiling method. MIMO signal processing has been described very briefly as newly developed 3-D MIMO imaging technique has been broadly discussed in Chapter 4.

**Chapter 4:** This chapter discusses the novel signal processing techniques developed for this thesis work. These include: range migration correction for phased array, shape matching algorithm for detecting hidden layers, vertical error correction by phase calibration and 3 dimensional MIMO imaging algorithm; these four techniques that have been developed for FMCW radar data analysis have been thoroughly discussed.

**Chapter 5:** Experimental results are shown in this chapter. All the algorithms that have been developed and simulated during this thesis were also put to the test by applying them to real data. The results are then analyzed and compared with the modelled results in this chapter.

**Chapter 6:** Along with concluding remarks, the overall contribution of the thesis is described in this chapter. Achievements and improvements made over the existing techniques has been discussed. Also, future work and scope for improvements of the developed algorithms are discussed.

## **Chapter 2**

### **Theoretical Background**

This chapter provides a basic theoretical outline of the whole research work starting from the fundamentals of radar. The majority of the discussion on the radar basics are based on [6] and [7]. After discussing these basics, FMCW radar theory is reviewed as it is the main experimental instrument for this research work. Theoretical discussion of Antarctic ice shelves is provided within the context of this particular research along with a review of previous research work in this field. This leads to a discussion of the need for updating the current technology and a description of the FMCW radar that was built at UCL for this particular project. A broad description of this Ice Shelf monitoring radar system is provided in the last section of this chapter.

#### **2.1 Fundamentals of Radar**

##### **2.1.1 Basic Concept of Radar**

After the formulation of electromagnetic radiation theory, the basic foundation of radio technology was established. The possibility of wireless communication was realized by the scientists, as electromagnetic waves propagate at the speed of light.

‘Radar’ (Radio Detection and Ranging) is a device that uses electromagnetic waves for the purpose of acquiring information at a distant location from that device. This operation is commonly known as ‘Remote Sensing’. Different bands of the electromagnetic spectrum are used in different applications of radar.



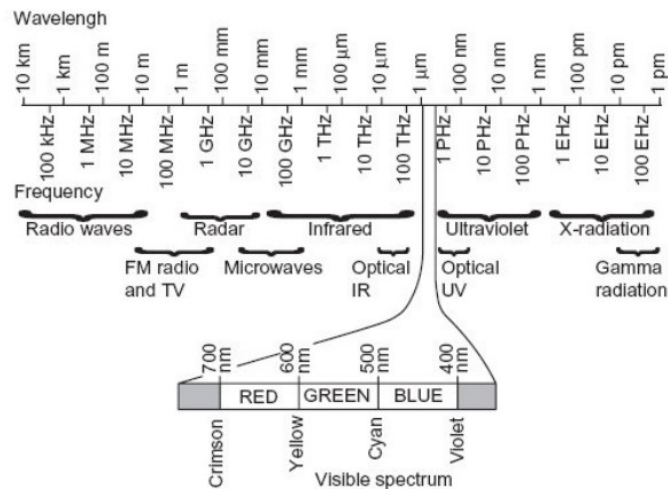


Figure 2.1 Electromagnetic Spectrum [4].

From the above figure, it can be seen that Radar just uses ‘light’ of a different wavelength. The visible spectrum is detected by the human eye. The advantage of the radar system is that it transmits electromagnetic waves, so the target can be detected from the reflection of the waves. The basic elements of a radar system are-

*Transmitter-* A transmitter produces a suitable amount of energy (usually in pulses) with the desired frequency by an oscillator. The transmitter power is radiated by an antenna. The antenna sends the signal with a specific directional distribution. These transmitted waves are scattered off objects that fall into the area encompassed by the antenna beam.

*Duplexer-* A duplexer is used in radar as a rapid switch to protect the radar receiver when the high-power transmitter is on. The duplexer also ensures the reflected signal from the target is channeled to the receiver, not the transmitter. During the reception mode, an antenna performs the identical operation in reverse direction.

*Receiver-* The receiver then amplifies (with minimal increase in noise) and demodulates the signal for subsequent detection and signal processing. Different types

of filter, attenuator, mixer and amplifier are used to achieve the desired signal. During the signal processing, with the information regarding antenna directivity, the time delay and phase/frequency change (with respect to the transmitted signal), the target position and/or speed can be determined.

A radar system includes a transmitter emitting electromagnetic radiation, in the form of a radio wave. The radio wave propagates through a medium (i.e. air, ice) and gets reflected from a target. The radar receiver collects the reflected wave. This reflected wave helps detect the presence of the target. As the speed of the wave propagation is known, the time of flight determines the range of the target (distance between the radar and the target). A moving target would produce a Doppler shift in the received signal, which can be used to calculate the velocity of the target. By using directional antennas, the direction of the target with respect to the radar (bearing) can also be determined.

The selection of wavelength for radar operation depends on the application. For instance, radar technology was mainly developed during World War II. The main objective at that time was to detect airplanes, combat ships or enemy vessels located beyond visual range [5]. The main advantage of such a technology is that no one has to rely on sunlight, or clear sky. Radar itself is the source of the emitted waves reflected by a target. So, radars mainly emit electromagnetic waves at such frequencies which, unlike visible light, easily pass through clouds, fogs with attenuation as less as possible (some radars use mm wave for high resolution, but signals get highly attenuated at that frequency).

### **2.1.2 Radar Equations**

One of the most basic aspects of radar theory is the radar range equation, derived from the equation for receiver power in relation to transmitted power and antenna gain.

A typical radar signal will have a two-way path propagation from transmitter to target and then reflected back from target to receiver. Due to this propagation through 3-dimensional space, the received signal power will depend on the reflectivity of the target, directivity of the signal, distance travelled by the signal and the gain of the antenna. So, it is very important to calculate the receiver power before performing any radar operation. The Radar receiver power equation [6] is as follows:

$$P_r = \frac{P_t G A_e \sigma}{(4\pi)^2 r^4} \quad (2.1)$$

In the above equation,  $P_t$  is the transmitted power,  $G$  is the maximum gain of antenna,  $A_e$  is effective area of the receiving antenna,  $\sigma$  is the radar cross section (RCS) of the target and  $r$  is the distance of the target from the radar receiver or target range.  $A_e$  is directly related to the physical area of the antenna aperture  $A$  where  $A_e = \rho_a A$ , where  $\rho_a$  is the efficiency of antenna aperture [6]. To determine the maximum range  $R_{max}$  of the radar,  $P_r$  needs to be at least equal to the minimum detectable signal  $S_{min}$ . Hence, rearranging Equation (2.1) gives the radar range equation [7]:

$$R_{max} = \left[ \frac{P_t G A_e \sigma}{(4\pi)^2 S_{min}} \right]^{1/4} \quad (2.2)$$

The above equation is also referred as the radar equation or range equation. During a radar design for any specific application, target RCS is taken into account to determine the choice of parameters, such as operating frequency and transmit power, since the reflectivity directly depends on frequency. After determining those, the power budget is calculated.

Another basic factor in radar operation is the range resolution. It determines the ability of the radar to distinguish between two targets. Before formulating the range resolution, it is necessary to first determine the range of the radar. By considering  $\tau$  as the two-way propagation delay time (time between the emission of the

electromagnetic wave from the receiver and the reflected wave coming back to the receiver). The propagation delay then can be written as,

$$\tau = \frac{2r}{c} \quad (2.3)$$

The above equation enables calculation of the radar range,  $r$ , where  $c$  is the speed of light in free space,  $3 \times 10^8 \text{ ms}^{-1}$ . In a different medium, the speed of light changes which is compensated by taking the dielectric constant of that medium into account. In its simplest form, range resolution can be calculated [7] as follows:

$$\Delta r = \frac{c\tau}{2} \quad (2.4)$$

In order to obtain two distinct signal returns from two separate targets, the targets have to be separated in range by at least half the distance corresponding to the pulse width. In most cases, radars send a modulated signal. So, it is more useful to determine range resolution in terms of signal bandwidth. That formula can be achieved by using the spectral density of the transmitted pulse.

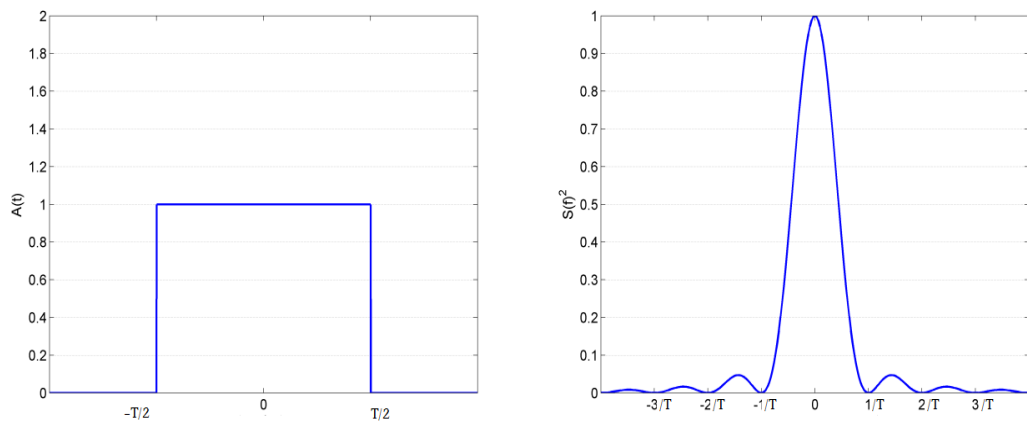


Figure 2.2: (a) Rectangular pulse in time domain; (b) Spectral density of that rectangular pulse first null occurring at  $1/T$ .

For a pulsed radar system, the time difference between two distinctly separated target echoes would be as follows,

$$T = \frac{2(r+\Delta r)}{c} - \frac{2r}{c} = \frac{2\Delta r}{c} \quad (2.5)$$

Because of the two-way propagation of the signal, the minimum range resolution would vary proportionally with one half of the pulse length,  $T$ . From the above figure, the bandwidth of the pulse can be determined with respect to the location of the first null (for a non-modulated signal, bandwidth would be the smallest positive frequency where the power spectral density is zero). According to the above figure, where a rectangular pulse is considered, the first null occurs at  $1/T$ . So, bandwidth can be written as:

$$B \approx \frac{1}{T} \quad (2.6)$$

Using Equation (2.5) and (2.6), range resolution can be re-written as:

$$\Delta r = \frac{c}{2B} \quad (2.7)$$

In order to calculate the maximum unambiguous range, the Pulse Repetition Frequency (PRF) needs to be considered. This is the rate at which the radar transmitter emits signal (in the case of pulsed radar system). If a signal is reflected from a target after the transmitter has sent another pulse, then the range of the target cannot be unambiguously determined, as the corresponding transmitted pulse is not defined with absolute certainty. This maximum unambiguous range,  $r_u$ , can be mathematically expressed as,

$$r_u = \frac{c}{2PRF} \quad (2.8)$$

### 2.1.3 Radar application for remote sensing of the environment

Even though radar was originally developed for military purpose during wars, it serves as a very important tool for civilian applications as well. The most commonly known civilian application for radar is air-traffic control [7]. The widely used Air Traffic Control (ATC) radar beacon system and the microwave landing system are mainly based on radar technology [7]. Radar is also widely used nowadays for

environmental monitoring. For weather observation, the Nexrad system is widely known [6]. It is a network of 160 S-band Doppler radars. Along with climatology, the precipitation data from Nexrad is used in the field of meteorology and hydrology as well [8]. Various SAR systems (both spaceborne and airborne) have been used for earth and climate monitoring as SAR images help imaging of the ocean current pattern, glacier motion pattern, volcanic movement, geological studies and so on. Examples include Magellan (NASA/JPL), RADARSAT-1(Canadian Space Agency), ERS-1/2 (European Remote Sensing satellite), J-ERS (Japanese Earth Remote Sensing satellite), SEASAT (performing first civilian SAR application), and CARABAS [9].

Radars are also used for ship safety [7]. In adverse weather conditions when visibility is poor, radars can be used for collision avoidance (with other naval vehicles, navigation buoys). Automated radar systems with detection and tracking capacity for collision avoidance are commercially available and usually small in size and not costly [7].

Another significant contribution of radar for remote sensing is the measurement of the earth's mean sea level (geoid- global mean sea level model to precisely measure the surface elevations of the earth) [10]. Downward looking spaceborne altimeters such as the ERS-1/2 have been used for this purpose [10].

## **2.2 FMCW Radar**

Generally, the transmitted radar signal is either a single pulse at a time or it is a continuous wave. Radar that uses the first technique is known as pulsed radar. Radar that uses the latter technique is known as Continuous Wave radar or CW radar for short. Frequency Modulated Continuous Wave (FMCW) radar is a type of CW radar that transmits frequency modulated waves, usually with linear frequency modulation.

The advantage obtained by modulating the frequency is the ability to determine range without requiring a short pulse. Modulation of the transmitted signal provides the timing mark which can be used to extract range information. For instance, a frequency modulated signal would have various frequency components. So, when a target echo is received, those different frequency components will arrive in different time thus creating the option for range measurement. The operation principle for FMCW is discussed below.

### 2.2.1 Linear FM Signal

The application and operation of FM signals is widespread in radio broadcasting. One advantage can be found in radar over radio broadcasting is that radar does not transmit any meaningful message. It just sends a signal in order to get a reflection. So, there is no need for complicated manipulation to decode any audio-visual message from the signal. A simple pulse containing numerous frequency components would suffice. One such simple FM signal is a chirp signal (or sweep signal, both terms are used interchangeably). It is basically an FM signal where the frequency is increased or decreased with time, just like the chirping noise of birds, or bats. It is simple in concept and easy to generate. A chirp signal can be generated in ascending order (up chirp) or descending order (down chirp). Also, the frequency differences can be linear or exponential. In order to maintain simplicity, most radars use a linear chirp signal. The transmitted signal equation is,

$$s(\tau) = \text{rect}\left(\frac{\tau}{T_p}\right) e^{(i2\pi f_c + i\pi K\tau^2)} \quad (2.9)$$

$K$  is the chirp rate which is the ratio of the bandwidth and pulse width  $B/T_p$ ,  $f_c$  is the center frequency and  $\tau$  is pulse duration.

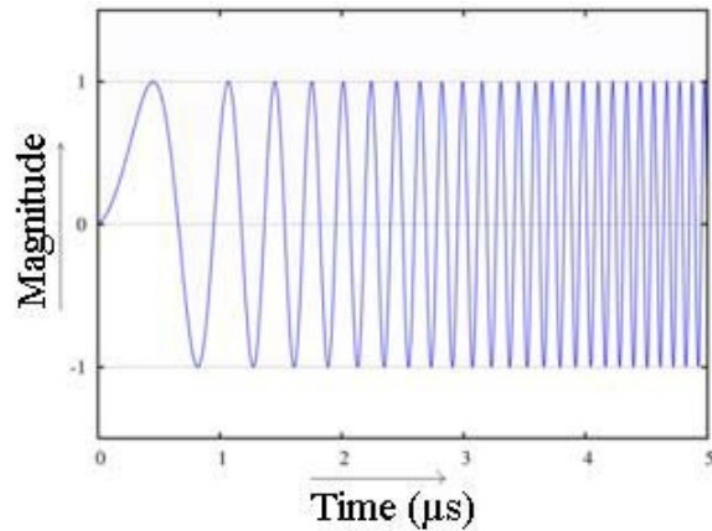


Figure 2.3 Linear FM (chirp) Signal.

### 2.2.2 Basic FMCW Radar Operation Principle

A linear FM chirp signal is used for transmission in FMCW radar. After receiving a target echo, the received signal is mixed with the transmit signal. The resultant frequency is known as a beat frequency of deramped frequency. This deramped frequency is used for calculating the target range. The mathematical depiction of FMCW radar operation is given in [11]. For a stationary target, the deramped frequency would be a constant value. In order to get an expression for the deramped frequency, the mathematical form of the instantaneous FM chirp transmitted signal in sinusoidal form can be written as:

$$s_{tr}(t) = a_o \sin 2\pi \left[ f_o t + \frac{Kt^2}{2} \right] \quad (2.10)$$

where  $K$  is the chirp rate and  $f_o$  is the centre frequency. The received signal will be a time delayed version of the above equation most likely with different amplitude due



to attenuation. If the time delay between transmit and receive is  $\tau$ , then the equation for the received signal becomes:

$$s_r(t) = b_o \sin 2\pi \left[ f_o(t - \tau) + \frac{K(t-\tau)^2}{2} \right] \quad (2.11)$$

The deramped signal  $s_{IF}(t)$  is the resultant of the mixing of these two signals:

$$s_{IF}(t) = s_{tr}(t) \cdot s_r(t) \quad (2.12)$$

The deramped signal will consist of both the sum of and difference of two signal phases. For FMCW operation, only the difference is required thus the sum is ignored or filtered out. Simplification after substituting values from Equation (2.10) and (2.11) to Equation (2.12) gives the following equation for the deramped signal:

$$s_{IF}(t) = c_o \cos 2\pi \left[ f_o\tau + Kt\tau - \frac{K\tau^2}{2} \right] \quad (2.13)$$

Here, the second term is the time linear phase term corresponding to the instantaneous frequency difference between the transmit signal and receive signal.

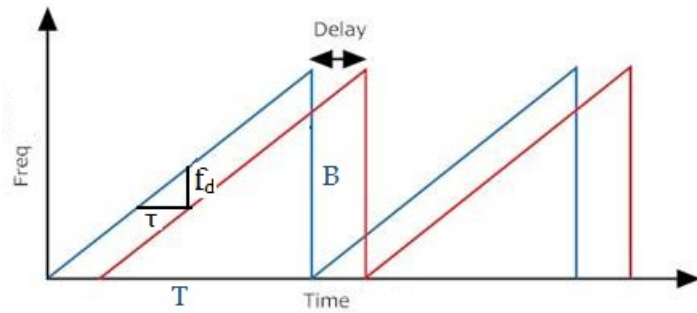


Figure 2.4: Plot of FMCW chirp transmit and receive signal to calculate the deramped frequency,  $f_d$ .

From the above figure, the equation for  $f_d$  can be obtained easily by using simple geometry. Here, the blue line is the transmitted chirp and the red line is the received signal, which is equal to the transmitted one but delayed by the round trip delay  $\tau$ . It can be seen that the triangle consisting of sides equal to bandwidth  $B$  and chirp duration  $T$  is similar to the triangle consisting of sides  $f_d$  and  $\tau$  (they have equal angles).

So, the ratio of their corresponding sides will be equal ( $f_d/B = \tau/T$ ). Hence, equation for the deramped signal can be written as:

$$f_d = \frac{B\tau}{T} \quad (2.14)$$

The round trip delay  $\tau$  can be written in terms of range  $R$  as  $\tau = 2R/c$ . So, Equation (2.14) then becomes:

$$f_d = \frac{2BR}{Tc} \quad (2.15)$$

It should be noted that in Figure (2.4), only the up-chirp has been used. In FMCW operation, it should be sufficient if Doppler information is of no concern. In order to extract Doppler information, both up and down chirps would be required. Otherwise, range-Doppler cross coupling will occur which can be mathematically seen by modifying Equation (2.13) for a moving target. If the target is moving with velocity  $v$ , then range becomes function of time ( $R(t) = R_o + vt$ ). So, the equation for round trip delay becomes:

$$\tau = \frac{2(R_o + vt)}{c} \quad (2.16)$$

The last term in Equation (2.13), which is a small phase offset, is neglected (as  $\tau \ll T$ ). Combining Equation (2.13) and (2.16):

$$s_{IF}(t) = c_o \cos 2\pi \left[ \frac{2f_o R_o}{c} + \left( \frac{2f_o v}{c} + \frac{2KR_o}{c} \right) t + \frac{2Kvt^2}{c} \right] \quad (2.17)$$

In the above equation, the third term corresponds to range-Doppler coupling.

So, when the FMCW radar application consists of a moving target, it uses chirps in both directions. The method can be understood from Figure (2.5). It can be seen that the deramped frequency will have different values depending on the chirp direction. For up and down chirps, equation for deramped frequencies would be as follows:

$$f_{d,down} = \frac{2BR}{Tc} - \frac{2f_o v}{c} \quad (2.18)$$

$$f_{d,up} = \frac{2BR}{T_c} + \frac{2f_0v}{c} \quad (2.19)$$

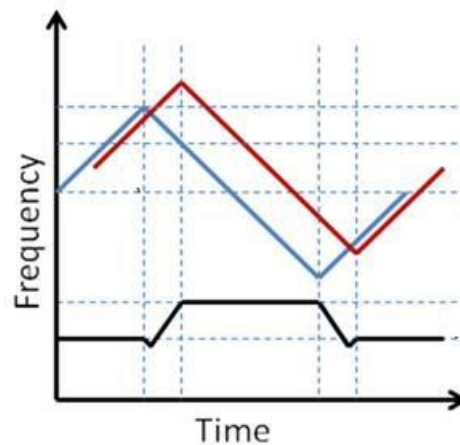


Figure 2.5: Plot of triangular FMCW transmit signal (blue) and return echo from a moving target (red) along with corresponding deramped frequency (black) [12].

It should be noted from Figure (2.5) that the deramped frequency of a moving target (assuming constant velocity) becomes an oscillating wave. The actual range will be the average of these two frequency values. Target velocity can also be obtained by substitution.

### 2.2.3 Advantages of using FMCW radar

FMCW radar has the advantage of achieving better range resolution for a given peak output power as it is easier to increase the bandwidth of a FMCW radar than to shorten the pulse width of a pulse radar. Another very important factor when deramp processing is used is that even though FMCW radar operates with a large bandwidth, after mixing, the frequency resolution becomes much smaller ( $1/T$ ). This is the equivalent processing gain often known as the time-bandwidth product [12]. The application for this specific project requires high precision through a lossy medium in order to measure the ice shelf melting rate precisely along with the need for yearlong data collection (which means low power transmission is desired). Also, as this is not

a military application, there is no concern with jamming or interference as is often the case with FMCW radar. Considering these factors, FMCW radar has been chosen for this project.

From a hardware perspective, the advantage of FMCW radar over conventional linear FM pulse compression radar is worth mentioning. In the latter one, good range resolution is also achieved by using wider bandwidth. But pulse compression is a time domain operation. So, data acquisition hardware is required to be quick enough to sample the whole transmit signal bandwidth. FMCW radar systems are able to mitigate this hardware constraint. This is because of the use of the deramped signal bandwidth instead of the entire transmission bandwidth. In a sense, Fourier transformation of deramped signal is equivalent to the matched filtering operation during pulse compression. Also, to compensate for the  $r^4$  attenuation ( $r^3$  for distributed targets), conventional pulse radars use gain controls to make sure the dynamic range of the IF signal stays within limit. Due to the linear relationship between range and deramped frequency, this can be achieved in the frequency domain in FMCW radars.

### **2.3 Overview of Antarctic Ice Shelf Research**

Antarctic ice shelf research is a part of environmental monitoring research work. One of the main research topics in modern science is the climate change research, which has escalated the need for research on worldwide environmental surveillance. One of the possible effects due to climate change is sea level rise across the world. The repercussion is quite alarming for areas close to oceans and sea shores [13]. So, it is of great importance that geoscientists are able to make accurate predictions of the sea level rise, as these calculations will have direct social, economic and political consequences. In the last 100 years, sea level rise has been approximately 4 to 8 inches

[13] which has been at least partly attributed to global warming [14] [15]. The average global surface temperature is also rising, which the scientific community attributes to man-made causes, such as increase in greenhouse gas. Due to this, polar ice caps and glaciers worldwide are showing a rapid melting rate hence contributing directly to sea level rise. This is the main incentive for scientific monitoring of Antarctic ice shelves. Proper quantification of the role of ice shelves in sea level rise is thus necessary. Before discussing the research work performed so far in this field, it would be appropriate to describe the basic theory of the ice shelves.

### 2.3.1 Ice Shelves in Antarctica

Ice shelves are floating extensions of ice sheets that link the ocean to the landmass. Even though there are ice shelves in the northern hemisphere, most of the ice shelves are on the Antarctica.

Understanding the formation of an ice shelf requires a brief description of an ice sheet. An ice sheet is a large chunk of land ice covering most of the Polar Regions.

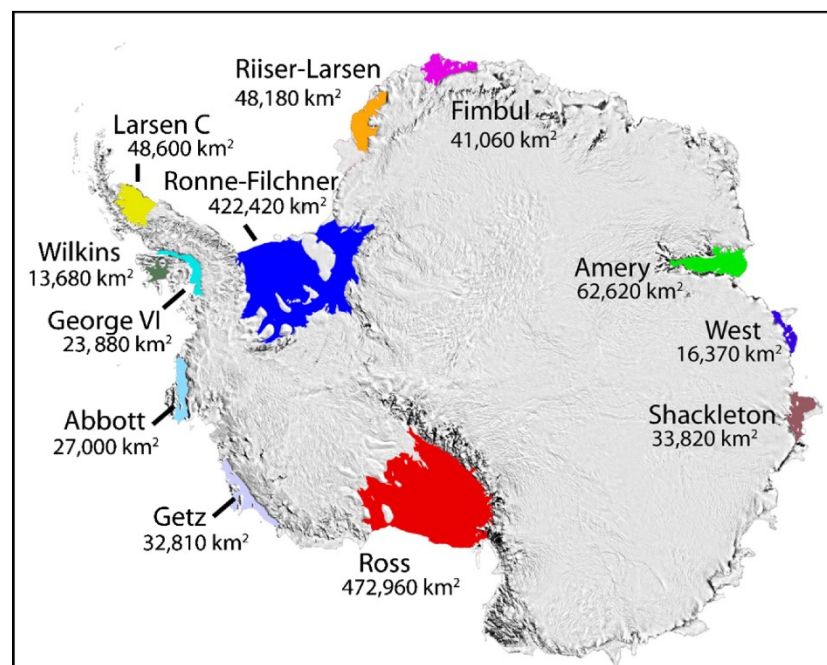


Figure 2.6: Map of Antarctica. The colored parts are various ice shelves, several of those have been used as experimental locations in this project. (Credit: Ted Scambos, NSIDC).

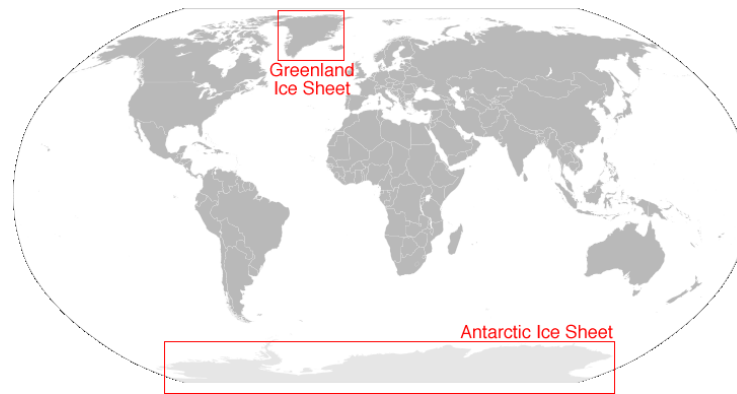


Figure 2.7 Ice sheets in both Polar Regions (Credit: NSIDC).

Large parts of North America and Scandinavia were also covered by ice sheets during the last ice age. The ice sheets in Antarctica and Greenland contain 99% of the freshwater ice of the whole world [16]. The Antarctic ice sheet extends over almost 5.4 million square miles, which is the largest single mass of ice in the world containing 7.2 cubic miles of ice [16]. The ice sheet in Greenland extends almost 656,000 square miles [16].

Ice shelves are basically floating extensions of ice sheets [1]. They are formed when the ice sheet extends to the coast and over the ocean. Through ice streams and glaciers, ice from the ice sheet gradually oozes into the sea. Due to differences in temperature, newly arrived ice does not melt into water straight away. It starts floating like an iceberg. But it also gains size as more ice gets added from behind before the small chunk gets the chance to dissolve into the ocean. The very important feature of these ice shelves is that they act as the interface between the huge ice sheet and the ocean. So, they are able to act as a bottleneck to slow down the ice flow into the ocean. This is very important in terms of sea level rise study because the implication is that ice shelves are very useful to limit the amount of water in the ocean by inhibiting the land ice (glacier and ice sheet) flow into the ocean. If these ice shelves are melting

rapidly or collapsing, that would have direct effect on sea level rise. There have been studies regarding ice shelf collapses and melting which even gives alarming prediction that the ice shelves may disappear within next 200 years [17] [18] [19].

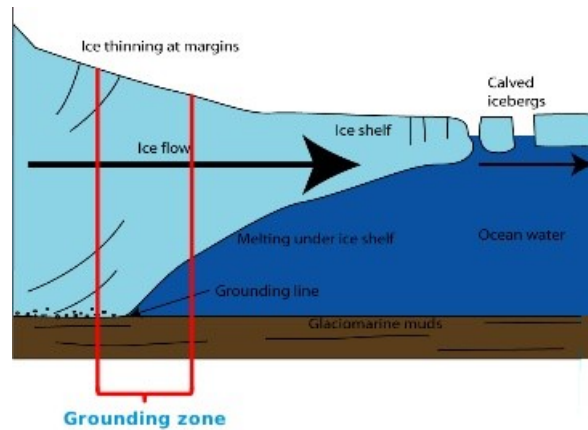


Figure 2.8: Simplified diagram of ice formation in Polar Regions showing the fragmentation from ice sheet to ice shelf and then to iceberg. Grounding line is the boundary between floating and grounded ice. (Credit: Mark R. Drinkwater; European Space Agency, ESA-ESTEC).

### 2.3.2 Use of remote sensing for Ice Shelf Monitoring

In 1933, it was discovered that high frequency radio signals can penetrate snow and ice, the foundation for use of remote sensing to study ice was then built [20]. It was 24 years later though when the first experiment was performed (by Amory Waite) that used radio echo-sounder (RES) for measuring ice thickness [20]. In 1963, a VHF system for echo sounding was built [21]. That pioneered the research field of RES systems for ice measurements. Radars built for this purpose were mostly pulse radars operating in the VHF and UHF frequency ranges. These instruments achieved quite good success in calculating ice thicknesses in Polar Regions [22]. After a decade, use of FMCW radar in this field also started. In 1980, an FMCW system [23] was used to

detect and measure water equivalence (amount of water contained in the snowpack) and snow stratigraphy (different layers of snow).

The current research on ice shelf monitoring is a major collaboration between government research organizations from various countries. Following are the names of few organizations that are leading the research work: British Antarctic Survey (BAS) from UK, National Snow and Ice Data Center (NSIDC) from USA, National Institute of Polar Research (NIPR) from Japan, Australian Antarctic Division (AAD) from Australia and Canadian Ice Service (CIS) from Canada.

Different kinds of radar have been used for ice monitoring. NSIDC have used a combination of satellite techniques (Interferometric Synthetic Aperture Radar (InSAR), visible-band imagery, and repeat-track laser altimetry) to map ice shelves [24]. They mapped the grounding line location of the entire Amery ice shelf. Results shown in [24] helped improve the understanding of the dynamic state of the ice shelf. Various other papers have been published encompassing analysis of data obtained from Ice Shelf measurements [25] [26]. In [27], Wilkins Ice Shelf break-up events that occurred in 28 February to 6 March, 27 May to 31 May, and 28 June to mid-July of 2008, are analyzed with the help of satellite remote sensing observation. The data provided great detail of the ice shelf calving while the break up was occurring. It was discovered that the break up was due to a unique type of ice shelf calving, referred to as 'disintegration'. The paper also goes into further details of this disintegration process. In [28], NSIDC used FMCW radar for ice measurements to study the discontinuities in the snowpack. C-Band (2-6 GHz), X-Band (8-12 GHz), and Ku-Band (14-18 GHz) frequencies were used for measurements (as the snow structure is not uniform, reflectivity from different sections of the snowpack for a single frequency band will vary. Hence, different frequency bands are used and the results are



compared). The magnitude and location of snow pack discontinuities have been calculated as function of frequencies to precisely determine snow cover properties from monostatic and bistatic radar systems. The experiments revealed that when radar operates at 14-18 GHz, it gives better information about the internal features of a dry snow pack. On the other hand, due to the high absorption loss in water, in wet snow areas, high frequency transmission provides very poor result. For the latter case, lower frequencies (2-6 GHz) were required to penetrate wet snow without significant attenuation. In [29], Landsat-7 ETM+ images had been used to map blue ice areas in Antarctica. In [30], IceBridge Ku-Band Radar altimeter data provides measurements from Polar Regions of both hemispheres. Time, latitude, longitude, elevation, and surface measurements, as well as flight path charts and echograms images were obtained from this radar system. The Ice, Cloud and Elevation Satellite (ICESat) [31] is another major project of NSIDC, for studying polar ice. This satellite performs operation based on Geoscience Laser Altimeter System (GLAS), a space-based laser-ranging instrument (LIDAR). It provides year-long data to determine ice sheet mass balance along with data of stratospheric clouds (clouds in the winter polar stratosphere) that are widespread over polar areas. The second version, ICESat-2, is expected to start operation in 2017.

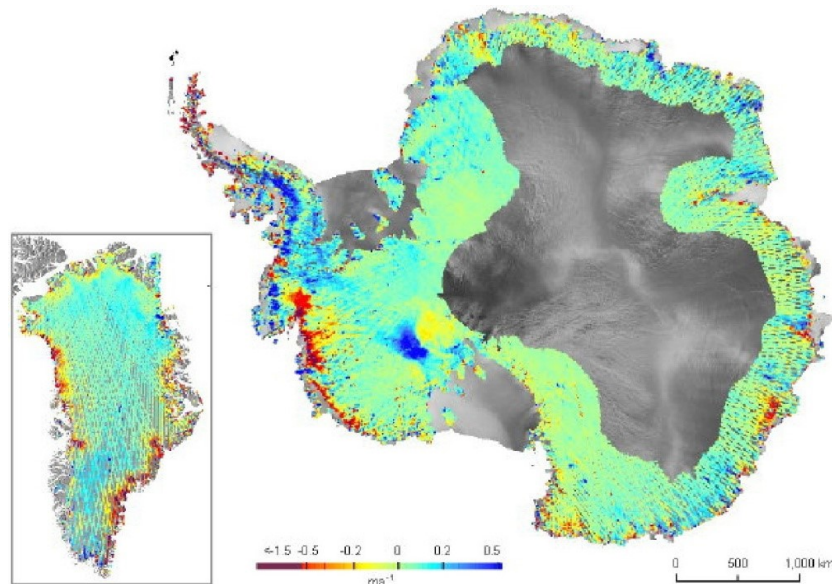


Figure 2.9: A map created from ICESat data demonstrating the extent of ice sheet thinning in Greenland and Antarctica [32]. (Credit: British Antarctic Survey/NASA).

Besides satellite based systems, numerous ground based systems are also used for ice monitoring in polar areas. An elaborate description of ground penetrating radar (GPR) systems can be found in [33]. Further details of the operating principles of GPR may be found in [33]. Figure (2.10) gives an example of processed data from GPR. An eskers are long ridges found in the glaciated regions of the northern hemisphere (Europe and North America). From Figure (2.10), properties underneath the esker can be seen from the GPR data. As it is seen, radar signal passes through the ice easily but gets reflected back from others (different types of rocks). Near the surface, reflection is coming from boulder and gravel (different classes of rock according to particle size). Below the ice, hyperbolic shaped reflections are due to sediment and bedrock. Most of the GPR operations are in VHF/UHF frequencies. Even though higher resolution can be achieved by using high frequency signals, for ice shelf thickness analysis where the depth of ice can be as long as 2 km, high frequency signals can be severely attenuated while propagating through the snow and ice [35]. In [36] [37], it was studied that ice shelf depth change measured by using satellite altimeters does not directly

correspond to the mass imbalance. Mass imbalance is the offset from the net balance between the accumulation and ablation of ice. This imbalance was measured by satellite altimeters to calculate the change in depth.

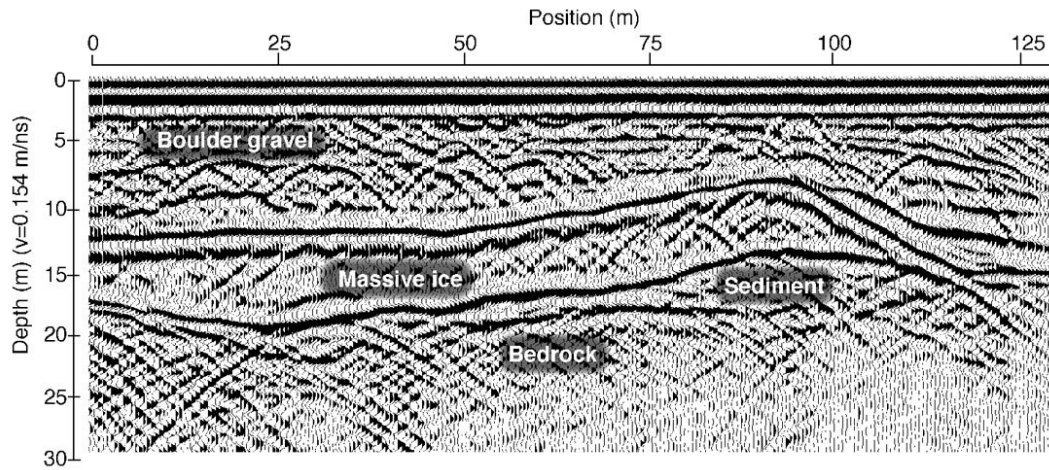


Figure 2.10: An example of GPR profile showing underneath the ice cored esker near Carat Lake, N.W. T. Canada [34].

Instead, thickness change can be attributed to other meteorological conditions that affect snow compaction such as densification process that changes snow morphology. In [37], GPR was used to measure snow compaction in two Antarctic Ice Shelves (McMurdo and Ross), which would compensate for the uncertainties from the altimeter data. The GPR operated at UHF band. In [38], GPR was used for crevasse detection across the Ross Ice Shelf. Here, the radar transmission frequency was also in UHF (400 MHz). In [39], debris characteristics (along with dynamics in the ablation region) of McMurdo Ice Shelf were studied using GPR data.

The British Antarctic Survey (BAS) have pioneered the use of phase sensitive radar for measuring ice shelf thicknesses with millimeter precision. As the image processing from the radar data is mainly done by the processing of the received signals, phase accuracy of the signals allows for more accurate imaging, hence enhancing the image quality. In [3], the potential was first shown of phase sensitive radar for precise

measurement of ice shelf basal layer melt rate. The pRES was first deployed at the George VI ice shelf to study ice-ocean interaction. The incentive for high precision melt rate measurement was that this event plays a major role on Antarctic ice mass balance, accounting for more than 20% ice discharge from the ice sheet [40]. Accurate study of the melting of the ice shelf basal layer also helps to understand the evolution cycle of water mass around polar areas, consequently having an effect on entire global water mass [41]. But many of the previous basal layer depth estimations had errors even by margins of 50% [42]. The pRES system helped to narrow down the error margin significantly. The radar system developed at BAS, known as pRES system, was based on the stepped-frequency principle. Stepped-frequency continuous wave radars are also FMCW radars where instead of a continuous frequency band sweep, the frequency is increased in discrete steps [33]. Advantages and disadvantages of stepped frequency radar have been detailed in [42]. Step frequency radars are used in glaciology mainly because of high bandwidth, which means high resolution, as well as for better signal-to-noise ratio (SNR) [43]. Other main factor is the radar systems capture not only amplitude, but also phase information. The amplitude was measured directly where the phase was calculated by comparing the reflected signal with the output of a precision reference oscillator. The carrier waveform was demodulated in both side bands from the centre frequency to the baseband [44]. The pRES system consists of a network analyzer (HP8751A). Transmit and receive antennas were identical broadband antennas with 10 dBi gain. The system operated at 305 MHz centre frequency where bandwidth being 160 MHz. The range plot from radar data is the Fourier transformed result of the return signal (in logarithmic scale) shown in Figure (2.11).

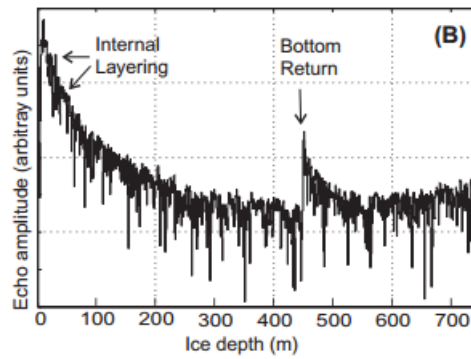


Figure 2.11: Logarithmically scaled amplitude plot after processing pRES data from George VI ice shelf, showing basal layer at around 450 meters [3].

In [2], the pRES system was used for basal layer depth measurement in the Ronne Ice Shelf near the grounding line of the Rutford Ice Stream. Here, the ice shelf is comparatively very deep (ranging from 1570 to 1940 meters). It was a year-long project that consisted of comprehensive analysis of the horizontal deformation, along with pRES data of the vertical displacement of the ice shelf base and internal layers. Even though in [3], basal melt rate measurement was quite straightforward (measuring thickness for 7 days), calculation of the Ronne Ice Shelf melt rate needed few more steps. Along with measuring oscillations in vertical strain rate and ice thickness due to the fluctuation of the ice shelf on the tide, depth variation measurement of the long-term strain rate was also done. The latter occurs when ice shelf surface deviates from the level of isostatic equilibrium.

### 2.3.3 Motivation for the radar system upgrade

The pRES system was able to provide very accurate data. But there were several limitations. Those are:

- Being based on a general-purpose network analyzer, the whole system is quite bulky, so not suitable for moving from one place to other
- High receiver noise (noise figure  $> 30$  dB)
- Hence a long measurement time is required (100-1000s)

- High power consumption, requiring a petrol generator
- Bulky and not suitable for low temperature operation
- Restricted to operation only during Austral summers and to a series of snapshots (due to the factors mentioned above, it is impractical to use the system for whole year)

The need for a new system that would be lightweight, consume low power and most importantly, be able to collect data throughout the year on its own has become apparent. Also, there was an ever-growing need for making numerous copies of the system to deploy simultaneously in various experimental sites. Copying a bulky and high power consuming pRES system would be very impractical. Hence, the project started by collaboration between BAS and UCL where first objective was to build a modified FMCW radar prototype with lower noise figure and power consumption along with all the radar components suitable for low temperature operation, hence making a year-long operation possible.

#### **2.3.4 Phase sensitive FMCW radar for Ice Shelf Monitoring built at UCL**

The FMCW radar built at UCL has the capacity to measure melt rates to mm/year precision. The radar is built to maintain this precision up to 1800 metres depth [45]. The main features of this radar are low power consumption, low noise figure and hence a short measurement time, lightweight and suitable for low temperature operation. The radar system is discussed below by using [45] as the main reference.

As seen from the Figure (2.12), the phase-sensitive FMCW radar is based on a Direct Digital Synthesizer (DDS) linear FM chirp generator. It consists of a low noise receiver/downconverter chain. A master clock is used to synchronize the DDS chirp generator and the data logger (Analog to Digital Converter (ADC)). This synchronization allows for precise phase measurement during processing and the

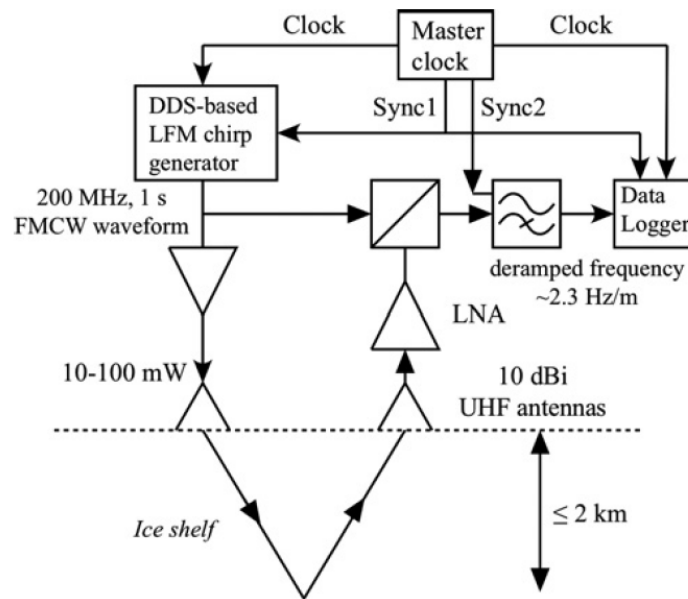


Figure 2.12: Simplified block diagram of the UCL built phase-sensitive FMCW radar [45]

achieved phase precision exceeds the standard range resolution. It is a low power consumption device with 5 W power needed during operation and 1 mW during standby. As the noise figure is far less than the pRES system, the signal collection time is also much smaller (1 s, as opposed to 100-1000 s for the pRES system). It should be noted that, as different gain settings are used during field test and to have the option for pulse-to-pulse averaging, the signal collection time is set to 1 min. If one minute of data is collected in every six hours (common operating mode), the mean power consumption is around 31 mW. So, to collect data for a whole winter, a modest accumulator of 40 Ah capacity would suffice. All the RF components of the radar system are able to work to temperatures as low as  $-40^{\circ}\text{C}$ . The radar system is designed for VHF-UHF band operation. The bandwidth is 200 MHz with centre frequency at 300 MHz. So, for a 1 s pulse duration, the deramped frequency is 2.35 Hz/m (using Equation (2.14)). The range resolution becomes 43 cm by considering the dielectric constant of the Antarctic ice as 3.1 (value provided by the British Antarctic Survey). Measuring the phase of the deramped pulses carefully, millimetre range precision is

achieved (discussed more in Chapter 3). Considering the maximum range as 2 km, the deramped frequency at this point would be 4.7 kHz. So, a simple data logger with a sampling rate of 12 ksamples/s or higher would be sufficient.

Table 1: Parameters of phase sensitive FMCW radar

Operating frequency (centre), $f_c$	300 MHz
FM sweep bandwidth, $B$	200 MHz
RF power, $P_r$	20 dBm
Antenna gains, $G_t, G_r$	10 dBi
Noise figure, $N$	6 dB ( $F = 4$ )
Associated standard range resolution, $\Delta R$	43 cm with $\epsilon_r = 3.1$
Depth precision in phase-sensitive mode	3 mm RMS, provided $\text{SNR} > 21$ dB
Pulse duration, $T$	1 s
Total acquisition time	Total acquisition time 60 s. Ten pulses each with four RF gain values
ADC sampling rate	>12 ksamples/s
Ice attenuation	0.015 dB/m
Maximum operating range, $R$	2 km
Reflection coefficient between internal layers	-60 to -90 dB
Reflection coefficient at ice sheet base	-2 dB



Figure 2.13 Prototype phase sensitive FMCW radar built at UCL [45].



Figure (2.13) shows the prototype radar system for the Antarctic ice shelf monitoring developed at UCL. The system is based on the block diagram shown in Figure (2.12), including added front-end filtering and digital clock generation and synchronization. The DDS synthesizer used for the system is an Analog Device AD9910. It generates a sweep signal of 200-400 MHz with a 1 GHz clock. The radar system is designed to provide good performance on data processing for reflections coming from both near and far from the radar. Different gain settings have been used for this purpose. A pair of Mini-Circuits ZX76-31-PP+ digital step attenuators are used to achieve this. These step attenuators have four combinations that set the RF gain values to 4, 16, 28 and 40 dB. During operation, the gain settings are changed in this sequence for successive chirps. A second-order high pass filter in the baseband path is also used in order to compensate for the signal attenuation in the ice. Considering the attenuation coefficient in the ice as 0.015 dB/m, the reflected signal degrades at a rate of 30 dB/decade within first 100 m range [45], then degrades more rapidly. The high-pass filter has 0 dB gain at frequencies under 50 Hz and the maximum gain is 80 dB at 5 kHz. It has a fixed slope of +40 dB/decade, over compensating at short ranges but undercompensating at longer ranges. This is an acceptable trade off, where the 60 dB or greater dynamic range of the ADC should work, which means effective number of bits of 10 should be enough. During operation, the system consumes 750 mA at 6 V. It consumes 0.24 mA during standby. This is a very low power consumption when compared to the pRES system which required hundreds of watts and a petrol generator. The reduced noise figure (6 dB) means shorter signal collection time, which saves energy consumption even more. During the link budget modelling for the system [45], the expected SNR was calculated as greater than 75 dB for ice shelf basal layer not exceeding 500 m. After that the SNR drops rapidly due to the inverse-cube dependence

and also the ice attenuation. At maximum range (2 km), the SNR is around 14 dB. The loop test results of the prototype system are discussed in Chapter 5.

## Chapter 3

# Signal Processing Algorithms for the Antarctic Data Analysis

This chapter concerns with the details of signal processing algorithms that have been used for analysing the FMCW radar data. The first part discusses the phase sensitive signal processing method. This method would be used to achieve the millimetre precision range values while measuring the ice shelf thickness. The FMCW radar has also been used for imaging of ice shelf base. Different experiments have been set up at Antarctic sites for imaging the layers underneath ice shelf surface. Three different algorithms have been used for image processing; phased array processing, SAR processing and MIMO processing. Discussions regarding the basics of all three algorithms have been done in this chapter. Along with the basics, simulation results for all the algorithms are shown and discussed. Instead of using some open source codes for simulation, MATLAB codes were written for all the conventional imaging algorithms, along with writing new code for the precision range profiling. One of the reasons for this is the received signal characteristics of the FMCW radar built at UCL. Usually conventional algorithms process the received signal (i.e. beamforming operation in phased array) in time domain. Meanwhile, in the FMCW radar, the received signal is mixed with the transmitted signal before the data logger saves it. So, the output from the data logger is the deramped signal, which is in frequency domain. Most of the open source codes use matched filtering operation to produce the range history, which is not the case for this system. A conventional SAR system simulation

scenario requires a platform in constant motion for a specific amount of time (which determines the Synthetic Aperture length). This constant motion creates the SAR Doppler history. The experimental setup used in the Antarctica was not a pure SAR system, but a simulated SAR system. The radar was mounted on a sledge and then moved in to different positions in a straight line to collect data. SAR signal history is obtained by the assumption of an entirely stationary targets within the period of data collection (a very assumption in the Antarctic Ice Shelves), so the Doppler information is extrapolated by using the phase information and the total data collection time. In case of MIMO processing, a conventional MIMO system transmits different transmitted signals (usually orthogonal to each other). In case of the Antarctic ice shelf monitoring, all the transmitters send same signal. MIMO system scenario is achieved by using time switching method. Considering all these factors, along with the need to ensure that the algorithms work properly with the FMCW radar system parameters, simulations for the conventional algorithms were performed by newly written MATLAB codes.

### **3.1 Phase Sensitive Range Profiling**

In [45] [46], techniques for millimetre precision range measurement with FMCW radar, along with the hardware design to achieve this, have been discussed. In [45], a processing technique for unambiguous range measurement with the newly built phase sensitive radar has been thoroughly discussed. The FMCW radar uses phase sensitivity to determine the high precision range. This high precision is achieved by using a Vernier-like process that accounts for the fine range measurement of the target along with its coarse range. The operation principle of a Vernier scale is using a major and a minor scale. The major scale gives the coarse length and combining it with the value obtained from the minor scale, precise length is measured. Similar approach is used

by the phase-sensitive FMCW radar system. This fine range is attained by measuring the phase of the signal in the particular range bin where the coarse range is obtained.

### 3.1.1 Phase sensitive processing steps

The mathematical description of the processing method [45] is given below:

The instantaneous frequency of the linear chirp transmission can be written as:

$$\omega_t(t) = \omega_c + K(t - T/2) \quad 0 \leq t \leq T \quad (3.1)$$

where  $\omega_c$  is the centre frequency,  $T$  is the pulse duration and  $K$  is the chirp rate. It should be noted that chirp rate is signal bandwidth divided by time (should be multiplied with  $2\pi$  to get the value in radians). By integrating Equation (3.1), equation for instantaneous phase can be obtained:

$$\varphi_t(t) = \omega_c t + \frac{K(t-T/2)^2}{2} + const \quad 0 \leq t \leq T \quad (3.2)$$

The equation for the phase of the received signal arriving from range  $R$  will be:

$$\varphi_r(t) = \omega_c(t - \tau) + \frac{K(t-\tau-T/2)^2}{2} + const \quad (3.3)$$

Here,  $\tau$  is the round trip delay which for any range  $R$ ,  $\tau = 2R\sqrt{\epsilon_r}/c$ .

Deramped signal phase can be mathematically obtained by subtracting the receive signal phase from the transmitted signal phase. So, from Equation (3.2) and (3.3):

$$\varphi_d(t) = \varphi_t - \varphi_r = \omega_c \tau + K\tau \left( t - \frac{T}{2} \right) - K\tau^2/2 \quad (3.4)$$

The first term of the phase is the important phase term used for high precision range calculation. Second term is the linear phase term and third one is the phase offset (generally negligible). It can be noted that equation for the deramped frequency can be obtained by differentiating Equation (3.4), which will be exactly the same as the deramped frequency derived in Chapter 2 (Equation (2.12)).

The range profile is obtained by Fourier transforming the deramped signal. The range bins will have resolution defined in Equation (2.5). The separation of range bins

will be  $1/T$  in frequency. The purpose-built radar uses Vernier-like process to obtain high resolution range. Along with measuring the coarse range of a target return, it also measures the phase of the range bin where target is located. Cumulating this fine range measurement within the range bin with the coarse range provides the millimetre range precision. In order to measure the accurate phase, the FM waveform and ADC need to be precisely synchronized.

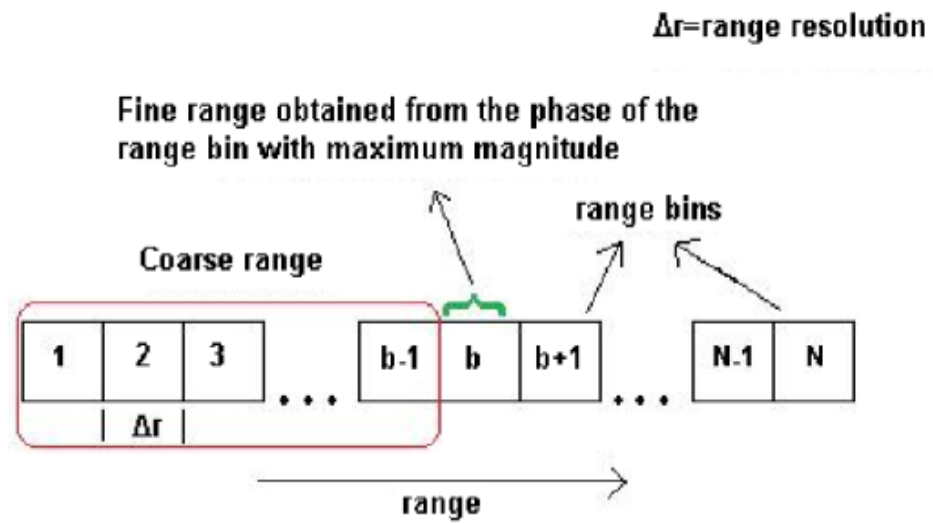


Figure 3.1: High precision range processing method for the phase-sensitive FMCW radar.

The synchronization is achieved by generating the FM signal with a frequency divided version of the ADC signal. Figure (3.1) shows the basic idea behind the range calculation method of this radar. At first, peak amplitude is searched within the range bins in the proximity of the expected target range. As can be seen in Figure (3.1), it is the range bin 'b'. Coarse range value with respect to range bin (b-1) is calculated by using the generic range formula,  $R_{coarse} = (b-1) c / 2B$ . After that, fine range within the range bin 'b' is attained from the phase of that range bin. The equation relating the instantaneous phase within the range bin and the range value can be found, as follows:

$$\varphi_d = \frac{2\pi}{\lambda} 2R_f$$

$$\Rightarrow R_f = \frac{\varphi_d \lambda}{4\pi} \quad (3.5)$$

Here is the instantaneous deramped signal and is the fine range. The factor 2 accommodates for the two-way travel of the radar signal. So, the final range output by this radar of a certain target is,

$$R = \frac{(b-1)c}{2B\sqrt{\epsilon_r}} + \frac{\varphi_d \lambda}{4\pi} \quad (3.6)$$

$\epsilon_r$  is the dielectric constant. There are some modifications required during signal processing. For instance, the phase variation of the deramped pulse from a target as it moves across the total range bin can be defined as:

$$\Delta(\omega_c \tau) = \omega_c (2\Delta R \sqrt{\epsilon_r/c}) = \frac{\omega_c}{B} = 2\pi f_c/B \quad (3.7)$$

To ensure phase unambiguously, the phase variation cannot exceed  $2\pi$ . Equation (3.7) shows that for unambiguous phase and hence range indication, the chirp bandwidth needs to be at least equal to the carrier frequency. Which means ultra-wideband operation would be necessary for the desired performance. But instead of putting strain on hardware design, the problem can be easily solved by signal processing. Before obtaining the range profile by Fourier transformation, the deramped signal can be zero padded, which decreases the range bin spacing by a factor of  $p$  to  $\Delta R/p$  (where  $\Delta R$  is the range resolution and  $p$  is the pad factor). A pad factor of 2 would be adequate for the current radar, reducing the phase variation from  $3\pi$  to  $1.5\pi$ .

One other observation can be made from the Equation (3.4). The phase centre of the deramped signal is at the centre of the sample, at  $t = T/2$ . But after Fourier transform, it would be at  $t=0$ . So, before calculating the range profile, the deramped signal needs to be rotated to align the waveform centre with the starting point of the sample. Figure (3.2) provides a visual understanding of the zero padding and rotating process.

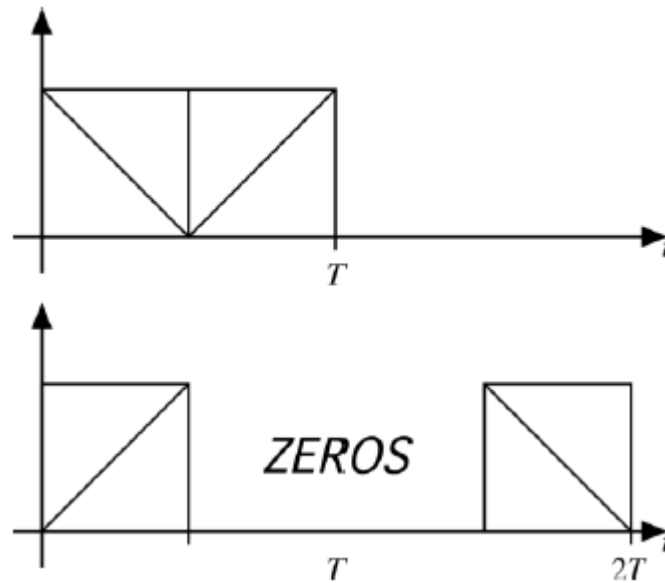


Figure 3.2: Visual depiction of the zero padding and rotating process for FMCW radar data processing [45].

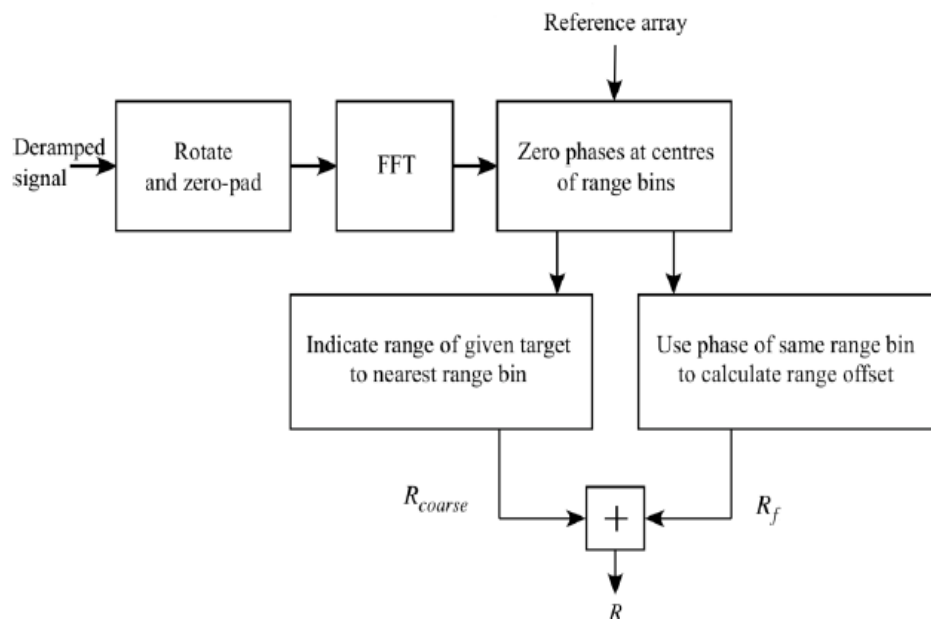


Figure 3.3: Diagram of the phase sensitive FMCW radar range measurement process [45].

When a point target return phase is at the centre of the range bin, it would be convenient if it is normalized to zero phase. By multiplying the deramped signal with the phase conjugate of the estimated phase at the centre of the range bin, return phase is normalized. The reference phase would be:



$$ref = e^{(-j(\omega_c\tau - K\tau^2/2))} \quad (3.8)$$

### 3.1.2 Time delay error correction

One important factor that has to be taken into account is the correction due to mis-synchronization of the ADC and the FM signal. This will cause a time delay which would correspond to a certain phase delay. The quantitative value of the phase delay  $\Delta\phi$  is,  $2\pi f_d \Delta t$ , where  $f_d$  is the deramped frequency of the target return and  $\Delta t$  is the time delay.  $f_d$  can be expressed in terms of range bin and pulse duration as  $(b-1)T$ . The threshold of time delay is  $\pm 0.75 \mu\text{s}$  while taking the desired range precision and maximum range into account. The value can be calculated by using Equation (3.5). The estimated range error due to the delay becomes as follows:

$$R_e = \frac{\lambda \Delta\phi_d}{4\pi} = \frac{BR\Delta t}{f_c T} \quad (3.9)$$

The value of  $\Delta t$  can be obtained from the above equation by considering maximum range for the radar application as 2 km and accepted range accuracy as 1 mm, which then becomes  $0.75 \mu\text{s}$ .

### 3.1.3 Range calibration with two reference values

During field operation, the radar will cycle through each 1 minute burst in every few hours. Due to the behaviour of the DDS development board, it was discovered that the precise timing of the initiation of the DDS chirp was subject to a small variation when the radar is turned on again from stand by. This causes a small error in timing that cannot be compensated by the time delay correction. The way to get around this is performing calibration by using two reference range values in order to correct both the timing errors that are independent of range (represented by  $a$ ) and the phase errors that result in range errors linearly-proportional to range (represented by  $b$ ). This method is practically plausible as many ice shelves have internal layers above the base

which are always constant in terms of depth. These internal layers ( $R1$  and  $R2$ ) can be taken as references and the range offset for calibration can be calculated as follows:

$$\begin{aligned}\Delta R1 &= a + bR1, \Delta R2 = a + bR2 \\ b &= \frac{\Delta R2 - \Delta R1}{R2 - R1}, a = \frac{\Delta R1 R2 - \Delta R2 R1}{R2 - R1} \\ -(\Delta R) &= -(a + bR)\end{aligned}\tag{3.10}$$

## 3.2 Phased Array Processing

Array processing deals with the processing of signals carried out by propagating wave phenomena [47]. The radar echo from a target is received by numerous sensors collectively referred to as an array. These sensors have a predefined spacing with respect to each other. The main goal of array processing is to extrapolate expedient features of the received signal field (its signature, direction, speed of transmission) [48].

The different sensors are expected to receive different but coherent signals from a target, although that may not be always the case. Receive signals can be non-coherent if those are independent of each other. The basic idea for array processing is to use the knowledge of sensor spacing and then synthesize the coherent signals from different sensors to obtain a combined result. According to the application, arrays are formed in different shapes [49] [50]. The simplest and widely used is the linear array, which is a collection of sensors in a straight line.

### 3.2.1 Developments in Phased Array Radar

Phased array radar uses the coherent phase change among the array elements of the received signal coming from a particular point target. Background on phased array radar system development can be found in [51]. The earliest development of phased array can be dated back to 1905, when Nobel laureate Karl Ferdinand Braun

experimentally proved the use of multiple antennas to dictate the direction of the transmitted wave [52]. The first phased array radar was built by Germany during World War II, which was named FuMG 41/42 Mammut [53]. It consisted of six or eight Freya [54] antennas and operated at VHF band (116-146 MHz). It had a range of around 325 metres and was deployed for detecting aerial vehicles of the enemy. Research on phased array theory became mainstream in the radar community in 1960s [6]. By the 1980s, due to advancement on antenna technology, phased array systems were being widely used for various applications and many research works had been published describing its working principle [55] [56] [57]. The main feature of a phased array system is the ability to electronically steer the beam. As mentioned above, phased array systems use antenna spacing information. This information is combined with the interference property of electromagnetic waves, eventually providing the ability to control the wave direction. It has the obvious advantage over any operation that requires to steer the antenna mechanically. Phased array radar has military applications as it is used by naval warships. The navies use it for finding ships on water and aircrafts and missiles on the air [58]. Phased array radars are also used for obtaining weather data [59] and even space research [60]. In this chapter the main focus would be on the signal processing aspect of phased array radar. The process phased array radar system uses for transmission or reception of signals from a certain desired direction, is known as beamforming.

### **3.2.2 Beamforming**

The beamforming technique is used to obtain information from a particular angle. It works according to the principle of constructive and destructive interference of electromagnetic waves. Due to antenna element spacing, each element will receive a signal coming from a certain direction at a slightly different time. This directly

corresponds to a phase shift among the signals. As the spacing is predetermined, the phase shifts would be coherent. By manipulating the phase, information from a desired direction can be achieved by inhibiting signals from all other directions. The process can be visualized from Figure (3.4). Assuming a signal arriving from certain angle  $\theta$ , it can be observed from the figure that there is a linear relationship between the times of arrival at different elements (as the element spacing being linear). If all the signals of antenna elements are added, due to the phase difference, the resultant signal will be a combination of both the sum and difference of the signals at various points. In order to achieve amplified signal, they all need to be aligned properly so while combining all of them, only constructive interference takes place. Thus, improved information from a certain direction is achieved by the help of combining signals from more than one antenna element.

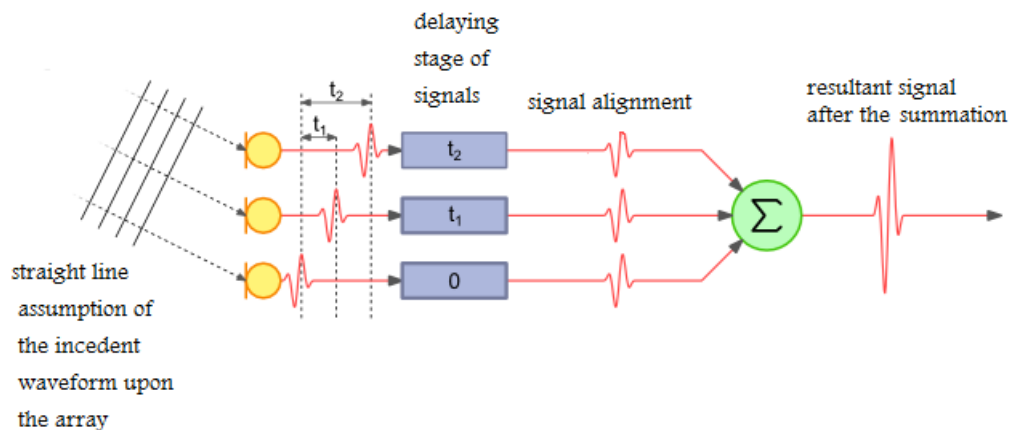


Figure 3.4: Illustration of basic concept for beamforming utilizing the constructive interference properties of waveforms.

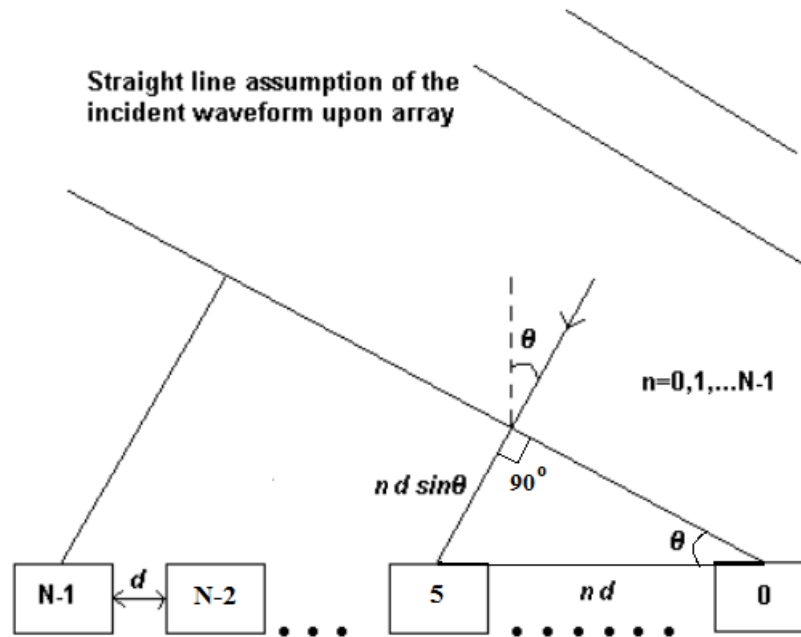


Figure 3.5: Geometric representation of a linear phased array system to calculate the time (hence phase) difference among array elements when arriving signal has an angle with normal to the array.

Figure (3.5) helps to quantify the phase difference between adjacent elements in a linear array. In this figure,  $N$  is the total number of elements in the array and  $n=0,1,2,3,\dots,N-1$  where  $d$  is the spacing between adjacent array elements. As all the elements are equally spaced, distance from any element from the reference (the first element) can be quantified as  $nd$ . Then, from the triangle drawn on Figure (3.5), extra path that a signal has to travel at each element due to the incident angle  $\theta$ , can be calculated. Converting the spatial value into phase gives the equation for phase delay:

$$\phi_{shift} = \frac{2\pi}{\lambda} nd \sin \theta \quad (3.11)$$

During beamforming, this equation is used for aligning the signals coming from angle  $\theta$ .  $4\pi$  should be used instead of  $2\pi$  to account for the two way signal propagation. It is obvious from Equation (3.11) that if the incident signal is parallel to the array ( $\theta=0$ ), there is no phase delay, hence just adding the original receive signals would give the resultant signal from that angle.

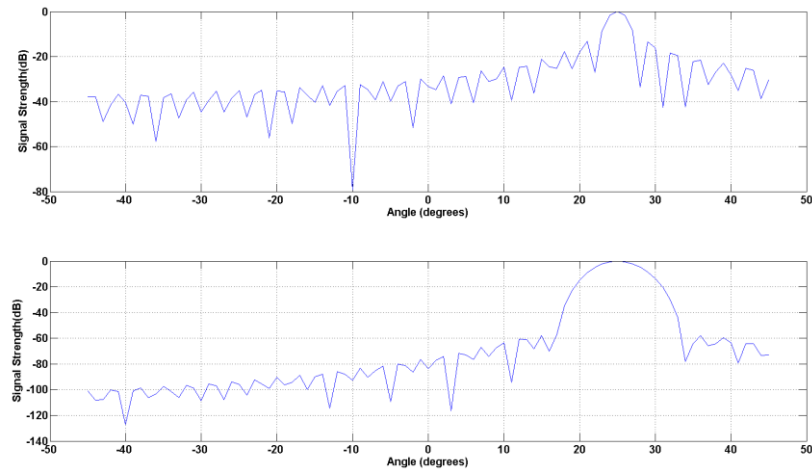


Figure 3.6: Beampattern for 25 degrees, without cross range windowing (above) and with cross range windowing (below). Blackman windowing has been used in both cases.

So, by delaying the array signals accordingly, scanning through any angle (obviously there would be limitation to scanning angle due to antenna beamwidth), normal to the array is possible. The beampattern equation for phased array signal processing becomes as follows:

$$B(\theta) = \sum_{n=0}^{N-1} w_n e^{\frac{-j2\pi n d \sin \theta}{\lambda}} \quad (3.12)$$

where  $w_n$  is the weighting coefficient defined by window function. Figure (3.6) shows how a signal from certain desired angle is synthesized by suppressing signals from all other angles. Here, a simulation is performed to calculate the beampattern for scanning angle +45 to -45 degrees. Beampattern is achieved by simulating a receive signal and then applying phase delay ( $\theta=25$  is used during simulation) to each elements and subsequently adding them all according to Equation (3.12). It can be seen from Figure (3.6) that signal strength at 25 degrees is around 15 dB higher than signal from other angles. It is seen that the peak sidelobe ratio (PSLR) is improved by implementing a window function across the array. During this simulation, Blackman window has been

used. The main lobe width has increased but this is the usual trade off while using window function.

Other important implication of Equation (3.11) and (3.12) is that the phase delay is a function of signal wavelength, hence frequency. So, using phase delay would not work properly in a wideband operation where this beam steering vector will not be a function of single frequency. So, beamforming in this process makes the beam squinted as perfect constructive interference is not achieved due to large number of frequency components. To resolve this issue, beamforming can be obtained by time delay instead of frequency delay. To convert the phase delay for beam steering into time delay one, relation between phase angle  $\varphi$  and time delay  $\Delta t$  can be used which is  $\varphi = 2\pi f * \Delta t$ . Here,  $f$  is the signal frequency. Substituting this value with Equation (3.10) gives the time delay factor and subsequently the time delay steering vector as follows:

$$\Delta t = \frac{nd \sin \theta}{c} \quad (3.13)$$

$$B(\theta) = \sum_{n=0}^{N-1} w_n e^{\frac{-jnd \sin \theta}{c}} \quad (3.14)$$

It can be seen that there are no frequency parameters in Equation (3.13) and (3.14).

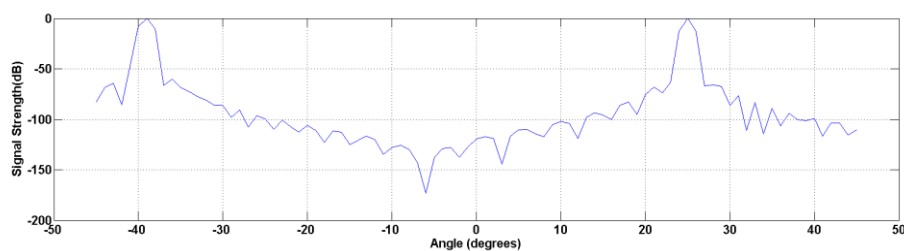


Figure 3.7: Illustration of grating lobe where on a beampattern for steering vector at 25 degrees, another strong signal appears at -38 degrees due to element spacing being more than half the signal wavelength.

### **3.2.3 Grating Lobes**

Phased array operation suffers from grating lobes (strong signal appearing at angles other than the beamforming angle) when unintended constructive interference occurs. This happens when the separation between antenna elements become quite large. If the distance corresponding to phase delay is multiple of the wavelength, combined signal cannot suppress the signals from unintended direction due to high correlation. To avoid grating lobes, the rule of thumb is to make sure the element spacing is less than half the wavelength.

### **3.2.4 Phased array imaging simulation**

A simulation for phased array imaging was performed to validate the ability of the algorithm to construct the image properly. The FMCW radar parameters were used according to Table 1. The simulation was made to image a point target with a range value of 1500 metres and 15 degrees away from the normal of the array. The simulation was performed for 36 elements (as in one of the field experiments, 36 element array was used). Element spacing is 0.25 metres which is well below the grating lobe threshold. Figure (3.8) and visually demonstrates the impact of phased array processing. As it can be seen, if the signals are added without phase shift, a constant signal return is observed throughout the scan. After the phased array processing, point target at 15 degree angle becomes properly noticeable. Both range and cross range windowing (Hanning window) have been used during simulation. The spreading of the point target even after beamforming is due to range migration effect which will be discussed in details in next chapter.



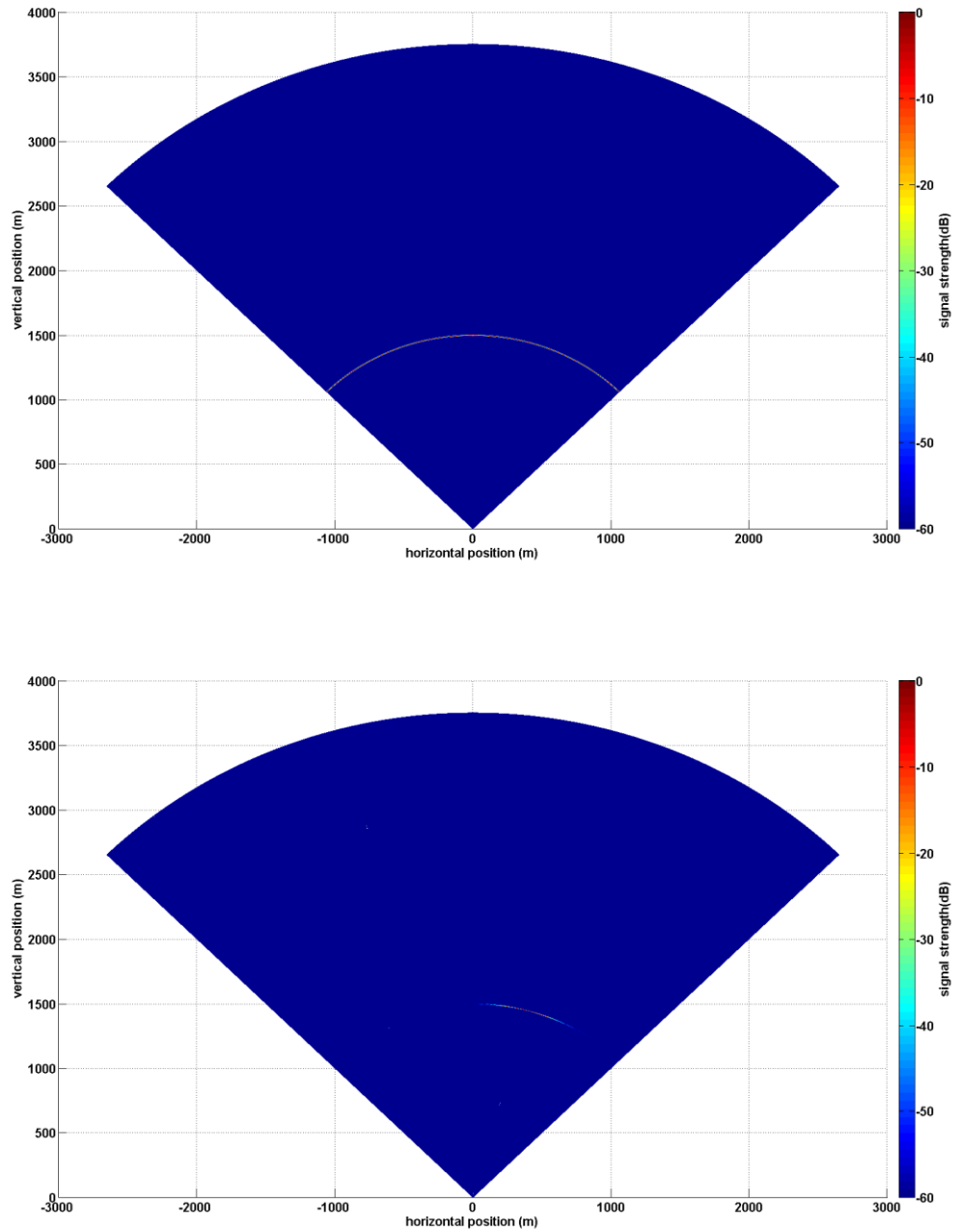


Figure 3.8: Image formation for a point target at 1500 metres range and 15 degree from the array: above one is without implementing beamforming and below one shows the image after applying phased array beamforming.

### **3.3 Synthetic Aperture Radar (SAR) Processing**

SAR signal processing is a 2-dimensional image processing operation. The fundamental difference between the conventional optical image processing and radar/SAR image processing is the latter one uses distances between the objects and the sensor to form an image. On the other hand, optical imaging systems such as cameras or optical telescopes use the angular differences of the signals incident upon the sensor to create an image. In the Antarctic Ice Shelf application that this thesis work concerns, no physical SAR system was deployed. But during the data collection, a SAR like experiment was performed by mounting the ice penetrating radar on a sledge and moving the sledge in a straight line while the radar collected data. This emulated the synthetic aperture. The idea was to analyse and compare the ice shelf images obtained from different processing methods (i.e. phased array and SAR).

#### **3.3.1 SAR geometry and Parameters**

SAR data acquisition method is similar to normal Radar when each transmitted pulse and corresponding reception is considered individually. After all, it is a standard Radar that is attached to the platform. However, there is one little assumption that is commonly made. Radar signal travels in the speed of light and the platform speed is extremely slow compared to light speed. So, during the time the Radar receives the reflection from the ground, platform's movement can be neglected. This is known as the 'start-stop' assumption [61].

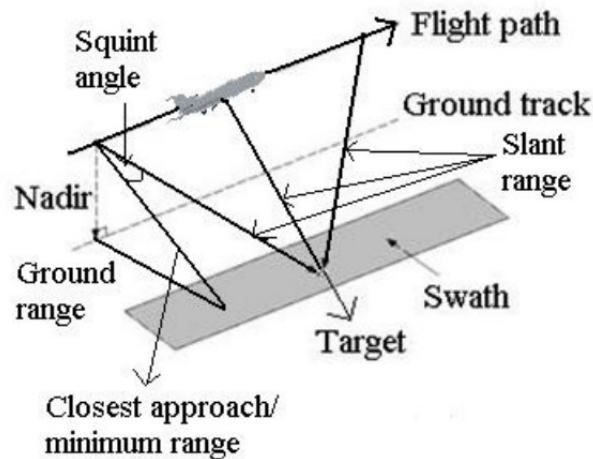


Figure 3.9: SAR operation geometry.

As mentioned, SAR typically requires precise positional information of the platform during its data gathering period. Method of SAR processing which is independent of platform position has also been researched [62]. With the platform information, it can calculate the necessary factors that determine the relationship between the platform and the target at each moment. These parameters that are obtained from the geometry along with the terms that are essential during computation are discussed below.

*Platform velocity:* All along the flight path, platform velocity is tried to be kept constant. In practical cases, the platform velocity does not remain the same for the whole flight. This causes error in computation and has to be accounted for, which is known as ‘motion compensation’.

*Integration angle:* It is basically the angle measured from the two endpoints of the flight path. It determines the eventual synthetic aperture length. The angle is determined by the lines drawn from two ends of the flight path joining the point target. Thus, Synthetic aperture length ( $L$ ) can be calculated from the integration angle.

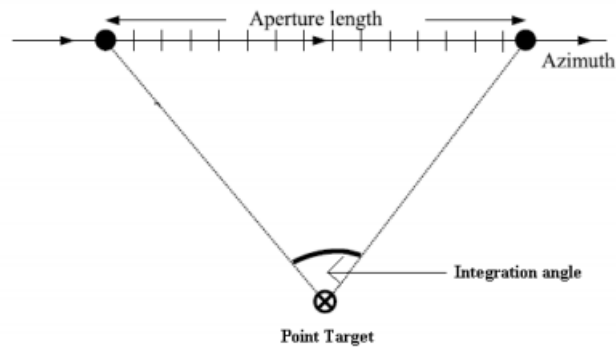


Figure 3.10: SAR integration angle and point target.

*Ground range:* This is the two dimensional measurement of the distance between the point target and a point from the flight path line's projection on the ground plane (Figure (3.9)).

*Slant range:* In the simplest of ways, it is the length of the straight line drawn between the platform and the target at each given moment during the flight. This is the evident distance travelled by the transmitted pulse from platform to target. As seen from Figure (3.9), it is a three dimensional measurement. At each transmission/reception period during flight, slant range is been determined. The measurement is done by simple Pythagorean geometry.

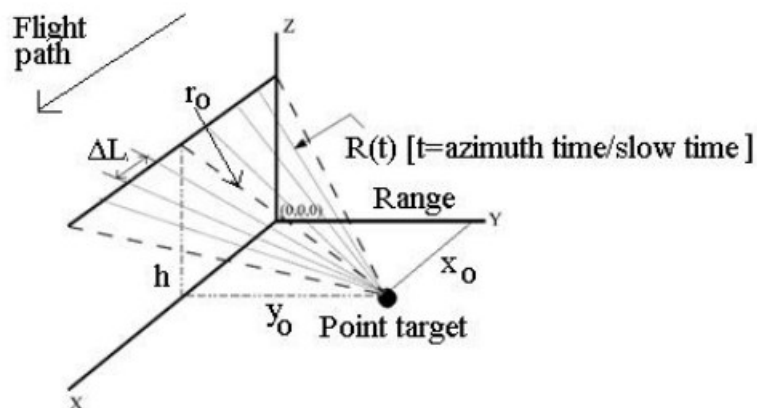


Figure 3.11: SAR range calculation geometry.

While calculating range, along with the known platform parameters, the minimum range to the target ( $r_o$ ) is known. An important fact should be known about monostatic SAR in terms of minimum range. The target and the platform should be separated by at least a distance defined by the time duration of switch of the radar from transmission to reception mode. The radar cannot detect target closer to this value as that target echo will reach the radar before it is switched into receiving mode. This time delay depends on the transmission pulse width as radar cannot switch before the whole pulse is been transmitted. From Figure (3.11), the point target has coordinate values in x and y axis. As seen from the geometry,  $x_o$  is just the displacement in x axis from the origin and  $y_o = \sqrt{r_o^2 - h^2}$ . Having  $v_{pl}$  and  $h$  as the platform velocity and altitude respectively, the equation becomes,

$$R(t) = \sqrt{(v_{pl}t - x_o)^2 + y_o^2 + h^2} \quad (3.15)$$

### 3.3.2 SAR Doppler History

For a single platform position, SAR signal return is collected. This recording is continued throughout the flight. For a given aperture length, the number of recorded sample depends on PRF. They are placed in the two dimensional memory of azimuth and range. This is the raw SAR signal history which is used for image processing by applying suitable algorithms. This is also called as SAR phase history. During the flight, range between the platform and target is changed. The range equation is hyperbolic, thus SAR signal history is the hyperbolic curve shown in the right side of Figure (3.12). The inherent Doppler Effect on SAR signal is apparent from the figure. The hyperbolic nature of the range migration curve depends in the integration angle. Larger integration angle makes the hyperbolic nature more obvious. During SAR

processing, the phase history obtained from the Doppler Effect is stored and used to create focused image.

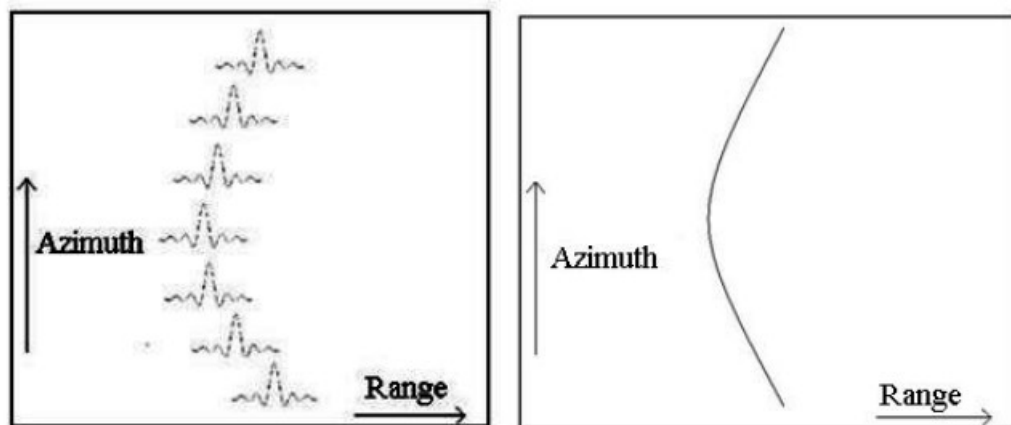


Figure 3.12: SAR signal history stored in a two dimensional memory and the corresponding range migration curve.

SAR treats the receiving echoes as being emitted from a stationary target and calculates accordingly for focusing. For a moving target, this coherence is lost and the image gets unfocused [63], [64].

### 3.3.3 SAR Image Processing Algorithms

SAR raw data is a hyperbolic curve which essentially represents the point target. It is the job of the implemented algorithm to translate this spread out data into a focused image. Two main factors for selecting an algorithm are accuracy and efficiency. SAR can deal with enormous amount of data depending on the integration angle. So, a very accurate algorithm might become inefficient in terms of computational load and processing time. Inversely, a very swift algorithm is often expected to produce poor image resolution. So, a seemingly simple solution for image processing by a 2-dimensional matched filter is practically not that proficient. Because conventional image processing algorithms do not account for the Doppler changes in general. So many corrections have to be made continuously to compensate for the induced offset due to varying range while progressing in azimuth. So, algorithms have been

developed to perform SAR image focusing distinctively. All the developed algorithms can be grouped into two main categories in terms of their working domains, Time domain algorithms and Frequency domain algorithms. As the signal history is obviously in time domain it can very easily be converted into frequency domain by performing two-dimensional Fourier transform. Actually, it is quite common in image processing field to manipulate images in frequency domain. The reason is its mathematical advantages as different features of an image correspond to different frequency elements.

Time domain algorithms by and large are based on the back-projection principle [65]. The concept is very simple. Each signal is back-projected onto the image plane according to their direction vector. Same value is assigned throughout that direction. The process is carried out for every sample and the summation gives out the final image. The most widely used such algorithm is Global Backprojection (GBP) algorithm [66]. It provides accurate image but it has high computational cost due to its large number of required operation to produce an image. Other algorithms such as Fast Backprojection (FBP) [66] or Fast Factorized Backprojection (FFBP) [9] have taken mainly developed to reduce this cost.

Meanwhile, frequency domain algorithms rely quite understandably on Fourier transformation. In fact, they use the Fast Fourier transform (FFT) technique. This feature brings the opportunity for speedy computation. These algorithms generally try to nullify the Doppler Effect by performing interpolation or complex multiplications in frequency domain. Prominent algorithms of this kind are Range Doppler algorithm (RDA) [5], Chirp Scaling Algorithm (CSA) [67] and Range Migration Algorithm (RMA) [61]. RDA was the first developed algorithm for SAR processing. In this algorithm, SAR data is converted into range Doppler domain by azimuth FFT

operation. Ensuing operations are performed in this domain to achieve the image. CSA is applied in a two-dimensional frequency domain that avoids the use of interpolation. This is an advantage because other frequency domain algorithms often require a very sophisticated interpolation procedure that inflicts computational cost on the processor. RMA on the other hand has the ability to operate with large integration angles and does not discard any phase term unlike CSA or RDA. Although, motion compensation in frequency domain is not that convenient, especially for large integration angle as it further increases the computational cost. In essence, algorithms of both domains inherit different features in terms of image quality and processing time [65] [66]. The selection of one algorithm will entirely depend on the integration angle, threshold for acceptable resolution, and platform parameters.

#### **3.3.4 Range Migration Algorithm**

The Range Migration Algorithm (RMA) has been used for SAR data processing during the period of the work presented in this thesis. RMA originated from seismic signal processing [61]. Seismic data processing can be considered strongly analogous to SAR. For seismic data processing, some numbers of geophones are placed along a straight line on the ground and a charge is detonated along that straight line. Every single geophones uses the received sound echo and the combined results are analyzed to figure out ground characteristics. This placement of geophones is similar to specific SAR platform positions at any given instance. During the processing of seismic received data, the phase function is hyperbolic just like SAR phase data.

RMA is a frequency domain algorithm where frequency conversion is applied in both azimuth and range axes. The technique that RMA uses to focus the image is based on downward continuation i.e. the downward propagation of the wave equation [61]. This property has also been drawn from seismic migration. The key feature of RMA



is the change of variable in the range axis. This method was developed by Stolt. He developed this variable swap method in frequency domain which is commonly known as ‘Stolt mapping’ or ‘Stolt interpolation’ [68]. He implemented this method to provide a proper solution for the wave equation for seismic data processing [5].

RMA has four major segments in terms of its functionality. Each has distinct features and purposes that sequentially lead to create a focused image of the area viewed by the radar aperture during the SAR flight.

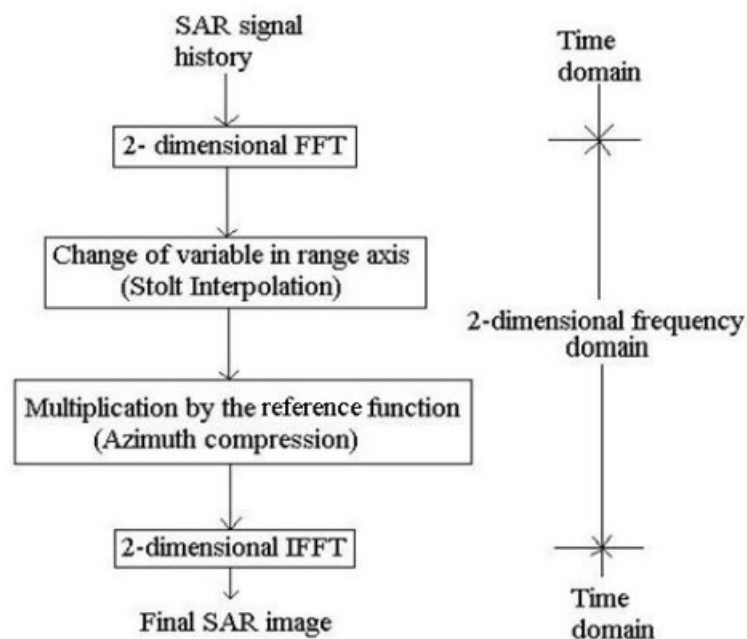


Figure 3.13: Block diagram of the Range Migration Algorithm.

The first and last steps are just domain converters. The main algorithm is applied to the signal data by the two steps in the middle. It is seen from the Figure (3.13) that those steps are performed in frequency domain. Just to set the premise before going into comprehensive discussion on these two steps in the middle, the image formation process of a single stationary point target will be the basis for the explanation of the algorithm. This will elucidate the algorithm at its core. When more point targets are added, the algorithm works exactly the same way on each of those point targets in

concern. Complex geometry of the scene increases complexity of the computation, not the complexity of the algorithm. However, it is prudent to keep in mind that the relativistic effect between platform and point target velocities has been ignored. This is done because every point target is assumed to be stationary relative to ground. So, this approximation of all point targets being stationary will not work while focusing a moving target.

*Stolt Interpolation:* FFT operation on both the axes of the SAR Doppler history turns the axes variables into their corresponding frequency domain variables. The coordinates of the SAR signal plot are the azimuth position of the platform and the fast time. Fast time is directly proportional to the range,  $R$ . So, the range axis variable is a time parameter whereas the variable for azimuth axis is spatial. So, the corresponding frequency domain representation of the data plane will contain a frequency term as well as a wavenumber term. Wavenumber is the spatial version of frequency. As frequency is measured in radian/sec, unit for wavenumber is radian/meter. Usually, the wavenumber is expressed by  $2\pi/\lambda$ . What is important here is that wavelengths for azimuth and range are direct products of the radar wavelength.

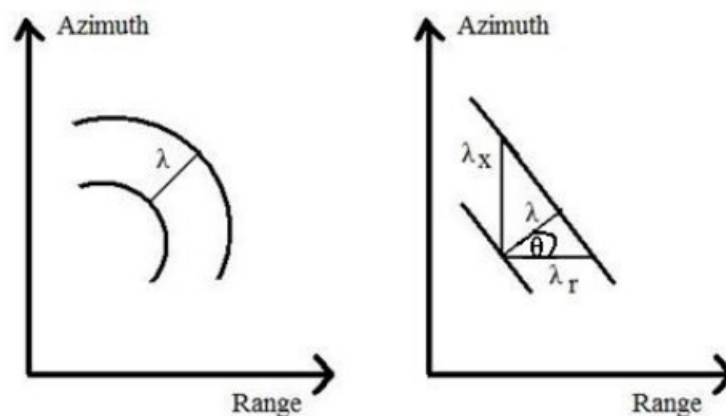


Figure 3.14: Far field radar antenna radiation pattern and its straight line approximation within minute area.

Figure (3.14) illustrates the wavelength concept of SAR signal. The radar wavelength  $\lambda$  projects to the direction of the wave propagation while  $\lambda_r$  is related to the ground range.  $\theta$  is the squint angle. As wavefronts are parallel to each other, the relationship among these three wavelengths can be evaluated from this above simple geometry [5].

$$\lambda_x = \frac{\lambda}{\sin \theta} \quad (3.16)$$

$$\lambda_r = \frac{\lambda}{\cos \theta} \quad (3.17)$$

Now, if the frequency domain representation of the azimuth axis spatial variable  $x$  is denoted as  $k_x$ , ground range variable  $r$  as  $k_r$  then,

$$k_x = \frac{2\pi}{\lambda_x} \quad (3.18)$$

$$k_r = \frac{2\pi}{\lambda_r} \quad (3.19)$$

The frequency domain representation for the remaining plane, the slant range plane, is merely the angular frequency of the signal as it changes proportionally with the fast time. Substituting the values of Equation (3.16) and (3.17) into Equation (3.18) and (3.19),

$$k_x = \frac{2\pi}{\lambda} \sin \theta \quad (3.20)$$

$$k_r = \frac{2\pi}{\lambda} \cos \theta \quad (3.21)$$

Squaring the above two equations and adding them together,

$$k_x^2 + k_r^2 = \left(\frac{2\pi}{\lambda}\right)^2 \quad (3.22)$$

Substituting the value of  $\lambda$  with wavelength  $\omega$ ,

$$\omega = \frac{c}{2} \sqrt{k_x^2 + k_r^2} \quad (3.23)$$

Equation (3.23) demonstrates the relationship between the frequency domain variable of the ground range with the frequency domain variables of the SAR signal history. Stolt interpolation uses this equation to change the variable of the range axis. Since the phase reference is assigned in closest approach point (between platform and target), the Doppler shift at that point is evidently zero. The main goal is to eradicate the Doppler shifts of other points taking the closest approach point as reference. In zero Doppler position, azimuth spatial frequency  $k_x=0$ . So, in zero Doppler, a relationship exclusively between  $k_r$  and  $\omega$  becomes  $k_r= 2\omega/c$ . This can be exploited to make the interpolation more efficient. For every  $k_x$ , the phase corrected value for  $\omega$  can be found by shifting it with respect to the center frequency. It gives the new value,  $\omega' = \omega - \omega_c$  ( $\omega_c$  is the centre frequency). This new value can be calculated from Equation (3.23) where  $k_r$  is measured according to the new value of  $k_r$  (with respect to reference phase). This whole manipulation produces the new equation:

$$\omega' = \frac{c}{2} \sqrt{k_x^2 + k_r^2} - \omega_c \quad (3.24)$$

By means of the above equation, the effect of the variable change is achieved. Interpolation operation is imposed to rectify the values of  $\omega$  with corresponding values of  $\omega'$ . What is to observe here is that the Stolt interpolation basically transforms the azimuth slant range plane of the raw SAR data into azimuth ground range plane. From the geometrical point of view, it can be said that the intrinsic Doppler Effect within the signal arises from the non-orthogonality of the two plains comprising the signal history. Since azimuth and ground range are orthogonal to each other, the deformation gets canceled out and so as the phase shift along the flight path [5] [69].

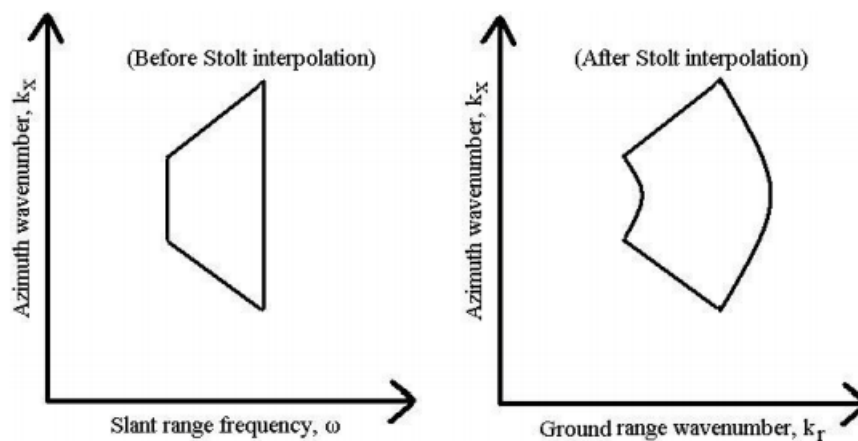


Figure 3.15: SAR signal energy in 2-dimensional frequency domain and Stolt interpolation's effect on it.

One interesting observation can be made from Figure (3.15) is the circular nature of the signal strength after the Stolt interpolation. Recalling Equation (3.23) helps to explain this matter. That is a rearranged version of the circle equation. This gives the visual representation of how frequency component replacement in range corrects the entire image spectrum. In practical cases, this nature hardly becomes apparent in narrowband SAR operation [61], but UWB SAR makes this property visually more evident because of its large integration angle.

*Azimuth Compression:* Before taking the signal data back into its original domain, azimuth compression needs to be performed to create the focused image. The signal information of a single point target which is spread out throughout the signal history has been taken care of by the Stolt interpolation in terms of phase shift correction. Still, the data needs to be converged in azimuth. This step is actually quite simple and does not possess any lengthy concept like the Stolt interpolation. An azimuth reference function [61] is multiplied with the signal spectrum to congregate the signal data smeared along the azimuth axis into the zero Doppler point. This is nothing but a matched filtering operation where the signal is correlated with the matched filter

function corresponding to the closest approach point. So, this multiplication just funnels in all the smeared data into the closest approach point or minimum range, for that is the ultimate position considered for the point target. After this compression, all the signal energy gets concentrated into this point hence creating the focused image (when it is transformed back to time domain). The azimuth reference function is as follows,

$$ref(k_x, k_r) = \frac{|k_r|}{\sqrt{k_x^2 + k_r^2}} e^{-j\left(r_o \sqrt{k_x^2 + k_r^2} - r_o k_r\right)} \quad (3.25)$$

The first part of the function is the magnitude which has little effect if integration angle is small. Combining the two operations, Stolt interpolation and azimuth compression, the raw SAR data is synchronized with the reference point phase, both in azimuth and range. In other words, the signal energy is entirely compressed. Now, the focused image with its full resolution can be seen simply by performing 2-D inverse Fourier transform of the signal spectrum.

### 3.3.5 SAR simulation Results

Before implementing SAR processing to real data, simulation of the range migration algorithm to focus point target was performed to ensure the algorithm was accurately translated into software routine (in this thesis work, MATLAB was used). Again, radar parameters were set according to Table 1. The simulation was performed to focus a point target with a range value of 350 metres (as the target return from SAR field experiment was from around that distance). Even though the integration angle in the field experiment was quite low (total SAR platform length was 22 metres only), simulation was performed for larger integration angle as well to visually observe its affect. Platform velocity was assumed to be very slow (1 m/sec.), as that would be the case during Antarctic experiment. Figure (3.16) shows the SAR signal history which

illustrates the conventional hyperbolic pattern. Figure (3.17) shows the focused point target after implementing RMA. Here, the integration angle is 30 degrees. Figure (3.18) shows the focused image of the same point target when the integration angle is very small (corresponding to 22 metres of synthetic aperture length only). The degradation of target focusing quality due to low integration angle can be easily compared from Figure (3.17) and (3.18).

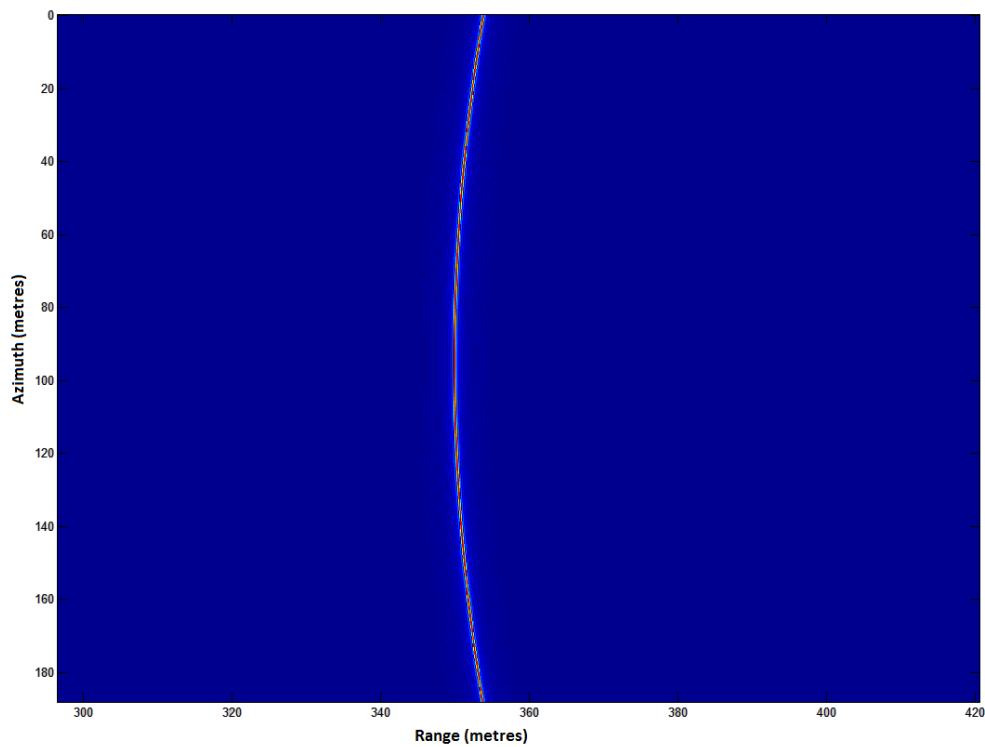


Figure 3.16: SAR signal history of a point target return before implementing focusing algorithm showing the characteristic hyperbolic curve (image is zoomed in).

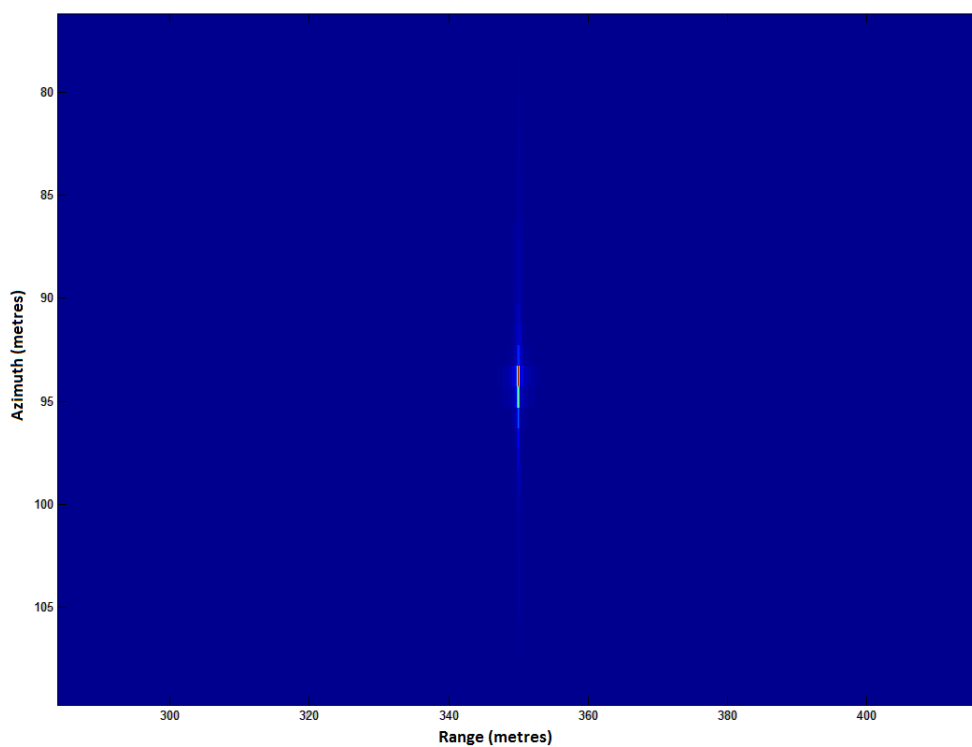


Figure 3.17: SAR point target image after implementing Range Migration Algorithm with 30 degree integration angle (image is zoomed in).

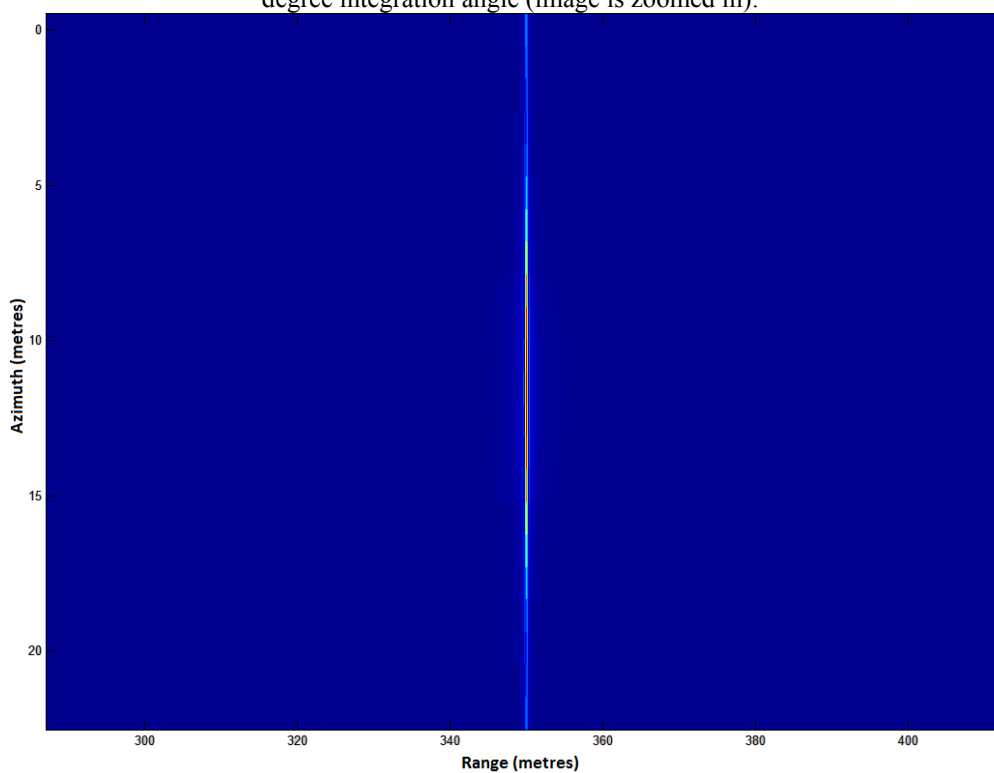


Figure 3.18: SAR point target image after implementing Range Migration Algorithm with 3.6 degree integration angle (image is zoomed in).



### 3.4 MIMO Processing

Multiple-input-multiple-output (MIMO) radar is a relatively new technology in the radar field. The basic idea is to add waveform diversity to the conventional phased array system, obtaining enhanced performance. According to [70], the definition of waveform diversity is the ability to adapt and diversify dynamically the waveform to the surrounding environment which would deliver enhanced performance compared to a non-adaptive system. In [71] [72] [73], the prospect of using multiple signals in an array system has been studied. The waveform diversity is also achieved by utilizing the antenna element spacing [74]. In various research papers, superiority of MIMO radars has been discussed (ability to obtain better resolution [73] [75], better capability for slowly moving target detection [76], ability to identify more targets than phased array [74], better application for adaptive array techniques [74] [77] [78]).

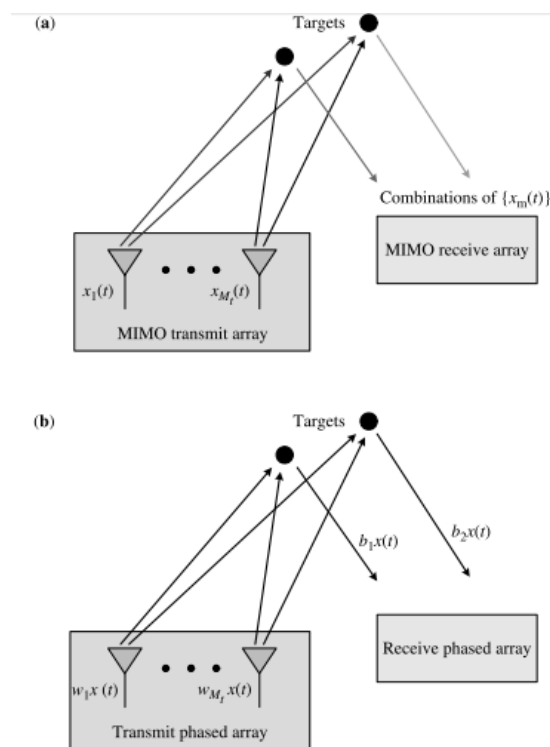


Figure 3.19: Comparison of (a) MIMO radar and (b) Phased array radar [79].

Figure (3.19) illustrates the fundamental geometric principle of MIMO radar compared to conventional phased array radar. It can be seen that MIMO radar uses waveform optimization by transmitting different signals from different elements. When compared to phased array, the potential to extract more information by MIMO can be intuitively understood. As a return echo from a target received by phased array system is essentially the same signal with a phase difference, signals on the receive side cannot be distinguished in terms of the parent transmitter. On the other hand, in MIMO, it can be easily achieved because each individual transmitter emits signals unique from the others. This information can be used for obtaining better angular information as well as for improved detection performance, resolution and SINR (signal to interference plus noise ratio) [80]. MIMO receivers use matched filtering to obtain information coming from each individual transmit antennas. The transmitted signals in MIMO radar are orthogonal even though it's not a strict requirement. The orthogonality is usually achieved by time or frequency division multiplexing. Also, in terms of antenna spacing, MIMO radar system can be colocated or sparse. For further details on those can be found in [79].

### **3.4.1 MIMO Virtual Array**

The unique concept of MIMO system is the virtual array configuration. Here, a brief understanding of this concept is provided. More elaborate studies on this can be found in [75] [76] [81].

Virtual array in MIMO is achieved by utilizing its waveform diversity property. The advantage of using virtual array is that during processing, the MIMO array consists of elements more than the number of elements that are physically in operation. MIMO

virtual array positions are determined by convoluting transmit and receive antenna element positions.

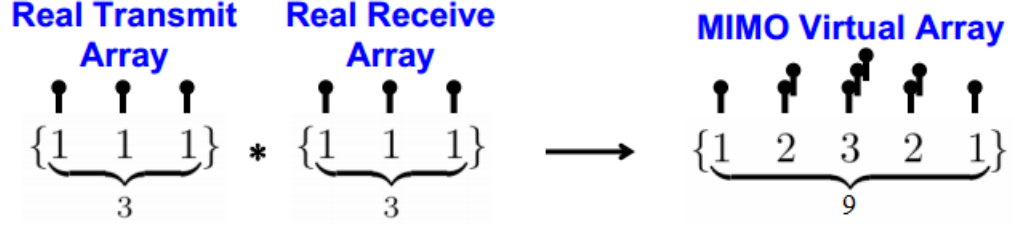


Figure 3.20: Illustration of MIMO virtual array positioning by convolution of real and receive array.

It can be seen from the above figure that the convolution process increases the number of array elements to 9 where there are 6 elements present physically. Assuming there are  $M$  number of transmit elements and  $N$  number of receive elements. The signal transmitted by  $m$ th antenna is denoted as  $\varphi_m(t)$ . Each transmitting signals are orthogonal. At each receive element,  $M$  number of matched filters are used to extract all orthogonal signals. Which corresponds to total number of receive signals equal to product of  $M$  and  $N$ . This is the total number of virtual array element which is  $NM$  produced by using  $N+M$  number of physical elements. According to the geometric position of the real array elements, equation for the target response at  $n$ th receiver for the signal coming from  $m$ th transmitter can be written as [80]:

$$y_{n,m}(t) = a e^{j\frac{2\pi}{\lambda} \mathbf{u}_t^T (\mathbf{x}_{T,m} + \mathbf{x}_{R,n})} \quad (3.26)$$

where  $\mathbf{u}_t$  is a unit vector pointing toward the target from the array.  $\mathbf{x}_{T,m}$  and  $\mathbf{x}_{R,n}$  are the positional vectors for  $m$ th transmitter and  $n$ th receiver respectively where  $a$  is the signal amplitude. It should be noted that location of transmission and receive elements attribute directly to the obtained phase difference.

### 3.4.2 MIMO Beamforming

For a uniform linear array, the location vectors can be defined according to Equation (3.11). So, the Equation (3.26) can be written as:

$$y_{n,m}(t) = a e^{j\frac{2\pi}{\lambda}(nd_r \sin \theta + md_t \sin \theta)} \quad (3.27)$$

where  $d_r$  and  $d_t$  are the element spacing for transmit and receive antennas respectively. This equation is used to create steering vector for MIMO. One basic difference of beamforming in MIMO from phased array beamforming is unlike phased array, virtual beamforming can only be achieved in receive side. During the thesis work, MIMO imaging simulation was performed to analyze the possibility of using MIMO radar for snow Avalanche imaging in Alps. Simulation results concluded that SNR would be too low to create any meaningful signal by MIMO transmission, hence the project stuck with its previous conventional phased array system. The simulation results are added in the appendix. MIMO beamforming and imaging technique developed for three-dimensional imaging of Ice Shelf base is described in Chapter 4.

## Chapter 4

# Novel Signal Processing Techniques for Antarctic Data Analysis

This chapter consists of the signal processing techniques that were developed during this thesis work for the Antarctic data analysis. The development of these techniques was motivated by the practical necessity stemmed from Antarctic ice shelf experiments. The first section describes the Shape matching algorithm that has been developed to detect internal layers underneath ice shelves. The second section describes the range migration correction for fixed phased array radar. The range migration within the array is caused by the high fractional signal bandwidth, which is the case with the FMCW radar that has been deployed in Antarctica. Also, adverse weather conditions in Antarctica make it very hard to maintain the geometric parameters of the experiments accurately. Hence, vertical antenna spacing error correction method by using reference phase has been developed. The final section encompasses the 3-D MIMO imaging algorithm that has been developed for 3-D imaging of the Antarctic Ice Shelf basal layer.

### 4.1 Shape Matching Algorithm to Detect Internal Layers

Due to unavoidable practical reasons, it is very difficult to take precise radar measurements in the challenging conditions in Antarctica. Events such as snowfall, heavy wind can change the vertical position of the radar antennas. When the expected accuracy is within millimeter range, slightest change of the radar position can become

significant. So, it is always desired to have some fixed reference that can be used for calibration and to compensate due to the displacement error.

The internal layer is a feature of the ice shelf that can be used for this calibration purpose. Internal layers can be detected by radio waves. Mainly, these are sections underneath the ice shelves where there is a contrast in the ice dielectric property [82]. Due to this contrast, an ice crystal fabric is formed that reflects radar transmitted signal. Most importantly, these layers are prone to have fixed coordinate. They usually do not drift vertically hence can be used as a very good reference point. If the radar shifts from its initial position, it can be easily determined by the presence of an internal layer. By adjusting for these discrepancies using internal layer, very high precision results can be achieved from the radar data. Not only for radar data calibration, but internal layers also have great interest among geoscientists in their field of study of ice in Polar Regions. Even though it is desirable to get an internal layer while profiling an ice shelf, it is not that straightforward to detect. Generally, internal layers are located close to the ice shelf surface (within 100 meter). Due to antenna direct coupling, radar range profiles consist of large returns in close range. These high signals tend to obscure return radar signal from any actual ice shelf feature (i.e. internal layer).

In radar signal processing, it is always a big challenge to detect the presence of a target with low signal to noise ratio (SNR) or signal to clutter ratio (SCR). A number of research works have been performed to solve these problems [83] [84]. In most cases, the problems are application specific, each requiring different modelling and assumption of the target scenario.

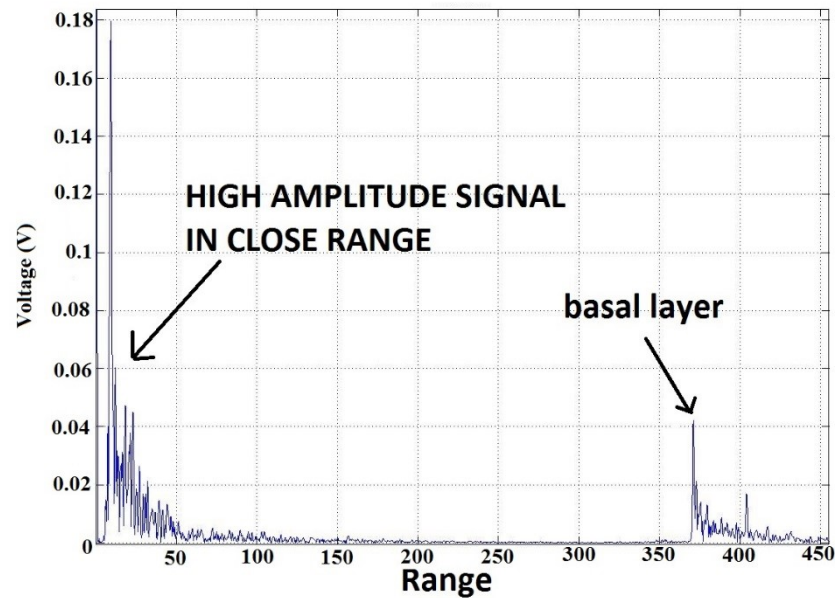


Figure4.1: Range profile of Antarctic ice shelf illustrating basal layer return and strong return in short range.

Figure (4.1) illustrates a typical FMCW radar range profile of Antarctic ice shelf. In Antarctic ice shelf measurement, the strongest return is expected from the basal layer. As it is seen in Figure (4.1), even though the basal layer is quite apparent, the profile encompasses high amplitude signals at very close range. These close range signals can be attributed to antenna direct coupling and/or clutter in the vicinity of the ice shelf surface. Usually, internal layers are expected to be present in that close range region. For radar target detection, a cross-correlation method is usually used when there is a known target [85]. In the particular case of ice shelves, the assumption is that both basal layer and internal layer will give out specular reflections. Hence, intuitively, cross-correlation method ought to work for internal layer detection. Meanwhile, it is obvious that cross-correlation will be ineffective when the antenna direct coupling signals are significantly higher than the actual basal layer return.

A proposed method to overcome this problem is described below. This shape matching algorithm works in following four steps-

- Defining the reference signal from the range profile.
- Amplitude scaling.
- Point by point difference calculation and ranking.
- Integration for overall rank

#### 4.1.1. Defining the reference signal from range profile

In shape matching algorithm, the first task is to determine the signal portion that is to be matched throughout the profile. In general case, a known expected pattern can be used or a copy of the transmitted signal (used in matched filtering). The idea is to slide this reference signal over the whole profile just like conventional correlation. In this case, the mathematical operation during the sliding would be entirely different. In the specific case of ice shelf, return signal from the ice shelf base will be the obvious reference. As mentioned before, internal layers are likely to have similar signal shape due to same reflection property. So, a portion of the signal from the profile that corresponds to basal return is chosen as a reference. Care is needed while selecting the reference cut, as that chunk would be the main reference. The more similar a signal portion in any other part of the profile is to this reference, the more likely that portion would be a radar target rather than clutter/noise. For the derivation, radar received signal is defined as  $s(n)$  having  $N$  number of samples ( $n=1,2,\dots,N$ ). Reference signal is defined as  $ref(m)$  consisting of  $M$  number of samples ( $m=1,2,\dots,M$ ).

#### 4.1.2. Amplitude scaling

The main reason that the generic cross-correlation weighting does not work for hidden feature detection is high amplitude values in close range. Amplitude scaling is thus performed to eradicate that problem. When  $ref(m)$  is being slid over  $s(n)$ , the portion of  $s(n)$  that  $ref(m)$  is covering will be scaled with respect to  $ref(m)$  amplitude.



It can be implemented in terms of average amplitude of the signal portion or the maximum amplitude. As the eventual goal of the algorithm is to determine whether the compared signal portion matches with the reference, using the maximum amplitude would be just fine. The portion of  $s(n)$  covered at a moment by  $ref(n)$  is defined as  $g(m)$ . It should be noted that  $g(m)$  will have less than  $M$  samples while  $ref(m)$  is crossing the edges. This will not hamper the scaling as only the maximum value is used. The scaling would be linear hence the scaling factor would be the difference between the maximum amplitudes. Thus, the scaling factor  $sc$  is,

$$sc = \text{maximum}(ref(m)) - \text{maximum}(g(m)) \quad (4.1)$$

Then,  $s_{sc}(m)$ , which is the scaled version of  $g(m)$  would become:

$$s_{sc}(m) = g(m) + sc \quad (4.2)$$

#### 4.1.3. Point by point difference calculation and ranking

After the amplitude scaling, the difference between the overlapping points will be calculated. This will be done for all the overlapping points ( $M$  number of points except when  $ref(m)$  is at the edge). When the difference is calculated, it should be then ranked in terms of likelihood. For instance, if the difference is zero, then it will get highest ranking point (1 for normalized calculation). For determining ranking value as zero, a threshold will be required. Zero would mean the points do not match at all. Choosing the threshold value is quite critical and would depend on the overall amplitude distribution of the actual signal. While defining threshold, it is better to calculate the amplitude ratio between overall maximum signal amplitude and maximum amplitude of the reference. A common case during specifying threshold is the tradeoff. Large threshold is likely to help revealing more features but also unwanted signals. On the other hand, very small threshold may lead to failed detection of a desired target.

If the point to point difference exceeds the threshold, the ranking value for that particular point would be assigned as zero. So, if the variable for ranking value is defined as  $rank$  and threshold is defined as  $thr$ , it can be written that if  $|s_{sc}(m) - ref(m)| > thr$ , then  $rank=0$ . Meanwhile, if that is not the case, a positive nonzero value would be assigned for rank. The Equation for  $rank$  then is as follows:

$$rank = \left(1 - \frac{|s_{sc}(m) - ref(m)|}{thr}\right)^2 \quad (4.3)$$

Equation (4.3) shows that when there is a perfect match (both the points have same values), the value of rank is 1. Absolute value is used as only the difference between the two points is of concern, it does not matter which point is greater or smaller. Validation of this criterion should also imply that the algorithm focuses solely on how similar the shape of the signal portion is to the reference. The squared term in the Equation is used to amplify the features, so a significant target return gets visibly more apparent than its neighboring points.

#### 4.1.4. Integration for overall rank

When values for rank are calculated for all the points, these values will be then added up to get the cumulative ranking value. So, overall ranking value  $rank_{cumul}$  for this one set of calculation would be,

$$rank_{cumul} = \sum_{m=1}^M rank(m) \quad (4.4)$$

It should be noted again that when part of  $ref(m)$  is outside  $s(n)$ ,  $m$  does not extend to  $M$  but calculations are carried out in same way. This is performed over the whole  $s(n)$ .  $ref(m)$  will slide over  $s(n)$  by gradually going forward by one sample point. As for every single increment  $rank_{cumul}$  is calculated, it will eventually have  $N+M-1$  values. Plotting all the values of  $rank_{cumul}$  (after normalizing) gives the new range profile.

The shape matching algorithm has been implemented on real Antarctic ice shelf data for internal layer location. The processed results are shown and discussed in the next chapter.

## **4.2 Range Migration Processing in FMCW Phased Array Radar**

As discussed in the previous chapter, range migration is a very common feature in SAR. This is due to the fact that SAR would usually have a very long synthetic aperture [86]. Methods for range migration correction are therefore an essential part of SAR research [61] [87]. In SAR, the radar moves along the aperture, so the range of particular target keeps changing. This creates migration and also Doppler shift. While processing the SAR image, this is taken into account (as the platform velocity is known) to create the resultant focused SAR image [88].

Meanwhile, in phased array/MIMO radar, the aperture length is very small compared to SAR. Also, there is no Doppler shift as the radar is stationary. This is why the range migration factor is generally neglected. Even though range migration is usually attributed to radar motion or very long aperture (in case of number of geophones placed for seismic activity measurement), phased array radar can also face range migration effect under certain conditions related to radar system parameters. Here, those conditions will be discussed. Along with that, the compensation process for range migration in phased array radar system, for instance an FMCW radar with deramped processing [11], is comprehensively described.

### **4.2.1. Criterion for significant range migration in phased array radar**

A phased array radar system is considered comprising a receive array with  $N$  elements of uniform spacing  $d$  arranged on a linear baseline, receiving an echo from

angle  $\theta$ . The target area is illuminated separately by transmit antenna. In order to understand the range migration effect, maximum scanning angle of the phased array radar needs to be determined in terms of array element spacing and signal wavelength. The equation (taking Nyquist sampling into account) is as follows;

$$\sin \theta_{\max} = \lambda_o / 2d \quad (4.5)$$

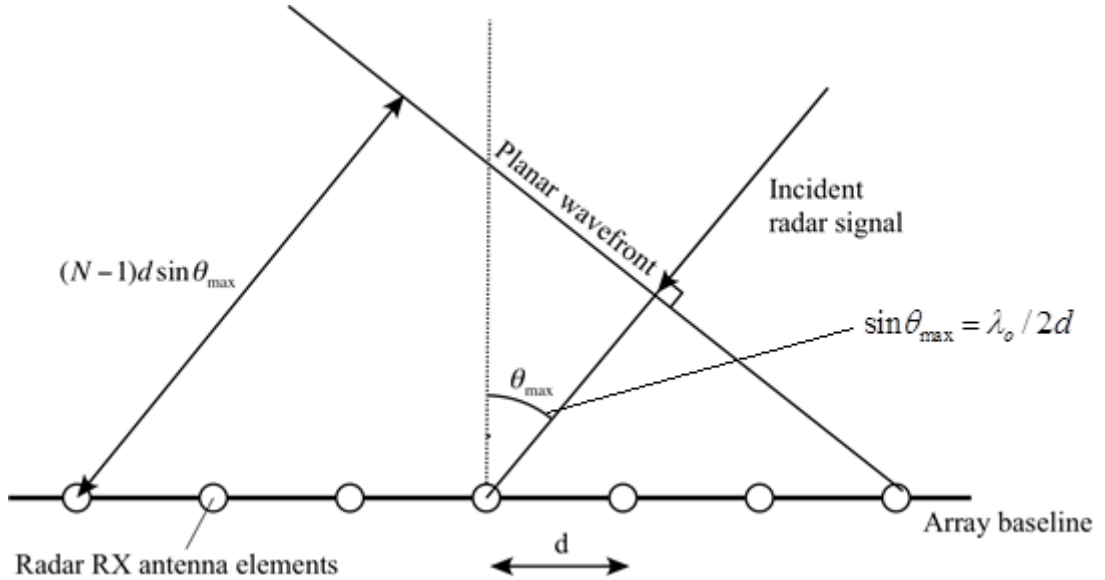


Figure 4.2: Construction used to determine the degree of range migration in a fixed phased array radar.

Where  $\lambda_o$  is the wavelength corresponding to the radar centre frequency. Equation (4.5) states that the radar may operate without grating lobes over a maximum angular range of  $\pm\theta_{\max}$  (for instance, the scanning range is then  $\pm 90^\circ$  if the array elements are spaced by half a wavelength). From Figure (4.2), it is evident that signal from non-zero angles will reach the different elements of the array at different times. Hence there will be variation in the signal distance within the array. For a target at the maximum scan angle, this path length variation,  $\Delta L$ , is given by,

$$\Delta L = (N-1)d \sin \theta_{\max} \equiv (N-1)\lambda_o / 2 \quad (4.6)$$

The range migration within the array will be equivalent to this (after compensating for two-way path). So,

$$range\_migration = \frac{\pm\Delta L}{2} = (N - 1)\lambda_o/4 \quad (4.7)$$

Where the expression for  $\Delta L$  is valid and accurate considering the target range is much greater than the array baseline. In cases where this does not apply, the signal distance variation within the array should be calculated in terms of exact geometry instead of using a  $\sin\theta_{\max}$  approximation.

By considering the general radar resolution,  $\Delta R = c/2B$ , range migration threshold will be twice  $\Delta R$  (as explained below).

The limit on array size beyond which range migration is significant may be established,

$$(N_{\max} - 1)\lambda_o / 4 = 2\Delta R = c / B \quad (4.8)$$

$$\text{hence, } N_{\max} = 1 + \frac{4c}{\lambda_o B} \equiv 1 + \frac{4f_o}{B} = 1 + \frac{4}{fractional\_bandwidth}$$

The above Equation determines the criteria of range migration correction for a particular phased array system. It can be seen that a system with high fractional bandwidth is prone to suffer from range migration within the array.

This gives a very simple expression to determine whether range migration processing needs to be applied. It indicates that a phased array radar with a higher fractional bandwidth and/or a larger number of elements is more likely to require range migration compensation. For instance, a radar with 20% fractional bandwidth may have up to 21 elements before range migration will have any effect. On the other hand, a phased array radar with 50% fractional bandwidth will cross the threshold if element

number is more than 9. Equation (4.8) does not have element spacing as a variable because it has direct correlation to the maximum scan angle.

To justify the assumption of maximum tolerable range migration being twice the range resolution, a point target response of a linear-FMCW range radar is considered, [89], with no windowing from a target at range,  $R_o$  along the  $x$ -axis. This gives the familiar result, [6], based on a sinc function:

$$V = A \operatorname{sinc}\left(\frac{\pi(x - R_o)}{\Delta R}\right) \quad (4.9)$$

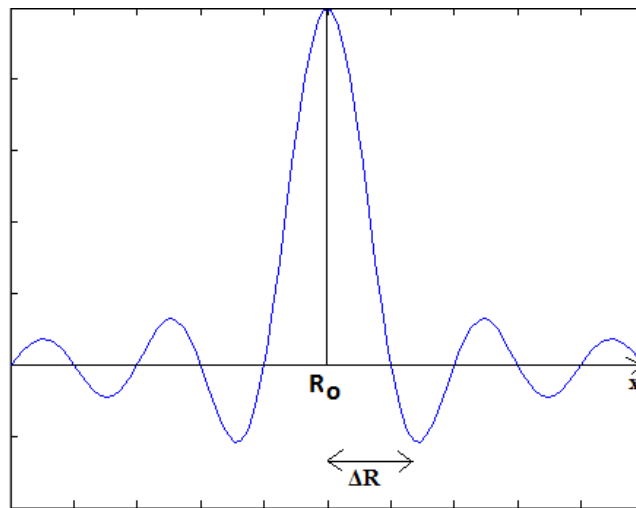


Figure 4.3: Illustration of maximum acceptable limit of range migration variation of range resolution,  $\pm\Delta R$  or  $2\Delta R$ .

Equation (4.9) implies that the point target response signal amplitude will fluctuate through the array. This is due to change in range,  $x$ . As can be seen from Figure (4.3), the value is maximum at the array centre of the sinc function and comparatively very low at the edges.

Therefore, the mean array output can be determined by taking an average across the array of the sinc function in Equation (4.9). The average is calculated by considering

a target at the maximum scan angle, where range migration threshold  $\pm\Delta R$ . The Equation then becomes as follows,

$$\begin{aligned}\bar{V} &= \frac{A}{2\Delta R} \int_{-\Delta R}^{\Delta R} \frac{\sin(\pi x / \Delta R)}{\pi x / \Delta R} dx \\ &= \frac{A}{2\pi} \int_{-\pi}^{\pi} \frac{\sin y}{y} dy \quad (4.10) \\ &= 0.590A \text{ or } -4.59 \text{ dB}\end{aligned}$$

Considering both Equation (4.8) and Equation (4.10), it can be deduced that there is a 4.6 dB reduction in coherent processing gain (CPG) at maximum scanning angle due to range migration (for a FMCW radar based operating with linear chirp waveform). The decrease in range resolution is  $\sqrt{2}$ , which can be determined by a convolution process. This deterioration in resolution results in smearing of the point target image after processing. It should be said that the acceptable limit for range resolution reduction in CPG is dependent on application, but the difference is minimal. For instance, by using modified integration limits in Equation (4.10), range migration threshold factor in terms of resolution can be determined as 1.6 times instead of 2. This is calculated by taking 3 dB decrease in CPG. This is done by using 3 instead of 4 in Equation (4.8).

#### 4.2.2. Mitigation of range migration in phased array radar

During received signal processing in phased array radar, true time delay processing can be used to remove of range migration. It is also commonly known that usual phase shifting for beamforming is not applicable in wideband operations nor in post-deramped FMCW. In [90] [91], it is described how beamforming is achieved in wideband systems by correcting the time delay within array elements. This advantage is not present for an FMCW/step frequency radar system. This is due to the fact that

the received signals in these systems are mixed with replica chirp to obtain the deramped signal. As it is known that deramped signal corresponds to frequency shift, time delay compensation for raw digitized received data cannot be achieved. Hence, time delay processing is not applicable. Instead, the range migration correction needs to be implemented by using the frequency shift.

There are number of papers that deal with the range migration correction. In [92], range migration correction is performed in frequency-wavenumber domain for 2-D wideband MIMO array where in [93], it is done for FMCW SAR system. In [94], range migration correction is also performed in frequency-wavenumber domain for ground penetrating radar system. In this thesis work, the range migration correction process has been developed for an FMCW/deramped phased array radar system that has been built for Antarctic Ice Shelf imaging. The radar system has quite large aperture due to low operating frequency and a wide fractional bandwidth.

For a phased array radar system that exceeds the criterion given in Equation (4.8) and therefore suffers from significant range migration, the obvious remedy is to re-align the returns from the elements in order to compensate for their displacement due to range migration, prior to cross-range processing. This may be achieved by time delay equalization in many cases, with the necessary delays easily calculated from basic geometry. From inspection of Figure (4.2), the necessary time delay equalization can be obtained from Equation (3.12). In the case of an FMCW radar with full-deramped processing, which is the basis of the experimental system in this thesis work, the frequencies of the deramped signals will be shifted by range migration and so it will instead be necessary to re-align the signals in the frequency domain. This may be affected by frequency-translating the signals from the  $n$ th element by the following amount:



$$\Delta f = \frac{B}{T} \Delta \tau = \frac{nBd \sin \theta}{cT} \quad (4.11)$$

The FFT resolution and range bin spacing is  $1/T$  (or  $1/T_p$  if zero-padding is used to increase the sample length by a factor of  $p$ ) and so it may be convenient to translate the signals at each element by appropriate increments of  $1/T_p$ , which is effectively nearest-neighbour interpolation. A pad factor,  $p$ , of two or more has been found to yield good results.

In addition to re-aligning the range profiles derived from elements across the array, attention must be paid to the relative phases of the signals that have been re-aligned, which is considered as follows. From Equation (3.4), deramped phase term, which is compensated by normal phased array processing is:

$$\psi = \omega_c \tau - \frac{k\tau^2}{2} \quad (4.12)$$

Equation (3.4) shows that the stationary phase point of the deramped FMCW radar signal is in the centre of the pulse, at  $t = T/2$ , at which point the signal phase is independent of the sweep parameters. An FFT function would normally indicate the phase at the beginning of the signal,  $t = 0$ , which depends strongly on the sweep parameters and will vary if the signal is frequency-shifted to achieve re-alignment in order to mitigate range migration. To consider the degree of phase change as a result of this re-alignment process, from Equation (3.4) the deramped signal from a given point target may be represented as a sinusoid within but not necessarily in the centre of a frequency bin, of frequency  $n\omega_o + \Delta\omega$ , where  $\omega_o$  is the FFT frequency resolution, which is equal to  $2\pi/T$ , with an instantaneous phase as follows,

$$\theta_d(t) = (n\omega_o + \Delta\omega)(t - T/2) + \psi \quad (4.13)$$

Where  $\omega_o = \frac{2\pi}{T}$  and  $|\Delta\omega| < \pi/T$

The FFT (or DFT) term relating to the  $n$ th range bin is given by

$$F_n(n\omega_o) = \int_0^T e^{-jn\omega_o t} \cdot e^{j(n\omega_o + \Delta\omega)(t-T/2)} \cdot e^{j\psi} dt \quad (4.14)$$

which reduces to the following expression,

$$F_n(n\omega_o) = T e^{j\psi} \cdot e^{-jn\pi} \cdot \text{sinc}(\Delta\omega T/2) \quad (4.15)$$

The sinc function factor in this expression is always positive, for  $|\Delta\omega| < 2\pi/T$  (a condition that is easily satisfied), and so this result shows that the phases of the FFT components obtained from sinusoids in adjacent range bins will alternate. Thus it is crucial that this phase change is corrected when the signals are translated to effect range migration mitigation. If zero-padding is used to sinc-interpolate the range profile by increasing the length of the time-domain sample by a factor of  $p$ , then the FFT resolution will be increased by a factor of  $p$  and the phase change between adjacent range bins will be correspondingly reduced, giving a more general result for the phase indicated by the  $n$ th range bin:

$$\angle F_n(n\omega_o) = \psi - \frac{n\pi}{p} \quad (4.15)$$

Thus, in addition to re-alignment of the signals to mitigate range migration it is also necessary to correct the phases, in the case of an FMCW radar with full-deramped processing. As an alternative, from Equation (3.4) it is clear that rotating the time domain signal by  $-T/2$  prior to taking the FFT will allow the phase reference to be moved to the centre of the pulse and yield a similar effect. This may be achieved by applying the standard Matlab *FFTshift* function.

### 4.2.3 Simulation results

Simulation was performed to validate the range migration correction. Simulation parameters were used according to Table 1. It should be noted that the FMCW radar has a high fractional bandwidth (66.7%). Using Equation (4.8), range migration processing will be required if the array contains more than 7 elements (which would

always be the case during Antarctic experiments). At first, the simulation was performed for a single point target to demonstrate the effect of range migration processing. Then, contiguous point targets parallel to the base line were simulated and imaging results have been shown here. During simulation, 64 elements were used and target range was 1500 metres.

Figure (4.4) shows the processed point target without applying range migration correction. It can be seen the image is smeared out (defocused). Figure (4.5) shows the immediate effect of the correction process as the image is now far better focused. It should be noted that if the point target is located at normal of the array, then image will be focused in the first place as no range migration will occur. Figure (4.6) shows the processed image of point targets with same range but instead of one, number of closely spaced point targets throughout the scanning angle has been imaged to simulate a basal layer return from ice shelf. It can be seen that as the scanning angle increases, the image becomes more defocused. Figure (4.7) shows the same image after range migration processing but without correcting the phase. It shows even though signal energies have converged, the overall image still has not been focused properly due to phases not being properly modified. Figure (4.8) shows the processed image after fully applying the correction process. Now, the target return looks like a straight line parallel to array, exactly the way target locations were defined during simulation.

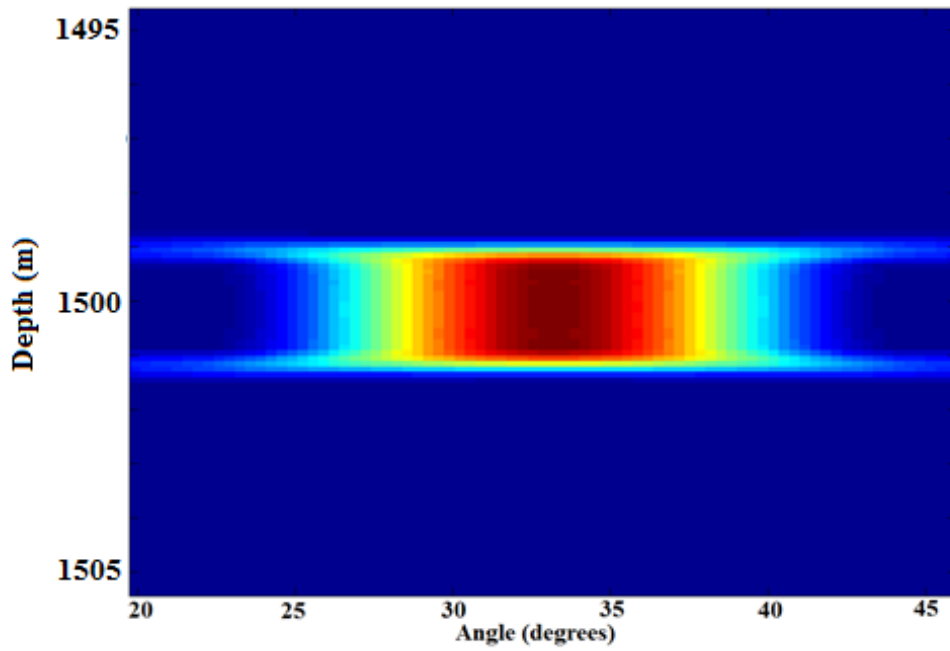


Figure 4.4: Point target simulation of a 64 element phased array radar operating with 66.7% fractional bandwidth without implementing range migration correction (target has a  $33^{\circ}$  angle array normal).

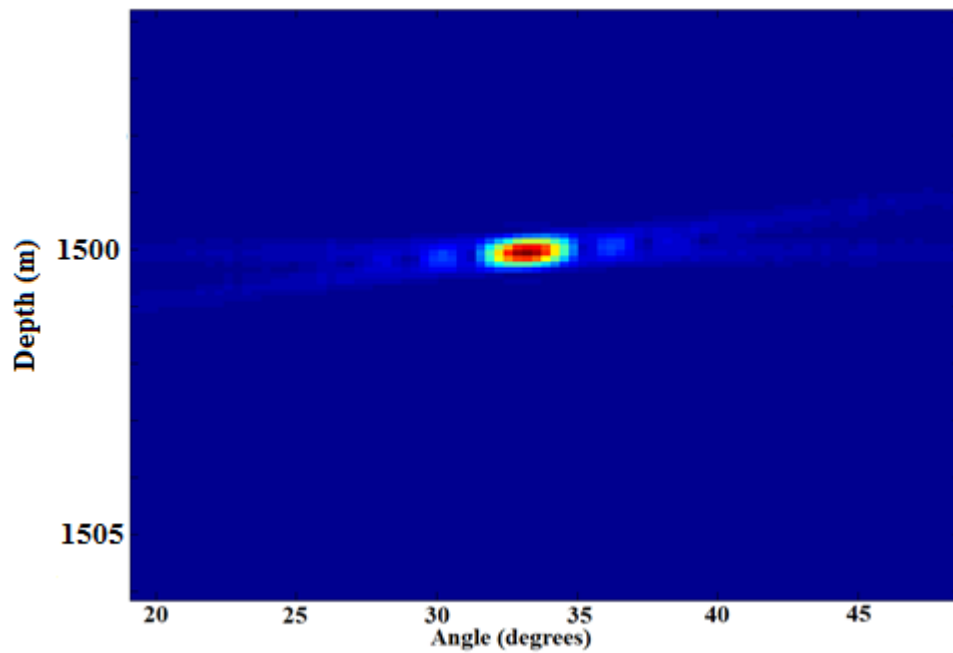


Figure 4.5: Point target simulation of a 64 element phased array radar operating with 66.7% fractional bandwidth after implementing range migration correction (target has a  $33^{\circ}$  angle from array normal).

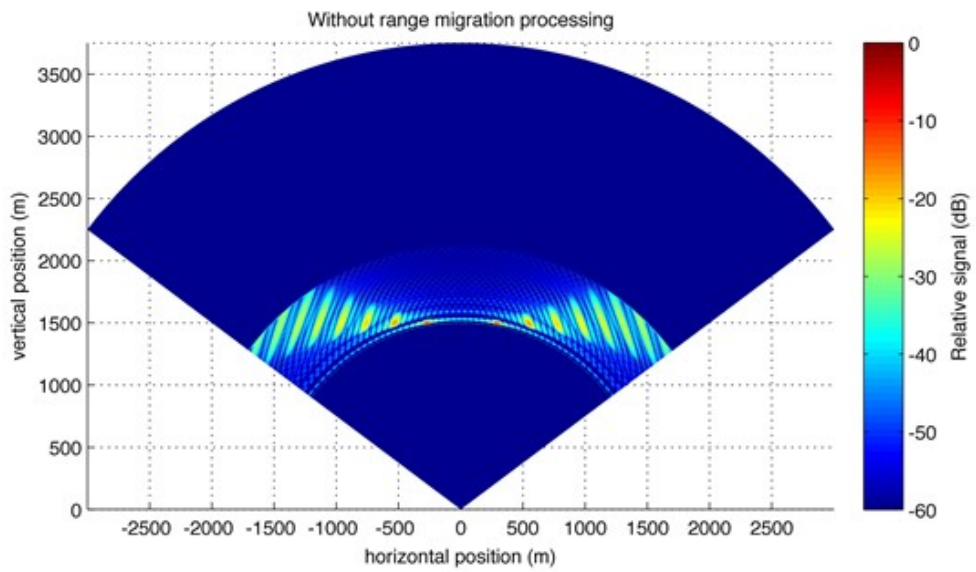


Figure 4.6: Phased array radar image of a target scene comprising a reflecting layer parallel to array at 1500 m range, without range migration processing.

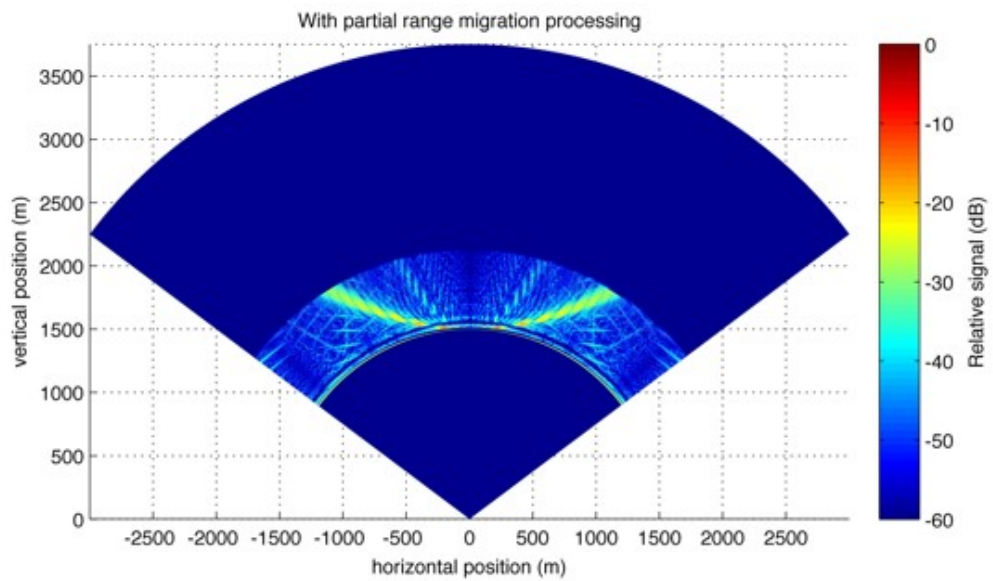


Figure 4.7: Phased array radar image of a target scene comprising a reflecting layer parallel to array at 1500 m range, after range migration processing but without phase correction.

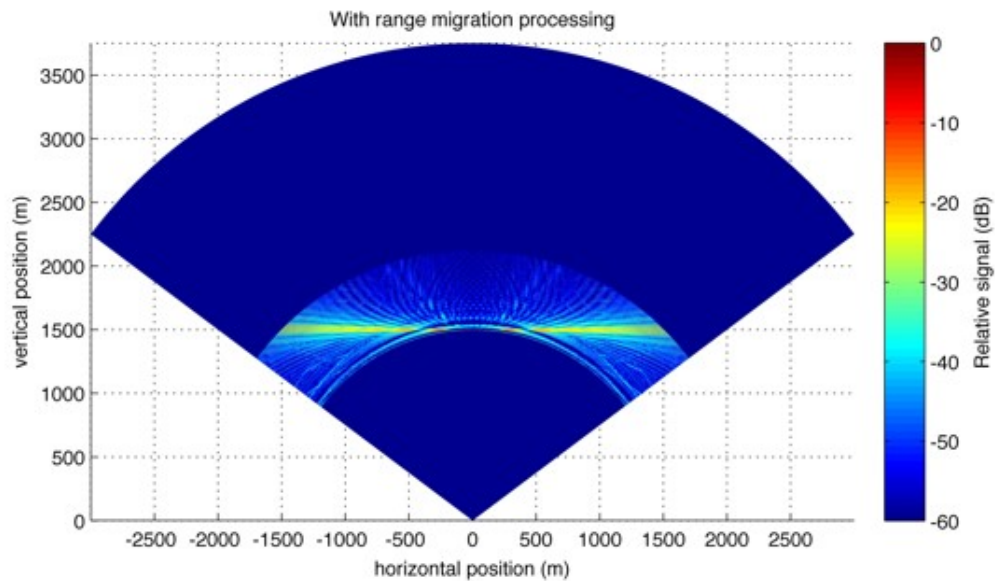


Figure 4.8: Phased array radar image of a target scene comprising a reflecting layer parallel to array at 1500 m range, after range migration processing.

### 4.3 Vertical Error Correction by Phase Calibration

As mentioned in the previous section, the vertical error in the Antarctic data signal processing occurs during experiments. An internal layer is hence very useful to resolve this but there is no guarantee of an internal layer being present all the time. In that case, some other means is required to correct the error to get a properly focused image. Radar motion compensation techniques have been well developed in the field of SAR [95] [96], or in case where Doppler shift is present due to moving target [97]. In this particular scenario of Antarctic Ice Shelf study, everything (radar and the target) is assumed to be stationary. So, an appropriate method needed to be developed which would enable minimization of the effect of vertical antenna displacement. This correction will also compensate for phase error occurred due to near-field effect, as it is an exact geometric correction. The main problem is antenna displacement errors cannot be determined in post processing if there is no reference present. But what can be done is the vertical displacement can be accommodated into the phased

array/MIMO beamforming equation. The previous figure shows the geometry of the vertical error and simple trigonometry to quantify it. If the vertical displacement is  $z$ , then by recalling Figure (3.5), extra distance travelled by the signal can be calculated as it is shown in Figure (4.9). It can be seen that the offset distance is  $z\cos\theta$ . The total extra distance corresponding to phase delay then becomes  $nd\sin\theta+z\cos\theta$ . So, by modifying Equation (3.11), the new beamforming equation becomes:

$$B(\theta) = \sum_{n=0}^{N-1} w_n e^{\frac{-j2\pi(nd\sin\theta+z\cos\theta)}{\lambda}} \quad (4.16)$$

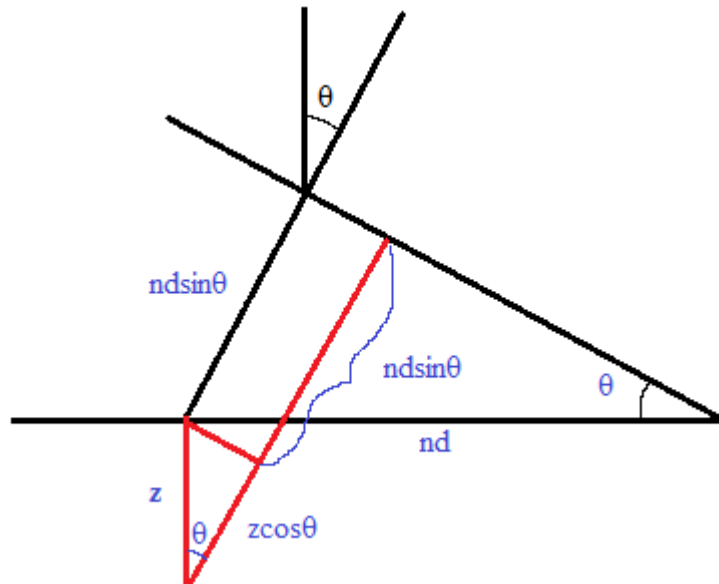


Figure 4.9: Geometry of vertical displacement of element from the array baseline.

The above figure illustrates the effect of vertical errors. Here, the same simulation was performed that obtained Figure (4.8), only difference being vertical spacing of the array elements has been imposed. 36 array elements are used here as that many elements were used during one of the experiments in Antarctica. The resultant image is totally defocused as during beamforming,  $z\cos\theta$  has not been taken into account. So, the signal phases lost coherence.

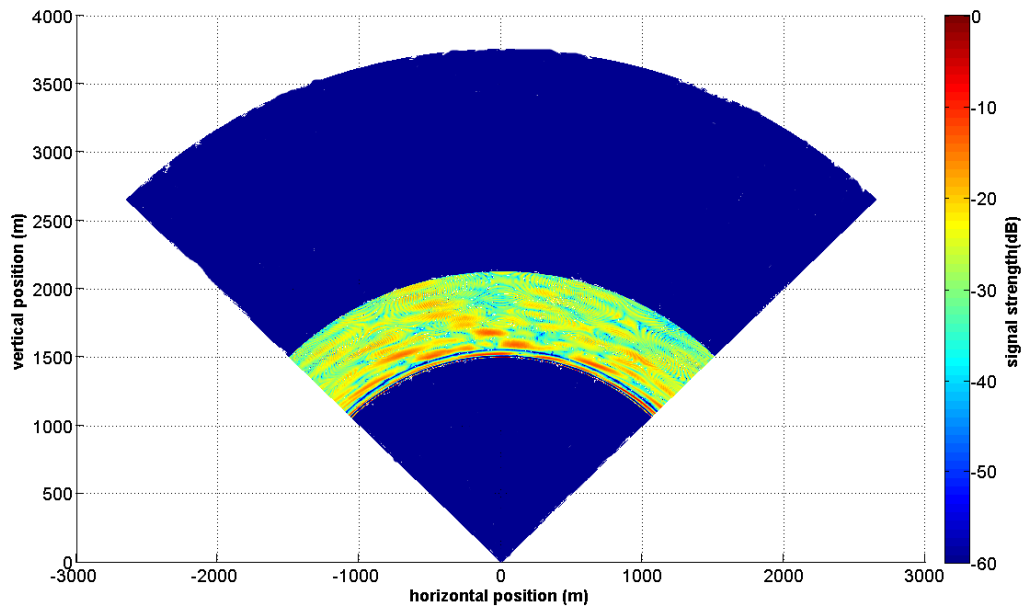


Figure 4.10: Phased array processed image of a reflecting layer at 1500 metres range without accounting for vertical spacing error of antennas.

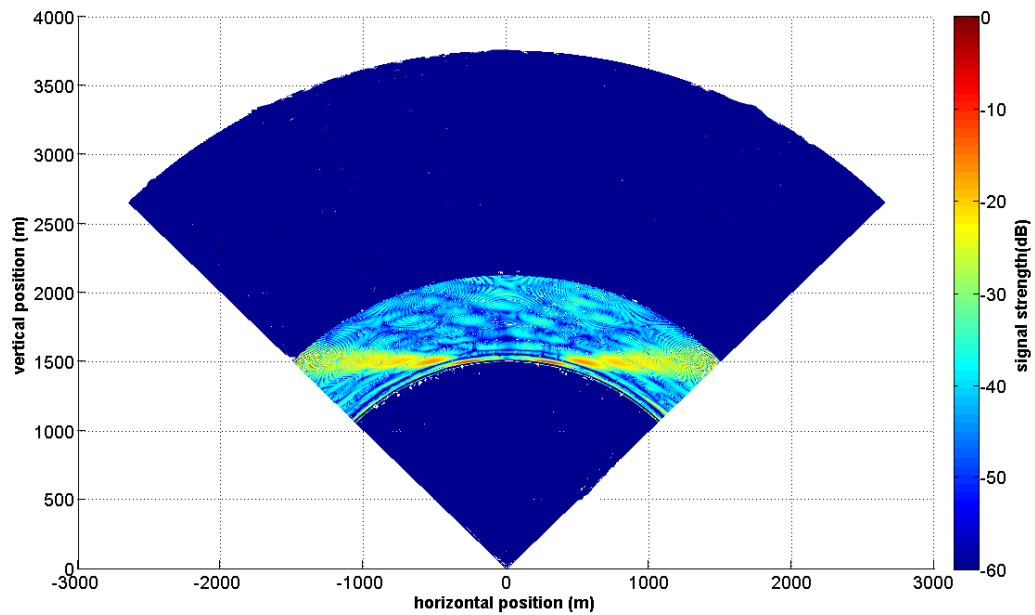


Figure 4.11: Phased array processed image of a reflecting layer at 1500 metres range where vertical spacing error of antennas has been taken into account.

Figure (4.11) shows that proper image is achieved when Equation (4.16) is used for creating steering vector, hence compensating for the vertical offset.



In practical cases, the value of  $z$  will be unknown. The proposed method for obtaining the value of  $z$  is to use the phase reference at zero degree angle. During post processing, the range bin corresponding to zero degree angle return for all the array elements can be determined by careful observation

Then, if phase of that range bin across the array is plotted, it is expected to give a flat phase response. This property can be used for estimating  $z$ .

For instance, Figure (4.12) shows the range profile of one of the 36 array elements. Here, the spike in range bin number 4001 (corresponding to 1500 metres range) is the return from the normal to the array. The trail that follows the spike (up to around range bin number 5700) corresponds to returns from other angles. Now, if the phase of the range bin number 4001 for the entire array elements (36 in this case) is plotted, more or less a straight line is expected as they all have almost same value (a slight difference will occur due to interference with contiguous signals from other angles).

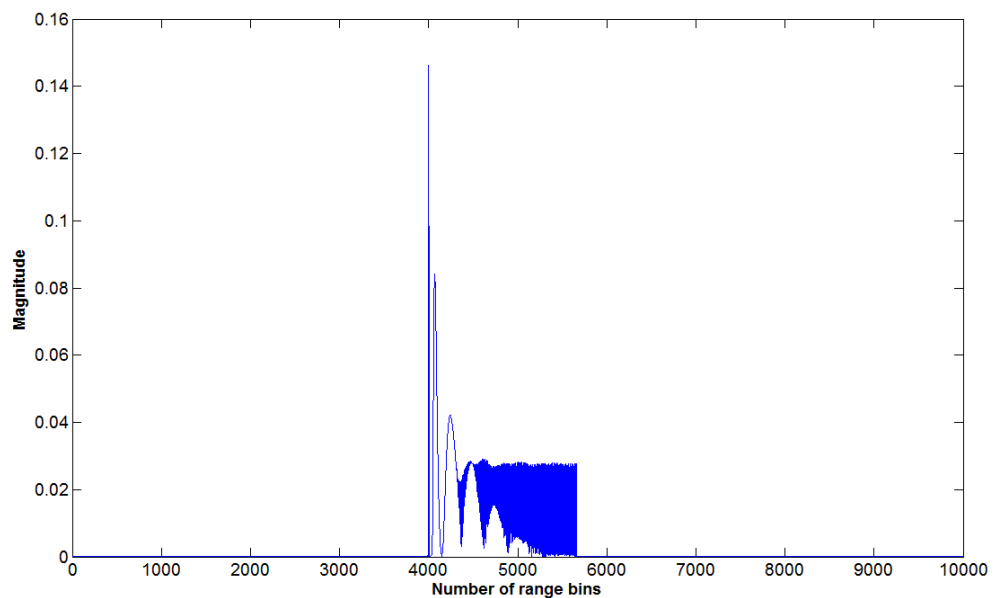


Figure 4.12: Range profile of one of the array elements showing the zero degree signal return at range bin number 4001.

It can be seen that phase is quite constant throughout (varying from only 1.95 to 1.65 radians). If there is any vertical displacement at the antenna elements, this flat phase response will not be seen. So, observing the phase response for this range bin somewhat gives an idea about the error.

Figure (4.14) demonstrates exactly that where due to antenna displacement, flat phase response is lost. The idea for determining  $z$  is using this arbitrary phase response and making it flat (assuming vertical displacement is the only/main cause for the phase offsets.). During simulation, the first value is taken as reference.

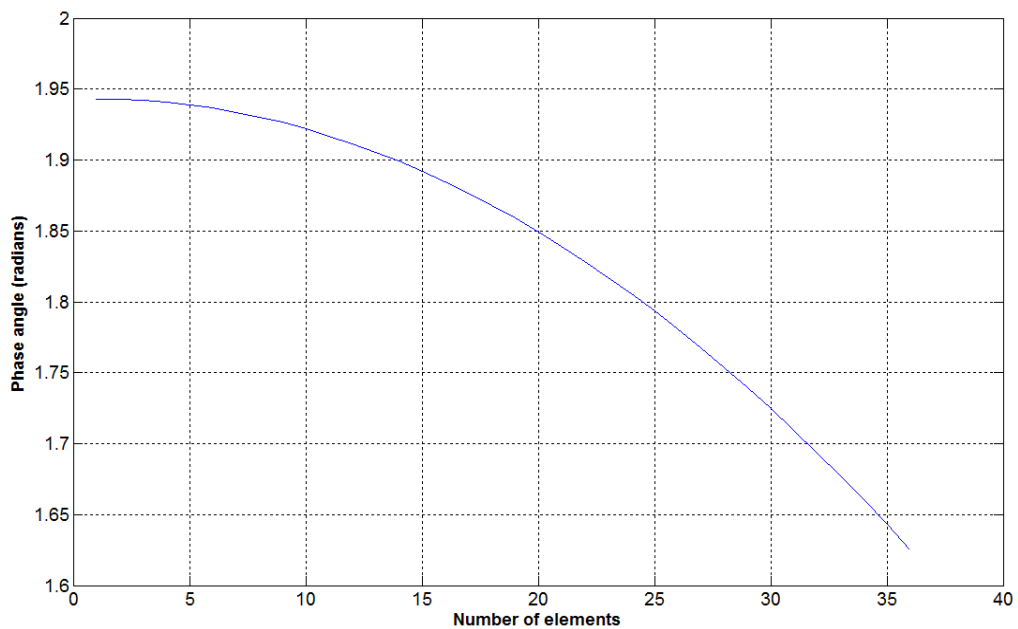


Figure 4.13: Plot of phase response for the zero degree return across the entire array.

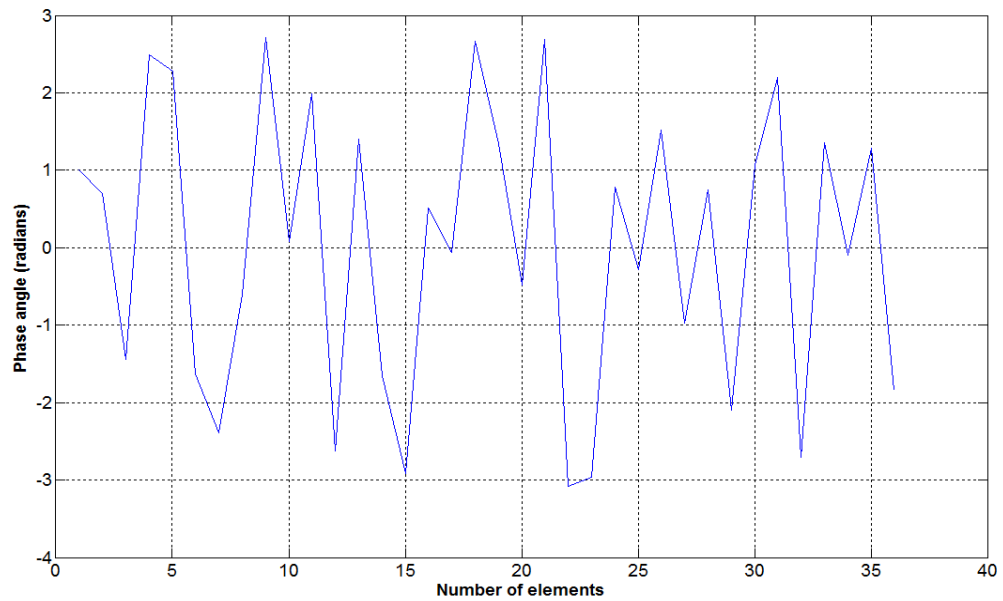


Figure 4.14: Plot of phase response for the zero degree return across the entire array where vertical displacement of the elements have been imposed.

Then the phase offset for other elements from that reference is calculated. That phase value is then converted into distance offset (dividing the phase by  $2\pi/\lambda$ ). These obtained distance offsets (0 for the first element as that is the reference) are used as the value for  $z$  at each element. It also implies that if the spacing is larger than the wavelength, this calibration method will not work. This obtained matrix of  $z$  is then used in Equation (4.16) for beamforming. Figure (4.15) shows the image obtained by using this method for compensating the vertical error. As can be seen, the result is quite promising as there is obvious improvement from Figure (4.10) where the image is very well focused and almost same as Figure (4.11). One disadvantage here is that target signal is not higher than the side lobes at optimum expected level. Here, the target signal strength is around 15 dB higher than the sidelobes.

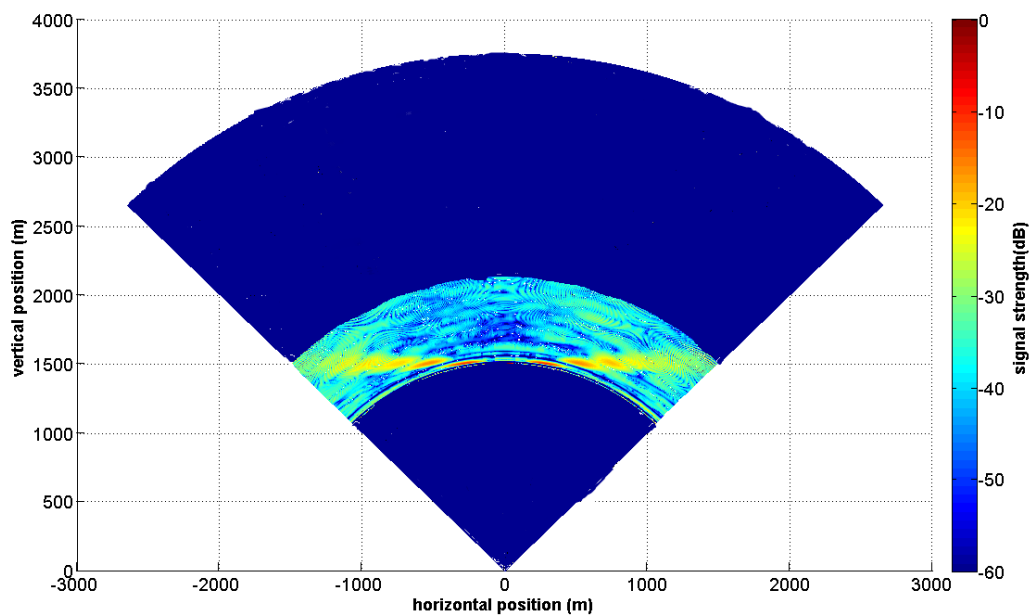


Figure 4.15: Phased array processed image of a reflecting layer at 1500 metres range where vertical spacing error of antennas has been compensated by phase calibration with reference to zero degree return.

#### 4.4 3-D MIMO Imaging Algorithm for Ice Shelf Basal Layer

The last signal processing algorithm developed during this thesis work was to create a 3-D image of basal layer of the Antarctic Ice Shelves. As the previous signal processing with linear phased array radar gave sectoral plots, 3-D image of the base would give more opportunity for more in depth analysis of the ice shelves. For this purpose, an 8x8 MIMO radar with 2-D array has been built at UCL, which have been deployed in Antarctica. It should be noted that no orthogonal waveform for MIMO transmission is used in this system (all transmitting antennas emit same waveform). The spatial diversity is obtained by time switching the antenna elements. New processing method needed to be developed to create proper 3-D images from the radar data.

There are research articles studying the 3-D MIMO beamforming technique. In [98], a beamforming equation for 2-D MIMO array has been illustrated. In [99] [100], 3-D imaging techniques have been described by combining MIMO and Synthetic Aperture Radar (SAR) geometry. In [101], imaging techniques for FMCW MIMO system have been shown, applicable to conventional two-dimensional imaging. In [102], a 3-D imaging technique has been described for co-located MIMO radar configuration. The paper proposes a 3-D imaging technique by taking one snapshot of a moving target, hence evading the need for motion compensation. But as the focus of the paper is to image a moving target, beamforming vector corresponds to only one direction (as opposed to both in azimuth and elevation). For the purpose of Ice Shelf Imaging, where the signal processing is done solely in terms of receive array, two dimensional steering vector is required to map the ice shelf base. So, it was necessary to create an algorithm for the entire 3-D MIMO image processing (based on FMCW radar receive array and focused on imaging base of ice shelf/sheet), which will include all the required operations. The steps for 3-D MIMO imaging algorithm developed for Antarctic Ice Shelf basal layer imaging are as follows:

- Range profile for all virtual elements.
- Applying 2-D window over the planar array.
- 3-D Beamforming (by compensating for phase offset at each elements according to azimuth and elevation angles).
- Range Migration compensation (if required).
- Selecting the range bin corresponding to zero degree return (using that range slice throughout the scan for 3-D basal layer imaging).
- Compensating for the curvature of the range slice.

- Image sidelobe suppression by 2-D correlation

The first two steps are quite straightforward, which basically deal with preparing the raw data before beginning the imaging process. As the algorithm is developed for FMCW based radar system, range profile should be obtained by simple FFT of the deramped signals (after applying range window to minimize range sidelobes). Also, in order to suppress cross range sidelobes, windowing over the arrays should be performed which is quite usual in phased array/MIMO processing. As the array is planar, the windowing operation needs to be performed on both axes. For instance, for an 8x8 array, windowing has to be applied on eight elements of each row as well as on each column.

#### 4.4.1 3-D Beamforming

The main objective of a beamforming operation is to find out the signal path offset in each element of the array, for every angle. It is determined with respect to a reference where the offset is zero. 3-D beamforming process for the imaging here is the same as described in Chapter 3, only with an added dimension, hence an added angle (instead of one incident angle, in this case both azimuth and elevation angles will be taken into account). So, even though the geometry requires a bit more visualization, the basic idea here for beamforming is the same as it is for a linear array.

Figure (4.16) illustrates a geometrical representation of the 3-D beamforming process. The red line represents the offset signal distance incident upon X and Y plane (corresponding to the X and Y axes of the planar array). ' $\theta$ ' and ' $\phi$ ' in the above figure define azimuth and elevation angles respectively. So, the beamforming matrix will contain a total number of elements equal to the product of azimuth and elevation angle sample numbers.

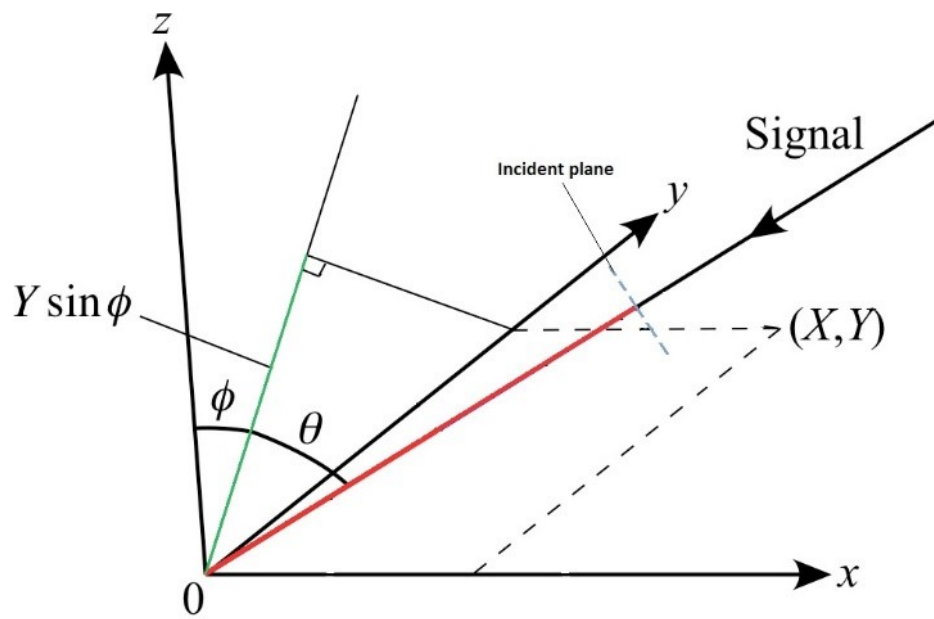


Figure 4.16: 3-D MIMO beamforming geometry for planar array.

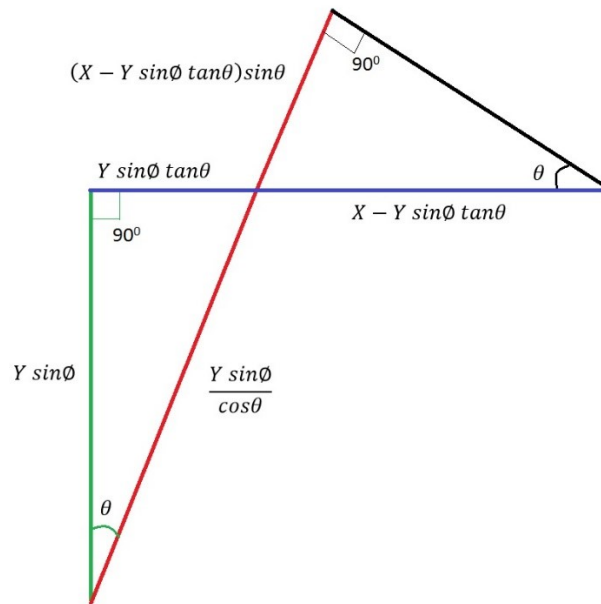


Figure 4.17: Two dimensional representation of the 3-D MIMO beamforming geometry for the purpose of Pythagorean calculation.

In order to determine the value of the signal offset, it will be convenient to draw a two dimensional representative diagram of the area of interest from Figure (4.16). As it can be seen in Figure (4.17), the red line corresponds to the same offset signal path from Figure (4.16). Similarly, the green lines in both figures correspond to same  $Y \sin \phi$ . The green line is simply the path offset at Y axis (where azimuth angle is 0 degree). Now, the signal direction would tilt by  $\theta$  degree when azimuth angle is present. This can be visualized from Figure (4.17). The horizontal blue line then would correspond to X distance. By using Pythagorean geometry, the values of lines can be defined which would lead to the calculation of the eventual distance required for beamforming (the red line). Those values are also shown in Figure (4.17). The calculation for beamforming distance then becomes as follows:

$$\begin{aligned}
 B_d(\theta, \phi) &= (X - Y \sin \phi \tan \theta) \sin \theta + \frac{Y \sin \phi}{\cos \theta} \\
 &= X \sin \theta + Y \frac{\sin \phi}{\cos \theta} (1 - \sin^2 \theta) \\
 &= X \sin \theta + Y \sin \phi \cos \theta
 \end{aligned} \tag{4.17}$$

This can be now easily translated into equation for phase compensation during beamforming, which is as follows,

$$B(\theta, \phi) = \frac{4\pi}{\lambda} (X \sin \theta + Y \sin \phi \cos \theta) \tag{4.18}$$

$4\pi/\lambda$  has been used as the distance to phase conversion factor instead of  $2\pi/\lambda$  here, due to two way wave propagation (the beamforming is only performed in receive array).

#### 4.4.2 Range Migration Compensation

The effect of range migration has been discussed earlier in this chapter. Equation (4.8) gives a proper indication whether a receive array processing would require the



migration correction. In order to implement the correction process for planar 2-D array, array length actually gets reduced. For instance, a 64 element linear array would have 8 times more length than an 8x8 planar array (considering same element spacing in both cases), even though in both cases there are there are 64 array elements. So, in later case, the condition check from equation (4.8) should be performed for 8 elements, instead of 64.

#### **4.4.3 Selecting the range bin corresponding to zero degree return**

In 3-D MIMO imaging, the obtained result would be a three dimensional form of the usual sectoral image achieved in 2-D phased array/MIMO processing. Imaging the whole 3-D image in such a way would be inconvenient for visual representation. So, instead of plotting the image for the entire range limit, single range slice is chosen. A 3-D image of that specific slice is then plotted (in terms of azimuth axis, elevation axis and signal strength at each point). In case of Antarctic Ice Shelf study, the range slice is chosen by determining the distance of basal layer from the surface. As the surface may not be plane (hence the reason for 3-D imaging), a reference point is chosen. Simplest way to choose the reference range bin is finding the range return from zero degree angle. During post processing, it is quite easy to find the range bin corresponding to zero degree return, as the signal strength would be usually quite higher. This is due to signal attenuation with distance increased and zero degree return being the closest distance. The corresponding range bin is found the same way it has been described in section 4.3.

#### **4.4.4 Compensating for the curvature of the range slice**

After the selection of the zero degree return range bin, range bins for other angles corresponding to the same range from the radar need to be identified. The range

changes with angle according to the Pythagorean geometry. Before plotting the image, this curvature of the surface need to be corrected.

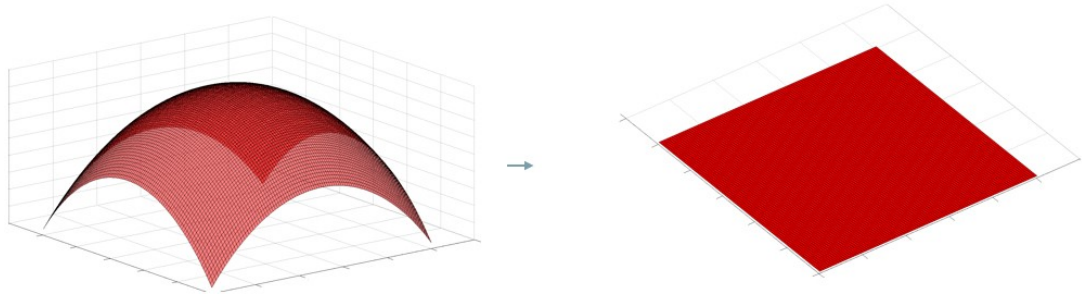


Figure 4.18: Surface curvature correction for 3-D MIMO imaging of one range slice.

The process of curvature correction can be visualized from Figure (4.18). It is evident from the figure that for each point of the image, the correction should be performed by using the Pythagorean cosine formula with respect to the reference range (the range obtained from zero degree return). The equation for determining the range bin number can be written as,

$$rangebin(n, m) = \frac{R_o}{\cos \theta_n \cos \phi_m \Delta R} \quad (4.19)$$

In the above equation,  $R_o$  is the reference range,  $\Delta R$  is the range resolution,  $\theta$  and  $\phi$  are azimuth and elevation angle respectively. Considering  $N$  and  $M$  being the total azimuth and elevation sample numbers respectively,  $n=1,2,3,\dots,N$  and  $m=1,2,3,\dots,M$ . In practice, the division may provide non integer values, which case it is rounded to the nearest integer.

#### 4.4.5 Image sidelobe suppression by 2-D correlation

A major problem faced during 3-D image processing is the sidelobes. In a conventional 2-D image, cross range sidelobes are common phenomena. In a 3-D image where the scanning is in two directions, sidelobes occur due to both. So, the final image quality becomes visually very poor as the sidelobes appear as circular ring

patterns (as will be shown in simulation results). To minimize that, 2-D correlation is performed over the image. It is done by taking the maximum value as reference in each row and column and performing correlation operation on each row and column with those references.

2	1	3	1
1			
3			
1			

Figure 4.19: Sample demonstration of 2-D correlation method used for sidelobe suppression in 3-D MIMO image

Figure (4.19) can be used to visualize the concept being used during reducing the circular rings from the image. The table corresponds to the image plane where each block represents a specific azimuth and elevation angle combination. The arbitrary numbers in the figure represent signal strength corresponding to that block. The blue box encases the first row in which 3 is the maximum value. That will be taken as a reference and the correlation would be performed over the entire row (within the blue box) with respect to the reference. Same thing will be repeated for all the rows. After that, this correlation process is applied on all the columns using the same method. In practice, instead of taking one value as reference, several values on both sides of the maximum value are also included to make a reference for correlation consisting of more than one sample. Also, due to correlation process, number of samples in each row and column will increase. To have same number of values in all the rows and all the columns for processing purpose, interpolation is used to scale that. In real case, the

surplus samples are very few in numbers compared to total samples in a row or column (as the reference signal size is comparatively very small). So, the change in image due to interpolation is negligible.

#### **4.4.6 Simulation results**

In order to verify the image processing method, a simulation has been performed in MATLAB. The parameters during simulation have been chosen according to the specification of the 8x8 FMCW based MIMO radar built at UCL. The radar operates at 300 MHz central frequency with 200 MHz bandwidth. The virtual array spacing is  $0.64\lambda$  in both axes. For the basic visual understanding of the images, the simulation has been performed for point targets. Also, from Equation (4.8), it is found that range migration is not going to have much of an effect on an 8x8 array. So, during simulation longer array length has been chosen just to validate the need of range migration correction for 3-D MIMO image processing algorithm.

Figure (4.20) shows the point target image at 1500 metre range and -8 degree angle from the array normal (both in azimuth and elevation). The scanning is performed for  $\pm 25^\circ$  in both directions. It can be seen that the point target image has wide spreading due to range migration. In Figure (4.21), the same point target can be seen after applying the range migration correction.

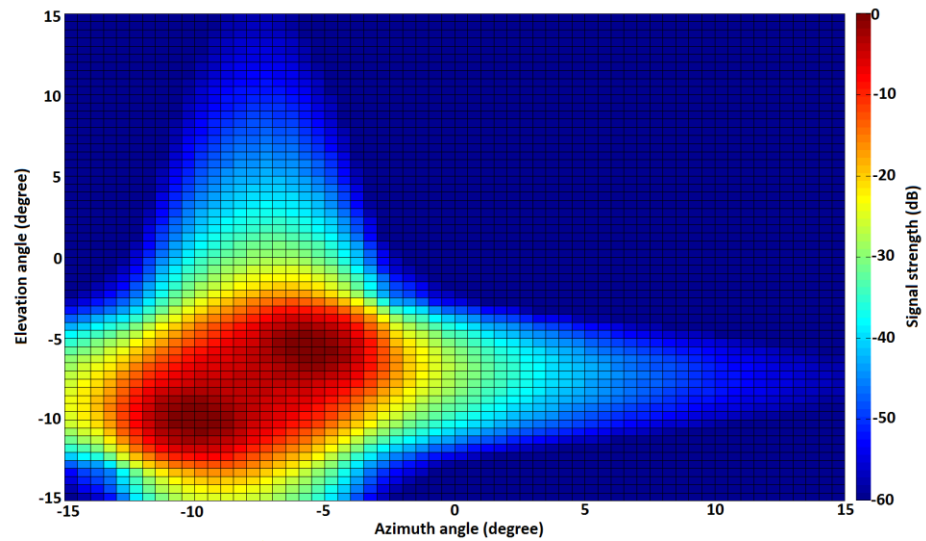


Figure 4.20: Simulated result of a defocused 3-D MIMO image for a point target (single range slice at 1500 m) with  $\pm 25^\circ$  scan angle in azimuth and elevation; for large array length and without applying range migration correction.

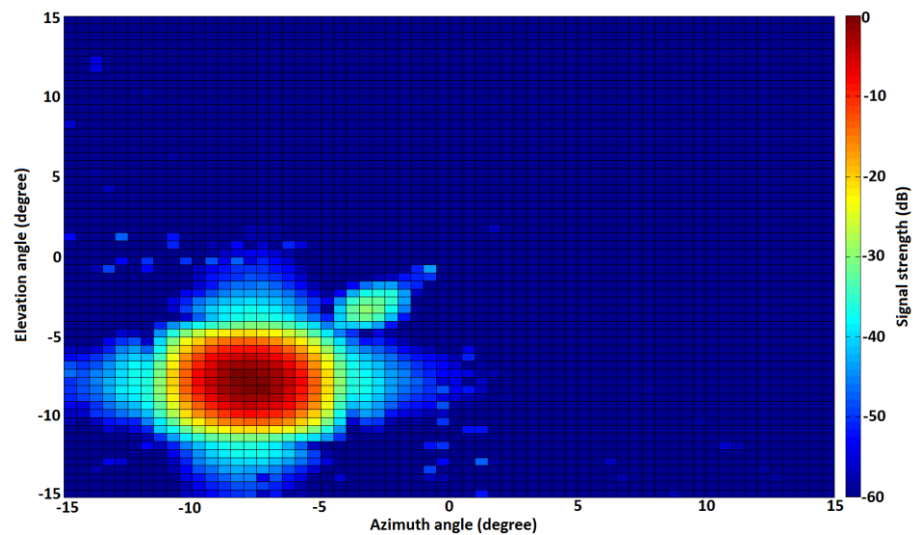


Figure 4.21: Simulated result of a focused 3-D MIMO image for a point target (single range slice at 1500 m) with  $\pm 25^\circ$  scan angle in azimuth and elevation; for large array length and after applying range migration correction.

The signals now get more concentrated, hence the point target image gets more focused. This justifies the use of range migration correction process for 3-D MIMO imaging as well if the planar array length exceeds the condition in Equation (4.8).

Simulation has also been performed to demonstrate the circular patterns in image and effect of 2-D correlation to remove it. This has been done by simulating a basal layer at 1500 metre range (by simulating very closely spaced targets everywhere within  $\pm 15^\circ$  area in azimuth and elevation). The image scanning coverage area is  $\pm 20^\circ$  in both directions. Figure (4.22) shows the simulated result where the circular pattern is easily visible. Figure (4.23) shows the resultant image after applying the 2-D correlation. It can be seen that the ring pattern is removed, giving a smooth 3-D image of the 1500 metre range slice. In the next chapter, the resultant images obtained from processing the raw Antarctic MIMO radar data are shown and discussed.

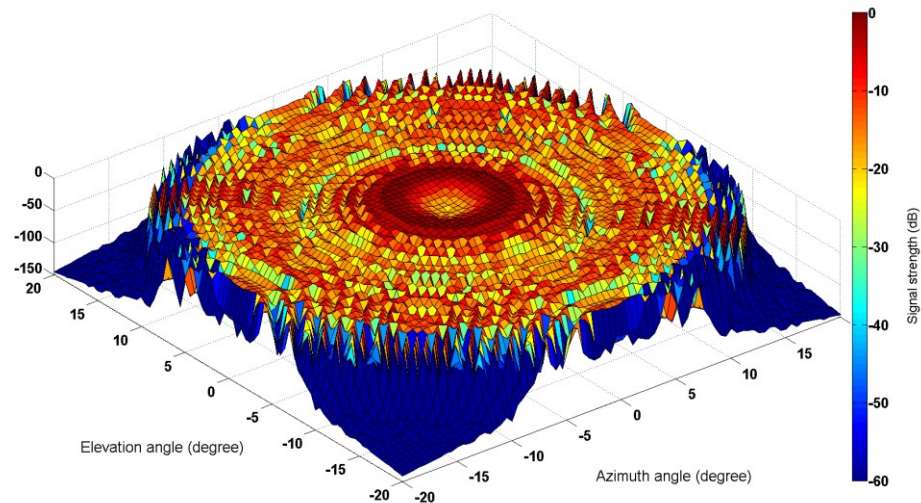


Figure 4.22: Simulated result of 3-D MIMO image of very closely spaced targets covering  $\pm 15^\circ$  area in both direction without sidelobe suppression hence showing the ring pattern (scan angle in both direction is  $\pm 20^\circ$ ).

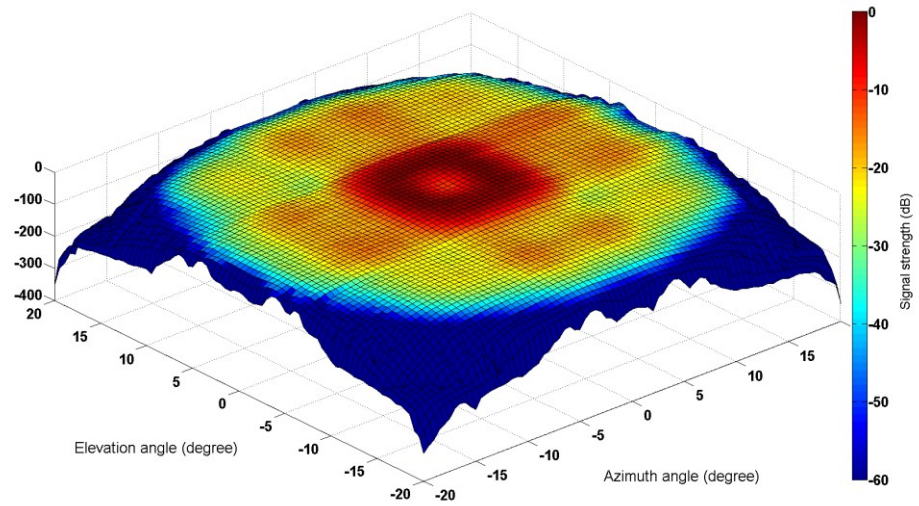


Figure 4.23: Simulated result of 3-D MIMO image of very closely spaced targets covering  $\pm 15^\circ$  area in both direction after applying hence removing the ring pattern (scan angle in both direction is  $\pm 20^\circ$ ).

## Chapter 5

# Processed Results from Antarctic Ice Shelf Experiments

This chapter demonstrates all the results obtained from processing Antarctic data. The signal processing techniques that have been discussed in Chapter 3 and 4, have been applied for real data processing. The data was obtained during field expedition organized by BAS in Austral summer of 2011-12. During this expedition, BAS deployed both their pRES system as well as the phase sensitive FMCW radar built at UCL. The FMCW radar was deployed in Larsen–C, Larsen South, Larsen North and George VI ice shelves. The pRES system made measurements in Foundation ice stream of Ronne Ice Shelf. Along with range profiling of the basal layer of these ice shelves, various imaging experiments were also carried out by constructing the array geometry. Generally, the radar was mounted on a sledge and then moved around carefully according to the specific geometry. Also, the radar performance was tested at UCL to validate radar performance, which is shown here. Finally, the 3-D imaging results from the Ronne Ice Shelf are also shown. The FMCW based 8x8 MIMO radar was deployed in the location in 2014-15 Austral summer. The radar was also taken to Greenland by BAS. Raw data from there has also been processed and the resultant images have been shown here.

### 5.1 Loop Test results at UCL

A trial had been conducted at UCL to verify the radar performance with a 240 meter cable (connecting the two endpoints to the transmitter and receiver). The



dielectric constant of the cable is 2.1. Phase sensitive FMCW radar signal processing algorithm [45] was implemented in MATLAB during the thesis to analyze the raw radar data. The codes were written in such a way so it would be able to segment the signals from different attenuator settings automatically and process the whole data file at one go. Figure (5.1) shows the received deramped signals for four different attenuator settings of the radar (which is to get optimum dynamic range while acquiring field data). Figure (5.2) shows the range plots.

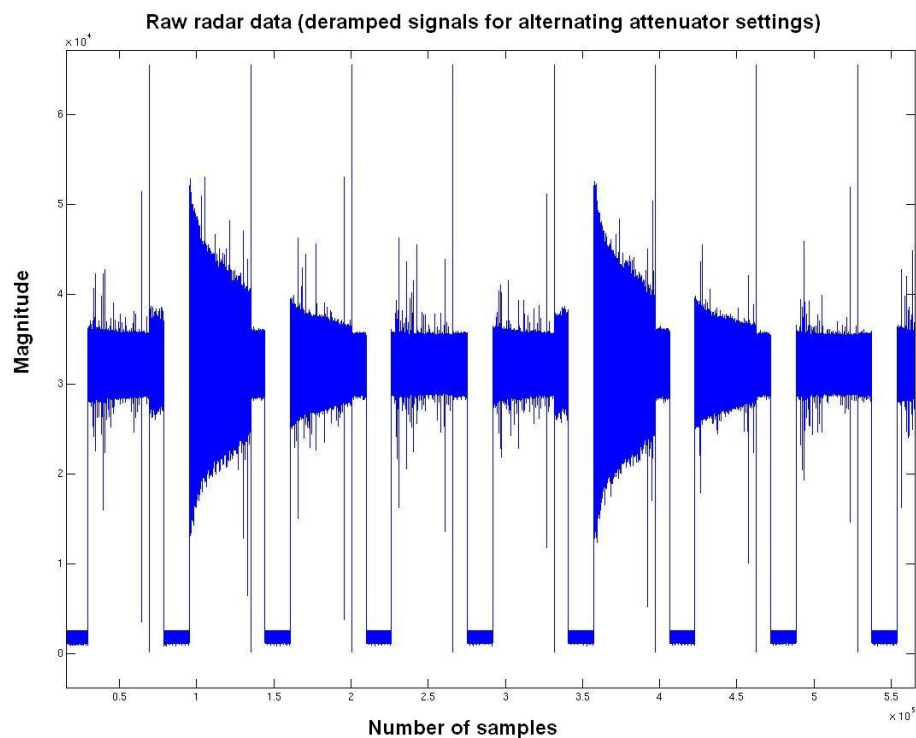


Figure 5.1: Raw received deramped signals of the FMCW phase sensitive radar with 4 alternating attenuator settings.

The above figure shows the MATLAB plot of the received signal from the FMCW radar. It shows 8 chirps (most of the 9<sup>th</sup> chirp is clipped off). As discussed in Chapter 2, each successive chirps correspond to four different gain values that change sequentially. The alternating attenuator setting allows to achieve the different gains (4, 16, 28 and 40 dB). The effect of time delay error correction (discussed in Chapter 3)

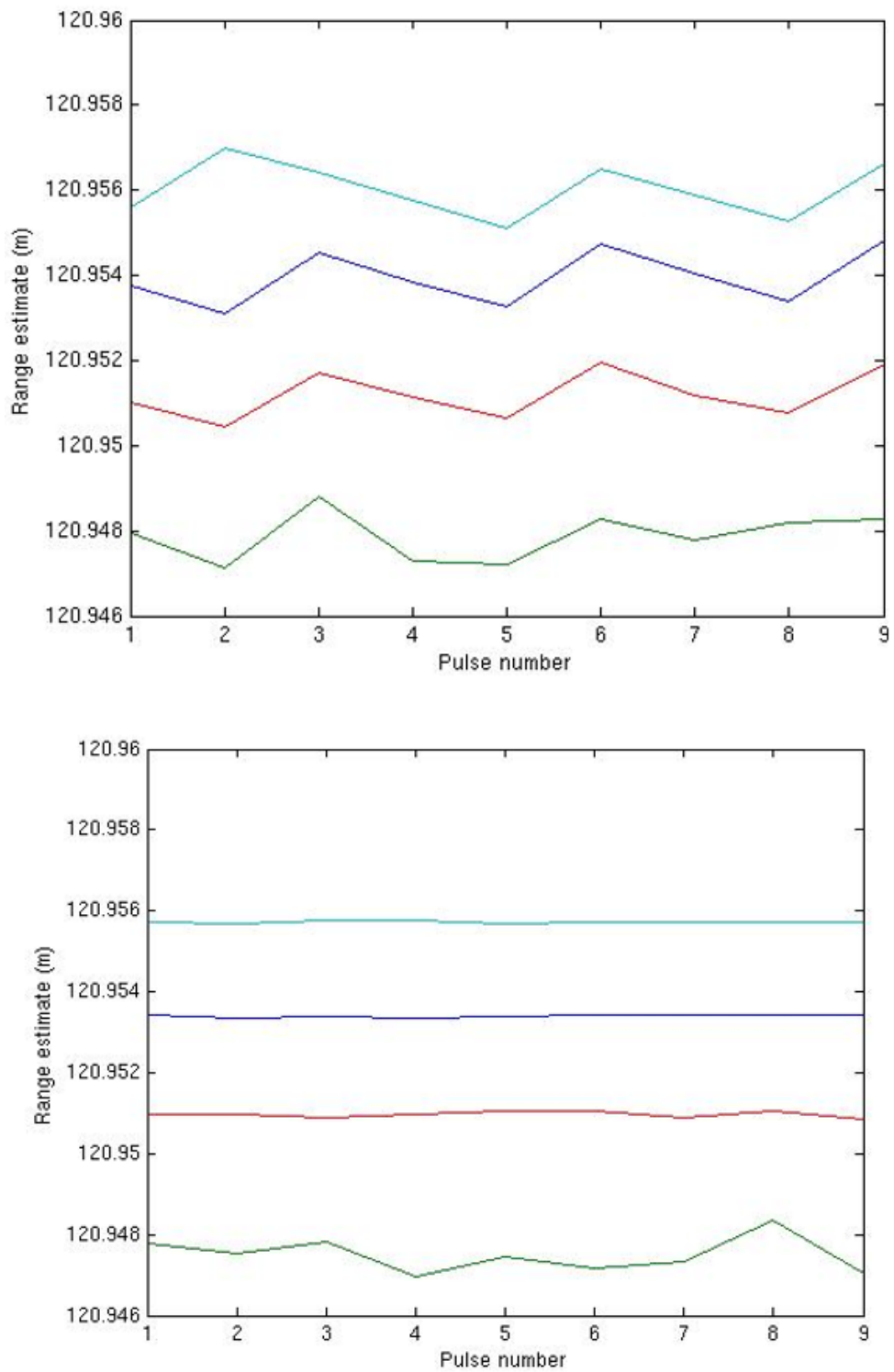


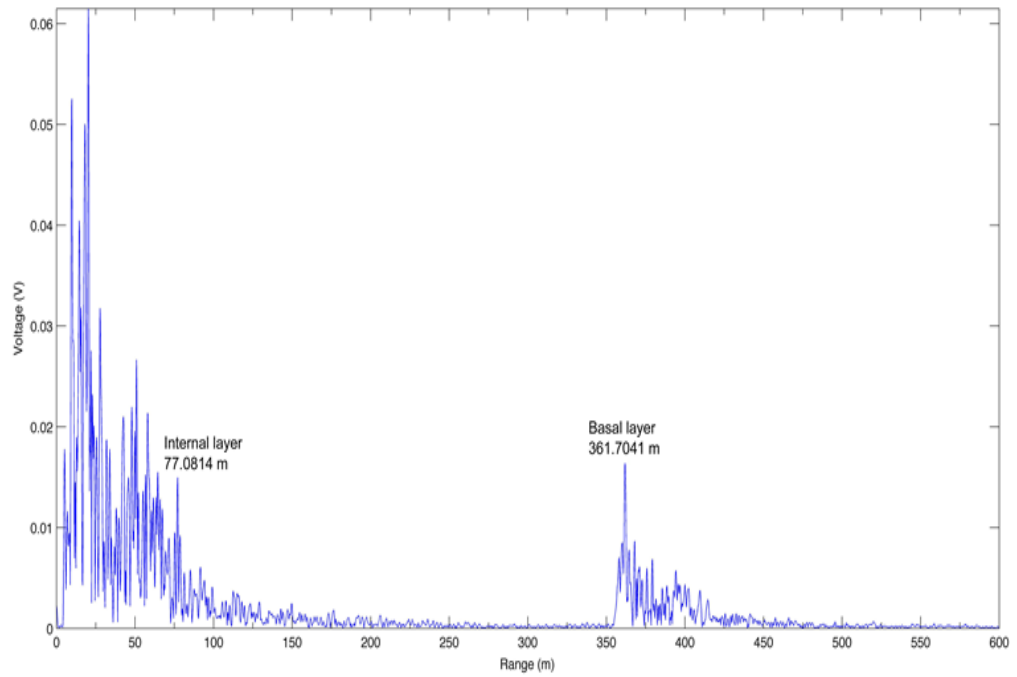
Figure 5.2: Range plots of the 240 meter round trip path through the cable for all 4 gain settings of the radar, without time error correction (above) and with time error correction (below).

can be visualized as the standard deviation decreases when the correction is applied. Even though the pulse duration is 1s, the radar runs for 1 minute as it switched over to the 4 attenuator settings. For each attenuator setting, it records 9 pulses as can be seen from Figure (5.2).

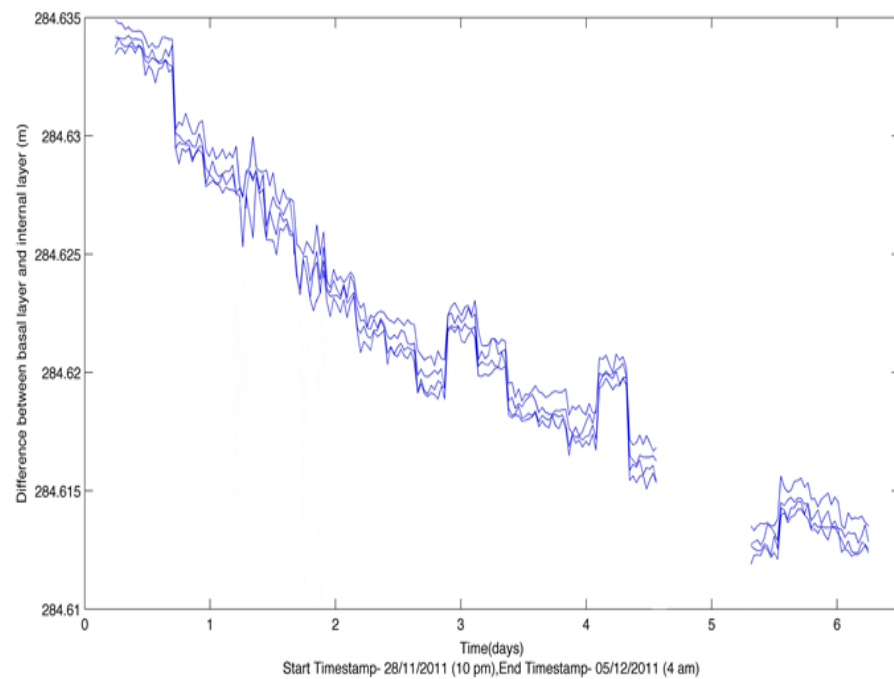
## 5.2 Range Profiling

The phase sensitive FMCW radar signal processing method described in Chapter 3 has been used for obtaining range profiles of radar. Figure (5.3(a)) shows the range profile of the Larsen-C ice shelf where the reflection from the basal layer can be seen at 362 metres. The high amplitude signal observed at close range is mainly due to antenna direct coupling or other features creating reflections. As can be seen from the range profile, there is a trail after the sharp peak from the basal layer. This trail corresponds to the returns coming from the non-zero angles. This trail is desired as it will provide more information about the basal layer when beamforming is applied. A shape matching algorithm is used to determine the presence of any feature among this. But in this case for Larsen-C ice shelf, a presence of an internal layer at around 77 metres is known from previous measurements (information obtained from BAS). So, the performance of the algorithm can be tested by applying it to the Larsen-C profile to see if it can detect the internal layer, which is shown later. Figure (5.3(b)) shows the differential plot of the basal layer and the internal layer over 6 days. The radar took continuous measurements from 28th November 2011 (10 pm) to 5th December 2011 (4am). The figure contains plots for measured results from all four attenuator settings. The gap seen in the plot is due to data loss during that period (data logger did not record any data). Even though there are fluctuations seen on the plot, the overall trend is clearly downward which infers the melting of the base. The average gives a melting

rate of around 3.3 mm/day. It should be considered that the data was collected for six days and during Austral summer, when melting of the ice shelf base is expected. So, the melt rate should not be used to calculate the total melting over the year. In order to do that, the year-long data needs to be processed. Only then the overall melting pattern of the ice shelf can be properly demonstrated. Figure (5.4) shows a continuous range plot for a single attenuator. A Total of 130 measurements were taken for 64.5 hours (1 in every 30 minutes). In the middle of the figure, there is an approximately 5 mm instant slump in range. That was due to a cable fault during measurement. The range return is from 386 metres. It can be seen that there has been a cyclic pattern in the range plot indicating ice accumulation and melt both occurring during the measurement period. Figure (5.5) shows range plot of Larsen South Ice Shelf where 495 continuous measurements were taken (1 in every 10 minutes). Within this 82 hours and 20 minutes, 1.6 mm melting is recorded. It should be noted that this is the first time a phase precision radar system is deployed in this area. So, there is no means for cross checking the data unless some other researchers take measurement at the same place at the same time. The confidence on the radar performance is achieved from the loop test results in the lab as it is not possible to place a known target deep under the ice shelf. A corner reflector can be used for a ground measurement, but that will not validate the radar performance underneath the ice shelf due to the dielectric constant factor. As the main goal is to measure the melt rate, the result should be valid anyways as long as the offset in measured range (if there is any) is constant. Two abrupt plunges in the plot are most probably due to wrong data recording of two samples as those two data points clearly seem to be an anomaly. The data is plotted for all four attenuator settings.



(a)



(b)

Figure 5.3: (a) Larsen-C range profile showing the ice shelf base at 362 metres (b) Plot of difference between measured ice shelf base and the internal layer over 6 days.

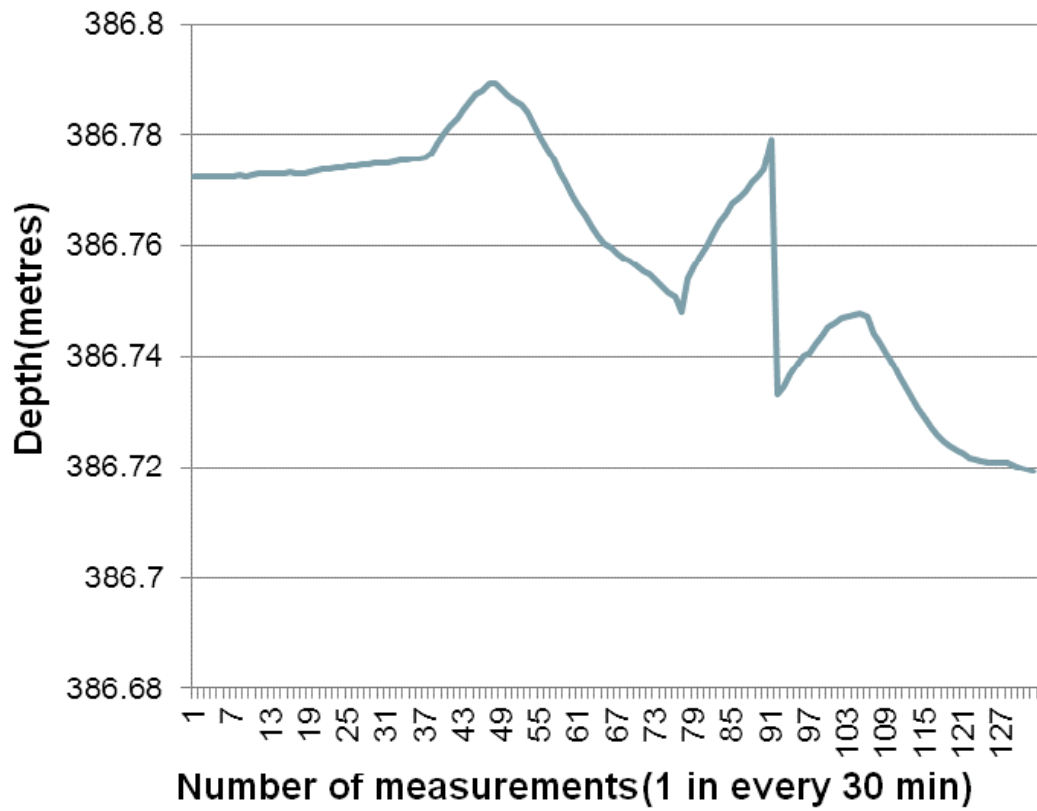


Figure 5.4: George VI Ice Shelf basal layer range (at 386 metres) plot for 130 continuous measurements, corresponding to 64.5 hours.

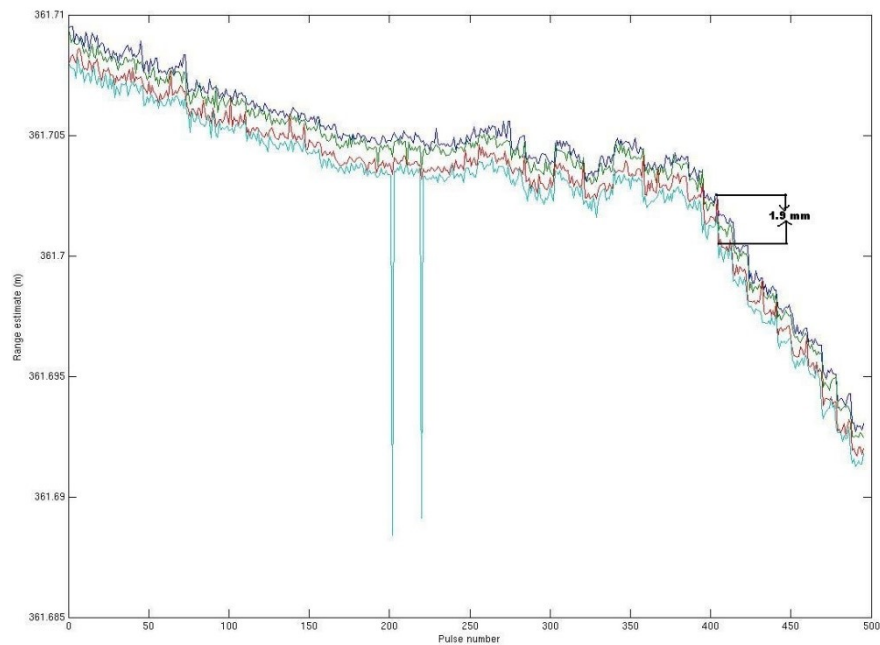


Figure 5.5: Larsen Ice Shelf basal layer plot for 495 continuous measurements (1 in every 10 minutes), corresponding to 82 hours and 20 minutes.

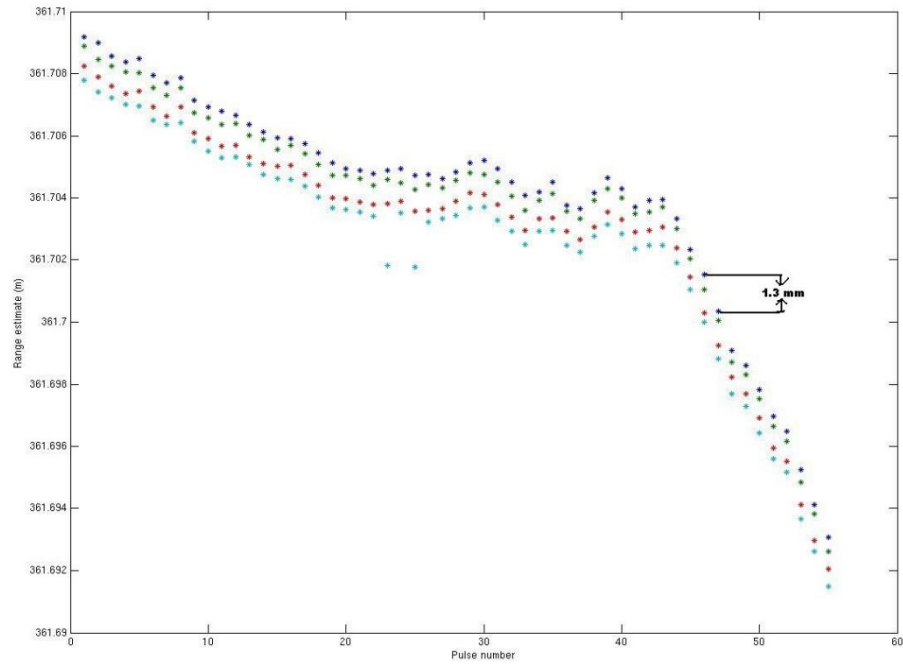


Figure 5.6: Larsen South Ice Shelf basal layer range (at 361 metres) plot for 495 continuous measurements (1 in every 10 minutes) after averaging of 9 chirps of each attenuator setting.

Figure (5.6) shows the same plot but with chirp averaging (averaging nine pulses of each attenuator settings). As it can be seen, no significant difference is observed. Hence, chirp averaging can be applied during range profiling.

Figure (5.7) shows a range profile of George VI ice shelf (basal layer return from 386 metres) along with the profile after conventional cross correlation is performed upon it. It can be seen that correlation does not help in terms of clutter/antenna direct coupling suppression or hidden feature detection. . Figure (5.8) shows the same profile but this time with shape matching algorithm implemented plot. The improvement can be very clearly observed. The base now has a very sharp high peak as that was the main reference (so *rank* (equation (4.4)) was maximum at that point).

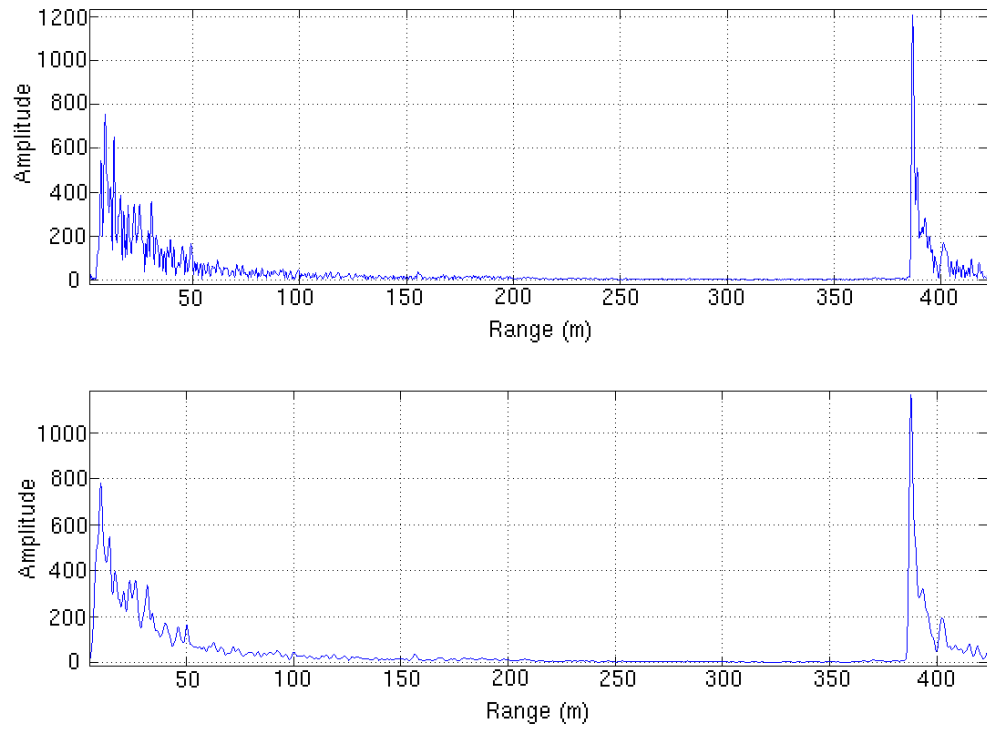


Figure 5.7: George VI range profile without cross-correlation (above) and with cross-correlation (below) having almost no impact.

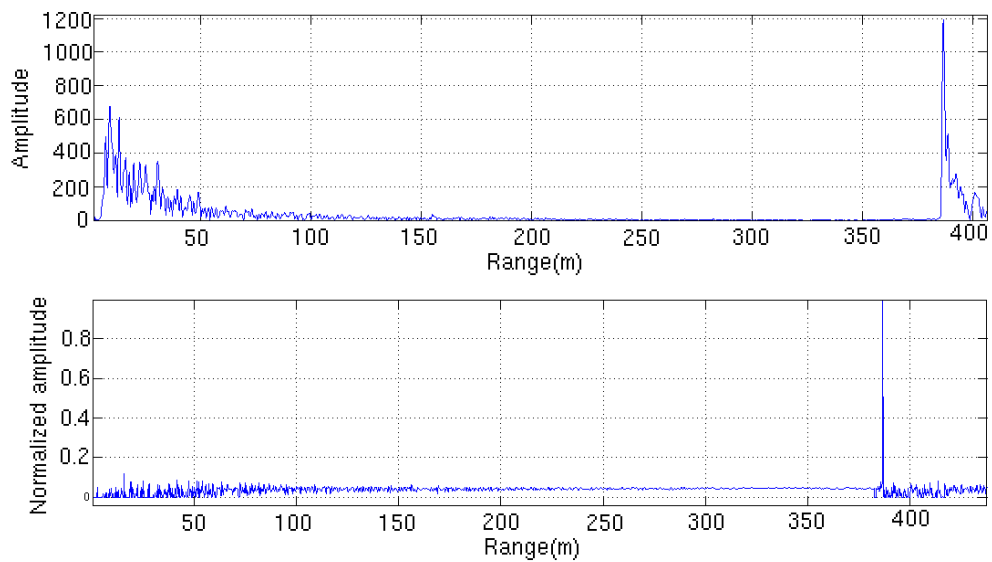


Figure 5.8: George VI range profile without shape matching algorithm implemented (above) and with the algorithm implemented (below) reducing the clutter/noise.



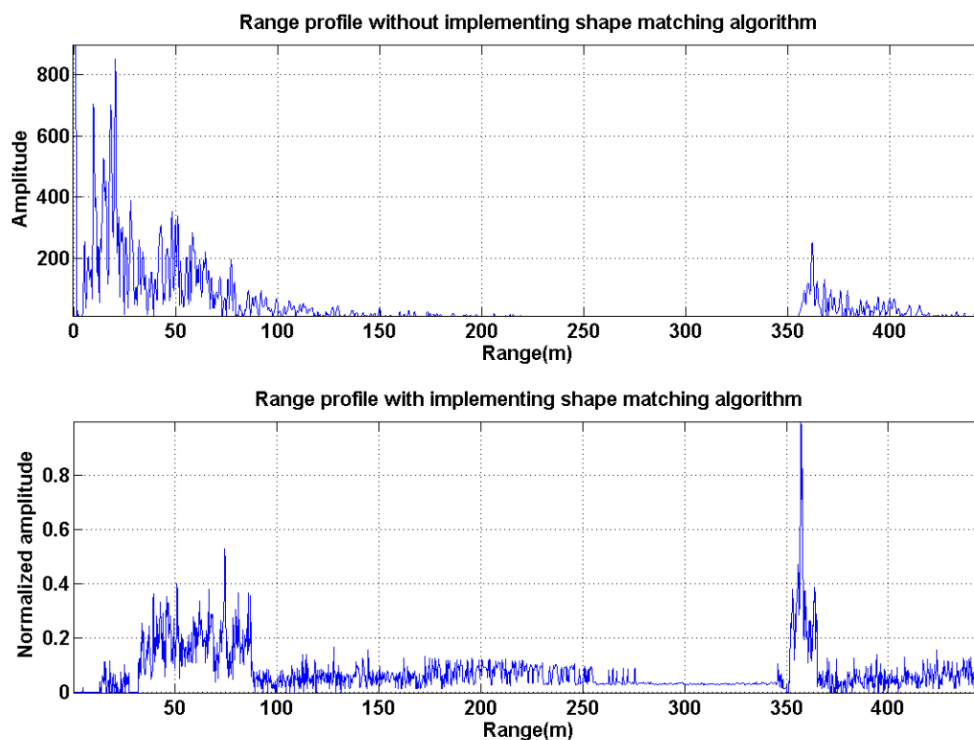


Figure 5.9: Normal Larsen ice shelf range profile (above) and the same profile after implementing shape matching algorithm showing a more than 50% match at around 77 metres.

Also, the algorithm successfully suppresses the antenna direct coupling/near surface clutter that is present throughout first 30 meters or so. However, no hidden feature is revealed from the plot. The straightforward implication of this is that there is perhaps no internal layer present in this ice shelf (or at least in the region that was covered by the antenna beam). Figure (5.9) shows the range profile of the Larsen ice shelf along with the profile with the shape matching algorithm processing. Along with close range signal attenuation, at around 77 metres, a feature is detected. As mentioned before, it is known that an internal layer exists around this region. Without the implementation of the algorithm, it is almost impossible to detect with just normal range processing. This justifies the algorithms' application as the processed result by using the shape matching algorithm agrees with the previously known result from the Larsen ice shelf.

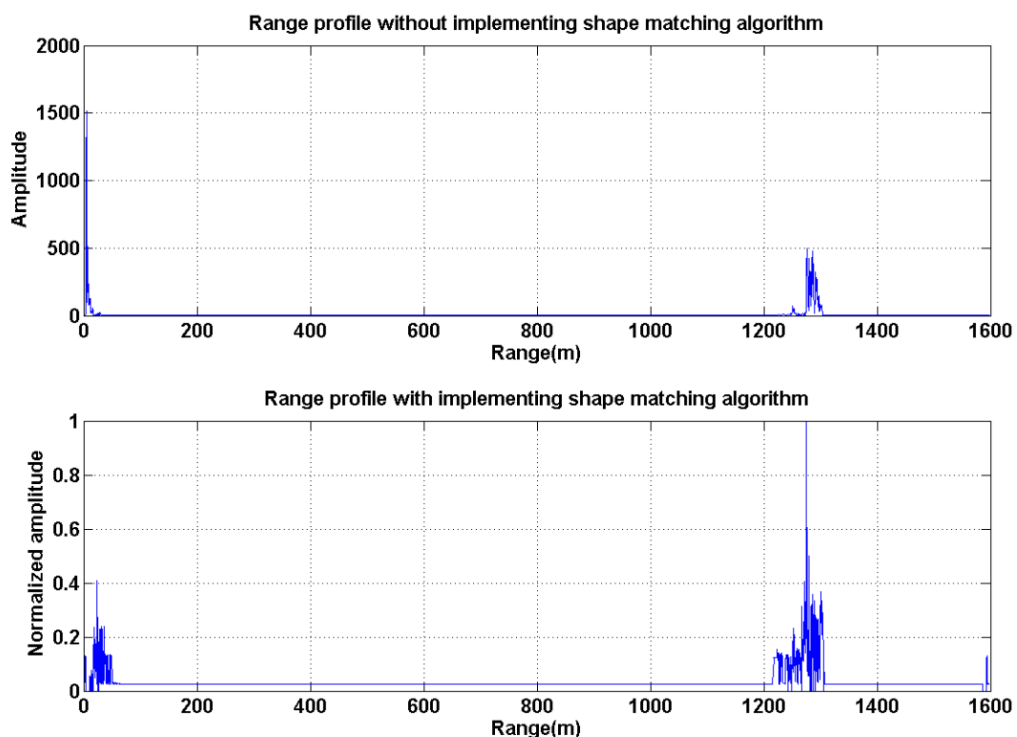


Figure 5.10: Normal Foundation ice stream range profile (above) and the same profile after implementing shape matching algorithm.

Figure (5.10) demonstrates a similar profile plot, this time for the Foundation ice stream. The basal layer return is coming from 1285 metres underneath the surface. Here, the algorithm also suppresses the high amplitude signals, but a feature is revealed around 23 meter. This internal layer is used for calibrating the radar data for melt rate measurement.

### 5.3 Foundation Ice Stream Imaging

For imaging Foundation Ice Stream, pRES radar was used. As mentioned earlier, it is similar to FMCW radar but during modulation it uses stepped frequency increase method. The parameters for the pRES system are given in Table 2. The imaging experiment done in this location was a MIMO experiment. Figure (5.11) shows the

MIMO array configuration, which is a 6x6 MIMO creating a 36 element linear virtual array (virtual array spacing being 0.14 metres).

Table 2: Parameters of pRES radar

<i>Parameters</i>	<i>Value</i>
Centre frequency, $f_c$	305 MHz
Step frequency, $f_{step}$	32 KHz
Bandwidth, $B$	160 MHz
Pulse duration, $T$	1s
Ice dielectric constant, $\epsilon_r$	3.1

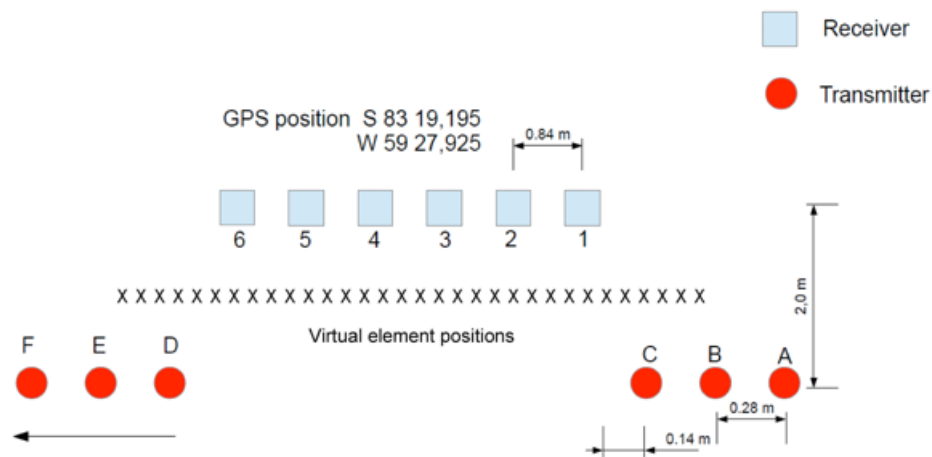


Figure 5.11: Foundation Ice Stream imaging experiment setup, comprising six transmitters and six receivers, creating 36 linear virtual array elements.

Both range migration algorithm (according to equation (4.8), it needs to be applied) and vertical error correction have been implemented during image processing. The effects of those on final image are demonstrated here as well. Figure (5.12) is the log plot of the range profile showing the basal return from 1285 metres. Phases of the corresponding range bin for all 36 elements (shown in figure (5.13)) have been used for vertical displacement correction. Figure (5.14) shows the processed MIMO image without any correction. As the virtual array is linear, the processing was the same as a conventional phased array image processing with  $\pm 60$  degree scanning. The basal layer can be seen from the image clearly. There is another return visible at

around 70 metres below the layer which is due to reflections from cables that are feeding the antenna. Figure (5.11) shows the image after applying range migration and phase correction. Comparing of Figure (5.14) and (5.15) demonstrates the effect of correction process for image quality improvement. It can be seen that after the correction process, the basal layer feature is now better defined showing more consistent intensity across the entire scan. Which means the signal energy has been more concentrated than it was before. Same conclusion can also be made by observing the clutter structure. In Figure (5.14), clutter is smeared out where in Figure (5.15), it is sharper which means energy is not smeared out anymore. As it was discussed in Chapter 4, image defocusing due to range migration becomes worse as scanning angle increases. Here in Figure (5.15), it is seen that image is better focused at wider angles as well where basal layer signal strength is around 15-20 dB higher compared to Figure (5.14). By observing Figure (5.12), it can be seen that there is another peak beyond 50 m from the basal return. This might suggest that there is a sharp slope present in that area causing another return, but no such slope is seen on Figure (5.14) and (5.15). This may be due to the fact that the antenna beam width was not sharp enough, so after beamforming this physical phenomena could not be observed. It can also be an artefact, in which case the beamforming will not show any slope which is not present. Also, the region between the surface and the base is quite plain, suggesting a homogenous ice structure throughout that area.

During the Foundation Ice Stream expedition, the MIMO experimentation was performed twice with a six day time gap. Both measurements were made in exactly the same location. So, phase differential was calculated by measuring the phases of the two processed images. The idea was to obtain a differential phase interferometry like information to observe minute changes in the basal layer during that 6 days. Figure

(5.16) shows the result of this attempt. After subtracting signal phase of one image from the other, the resultant differential phases were converted into distance values.

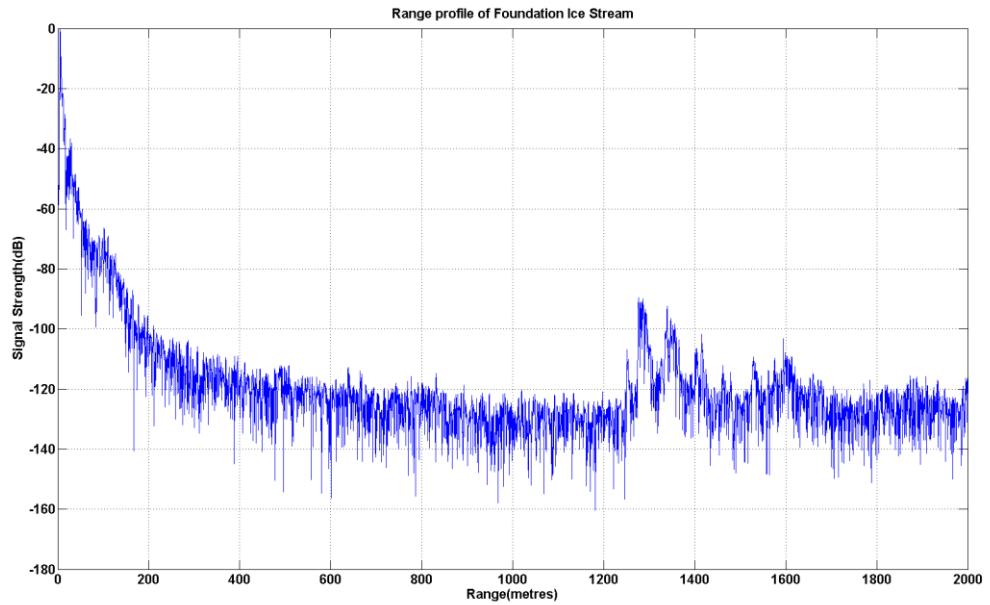


Figure 5.12: Measured result taken from Foundation Ice Stream, Antarctica showing range profile.

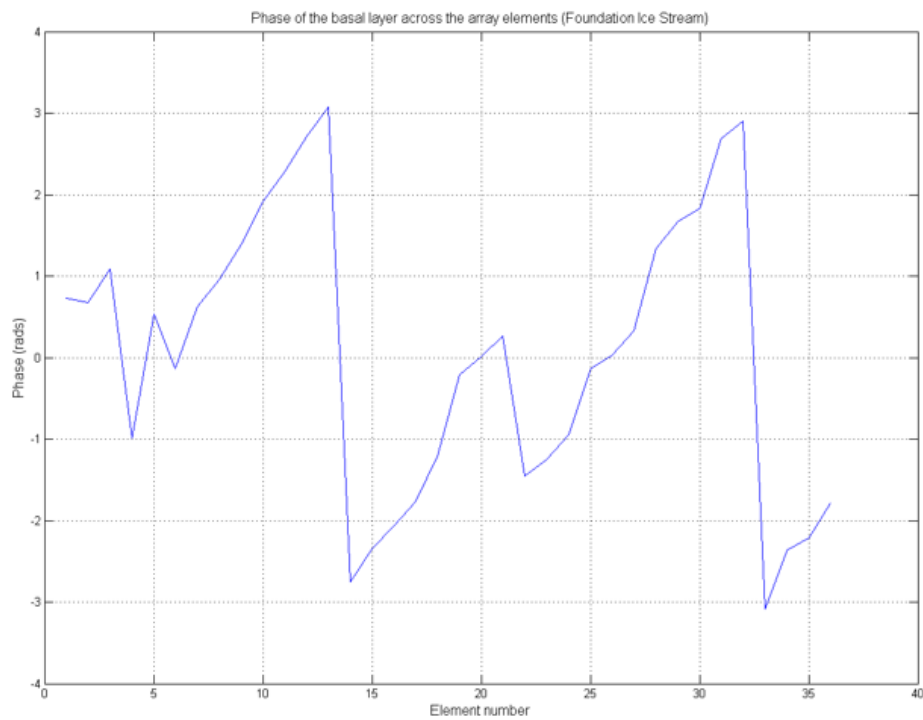


Figure 5.13: Measured phase of the range bin corresponding to the basal layer across each of the 36 array elements, to act as an array phase calibration.

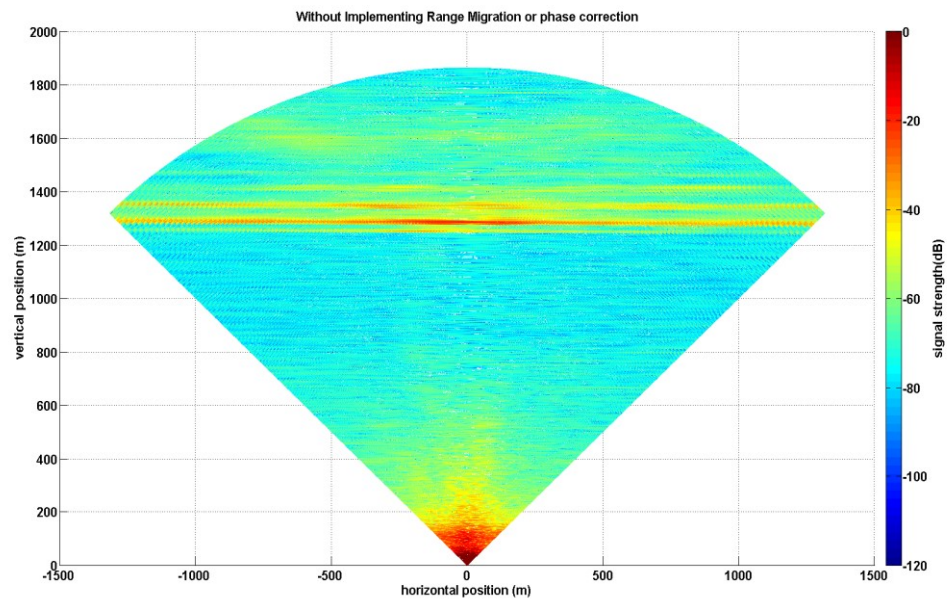


Figure 5.14: MIMO image of the Foundation Ice Stream, without range migration processing or vertical error correction; the basal layer return echo is at 1285 metres.

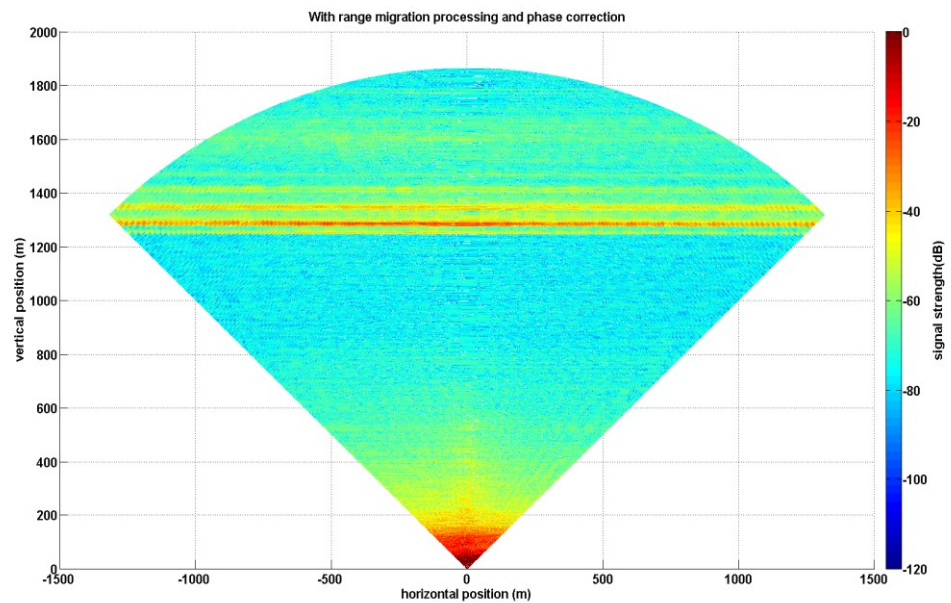


Figure 5.15: MIMO image of the Foundation Ice Stream, using range migration processing and phase calibration, showing consistent image intensity at the basal layer and better-defined clutter structure.

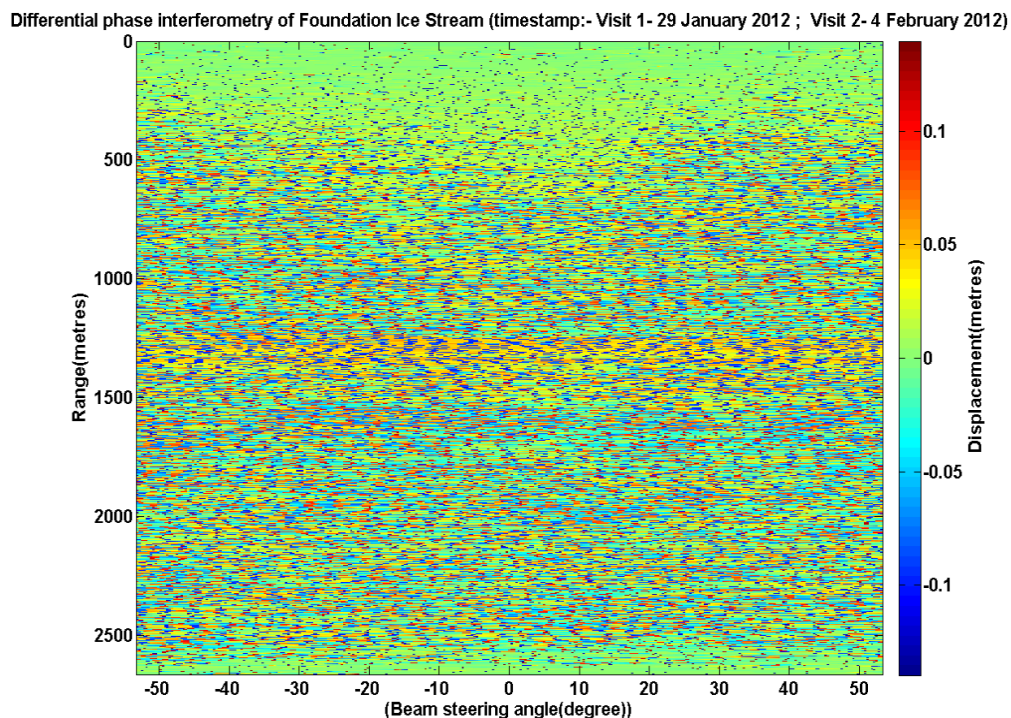


Figure 5.16: Differential phase interferometry image of Foundation ice stream to observe the basal layer melt within 6 days using the detected internal layer for calibration.

The figure shows the eventual image where a thin continuous line can be observed at basal layer (around 1285 metres). This line corresponds to around 40-50 millimetre (around 7-8 mm/day melt rate). It should also be noted that there is no distinct second line present in the interferometry image. It suggests that the second peak seen in Figure (5.12) is most likely an artefact, which got cancelled out.

## 5.4 Larsen North Imaging

In the Larsen North expedition, the experimental setup was done by mounting the FMCW radar on a sledge and taking measurement 109 times along a straight line. The objective was to simulate a linear uniform array with 20 cm spacing (approximately) between each other. The spacing has been chosen to ensure that no grating lobe occurs. A SAR like processing was done by making few assumptions. SAR processing requires the antenna aperture to be in constant motion during data

acquisition. This assumption can be made for the Larsen North processing as well. The fact that the radar was stationary during each element position can be related to the usual ‘stop and go’ assumption for the SAR processing. The Range Migration Algorithm (RMA) for SAR processing discussed in Chapter 3 is quite suitable for this scenario as it stems from seismic data processing technique [53].

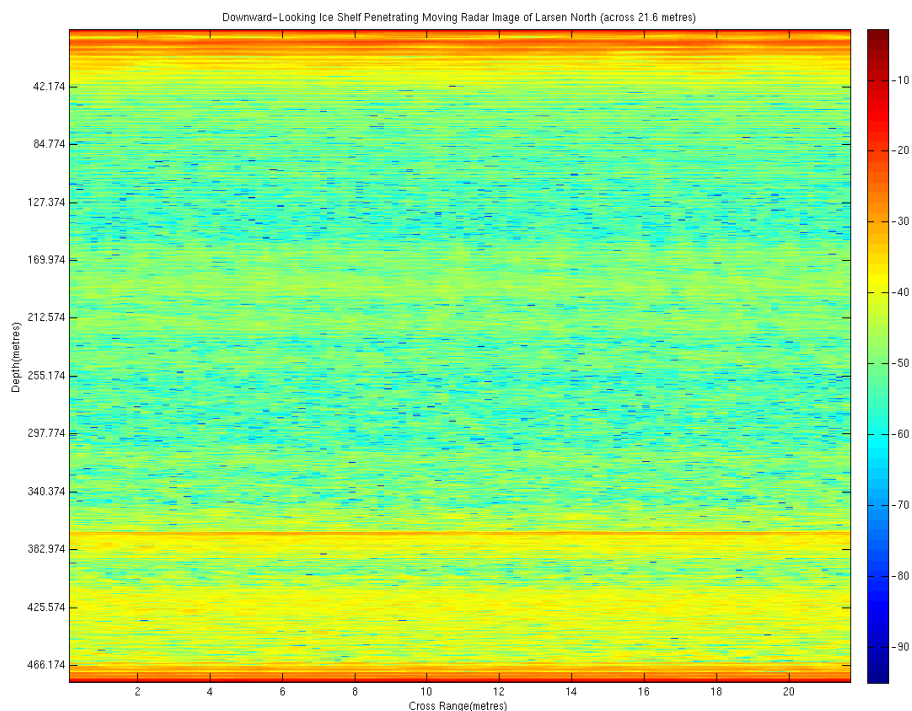


Figure 5.17: SAR processed 2D cross-sectional image of Larsen North ice shelf.

For seismic data processing, some numbers of geophones are placed along a straight line on the ground and a charge is detonated along that straight line. Every single geophone uses the received sound echo and the combined results are analyzed to figure out ground characteristics. This placement of geophones is similar to specific SAR platform positions at any given instance, as well as the experimental arrangement of Larsen North.



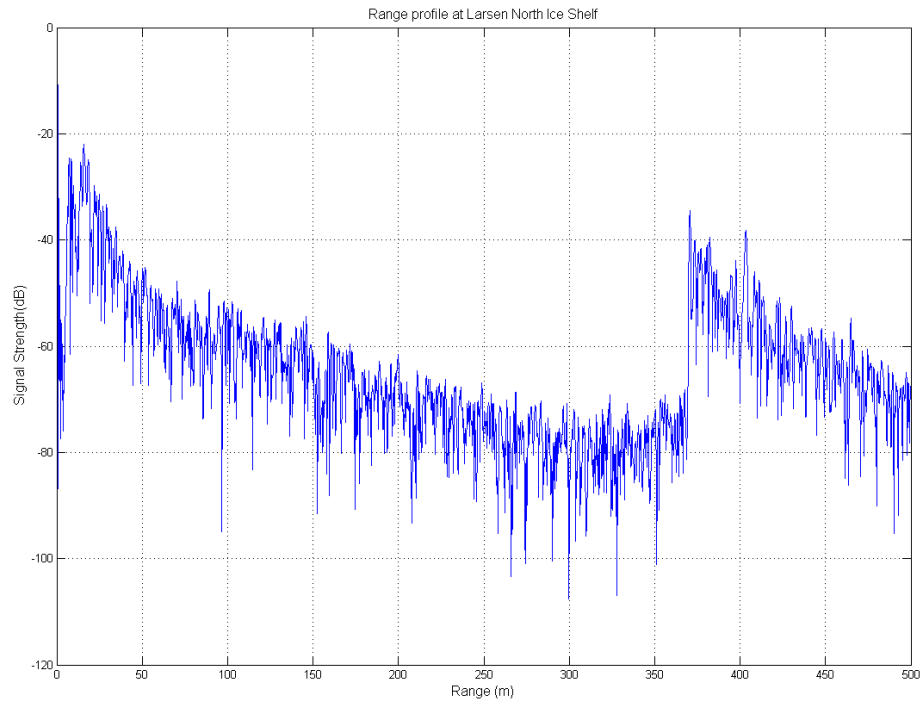


Figure 5.18: Measured result taken from Larsen North Ice Shelf, Antarctica showing range profile.

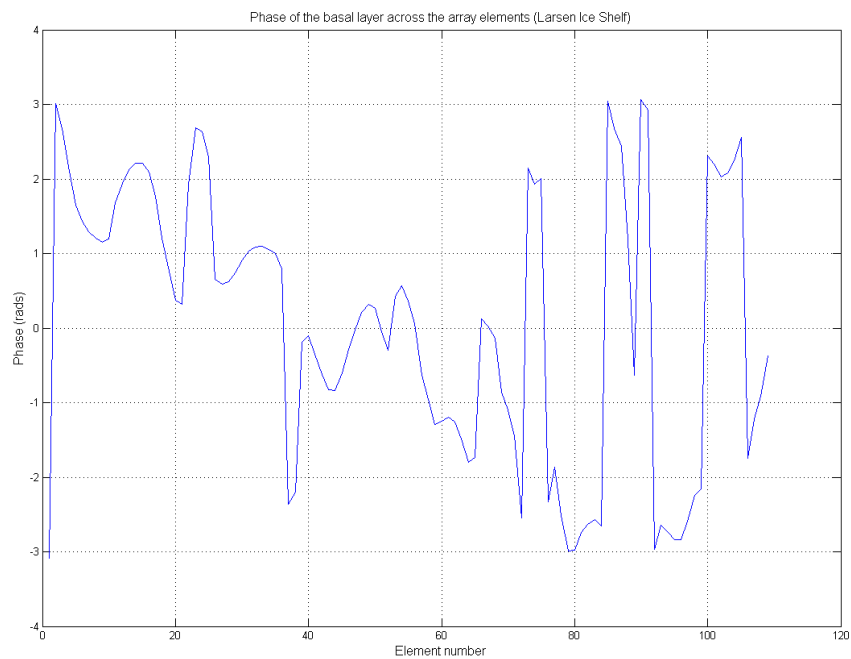


Figure 5.19: Measured phase to the basal layer across each of the 109 array elements, to act as an array phase calibration.

Figure (5.17) shows the SAR processed cross sectional image of the Larsen North data. Strong reflection from the basal layer can be seen at around 370 metres. It can be observed the cross range view is quite smaller (21.6 metres) compared to the depth due to very low integration angle relative to basal layer. The strong returns from first 20-30 metres are due to the direct coupling between transmitting and receiving antennas. The Same data was also processed by phased array imaging algorithm which would be able to provide larger cross range view compared to SAR. The processing technique used for MIMO imaging, was also used for Larsen North. During this processing, the beam has been scanned  $\pm 60^\circ$ . This covers roughly 700 meters in cross range, so more features can be seen of the basal layer as well as any other special features above if present. Figure (5.18) shows the range profile of Larsen North showing the basal layer return from 370 metres. The phase of the range bin corresponding to this of each element is plotted in Figure (5.19). This is used for phase calibration to compensate for vertical displacement error if there is any. Figure (5.20) is the processed image without range migration correction. Figure (5.21) is the image after applying the range migration and phase correction. Comparison of Figure (5.20) and Figure (5.21) gives same conclusion as it was obtained from Foundation Ice Stream image processing. The implementation of the corrections improves the image quality by increasing the signal strength which means now there is better coherence. Also, Figure (5.21) shows the basal layer having a slope rather than a straight line, which is often natural as in reality a basal layer of the ice shelf will not be a horizontal base all the way through. Figure (5.22) shows the Larsen North phased array image with increased range profile. This is to get a visual observation of the radar signal reflecting from the base, coming back and reflected downward again from the surface of the ice shelf, then coming back up again and captured by radar. This is known as

the double bounce affect. As the signal propagates more distance, it gets attenuated quite a lot through ice, which is evident by the lower signal strength of the double bounce from the basal return.

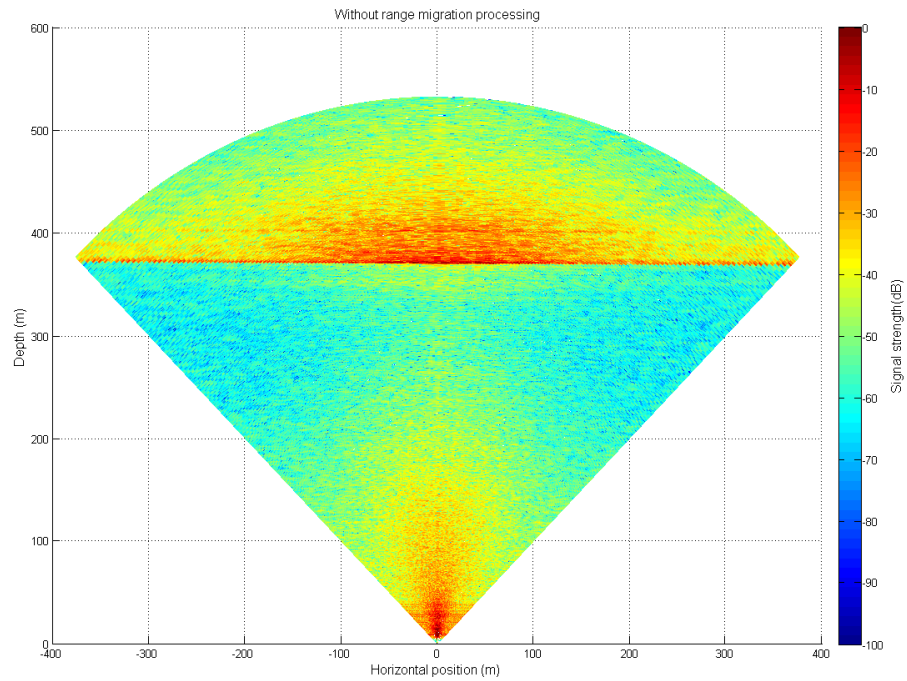


Figure 5.20: Phased array image of the Larsen North Ice Shelf, without any correction; the basal layer return echo is at 370 metres.

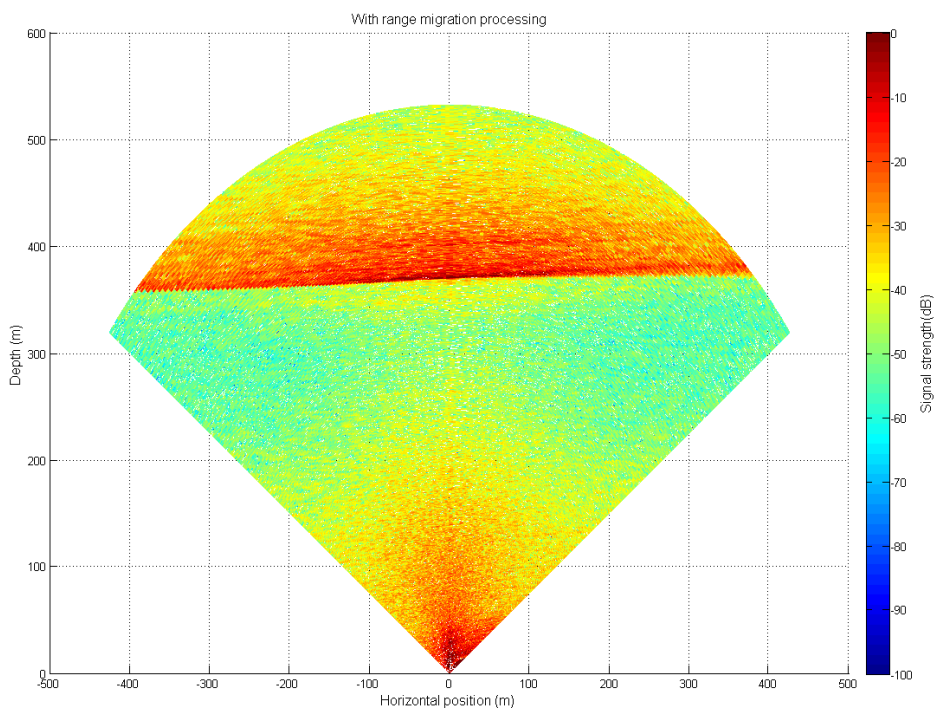


Figure 5.21: Phased array image of the Larsen Ice Shelf, using range migration processing and phase calibration, showing consistent image intensity at the basal layer and better-defined clutter structure.

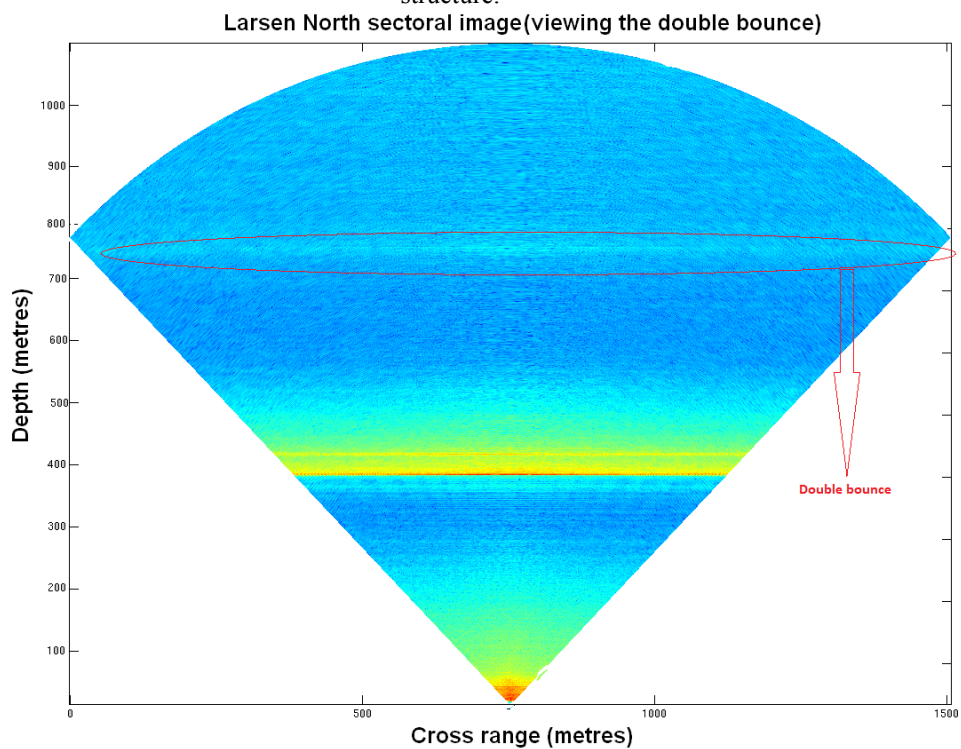


Figure 5.22: Phased array image of the Larsen North Ice Shelf showing the thin line showing the double bounce return coming from twice the basal layer range.

### 5.5 3-D MIMO Imaging of Ice Shelf Base

The 8x8 MIMO radar was deployed in the Ronne Ice Shelf during 2014-15 Austral summer. The MIMO array configuration can be seen at Figure (5.23). The physical length of the array is 6.42 x 6.47 metres. The 1.08 m gap between the adjacent transmitter and the receiver on the top left of the geometry is to get proper antenna isolation. As trenches were made to mount each antenna element, some were 5 metres deep and some were 10 metres. This has been taken into account during processing. The antennas used are bow tie antennas operating at 300 MHz centre frequency with 200 MHz bandwidth. The radar operation is based on FMCW principle where it transmits up-chirp signals (with time switching) and the receiver outputs deramped signal after mixing with reference chirp. The radar measurement was taken in two different sites in Ronne Ice Shelf, as well as in Greenland.

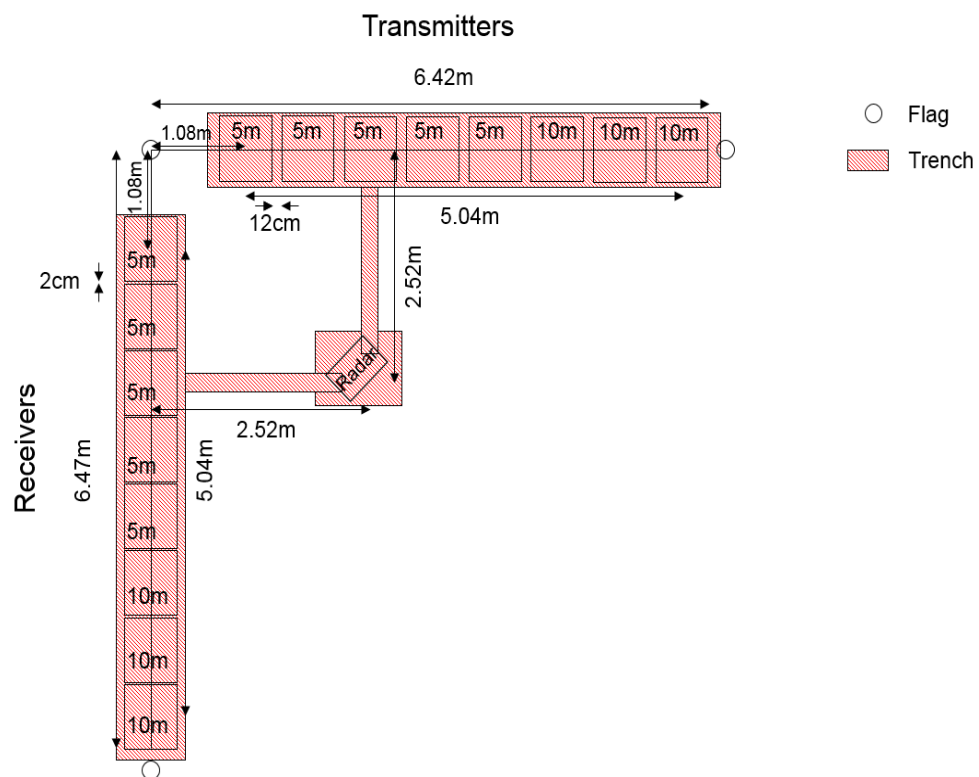


Figure 5.23: Planar MIMO antenna array geometry for 3-D imaging of Antarctic Ice Shelf.

Figure (5.24) shows the processed image of the ice shelf base at 566 metres depth. Range slice corresponding to that depth has been imaged here. Scan angle is  $\pm 25^\circ$  both in azimuth and elevation. At 566 metres deep, this corresponds to  $653.5 \times 653.5$  square metres. Array baseline length is not large enough for range migration to take effect, so no correction has been applied. But the image is processed without 2-D correlation. The ring patterns in the image are very apparent, which makes it very hard to properly visualize and analyze the result. Figure (5.25) shows the same image after the correlation is applied. The significant change in the image quality is easily noticeable. The curves within the curvature of the basal layer can now be visualized. The image shows high signal strength in the middle region and reduced signal strength elsewhere, which is expected as the scanning angle increases. The basal layer is not plain and regular bumps are observed throughout the scanning area. This suggests that there is mass imbalance in the ice shelf causing deformation [103]. Mass imbalance occurs when there is a difference between amount of ice accumulation and ablation, causing ice flow underneath.

Figure (5.26) illustrates the processed image at a different site in the same ice shelf without the 2-D correction. Here the basal layer is at 514 metres. The scanning angle is the same as it was for the previous site, which corresponds to coverage area of  $593.5 \times 593.5$  square metres. Figure (5.27) shows the same image after suppressing the circular rings. Figure (5.28) and (5.29) shows image processed in same method and parameters, but with data obtained from Greenland. The basal layer here is at 638 metres, which means the coverage area is approximately  $736.72 \times 736.72$  square metres. Figure (5.29) shows a very plain basal layer compared to the images from the Antarctica. This is expected as this is an ice sheet base instead of an ice shelf. Ice sheets are more stable than ice shelves which melts rapidly compared to an ice sheet.

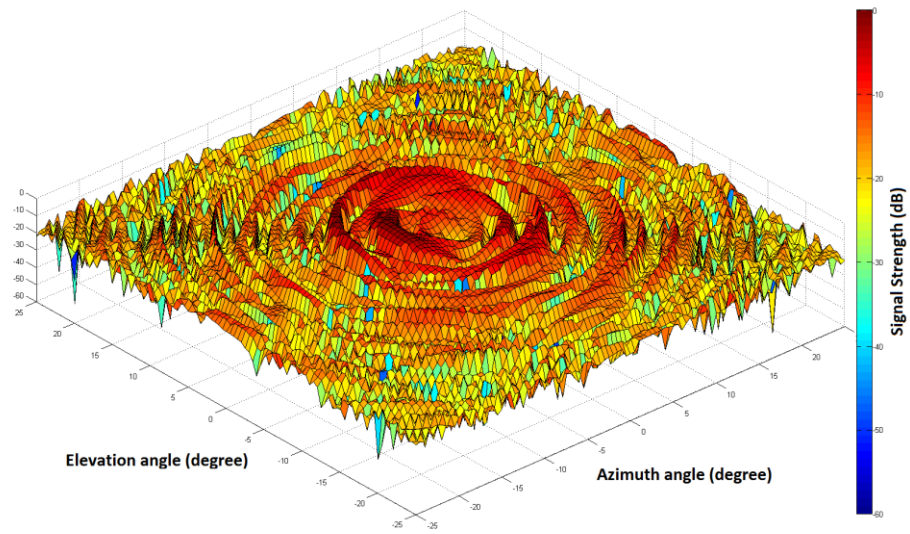


Figure 5.24: 3-D MIMO image of Ronne Ice Shelf (site1) basal layer at 566 metres (without 2-D correlation).

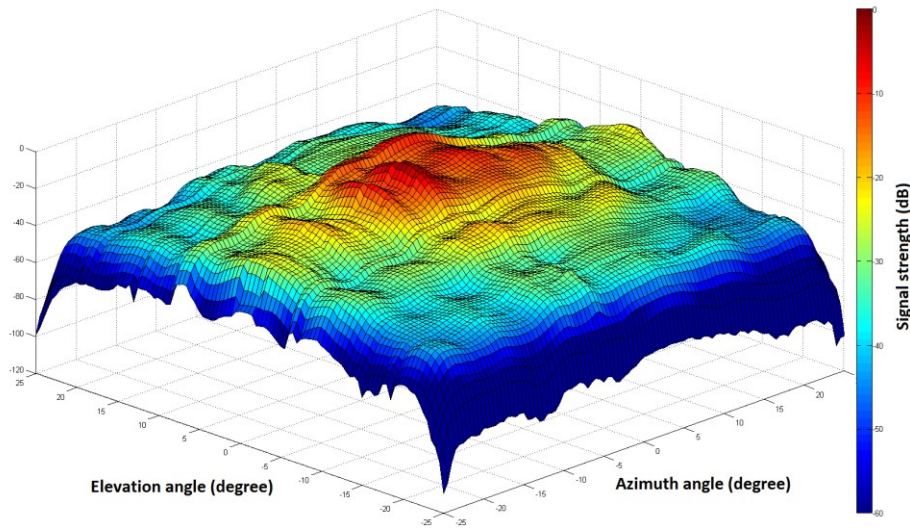


Figure 5.25: 3-D MIMO image of Ronne Ice Shelf (site1) basal layer at 566 metres (after 2-D correlation).

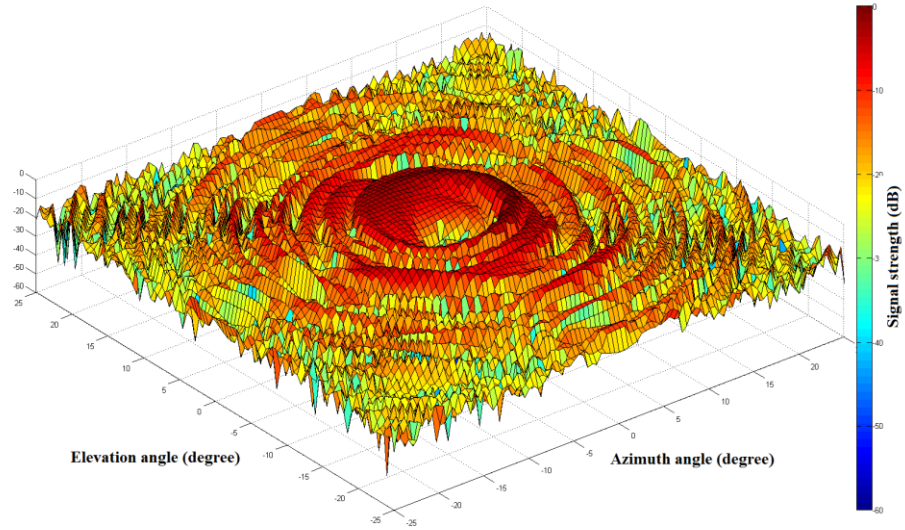


Figure 5.26: 3-D MIMO image of Ronne Ice Shelf (site2) basal layer at 514 metres (without 2-D correlation).

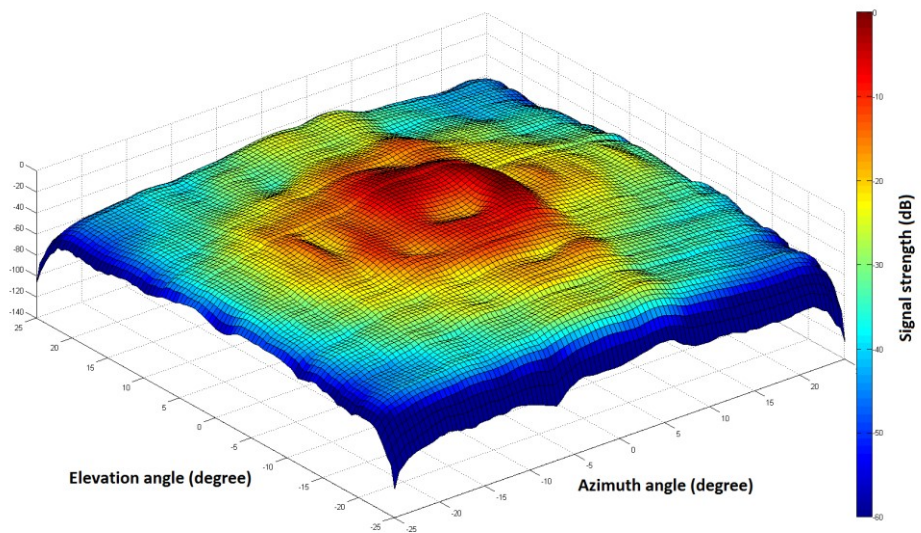


Figure 5.27: 3-D MIMO image of Ronne Ice Shelf (site2) basal layer at 514 metres (after 2-D correlation).



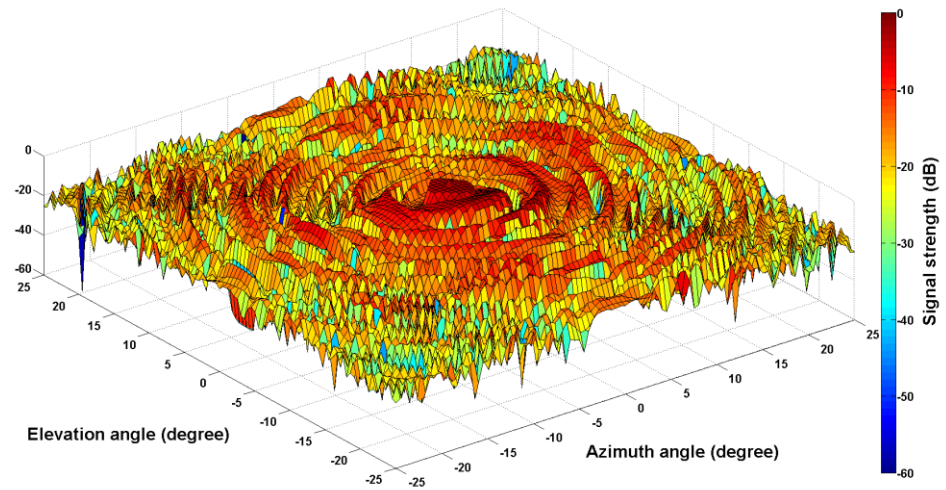


Figure 5.28: 3-D MIMO image of Greenland Ice Sheet basal layer at 638 metres (without 2-D correlation).

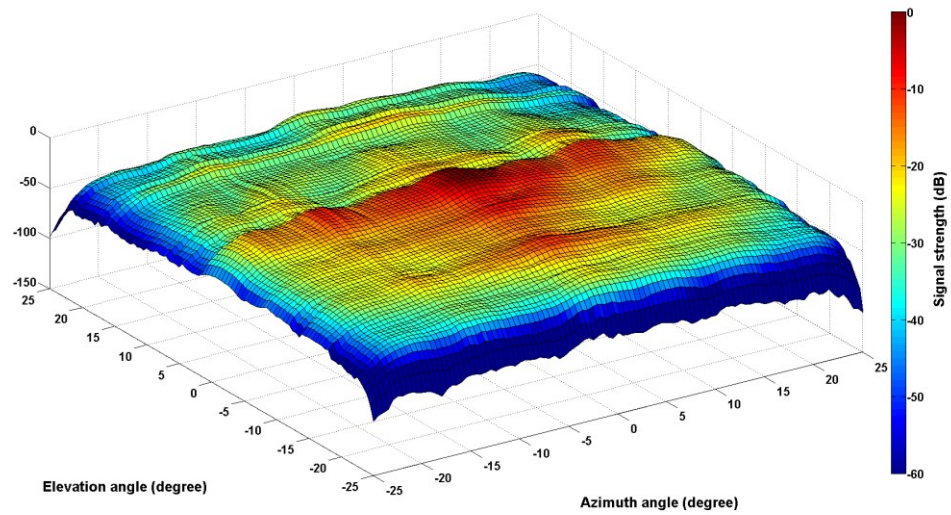


Figure 5.29: 3-D MIMO image of Greenland Ice Sheet basal layer at 638 metres (after 2-D correlation).

## Chapter 6

# Conclusions and Future Work

### 6.1 Conclusions

The main goal of the thesis work was to develop novel signal processing algorithms for phase-sensitive FMCW radar data to profile and image Antarctic Ice Shelves. Along with developing new processing techniques, conventional processing methods also had to be learned and used for data processing, which are phased array/SAR/MIMO imaging and FMCW radar range profiling. The first part of the thesis encompassed mostly studying and implementing those algorithms on raw radar data from Antarctica. The algorithms had been implemented in MATLAB in a way that they would accommodate the nature of the FMCW radar data. While processing the real data, new signal processing challenges had been faced which motivated to develop new processing techniques. These techniques had been simulated in terms of Antarctic ice shelf experiment scenario at first to validate the principle. The results obtained by processing the Antarctic data have validated the use of developed methods towards better analysis of the ice shelves.

The list of achievements and advances made over the existing techniques during the thesis are as follows-

***Millimetre precision range profiling of the Antarctic dataset-*** The phase-sensitive FMCW radar built for the year-long Antarctic ice shelf monitoring is a novel system. In [46], FMCW radar has been used for similar purpose but with no phase processing and with reduced resolution (1 m) compared to the radar built at UCL. In [104] [105],

large area of the Antarctic ice shelves have been mapped by using satellite radar altimeter. The precision achieved from this system is not satisfactory (around 20 cm). Also, these are pulsed radar systems which are not favourable for high precision measurement in a lossy medium like ice with low power requirements. Hence, new signal processing techniques to process the large amount of data at once to obtain the ice shelf melt rate over a long period was required. The first achievement during this PhD thesis was the MATLAB code written to process the large dataset from the Antarctica, which outputs the range profile by calculating the fine range and adding it with the coarse range, to get the millimetre range precision.

***Shape matching algorithm development-*** Detection of internal layers has been a big challenge during the monitoring of the Antarctic ice shelves. There is no literature present which deals with this particular issue. The conventional method for finding any hidden target is to take some known signal as reference and performing a cross-correlation along the range profile. Due to the nature of the FMCW radar data, this conventional method does not work, as in close range antenna direct coupling produces larger amplitudes than the reference signal. So, to solve this problem, shape matching algorithm was developed which uses a point by point ranking system instead of multiplication, so the amplitude does not have any affect. The validation of this developed algorithm could only be made by applying it on a real data obtained from a place where there is a presence of a known internal layer. The algorithm was applied on such dataset from the Larsen ice shelf and was successfully able to detect the internal layer which was not visible from the normal range profile plot.

***Range migration algorithm for FMCW phased array radar-*** Range migration problem is not usually faced in a phased array system, but due to the system parameters of the FMCW radar and the nature of the received data (deramped signal, so time delay

beamforming is not possible) for the ice shelf monitoring, this became an issue. The system operates with a very high fractional bandwidth, so range migration is occurred when beamforming is applied on the received signal at large angles. New mathematical concept was developed that shows the relationship between the number of array elements and the fractional bandwidth. Also, it was mathematically shown how the range migration compensation relates to the signal phase and how to compensate for it. This algorithm was applied on the field data from the Antarctica and it was successfully able to enhance the phased array image quality at wide angles.

***Vertical error correction by phase calibration-*** A very practical consideration in the Antarctic field test is the adverse weather, which is disadvantageous to precise radar measurement. As this Antarctic ice shelf monitoring radar system was built for measuring the melt rate with very high precision, it was a big challenge to compensate for any experimental errors. The error correction method was developed that would not require any prior knowledge of the vertical displacements of the antenna. The method is based on by using phase of the zero degree return from the ice shelf base as reference. Applying this method compensated for any ambiguity in range calculation. By doing this, it allowed for the accurate calculation of the ice shelf melt rate and improved the processed image quality.

***3-D MIMO imaging of the ice shelf-*** A planer array MIMO radar system was developed at UCL for the 3-D imaging purpose. Considering its application and the radar system parameter, an entirely new processing technique was required. During the last period of the thesis, this was achieved. There are 3-D MIMO imaging techniques developed by other researchers which have been discussed in the previous chapters, but those are not suitable for this application, where a large area underneath the ice shelf will be imaged, which would create serious sidelobe issues by

conventional processing. So, a step by step method was developed which would take into account all the factors regarding the radar system parameters and the experimental scenario. The beamforming technique with the 3-D geometry, range migration correction requirements, surface curvature correction and sidelobe suppression; all of these are sequentially applied to obtain the final image. This algorithm was not only successfully applied on the Antarctic ice shelf data, but also on the Greenland ice sheet data to visualize the image of the base.

## 6.2 Future Works

The phase-sensitive radar system requires very rigorous stability in phase. In practice, it was observed that the temperature variation within the components in the RF chain has effect on eventual range values.

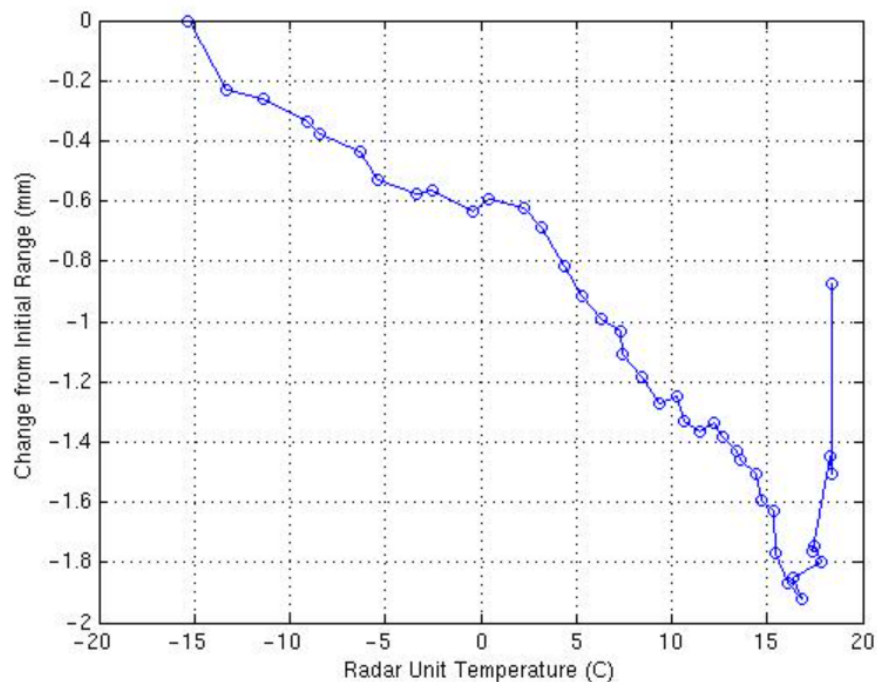


Figure 6.1: Plot of change in range values with respect to the temperature change in the radar system

The above figure shows the variation in range values when temperature of the radar unit is varied. This plot was created by using the data from one of the loop tests with

the 240 m cable. In this test, the radar was started from a low temperature ( $-15^{\circ}\text{C}$ ) and was let to operate until it reached around room temperature. A total of 40 bursts were recorded along with the radar unit temperature. All the values are for the maximum receiver gain setting (i.e. 40 dB). It shows around 1.8 mm range change in range over some  $35^{\circ}\text{C}$  temperature swing, 0.05 mm per degree Celsius. Usually, this variation is compensated by the time delay error correction and such large temperature variation is not common. Still, considering the year-long application of the radar system, such extreme condition may occur. As the time delay has a threshold of  $\pm 0.75\ \mu\text{s}$  due to expected range precision and the maximum range, the correction process will not work if the threshold is exceeded. Future researchers can look into this problem and development of an internal calibration process within the radar hardware will be a good solution (as option for external calibration with respect to some reference may not always be available during field experiments in the Antarctica).

The developed shape matching algorithm for detecting hidden layers use the shape of the basal layer return as reference. This means it will be able to detect a hidden target with similar reflection characteristics. Even though this is a valid assumption in case of the Antarctic ice shelf data, in extreme case where there are more than one internal layers very close to each other, the return signal shape might change. In that case, the algorithm may not be able to detect the hidden layer or distinguish between one internal layer from the other. Further work can be done to upgrade the algorithm which would be able to deal with the non-linear factors arising due to the geophysical structure underneath the ice shelf.

The vertical error correction method compensates for the antenna spacing errors, but there is scope for improving the final image quality. During the modelling of the algorithm, it was seen that the target signal strength was around 15 dB higher than the

sidelobe. This was good enough during simulation as the modelled basal layer was clearly visible. Also, it is expected to provide good results in a relatively low noise scenario which is the case in the Antarctica. Further improvement of the sidelobe can still be practical, as then it will allow the algorithm to be used in a much noisier scenario.

In the case of 3-D MIMO imaging, only one range slice was taken as the result was supposed to be interpreted on a 2-D plane. As the signal processing results contain images of all range slices, a 3-D interpretation method of the whole set can be developed. Also, a slice by slice comparison method can be modelled which would enable to find out about other features (if present) within the ice shelf. In this way, there is opportunity to study the entire ice shelf from surface to base (the volume covered by the radar), instead of just the basal layer. By comparing the 3-D data taken at different time, the characteristics of the ice between the surface and the base can be better understood. It will provide the geoscientists an improved understanding of the whole ice shelf underneath the surface. It should be noted that to achieve this, very large dataset have to be processed. So, the computational load will be quite heavy which should be taken into account.

Finally, after the successful operation with the prototype system in the Antarctica, the system has been replicated and number of radars have been placed at various ice shelves in the Antarctica to obtain data covering the whole year. Processing of a year-long data will give more in depth knowledge about the melting pattern of the ice shelves with respect to the seasonal change in the temperature and the ocean current.

## Appendix A

### Simulation for MIMO Snow Avalanche Imaging

Snow avalanches cause significant loss of life around the world. Understanding the dynamics of these flows is complex because the majority of the mass is in a thin granular layer that is often hidden by a cloud of suspended material.

The snow avalanche imaging radar built at UCL was a radar system consisting of 8 transmitter and 8 receivers. The radar is mounted at the avalanche bunker and looks up the path. It has an operating frequency of 5.3GHz which is a long enough wavelength to pass through the powder cloud and obtain an image of the denser layers of flowing snow beneath.

The first idea while the project for upgrading the radar system was to develop an 8x8 MIMO radar that would create a linear virtual array of 64 elements. There was option for using time switching transmitters but it would not have been suitable for imaging moving target scenario as avalanche. Hence, for MIMO operation, it was proposed that the antennas would transmit coded waveforms. Hence, all the 8 transmitters can operate at once. Before hardware implementation, the obvious work was to create a simulation of the image processing with the relevant parameters of radar. The main concern was the signal leakage due to all antenna elements being switched on at once. So, simulation was performed to view the effect of leakage on final image quality.

During simulation, Pseudo Random Noise (PN) spreading code was used. A 10000 point transmit signal was produced by generating a 1000 point random spreading code.



It is basically phase coding as the values were randomly either +1 or -1. Then, this 1000 point code was applied on a sine wave and finally was spread into a 10000 point coded signal. All the 8 transmitters had unique codes. In receive side, matched filtering (according to the code) was applied to extract signal for every transmitter receiver combination.

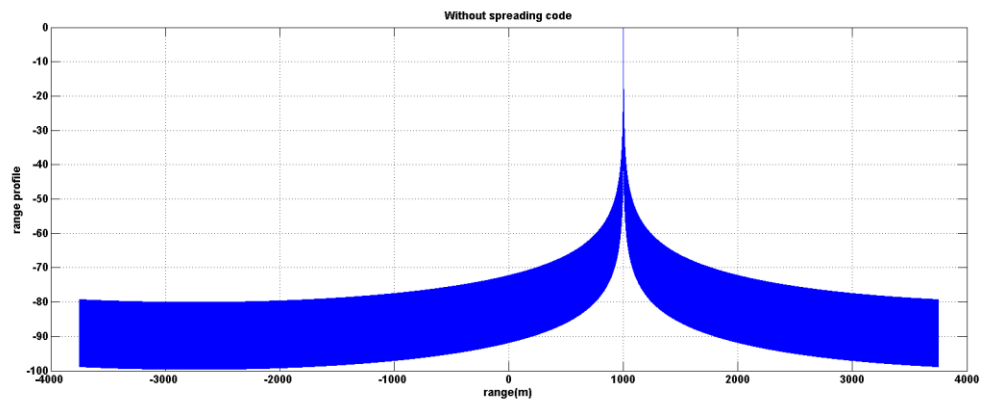


Figure app.1: Range profile for point target at 1000 metres without any spreading code applied.

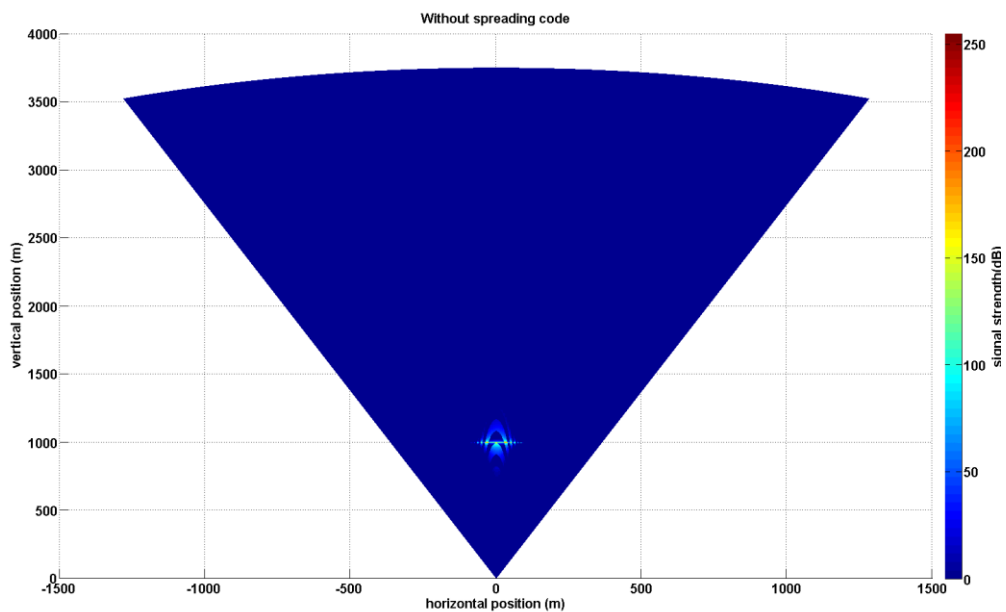


Figure app.2: Point target image (at 1000 metres range) of a MIMO radar without any spreading code applied.

The above two figures demonstrate the results of a point target image without any spreading code. Basically, there is no leakage between elements. As it can be seen, the

target mainlobe is around 70 dB higher than the sidelobe level, indicating very good image quality.

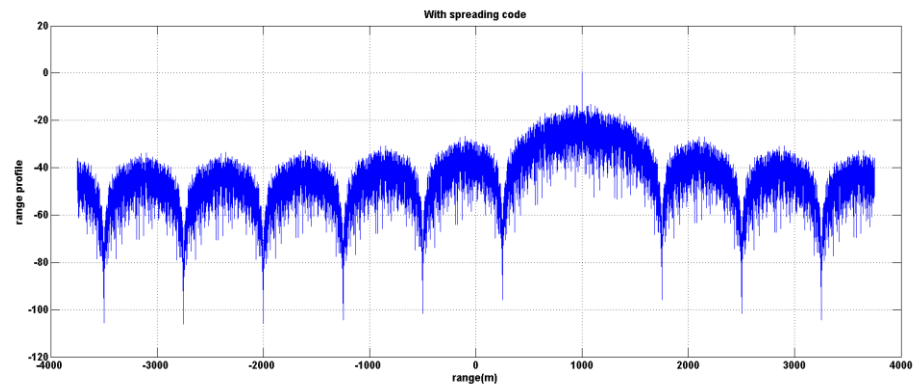


Figure app.3: Range profile for point target at 1000 metres after applying spreading code.

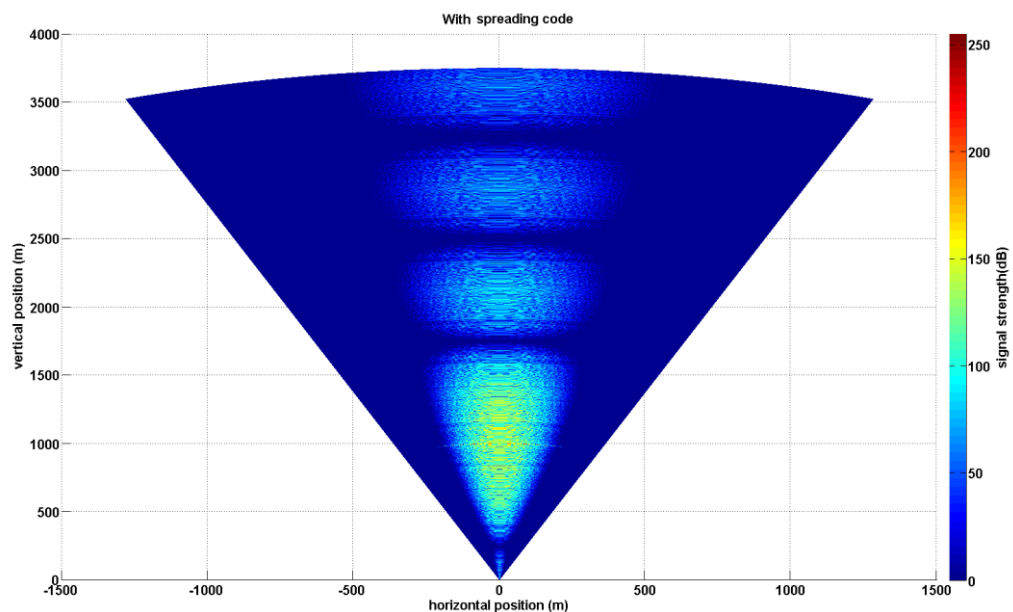


Figure app.4: Point target image (at 1000 metres range) of a MIMO radar after applying spreading code.

The above two figures show the effect of using spreading code signal leakage while all array elements are operating at once. It can be seen from the range profile that mainlobe is only around 20 dB higher than the sidelobe level and the sidelobe level throughout the range is quite visible on the resultant image. Even after the matched filtering, the leakage from the other elements attribute to increased sidelobe level. It should be noted that this simulation was only performed for a stationary point

target. In practical case, the target will be snow avalanche which would be moving and distributed. So, it was concluded that MIMO radar development with coded signal transmission may become a big risk for snow avalanche imaging as it takes huge amount of time and effort to build the hardware. The final decision was to resort to a conventional phased array radar development.

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