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A Simplified Rectification Method for Efficient Hardware Implementation

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Abstract

Rectification is used as a pre-processing step in all stereo matching algorithms. The rectification algorithms are hard to implement in hardware circuit because of the complexity of its mathematical calculations. To solve this problem, this research proposes an efficient calculation method and rectification hardware architecture optimized for stereo matching. The proposed calculation method optimizes previous rectification calculations methods using the linearity of pixel unit calculation and approximation of coordinate translation. The experimental results show that the proposed rectification method has marginal differences in terms of the rectification results compared with the previous rectification method while consuming relatively less hardware resources. The proposed rectification hardware architecture can be implemented in all stereo vision systems and can be applied in various applications such as autonomous vehicles, intelligent robots, and 3D multimedia contents.

Keywords: Stereo vision; stereo matching; rectification; hardware implementation

1. Introduction

A stereo vision system calculates the 3D distances by finding the disparity of objects in stereo image pairs captured by a stereo camera. The 3D distance information of a stereo vision system can be applied to various applications such as autonomous vehicles, intelligent robots, and 3D multimedia contents [1-5]. In the field of multimedia, 3D contents generation by stereo vision is also an active research area. For efficient 3D multimedia content creation, efficient real-time performance is required. In order to adapt stereo vision systems for 3D multimedia systems, there has been an abundance of research on hardware implementations of stereo vision systems proposed for real-time performance [6-8]

In general, stereo vision systems for hardware implementation go through pre-process, matching process, and post-process. In the pre-process step, input images from the stereo camera are processed to be easily used the disparity extraction. The matching process step extracts the disparity map by finding correspondence of the same object. The post-process step refines the disparity map by post-correcting the disparity map extracted by the matching process step. The matching process of the stereo vision system includes various algorithms such as the local matching method, the global matching method, and the SGM (semi-global matching) method, and these various matching algorithms [9] must go through the pre-process step known as rectification. The

rectification reduces the disparity search range, which is used to discover the disparity in the matching process, to 1D. Thus, rectification is commonly required for all matching algorithms to reduce calculation overhead [10]. Because rectification requires complex matrix calculations such as coordinate translation and image rotation, it causes difficulties in hardware implementation. Thus, this research proposes an optimized rectification calculation method that reduces use of hardware resources by optimizing previous calculation methods.

2. Conventional Rectification Method

The rectification in stereo vision systems is the process of aligning the epipolar line of the epipolar geometry of the stereo camera. By aligning the epipolar line of the epipolar geometry, the height differences between the stereo images taken by the stereo camera are removed. So, the disparity search range for matching is reduced to 1D. This process is shown in Figure 1. As shown in Figure 1-(a), to find point (W) in an image, the entire other image must be scanned, but in the case of rectified images, because there is no height difference between the stereo images as shown in (b), the disparity search range to find the correspondence is reduced to one line. This rectification is a critical process in real-time stereo vision systems that extract dense disparity maps.

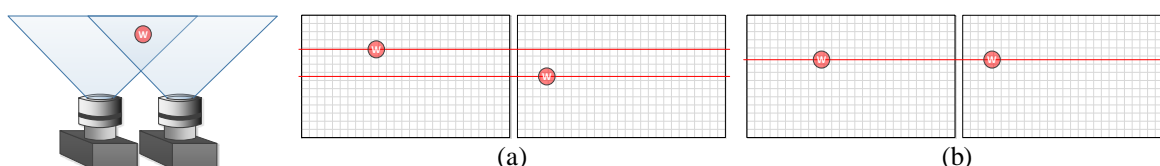


Figure 1. Stereo image rectification. (a) Pre-rectification. (b) Post-rectification.

In general the process of rectification with the distortion correction of the lens omitted is as shown in Figure 2. A detailed explanation of this is as follows.

1. Assume a 2D pixel coordinates of rectified image.
2. The 2D pixel coordinates of the rectified image are translated to 3D camera coordinates.
3. The translated coordinates are rotated to the 3D camera coordinates, or in other words the 3D camera coordinates before rectification.
4. The rotated coordinates are projected to 3D homogeneous coordinates of an unrectified image.
5. The projected coordinates are translated to 2D pixel coordinates of the unrectified image.
6. The pixel values of the calculated coordinates are mapped to the coordinates of the originally assumed rectified image using linear interpolation.

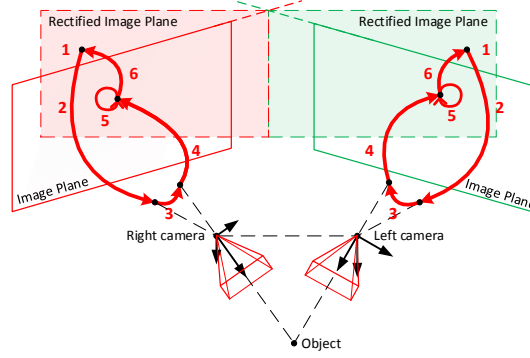


Figure 2. Rectification process.

The formulaic expression of this process is defined as

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = A \cdot R_{rect}^T \cdot A_{rect}^{-1} \cdot \begin{bmatrix} x_{rect} \\ y_{rect} \\ 1 \end{bmatrix}, \quad (1)$$

where $[x \ y \ z]^T$ is the 3D homogeneous coordinates of the unrectified image, and $[x_{rect} \ y_{rect} \ 1]^T$ is the 3D homogeneous coordinates of the rectified image. A is the intrinsic parameter matrix of the camera, A_{rect} is the intrinsic parameter matrix of the rectified image and R_{rect} is the rectification rotation matrix. These parameter matrices are 3×3 matrices that can be calculated using the process of stereo camera calibration [11].

3. Proposed Rectification Method

In general, the process of rectification, as described above, includes complex matrix calculations. Despite the pre-processing step, these complex calculations require significant hardware resources. To overcome the problem, this research proposes a novel rectification method which requires less hardware resources for hardware implementation by optimizing this general rectification calculation method.

A , A_{rect} , and R_{rect} , which are the parameters of Equation(1), are the unique 3×3 matrix parameters of the stereo camera and these parameters are an inherent property of the stereo camera calibration. Therefore $A \cdot R_{rect}^T \cdot A_{rect}^{-1}$ of Equation(1) can be transposed to one coefficient matrix as

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix} \equiv A \cdot R_{rect}^T \cdot A_{rect}^{-1}, \quad (2)$$

and Equation(1) can be rewritten as

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix} \cdot \begin{bmatrix} x_{rect} \\ y_{rect} \\ 1 \end{bmatrix} = \begin{bmatrix} c_{11} \cdot x_{rect} + c_{21} \cdot y_{rect} + c_{13} \\ c_{21} \cdot x_{rect} + c_{22} \cdot y_{rect} + c_{23} \\ c_{31} \cdot x_{rect} + c_{32} \cdot y_{rect} + c_{33} \end{bmatrix}. \quad (3)$$

Through the process of Equation(3), the 3D homogeneous coordinates of the unrectified image can be calculated from the 2D pixel coordinates of the rectified image. By dividing both the x and y axis by the z axis of the 3D homogeneous coordinates, the 2D pixel coordinates of the camera image can be calculated. As the commonly used stereo camera sends stereo images in pixel units, Equation (3) is further simplified by using accumulation.

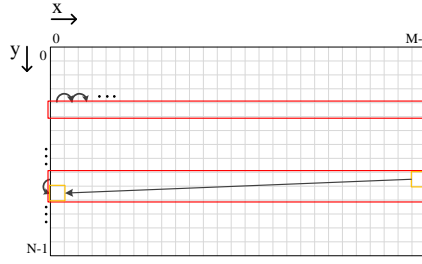


Figure 3. Rules of pixel unit calculations.

As shown in Figure 3, when processing resolution $(M \times N)$ images in pixel units, 2 types of processing operation are required. When the position of a pixel increases in the x -axis direction, the position of y -axis is fixed and the position of x -axis is increased from 0 to $M - 1$, but when the position of a pixel increases in the y -axis direction, the position of x -axis decreases from $M - 1$ to 0, and the position of y -axis increases from 0 to $N - 1$. By applying this uniformity of pixel unit calculation from Equation(3), the $[x \ y \ z]^T$ of the 3D homogeneous coordinates of the unrectified image can be optimized as an accumulation according to the increase in the x -axis and the increase in the y -axis. Therefore, Equation(3) can be rewritten as

$$\begin{bmatrix} x_n \\ y_n \\ z_n \end{bmatrix} = \begin{bmatrix} x_{n-1} \\ y_{n-1} \\ z_{n-1} \end{bmatrix} + \begin{cases} \begin{bmatrix} c_{11} \\ c_{21} \\ c_{31} \end{bmatrix}, & \text{when increase column,} \\ \begin{bmatrix} c_{12} - (M - 1) \cdot c_{11} \\ c_{22} - (M - 1) \cdot c_{21} \\ c_{32} - (M - 1) \cdot c_{31} \end{bmatrix}, & \text{when increase row.} \end{cases} \quad \left(\begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix} = \begin{bmatrix} c_{13} \\ c_{23} \\ c_{33} \end{bmatrix} \right) (4)$$

When calculating along the x -axis, because there is no change in rows, the only position of the x -axis is variable and $[x \ y \ z]^T$ can be each expressed as linear equations that contain gradient $[c_{11} \ c_{21} \ c_{31}]^T$. When calculating along the y -axis, the change in the x -axis point is always from $N - 1$ to 0 and the position of the y -axis increases from 0 to $M - 1$. Therefore when calculating along the y -axis direction, the gradient of the x -axis is constant, and the position of y -axis can be expressed as a linear equation with the gradient $[c_{12} - (M - 1) \cdot c_{11} \ c_{22} - (M - 1) \cdot c_{21} \ c_{32} - (M - 1) \cdot c_{31}]^T$. Therefore a complex matrix operation like Equation (1) can be optimized using the simple accumulation of the constant coefficient values according to the increase in rows and columns from the initial values as shown in Equation(4). The initial values $[x_0 \ y_0 \ z_0]^T$ in the Equation(4) are the 3D homogeneous coordinates of the unrectified image that correspond to the (0,0) in the rectified image coordinates, where $[x_n \ y_n \ z_n]^T$ is the position of the 3D homogeneous coordinates of the unrectified image corresponding to the currently calculated position of the rectified image coordinates and $[x_{n-1} \ y_{n-1} \ z_{n-1}]^T$ is the position of the 3D homogeneous coordinates of the unrectified image corresponding to the previously calculated position of the rectified image coordinates.

$[x_n \ y_n \ z_n]^T$ is the 3D homogeneous coordinates of the unrectified image. So these coordinates must be translated into 2D pixel coordinates. In order to do this translation, the x -axis and y -axis are each divided by the z -axis. From Equation (3), $z = c_{31} \cdot x_{rect} + c_{32} \cdot y_{rect} + c_{33}$. The variables that influence the value of z are the 2D pixel coordinates of the rectified image and the 3rd row vector $[c_{31} \ c_{32} \ c_{33}]^T$ of the coefficient matrix of Equation(2). The 3rd row vector of the coefficient matrix of Equation(2) can be rewritten as

$$[c_{31} \ c_{32} \ c_{33}] = \left[r_{31}/f_x \quad r_{32}/f_y \quad r_{33} \right] \cong [0 \ 0 \ 1]. (5)$$

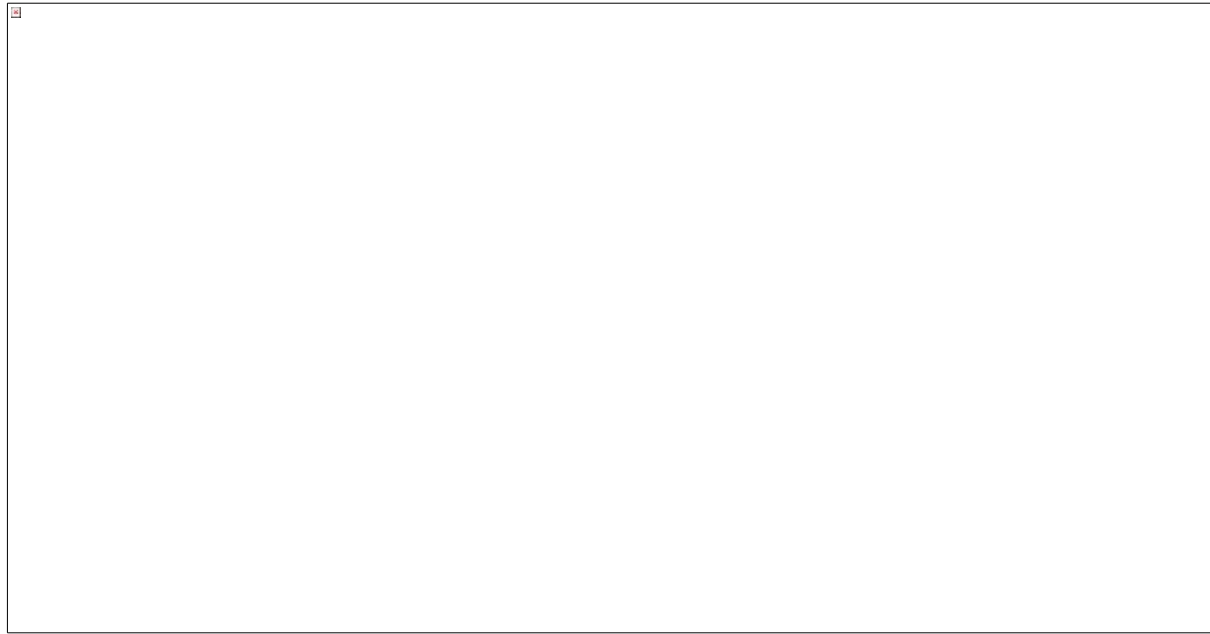
$$(f_x \gg r_{31}, f_y \gg r_{32}, r_{33} \approx 1)$$

where f_x and f_y are the focal lengths of the camera, and $[r_{31} \ r_{32} \ r_{33}]$ is the 3rd column of the rectification rotation matrix. According to the characteristics of the camera intrinsic parameter and rectification rotation matrix, z can be simplified as 1 as shown in Equation (5). First, because the rectification rotation matrix takes the form of a symmetric matrix, the diagonal elements are mostly 1 and the rest of the elements are smaller than 1. Because the focal length is a pixel unit coefficient that is the intrinsic parameter of camera, it is much larger than 1. Therefore the z value from Equation(3) can be considered as 1 as in Equation (5), which means that when translating the 3D homogeneous coordinates to 2D pixel coordinates in the process of Equation (4), the division by z can be omitted. Therefore, the process of Equation(4) can be simplified as

$$\begin{bmatrix} x_n \\ y_n \end{bmatrix} = \begin{bmatrix} x_{n-1} \\ y_{n-1} \end{bmatrix} + \begin{cases} \begin{bmatrix} c_{11} \\ c_{21} \end{bmatrix}, \text{when increase column,} \\ \begin{bmatrix} c_{12} - (M - 1) \cdot c_{11} \\ c_{22} - (M - 1) \cdot c_{21} \end{bmatrix}, \text{when increase row.} \end{cases} \left(\begin{bmatrix} x_0 \\ y_0 \end{bmatrix} = \begin{bmatrix} c_{13} \\ c_{23} \end{bmatrix} \right) \quad (6)$$

4. Proposed Rectification Hardware Architecture

Previously proposed rectification hardware architectures are broadly divided into two types. The first is the LUT-based method and the second is the computational logic-based method. Because the LUT-based method must store all the correspondence of coordinate translation for the rectification images, this method requires a significant amount of memory. Also, as the resolution of the image increases, the amount of memory also increased proportionally. On the other hand, the computational logic-based method uses a minimum amount of memory and has the advantage of being able to calculate the correspondence coordinates for the rectification images. However, the computational logic-based method rectification has a complex logic due to complex matrix operations. Therefore in this research, the complex operations of rectification are optimized using the linearity of pixel unit calculation and approximation of coordinate translation and a novel hardware architecture is proposed which uses a proposed rectification calculation method. Figure 4 shows both the optimized rectification hardware architecture and also the conventional rectification hardware architecture.



(a)

(b)

Figure 4. Rectification hardware architecture. (a) Conventional rectification. (b) Proposed rectification.

The proposed rectification optimizes the process of calculating the correspondence coordinates of the camera image. As shown in Figure 4, when compared to conventional rectification hardware architecture, the multipliers and the dividers from the correspondence coordinates calculating part have been removed.

5. Experimental Results

The proposed rectification hardware architecture was designed using Verilog HDL and implemented using Xilinx FPGA Virtex-7. Figure 5 is a comparison of the results of the conventional rectification and the proposed rectification, where Table 1 is the pixel coordinate error of the proposed rectification based on the results of the conventional rectification. As shown in Table 1, the average pixel coordinate error is under 1 pixel, which is an error that occurs due to the omitted process, division by z when, translating the 3D homogeneous coordinates of the unrectified image to 2D pixel coordinates. However, as shown in Figure 5, this error is marginal and the proposed rectification also finely aligns with the epipolar line.

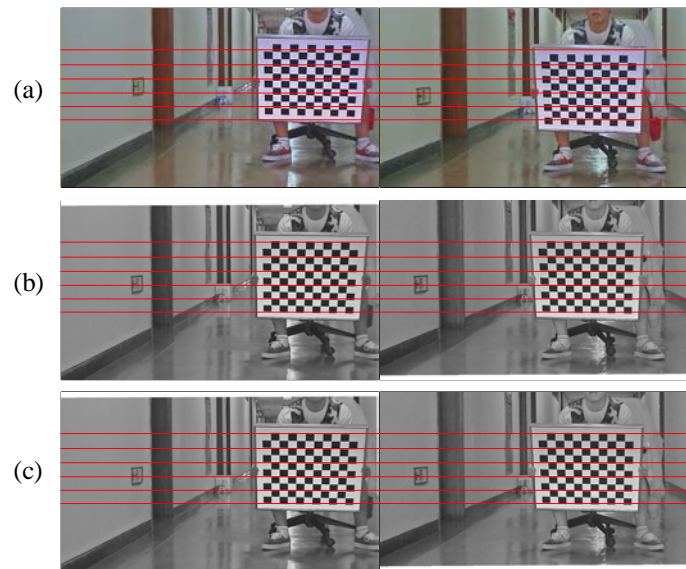


Figure 5. Rectification results. (a) Original left and right images. (b) Conventional rectified images. (c) Proposed rectified images.

Table 1. Number of pixel coordinate errors of the proposed rectification.

Stereo Images	Average number of error pixels		Max number of error pixels	
	x-axis	y-axis	x-axis	y-axis
Left image	0.4465	0.2255	2	2
Right image	0.5028	0.2880	3	2

The hardware resources consumed when the conventional rectification and proposed rectification were implemented using FPGA are shown in Table 2. The proposed rectification hardware architecture uses less than 1/3 of the hardware resources compared with that of conventional rectification hardware architecture because of the simplified rectification calculation.

Table 2. Synthesis results of the FPGA hardware implementations.

Method	Number of Slice Registers	Number of Slice LUTs
Conventional rectification	12837	9508
Proposed rectification	593	3718

6. Conclusions

For real-time performance of stereo vision which is used for 3D multimedia system, this research proposed a novel rectification method by optimizing previous rectification calculations methods using the linearity of pixel unit calculation and approximation of coordinate translation and also proposed rectification hardware architecture using the proposed rectification calculation method. The experimental results show that when comparing the proposed rectification to the conventional rectification there were marginal differences in the results while reducing the use of hardware resources by a significant amount. Because the rectification is necessary for stereo matching algorithms, to adapt to the stereo vision based real-time 3D multimedia system, it must be considered when designing the rectification hardware architecture for real-time performance and hardware resource minimization. The proposed rectification hardware architecture shows marginal errors and uses less than 1/3 hardware resources with compared with conventional rectification hardware architecture. So, the proposed rectification method can be efficiently applied to real-time multimedia system based on stereo vision.

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