

Measurement of the Cam Spacing on Camshaft by Binocular Vision

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Abstract

This paper proposed a measurement method of the cam spacing of camshaft based on binocular vision. Firstly, internal and external parameters of the camera were calibrated by the improved calibration. Secondly, the objective function is established by the distance of pixel coordinates and their corresponding epipolar lines, and the fundamental matrix is obtained by optimizing the objective function. Finally, the translation vector and the rotation matrix are determined by the singular value decomposition. Using the results of calibration, a method of measuring cam spacing of camshaft is presented. In the experiment condition, cam spacing of camshaft is measured. Experiment result shows that accuracy of the proposed method could meet the requirements of monitoring the cam spacing of combined type camshaft.

1. Introduction

Camshaft is one of the important components in engine. Modern technology mostly adopted a combination of hollow structure. Process requires the precision is smaller than $\pm 0.3\text{mm}$ [1-2]. Because there is a phase difference of every cam along the circumference of the cam, the distance between two cams is a 3-dimensional size.

In the past, many researchers have developed algorithms for the 3-dimensional reconstruction by binocular vision. K. Zhang [3] explores a model of the binocular vision system focused on 3D reconstruction and describes an improved genetic algorithm aimed at estimating camera system parameters. In order to enhance the calibration accuracy, many corners should be treated as feature points. W. Sun [4] presents a study investigating the effects of training data quantity, pixel coordinate noise on binocular vision accuracy. H. H. Cui [5] discusses an improved method for an accurate 3-dimensional measurement. The system accuracy is improved considering the nonlinear measurement error. An accurate phase-height mapping algorithm is proposed by Z. W. Li [6] to improve the performance of the structured light system with digital fringe projection. By means of a training network, the relationship between the 2D image coordinates and the 3D object coordinates can be achieved. However, their experiments involve fixed system structure parameters and provide a synthetic evaluation on flexible binocular system parameters to verify the accuracy results.

This article proposed a measurement method of can spacing by binocular vision. This work can meet the requirement of practical measurement, and realize the automatic detection.

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1.1. Binocular Vision Model

Binocular vision model could be expressed based on the perspective transformation model, which is shown in Figure 1.

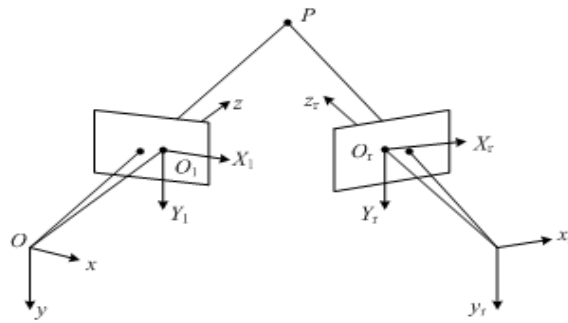


Figure 1. Binocular Vision Model

$$s_l \begin{bmatrix} X_l \\ Y_l \\ 1 \end{bmatrix} = \begin{bmatrix} f_l & 0 & 0 \\ 0 & f_l & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad (1)$$

$$s_r \begin{bmatrix} X_r \\ Y_r \\ 1 \end{bmatrix} = \begin{bmatrix} f_r & 0 & 0 \\ 0 & f_r & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_r \\ y_r \\ z_r \end{bmatrix} \quad (2)$$

Where, $oxyz$ is the world coordinate system (coincide with the left camera coordinate system), $o_r x_r y_r z_r$ is the right camera coordinate system, $O_l X_l Y_l$, $O_r X_r Y_r$ are the left and right cameras' image coordinate systems. f_l , f_r are effective focal lengths of the left and right cameras. P is the measured point in the public view. Relationship between the right camera coordinate system and the the world coordinate system can be expressed as:

$$\begin{bmatrix} x_r \\ y_r \\ z_r \\ 1 \end{bmatrix} = M_{lr} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} r_1 & r_2 & r_3 & t_x \\ r_4 & r_5 & r_6 & t_y \\ r_7 & r_8 & r_9 & t_z \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \quad (3)$$

Where, $M_{lr} = [R \quad T]$, $R = \begin{bmatrix} r_1 & r_2 & r_3 \\ r_4 & r_5 & r_6 \\ r_7 & r_8 & r_9 \end{bmatrix}$, $T = \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix}$ are the ration matrix and translation vector between two coordinate systems.

From Eq. (1), (2) and (3), coordinates of P in the 3 dimensional world coordinate system can be obtained:

$$\left\{ \begin{array}{l} x = z X_1 / f_1 \\ y = z Y_1 / f_1 \\ z = \frac{f_1(f_r t_x - X_r t_z)}{X_r(r_7 X_1 + r_8 Y_1 + f_1 r_9) - f_r(r_1 X_1 + r_2 Y_1 + f_1 r_3)} \\ = \frac{f_1(f_r t_y - Y_r t_z)}{Y_r(r_7 X_1 + r_8 Y_1 + f_1 r_9) - f_r(r_4 X_1 + r_5 Y_1 + f_1 r_6)} \end{array} \right. \quad (4)$$

2. Key Techniques of Binocular Vision

From Eq.(4), some work has to be carried out in the visual measurement:(1) the calibration of effective focal length f_l 、 f_r ; (2) relationship between two camera coordinate system; (3) matching of corresponding points in the left and right image.

2.1. Calibration of Effective Focal Length

The calibration of CCD camera can be divided to the calculating of interior camera parameters and exterior camera parameters, which directly affect the accuracy of measurement. A large number of research work have been carried out by scientists [7-8]. An improved method [9] is used to calibrate the effective focal length.

2.2. Matching of Corresponding Points and Fundamental Matrix

Polar geometric constraint is one of the most important constraints in the matching of corresponding points in the left and right images, which can be expressed by fundamental matrix. The ration matrix and translation vector between two cameras can be obtained by the decomposing of fundamental matrix. As a result, the calibration of fundamental matrix is the most important technology in the binocular vision measurement.

The fundamental matrix can be written as [10]:

$$\mathbf{F} = \mathbf{A}_r^{-T} \mathbf{S} \mathbf{R} \mathbf{A}_l^{-1} \quad (5)$$

\mathbf{A}_l 、 \mathbf{A}_r are the interior camera parameters of the left and right cameras. \mathbf{S} is anti-symmetric matrix, which can be defined by the translation vector between two cameras:

$$\mathbf{S} = [\mathbf{t}]_x = \begin{bmatrix} 0 & -t_z & t_y \\ t_z & 0 & -t_x \\ -t_y & t_x & 0 \end{bmatrix} \quad (6)$$

Let $\mathbf{E} = \mathbf{S} \mathbf{R} = \mathbf{A}_r^T \mathbf{F} \mathbf{A}_l$ is the essential matrix. The ration matrix and translation vector between two cameras can be calculated by singular value decomposition of \mathbf{E} .

$$\mathbf{R} = \mathbf{U} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{V} \quad (7)$$

$$T = kU \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad (8)$$

2.3. Calculation of Fundamental Matrix

There were a lot of methods to calculate fundamental matrix. Armanque X divided these methods into three types: linear method, iterative method and robust method. 8 points method is the most representative method in linear method. It is sensitive to noise, and the robustness is poor. So it is difficult to use in practice. Although the accuracy of the iterative method has been improved, the calculation time is longer. At the same time, the problem of noise is not solved. Robust method's accuracy is improved, and the problem of noise is solved. But the algorithm is too complicated, which is not easy to achieve.

Because the pixel coordinates extracted from model plane have the feature of high accuracy and good stability, they were used to obtain a precise and stable fundamental matrix. Based on the relationship of polar geometric constraint, the distance between the pixel point and its corresponding polar line is used as the objective function of optimization to calculate the fundamental matrix.

Let $I' = (l'_1, l'_2, l'_3) = Fm$, $I = (l_1, l_2, l_3) = m^T F$, distance between the pixel point and its corresponding polar line can be expressed as:

$$d(m', l') = \frac{|m'^T I'|}{\sqrt{(l'_1)^2 + (l'_2)^2}} \quad (9)$$

$$d(m, l) = \frac{|Im|}{\sqrt{(l_1)^2 + (l_2)^2}} \quad (10)$$

Where, $d(m', l')$ is the distance between point in the image of left camera and its corresponding polar line in the image of right camera; $d(m, l)$ is the distance between point in the image of right camera and its corresponding polar line in the image of left camera; m' is the pixel coordinates of point in the left camera; m is the pixel coordinates of point in the right camera.

As a result, the objective function can be defined as:

$$\min_F \sum_{i=1}^n [k_i'^2 d^2(q'_i, I'_i) + k_i^2 d^2(q_i, I_i)] \quad (11)$$

$$\text{Where } k' = \sqrt{(l'_1)^2 + (l'_2)^2}, k = \sqrt{(l_1)^2 + (l_2)^2}$$

3. Measurement of Cam Spacing on Assembled Camshaft

3.1. Calibration of Parameter

In the laboratory, two JAI CCD camera with the resolution of 1392×1040 pixel was used to establish a binocular vision measurement system. The precious of model plane is $\pm 1\mu\text{m}$. The measured camshaft is shown in Figure 2.



Figure 2. Measured Camshaft

The effective focal lengths of two cameras are obtained by the method proposed in^[11]:

$$f_l = 8.8481\text{mm}, \quad f_r = 8.5061\text{mm}.$$

Using the corner points on the model plane, the fundamental matrix is fitted:

$$F = \begin{bmatrix} -1.4999\text{E-}5 & -1.2035\text{E-}4 & 0.0324 \\ 1.1852\text{E-}4 & -3.7553\text{E-}5 & -0.0721 \\ -0.0073 & 0.0720 & -8.5615 \end{bmatrix}$$

From the fundamental matrix, the relationships between the points and their corresponding polar lines are shown in Figure 3:

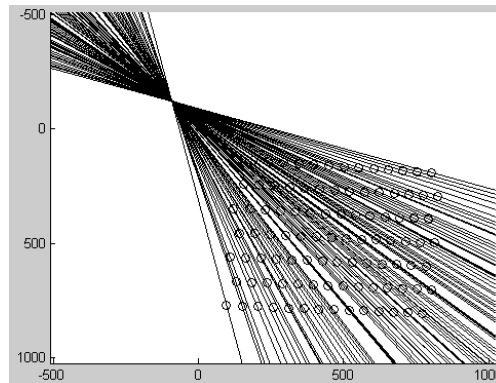


Figure 3. Relationships Between the Points and their Corresponding Polar Lines

Decomposing the fundamental matrix, the rotation matrix and translation vector between two cameras could be obtained:

$$R = \begin{bmatrix} -0.9757 & 0.2184 & -0.0163 \\ -0.2123 & -0.9613 & -0.1753 \\ -0.0540 & -0.1676 & 0.98436 \end{bmatrix} \quad T = \begin{bmatrix} -2.8002 \\ -8.5672 \\ 89.1906 \end{bmatrix}$$

The fundamental matrix provides a constraint for the matching of corresponding points. The edge of the cam can be taken as the other constraint, which is shown in Figure4.

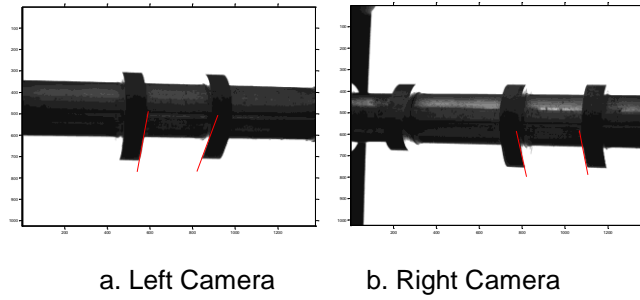


Figure 4. Image for Measurement

3.2. Experimental Result

From the corresponding edge points, the 3-dimensional points on the cam edge was reconstructed, which is shown in Figure 5. By the linear fitting and calculating of distance between straight lines, the cam spacing was obtained.

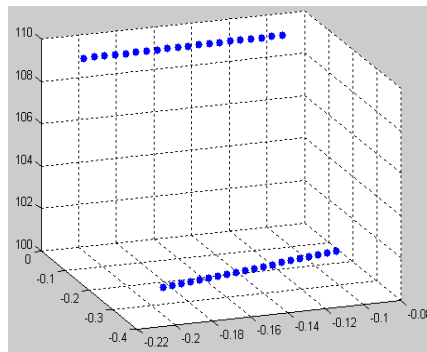


Figure 5. Result of 3-Dimensional Reconstruction

The result is contrasted to the measurement by vernier caliper, the results repeated 5 times for each cam spacing is shown in Table 1.

Table 1. Result of Measuring Cam Spacing (mm)

	1	2	3	4	5	Mean	Standard deviation
Proposed method	30.0	30.0	30.1	30.0	30.1	30.0	0.0
	941	894	000	899	201	987	127
Vernier caliper	30.0	30.0	30.0	30.0	30.0	30.0	0.0
	31	26	32	43	50	364	0987

From the measurement result, the error of the proposed method is smaller than ± 0.1 mm, which can meet the requirement of practical measurement.

4. Conclusions

A vision measurement method is proposed for measuring cam spacing on assembled camshaft. This method can meet the requirement of practical measurement, which will reduce the labor intensity of workers.

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