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FPGA Implementation of the Range-Doppler Algorithm for Real-Time Synthetic Aperture Radar Imaging

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Abstract: In this paper, we propose a range-Doppler algorithm (RDA)-based synthetic aperture radar (SAR) processor for real-time SAR imaging and present FPGA-based implementation results. The processing steps for the RDA include range compression, range cell migration correction (RCMC), and azimuth compression. A matched filtering unit (MFU) and an RCMC processing unit (RPU) are required for real-time processing. Therefore, the proposed RDA-based SAR processor contains an MFU that uses the mixed-radix multi-path delay commutator (MRMDC) FFT and an RPU. The MFU reduces the memory requirements by applying a decimation-in-frequency (DIF) FFT and decimation-in-time (DIT) IFFT. The RPU provides a variable tap size and variable interpolation kernel. In addition, the MFU and RPU are designed to enable parallel processing of four 32-bit which are transferred via a 128-bit AXI bus. The proposed RDA-based SAR processor was designed using Verilog-HDL and implemented in a Xilinx UltraScale+ MPSoC FPGA device. After comparing the execution time taken by the proposed SAR processor with that taken by an ARM cortex-A53 microprocessor, we observed a 85-fold speedup for a 2048 × 2048 pixel image. A performance evaluation based on related studies indicated that the proposed processor achieved an execution time that was approximately 6.5 times less than those of previous FPGA implementations of RDA processors.

Keywords: synthetic aperture radar (SAR); range-Doppler algorithm (RDA); real-time processing; matched filter; range cell migration correction (RCMC); field programmable gate array (FPGA)

1. Introduction

A synthetic aperture radar (SAR) is an active sensor system that operates in the microwave band. The primary strength of the SAR is that it can provide high-quality images independently of light and weather conditions [1–4]. Traditional SAR systems have been limited to being mounted on large platforms, such as satellites and aircraft, because of their high power consumption and the processing requirements of large datasets. However, recent advances in CMOS process technologies and signal processing have enabled the development of compact and lightweight SAR systems, and research on SAR systems mounted on small platforms, such as unmanned aerial vehicles (UAVs), is increasing [5–15].

SAR data processing is computationally intensive, and hardware accelerators, such as graphics processing units (GPUs) and field programmable gate arrays (FPGAs), are required for real-time processing [16–26]. Although GPUs have high processing capabilities, their large power consumption makes them unsuitable for small platforms. FPGAs have made significant progress in terms of high throughput, on-chip storage resources, arithmetic logic resources, and low power consumption [27,28]. Therefore, an FPGA-based SAR system is appropriate for a small platform with limited power. Several SAR imaging algorithms have been implemented, including the range-Doppler algorithm (RDA) [17–19], back-projection algorithm (BPA) [20–22], and polar format algorithm (PFA) [23–26]. Although these algorithms are effective, RDA is the most popular one because it is simple,
offers easy motion compensation, and provides a flexible tradeoff between accuracy and the number of computations [29]. RDA contains three primary stages: range compression, range cell migration correction (RCMC), and azimuth compression [30–32]. Range compression and azimuth compression are also known as range-matched filtering and azimuth-matched filtering, respectively, and they are both performed via multiplication in the frequency domain. Efficient transformation of signals to and from the frequency domain is achieved using the fast Fourier transform (FFT). Meanwhile, RCMC is realized via sinc interpolation [33,34]. Therefore, the FFT and interpolation processors are the two key components in the implementation of an RDA.

In the RDA, the azimuth resolution depends on the FFT length [35]. FFT processors for the RDA typically support a fixed length, making it challenging to apply this algorithm in various SAR applications. Therefore, a variable-length FFT processor is required [36]. Many FFT algorithms are available, including radix-2, radix-4, radix-8, radix-2², radix-2³, and mixed-radix. The mixed-radix algorithm can reduce the number of non-trivial multiplications better than the radix-2 or radix-4 algorithms can. Trivial multiplication can be simply implemented by using shifters and adders, and therefore, its complexity is much lower than that of non-trivial multiplication. Therefore, the mixed-radix algorithm can be implemented with lower complexity than the radix-2 or radix-4 algorithms can [37]. In addition, it can support more flexible FFT lengths than the radix-4 or radix-8 algorithms, and it has double the throughput of the radix-2³ algorithm. It is also suitable as an area-efficient variable-length FFT processor [38,39].

FFT hardware architectures can be roughly classified as single-butterfly, pipeline, and parallel architectures. In particular, the pipeline architecture offers an appropriate tradeoff between hardware complexity and throughput. Pipeline architectures are classified as single-path delay feedback (SDF) and multi-path delay commutator (MDC) architectures. The SDF architecture operates at a lower throughput than the MDC architecture because of its single path [40]. In a real-time SAR system, an FFT processor should provide high throughput rates. Therefore, the MDC architecture is more appropriate than the SDF architecture in such systems [41,42].

The tap size of the interpolation processor in the RDA depends on the quality of the RCMC. Although a longer tap provides higher accuracy, the computational complexity increases with the tap size. Analogously, a short tap size implies lower computational complexity but lower accuracy. Because the required accuracy and computational complexity vary between SAR applications, the tap size should ideally be variable so that the interpolation processor is suitable for various SAR applications. Interpolation processors generally use several types of windows, such as the Kaiser, Hamming, and Hanning windows, to reduce the sidelobes of the sinc kernel. Because different windowed sinc kernels offer different tradeoffs between sidelobes and resolution, an appropriate windowed sinc kernel must be selected for each SAR application [35]. In addition, interpolation operations must be accelerated through parallel processing because of their high computational complexity.

Many studies have been conducted to implement RDA using FPGAs. Araujo et al. used an Altera Cyclone E IV FPGA to implement the RDA with a mixed-radix SDF FFT processor, and RCMC was implemented using a data-shifting method. This FPGA acquired a 2048 × 2048 pixel image in 20.31 s at a speed of 130 MHz [17]. Hou et al. used a Xilinx Virtex-6 FPGA to implement the RDA with a radix-2 single-butterfly FFT processor. It acquired a 2048 × 4096 pixel image in 12.03 s at a speed of 200 MHz [18]. In addition, Hossain et al. used a Xilinx Virtex-6 FPGA to implement the RDA with a radix-2 SDF FFT processor. It acquired a 2048 × 900 pixel image in 2.08 s at a speed of 200 MHz [19].

In this study, we propose an RDA-based SAR processor containing a matched filtering unit (MFU) using a mixed-radix multi-path delay commutator (MRMDC) FFT processor and an RCMC processing unit (RPU). The MRMDC FFT processor supports variable length and offers a higher throughput and lower area. The MFU reduces the memory requirements by applying the decimation-in-frequency (DIF) FFT and decimation-in-time (DIT) IFFT so that the FFT processor’s output data are entered into the IFFT processor without reordering.
The RPU provides a 2/4/6/8/10/12/14/16 variable tap size, variable interpolation kernel suitable for various SAR applications, and parallel architecture. In summary, the main contribution of this study is the proposal of a high-speed and area-efficient hardware structure for designing an RDA-based SAR processor and the presentation of its implementation and experimental results.

The remainder of this paper is organized as follows. Section 2 reviews the RDA. Section 3 describes the hardware architecture of the proposed RDA-based SAR processor. Section 4 presents and discusses the implementation and verification results, and also compares the speed performance of the proposed processor with results from previous studies. Section 5 concludes the paper.

2. Range-Doppler Algorithm

A procedure of the RDA is shown in Figure 1. The RDA comprises three processing steps: range-matched filtering, RCMC, and azimuth-matched filtering. Range-matched filtering is performed on raw data, which are multiplied by the range reference signal after the range FFT. The range IFFT is then performed. The RDA obtains low-quality SAR images owing to relative motion between the radar and target. Therefore, to obtain high-quality SAR images, signals from the same scatter should be arranged in one range cell. This operation is called RCMC. Before proceeding with RCMC, however, the azimuth FFT must be performed. Finally, SAR image generation is complete once azimuth-matched filtering is performed via multiplication by the azimuth reference signal in the azimuth frequency domain, followed by the azimuth IFFT.

![Figure 1. Procedure of the range-Doppler algorithm.](image-url)

The transmission signal of a pulse-Doppler radar is assumed to be a linear frequency modulation (FM) chirp signal with an FM rate of \(K_r\). The received signal is demodulated into a baseband and can be described as

\[
s_0(\tau, \eta) = A_0w_r(\tau - R'(\eta))w_a(\eta - \eta_c)\exp\left\{-j2\pi f_0 R'(\eta) + j\pi K_r(\tau - R'(\eta))^2\right\} \tag{1}
\]

where \(A_0\) is the amplitude of the signal, \(R'(\eta)\) is equivalent to \(2R(\eta)/c\), \(c\) is the velocity of light, \(\tau\) is the range time, \(\eta\) is the azimuth time, \(\eta_c\) is the beam center crossing time, \(w_r(\tau)\) is the range envelope (a rectangular function), \(w_a(\eta)\) is the azimuth envelope (a sinc function), \(f_0\) is the radar center frequency, \(K_r\) is the range chirp FM rate, and \(R(\eta)\) is the instantaneous slant range. The discrete-time expression of Equation (1) is given by

\[
s_0(n, k) = A_0w_r(n - R'(k))w_a(k - k_c)\exp\left\{-j2\pi f_0 R'(k) + j\pi K_r(n - R'(k))^2\right\} \tag{2}
\]

where \(n\) is the range time index \((\tau = nT_s)\) and \(T_s\) is the sampling interval. \(k\) is the azimuth time index \((\eta = kT_p)\) and \(T_p\) is the pulse-repetition time. Let \(S_0(f_n, k)\) be the range Fourier transform of \(s_0(n, k)\) from Equation (2) and let \(G(f_n)\) be the frequency-domain range reference signal. The output of the range-matched filter can be described as

\[
s_{rc}(n, k) = \text{IFFT}_n(S_0(f_n, k)G(f_n)) = A_0p_r(n - R'(k))w_a(k - k_c)\exp\left\{-j\pi R(k)/\lambda\right\} \tag{3}
\]

where \(p_r\) is the range envelope (a sinc function), \(\lambda\) is the wavelength, \(\text{IFFT}_n\) is the range IFFT, and \(2R(k)/c\) is the target range migration incorporated via the azimuth time. The
range-matched filtered signal requires the azimuth FFT for RCMC and azimuth-matched filtering. In low-squint cases, the range equation can be approximated using Equation (4).

\[ R(k) = \sqrt{R_0^2 + V_r^2 k^2} \approx R_0 + \frac{V_r^2 k^2}{2R_0} \]  

(4)

where \( R_0 \) is the slant range of the closest approach, and \( V_r \) is the platform velocity. By substituting \( R(k) \) in the phase component of Equation (3) for the range-matched filtered signal, Equation (5) can be obtained as

\[ s_{rc}(n, k) \approx A_0 p_r \left\{ n - \frac{2R(k)}{c} \right\} w_a(k - k_c) \exp \left( -j \frac{4\pi f_0 R_0}{c} \right) \exp \left( -j \pi \frac{2V_r^2 k^2}{\lambda R_0} \right) \]  

(5)

Because the phase in Equation (5) is a function of \( k^2 \), the signal has linear FM characteristics in the azimuth direction, and the azimuth FM rate can be expressed as

\[ K_a \approx \frac{2V_r^2}{\lambda R_0} \]  

(6)

By substituting Equation (6) into Equation (5), the signal after the azimuth FFT can be rewritten as

\[ S_1(n, f_k) = A_0 p_r \left\{ n - \frac{2R_{rd}(f_k)}{c} \right\} W_a(f_k - f_{kc}) \exp \left( -j \frac{4\pi f_0 R_0}{c} \right) \exp \left( j\pi \frac{f_k^2 K_a}{f_k} \right) \]  

(7)

where \( W_a(f_k) \) is the frequency-domain version of the azimuth beam pattern \( w_a(k) \). The first phase term carries the inherent phase information of the target and is not important in an intensity image. The second phase term represents the azimuth modulation. The range cell migration (RCM) in the range time and azimuth frequency domain \( R_{rd}(f_k) \) can be expressed as

\[ R_{rd}(f_k) \approx R_0 + \frac{V_r^2}{2R_0} \left( \frac{f_k}{K_a} \right)^2 = R_0 + \frac{\lambda^2 R_0 f_k^2}{8V_r^2} \]  

(8)

RCMC is performed via sinc interpolation in the range direction. The sinc kernel is weighted by a tapering window, such as the Kaiser, Hamming, or Hanning windows. The second term in Equation (8) represents the amount of RCM to be corrected, expressed as

\[ \Delta R(f_k) = \frac{\lambda^2 R_0 f_k^2}{8V_r^2} \]  

(9)

The range envelope \( p_r \) is independent of \( f_k \), indicating that the RCM has been corrected. In addition, the energy is arranged at \( n = 2R_0/c \), which is the range of the closest approach. To perform multiplication by the azimuth reference signal, the signal after RCMC, i.e., \( S_2(n, f_k) \) in Equation (10), is multiplied by the frequency-domain azimuth reference signal \( H_{az}(f_k) \). The azimuth reference signal can be expressed as

\[ H_{az}(f_k) = \exp \left( -j\pi \frac{f_k^2}{K_a} \right) \]  

(11)
The frequency-domain azimuth reference signal can be obtained as a complex conjugate of the second phase term of Equation (10), where $K_a$ is a function of $R_0$. The resulting signal after azimuth reference multiplication can be expressed as

$$S_3(n, f_k) = S_2(n, f_k)H_{az}(f_k) = A_0 p_r \left( n - \frac{2R_0}{c} \right)W_a(f_k - f_{kc})\exp\left(-j \frac{4\pi f_0 R_0}{c}\right)$$  \hspace{1cm} (12)

Finally, the RDA values are obtained by performing an azimuth IFFT on Equation (12), expressed as

$$s_{ac}(n, k) = \text{IFFT}_k(S_3(n, f_k)) = A_0 p_r \left( n - \frac{2R_0}{c} \right)p_a(k)\exp\left(-j \frac{4\pi R_0}{\lambda} + j2\pi f_{kc}k\right)$$  \hspace{1cm} (13)

where $p_a$ is a sinc-like azimuth envelope. In Equation (13), the range envelope and azimuth envelope indicate that the target is now positioned at $n = 2R_0/c$ and $k = 0$.

### 3. Proposed Hardware Architecture

Figure 2 shows the hardware architecture of the proposed RDA-based SAR processor, which contains an MFU in order to perform the range-matched filtering, azimuth-matched filtering, and azimuth FFT, as well as an RPU to correct the RCM. The MFU and RPU comprise master and slave interfaces for communicating with the double-data-rate (DDR) memory controller and microprocessor, respectively, a register for changing the operation mode of each unit, and a cache RAM for temporarily storing input/output data. In our design, the master interfaces are connected to the DDR memory controller via a 128-bit AXI bus. Thus, we can transmit four 32-bit data per clock cycle. Therefore, the MFU and RPU can efficiently perform parallel operations on the four 32-bit data. To achieve this, the FFT included in the MFU was designed with an MRMDC architecture, and the RPU was designed with an architecture comprising four interpolation modules.

![Figure 2. Block diagram of the proposed RDA-based SAR processor.](image)

#### 3.1. Matched Filtering Unit (MFU)

The proposed MFU consisted of a DIF MRMDC FFT module, DIT MRMDC IFFT module, and reference signal RAM, as shown in Figure 3. The reference signal RAM stored reference signals in the frequency domain for range- and azimuth-matched filtering. The DIF MRMDC FFT and DIT MRMDC IFFT modules employed a mixed-radix FFT algorithm to support flexible FFT lengths and to reduce the number of non-trivial multipliers, thereby facilitating a low-area implementation. The FFT modules employed the MDC architecture to enable high-throughput processing of four 32-bit data per clock cycle. In particular, the FFT and IFFT modules applied the DIF and DIT algorithms, respectively, so that the FFT module output could be used as an input to the IFFT module without reordering. This reduced the memory requirements and reordering time by eliminating the need for a...
reordering buffer. In addition, the MFU was designed to enable the azimuth FFT operation before RCMC by allowing it to output to the DIF MRMDC FFT module according to the slave register setting. In such cases, FFT reordering was performed while writing to the cache RAM.

Figure 3. Block diagram of the proposed matched filtering unit.

3.2. Range Cell Migration Correction Processing Unit (RPU)

The proposed RPU consisted of registers, coefficient RAM, and interpolation modules, as shown in Figure 4. Each register stored 32 bits of data, which were shifted to the next register at every clock cycle. The coefficient RAM was wired to the slave interface, which was designed to allow the microprocessor to store various kernel coefficients. The interpolation modules performed a dot product with the kernel coefficients by using the input data stored in the registers. The RPU was designed to have four interpolation modules to facilitate parallel computation of the data at each clock cycle. In addition, the desired tap could be set through the slave interface, and multiplexers were located between registers to allow the tap size of interpolation to be changed to 4/6/8/10/12/14/16 taps.

Figure 4. Block diagram of the proposed range cell migration correction processing unit.

4. Implementation and Acceleration Results

The MFU and RPU included in the proposed RDA-based SAR processor were designed using the Verilog hardware description language (HDL) and were implemented on a Xilinx Zynq Ultrascale+ FPGA device. Thus, the MFU was implemented with 8836 CLBs, 48,077 CLB LUTs, eight block RAMs, and 112 DSPs, whereas the RPU was implemented with 914 CLBs, 3465 CLB LUTs, four block RAMs, and 256 DSPs, as listed in Table 1. The MFU and RPU could process at maximum operating frequencies of 314 and 312 MHz, respectively. The power consumption of the proposed RDA-based SAR processor was
measured to be 1.31 W. Figure 5 shows the verification environment used for the FPGA platform.

Table 1. Implementation results based on the Xilinx Zynq Ultrascale+ FPGA device.

| Unit | CLB | CLB LUT | Block RAMs | DSP | Max. Operating Clock Freq. |
|------|-----|---------|------------|-----|---------------------------|
| MFU  | 8836| 48,077  | 8          | 112 | 314 MHz                   |
| RPU  | 914 | 3465    | 4          | 256 | 312 MHz                   |

Figure 5. Verification environment for the proposed FPGA implementation.

The raw data used for verifying our RDA-based SAR processor were loaded into the DDR memory. Subsequently, when the start signal of the MFU was entered through the microprocessor, the raw data in the DDR memory were transferred to the cache RAM of the MFU through the master interface. The MFU then performed range compression. Next, the results of the MFU were output to the DDR memory through the master interface. In addition, to perform the azimuth FFT, the MFU was set to the FFT mode using the microprocessor. Subsequently, when the start signal of the FFT-mode MFU was entered through the microprocessor, the range-compressed data in the DDR memory were transferred to the cache RAM of the FFT-mode MFU. After completing the FFT operation, the data stored in the cache RAM of the FFT-mode MFU were transferred back to the DDR memory. RCMC and azimuth compression were performed similarly.

Figure 6 presents the point-target-based verification results for the RDA’s intermediate step. The measured SAR image quality parameters for the peak-to-sidelobe ratio (PSLR) were \(-13.04\) dB (range) and \(-13.37\) dB (azimuth). Regarding the integrated sidelobe ratio (ISLR), the values of \(-10.24\) dB (range) and \(-10.54\) dB (azimuth) were measured. In addition, we used a RADARSAT-1 dataset that was collected on 16 June 2002 to verify the proposed hardware using actual SAR data. More specifically, we used an image of Vancouver, Canada from RADARSAT-1’s Fine Beam 2 [35]. The results of ARM Cortex-A53-based software processing were used as a reference to evaluate the resulting image quality of the proposed hardware. The peak signal-to-noise ratio (PSNR) of the actual SAR data
processed by the proposed hardware was 55.8 dB. Wei Di et al. achieved a PSNR of 41.8 dB [43]. Yohei Sugimoto et al. measured a PSNR of 45.2 dB [44]. By comparing the PSNR quality parameter, it was verified that the proposed RDA-based SAR processor presented better performance than those of [43,44]. Figure 7 illustrates the SAR image obtained after processing the actual SAR data.

**Figure 6.** Point target simulation results: (a) raw data; (b) range FFT results; (c) range compression results; (d) azimuth FFT results; (e) RCMC results; (f) azimuth compression results (point target results).

Table 2 presents the evaluation results in terms of RDA execution time. To evaluate the speed performance of the designed MFU and RPU, we measured the execution times of the different sub-operations of the RDA in the ARM Cortex-A53-based software implementation. The RCMC operation was accelerated by the RPU, and the matched filtering and FFT operations were accelerated by the MFU. We calculated the acceleration in the execution times for the RCMC, matched filtering, and FFT operations. The experimental results indicate that the calculation time decreased from approximately 70.33 s to approximately 0.823 s for a 2048 × 2048 pixel image, resulting in an 85-fold acceleration. Considering the sampling rate of 60 MHz for the raw data of the size of 2048 × 2048, the time taken to load the data into the DDR memory was expected to be around 0.069 s. Since the execution time of the proposed SAR processor was 0.823 s, the total processing time was 0.892 s, and the SAR images were expected to be generated at the rate of 1.1 Hz. If the proposed SAR processor is implemented with a high-end FPGA or very-large-scale integrated circuit (VLSI), it is expected that the imaging speed will be further improved.
Figure 7. Actual SAR images derived using the (a) ARM Cortex-A53-based software and (b) the proposed FPGA-based hardware.

Table 2. RDA execution time.

| Image Size | Execution Time (s) | Full SW | RCMC Accel. | MF/FFT Accel. | Full Accel. |
|------------|--------------------|---------|-------------|---------------|-------------|
| 256 × 256  | 0.805              | 0.691   | 0.122       | 0.007         |
| 512 × 512  | 3.648              | 3.114   | 0.574       | 0.030         |
| 1024 × 1024| 16.114             | 13.850  | 2.420       | 0.188         |
| 2048 × 2048| 70.330             | 61.276  | 9.951       | 0.823         |

Table 3 presents the results of the comparison in terms of the normalized execution time for the proposed RDA-based SAR processor and the RDA-based SAR processors presented in References [17–19]. Making an accurate comparison between the proposed RDA-based SAR processor and those from related works is problematic because they were implemented in different FPGA devices. However, to make the fairest comparison possible, we compared the speed performance in terms of the normalized execution time, which was calculated based on the FPGA process technology and image size. The normalized execution time used for comparison can be expressed as

$$T_{norm} = \frac{16 \text{ nm}/\text{Technology}}{\text{Image Size}} \times \text{Execution Time} \quad (14)$$

where “Technology” denotes the CMOS process technology expressed in nanometers. As this value increases, the operation speed decreases, thereby increasing the execution time. With an increase in image size, which is the size of the input data, the execution time increases. According to the results of the comparison, the proposed RDA-based SAR processor offers the fastest execution time compared with those from the related studies.
Table 3. Comparison with previous implementations.

| Work          | Platform         | Image Size | Hardware Resources                      | RCMC | Operating Freq. (MHz) | Exec. Time (s) | Tech. (nm) | $T_{norm}$ (µs) |
|---------------|------------------|------------|----------------------------------------|------|-----------------------|----------------|------------|-----------------|
| Proposed      | Zynq UltraScale+ | 2048 × 2048| CLB: 9750, CLB LUT: 51,542 Block RAMs: 12 DSP: 368 | O    | 300                   | 0.823          | 16         | 0.2             |
| [17] Intel Altera Cyclone E IV | 2048 × 2048 | Logic Elements: 65,209 Logic Registers: 35,069 Memory (kbit): 814 Embedded Multipliers: 297 | O    | 500                   | 20.31          | 60         | 1.3             |
| [18] Xilinx Virtex-6 | 2048 × 4096 | N/A        | X                                      | 130  | 12.03                 | 40             | 0.57       |                 |
| [19] Xilinx Virtex-6 | 900 × 2048 | N/A        | O                                      | 200  | 2.08                  | 40             | 0.45       |                 |

5. Conclusions

In this study, we proposed a design for an RDA-based SAR processor architecture consisting of an MFU and an RPU. The FFT module included in the MFU was designed to support a variable FFT length and reduce the number of non-trivial multipliers by employing a mixed-radix algorithm. The FFT module used an MDC pipeline architecture to enable the high-throughput processing of four 32-bit data per clock cycle. In addition, the MFU reduced the memory requirements due to the DIT FFT and DIF IFFT modules. The RCMC processor provided a variable tap size and variable interpolation kernel and was able to process the four 32-bit data in parallel. The proposed processor was implemented with 9750 CLBs, 51,542 CLB LUTs, 12 block RAMs, and 368 DSPs on a Xilinx Zynq UltraScale+ FPGA device. For an image with 2048 × 2048 pixels, we achieved an approximately 85-fold acceleration compared with the available software. We computed the normalized execution time and compared the results with those from related studies. The proposed RDA-based SAR processor exhibited the fastest normalized execution time compared with the RDA-based SAR processors from previous studies, despite supporting the RCMC operation.

Future work will involve a focus on other SAR imaging algorithms, such as BPA or PFA. We will then conduct research on the design and implementation of the hardware architecture for such algorithms. Finally, we will attempt to develop an integrated processor that can compute the RDA, BPA, and PFA to flexibly exploit the advantages of each algorithm.

Author Contributions: Y.C. designed the MFU and RCU, performed the simulation and experiment, and wrote the paper. D.J., M.L., and W.L. implemented the processor and performed the revision of this manuscript. Y.J. conceived of and led the research, analyzed the experimental results, and wrote the paper. All authors read and agreed to the published version of the manuscript.

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